## Inductive Temporal Logic Programming

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### 1. Introduction

The field of *Inductive Logic Programming* (see [121], [119], [120], [118], [99] and [126]) is an area of active research. Inductive Logic Programming (ILP) deals with the topic of extracting a suitable explanation of a phenomenon from finite sets of examples. While this is the basic topic in most areas of *Artificial Intelligence* (AI), the target concept in ILP (i.e. the phenomenon to be *learned*) is a logic program. Logic programs are finite sets of clauses which allow a very natural interpretation as declarations of procedural rules. While this natural interpretation of a clause

$$A \leftarrow B_1, \ldots, B_n$$

is "to solve the task A, solve the sequence  $B_1, \ldots, B_n$  of subtasks", the interpretation in ILP is more rule based: if  $x_1, \ldots, x_n$  are the variable symbols occurring in the head of the clause (i.e. in A), then  $x_1, \ldots, x_n$  will be assumed to have the property A if they have the properties  $B_1, \ldots, B_n$ . This gives the interpretation "if  $B_1, \ldots, B_n$  hold, then A does also hold".

Since the concepts of interest are logic programs, examples for any ILP-based learning system can (or should) be *ground atoms*. Each such example can be either a *positive* or a *negative* example: if P is the program to be learned and e is an example, then

- if e is positive, then  $P \models e$  and
- if e is negative, then  $P \not\models e$ .

ILP techniques have been applied in many branches of science, notable in computational biology. There, several areas of application have been identified, notably

- Drug-activity comparison (see [88], [148], [24] and [87]),
- Diagnosis of rheumatic diseases (see [33] and [100]),
- Mutagenesis prediction (see [40], [151], [152] and [153]),
- Protein-structure prediction (see [122] and [86]) and
- Design of Medical Diagnosis Tools (see [136] and [101]).

Other areas of application are prediction of strategies for chess games (see [16]), finite element methods (see [50], [52], [51], [54] and [55]), data mining (see [173], [115] and [116]) and learning models for dynamic systems (see [25] and [55]).

This thesis deals with a natural extension of the classical ILP paradigm, namely synthesizing *temporal* logic programs from given examples. Temporal Logic is a natural way to describe relations which may change over time. We will distinguish two different areas:

- Propositional Inductive Temporal Logic Programming Here the language used in order to describe the programs is LTL, a simple temporal logic language allowing the use of operators such as X, G, F, U and R for modeling time-dependent relations. LTL is very popular in the fields of *Model Checking* and *Supervisory Control*. We will exploit the fact that each LTL-formula can be represented as a nondeterministic Büchi-automaton (an automaton accepting infinite sequences of letters) in order to define operators which manipulate such an automaton in order to fit the specification given by the examples.
- First Order Inductive Temporal Logic Programming Here the language of interest is much more flexible and expressive than in the propositional case. The major drawback is the undecidability of first order logic (which has been proven by Church in

1936) and therefore of the full first order temporal logic. To keep first order temporal logic tractable, we identify a simple extension of a PROLOG-style language which allows the usage of temporal the operators X, G, F, U and P in front of literals. We will see that this programming language (which we will call PROLOG(+T)) has a very natural procedural interpretation in terms of some well known rewrite rules for temporal logic formulas. These rules will be used in order to give a *saturation* based calculus.

The main part of this thesis will be devoted to the field of First Order Inductive Temporal Logic Programming. We will see that the lattice properties (with respect to the *subsumption*-ordering) of first order atoms, literals and clauses can be extended to PROLOG(+T)-atoms, -literals and -clauses. So the existence of least generalizations and greatest specializations of PROLOG(+T)-clauses can be ensured. We will give algorithms which allow the computation of such generalizations and specializations and use the techniques from these algorithms in order to define *refinement operators* for PROLOG(+T)programs.

The thesis is structured as follows: in the first part we will define some basic notations from first order logic, temporal logic and logic programming. The chapters in this first part are kept rather short since we assume that the reader is familiar with these topics. After having introduced these basic concepts we will briefly introduce some concepts from the field of ILP.

The second part is dedicated to an in-depth treatment of First Order Temporal ILP. This includes the definition of PROLOG(+T) and the discussion of its declarative semantics. Having achieved this, we will present a proof procedure, discuss the lattice properties of PROLOG(+T) objects and study refinement operators.

The third part is then dedicated to Propositional Temporal ILP. After having defined basic concepts from the field of  $\omega$ -automata, we will present two operations for refining LTL-programs by manipulating their representing automata. The final chapter of this third part will be devoted to the question of the complexity of the identification task. Therefore we will derive upper bounds for the VC-dimension of certain classes of LTLprograms. These VC-dimensions allow a direct extraction of the number of examples which are needed in order to identify the program under consideration (or more precise: a program which is equivalent to the program under consideration). Part I.

## Temporal Logic and (inductive) Logic Programming

## 2. Preliminaries

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In this chapter we will briefly define and review some of the basic and most important concepts which we will use throughout the rest of this thesis. This includes propositional logic, first order logic and temporal logic. Each of these three logics is equipped with both a semantical consequence relation which we will as usual denote as  $\models$ . The three logics will be defined by first defining the sets of formulas which can be built from some atomic objects and connectives and can be seen as a more or less detailed way to describe mathematical concepts in a syntactic way. The properties of the logics are only mentioned. We will not prove them since the literature on propositional, first order and temporal logic is rich (see [18], [147], [31], [20], [89] and [56]).

#### 2.1. Propositional Logic

The simplest logic which we will define is the classical *propositional logic*. Propositional logic is a formalism which has been studied very well. Early studies were done by

Boole (see [23]) in the 19th century. But to this date the study of propositional logic was motivated by the circumstance that researchers wanted to formalize the process of mathematical reasoning. Propositional logic seemed to be a good starting point for such formalizations. Later the usefulness of propositional logic for the description of electrical and electronical circuits was pointed out. We refer to [104] for an introduction to applications of propositional logic.

Now assume that a countable infinite set  $X = \{p_i \mid i \in \mathbb{N}\}$  is given. Each element of X will be called a *propositional variable* or simply a *variable* if there is no way of confusion. Formulas of the propositional language defined over X are defined inductively as follows (roughly following the treatment from [20]).

#### Definition 2.1.1 (Propositional Logic)

Let a set X of propositional variables be given. The set of *formulas* over X is defined as:

- 1. true and false are formulas,
- 2. each  $p \in X$  is a formula,
- 3. if  $\varphi$  is a formula, then  $\neg \varphi$  is a formula and
- 4. if  $\varphi_1$  and  $\varphi_2$  are formulas, then so are  $(\varphi_1 \land \varphi_2)$ ,  $(\varphi_1 \lor \varphi_2)$ ,  $(\varphi_1 \rightarrow \varphi_2)$  and  $(\varphi_1 \leftrightarrow \varphi_2)$ .

The set of *all* formulas over X will be denoted as  $\mathcal{F}(X)$ .

The above definition of propositional formulas models the syntactic level of propositional reasoning. To model the semantic part, i.e. the logical consequence relation, we will now define a way to *evaluate* propositional formulas to values 1 (identified as true) and 0 (identified as *false*). This will be done using a suitable concept of *evaluation*  functions.

#### Definition 2.1.2 (Evaluation)

Let X be a set of propositional variables. An evaluation for X is a function  $v : X \cup \{\texttt{true}, \texttt{false}\} \rightarrow \mathbb{Z}_2$  (where  $\mathbb{Z}_2$  denotes the Galois field of cardinality 2) satisfying v(true) = 1 and v(false) = 0.

The concept of an evaluation can now be extended to functions  $\hat{v} : \mathcal{F}(X) \to \mathbb{Z}_2$  in the obvious (homomorphic) way. Let  $\varphi \in \mathcal{F}(X)$  be any formula. Then

- 1. if  $\varphi = p \in X$ , then  $\hat{v}(p) = v(p)$ ,
- 2. if  $\varphi = \neg \psi$ , then  $\hat{v}(\varphi) = 1 \hat{v}(\varphi)$ ,
- 3. if  $\varphi = (\varphi_1 \land \varphi_2)$ , then  $\hat{v}(\varphi) = \min{\{\hat{v}(\varphi_1), \hat{v}(\varphi_2)\}},$
- 4. if  $\varphi = (\varphi_1 \lor \varphi_2)$ , then  $\hat{v}(\varphi) = \max{\{\hat{v}(\varphi_1), \hat{v}(\varphi_2)\}},$
- 5. if  $\varphi = (\varphi_1 \to \varphi_2)$ , then  $\hat{v}(\varphi) = \max \{ \hat{v}(\neg \varphi_1), \hat{v}(\varphi_2) \}$  and
- 6. if  $\varphi = (\varphi_1 \leftrightarrow \varphi_2)$ , then  $\hat{v}(\varphi) = \min \{ \hat{v}(\varphi_1 \to \varphi_2), \hat{v}(\varphi_2 \to \varphi_1) \}.$

Since there is no way of confusion we will from now on identify  $\hat{v}$  and v writing  $v(\varphi)$  for the result of  $\hat{v}(\varphi)$  for any formula  $\varphi \in \mathcal{F}(X)$ . The set of all evaluations (or valuations from now on) will be denoted as VAL.

For the sake of simplicity we will introduce two more concepts:

1. Let  $\varphi_1, \ldots, \varphi_n$  be any finite sequence of formulas from  $\mathcal{F}(X)$ . Then the formulas  $\bigwedge_{i=1}^n \varphi_i$  and  $\bigvee_{i=1}^n \varphi_i$  are defined as

$$\bigwedge_{i=1}^{n} \varphi_i = \left(\varphi_n \land \bigwedge_{i=1}^{n-1} \varphi_i\right) \text{ and}$$
$$\bigvee_{i=1}^{n} \varphi_i = \left(\varphi_n \lor \bigvee_{i=1}^{n-1} \varphi_i\right).$$

- 2. We assume that the connectives have the following binding priorities:
  - a)  $\neg$  has a higher binding priority than  $\land$ ,
  - b)  $\wedge$  has a higher binding priority than  $\vee$ ,
  - c)  $\lor$  has a higher binding priority than  $\rightarrow$  and
  - d)  $\rightarrow$  has a higher binding priority than  $\leftrightarrow$ .

Applying the binding priority rules allows the omission or brackets in many formulas which improves the readability. For example, the formula

$$(\neg p_1 \lor (p_2 \land p_3)) \to p_4$$

can be written as

$$\neg p_1 \lor p_2 \land p_3 \to p_4$$

Now let  $\varphi$  be any formula. A valuation v is called a *model* of  $\varphi$  if and only if  $v(\varphi) = 1$ . The set of all models of  $\varphi$  will be denoted as  $MD(\varphi)^1$ . Similarly for sets  $\Phi$  of formulas we define a valuation v to be a model of  $\Phi$  if  $v(\varphi) = 1$  for every  $\varphi \in \Phi$ . We then have  $MD(\Phi) = \bigcap_{\varphi \in \Phi} MD(\varphi)$ .

We will call pairs  $(\varphi_1, \varphi_2) \in \mathcal{F}(X)^2$  (semantically) equivalent (written  $\varphi_1 \equiv \varphi_2$ ) if and only if  $v(\varphi_1) = v(\varphi_2)$  for every valuation v. Equivalently we could define semantical equivalence as follows:

> $\varphi_1 \equiv \varphi_2$  if and only if  $\{\varphi_1\} \models \varphi_2$  and  $\{\varphi_2\} \models \varphi_1$ if and only if  $\emptyset \models \varphi_1 \leftrightarrow \varphi_2$

<sup>1</sup>So  $MD(\varphi) = \{ v \in \text{VAL} \mid v(\varphi) = 1 \}.$ 

where  $\models$  denotes the semantical consequence relation which will be defined below.

Note that  $\equiv$  is an equivalence relation on  $\mathcal{F}(X)$ . Furthermore note that  $\varphi_1 \equiv \varphi_2$  if and only if  $MD(\varphi_1) = MD(\varphi_2)$ .

A formula  $\varphi \in \mathcal{F}(X)$  is called

- satisfiable if  $MD(\varphi) \neq \emptyset$ ,
- valid if  $MD(\varphi) = VAL$  and
- unsatisfiable if  $MD(\varphi) = \emptyset$ .

Similar concepts can be defined for sets of formulas. A set  $\Phi \subseteq \mathcal{F}(X)$  is called

- satisfiable if  $MD(\Phi) \neq \emptyset$ ,
- valid if  $MD(\Phi) = VAL$  and
- unsatisfiable if  $MD(\Phi) = \emptyset$ .

Satisfiable sets of propositional formulas can be characterized by Theorem 2.1.1 which is commonly known as the *finiteness theorem*.

#### Theorem 2.1.1 (Finiteness Theorem)

Let  $\Phi \subseteq \mathcal{F}(X)$  be a set of formulas. Then  $\Phi$  is satisfiable if and only if every finite set  $\Psi \subseteq \Phi$  is satisfiable.

The concept of *logical consequence* is commonly modeled as follows: A formula  $\varphi$  is a logical consequence of a set  $\Phi \subseteq \mathcal{F}(X)$  if every model of  $\Phi$  is also a model of  $\varphi$ .

#### Definition 2.1.3 (Logical Consequence)

Let  $\Phi \subseteq \mathcal{F}(X)$  be a set of formulas and let  $\varphi \in \mathcal{F}(X)$  be a formula. Then  $\varphi$  is a logical consequence of  $\Phi$  (written as  $\Phi \models \varphi$ ) if and only if for every  $v \in \text{VAL}$  such as  $v(\Phi) \subseteq \{1\}^2$  it holds that  $v(\varphi) = 1$ .

Equivalently one can define  $\Phi \models \varphi$  if and only if  $MD(\Phi) \subseteq MD(\varphi)$ . Moreover we can define the following: Let  $\Phi$  be as above and let  $\Psi \subseteq \mathcal{F}(X)$  be any set of formulas. Then  $\Phi \models \Psi$  if and only if  $\Phi \models \psi$  for every  $\psi \in \Psi$ . If  $\Psi$  is a finite set, say  $\Psi = \{\psi_1, \ldots, \psi_n\}$ , then  $\Phi \models \Psi$  if and only if  $\Phi \models \bigwedge_{i=1}^n \psi_i$ .

Equivalently one can use the following characterization of unsatisfiable sets of formulas.

#### Theorem 2.1.2

Let  $\Phi \subseteq \mathcal{F}(X)$  be a set of formulas. Then  $\Phi$  is unsatisfiable if and only if  $\Phi \models \texttt{false}$ .

The logical consequence relation enjoys the following nice properties.

#### Theorem 2.1.3

Let  $\Phi \subseteq \mathcal{F}(X)$  be a set of formulas and let  $\varphi \in \mathcal{F}(X)$  be a formula. Then

1.  $\Phi \models \varphi$  if and only if  $\Phi \cup \{\neg \varphi\} \models \texttt{false}$ .

2. If  $\Phi \models \texttt{false}$ , then there is a finite set  $\Psi \subseteq \Phi$  such that  $\Psi \models \texttt{false}$ .

Part 2 of Theorem 2.1.3 is also known as the *compactness theorem*. It is an easy corollary of Theorem 2.1.1.

For finite sets  $\Phi \subseteq \mathcal{F}(X)$  of propositional logic formulas and formulas  $\varphi \in \mathcal{F}(X)$  it is *decidable* whether  $\Phi \models \varphi$  holds or not. However, the related *satisfiability* problem is  $\mathcal{NP}$ -complete (see [38] and [71]), that is one cannot (or better *should not*) hope for efficient procedures which are capable of deciding if a formula  $\varphi$  (or a set of  $\Phi$ formulas) is satisfiable. Moreover the problem of deciding the logical consequence relation is  $co - \mathcal{NP}$ -complete which indicates that is in some sense even more difficult to decide than satisfiability.

#### 2.2. First Order Logic

In contrast to propositional logic, *first order logic* or *(first order) predicate logic* allows a more precise formalization of (mathematical) relations. Using a more flexible language of

logic was motivated by the limitations arising from the usage of propositional logic. For example propositional logic only allows reasoning about propositions, i.e. things which are either *true* or *false*. But in Mathematics the truth or falsity of a proposition often depends on the values of certain variables occurring as inputs to functions. For example the formula f(x) = 0 should be evaluable to true or false. But the truth-value of this formula depends on

- the function which is represented by the function-symbol f and
- the value which is assigned to the variable x.

So propositional logic is not an adequate formal system for modeling this formula. Early studies of first order logic were presented at the beginning of the 20th century e.g. by Frege (see [69]), Gentzen (see [75]), Russell and Whitehead (see [170]) and several others. At this point of time sound and complete calculi have been developed. But practical applications arose much later.

As we have already pointed out formulas are not built from *propositions* alone but from a more general concept which we will call *atomic formulas* or simply *atoms* from now on. Therefore we will have to refine the syntax of the logical language to be used in a suitable way.

Recall that in the case of propositional logic, the syntax (i.e. the formulas of a logical language) only depends on the set X of propositional variables. A similar concept for first order logic is given by the concept of *signatures*.

#### Definition 2.2.1 (Signature)

A signature is a tuple sig =  $(\mathcal{X}, F, P, \alpha)$  where

- 1.  $\mathcal{X} = \{x_i \mid i \in I \text{ for some set } I \subseteq \mathbb{N} \text{ of indices}\}$  is a countable set of variable symbols,
- 2. F and P are finite sets of function-resp. predicate-symbols and

3.  $\alpha: F \cup P \to \mathbb{N}$  is a function which maps each symbol to a natural number (its arity) and is therefore called the arity-function.

If  $\sigma \in F \cup P$  is a function-respectively predicate-symbol and  $\alpha(\sigma) = n$  for some  $n \ge 0$ , we will say that  $\sigma$  has arity n. In the case that n = 0 we will also say that  $\sigma$  is a constant symbol.

The simplest objects which can be built from a signature are terms.

#### Definition 2.2.2 (Terms)

Let sig =  $(\mathcal{X}, F, P, \alpha)$  be a signature. The set  $\mathcal{T}(sig)$  (or simply  $\mathcal{T}$ ) of terms over sig is defined inductively as follows:

- each x ∈ X is a term and
   if f ∈ F is a function-symbol, n = α(f) and t<sub>1</sub>,..., t<sub>n</sub> ∈ T are terms, then so is

Formulas are now defined to be either *atomic* ones or formulas composed from simpler ones.

#### **Definition 2.2.3 (Atomic Formulas)**

Let sig =  $(\mathcal{X}, F, P, \alpha)$  be a signature. The set  $\mathcal{A}(sig)$  (or simply  $\mathcal{A}$ ) of all *atomic* formulas (or simply *atoms*) over sig is defined inductively as follows:

- *rmulas* (or  $\ldots$  ). 1. true and false are in  $\mathcal{A}$  and 2. if  $p \in P$  is a predicate symbol,  $n = \alpha(p)$  and  $t_1, \ldots, t_n \in \mathcal{T}$  are terms, then t ) is in  $\mathcal{A}$ .

In contrast to formulas of the propositional logic language, first order logic formulas are also capable to model terms like "for all x it holds that ..." and "there is an x such that ...". This is achieved by introducing two quantifiers  $\forall$  (the universal quantifier) and  $\exists$  (the existence quantifier).

#### Definition 2.2.4 (First Order Formulas)

Let sig =  $(\mathcal{X}, F, P, \alpha)$  be a signature. The set  $\mathcal{F}(sig)$  (or simply  $\mathcal{F}$ ) of formulas over sig is defined inductively as follows:

- 1. every  $\varphi \in \mathcal{A}$  is a formula,
- 2. if  $\varphi$  is a formula, then so is  $\neg \varphi$ ,
- 3. if  $\varphi_1$  and  $\varphi_2$  are formulas, then so are  $(\varphi_1 \land \varphi_2)$ ,  $(\varphi_1 \lor \varphi_2)$ ,  $(\varphi_1 \to \varphi_2)$  and  $(\varphi_1 \leftrightarrow \varphi_2)$  and
- 4. if  $\varphi$  is a formula and  $x \in \mathcal{X}$  is a variable symbol, then  $\forall x \varphi$  and  $\exists x \varphi$  are formulas.

Formulas from the set

$$\mathcal{A} \cup \{\neg \varphi \mid \varphi \in \mathcal{A}\}$$

are called *literals*. Since the formulas of first order logic allow finer reasoning about mathematical concepts, their *interpretation* also has to be more detailed. This includes the interpretation of the function symbols and the interpretation of the predicate symbols. So the concept of *evaluation functions* as introduced for propositional logic is not adequate anymore. An interpretation is given as a tuple consisting of a set of possible *values* of the variables (the *universe*), two mappings assigning functions to the function symbols and predicates to the predicate symbols and a mapping assigning values to the variable symbols.

#### Definition 2.2.5 (Interpretation)

Let sig =  $(\mathcal{X}, F, P, \alpha)$  be a signature. An *interpretation* (for sig–formulas) is a tuple

$$\mathcal{J} = (U_{\mathcal{J}}, \mathfrak{F}, \mathfrak{P}, w) \,,$$

where

- 1.  $U_{\mathcal{J}}$  is a nonempty set of objects (the *universe* of  $\mathcal{J}$ ),
- 2.  $\mathfrak{F}$  is a function which maps each  $f \in F$  to a function  $f^{\mathcal{J}}: U_{\mathcal{J}}^{\alpha(f)} \to U_{\mathcal{J}},$
- 3.  $\mathfrak{P}$  is a function which maps each  $p \in F$  to a predicate  $p^{\mathcal{J}} : U_{\mathcal{J}}^{\alpha(p)} \to \{0,1\}$  and
- 4.  $w: \mathcal{X} \to U_{\mathcal{J}}$  maps each variable symbol x to an element  $w(x) \in U_{\mathcal{J}}$ .

Now let  $\mathcal{J}$  be an interpretation. The *evaluation of terms* is straightforward depending on the structure of the term to be evaluated:

1. 
$$\mathcal{J}(x) =: x^{\mathcal{J}} = w(x)$$
 for every  $x \in \mathcal{X}$  and  
2.  $\mathcal{J}(f(t_1, \dots, t_n)) = f^{\mathcal{J}}(t_1^{\mathcal{J}}, \dots, t_n^{\mathcal{J}})$  for each  $f(t_1, \dots, t_n) \in \mathcal{T}$ .

Having defined how to evaluate terms, every formula can be evaluated in a straightforward way:

- 1.  $\mathcal{J}(p(t_1,...,t_n)) = p^{\mathcal{J}}(t_1^{\mathcal{J}},...,t_n^{\mathcal{J}}) \in \{0,1\},\$
- 2.  $\mathcal{J}(\neg \varphi) = 1 \mathcal{J}(\varphi),$
- 3.  $\mathcal{J}(\varphi_1 \land \varphi_2) = \min \{ \mathcal{J}(\varphi_1), \mathcal{J}(\varphi_2) \},\$
- 4.  $\mathcal{J}(\varphi_1 \lor \varphi_2) = \max{\{\mathcal{J}(\varphi_1), \mathcal{J}(\varphi_2)\}},\$
- 5.  $\mathcal{J}(\varphi_1 \to \varphi_2) = \max{\{\mathcal{J}(\neg \varphi_1), \mathcal{J}(\varphi_2)\}},\$

- 6.  $\mathcal{J}(\varphi_1 \leftrightarrow \varphi_2) = \min \{\mathcal{J}(\varphi_1 \to \varphi_2), \mathcal{J}(\varphi_2 \to \varphi_1)\},\$
- 7.  $\mathcal{J}(\forall x\varphi) = \min \left\{ \mathcal{J}_x^t(\varphi) \mid t \in U_{\mathcal{J}} \right\}$  and
- 8.  $\mathcal{J}(\exists x\varphi) = \max \left\{ \mathcal{J}_x^t(\varphi) \mid t \in U_{\mathcal{J}} \right\}.$

where  $\mathcal{J}_x^t$  is defined as  $\mathcal{J}_x^t = (U_{\mathcal{J}}, \mathfrak{F}, \mathfrak{P}, \tilde{w})$  with  $\tilde{w} : \mathcal{X} \to U_{\mathcal{J}}$  being defined as

$$\tilde{w}(\bar{x}) = \begin{cases} w(\bar{x}) & \Leftrightarrow \bar{x} \neq x \\ t & \Leftrightarrow \bar{x} = x \end{cases}.$$

We will write  $\mathcal{J} \models \varphi$  if  $\mathcal{J}(\varphi) = 1$ . Similarly we will write  $\mathcal{J} \models \Phi$  (for a set  $\Phi \subseteq \mathcal{F}$ ) if and only if  $\mathcal{J} \models \varphi$  for every  $\varphi \in \Phi$ .

As in the case of propositional logic an interpretation  $\mathcal{J}$  is called a *model* of a formula  $\varphi$  (respectively a *model* of a set  $\Phi$  of formulas) if  $\mathcal{J} \models \varphi$  (respectively  $\mathcal{J} \models \Phi$ ). The set of all models of  $\varphi$  is defined to be

$$MD(\varphi) = \{\mathcal{J} \mid \mathcal{J} \models \varphi\}$$

and the set of all models for  $\Phi$  is

$$MD(\Phi) = \bigcap_{\varphi \in \Phi} MD(\varphi).$$

We will adopt the following terms from propositional logic: A formula  $\varphi \in \mathcal{F}$  is called

- satisfiable if  $MD(\varphi) \neq \emptyset$ ,
- valid if  $\mathcal{J} \models \varphi$  for every interpretation  $\mathcal{J}$  and
- unsatisfiable if  $MD(\varphi) = \emptyset$ .

Similar concepts can be defined for sets of formulas. A set  $\Phi \subseteq \mathcal{F}$  is called

- satisfiable if  $MD(\Phi) \neq \emptyset$ ,
- valid if every  $\varphi \in \Phi$  is valid and
- unsatisfiable if  $MD(\Phi) = \emptyset$ .

As in the case of propositional logic we define a relation  $\models$  by defining:

#### Definition 2.2.6 (Logical Consequence)

Let  $\Phi \subseteq \mathcal{F}$  be a set of formulas and let  $\varphi \in \mathcal{F}$  be a formula. Then  $\varphi$  is a logical consequence of  $\Phi$  ( $\Phi \models \varphi$ ) if and only if for every interpretation  $\mathcal{J}$  such as  $\mathcal{J}(\Phi) = 1$  it holds that  $\mathcal{J}(\varphi) = 1$ .

Again we can express  $\models$  by  $\Phi \models \varphi$  if and only if  $MD(\Phi) \subseteq MD(\varphi)$ .

The properties of  $\models$  are still the same as in propositional logic.

#### **Theorem 2.2.1**

Let  $\Phi \subseteq \mathcal{F}$  be a set of formulas. Then  $\Phi$  is unsatisfiable if and only if  $\Phi \models \texttt{false}$ .

#### Theorem 2.2.2

Let  $\Phi \subseteq \mathcal{F}$  be a set of formulas. Then  $\Phi$  is satisfiable if and only if every finite set  $\Phi' \subseteq \Phi$  is satisfiable.

#### Theorem 2.2.3

Let  $\Phi \subseteq \mathcal{F}$  be a set of formulas and let  $\varphi \in \mathcal{F}$  be a formula. Then

- 1.  $\Phi \models \varphi$  if and only if  $\Phi \cup \{\neg \varphi\} \models \texttt{false}$ .
- 2. If  $\Phi \models \texttt{false}$ , then there is a finite set  $\Psi \subseteq \Phi$  such that  $\Psi \models \texttt{false}$ .

In contrast to propositional logic, where testing for satisfiability is decidable (but  $\mathcal{NP}$ complete), testing for satisfiability is undecidable in first order logic. This is due to the
following theorem proved by Church in 1936 (see [32]).

#### Theorem 2.2.4

Let  $\Phi \subseteq \mathcal{F}$  be a set of formulas. Then the following problem is undecidable:

Input:  $\Phi$ 

**Output:** 
$$\begin{cases} 1 \quad \Leftrightarrow MD(\Phi) \neq \emptyset \\ 0 \quad \Leftrightarrow \text{else} \end{cases}$$

#### 2.3. Temporal Logic

In contrast to the logics which we have introduced so far, *temporal logic* is concerned with reasoning about *time-dependent properties*. An example might be the operator X which has the following intuitive interpretation:

If  $\varphi$  is true at the next point of time, then  $X\varphi$  is true at the current point of time.

Consequently X will be referred to as Next-State-Operator. Here we can already notice that in our temporal logics time will be of discrete nature. Consequently any sequence of *points of time* can only contain countably many such *points*.

Again we wish to distinguish temporal logics according to the primitive objects under consideration. So we will have propositional temporal logic and first order temporal logic as alternatives while the former is properly contained in the latter. Another possible criterion of differentiation is between linear time (see [110]) and branching time (see [58], [59]) logics. Linear time logics allow reasoning about one possible continuation of the current point of time while branching time logics are equipped with operators quantifying over sequences of continuations, so called paths (and are therefore called path quantifiers). We will only consider linear time temporal logics since they are well suited for our purposes. Perhaps the most prominent of these linear time temporal logics is LTL which has been subject of both theoretical research and practical applications.

Since temporal logics allow reasoning about time-dependent aspects of objects, the concept of interpretations will have to be extended to *sequences of interpretations*. This

will be the subject of consideration for the rest of this chapter.

#### 2.3.1. Propositional Temporal Logic

Again assume that X is a given set of *propositional variables* as defined in section 2.1, that is  $X = \{p_i \mid i \in \mathbb{N}\}$ . The language LTL of linear time temporal logic formulas is built from the language  $\mathcal{F}(X)$  by introducing several temporal operators.

#### Definition 2.3.1 (Propositional Linear Time Temporal Logic, e.g. [110])

The language LTL of *linear time temporal logic formulas* is inductively defined as

- every φ ∈ F(X) is in LTL,
   if φ is in LTL, then so are Xφ, Gφ and Fφ and
   if φ<sub>1</sub>, φ<sub>2</sub> are in LTL, then so are φ<sub>1</sub>Uφ<sub>2</sub> and φ<sub>1</sub>Rφ<sub>2</sub>

The temporal operators X, G, F, U and R will have the following intuitive interpretation:

- 1.  $X\varphi$ : if  $\varphi$  is true at the next point of time, then  $X\varphi$  is true at the actual point of time (*Next-State-Operator*).
- 2.  $\mathbf{G}\varphi$ :  $\varphi$  is true at every point of time (Always-Operator).
- 3. F $\varphi$ : there is a point of time such that  $\varphi$  is true at this point (*Eventually-Operator*).
- 4.  $\varphi_1 \cup \varphi_2$ :  $\varphi_1$  holds until  $\varphi_2$  is true (Until-Operator).
- 5.  $\varphi_1 \mathsf{R} \varphi_2$ :  $\varphi_1$  has to be true *before*  $\varphi_2$  is true (*Release-Operator*).

Formally LTL-formulas are evaluated in sequences of states each of which is a single evaluation of the propositional symbols of the language which is defined by the set X. Sequences of states are assumed to be

• infinite and

• countable.

So each such sequence is isomorphic to the set  $\mathbb N$  of natural numbers and the time points in these sequences are *discrete*. Consequently the set of *all* sequences of states is uncountable.

# Definition 2.3.2 (Temporal State) A temporal state is a set $s \subseteq X$ .

We can interpret a temporal state s as an evaluation  $v_s: X \to \mathbb{Z}_2$  defined by

$$v_s(x) = \begin{cases} 1 & \Leftrightarrow x \in s \\ \\ 0 & \Leftrightarrow \text{else} \end{cases}$$

and extend this evaluation from X to  $\mathcal{F}(X)$  in the obvious way.

So far we are not able to assign a meaning to the temporal operators G, F, X, U and P. Therefore we extend the concept of an evaluation (as introduced in section 2.1) to temporal interpretations defined formally as follows.

#### Definition 2.3.3 (Temporal Interpretation)

A temporal interpretation (or interpretation for short) is an infinite sequence  $\mathcal{J} = (s_0, s_1, \dots, s_i, \dots)$  of temporal states. For  $j \in \mathbb{N}$  the notation  $\mathcal{J}^j$  will denote the temporal interpretation starting at time point j, i.e.  $\mathcal{J}^j = (s_j, s_{j+1}, \dots, s_k, \dots)$ .

Now let  $\varphi \in LTL$  be a formula and let  $\mathcal{J} = (s_0, s_1, \dots, s_i, \dots)$  be a temporal interpretation. We extend the relation  $\models$  as follows:

1. if  $\varphi \in X$ , then  $\mathcal{J} \models \varphi$  if and only if  $\varphi \in s_0$ ,

- 2. if  $\varphi = \neg \psi$ , then  $\mathcal{J} \models \varphi$  if and only if  $\mathcal{J} \not\models \psi$ ,
- 3. if  $\varphi = \varphi_1 \land \varphi_2$ , then  $\mathcal{J} \models \varphi$  if and only if  $\mathcal{J} \models \varphi_1$  and  $\mathcal{J} \models \varphi_2$ ,
- 4. if  $\varphi = \varphi_1 \lor \varphi_2$ , then  $\mathcal{J} \models \varphi$  if and only if  $\mathcal{J} \models \varphi_1$  or  $\mathcal{J} \models \varphi_2$ ,
- 5. if  $\varphi = \varphi_1 \to \varphi_2$ , then  $\mathcal{J} \models \varphi$  if and only if  $\mathcal{J} \not\models \varphi_1$  or  $\mathcal{J} \models \varphi_2$ ,
- 6. if  $\varphi = \mathsf{X}\psi$ , then  $\mathcal{J} \models \varphi$  if and only if  $\mathcal{J}^1 \models \psi$ ,
- 7. if  $\varphi = \mathsf{G}\psi$ , then  $\mathcal{J} \models \varphi$  if and only for every  $i \ge 0$  it holds that  $\mathcal{J}^i \models \psi$ ,
- 8. if  $\varphi = \mathsf{F}\psi$ , then  $\mathcal{J} \models \varphi$  if and only if there is  $i \ge 0$  such that  $\mathcal{J}^i \models \psi$ ,
- 9. if  $\varphi = \varphi_1 \cup \varphi_2$ , then  $\mathcal{J} \models \varphi$  if and only if there is  $i \ge 0$  such that  $\mathcal{J}^i \models \varphi_2$  and for every j such that  $0 \le j < i$  it holds that  $\mathcal{J}^j \models \varphi_1$  and
- 10. if  $\varphi = \varphi_1 \mathbb{R} \varphi_2$ , then  $\mathcal{J} \models \varphi$  if and only if for every  $i \ge 0$  such that  $\mathcal{J}^i \not\models \varphi_2$  there is j such that  $0 \le j < i$  and  $\mathcal{J}^j \models \varphi_1$ .

As in the case of propositional and first order logic, an interpretation  $\mathcal{J}$  with  $\mathcal{J} \models \varphi$ for some LTL-formula  $\varphi$  is called a *model* of  $\varphi$ . The set of all models of  $\varphi$  is again denoted as  $MD(\varphi)$ . As before we define sets  $\Phi$  of formulas to be satisfied by an interpretation  $\mathcal{J}$ if every formula in  $\Phi$  is satisfied by  $\mathcal{J}$ . Formally:  $\mathcal{J} \models \Phi$  if and only if  $I \models \varphi$  for each  $\varphi \in \Phi$ . The notation MD is extended to sets of LTL-formulas as before:

$$MD(\Phi) = \bigcap_{\varphi \in \Phi} MD(\varphi).$$

As before we will call a formula  $\varphi$  (respectively a set  $\Phi$  of formulas)

- satisfiable if  $MD(\varphi) \neq \emptyset$  (respectively  $MD(\Phi) \neq \emptyset$ ),
- valid if  $\mathcal{J} \models \varphi$  for each  $\mathcal{J}$  (respectively if every  $\varphi \in \Phi$  is valid) and
- unsatisfiable if  $MD(\varphi) = \emptyset$  (respectively  $MD(\Phi) = \emptyset$ ).

As one might already expect, there is also an extension of the *logical consequence* relation  $\models$  known from propositional logic to LTL. Again we have

$$\Phi \models \varphi$$
 if and only if  $MD(\Phi) \subseteq MD(\varphi)$ .

We will also write  $\varphi \models \psi$  if the set  $\Phi$  only consists of the single formula  $\varphi$ , that is  $\varphi \models \psi$  denotes  $\{\varphi\} \models \psi$ . The properties of  $\models$  carry over from propositional logic to LTL. Furthermore we have the following lemma.

#### Lemma 2.3.1

Let  $\Phi$  be a set of LTL-formulas and let  $\varphi$  be an LTL-formula.

- 1.  $\Phi$  is unsatisfiable if and only if  $\Phi \models \texttt{false}$ .
- 2. If  $\Phi \models \varphi$ , then there is a finite subset  $\Psi \subseteq \Phi$  such that  $\Psi \models \varphi$ .

In particular, testing for unsatisfiability can be accomplished by applying Lemma 2.3.1.

#### Corollary 2.3.1

Let  $\Phi$  be a set of LTL-formulas. If  $\Phi \models \texttt{false}$ , then there is some finite subset  $\Phi_0 \subseteq \Phi$  such that  $\Phi_0 \models \texttt{false}$ .

The relation  $\equiv$  is again extended in the obvious way: for every pair  $\varphi_1, \varphi_2$  of LTLformulas we have  $\varphi_1 \equiv \varphi_2$  if and only if  $\varphi_1 \models \varphi_2$  and  $\varphi_2 \models \varphi_1$  or equivalently if and only if  $MD(\varphi_1) = MD(\varphi_2)$ .

#### 2.3.2. First Order Temporal Logic

This last section of this chapter will deal with the extension of the propositional temporal logic LTL introduced in chapter 2.3.1 to the field of first order logic. The resulting logic will consequently be denoted as FOLTL (standing for First Order LTL).

Assume that a signature sig =  $(\mathcal{X}, F, P, \alpha)$  as defined in chapter 2.2 is given. Therefore the set  $\mathcal{T} = \mathcal{T}(\text{sig})$  is defined. We will introduce the set of FOLTL-formulas stepwise.

#### **Definition 2.3.4 (Temporal Atoms)**

The set of all *temporal atoms* over sig (denoted as  $\mathcal{A}_t(sig)$ ) is defined as the smallest set of objects closed under the following rules:

- 1. if  $\varphi \in \mathcal{A}(\text{sig})$  is a first order atomic formula, then  $\varphi \in \mathcal{A}_t(\text{sig})$ ,
- 2. if  $\varphi$  is a temporal atom from  $\mathcal{A}_t(\text{sig})$ , then  $X\varphi$ ,  $F\varphi$  and  $G\varphi$  are in  $\mathcal{A}_t(\text{sig})$  and
- 3. if  $\varphi_1$  and  $\varphi_2$  are in  $\mathcal{A}_t(\text{sig})$ , then so are  $\varphi_1 \mathsf{U} \varphi_2$  and  $\varphi_1 \mathsf{P} \varphi_2$ .

The definition of *temporal literals* is very similar to the definition of temporal atoms.

#### Definition 2.3.5 (Temporal Literals)

The set of all *temporal literals* over sig (denoted as  $\mathcal{L}_t(sig)$ ) is defined as the smallest set of objects closed under the following rules:

- 1. if  $\varphi \in \mathcal{A}_t(\text{sig})$  is a temporal atomic formula, then  $\varphi \in \mathcal{L}_t(\text{sig})$ ,
- 2. if  $\varphi$  is a temporal literal from  $\mathcal{L}_t(\text{sig})$ , then  $X\varphi$ ,  $F\varphi$  and  $G\varphi$  are in  $\mathcal{L}_t(\text{sig})$ ,
- 3. if  $\varphi \in \mathcal{L}_t(\text{sig})$  is a temporal literal, then so is  $\neg \varphi$  and
- 4. if  $\varphi_1$  and  $\varphi_2$  are in  $\mathcal{L}_t(\text{sig})$ , then so are  $\varphi_1 \mathsf{U} \varphi_2$  and  $\varphi_1 \mathsf{P} \varphi_2$ .

Formulas from FOLTL are now defined as in the case of a first order logic language.

#### Definition 2.3.6 (First Order Linear Time Temporal Logic, e.g. [3])

The set of FoLTL-formulas is the smallest set of objects closed under the following rules:

1. each  $\varphi \in \mathcal{L}_t(\text{sig})$  is a formula in FOLTL,

- 2. if  $\varphi_1$  and  $\varphi_2$  are formulas from FOLTL, then so are  $(\varphi_1 \land \varphi_2)$ ,  $(\varphi_1 \lor \varphi_2)$ ,  $(\varphi_1 \to \varphi_2)$ and  $(\varphi_1 \leftrightarrow \varphi_2)$ ,
- 3. if  $\varphi$  is a formula from FOLTL, then so are  $X\varphi$ ,  $G\varphi$  and  $F\varphi$ ,
- 4. if  $\varphi_1$  and  $\varphi_2$  are formulas from FOLTL, then so are  $\varphi_1 \mathsf{U} \varphi_2$  and  $\varphi_1 \mathsf{P} \varphi_2$  and
- 5. if  $\varphi$  is a formula from FOLTL and  $x \in \mathcal{X}$ , then  $\forall x \varphi$  and  $\exists x \varphi$  are formulas from FOLTL.

The extension of the connectives  $\wedge$  and  $\vee$  is extended to include arbitrary many formulas as described for first order logic formulas on page 9. We will also make use of the binding priority for the connectives  $\neg$ ,  $\wedge$ ,  $\vee$ ,  $\rightarrow$  and  $\rightarrow$  omitting brackets whenever this is possible.

In contrast to the propositional temporal logic LTL one can distinguish between two kinds of symbols: *rigid* symbols and *flexible* symbols. Rigid symbols are symbols which are required to be interpreted to the same operation regardless of the point of time under consideration while flexible symbols may be interpreted as different operations at different points of time. We assume that each symbol is either flexible or rigid.

The semantics of FOLTL is described by a suitable extension of the concept of *temporal interpretations* as introduced for LTL in chapter 2.3.1. We will follow notations from [3] which present an adaption of the so called *possible worlds semantics* which had been originally developed by Hintikka (see [81]) and Kripke (see [97]). An *interpretation* is given as a tuple

$$\mathcal{J} = (U_{\mathcal{J}}, S, s_0, \delta_1, \delta_2, w, \mathcal{I}),$$

where

- $U_{\mathcal{J}}$  is a nonempty set, called the *universe* of  $\mathcal{J}$ ,
- S is a set of states (also called *possible worlds*) which contains the distinguished

element  $s_0$ , the *initial state* (or *actual world*),

- $\delta_1, \delta_2 \subseteq S \times S$  are accessibility relations,
- $w: \mathcal{X} \to U_{\mathcal{J}}$  is an evaluation of the variable symbols and
- $\mathcal{I}$  is a first order interpretation for the symbols of sig which maps each symbol  $\sigma \in F \cup P$  in each state s to an operation  $\mathcal{I}(s,\sigma) : U_{\mathcal{J}}^{\alpha(\sigma)} \to U_{\mathcal{J}}$  (if  $\sigma \in F$ ) or to a predicate  $\mathcal{I}(s,\sigma) : U_{\mathcal{J}}^{\alpha(\sigma)} \to \{0,1\}$  (if  $\sigma \in P$ ).  $\mathcal{I}$  is assumed to have the following properties:
  - if  $\sigma$  is a rigid symbol, then  $\mathcal{I}(s_1, \sigma) = \mathcal{I}(s_2, \sigma)$  for every  $s_1, s_2 \in S$  and
  - if  $\sigma$  is a flexible symbol, then there are  $s_1, s_2 \in S$  such that  $\mathcal{I}(s_1, \sigma) \neq \mathcal{I}(s_2, \sigma)$ .

The evaluation of terms in such an interpretation is accomplished as expected: let  $t \in \mathcal{T}(sig)$  be given.

- 1. if  $t = x \in \mathcal{X}$ , then  $\mathcal{J}(x) = w(x)$  and
- 2. if  $t = f(t_1, \ldots, t_n)$  for some  $f \in F$  with  $\alpha(f) = n$  and  $t_1, \ldots, t_n \in \mathcal{T}(sig)$ , then

$$\mathcal{J}(t) = \mathcal{I}(s_0, f) \left( \mathcal{J}(t_1), \dots, \mathcal{J}(t_n) \right).$$

The interpretation of formulas is now defined similarly to the interpretation of formulas in first order logic.

- $\mathcal{J}(\texttt{true}) := 1$ ,
- $\mathcal{J}(\texttt{false}) := 0$  and
- if  $\varphi = p(t_1, \ldots, t_n) \in \mathcal{A}(\text{sig})$  for  $p \in P$  with  $\alpha(p) = n$  and  $t_1, \ldots, t_n \in \mathcal{T}(\text{sig})$ , then

$$\mathcal{J}(\varphi) = \mathcal{J}(p(t_1, \dots, t_n)) := \mathcal{I}(s_0, p)(\mathcal{J}(t_1), \dots, \mathcal{J}(t_n)).$$

For the connectives  $\land, \lor, \rightarrow$  and  $\leftrightarrow$  which we will call *first order connectives* from now on and the quantifiers  $\forall$  and  $\exists$  the semantics is defined as usual. Let  $\varphi, \varphi_1$  and  $\varphi_2$  be FOLTL-formulas and let  $x \in \mathcal{X}$  be a variable symbol.

- $\mathcal{J}(\neg \varphi) := 1 \mathcal{J}(\varphi),$
- $\mathcal{J}(\varphi_1 \wedge \varphi_2) := \min{\{\mathcal{J}(\varphi_1), \mathcal{J}(\varphi_2)\}},$
- $\mathcal{J}(\varphi_1 \lor \varphi_2) := \max{\{\mathcal{J}(\varphi_1), \mathcal{J}(\varphi_2)\}},\$
- $\mathcal{J}(\varphi_1 \to \varphi_2) := \max{\{\mathcal{J}(\neg \varphi_1), \mathcal{J}(\varphi_2)\}},$
- $\mathcal{J}(\varphi_1 \leftrightarrow \varphi_2) := \min \{ \mathcal{J}(\varphi_1 \to \varphi_2), \mathcal{J}(\varphi_2 \to \varphi_1) \},\$
- $\mathcal{J}(\forall x\varphi) := \min \left\{ \mathcal{J}_x^t(\varphi) \mid t \in U_{\mathcal{J}} \right\}$  and
- $\mathcal{J}(\exists x\varphi) := \max \{ \mathcal{J}_x^t(\varphi) \mid t \in U_{\mathcal{J}} \}.$

where  $\mathcal{J}_x^t$  emerges from  $\mathcal{J}$  in a similar way as in first order logic (see page 17).

What remains to be defined is the semantics of the temporal operators. This is done via the reachability relations  $\delta_1$  and  $\delta_2$  which model the *next state*-relation ( $\delta_1$ ) and its transitive closure ( $\delta_2$ ). For modeling the semantics we will need another concept. Let  $s \in S$  be any state. The interpretation  $\mathcal{J}[s]$  emerges from  $\mathcal{J}$  by setting its initial state (or its *actual world* which gives a better intuition in this case) from  $s_0$  to s. Now let  $\varphi, \varphi_1$  and  $\varphi_2$  be given.

- $\mathcal{J}(\mathsf{X}\varphi) := 1$  if and only if there is  $s_1 \in S$  such that  $s_0 \delta_1 s_1$  and  $\mathcal{J}[s_1](\varphi) = 1$ ,
- $\mathcal{J}(\mathsf{F}\varphi) := 1$  if and only if there is  $s_1 \in S$  such that  $s_0 \delta_2 s_1$  and  $\mathcal{J}[s_1](\varphi) = 1$ ,
- $\mathcal{J}(\mathsf{G}\varphi) := 1$  if and only if  $\mathcal{J}[s_1](\varphi) = 1$  for every  $s_1 \in S$  such that  $s_0 \delta_2 s_1$ ,
- $\mathcal{J}(\varphi_1 \cup \varphi_2) = 1$  if and only if for every  $s_1 \in S$  such that  $s_0 \delta_2 s_1$  it holds that  $\mathcal{J}[s_1](\varphi_1) = 1$  or there is  $s_2 \in S$  such that  $s_0 \delta_2 s_2$ ,  $s_2 \delta_2 s_1$  and  $\mathcal{J}[s_2](\varphi_2) = 1$  and

•  $\mathcal{J}(\varphi_1 \mathsf{P}\varphi_2) = 1$  if and only if there is  $s_1 \in S$  such that  $s_0\delta_2 s_1$  and  $\mathcal{J}[s_1](\varphi_1) = 1$ and for each  $s_2 \in S$  such that  $s_0\delta_2 s_2$  and  $s_2\delta_2 s_1$  it holds that  $\mathcal{J}[s_2](\neg \varphi_2) = 1$ .

As usual we will write  $\mathcal{J} \models \varphi$  if  $\mathcal{J}(\varphi) = 1$  and call  $\mathcal{J}$  a *model*. The set of all models of  $\varphi$  is again denoted as  $MD(\varphi)$ . For sets  $\Phi$  of formulas we have the obvious extension:

$$MD(\Phi) = \bigcap_{\varphi \in \Phi} MD(\varphi).$$

The notations of *satisfiability*, *validity* and *unsatisfiability* are extended in a straightforward way. To conclude the definition of the semantics of FOLTL we extend the logical consequence relation  $\models$  to FOLTL-formulas and sets of FOLTL-formulas by adjusting the notations from first order logic given on page 18.

Now that both propositional and first order linear time temporal logic is defined we are ready to introduce the remaining concepts which will be the subject of the theory to be developed in this thesis, namely Logic Programming and Inductive Logic Programming.

# 3. Logic Programming

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This chapter briefly reviews the basic concepts of Logic Programming as introduced by Kowalski (see [94] and [93]). Logic Programming is a form of *Declarative Programming* which is a programming concept based on the philosophy that the programmer should not be concerned with the way a solution is searched for by a programming system but (s)he should be allowed to concentrate on the description of the properties of the solutions. Various realizations of declarative languages, mostly functional languages such as LISP or HASKELL, have been proposed. Logic Programming is another form of declarative programming which is concerned with describing *relationships* between objects with certain properties.

## 3.1. Predicate Logic as a Programming Language

The basic objects of a logic programming language are formulas of a special type, so called *clauses*.

# Definition 3.1.1 (Clause, Robinson [140])

A clause C is a disjunction of literals  $l_i$ :

$$C = \bigvee_{i=1}^{n} l_i.$$

Assume for now that a clause C is given. C can also be seen as a set of literals  $C = \{l_1, \ldots, l_n\}$ . Since every literal is either an atom from  $\mathcal{A}$  or the negation of an atom, we can partition C into two sets Pos(C) and NEG(C) containing the *positive* literals of C (i.e. the atoms<sup>1</sup>) and the *negative* ones (i.e. the negations of atoms):

$$C = \underbrace{\{l_i \mid i \in \{1, \dots, n\}, l_i \in \mathcal{A}\}}_{=: \operatorname{Pos}(C)} \cup \underbrace{\{l_i \mid i \in \{1, \dots, n\}, \neg l_i \in \mathcal{A}\}}_{=: \operatorname{Neg}(C)}$$

Now we can identify several classes of clauses: C is called a

- hornclause if and only if  $|Pos(C)| \le 1$ ,
- definite hornclause if and only if |Pos(C)| = 1 and
- unit clause (or fact) if and only if |Pos(C)| = 1 and  $Neg(C) = \emptyset$ .

A goal (or a query) G is a finite sequence of atoms which are considered to be conjunctively connected:  $G = G_1 \wedge \cdots \wedge G_m$ . The atoms  $G_i$  are called the *subgoals of* G.

The distinguished clause which neither contains positive nor negative literals, and which is therefore represented as the set  $\{\}$ , is called the *empty clause* which we will denote as  $\Box^2$ . The empty clause is considered equivalent to any unsatisfiable first order formula.

<sup>&</sup>lt;sup>1</sup>Here we identify  $\neg \neg \varphi$  and  $\varphi$ .

<sup>&</sup>lt;sup>2</sup>Note that  $\Box$  is also a goal, namely the goal which does not contain any subgoals. We will therefore also refer to  $\Box$  as the *empty goal*.

The philosophy of Logic Programming is the following (see e.g. [94] and [164]):

- Statements which are known to be correct are modeled by unit clauses (facts).
- Relations between objects are modeled by definite horn clauses (which are interpreted as *rules*).
- Program calls are modeled by goals.

We will assume that every rule represents a hornclause which is *implicitly universally* closed, that is every variable symbol which occurs in a rule is assumed to be inside the scope of a universal quantifier. Formally this means that if

$$C = \varphi_1 \wedge \dots \wedge \varphi_n \to \psi$$

is a rule which contains the variables  $\{x_1, \ldots, x_m\}$  then we merely work with the formula

$$\forall x_1 \dots \forall x_m (\varphi_1 \wedge \dots \wedge \varphi_n \to \psi).$$

#### Example 3.1.1

Let sig be a signature which contains a relation symbol is\_even with  $\alpha(is_even) = 1$ and function symbols null and s with  $\alpha(null) = 0$  and  $\alpha(s) = 1$ . Then the concept of even numbers is modeled by the following set of definite horn clauses:

$$C_1 = is\_even(null) and$$
  
 $C_2 = is\_even(x) \rightarrow is\_even(s(s(x)))$ 

Programs in a logic programming language are now given as sets of facts and rules.

## Definition 3.1.2 (Logic Program, Lloyd [105])

Let sig =  $(\mathcal{X}, F, P, \alpha)$  be a signature. A (logic) *program* over sig is a finite set of definite hornclauses over sig.

#### Example 3.1.2

Let P be the set =  $\{C_1, C_2\}$  from Example 3.1.1. Then P is a logic program over the signature given there.

Due to the special structure of definite hornclauses and goals one can introduce special notations for them. Assume that C is a definite horn clause and  $Pos(C) = \{A\}$ ,  $NEG(C) = \{\neg B_1, \ldots, \neg B_n\}$  for  $A, B_1, \ldots, B_n \in \mathcal{A}$ . Then we have

$$C = \{A, \neg B_1, \dots, \neg B_n\}$$
$$= A \lor \neg B_1 \lor \dots \lor \neg B_n$$
$$\equiv A \lor \neg (B_1 \land \dots \land B_n)$$
$$\equiv (B_1 \land \dots \land B_n) \to A$$
$$=: A \leftarrow B_1, \dots, B_n.$$

This can be seen as a *procedure declaration* for a procedure labeled A as described by Kowalski in [94]. The interpretation is then given as follows:

To solve A, solve 
$$B_1, B_2, \ldots, B_n!$$

Now assume that G is a goal consisting of the subgoals  $G_i$ , that is  $G = G_1 \wedge \cdots \wedge G_m$ . By analogy we have

$$\neg G = \neg (G_1 \land \dots \land G_m)$$
$$\equiv \neg G_1 \lor \dots \lor \neg G_m$$
$$=: \leftarrow G_1, \dots, G_m.$$

Using the procedural interpretation from above we can see a goal as the statement

Solve 
$$G_1, G_2, \ldots, G_m!$$

Now if one wants to run a logic program a goal G is added to the program and a

theorem proving procedure tests if the goal is a logical consequence of the program under consideration. The task how this is achieved is the subject of the next section of this chapter.

### 3.2. The Concept of SLD-Resolution

This section will be concerned with a brief explanation of the concept of *SLD-resolution* (originally introduced in [95]) which is a theorem proving procedure designed to handle (definite) hornclauses. It is a refinement of the classical *Resolution* procedure introduced by Robinson (see [140]). Since there are many excellent texts on theorem proving in general (and especially on the topic of logic programming) the discussion will be rather short. We refer the interested reader to the literature (see [31], [8], [7], [105] and [126]).

The key result for understanding SLD–Resolution is given by the following lemma which is a special case of Proposition 3.1 from [105].

#### Lemma 3.2.1 (Lloyd [105])

Let P be a logic program and let G be a goal. Then  $P \models G$  if and only if  $P \cup \{\leftarrow G\} \models \Box$ .

This lemma can also be seen as an easy consequence of Theorem 2.2.3. So if  $G_1 \wedge \cdots \wedge G_m$  is a logical consequence of P one only has to deduce the empty clause from  $P \cup \{\neg (G_1 \wedge \cdots \wedge G_m)\}$ . This is achieved by application of the principle of SLD-resolution which we will define now.

Several other approaches for implementing logical programming languages have been proposed. In principle it is possible to take *any* complete proof procedure for first order logic in order to achieve this goal. But due to implementation difficulties and performance problems one concentrates on refutation–complete calculi. Popular approaches are based on tableaux techniques (see e.g. [67] and [4] for calculi for definite logic programs and [19] and [130] for calculi for disjunctive logic programming languages) and the Model Elimination Technique as described by Loveland in [107] and [108] (see e.g. [70]). However, the SLD-resolution approach has been the first method to be implemented in PROLOG systems and is therefore still a dominant procedure in logic programming systems. Although SLD–Resolution is sound and refutation–complete many implementations omit certain operations which are necessary in order to guarantee these properties. In particular, the occur-check which is necessary during the unification process is a very expensive operation and therefore many implementations skip this check. Although there are programs which allow skipping this check (see [9]) omitting the check results in losing the property of soundness (see e.g. [112]). Another expensive operation is the breadth-first search strategy which is necessary in order to guarantee refutation-completeness. Most PRO-LOG systems simply carry out one inference step (namely the first one which is possible) without trying other steps which might be applicable. This might result in non-halting derivations which have trivial solutions.

In order to reason about instantiations of formulas one has to define a suitable concept of *substitution*. Intuitively a substitution replaces variables by terms. Formally a substitution is defined as a certain type of homomorphism on terms and formulas.

**Definition 3.2.1 (Substitution, Robinson [140])** A substitution is a mapping  $\sigma : \mathcal{X} \to \mathcal{T}$  such that  $\{x \in \mathcal{X} \mid \sigma(x) \neq x\}$  is finite.

Since the set of variables which are changed by the substitution  $\sigma$  is required to be finite, we can write down substitutions by stating which variables are replaced by which terms and omitting the variables which remain unchanged. The set of all these variables will be called the *domain* of the substitution  $\sigma$  and will be denoted as  $DOM(\sigma)$ . Assume that  $\text{Dom}(\sigma) = \{x \mid \sigma(x) \neq x\} = \{x_{i_1}, \dots, x_{i_n}\}$  and that  $\sigma(x_{i_j}) = t_j \in \mathcal{T}$  for  $j = t_j \in \mathcal{T}$  $1, \ldots, n$ . Then  $\sigma$  will be identified by the set of *bindings* 

$$\left\{\frac{x_{i_1}}{t_1},\ldots,\frac{x_{i_n}}{t_n}\right\} = \left\{\frac{x_{i_1}}{\sigma(x_{i_1})},\ldots,\frac{x_{i_n}}{\sigma(x_{i_n})}\right\}$$

where each of the *n* bindings  $\frac{x_{i_j}}{t_j}$  denotes the substitution  $\sigma_j$  with

$$\sigma_j(x) = \begin{cases} t_j & \Leftrightarrow x = x_{i_j} \\ x & \Leftrightarrow \text{else} \end{cases}$$

Substitutions are extended to terms and formulas by defining homomorphic extensions:

$$\sigma(f(t_1, \dots, t_n)) = f(\sigma(t_1), \dots, \sigma(t_n)),$$

$$\sigma(p(t_1, \dots, t_n)) = p(\sigma(t_1), \dots, \sigma(t_n)),$$

$$\sigma(\neg \varphi) = \neg \sigma(\varphi),$$

$$\sigma(\varphi_1 \land \varphi_2) = \sigma(\varphi_1) \land \sigma(\varphi_2),$$

$$\sigma(\varphi_1 \lor \varphi_2) = \sigma(\varphi_1) \lor \sigma(\varphi_2),$$

$$\sigma(\varphi_1 \rightarrow \varphi_2) = \sigma(\varphi_1) \rightarrow \sigma(\varphi_2),$$

$$\sigma(\varphi_1 \leftrightarrow \varphi_2) = \sigma(\varphi_1) \leftrightarrow \sigma(\varphi_2),$$

$$\sigma(\varphi_1 \leftrightarrow \varphi_2) = \sigma(\varphi_1) \leftrightarrow \sigma(\varphi_2),$$

$$\sigma(\forall x \varphi) = \forall x \sigma(\varphi) \text{ and}$$

$$\sigma(\exists x \varphi) = \exists x \sigma(\varphi).$$

Substitutions can be *composed* in order to build *complex* substitutions from *simpler* ones. This will be interesting for us in the following part of this section when we will define the result computed by a logic program P given a goal G as input.

Definition 3.2.2 (Composition of Substitutions, Robinson [140]) Let  $\sigma_1 = \left\{ \frac{x_1^{(1)}}{t_1^{(1)}}, \dots, \frac{x_n^{(1)}}{t_n^{(1)}} \right\}$  and  $\sigma_2 = \left\{ \frac{x_1^{(2)}}{t_1^{(2)}}, \dots, \frac{x_m^{(2)}}{t_m^{(2)}} \right\}$  be substitutions. Then the substitution  $\sigma_1 \circ \sigma_2$  (the *composition* of  $\sigma_1$  and  $\sigma_2$ ) is defined as follows:

$$\sigma_1 \circ \sigma_2 = \left\{ \frac{x_1^{(1)}}{\sigma_2\left(t_1^{(1)}\right)}, \dots, \frac{x_n^{(1)}}{\sigma_2\left(t_n^{(1)}\right)}, \frac{x_1^{(2)}}{t_1^{(2)}}, \dots, \frac{x_m^{(2)}}{t_m^{(2)}} \right\}$$

$$\setminus \left( \left\{ \frac{x_i^{(1)}}{\sigma_2\left(t_i^{(1)}\right)} \mid x_i^{(1)} = \sigma_2\left(t_i^{(1)}\right) \right\} \cup \left\{ \frac{x_j^{(2)}}{t_j^{(2)}} \mid x_j^{(2)} \in \left\{ x_1^{(1)}, \dots, x_n^{(1)} \right\} \right\} \right)$$

That is in order to compute the composition of  $\sigma_1$  and  $\sigma_2$  one first applies  $\sigma_2$  to the terms which  $\sigma_1$  replaces for the variables in its domain, then adds the bindings of  $\sigma_2$  (in order to add the elements which are not yet in the domain of  $\sigma_1$ ) and then reduces the resulting set by removing identical bindings and such bindings which won't have any effect.

We will identify a distinguished element  $\varepsilon$  as the substitution which does not contain any binding and so will have no effect on the objects to which it is applied. This element is given as  $\varepsilon = \emptyset$  and is denoted as the *empty substitution*. The composition operation  $\circ$ enjoys the following properties (Proposition 4.1 from [105]):

#### Lemma 3.2.2 (Properties of Substitutions, Lloyd [105])

Let  $\sigma_1, \sigma_2$  and  $\sigma_3$  be substitutions, let t be a term and let  $\varphi$  be a formula. Then

- 1.  $\sigma_1 \circ \varepsilon = \varepsilon \circ \sigma_1 = \sigma_1$ ,
- 2.  $\sigma_2(\sigma_1(t)) = (\sigma_1 \circ \sigma_2)(t),$
- 3.  $\sigma_2(\sigma_1(\varphi)) = (\sigma_1 \circ \sigma_2)(\varphi)$  and
- 4.  $(\sigma_1 \circ \sigma_2) \circ \sigma_3 = \sigma_1 \circ (\sigma_2 \circ \sigma_3).$

Let  $\varphi_1$  and  $\varphi_2$  be arbitrary formulas from  $\mathcal{F}$ . We say that  $\varphi_1$  and  $\varphi_2$  are variants if there are substitutions  $\sigma_1$  and  $\sigma_2$  such that  $\sigma_1(\varphi_1) = \varphi_2$  and  $\sigma_2(\varphi_2) = \sigma_1$ .

Now let  $t_1, t_2$  be terms and let  $\varphi_1, \varphi_2$  be literals. Substitutions  $\sigma$  which yield syntactically identical objects, i.e. substitutions  $\sigma$  such that  $\sigma(t_1) = \sigma(t_2)$  or  $\sigma(\varphi_1) = \sigma(\varphi_2)$  play an important role.

#### Definition 3.2.3 (Unification)

Let  $t_1$  and  $t_2$  be terms, let  $\varphi_1$  and  $\varphi_2$  be literals.  $t_1$  and  $t_2$  (respectively  $\varphi_1$  and  $\varphi_2$ ) are said to be *unifiable* if and only if there is a substitution  $\sigma$  such that  $\sigma(t_1) = \sigma(t_2)$ (respectively  $\sigma(\varphi_1) = \sigma(\varphi_2)$ ).  $\sigma$  is called a *unifier*.

Some unifiers can be characterized as unifiers which only change the objects to be unified as much as necessary.

#### Definition 3.2.4 (Most General Unifier)

Let  $t_1$  and  $t_2$  be terms and let  $\varphi_1$  and  $\varphi_2$  be literals. A unifier  $\sigma$  for  $t_1$  and  $t_2$ (respectively for  $\varphi_1$  and  $\varphi_2$ ) is called a *most general unifier* (or *mgu*) if and only if for every unifier  $\sigma_1$  for  $t_1$  and  $t_2$  (respectively  $\varphi_1$  and  $\varphi_2$ ) there is a substitution  $\sigma_2$ such that  $\sigma = \sigma_1 \circ \sigma_2$ . We will denote this by writing  $\sigma = \text{mgu}(t_1, t_2)$  (respectively  $\sigma = \text{mgu}(\varphi_1, \varphi_2)$ ).

Having defined which properties a most general unifier satisfies it is necessary to ensure that in the case of unifiable objects there is indeed an algorithm which can compute such a substitution. The following lemma from [140] ensures this.

#### Lemma 3.2.3 (Unification Theorem, Robinson [140])

Let  $t_1$  and  $t_2$  be terms and let  $\varphi_1$  and  $\varphi_2$  be literals. If  $t_1$  and  $t_2$  (respectively  $\varphi_1$  and  $\varphi_2$ ) are unifiable, then there exists a mgu of  $t_1$  and  $t_2$  (respectively  $\varphi_1$  and  $\varphi_2$ ) which is uniquely determined up to renaming of variables and which can be effectively computed.

Several algorithms have been proposed for computing most general unifiers. The first and also most simple one was presented by Robinson in 1965 (see [140]) which has the drawback that its worst case runtime is exponential in the length of the objects to be identified. It has also been shown that the unification problem is solvable in linear time by Paterson and Wegman which presented an algorithm which operates on directed acyclic graphs representing the objects (see [129]). However, due to its simplicity another algorithm introduced by Martelli and Montanari in 1982 (see [113]) is used very often in Logic Programming systems. We will refine this algorithm in order to be capable of handling temporal objects in a later chapter.

Now let  $P = \{P_1, \ldots, P_n\}$  be a logic program. Furthermore assume that  $G = G_1 \wedge \cdots \wedge G_m$  is a goal. Due to the special structure of definite horn clauses we can assume that each  $P_i$  can be written as

$$P_i = A_i \leftarrow B_1^{(i)}, \dots, B_{n_i}^{(i)}$$

for some  $n_i$  and  $A_i, B_1^{(i)}, \ldots, B_{n_i}^{(i)} \in \mathcal{A}$ . Assume that *i* and *j* are such that  $G_j$  and  $A_i$  are unifiable with  $\sigma = \text{mgu}(A_i, G_j)$ . The resolvent (or *SLD*-resolvent) of *G* and  $P_i$  with respect to  $\sigma$  is the goal

$$\sigma\left(G_{1}\wedge\cdots\wedge G_{j-1}\wedge B_{1}^{(i)}\wedge\cdots,B_{n_{i}}^{(i)}\wedge G_{j+1}\wedge\cdots\wedge G_{m}\right)$$

$$= \sigma\left(G_{1}\right)\wedge\cdots\wedge\sigma\left(G_{j-1}\right)\wedge\sigma\left(B_{1}^{(i)}\right)\wedge\cdots\wedge\sigma\left(B_{n_{i}}^{(i)}\right)\wedge\sigma\left(G_{j+1}\right)\wedge\cdots\wedge\sigma\left(G_{m}\right)$$

$$=:\left(G_{1}\wedge\cdots\wedge G_{j-1}\wedge B_{1}^{(i)}\wedge\cdots\wedge B_{n_{i}}^{(1)}\wedge G_{j+1}\wedge\cdots\wedge G_{m}\right)\sigma$$

#### Definition 3.2.5 (SLD-derivation-step, Kowalski and Kuehner [95])

Let P be a logic program and let  $G = G_1 \land \cdots \land G_m$  be a goal. An *SLD-derivation-step* is a sequence of actions carried out as follows:

- 1. A subgoal  $G_i$  is chosen.
- 2. A clause  $A \leftarrow B_1, \ldots, B_n$  from P is chosen such that A and  $G_i$  are unifiable. Let  $\sigma$  be a most general unifier for A and  $G_i$ .
- 3. The SLD-resolvent G' of G and  $A \leftarrow B_1, \ldots, B_m$  with respect to  $\sigma$  is constructed.

Given P we will write  $G \vdash_{\text{Res}} G'$  if there is an SLD-derivation-step yielding G' from G. The relation  $\vdash$  is then defined as the reflexive-transitive closure of  $\vdash_{\text{Res}}$ :

$$\begin{array}{lll} G \stackrel{0}{\vdash} G' & :\Leftrightarrow & G = G', \\ G \stackrel{1}{\vdash} G' & :\Leftrightarrow & G \vdash_{\operatorname{Res}} G' \text{ and} \\ G \stackrel{n+1}{\vdash} G' & :\Leftrightarrow & \operatorname{there} \operatorname{is} \bar{G} \text{ such that } G \stackrel{n}{\vdash} \bar{G} \text{ and } \bar{G} \stackrel{1}{\vdash} G' \end{array}$$

So we can define  $G \vdash G'$  if there is some  $n \ge 0$  such that  $G \stackrel{n}{\vdash} G'$ . In other words  $\vdash = \vdash_{\text{Res}}^*$ .

An *SLD-derivation* of *G* from *P* is a sequence  $(G_i)_{i\geq 0}$  such that  $G_0 = G$  and for each *i* there is a clause from *P* which can be used in order to carry out an SLD-derivation-step yielding  $G_{i+1}$ .

An SLD-derivation  $(G_i)_{i\geq 0}$  of G from P which is of finite length, say  $(G_i)_{i=1}^n$  is called an *SLD-derivation of*  $G_n$  from P given input G.

An SLD-derivation  $(G_i)_{i=1}^n$  of  $G_n$  from P given input  $G_0$  is called *successful* if  $G_n = \Box$ . In the case that there is a successful SLD-derivation of  $\Box$  from P given input G we will write  $P \vdash G$ , otherwise we will write  $P \nvDash G$ . Furthermore we will call successful SLD-derivations of  $\Box$  from P given input G SLD-derivations of  $\leftarrow G$  from P. Sometimes we will also refer to SLD-derivations of goals from P given input G as SLD-derivations of  $P \cup \{\leftarrow G\}$ .

So given a program P and a goal G one can construct all possible derivations of  $P \cup \{\leftarrow G\}$  and check if there is a refutation of  $P \cup \{\leftarrow G\}$ . If such a refutation exists, then it can be found by breadth-first-search. If no such refutation exists, then in general there is no way to detect this since the fragment of clausal logic is undecidable (see [142]).

#### Example 3.2.1

Again consider the program from Examples 3.1.1 and 3.1.2. Assume that the goal G is given by  $G = is\_even(s(s(s(s(null))))))$ . Then an SLD-refutation of  $P \cup \{\leftarrow G\}$  can be visualized as depicted in Figure 3.1 where an arrow between two goals means that the

```
is_even(s(s(s(null)))))
     ..s(nul)
↓
is_even(null)
↓
  is_even(s(s(null)))
```

Figure 3.1.: SLD-refutation

second goal is a resolvent of the first one (with respect to a suitable mgu).

#### 3.3. Soundness and Completeness

As we have seen computations by logic programs are carried out by giving a goal G to a program P and proving that the set  $P \cup \{\leftarrow G\}$  is unsatisfiable, that is proving that  $P \cup \{\leftarrow G\} \vdash \Box$  holds. The *result* of such a computation is given by the substitution which emerges from composing the unifiers used in this refutation. Let  $\sigma$  be this substitution. We first state the following lemma:

#### Lemma 3.3.1 (Soundness, Apt and v. Emden [10])

The SLD-resolution rule is sound. That is if  $P \vdash \varphi$ , then  $P \models \varphi$ .

This lemma is proved directly in [10] but it is also a consequence of the soundness of the general resolution rule presented in [140].

So assume that  $P \cup \{\leftarrow G\} \vdash \Box$ . Then we have  $P \cup \{\leftarrow G\} \models \Box$  and due to Lemma 3.2.1 we have  $P \models G$ . Since G is a goal, we have  $G \models G_1, \ldots, G_m$  for atoms  $G_i$  and therefore  $P \models G_1 \land \cdots \land G_m$ .

We introduce two concepts of substitutions which will turn out to be useful.

**Definition 3.3.1 (Answer, Lloyd [105])** Let P be a logic program and let  $G = G_1, \ldots, G_m$  be a goal.

- 1. An answer for  $P \cup \{\leftarrow G\}$  is a substitution for the variables occurring in G.
- 2. A correct answer for  $P \cup \{\leftarrow G\}$  is an answer  $\sigma$  such that  $P \models (G_1 \land \cdots \land G_m) \sigma$ .
- A computed answer for P∪{← G} is the composition of the most general unifiers used in a refutation of P ∪ {← G}.

So a computed answer for  $P \cup \{\leftarrow G\}$  can be seen as a result computed by the program P given a goal G as *input*. Indeed, this interpretation of logic programs and goals is an adequate way to carry out computations as the next theorem shows.

# Theorem 3.3.1 (Refutation–Completeness, v. Emden and Kowalski [164]) Let P be a logic program and let G be a goal.

- 1. Every computed answer for  $P \cup \{\leftarrow G\}$  is a correct answer for  $P \cup \{\leftarrow G\}$ .
- 2. For every correct answer  $\sigma$  for  $P \cup \{\leftarrow G\}$  there is a computed answer  $\sigma_1$  for  $P \cup \{\leftarrow G\}$  and a substitution  $\sigma_2$  such that  $\sigma = \sigma_1 \circ \sigma_2$ .

So if an answer is computed by using SLD-resolution this answer is a correct solution of the problem modeled by the program P under consideration (soundness). Additionally it is possible to compute any answer which can be instantiated to a correct one (completeness). Furthermore it is possible to model every *computable* function by a suitable logic program (see [5]). So predicate logic can indeed be seen as an adequate formalism for computation.

# 4. Inductive Logic Programming

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This chapter will be concerned with a brief introduction of the concepts from *Inductive Logic Programming* (or *ILP* for short) which we will extend to the temporal logic to be defined in the next part of this thesis. Therefore we will first present a description of the tasks which ILP systems have to perform. After this we will introduce several important concepts from lattice theory which will be used throughout the next chapters.

#### 4.1. The basic Framework

Inductive Logic Programming is concerned with synthesizing general rules from examples. Henceforth it is a special case of the theory of algorithmic learning. In general algorithmic learning is a generic term for every theory which is concerned with determining explanations for certain phenomena. In ILP, the objects (or concepts) to be learned are logic programs. The hints about the concept to be learned are given by sets  $\mathcal{E}^+$ and  $\mathcal{E}^-$  consisting of ground atoms. These sets are considered to be examples for the (unknown) program P to be learned.  $\mathcal{E}^+$  contains the positive examples and  $\mathcal{E}^-$  contains the negative ones. The interpretation of positive end negative examples is then given by

- $P \models e_+$  for every  $e_+ \in \mathcal{E}^+$  and
- $P \not\models e_{-}$  for every  $e_{-} \in \mathcal{E}^{-}$ .

The problem setting which we will use here is usually referred to as the normal problem setting for  $ILP^1$ . Some properties of a program P and sets  $\mathcal{E}^+$  and  $\mathcal{E}^-$  can be defined formally as follows (following [126]).

#### Definition 4.1.1 (Properties of Programs, e.g. [126])

Let P be a logic program and let  $\mathcal{E}^+$  and  $\mathcal{E}^-$  be (finite) sets of ground atoms. P is called

- complete wrt. E<sup>+</sup> if P ⊨ e<sub>+</sub> for every e<sub>+</sub> ∈ E<sup>+</sup>,
  consistent wrt. E<sup>-</sup> if P ∪ {¬e<sub>-</sub> | e<sub>-</sub> ∈ E<sup>-</sup>} ⊭ □ and
  correct wrt. E<sup>+</sup> and E<sup>-</sup> if P is complete wrt. E<sup>+</sup> and consistent wrt. E<sup>-</sup>.

Additionally the following definitions allow a closer classification of programs relative to given sets of examples.

#### Definition 4.1.2 (Further Properties, e.g. [126])

Let P be a logic program and let  $\mathcal{E}^+$  and  $\mathcal{E}^-$  be sets of positive and negative examples. P is called

- too strong wrt. \$\mathcal{E}^-\$ if \$P\$ is not consistent wrt. \$\mathcal{E}^-\$,
  too weak wrt. \$\mathcal{E}^+\$ if \$P\$ is not complete wrt. \$\mathcal{E}^+\$,
  overly general wrt. \$\mathcal{E}^+\$ and \$\mathcal{E}^-\$ if \$P\$ is complete wrt. \$\mathcal{E}^+\$ and not consistent wrt.  $\mathcal{E}^{-}$  and

<sup>&</sup>lt;sup>1</sup>In contrast to the normal setting the nonomontonic setting for ILP has been defined (see [79]).

overly specific wrt. E<sup>+</sup> and E<sup>-</sup> if P is consistent wrt. E<sup>-</sup> and not complete wrt.
 E<sup>+</sup>.

The task, which an ILP system has to perform is to find (or *synthesize*) a program P which is correct wrt. given sets  $\mathcal{E}^+$  and  $\mathcal{E}^-$ .

Usually this normal setting is extended in a way that the usage of background knowledge is possible. Background knowledge is given as a finite set  $\mathcal{B}$  of clauses which are interpreted as rules which are known to be correct. The program P to be synthesized is now required to have the following properties:

- $P \cup \mathcal{B} \models e_+$  for every  $e_+ \in \mathcal{E}^+$  and
- $P \cup \mathcal{B} \not\models e_{-}$  for every  $e_{-} \in \mathcal{E}^{-}$ .

It is possible to distinguish between several ways the examples are presented to an ILP system.

- **Batch / Incremental** In a *batch learning system* the examples from  $\mathcal{E}^+$  and  $\mathcal{E}^-$  are given to the system at the beginning of the learning task. In contrast, an *incremental learning system* receives the examples at different points of time. An example for a batch learning system is FOIL (see [137]).
- **Top down / Bottom up** The distinction between top down and bottom up systems comes from the *direction* in which the system searches for a correct program. While in a top down system an overly general set  $P \cup \mathcal{B}$  is *specialized*, in a bottom up system an overly specific set  $P \cup \mathcal{B}$  is *generalized*. Generalization and specialization will be treated in depth in a later chapter taking temporal literals and clauses into account.
- Interactive / Noninteractive An interactive system is capable of interacting with the user. Therefore such a system can ask if some assumptions it has generated while

searching for a correct program are fulfilled or not. This allows the generation of *better* programs.

In all cases, the program which is generated from the sets  $\mathcal{E}^+$ ,  $\mathcal{E}^-$  and  $\mathcal{B}$  cannot be guaranteed to be exactly the program which the person who has generated the examples has in mind. Rather it is (in most cases<sup>2</sup>) possible to construct a program which is correct with respect to examples seen so far and which has a very high probability of being correct for *other examples* which have not yet been seen.

#### 4.2. Generalization and Specialization

In order to modify a logic program to fit its specification one has to refine the program by manipulating clauses in order to imply more or less atoms. The basis of all refinement operations is some fixed generality ordering, mostly the subsumption ordering which is both relatively powerful and still decidable (in contrast to the ordering induced by logical implication which in general is undecidable). We will here restrict our attention on refinement operations on pairs of clauses. Assume that a quasi ordering  $\preceq^3$  on the set of clauses over some signature is given. Further assume that  $C_1$  and  $C_2$  are clauses. A clause C is then called (following [133])

- a generalization of  $C_1$  and  $C_2$  if  $C \succeq C_1$  and  $C \succeq C_2$  and
- a specialization of  $C_1$  and  $C_2$  if  $C \leq C_1$  and  $C \leq C_2$ .

Certain specializations and generalizations are of special interest in ILP. A clause C is called

• a least generalization of  $C_1$  and  $C_2$  if C is a generalization of  $C_1$  and  $C_2$  and for every generalization D of  $C_1$  and  $C_2$  it holds that  $D \succeq C$  and

<sup>&</sup>lt;sup>2</sup>Note that there may exist (nontrivial)  $\mathcal{E}^+$  and  $\mathcal{E}^-$  such that no correct P may exist, see [126] for a proof of this.

<sup>&</sup>lt;sup>3</sup>  $\preceq$  is called a *quasi-ordering* if  $\preceq$  is reflexive and transitive. Given  $\preceq$  the notation  $\succeq$  will be used as expected. We will write  $\approx$  if both  $\preceq$  and  $\succeq$  holds and  $\prec$  (resp.  $\succ$ ) if  $\preceq$  (res.  $\succeq$ ) and not  $\approx$  holds for pairs of objects of the underlying set.

• a greatest specialization of  $C_1$  and  $C_2$  if C is a specialization of  $C_1$  and  $C_2$  and for every specialization D of  $C_1$  and  $C_2$  it holds that  $C \succeq D$ .

A pair  $(S, \preceq)$  consisting of a nonempty set S and a quasi-ordering  $\preceq$  is called a *lattice* if for every pair  $x_1, x_2$  of elements from S there exists a least generalization and a greatest specialization wrt.  $\preceq$  in S.

Assuming that the chosen ordering  $\leq$  yields a lattice structure<sup>4</sup>, operations for refinement can be implemented in two ways:

- **Upward Refinement** Given  $C_1$  and  $C_2$  construct a least generalization of  $C_1$  and  $C_2$  wrt.  $\leq$ .
- **Downward Refinement** Given  $C_1$  and  $C_2$  construct a greatest specialization of  $C_1$  and  $C_2$  wrt.  $\leq$ .

#### 4.3. Refinement Operators and their Properties

As we have described in chapter 4.2, specialization and generalization are central operations which any ILP system has to perform. An algorithm which is capable of generalizing and/or specializing clauses is called a *refinement operator*. Consequently one can distinguish between *Upward Refinement Operators* and *Downward Refinement Operators* depending on the *direction* in which the refinement is performed, that is depending on the question whether it constructs a generalization or a specialization of the input clauses.

Formally we assume that a generality ordering  $\leq$  (a quasi ordering) is given as described above. A *downward refinement operator* is a function  $\rho_d$  mapping clauses to sets of clauses such that for every clause C it holds that

 $\rho_d(C) \subseteq \{D \mid C \succeq D\}.$ 

<sup>&</sup>lt;sup>4</sup>In chapter 6 we will see that there are indeed orderings which have this property.

Consequently an *upward refinement operator* is given as a function  $\rho_u$  mapping clauses to sets of clauses such that for each clause C it holds that

$$\rho_u(C) \subseteq \{D \mid D \succeq C\}.$$

Let  $\rho$  be any refinement operator (upward or downward) and let C be a clause. According to the notational conventions from [126] we define *1-step-refinement*, *n-step-refinements* and *refinements* as follows:

$$\begin{split} \rho^1(C) &= \rho(C) \quad (1\text{-step-refinement}), \\ \rho^n(C) &= \rho(\rho^{n-1}(C)), n > 1 \quad (n\text{-step-refinement}) \text{ and} \\ \rho^*(C) &= \bigcup_{i \ge 1} \rho^i(C) \quad (\text{refinement}) \end{split}$$

 $\rho$  is called

- locally finite if for every C the set  $\rho(C)$  is finite and computable,
- complete if for every pair  $C_1, C_2$  such that  $C_1 \succ C_2$  (or  $C_1 \prec C_2$  for upward refinement operators) there is an element  $C \in \rho^*(C_1)$  such that  $C \approx C_2$ ,
- proper if for every C it holds that  $\rho(C) \subseteq \{D \mid D \succ C\}$  (or  $\rho(C) \subseteq \{D \mid D \prec C\}$ ) and
- *ideal* if  $\rho$  is locally finite, complete and proper.

Ideality seems to be a desirable property of refinement operators. However, in general it is not possible to guarantee the existence of such ideal refinement operators (see for example [162]). Part II.

# First Order Inductive Temporal Logic Programming

# 5. The Programming Language Prolog(+T)

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We will now introduce the programming language of interest. This language will be similar to PROLOG (see [154] for an introduction) but will allow the usage of temporal operators. Therefore we will call it PROLOG(+T). PROLOG(+T) will allow list processing (as PROLOG does) but *constraints* have not yet been integrated.

## 5.1. Syntax of Prolog(+T)

### 5.1.1. Terms in Prolog(+T)

Since PROLOG(+T) is essentially a logic programming language which is enriched with the temporal operators X, G, F, U and P, the basic concepts are the same as described in chapter 3.1. However, to become a *practically usable* language, the usage of terms has to be simplified in order to allow more comfortable manipulating techniques. In particular, PROLOG(+T) allows *lists* as terms which is probably the most comfortable feature of PROLOG.

## Definition 5.1.1 (PROLOG(+T)-terms)

Let sig =  $(\mathcal{X}, F, P, \alpha)$  be a signature. The set of PROLOG(+T)-terms over sig is defined to be the smallest set which is closed under the following rules:

- 1. \_ is a PROLOG(+T)-term,
- 2. every string representing an *integer* is a PROLOG(+T)-term,
- 3. every term  $t \in \mathcal{T}(sig)$  is a PROLOG(+T)-term,
- 4. if  $t_1, \ldots, t_n$  is a finite (possibly empty) sequence of PROLOG(+T)-terms, then  $[t_1, \ldots, t_n]$  is a PROLOG(+T)-term and
- 5. if t is a PROLOG(+T)-term and  $t_1, \ldots, t_n$  is a finite (possibly empty) sequence of PROLOG(+T)-terms, then  $[t|t_1, \ldots, t_n]$  is a PROLOG(T)-term.

We will make use of the following *convention*: Variable identifiers will start with upper case letters while other identifiers such as function symbols and predicate symbols have to start with lower case letters.

The term \_ is referred to as an *anonymous variable*. Anonymous variables are a concept which is recently used in PROLOG-programs in order to define predicates which have variables as parameters which are not used in the definition. An example might be the definition of the nonnegative subtraction function:

nndiff
$$(x, y) = \begin{cases} x - y & \Leftrightarrow x \ge y \\ 0 & \text{else} \end{cases}$$

If x is a variable symbol and O is a term representing the natural number O, then the following 3-ary predicate might be *part* of the definition of the function nndiff :  $\mathbb{N}^2 \to \mathbb{N}$ :

$$p_{nndiff}(0, X, 0)$$

This models the following part of the definition of nndiff: nndiff(0, x) = 0 for every value of x. Consequently, the value of x itself does not play any role in this case. So the occurrence of x can be replaced by the anonymous variable \_ here which yields the predicate

$$p_{nndiff}(0, \_, 0).$$

In general, anonymous variables might be used whenever one is only concerned with proving that a solution *exists* without being interested in the actual value of this solution.

Integers are included in order to yield more comfortable programming facilities. It is obvious that integers are not essential for the completeness of the language since it is possible to define the natural numbers  $\mathbb{N}$  (and therefore the integers  $\mathbb{Z}$ ) in terms of a constant symbol null and an unary function **s** realizing the successor function.

The terms built up using the last two points from the above definition are referred to as *lists*. They represent *collections* of elements. The term [] represents the so called empty list which does not contain any elements. If a list can be described by completely writing down all its elements, then it might be realized as a list of the form  $[t_1, \ldots, t_n]$ . Note that the empty list is just a special case of such a list and also note that some of the elements  $t_1, \ldots, t_n$  might again be lists since the definition of PROLOG(+T)-terms is recursive. So the following construct is a well-formed PROLOG(+T)-term<sup>1</sup>:

$$[\texttt{f}(\texttt{X}),\texttt{g}(\texttt{f}(\texttt{X}),\texttt{a}),[],[\texttt{Y}|\texttt{a},\texttt{b},\texttt{c}]].$$

In lists of the form  $t = [\bar{t}|t_1, \ldots, t_n]$ , the term  $\bar{t}$  is usually referred to as the *head* of the list while the list  $[t_1, \ldots, t_n]$  is referred to as the *tail* of the list.

For all kinds of PROLOG(+T)-terms t we define the set VAR(t) to contain all variables occurring in t. Formally this is accomplished as follows.

#### Definition 5.1.2 (Variables in PROLOG(+T)-terms)

Let t be a PROLOG(+T)-term. Then the set VAR(t) of variables of t is defined as follows:

- 1. if  $t = \_$  or t = [], then  $VAR(t) = \emptyset$ , 2. if t = X for some variable identifier X, then  $VAR(t) = \{X\}$ ,

3. if 
$$t = \mathbf{f}(t_1, \dots, t_n)$$
 or  $t = [t_1, \dots, t_n]$ , then  $\operatorname{VAR}(t) = \bigcup_{i=1}^n \operatorname{VAR}(t_i)$  and

4. if 
$$t = [\bar{t}|t_1, \dots, t_n]$$
, then  $\operatorname{VAR}(t) = \operatorname{VAR}(\bar{t}) \cup \bigcup_{i=1}^n \operatorname{VAR}(t_i)$ .

A variable is said to *occur* in a term t if and only if it is a member of the set VAR(t).

Anonymous variables and the list concept are only introduced in order to improve the usability of PROLOG(+T) for practical applications. It is possible to show that omitting these concepts does not effect the expressivity of the language. Therefore we will not

<sup>&</sup>lt;sup>1</sup>Provided that the signature of interest does contain definitions for the symbols used in this term.

deal with lists and anonymous variables in proofs of properties in the sequel.

#### 5.1.2. Facts

As in PROLOG, PROLOG(+T) allows modeling relationships between objects as rules and *facts* (which are a special case of rules, namely rules without premises). Roughly speaking, rules are models of inferences which may be carried out. In general such inferences may only be applicable if some *premises* are fulfilled. For facts no such premises are present: any fact is interpreted as something which is true without having to be proved. As in PROLOG, facts in PROLOG(+T) are equivalent to atomic formulas.

#### Definition 5.1.3

A  $\operatorname{Prolog}(+T)\text{-}\text{fact}$  is a string of the form

 $\varphi$ .,

where  $\varphi \in \mathcal{A}_t(\text{sig})$  (for some suitable signature sig) is an atomic formula of FOLTL.

Note the symbol . at the end of the definition of a fact. This comes from the circumstance that facts are special cases of rules.

#### 5.1.3. Rules

As described above, *rules* in general model allowed inferences which may be applied if some premises are known to be fulfilled. Rules are (as described in chapter 3.1) modeled as clauses. However, we will not restrict ourselves on (definite) hornclauses here.

Definition 5.1.4 (Rules in PROLOG(+T)) Let sig =  $(\mathcal{X}, F, P, \alpha)$  be a signature. A PROLOG(+T)-rule is either a PROLOG(+T)-

fact or a statement of the form

$$\psi:=\varphi_1,\ldots,\varphi_n.$$

for  $\psi \in \mathcal{A}_t(\text{sig})$  and  $\varphi_1, \ldots, \varphi_n \in \mathcal{L}_t(\text{sig})$ . The atom  $\psi$  is called the *head* of the rule while the set  $\{\varphi_1, \ldots, \varphi_n\}$  is called the *tail* of the rule.

So the tail of a rule describes the premises which have to be proved while the head describes the proposition which can be inferred. In general the tail may contain *negative* literals, that is  $\varphi_i \in \mathcal{L}_t(\text{sig}) \setminus \mathcal{A}_t(\text{sig})$  is permitted. This has both advantages and disadvantages. On the one hand, allowing negated premises yields better readable, more elegant and shorter programs, on the other hand the definition of the semantics of programs containing rules with negated premises is more complicated to describe. We will see how this can be overcome in a later section of this chapter.

For reasons of readability we will make the following convention: if

$$\psi := \varphi_1, \ldots, \varphi_n.$$

is a PROLOG(+T)-rule and  $\varphi_i$  is from  $\mathcal{L}_t(\operatorname{sig}) \setminus \mathcal{A}_t(\operatorname{sig})$ , that is  $\varphi_i = \neg \varphi$  for some  $\varphi \in \mathcal{L}_t(\operatorname{sig})$ , then we will write  $\varphi_i = \operatorname{not}(\varphi)$  instead. Another reason for using the above notation is the fact that we will not be dealing with *classical* negation but merely with *negation as failure* (or better: with an adaption of this negation-approach). Intuitively  $\operatorname{not}(\varphi)$  is assumed to be a logical consequence of a program P if  $\varphi$  cannot be *proved* from P in a finite number of proof steps or equivalently if every attempt to prove  $\varphi$  from P fails after a finite number of steps. We will see in later sections how the negation as failure-approach can be adapted in order to handle  $\operatorname{PROLOG}(+T)$ -programs.

#### 5.1.4. Programs

As in PROLOG, a program in PROLOG(+T) is given as a finite set of PROLOG(+T)-rules (and facts). The variables in the rules will be assumed to be universally quantified. This is a useful assumption as we will see when studying the semantics of PROLOG(+T)programs in chapter 5.2.

**Definition 5.1.5 (PROLOG(+T)-program)** A PROLOG(+T)-program is a finite set P of PROLOG(+T)-rules.

PROLOG(+T)-programs (or simply *programs* if there's no way of confusion) will be denoted as  $P, \overline{P}, P' \dots$  from now on. Due to the special form of the formulas in programs it is possible to build a formula which is semantically equivalent to a PROLOG(+T)program and which has a very simple form. Let P be a program containing n rules. That is let  $P = \{P_1, \ldots, P_n\}$  where each  $P_i$  has the form

$$P_i = \psi^{(i)} :- \varphi_1^{(i)}, \dots, \varphi_{n_i}^{(i)}.$$

for  $\psi^{(i)} \in \mathcal{A}_t(\text{sig})$  and  $\varphi_1^{(i)}, \ldots, \varphi_{n_i}^{(i)} \in \mathcal{L}_t(\text{sig})$  and some  $n_i \in \mathbb{N}$ . For formal reasons we will identify  $\psi^{(i)}$  :- and  $\psi^{(i)}$ . Due to the interpretation of clauses as sets of literals we therefore have

$$P_i \equiv \neg \left(\varphi_1^{(i)} \land \dots \land \varphi_{n_i}^{(i)}\right) \lor \psi^{(i)}$$
$$\equiv \neg \varphi_1^{(i)} \lor \dots \lor \neg \varphi_{n_i}^{(i)} \lor \psi^{(i)}$$
$$\equiv \psi^{(i)} \lor \bigvee_{j=1}^{n_i} \neg \varphi_j^{(i)}$$

for  $i = 1, \ldots, n$ . Since sets of clauses are satisfied under interpretations if all clauses are satisfied simultaneously we have  $\mathcal{J} \models P$  if and only if  $\mathcal{J} \models P_i$  for i = 1, ..., n if and only if  $\mathcal{J} \models P_1 \land \dots \land P_n$  if and only if  $\mathcal{J} \models \bigwedge_{i=1}^n P_i$ . This gives  $P \equiv \bigwedge_{i=1}^n P_i$  and therefore

$$P \equiv \bigwedge_{i=1}^{n} P_{i}$$
$$= \bigwedge_{i=1}^{n} \left( \psi^{(i)} \lor \bigvee_{j=1}^{n_{i}} \neg \varphi_{j}^{(i)} \right)$$

is a relatively simple formula which might characterize the semantics of a PROLOG(+T)program P.

From now on we will assume that the rules in a program P are standardized apart.

**Definition 5.1.6** Let  $\varphi_1$  and  $\varphi_2$  be FOLTL-formulas.  $\varphi_1$  and  $\varphi_2$  are called *standardized apart* if  $VAR(\varphi_1) \cap VAR(\varphi_2) = \emptyset$ .

It is obvious that two rules from a program can be easily standardized apart. This is a consequence of the assumption that all variables in rules are implicitly universally quantified: let P be a program as described above and let  $i, j \in \{1, \ldots, n\}$  be such that  $\operatorname{VaR}(P_1) \cap \operatorname{VaR}(P_2) = \{\mathbf{X}_{i_1}, \dots, \mathbf{X}_{i_k}\} \neq \emptyset$ . We now fix two sets  $\{\mathbf{X}_{i_1}^{(1)}, \dots, \mathbf{X}_{i_k}^{(1)}\}$ and  $\{\mathbf{X}_{i_1}^{(2)}, \dots, \mathbf{X}_{i_k}^{(2)}\}$  such that  $\{\mathbf{X}_{i_1}^{(1)}, \dots, \mathbf{X}_{i_k}^{(1)}, \mathbf{X}_{i_1}^{(2)}, \dots, \mathbf{X}_{i_k}^{(2)}\} \cap (\operatorname{VaR}(P_1) \cup \operatorname{VaR}(P_2)) = \emptyset$ , define substitutions

$$\sigma_i = \left\{ \frac{\mathbf{X}_{i_1}}{\mathbf{X}_{i_1}^{(i)}}, \dots, \frac{\mathbf{X}_{i_k}}{\mathbf{X}_{i_k}^{(i)}} \right\} \text{ and}$$
  
$$\sigma_j = \left\{ \frac{\mathbf{X}_{i_1}}{\mathbf{X}_{i_1}^{(j)}}, \dots, \frac{\mathbf{X}_{i_k}}{\mathbf{X}_{i_k}^{(j)}} \right\}$$

and replace the program P by  $\overline{P} = (P \setminus \{P_i, P_j\}) \cup \{\sigma_i(P_i), \sigma_j(P_j)\}$ . Due to the implicit universal quantification of all variables in the rules of P we have  $P \models \varphi$  if and only if  $\bar{P} \models \varphi$  for any  $\varphi$ .

#### 5.1.5. Queries

As in PROLOG, a PROLOG(T)-program is *run* by giving a *query* to the program and searching for a substitution such that the result of applying this substitution to the query yields a logical consequence of the program. Formally we will adopt the philosophy of logic programming as described in chapter 3.1. A *query* in PROLOG(+T) is therefore interpreted as a sequence of single queries which have to be solved one by one. As in rules we will also allow negation as failure in queries.

## Definition 5.1.7 (PROLOG(+T)-query)

Let sig =  $(\mathcal{X}, F, P, \alpha)$  be a signature. A PROLOG(+T)-query over sig is a formula of the form

$$G = \varphi_1, \ldots, \varphi_n$$

such that  $\varphi_1, \ldots, \varphi_n \in \mathcal{L}_t(\text{sig}).$ 

As we are interested in substitutions for the variables in a goal, we need the concept of an answer as described by [105]. An answer for a query G is a substitution  $\sigma$  such that  $DOM(\sigma) = VAR(G)$ . If P is a PROLOG(+T)-program and  $G = \varphi_1, \ldots, \varphi_n$ , then an answer  $\sigma$  is called *correct* for G if  $P \models \sigma(\varphi_1) \land \cdots \land \sigma(\varphi_n)$ . Obviously the semantics of a program P can be characterized in terms of the set of all goals G consisting of a single query (i.e.  $G = \varphi$ . for some  $\varphi \in \mathcal{L}_t(sig)$ ) such that there is a correct answer  $\sigma$  for G. In this case  $\sigma(\varphi)$  is contained in the set characterizing the semantics of the program. How these characterization can be formally described will be the subject of the following section of this chapter.

# 5.1.6. The Relation of Prolog(+T) to other Temporal Logic programming Languages

This section will deal with the question of how PROLOG(+T) differs from other temporal logic programming languages introduced so far. In particular we will discuss the differences between PROLOG(+T), TEMPLOG and TOKIO. It will turn out that on the one hand PROLOG(+T) is syntactically closely related to TEMPLOG although there are some differences which are worth pointing out. On the other hand we will discuss TOKIO which is also similar to PROLOG(+T) regarding the temporal operators used in its definition but which has a completely different philosophical origin.

#### Templog

TEMPLOG is a first-order temporal programming language which has originally been introduced and defined by Abadi and Manna in [3] and [2]. The underlying logic of TEMPLOG allows the usage of the same set of temporal operators as PROLOG(+T) while the authors distinguish between *flexible* and *rigid* symbols. Flexible symbols may be interpreted as operations with a semantics which changes over time while the interpretations of rigid symbols must not be depending on the point of time at which the symbol is evaluated.

In [3] the definition of TEMPLOG is carried out as follows:

- 1. The authors define a temporal logic on which TEMPLOG is based.
- 2. They define a fragment of TEMPLOG in order to present the basic ideas.
- 3. Finally they introduce the full logic by enhancing the set of temporal operators which are allowed in the definition of the programming statements of the programming language TEMPLOG.

The fragment of TEMPLOG which is defined in [3] introduces the concepts of *initial* temporal hornclauses and permanent temporal hornclauses as an extension of the hornclauses which are used as programming statements in PROLOG. An initial temporal hornclause is a construct which has the form

$$\forall x_1 \dots \forall x_k \, (\varphi_1 \wedge \dots \wedge \varphi_n \to \psi)$$

for so called *next-atomic* formulas  $\varphi_1, \ldots, \varphi_n$  and  $\psi$  where  $\bigcup_{i=1}^n \operatorname{VAR}(\varphi_i) \cup \operatorname{VAR}(\psi) = \{x_1, \ldots, x_n\}$ . Here a formula  $\varphi$  is called *next-atomic* if  $\varphi$  has the form  $\varphi = X^i \overline{\varphi}$  for some  $i \ge 0$  and some atom  $\overline{\varphi}$ . Due to the fact that the definition TEMPLOG is motivated by the wish to enhance PROLOG with temporal operators the authors of [3] introduce the abbreviation

$$\psi \leftarrow \varphi_1, \dots, \varphi_n := \forall x_1 \dots \forall x_k \, (\varphi_1 \wedge \dots \wedge \varphi_n \to \psi)$$

which clearly shows the relation between TEMPLOG and PROLOG.

Additionally permanent temporal hornclauses are programming statements of the form

$$\forall x_1 \dots \forall x_k \mathsf{G} (\varphi_1 \wedge \dots \wedge \varphi_n \to \psi)$$

for next-atomic formulas  $\varphi_1, \ldots, \varphi_n$  and  $\psi$  and  $\{x_1, \ldots, x_k\}$  as before. Similar to initial temporal hornclauses permanent temporal hornclauses may be abbreviated using

$$\psi \leftarrow \varphi_1, \dots, \varphi_n := \forall x_1 \dots \forall x_k \mathsf{G} (\varphi_1 \land \dots \land \varphi_n \to \psi).$$

Programs in the fragment defined in this way are defined as sets consisting of permanent and initial hornclauses. Queries to such programs are defined as conjunctions of nextatomic formulas. It is possible to show that the programming language defined by this fragment can be evaluated using a resolution-style theorem proving procedure (see e.g. [1] for a discussion of such a procedure).

After having defined the fragment presented above the authors introduce the full logic programming TEMPLOG by allowing usage of G in the head and F in the tail of program

statements (both with restrictions).

As a summary we can see:

- 1. TEMPLOG offers a distinction between flexible and rigid symbols which allows to interpret symbols in a different way at different points of time.
- 2. PROLOG(+T) does not have a counterpart to the operator  $\leftarrow$  of TEMPLOG.
- 3. In contrast to PROLOG(+T) TEMPLOG limits the application of G and F to the head (resp. the tail) of clauses.
- 4. The operators U and P are forbidden in TEMPLOG programs.

#### Tokio

TOKIO is another programming language which allows the usage of certain temporal operators in programming statements. TOKIO has been presented in [6] as an extention of the logic ITL (see e.g. [117]). In contrast to the logic underlying both TEMPLOG and PROLOG(+T) TOKIO and its predecessor ITL are *interval based* logic programming languages. This means that the main goal of a proof procedure executing a TOKIO program is not to prove a goal but to find an interval of time in which the goal holds. We will make this clear soon.

The syntax of TOKIO's programming constructs is defined in [6] (in another paper an interpreter for TOKIO written in PROLOG is presented. It is noted that the execution time of TOKIO programs using this interpreter is slowed down by a factor of 40 in contrast to ordinary PROLOG programs; we refer to [92] for implementation details). Similarly to PROLOG we can use programming statements which do not contain temporal operators. So every PROLOG-statement  $\psi \leftarrow \varphi_1, \ldots, \varphi_n$  is also a well-formed TOKIO-statement.

Additionally to PROLOG TOKIO allows the usage of several temporal constructs. These are

• sequential execution,

- the Next–operator,
- the Always–operator,
- the Sometimes-operator,
- the *keep*-operator and
- the *fin*-operator.

Since TOKIO is an interval-based programming approach the main goal of a prover executing a TOKIO formula is to define an interval of time in which this formula is satisfied. Time in TOKIO is considered discrete, so an interval I can be described by specifying its start- resp. endpoint in terms of natural numbers. If  $I = [I_{\text{beg}}, I_{\text{fin}}]$  is such an interval we have to require that  $I_{\text{beg}} \leq I_{\text{fin}}$ . Since the motivation for the definition of TOKIO is the description of hardware the authors of [6] prefer the term *execution* in order to denote the satisfaction of a formula.

We will now make the concepts of TOKIO more clear.

- TOKIO allows the specification of the sequential execution of goals. This is carried out by the so called *chop*-operator which is denoted by &&. So a statement ψ ← φ<sub>1</sub>&&...&&φ<sub>n</sub> is executed in an interval I = [I<sub>beg</sub>, I<sub>fin</sub>] if I can be divided into intervals I<sub>1</sub>,..., I<sub>n</sub> such that I<sub>1</sub> = [I<sub>beg</sub>, t<sub>1</sub>],..., I<sub>j</sub> = [t<sub>j-1</sub>, t<sub>j</sub>],..., I<sub>n</sub> = [t<sub>n-1</sub>, I<sub>fin</sub>] for t<sub>1</sub>,..., t<sub>n-1</sub> ∈ N such that t<sub>i</sub> ≤ t<sub>i+1</sub> for i = 1,..., n-2 and each φ<sub>i</sub> is executed in the interval I<sub>i</sub>.
- 2. The Next-operator is intended as a similar concept to the operator X from PRO-LOG(+T). In TOKIO the Next-operator is denoted as @ and a statement  $\psi \leftarrow @\varphi$  is intended to be executed in an interval  $I = [I_{beg}, I_{fin}]$  if  $\psi$  is executed in the interval  $[I_{beg} + 1, I_{fin}]$ .
- 3. Similarly the Always– and Sometimes–operators which are denoted as # resp. <>

require the execution of a statement at certain points of time in an interval. If  $I = [I_{beg}, I_{fin}]$  is a given interval, then

- a)  $\psi \leftarrow \#\varphi$  is executed in I if  $\varphi$  is executed at every point  $I_{\text{beg}}, \ldots, I_{\text{beg}} + j, \ldots, I_{\text{fin}}$  of time and
- b) ψ ← <>φ is executed in I if there is (at least) one j with 0 ≤ j ≤ fin − beg such that φ is executed at I<sub>beg</sub> + j.
- 4. The keep-operator allows reasoning about the execution of a statement at serval points of time in an interval. If I is given then ψ ← keep(φ) is executed in I if φ is executed at every point of time in I except of the last point. More formally if I = [I<sub>beg</sub>, I<sub>fin</sub>] then ψ ← keep(φ) is executed in I if φ is executed at time I<sub>beg</sub>,..., I<sub>fin</sub> 1 and φ is not executed at time point I<sub>fin</sub>.
- Finally the fin-operator is introduced in order to reason about the execution of a statement at a final point of time in an interval. If I = [I<sub>beg</sub>, I<sub>fin</sub>] is given then ψ ← fin(φ) is executed in I if φ is executed at the time I<sub>fin</sub>.

Although the operators #, && and @ are the same operators as their counterparts in PROLOG(+T) the *keep* and *fin*-operators cannot be mapped to PROLOG(+T) adequately. The only relationships we can derive is that

- 1. statements of the form  $\psi \leftarrow keep(\varphi)$  are related to statements of the form  $\psi :-\varphi U\bar{\varphi}$ for some  $\bar{\varphi}$  and
- 2. statements of the form  $\psi \leftarrow fin(\varphi)$  are related to statements of the form  $\psi :-\varphi \mathsf{P}\bar{\varphi}$  for some  $\bar{\varphi}$ .

As a summary we can therefore point out the following differences between TOKIO and PROLOG(+T):

1. PROLOG(+T) does not have direct counterparts for the operators keep, fin and && of TOKIO.

- 2. In PROLOG(+T) we do not have any limits for the time points in which formulas are satisfied. In contrast TOKIO statements are executed in intervals. Since such intervals are assumed to be given by time points which are represented as natural numbers, they always have a finite length.
- 3. The philosophies of TOKIO and PROLOG(+T) are completely different since the main task in TOKIO is the construction of a model (i.e. the detection of an interval) while the main task in PROLOG(+T) is to refute a formula, i.e. proving that there cannot exist any models.

# 5.2. Declarative Semantics of Prolog(+T)

This section will deal with the *declarative semantics* of the programming language PRO-LOG(+T) introduced in the last section. The declarative semantics is defined in terms of the logical consequence relation  $\models$  and it is important to distinguish this semantics from the *operational semantics* which is defined in terms of some suitable calculus  $\vdash$ . This operational semantics will be the subject of chapter 5.3.

# 5.2.1. Preliminaries

We will now extend the concepts of *Herbrand-interpretations* as introduced in first order logic (see [140] or [105]) to the logic FOLTL. We will see that the results from first order logic carry over to FOLTL.

Let sig =  $(\mathcal{X}, F, P, \alpha)$  be any signature. We will need the concept of *ground objects* in order to define Herbrand-interpretations.

- A term  $t \in \mathcal{T}(\text{sig})$  is called a ground term if  $\text{VAR}(t) = \emptyset$ .
- A FOLTL-formula  $\varphi$  is called a ground formula if  $VAR(\varphi) = \emptyset$ .

Similarly a ground atom is a ground formula which is an atom, a ground literal is a

ground formula which is a literal and a *ground clause* is a ground formula which is a clause.

For the rest of this chapter we will assume that sig =  $(\mathcal{X}, F, P, \alpha)$  is any fixed signature. Let  $\Phi$  be any set of formulas over sig. The set of all ground terms which can be built from symbols occurring in  $\Phi$  will be called the *Herbrand-universe* of  $\Phi$  and will be denoted as  $U_{\Phi}$ . In particular we will be interested in  $U_P$  for PROLOG(+T)-programs P. Similarly one defines sets  $B_{\Phi}$  and  $B_{\Phi}^{FoLTL}$  as follows:

$$B_{\Phi} = \{ \varphi \in \mathcal{A}(\operatorname{sig}) \mid \operatorname{Var}(\varphi) = \emptyset \} \text{ and}$$
$$B_{\Phi}^{\operatorname{FoLTL}} = \{ \varphi \in \mathcal{A}_t(\operatorname{sig}) \mid \operatorname{Var}(\varphi) = \emptyset \}$$

The set  $B_{\Phi}$  is the well known *Herbrand-base*. The set  $B_{\Phi}^{\text{FoLTL}}$  is an extension of the Herbrand-base which also allows the inclusion of temporal atoms. We will also refer to  $B_{\Phi}^{\text{FoLTL}}$  as the *Herbrand-base* of  $\Phi$  since there is no way of confusion<sup>2</sup>.

We will see that a well-known result from first order logic can be extended to FOLTL. Therefore we will define the set FREE(o) for some logical object o to be the set of *free variables* occurring in this object. Formally:

- for terms from  $\mathcal{T}(sig)$  we define
  - $FREE(X) = \{X\}$  if  $X \in \mathcal{X}$  is a variable symbol and
  - FREE $(\mathbf{f}(t_1,\ldots,t_n))$  =  $\bigcup_{i=1}^n \text{FREE}(t_i)$  if  $\mathbf{f} \in F$  is a function symbol with  $\alpha(\mathbf{f}) = n \text{ and } t_1,\ldots,t_n \in \mathcal{T}(\text{sig}).$

Equivalently we could define FREE(t) = VAR(t) for any  $t \in \mathcal{T}(sig)$ .

• For formulas we define

<sup>&</sup>lt;sup>2</sup>Due to the fact that syntactically different literals might be logically equivalent we have that a *positive* literal can be equivalent to a *negative* one, e.g.  $Gp(X) \equiv not(Fnot(p(X)))$ . Therefore we will make the following convention:  $B_P^{FoLTL}$  contains all temporal literals which are equivalent to some positive temporal literal.

- $\operatorname{FREE}(\mathtt{true}) = \operatorname{FREE}(\mathtt{false}) = \emptyset,$
- if  $\mathbf{p} \in P$  is a predicate symbol with  $\alpha(\mathbf{p}) = n$  and  $t_1, \ldots, t_n \in \mathcal{T}(\text{sig})$ , then  $\operatorname{FREE}(\mathbf{p}(t_1, \ldots, t_n)) = \bigcup_{i=1}^n \operatorname{FREE}(t_i)$ ,
- if  $\varphi$  is a formula, then

$$FREE(not(\varphi)) = FREE(X\varphi) = FREE(G\varphi) = FREE(F\varphi) = FREE(\varphi),$$

- if  $\varphi$  is formula, then  $\text{FREE}(\forall \mathbf{X}\varphi) = \text{FREE}(\exists \mathbf{X}\varphi) = \text{FREE}(\varphi) \setminus \{\mathbf{X}\}$  and
- if  $\varphi_1$  and  $\varphi_2$  are formulas, then  $\operatorname{FREE}(\varphi_1 \cup \varphi_2) = \operatorname{FREE}(\varphi_1 \cap \varphi_2) = \operatorname{FREE}(\varphi_1 \wedge \varphi_2) = \operatorname{FREE}(\varphi_1 \vee \varphi_2) = \operatorname{FREE}(\varphi_1 \to \varphi_2) = \operatorname{FREE}(\varphi_1 \leftrightarrow \varphi_2) = \operatorname{FREE}(\varphi_1) \cup \operatorname{FREE}(\varphi_2).$

A formula  $\varphi$  is called *closed* if  $\text{FREE}(\varphi) = \emptyset$ . Similarly a formula  $\varphi$  is called *universally* closed if  $\varphi$  is a closed formula which does not contain the quantifier  $\exists$ .

We will now see that certain subsets of the Herbrand-base of a program can be considered as interpretations. This is achieved in a similar way as in the case of first order logic programs. However, we will have to put some restrictions on the subsets of interest. After this we will make a link between these sets of atoms (which we will refer to as *setbased-interpretations*) and the interpretations of FOLTL-formulas as defined in chapter 2.3.2 (which we will refer to as *structure-based interpretations*).

Let  $P = \{P_1, \ldots, P_n\}$  be any PROLOG(+T)-program and let  $I \subseteq B_P^{FOLTL}$  be any set of ground atoms built from symbols occurring in P. Furthermore let  $\varphi$  be any universally closed FOLTL-formula<sup>3</sup>. I will be seen as an interpretation for  $\varphi$  as follows:

 if φ = X<sup>i</sup>ψ for any i ≥ 0 and any nontemporal atom ψ from B<sub>P</sub>, then I ⊨ φ if and only if φ ∈ I,

<sup>&</sup>lt;sup>3</sup>Similarly as in First Order Logic these construction strongly relies on the assumption that every variable symbol in the formula under consideration is universally quantified. For general formulas the construction fails.

- 2. if  $\varphi = \varphi_1 \land \varphi_2$ , then  $I \models \varphi$  if and only if  $I \models \varphi_1$  and  $I \models \varphi_2$ ,
- 3. if  $\varphi = \varphi_1 \lor \varphi_2$ , then  $I \models \varphi$  if and only if  $I \models \varphi_1$  or  $I \models \varphi_2$ ,
- 4. if  $\varphi = \neg \psi$ , then  $I \models \varphi$  if and only  $I \not\models \psi$ ,
- 5. if  $\varphi = \varphi_1 \to \varphi_2$ , then  $I \models \varphi$  if and only if  $I \models \neg \varphi_1$  or  $I \models \varphi_2$ ,
- 6. if  $\varphi = \varphi_1 \leftrightarrow \varphi_2$ , then  $I \models \varphi$  if and only if  $I \models \varphi_1 \rightarrow \varphi_2$  and  $I \models \varphi_2 \rightarrow \varphi_1$  and
- 7. if  $\varphi = \forall \mathbf{X} \psi$  then
  - a) if  $X \in VAR(\psi)$ , then  $I \models \varphi$  if and only if for every substitution  $\sigma = \left\{\frac{X}{t}\right\}$  for some t such that  $VAR(t) = \emptyset$  it holds that  $I \models \sigma(\psi)$  and
  - b) if  $X \notin VAR(\psi)$ , then  $I \models \varphi$  if and only if  $I \models \psi$ .

This definition of the semantics in terms of subsets of  $B_P^{\rm FoLTL}$  only allows the interpretation of formulas which contain the temporal operator X. But in order to handle formulas involving G, F, U and P we have to restrict the subsets of interest to such subsets which are *temporally closed*.

# Definition 5.2.1 (Temporally closed set)

A set  $I \subseteq B_P^{\text{FoLTL}}$  is called *temporally closed* if and only if for every  $\varphi, \varphi_1$  and  $\varphi_2$  from  $B_P^{\text{FoLTL}}$  and every  $i \ge 0$  the following conditions are fulfilled:

- 1.  $\mathsf{X}^{i}\mathsf{G}\varphi \in I$  if and only if  $\mathsf{X}^{i+j}\varphi \in I$  for every  $j \ge 0$ ,
- X<sup>i</sup>Fφ ∈ I if and only if X<sup>i+j</sup>φ ∈ I for some j ≥ 0,
   X<sup>i</sup>φ<sub>1</sub>Uφ<sub>2</sub> ∈ I if and only if X<sup>i</sup>φ<sub>2</sub> ∈ I or X<sup>i</sup>φ<sub>1</sub> ∈ I and X<sup>i+1</sup>φ<sub>1</sub>Uφ<sub>2</sub> ∈ I,
   X<sup>i</sup>φ<sub>1</sub>Pφ<sub>2</sub> ∈ I if and only if X<sup>i</sup>φ<sub>2</sub> ∉ I and X<sup>i</sup>φ<sub>1</sub> ∈ I or X<sup>i+1</sup>φ<sub>1</sub>Pφ<sub>2</sub> ∈ I and
- 5.  $\varphi \in I$  if and only if  $\{\psi \in B_P^{\text{FoLTL}} \mid \psi \equiv \varphi\} \subseteq I$ .

The motivation for this definition should immediately be clear since it is directly derived from the properties of the semantical equivalence relation for FoLTL-formulas. So a set I is temporally closed if satisfaction of *one* member of an equivalence class implies satisfaction of *all* members of this equivalence class. Therefore considering temporally closed subsets of  $B_P^{\text{FoLTL}}$  as the interpretations of interest is reasonable.

It is important to note that one can in many cases enrich a set  $I \subseteq B_P^{\text{FoLTL}}$  by adding atoms in order to receive a temporally closed set. This procedure will later be referred to as *building* (or *constructing*) the temporal closure. In general, a temporally closed superset of a set I is not uniquely determined.

#### Example 5.2.1

Consider the set  $I = {Fp(a)}$ . every temporally closed superset of I contains  $X^{j}p(a)$  for some  $j \ge 0$ .

The temporally closed supersets of a set I can be seen as the unions of sets on (infinite) maximal paths in an infinite tree. We will therefore construct a labeled graph T(I) = (V, E, l) from I as described below<sup>4</sup>.

T(I) = (V, E, l) with  $l : V \to 2^{B_P^{\text{FoltL}}} \cup \{\text{fail}\}$  is the infinite tree satisfying the following conditions:

- 1. There is a uniquely determined  $v_0 \in V$  such that  $(v, v_0) \notin E$  for each  $v \in V$  (the *root* node),
- 2.  $l(v_0) = I$  and
- 3. for each  $v \in V$  the following is true:
  - a) if there is an atom  $X^i G \varphi \in l(v)$ , then there is  $v' \in V$  such that  $(v, v') \in E$  and  $l(v') = l(v) \cup \{X^{i+j}\varphi \mid j \ge 0\},\$

<sup>&</sup>lt;sup>4</sup>Here V is a nonempty set of *nodes* (also called *vertices*),  $E \subseteq V \times V$  is a set of *edges* connecting these nodes and  $l: v \to 2^{B_P^{\text{FoLTL}}} \cup \{\text{fail}\}$  is a mapping which *labels* the nodes.

- b) if there is an atom  $X^i F \varphi \in l(v)$  then there are  $v'_1, \ldots, v'_j, \cdots \in V$  such that  $(v, v'_j) \in E$  and  $l(v'_j) = l(v) \cup \{X^{i+j}\varphi\}$  for each  $j \ge 0$ ,
- c) if there is an atom  $X^i \varphi_1 U \varphi_2 \in l(v)$ , then there are  $v'_1, v'_2 \in V$  such that  $(v, v'_1), (v, v'_2) \in E$  and
  - $l(v'_1) = l(v) \cup \{\mathsf{X}^i \varphi_2\}$  and

• 
$$l(v'_2) = l(v) \cup \{\mathsf{X}^{i+1}\varphi_1\mathsf{U}\varphi_2\}$$

if  $\mathsf{X}^i \varphi_2 \notin l(v)$ ,

d) if there is an atom  $X^i \varphi_1 P \varphi_2 \in l(v)$  then there are  $v'_1, v'_2 \in V$  such that  $(v, v'_1), (v, v'_2) \in E$  and

• if 
$$X^i \varphi_2 \in l(v)$$
, then  $l(v'_1) = l(v'_2) = \texttt{fail}$  and

- if  $\mathsf{X}^i \varphi_2 \notin l(v)$ , then  $l(v_1') = l(v) \cup \{\mathsf{X}^i \varphi_1\}$  and  $l(v_2') = l(v) \cup \{\mathsf{X}^{i+1} \varphi_1 \mathsf{P} \varphi_2\}$ . and
- e) for each  $\varphi \in l(v)$  it holds that

$$\{\psi \in B_P^{\text{FoLTL}} \mid \psi \equiv \varphi\} \subseteq l(v).$$

Given T(I) the set of all (possibly infinite) maximal paths  $\pi$  starting at  $v_0$  such that there is no v with  $l(\pi) = \texttt{fail}$  occurring on  $\pi$  will be denoted as p(T(I)).

Given  $\pi \in p(T(I))$ , the set  $l(\pi)$  denotes the union of all sets with which the nodes on  $\pi$  are labeled. Formally if  $V(\pi)$  denotes the set of nodes occurring on  $\pi$ , then

$$l(\pi) = \bigcup_{v \in V(\pi)} l(v).$$

Then the following claims are immediate:

- 1. For every  $\pi \in p(T(I))$  the set  $l(\pi)$  is temporally closed.
- 2. For every  $\pi \in p(T(I))$  it holds that  $I \subseteq l(\pi)$ .

We will from now on denote the set consisting of all sets computable in the way sketched above as the *temporal closure* of I and denote it as

TEMPCLOSURE
$$(I) = \{l(\pi) \mid \pi \in p(T(I))\}$$
.

Definition 5.2.2 (Set-based Herbrand-Interpretation)

A set-based Herbrand-Interpretation for a program P is a temporally closed subset  $I \subseteq B_P^{\text{FoLTL}}$ . A set-based Herbrand-model for P is any temporally closed set  $I \subseteq B_P^{\text{FoLTL}}$  such that  $I \models P$ .

It is worth noticing that considering only temporally closed sets as interpretations has the drawback that there are some programs which are not satisfiable. This is one difference to pure first order logic programs which are always satisfied by the interpretation  $I = B_P$ . Consider the following program  $P = \{P_1, P_2\}$  where

$$P_1 = \mathbf{p}(\mathbf{a}).$$
 and  
 $P_2 = \mathbf{q}(\mathbf{X})\mathsf{Pp}(\mathbf{X}):-\mathbf{p}(\mathbf{X})$ 

Now fix any temporally closed  $I \subseteq B_P^{\text{FoLTL}}$ . If  $I \models P$ , then in particular we have  $I \models P_1$ , that is  $I \models P_1 = p(\mathbf{a})$ . So  $p(\mathbf{a}) \in I$ . On the other hand we have  $I \models P_2 = q(\mathbf{X})Pp(\mathbf{X})$ :  $-p(\mathbf{X})$  and since  $P_2$  is considered universally closed we have  $I \models q(\mathbf{a})Pp(\mathbf{a}) :-p(\mathbf{a})$ . But since I is temporally closed we have  $I \not\models p(\mathbf{a})$  and  $I \models q(\mathbf{a})$  or  $I \models Xq(\mathbf{a})Pp(\mathbf{a})$ . This is a contradiction. So  $I \not\models P$  and therefore P has no set-based Herbrand-model.

We will now prove that for universally closed sets of FOLTL-formulas the concepts of set-based Herbrand-Interpretations and structure-based Herbrand-Interpretations as introduced in chapter 2.3.2 are equivalent. This allows reasoning about properties of programs by considering interpretations as sets of literals instead of the formally more complicated structures.

We will first prove the *easier* direction.

#### Lemma 5.2.1

Let P be a PROLOG(+T)-program. If P has a structure-based Herbrand-model, then P has a set-based Herbrand-model.

**Proof.** Let P be a PROLOG(+T)-program and let  $\mathcal{J} = (U_{\mathcal{J}}, S, s_0, \delta_1, \delta_2, w, \mathcal{I})$  be a structure-based Herbrand-model of P. Set  $I_{\mathcal{J}} = \{\varphi \in B_P^{\text{FoLTL}} \mid \mathcal{J} \models \varphi\}$ , i.e.  $I_{\mathcal{J}}$  becomes the set of all ground instances satisfied by  $\mathcal{J}$ . Then every  $I \in \text{TEMPCLOSURE}(I_{\mathcal{J}})$  is easily seen to be a set-based Herbrand-model of P.

The opposite direction is also true although it is much more complicated to prove (due to the more complicated definition of structure-based interpretations).

#### Lemma 5.2.2

Let P be a PROLOG(+T)-program. If P has a set-based Herbrand-model, then P has a structure-based Herbrand-model.

**Proof.** Let  $I \subseteq B_P^{\text{FoLTL}}$  be a set-based Herbrand-model of P, that is I is temporally closed and  $I \models P$ . We will construct a structure-based interpretation  $\mathcal{J}_I = (U_P, S, s_0, \delta_1, \delta_2, w, \mathcal{I})$  from I such that  $\mathcal{J}_I \models P$ . Therefore define  $S = \{s^{(i)} \mid i \in \mathbb{N}\}$  and  $s_0 = s^{(0)}$ . Furthermore define for  $s \in S$  and any  $t = f(t_1, \ldots, t_n) \in U_P$ :  $\mathcal{I}(s, t) = \mathcal{I}(s, f)(t_1, \ldots, t_n) = f(t_1, \ldots, t_n)$  as obvious. Since every element of P is considered universally closed we can set w as any arbitrary mapping.

We then define  $\delta_1 := \{(s^{(i)}, s^{(i+1)}) \mid i \ge 0\}$  and  $\delta_2 := \{(s^{(i)}, s^{(i+j)}) \mid i, j \ge 0\}$ . After this we proceed as follows:

- 1. Take some  $\varphi$  from I and set  $I = I \setminus \{\varphi\}$ .
- 2. Case 1 if  $\varphi = X^i p(t_1, \ldots, t_n)$  for some  $i \ge 0$ , some predicate symbol **p** of arity n

and  $t_1, \ldots, t_n \in U_P$ , then we set

$$\mathcal{I}\left(s^{(i)},\mathbf{p}\right)\left(t_1,\ldots,t_n\right)=1,$$

**Case 2** if  $\varphi = \mathsf{X}^i \mathsf{G} \psi$  for some  $i \geq 0$  and some  $\psi \in B_P^{\text{FoLTL}}$ , then set

$$\mathcal{I}\left(s^{(i+j)},\psi\right) = 1$$

for every  $j \ge 0$ ,

**Case 3** if  $\varphi = \mathsf{X}^i \mathsf{F} \psi$  for some  $i \geq 0$  and some  $\psi \in B_P^{\text{FoLTL}}$ , then set

$$\mathcal{I}\left(s^{(i+j)},\psi\right) = 1$$

for some  $j \ge 0$ ,

**Case 4** if  $\varphi = \mathsf{X}^i \varphi_1 \mathsf{U} \varphi_2$  for some  $i \ge 0$  and  $\varphi_1, \varphi_2 \in B_P^{\text{FoLTL}}$ , then set

$$\mathcal{I}\left(s^{(i)},\varphi_2\right) = 1$$

or

$$\mathcal{I}\left(s^{(i)},\varphi_{1}\right) = 1 \text{ and } \mathcal{I}\left(s^{(i+1)},\varphi_{1}\mathsf{U}\varphi_{2}\right) = 1$$

and

**Case 5** if  $\varphi = \mathsf{X}^i \varphi_1 \mathsf{P} \varphi_2$  for some  $i \ge 0$  and  $\varphi_1, \varphi_2 \in B_P^{\text{FoLTL}}$ , then set

$$\mathcal{I}\left(s^{(i)},\varphi_2\right) = 0$$

and

$$\mathcal{I}\left(s^{(i)},\varphi_{1}\right) = 1 \text{ or } \mathcal{I}\left(s^{(i+1)},\varphi_{1}\mathsf{P}\varphi_{2}\right) = 1.$$

3. If  $I \neq \emptyset$ , then go back to step 1.

It is obvious that the structure-based interpretation  $\mathcal{J}_I$  which emerges by applying these steps as long as  $I \neq \emptyset$  (i.e. the *limit interpretation*) is a structure-based Herbrand-model of P.

Theorem 5.2.1 (Equivalence of set- and structure-based interpretations) Let P be any PROLOG(+T)-program. Then P has a structure-based Herbrand-model if and only if P has a set-based Herbrand-model.

**Proof.** Immediately from Lemma 5.2.1 and Lemma 5.2.2.  $\Box$ 

# 5.2.2. Reduction of Literals

It will turn out useful to introduce a concept of *reduction* for temporal literals. Intuitively a reduced form of some temporal literal  $\varphi$  is a *normal form*  $\text{RED}(\varphi)$ . We will therefore define how a reduced literal looks like and how it can be effectively computed given the (unreduced) literal.

An approach to define a certain type of normal-form for temporal logic formulas has been presented in [65] and [64] for the propositional logic LTL enriched with past operators. However, this *separated normal form* is defined for a much larger class of formulas than only atoms and literals. Consequently the structure of this normal form is much more complicated than necessary for our purposes.

In order to compute reduced forms of literals we will exploit several simple semantical equivalences. The basic idea is to first *pull out* the negation operator (if it is contained), so that each reduced literal  $\varphi$  is either of the form  $\varphi = \psi$  for some literal  $\psi$  not containing **not** or  $\varphi = \mathbf{not}(\psi)$  for some  $\psi$  not containing **not**. The following set of logical equivalences will be the basis of our reduction concept.

GGarphi	$\mapsto$	$G \varphi,$		$FF\varphi$	$\mapsto$	$F\varphi,$
$\texttt{Gnot}(\varphi)$	$\mapsto$	$\mathtt{not}(F\varphi),$		${\tt Fnot}(\varphi)$	$\mapsto$	$\mathtt{not}(G\varphi),$
$\texttt{Xnot}(\varphi)$	$\mapsto$	$\mathtt{not}(X\varphi),$		$\texttt{not}(\texttt{not}(\varphi))$	$\mapsto$	$\varphi,$
Xfalse	$\mapsto$	false,		Xtrue	$\mapsto$	true,
Gfalse	$\mapsto$	false,		Gtrue	$\mapsto$	true,
Ffalse	$\mapsto$	false,		Ftrue	$\mapsto$	true,
$\texttt{trueU}\varphi$	$\mapsto$	$F\varphi,$		$\texttt{trueP}\varphi$	$\mapsto$	$\texttt{not}(\varphi),$
arphiUtrue	$\mapsto$	true,		arphiPtrue	$\mapsto$	false,
${\tt falseU}arphi$	$\mapsto$	$\varphi,$		${\tt false}{\sf P}arphi$	$\mapsto$	$\texttt{not}(F\varphi),$
arphiUfalse	$\mapsto$	${\sf G}arphi$	$\operatorname{and}$	arphi Pfalse	$\mapsto$	$F\varphi.$

Figure 5.1.: Rewrite System for computing reduced literals

# Lemma 5.2.3

Let  $\varphi$  be any formula from FOLTL. Then

GGarphi	≡	$G\varphi,$		FFarphi	≡	$F\varphi,$
$\texttt{Gnot}(\varphi)$	≡	$\mathtt{not}(F\varphi),$		$\texttt{Fnot}(\varphi)$	≡	$\mathtt{not}(G\varphi),$
$\texttt{Xnot}(\varphi)$	≡	$\mathtt{not}(X\varphi),$		$\texttt{not}(\texttt{not}(\varphi))$	≡	$\varphi,$
Xfalse	≡	false,		Xtrue	≡	true,
Gfalse	≡	false,		Gtrue	≡	true,
Ffalse	≡	false,		Ftrue	≡	true,
$\texttt{trueU}\varphi$	≡	$F\varphi,$		$\texttt{trueP}\varphi$	≡	$\texttt{not}(\varphi),$
arphiUtrue	≡	true,		arphiPtrue	≡	false,
${\tt falseU}arphi$	≡	$\varphi,$		${\tt false}{\sf P}arphi$	≡	$\mathtt{not}(F\varphi),$
arphiUfalse	≡	${\sf G}arphi$	and	arphiPfalse	≡	$F\varphi.$

These equivalences are easily seen to be correct. In order to define a suitable concept of reducedness we will convert the equivalences into a terminating and confluent rewrite system. The rules of this system are then applied exhaustively to a literal and the resulting nonreducible literal is said to be the *reduced form* of the original literal. The set of rules is given in Figure 5.1.

In order to analyze the properties of the introduced rewrite system we will have to

review some of the basic concepts from term rewriting. The presentation will be standard (see [13], [12] or [83]).

Let  $t \in \mathcal{T}$  be a term. The set of *positions* of t is defined as

- 1.  $\operatorname{Pos}(\mathbf{X}) = \{\varepsilon\}$  if  $t = \mathbf{X} \in \mathcal{X}$  and
- 2.  $\operatorname{Pos}(\mathbf{f}(t_1,\ldots,t_n) = \{\varepsilon\} \cup \bigcup_{i=1}^n \{ip \mid p \in \operatorname{Pos}(t_i)\}.$

Similarly we define positions in literals as follows:

$$\begin{aligned} \operatorname{Pos}(\operatorname{true}) &= \operatorname{Pos}(\operatorname{false}) = \{\varepsilon\},\\ \operatorname{Pos}(\operatorname{p}(t_1, \dots, t_n)) &= \{\varepsilon\} \cup \bigcup_{i=1}^n \{ip \mid p \in \operatorname{Pos}(t_i)\},\\ \operatorname{Pos}(\neg \varphi) &= \operatorname{Pos}(\varphi),\\ \operatorname{Pos}(\mathsf{X}\varphi) &= \operatorname{Pos}(\mathsf{G}\varphi) = \operatorname{Pos}(\mathsf{F}\varphi)\\ &= \{\varepsilon\} \cup \{1p \mid p \in \operatorname{Pos}(\varphi) \text{ and} \\ \operatorname{Pos}(\varphi_1 \mathsf{U}\varphi_2) &= \operatorname{Pos}(\varphi_1 \mathsf{P}\varphi_2)\\ &= \{\varepsilon\} \cup \{1p \mid p \in \operatorname{Pos}(\varphi_1)\} \cup \{2p \mid p \in \operatorname{Pos}(\varphi_2)\}. \end{aligned}$$

So  $\operatorname{Pos}(o) \subseteq \mathbb{N}^*$  for any logical object o.

The term respectively literal at any given position can then be extracted as follows:

- 1.  $t|_{\varepsilon} = t$  for each  $t \in \mathcal{T}$  and
- 2.  $\mathbf{f}(t_1,\ldots,t_i,\ldots,t_n)|_{ip} = t_i|_p$  for  $p \in \operatorname{Pos}(t_i)$ .

Similarly we can extract subparts from literals:

- 1.  $\varphi|_{\varepsilon} = \varphi$  for every literal  $\varphi$ ,
- 2.  $p(t_1, ..., t_i, ..., t_n)|_{ip} = t_i|_p$  for  $p \in Pos(t_i)$ ,
- 3.  $\mathsf{X}\varphi|_{1p} = \mathsf{G}\varphi|_{1p} = \mathsf{F}\varphi|_{1p} = \varphi|_p$  for  $p \in \operatorname{Pos}(\varphi)$ ,

- 4.  $(\varphi_1 \mathsf{U} \varphi_2)|_{1p} = (\varphi_1 \mathsf{P} \varphi_2)|_{1p} = \varphi_1|_p$  for  $p \in \operatorname{Pos}(\varphi_1)$  and
- 5.  $(\varphi_1 \mathsf{U}\varphi_2)|_{2p} = (\varphi_1 \mathsf{P}\varphi_2)|_{2p} = \varphi_2|_p$  for  $p \in \operatorname{Pos}(\varphi_2)$ .

# Example 5.2.2

Let t = f(f(g(X, f(a)))) and  $\varphi = p(f(g(a, f(a))))$  be given. Then

$$Pos(t) = \{\varepsilon, 1, 11, 111, 112, 1121\},$$

$$Pos(\varphi) = Pos(p(f(g(a, f(a)))))$$

$$= \{\varepsilon\} \cup \{1p \mid p \in Pos(f(g(a, f(a))))\}$$

$$= \{\varepsilon\} \cup \{1\varepsilon, 11, 111, 1112, 11121\}$$

$$= \{\varepsilon, 1, 11, 111, 1111, 1112, 11121\}$$

and

$$t|_{112} = f(a)$$
  $t|_{1121} = a$   
 $\varphi|_{11} = f(g(a, f(a)))$   $\varphi|_{11121} = a$ 

The operation of *replacement* is now defined as follows (following the formalisms from [13]):

- 1.  $t'[t]|_{\varepsilon} = t$  for  $t, t' \in \mathcal{X}$ ,
- 2.  $\mathbf{f}(t_1, \dots, t_i, \dots, t_n)[t]|_{ip} = \mathbf{f}(t_1, \dots, t_{i-1}, t_i[t]|_p, t_{i+1}, \dots, t_n)$  for  $t \in \mathcal{X}$  and  $p \in Pos(t_i)$ ,
- 3.  $\varphi[\psi]|_{\varepsilon} = \neg \varphi[\psi]|_{\varepsilon} = \psi$  for literals  $\varphi, \psi$ ,
- 4. a)  $(\mathsf{X}\varphi[\psi])|_{1p} = \mathsf{X}(\varphi[\psi]|_p)$  for  $p \in \operatorname{Pos}(\varphi)$  if  $\varphi|_p$  is a literal and
  - b)  $(\mathsf{X}\varphi[t])|_{1p} = \mathsf{X}(\varphi[t]|_p)$  for  $p \in \operatorname{Pos}(\varphi)$  if  $\varphi|_p$  is a term,
- 5. a)  $(\mathsf{G}\varphi[\psi])|_{1p} = \mathsf{G}(\varphi[\psi]|_p)$  for  $p \in \operatorname{Pos}(\varphi)$  if  $\varphi|_p$  is a literal and
  - b)  $(\mathsf{G}\varphi[t])|_{1p} = \mathsf{G}(\varphi[t]|_p)$  for  $p \in \operatorname{Pos}(\varphi)$  if  $\varphi|_p$  is a term,

7. 
$$(\varphi_1 \mathsf{U} \varphi_2)[\psi]|_{1p} = (\varphi_1 \mathsf{P} \varphi_2)|_{1p} = \varphi_1|_p$$
 for  $p \in \operatorname{Pos}(\varphi_1)$  and

8.  $(\varphi_1 \mathsf{U}\varphi_2)[\psi]|_{2p} = (\varphi_1 \mathsf{P}\varphi_2)|_{2p} = \varphi_2|_p$  for  $p \in \operatorname{Pos}(\varphi_2)$ .

Using replacement we define *reduction* formally as follows. Let a set  $\{l_i \mapsto r_i \mid i = 1, \ldots, n\}$  of rules be given. Then

- 1.  $\varphi \mapsto \psi$  if and only if there is  $p \in \text{Pos}(\varphi)$ ,  $i \in \{1, \dots, n\}$  and a substitution  $\sigma$  such that  $\sigma(l_i) = \varphi|_p$  and  $\psi = \varphi[\sigma(r_i)]|_p$ ,
- 2.  $\varphi \stackrel{n}{\mapsto} \psi$  for  $n \in \{1, 2, ...\}$  if and only if there is a sequence  $\varphi_0, \varphi_1, ..., \varphi_n$  such that  $\varphi_0 = \varphi, \varphi_n = \psi$  and  $\varphi_i \mapsto \varphi_{i+1}$  for  $i \in \{0, 1, ..., n-1\}$  and
- 3.  $\varphi \stackrel{*}{\mapsto} \psi$  if and only if  $\varphi = \psi$  or there is an *n* such that  $\varphi \stackrel{n}{\mapsto} \psi$ .

A rewrite system given by a set of rules  $\{l_i \mapsto r_i \mid i = 1, ..., n\}$  (for some finite *n*) is called *terminating* if the length of reductions is finite. Formally: there is no sequence  $(\varphi_i)_{i \in \mathbb{N}}$  such that  $\varphi_i \neq \varphi_{i+1}$  and  $\varphi_i \mapsto \varphi_{i+1}$  for every  $i \ge 0$ . So in a terminating rewrite system every literal will be reduced to some literal which cannot be reduced any further. Similarly we will call literals  $\varphi_1$  and  $\varphi_2$  *joinable* if there is  $\psi$  such that  $\varphi_1 \stackrel{*}{\mapsto} \psi$  and  $\varphi_2 \stackrel{*}{\mapsto} \psi$ .  $\mapsto$  is called *confluent* (*locally confluent*) if for every  $\varphi$  such that  $\varphi \stackrel{*}{\mapsto} \varphi_1$  $(\varphi \mapsto \varphi_1)$  and  $\varphi \stackrel{*}{\mapsto} \varphi_2$  ( $\varphi \mapsto \varphi_2$ ) there is  $\psi$  such that  $\varphi_1 \stackrel{*}{\mapsto} \psi$  and  $\varphi_2 \stackrel{*}{\mapsto} \psi$ . Termination and confluence are properties of a rewrite system which are essential if one wants to compute normal forms.

In order to analyze the confluence of a rewrite system it suffices to concentrate on a finite set of reductions, so called *critical pairs*. Assume that  $r^{(1)} = l_{i_1} \mapsto r_{i_1}$  and  $r^{(2)} = l_{i_2} \mapsto r_{i_2}$  are two rules. We will call  $r^{(1)}$  and  $r^{(2)}$  overlapping (at position p) if and only if there is a position  $p \in \text{Pos}(l_{i_1})$  and a substitution  $\sigma$  such that  $\sigma = \text{mgu}(l_{i_1}|_p, l_{i_2})$ . Critical pairs are pairs of literals which can be derived from the application of a restricted kind of overlapping rules.

# Definition 5.2.3 (Critical Pair, e.g. Baader and Nipkow [13])

Let  $r^{(1)} = l_{i_1} \mapsto r_{i_1}$  and  $r^{(2)} = l_{i_2} \mapsto r_{i_2}$  be overlapping at position  $p \in \text{Pos}(l_{i_1})$ such that  $l_{i_1}|_p \notin \mathcal{X}$  and let  $\sigma = \text{mgu}(l_{i_1}|_p, l_{i_2})$ . Then  $\langle \sigma(r_{i_1}), \sigma(l_{i_1}[\sigma(r_{i_2})]|_p \rangle$  is called a *critical pair* (with respect to  $\mapsto$ ).

The following classical result shows that confluence of a terminating rewrite system  $\mapsto$  can be proved by checking if critical pairs are joinable.

#### Lemma 5.2.4 (Critical–Pair–Lemma, Knuth and Bendix [90])

A terminating rewrite system  $\mapsto$  is confluent if and only if all critical pairs (with respect to  $\mapsto$ ) are joinable.

From now on we will concentrate on the rewrite system described in Figure 5.1. This system will therefore be denoted as  $\mapsto$ . We will see that  $\mapsto$  indeed has the desired properties. The first property is immediate.

# Lemma 5.2.5

 $\mapsto$  is terminating.

**Proof**. First observe that no application of a rewrite step yields a literal which is longer than the original literal. In particular there are several rules which shorten the literals. Since every literal consists of a finite number of symbols, these rules can only be applied a finite number of times. So now consider the *length-preserving* rules  $Gnot(\varphi) \mapsto not(F\varphi)$ ,  $Fnot(\varphi) \mapsto not(G\varphi)$  and  $Xnot(\varphi) \mapsto not(X\varphi)$ . It is obvious that these rules push negations to the left. But this can also be done only a finite number of times (due to the finite length of literals), so the rules cannot be applied infinitely often. This proves the lemma.

In order to show that  $\mapsto$  is also confluent we will prove that all critical pairs are joinable.

1. Let the rule  $GG\varphi \mapsto G\varphi$  overlap with itself. Then there are two possible rewrite steps (the replaced subliteral is underlined):

$$\begin{array}{rcl} \mathsf{G}\underline{\mathsf{G}}\mathbf{G}\varphi & \mapsto & \mathsf{G}\mathbf{G}\varphi \mbox{ and} \\ \\ \underline{\mathsf{G}}\mathbf{G}\mathbf{G}\varphi & \mapsto & \mathsf{G}\mathbf{G}\varphi. \end{array}$$

Since the literals which emerge from applying these two steps are identical, this pair is clearly joinable.

2. Similarly it can be shown that an overlapping of  $\mathsf{FF}\varphi \mapsto \mathsf{F}\varphi$  with itself is joinable using the following steps:

$$\begin{array}{rcl} \mathsf{F}\underline{\mathsf{F}}\overline{\mathsf{F}}\varphi & \mapsto & \mathsf{F}\mathsf{F}\varphi \mbox{ and} \\ \\ \mathsf{F}\mathsf{F}\mathsf{F}\varphi & \mapsto & \mathsf{F}\mathsf{F}\varphi. \end{array}$$

3. If  $\operatorname{Gnot}(\varphi) \mapsto \operatorname{not}(\mathsf{F}\varphi)$  overlaps with  $\operatorname{not}(\operatorname{not}(\varphi)) \mapsto \varphi$  we have

4. If  $\operatorname{Fnot}(\varphi) \mapsto \operatorname{not}(\mathsf{G}\varphi)$  overlaps with  $\operatorname{not}(\operatorname{not}(\varphi)) \mapsto \varphi$  we have

$$\begin{array}{lcl} \operatorname{Fnot}(\operatorname{not}(\varphi)) & \mapsto & \operatorname{F}\varphi \ \mathrm{and} \\ \\ \overline{\operatorname{Fnot}(\operatorname{not}(\varphi))} & \mapsto & \operatorname{not}(\overline{\operatorname{Gnot}(\varphi)}) \mapsto \underline{\operatorname{not}(\operatorname{not}(\operatorname{F}\varphi))} \mapsto \operatorname{F}\varphi. \end{array}$$

5. If  $\operatorname{Xnot}(\varphi) \mapsto \operatorname{not}(\mathsf{X}\varphi)$  overlaps with  $\operatorname{not}(\operatorname{not}(\varphi)) \mapsto \varphi$  we have

$$\begin{array}{rcl} X\underline{\operatorname{not}(\operatorname{not}(\varphi))} & \mapsto & X\varphi \text{ and} \\ \\ \underline{X}\underline{\operatorname{not}(\operatorname{not}(\varphi))} & \mapsto & \operatorname{not}(X\underline{\operatorname{not}(\varphi)}) \mapsto \underline{\operatorname{not}(\operatorname{not}(X\varphi))} \mapsto X\varphi. \end{array}$$

The other rules of the rewrite system can only overlap in a noncritical way, so each critical pair with respect to  $\mapsto$  is joinable and the following lemma is proved.

# Lemma 5.2.6

 $\mapsto$  is confluent.

From now on let  $\text{Red}(\varphi)$  denote the uniquely determined reduced literal which emerges from the application of the above rewrite system. We have

$$\varphi \equiv \operatorname{Red}(\varphi)$$

(due to the fact that  $\mapsto$  is constructed from a set of semantical equivalences), and therefore the following theorem holds.

# Theorem 5.2.2

For every FOLTL-literal  $\varphi$  there is a uniquely determined normal form  $\text{Red}(\varphi)$  with  $\varphi \equiv \text{Red}(\varphi)$  which can be effectively computed.

In some of the following chapters we will restrict our analysis to reduced literals since these can be handled much easier than general ones.

# 5.2.3. Semantics for programs

We will now show how the semantics of a FOLTL-program P can be characterized in terms of the *stable model semantics* introduced by Gelfond and Lifschitz in [73]. The adaption of the Gelfond-Lifschitz constructions is necessary since rules in PROLOG(+T)-programs may contain negated literals in the tail (note that the problem of inducing stable models, i.e. models for normal (nontemporal) logic programs has been already addressed, i.e. in [128] and [125]). So the results from logic programming regarding definite logic programs such as the existence of monotonic and continuous operators which have fixpoints which can be seen as *least models* (in fact they are identical to the intersection of all Herbrandmodels, see [164]) cannot be easily generalized to PROLOG(+T)-programs anymore. For nontemporal logic programming languages, Gelfond and Lifschitz have extended the classical *immediate consequence operator*  $T_P$  which characterizes the least Herbrand-Model to an operator which, given any subset M of  $B_P$ , computes a least Herbrandmodel of a modified definite program  $P_M$  which (if M is chosen in the *right* way) has the property of being a *stable model* of P (see [73]).

We will now adapt the necessary concepts introduced in [73] in order to deal with PROLOG(+T)-programs.

So assume that a set P consisting of reduced ground rules is given and  $M \subseteq B_P^{\text{FoLTL}}$ is any set of reduced ground atoms built from symbols occurring in P. The program  $P_M$ is then constructed as follows:

- 1. if there is a rule in P such that  $not(\varphi)$  for some  $\varphi \in M$  occurs in the tail of this rule, then this rule is deleted and
- 2. negated atoms in the tails of the remaining clauses are also deleted.

Then  $P_M$  is clearly negation-free in the sense that no rule in  $P_M$  contains a negated atom in its tail. We will see that negation-free sets of PROLOG(+T)-ground-rules have models (in the case of satisfiability). Note that P is in general *not* satisfiable as shown in the example after Definition 5.2.2 on page 71.

For satisfiable negation-free programs we can indeed extend the fixpoint semantics of first order logic programming languages in a straightforward way. Recall that P has been required to consist only of ground rules, that is in general, P is not a finite set of rules anymore. We proceed as follows:

- 1. starting from the empty set  $\emptyset$  we will construct an infinite tree consisting of nodes labeled with subsets of  $B_P^{\text{FoLTL}}$  and
- 2. show that the union of all labels of nodes on any maximal path in this tree is a set–based Herbrand–model of P.

We will use the following extension of the *immediate consequence operator*  $T_P$  (see [164] or [105]). This operator is used in the theory of first order Logic Programming to derive characterizations for the semantics of a definite program. The semantics of a program Pis there given as the *smallest* set of ground atoms of the underlying first order language which P satisfies. Equivalently the semantics is given as the set of all implied ground atoms implied by the program. This set can be characterized as the least fixpoint of an operator (namely the operator  $T_P$  as introduced in [164]) and is easily seen to be uniquely determined (since Herbrand-models of definite logic programs can be intersected yielding Herbrand-models). For characterizing the semantics of PROLOG(+T)-programs we will change the original operator  $T_P$  in a way that allows the treatment of FOLTL-objects rather than only first order objects.

# Definition 5.2.4 (Immediate Consequence Operator for FoLTL)

Let P be any negation-free set of ground rules. The mapping  $T_P^{\text{FoLTL}}: 2^{B_P^{\text{FoLTL}}} \rightarrow 2^{B_P^{\text{FoLTL}}}$  is defined as follows:

$$T_P^{\text{FoLTL}}(I) = \left\{ \varphi \in B_P^{\text{FoLTL}} \mid \text{there is a rule } \varphi := \psi_1, \dots, \psi_k. \text{ in } P \text{ such that} \right\}.$$
$$I \models \psi_1, \dots, \psi_k$$

This operator will be used together with the operator TEMPCLOSURE in alternating order.

Now assume that any negation-free PROLOG(+T)-program P which only contains ground rules is given. We construct an infinite labeled tree T(P) = (V, E, l) with  $l: V \rightarrow 2^{B_P^{FOLTL}}$  such that the following conditions are satisfied:

- 1. V contains a uniquely determined root node  $v_0$ ,
- 2.  $l(v_0) = \emptyset$  and
- 3. for each  $v \in V$  it holds that
  - a) if v is on an even level then there is  $v' \in V$  such that  $(v, v') \in E$  and  $l(v') = T_P^{\text{FoLTL}}(l(v))$  and
  - b) if v is on an odd level then assume that

TEMPCLOSURE
$$(l(v)) = \{\mathcal{T}_1, \ldots, \mathcal{T}_i, \ldots\}$$

and there are nodes  $v'_1, \ldots, v'_i, \cdots \in V$  such that  $(v, v'_i) \in E$  and  $l(v'_i) = \mathcal{T}_i$  for every  $i \geq 1$ .

Given T(P) we define the set p(T(P)) to consist of all maximal paths in T, that is of all (in general infinite) maximal sequences of nodes  $(v_i)_{i \in \mathbb{N}}$  such that  $(v_i, v_{i+1}) \in E$ . If  $\pi \in p(T(P))$  is a maximal path then  $V(\pi)$  denotes the set of nodes which are visited while traversing  $\pi$ .

## Lemma 5.2.7

For every satisfiable set P of negation-free ground rules and for every  $\pi \in p(T(P))$  it holds that

- 1.  $\bigcup_{v \in V(\pi)} l(v) \models P$  and
- 2.  $\bigcup_{v \in V(\pi)} l(v) \models P$  is temporally closed.

**Proof**. The first claim is clear by definition of  $T_P^{\text{FoLTL}}$ . For the proof of the second claim fix any  $\pi \in p(T(P))$ . If  $\bigcup_{v \in V(\pi)} l(v)$  is not temporally closed, then an application of the

operator TEMPCLOSURE would yield a temporally closed superset which shows that  $\pi$  is not maximal in this case. This contradicts the assumption on  $\pi$ . So  $\bigcup_{v \in V(\pi)} l(v)$  is temporally closed.

So the models of P derived by the above construction can be collected in a set  $\Theta_P$  as follows:

$$\Theta_P = \left\{ \bigcup_{v \in V(\pi)} l(v) \mid \pi \in p(T(P)) \right\}.$$

In order to deal with programs containing negation, we will now concentrate on programs  $P_M$  as described above.

# Lemma 5.2.8

Let P be a set of ground PROLOG(+T)-rules and let  $M \subseteq B_P^{FoLTL}$  be a set of atoms occurring in P. If P is satisfiable, then every  $\mathcal{J} \in \Theta_{P_M}$  is a Herbrand-model of P.

**Proof.** Let M be as required. Then  $\mathcal{J} \models P_M$  for every  $\mathcal{J} \in \Theta_{P_M}$ . Assume that  $P_M = \{P_1, \ldots, P_n\}$ , fix  $\mathcal{J} \in \Theta_{P_M}$  and  $i \in \{1, \ldots, n\}$ . Since  $\mathcal{J} \models P_M$  we have  $\mathcal{J} \models P_i$ . Consider the following cases.

**Case 1**  $P_i \in P$ . Then the claim is immediate.

**Case 2**  $P_i \notin P$ . Then  $P_i$  emerges from a rule from P by deleting negated atoms in the tail of  $P_j$ . That is for  $P_i = \varphi := \psi_1, \ldots, \psi_k$ . there is  $P_j = \varphi := \psi_1, \ldots, \psi_k, \bar{\psi}_1, \ldots, \bar{\psi}_l$  in P for  $\bar{\psi}_1, \ldots, \bar{\psi}_l \in B_P^{\text{FoLTL}}$ . Since  $\mathcal{J} \models P_i$  we have  $\mathcal{J} \models \varphi \lor \neg \psi_1, \ldots, \neg \psi_k$  which implies  $\mathcal{J} \models \varphi \lor \neg \psi_1 \lor \cdots \lor \neg \psi_k \lor \bar{\psi}_1 \lor \cdots \lor \bar{\psi}_l$  and therefore  $\mathcal{J} \models P_j$ .

It remains to prove that also the rules C from P which have been deleted during the construction of  $P_M$  are satisfied by  $\mathcal{J}$ . If C is such a rule, then  $C = \varphi :-\psi_1, \ldots, \psi_k, \mathsf{not}(\psi)$ . for some  $\psi \in M$ . Since  $\mathcal{J} \models M$  the claim is immediate. So  $\mathcal{J}$  is a model of P and the lemma is proved.

# 5.3. A Saturation-based temporal Proof Procedure

As we have already mentioned in chapter 5.2 we will now study the operational semantics of PROLOG(+T). This semantics will be studied by presenting a calculus which is both sound and refutation complete. Basically this calculus will be an extension of the wellknown tableaux principle which is a quite popular principle in many areas of automated reasoning. In contrast to PROLOG where an eventually modified (and speed up) version of the *SLD-resolution* principle is chosen for proving logical consequence of goals (and therefore for computing answers to programs) resolution-based approaches are not the best choice for our problem. Consider the following problem: given a program P = $\{P_1, \ldots, P_n\}$  and a goal  $G =: -\varphi_1, \ldots, \varphi_m$ . Assume that *i* is such that  $\varphi_i = \text{Gp}(X)$ . Then  $\varphi_i \equiv p(X) \wedge X\text{Gp}(X)$ . If there is for example a rule  $P_j = XXXp(X) :-q(Y,Y)$ . in *P* the normal SLD-principle will not be able to compute a resolvent. But it is obvious that  $G' = \varphi_1, \ldots, \varphi_{i-1}, p(X), Xp(X), XXp(X), q(Y,Y), XXXXGp(X), \varphi_{i+1}, \ldots, \varphi_m$  should be a resolvent of *G* and  $P_j$ . Therefore we will adapt the tableaux calculus in order to deal with temporal constructs. We will see that the resulting proof procedure is indeed sound and (in some sense) refutation complete.

The study of sound and complete proof-procedures for FoLTL and its fragments dates as far back as to the first contributions of Abadi and Manna (see [3]). It has been pointed out quite early that the whole logic FoLTL is not only undecidable (which is clear since it contains the whole first order logic) but also not recursively enumerable (see e.g. [82] or [30]). Consequently no complete proof-procedure for FoLTL can exist. But restrictions of FoLTL yield recursively enumerable fragments. For example in [82] a restricted usage of temporal operators yields the *monodic fragment* which can be embedded into the monadic second order theory<sup>5</sup>. Omitting the use of U and P yields a logic which can be recursively enumerated. Enumeration can be carried out by a sequent-style calculus (see

<sup>&</sup>lt;sup>5</sup>A formula  $\varphi$  is called *monodic* if every subformula  $\psi$  of  $\varphi$  which has the form  $\psi = \psi_1 \oplus \psi_2$  for  $\oplus \in \{U, P\}$  has at most one free variable. Consequently the *monodic fragment* of first order temporal logic consists of all monodic formulas.

[14]). Other types of proof-procedures for FoLTL-fragments are Hilbert-style calculi (see e.g. [53]), Resolution-style calculi (see [91], [41], [43], [44], [91], [41] and [42]) and ASM-based procedures (see [174]). A good survey of first order temporal logic proof-procedures has been presented recently by C. Castellini in his Ph.D. Thesis (see [27]).

Our proof-procedure which we will present on this section is a tableaux-style procedure. Several modifications of the tableaux-construction for first order logic have been presented in order to derive proof-procedures for first order temporal logic languages (see [28] and [29] for FoLTL and [114] for the first order version of CTL). Nevertheless our method has its justification since it is defined especially for PROLOG(+T)-formulas and since it integrates the reduction-concept for temporal literals introduced above.

# 5.3.1. Tableaux Rules

Since PROLOG(+T) is first order logic based we will have to define a proof procedure which enables us to handle first order logic constructs. The tableaux method (see [66]) is one such method. Basically it constructs a tableaux (which is represented as a directed graph) consisting of nodes which are labeled with sets of formulas. The key property for the soundness and completeness is the following: if  $n_1$  and  $n_2$  are such nodes and  $F_1$ and  $F_2$  are the sets of formulas labeling these two nodes and  $n_2$  is a successor of  $n_1$  then unsatisfiability of  $F_2$  implies unsatisfiability of  $F_1$ .

We will distinguish two principle kinds of tableaux rules here: expansion rules and saturation rules. Expansion rules are rules which consider the first order part of a formula. An example might be the following: if  $\varphi_1 \wedge \varphi_2$  is to be satisfied, then both  $\varphi_1$  and  $\varphi_2$ have to be satisfied at the same time. On the other hand, saturation rules consider the temporal part of formulas. Here we can argue with the example presented above: if  $G\varphi$ is to be satisfied, then  $\varphi$  and  $XG\varphi$  have to be satisfied (in other words  $X^i\varphi$  has to be satisfied for every  $i \geq 0$ ).

The discussion of the proof procedure will now proceed as follows: first we will have

to define certain concepts which are needed for the presentation of the proof procedure. This will include a node format for the nodes of the tableaux. After having introduced the basics we will consider the expansion rules and the saturation rules separately. We will see that the expansion rules are the rules from the *classical* tableaux calculus for first order predicate logic (moreover, the proof procedure to be introduced is capable of handling formulas which are not restricted to the formulas occurring in PROLOG(+T)programs). In order to be applicable we will then present two kinds of termination criteria: criteria for termination in the case of success and in the case of failure. We will see that termination by failure is a much more difficult task than successful termination. The chapter will then be continued by proving the soundness and refutation completeness of the procedure.

#### **Basic Formalisms**

We will now describe the format of the nodes from which a tableaux is built up.

# Definition 5.3.1 (Tableaux node)

A tableaux node is a tuple  $n = (\Phi, F, S)$  where  $\Phi$  is a nonempty set of FoLTL-formulas and F and S are (possibly empty) sets of tableaux nodes. The sets F and S will be referred to as the *father nodes* and the *son nodes* of n.

For the proofs in later sections we will introduce the following abbreviation. For a node n we will denote the set of formulas *labeling* this node by  $\mathcal{F}(n)$ . In other words if  $n = (\Phi, F, S)$ , then  $\mathcal{F}(n) = \Phi$ .

**Definition 5.3.2 (Path)** A path is a sequence  $\pi = (n_0, n_1, \dots, n_k, \dots)$  of tableaux nodes such that  $n_i = (\Phi_i, F_i, S_i)$  for  $i = 0, 1, \dots, F_0 = \emptyset$  and for each  $i \in \{0, 1, \dots\}$  it holds that

 $n_{i+1} \in S_i$ . A node *n* is said to *occur* on  $\pi$  if there is an index *i* such that  $n_i = n$ . If  $\pi = (n_0, n_1, \dots, n_{k-1})$ , the integer *k* is called the *length* of  $\pi$  and  $\pi$  is called *finite* (or *of finite length*). Otherwise (i.e. if no such *k* exists),  $\pi$  is called *infinite* (or *of infinite length*).

Note that in general paths in a tableaux will be of infinite length. This will be of interest in the proof of refutation completeness which we will present later on.

We will now describe the construction of the *initial tableaux node* for the tableaux to be constructed from a formula  $\varphi$ . This node is given as follows:

$$n_0 = (\{\varphi\}, \emptyset, \emptyset) .$$

Whenever we will refer to the *initial node* of a tableaux from now on we will mean a node  $n_0$  constructed in this way.

#### **Expansion Rules**

We will now describe how to construct new nodes from nodes which have already been constructed. The rules to be described in this section will later on be referred to as expansion rules. We assume that  $n = (\Phi, F, S)$  is a node which has been chosen from a set N of nodes. A fixed selection rule has to be chosen in order to select formulas from  $\mathcal{F}(n)$  to which the rules can be applied. The only requirement which we will have to put on the selection rule is the requirement of fairness. Informally fairness means that no application of a rule to a formula from  $\mathcal{F}(n)$  is retarded for an infinite time. Equivalently we can say that every possible inference step is eventually carried out (as long as the procedure does not terminate).

Now let  $\varphi$  be the selected formula. We distinguish the following cases:

1.  $\varphi = \varphi_1 \wedge \varphi_2$ . In this case we create a node  $n' = (\mathcal{F}(n'), F', S')$  as follows:

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\varphi_1 \land \varphi_2\}) \cup \{\varphi_1, \varphi_2\},\$$

 $F' = \{n\}, S' = \emptyset$ . Furthermore we update the set S of sons of n to  $S = S \cup \{n\}$ . This rule will from now on be referred to as  $(\wedge \mathbf{1})$ .

2.  $\varphi = \neg (\varphi_1 \lor \varphi_2)$ . In this case we create a node  $n' = (\mathcal{F}(n'), F', S')$  as follows:

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\neg (\varphi_1 \land \varphi_2)\}) \cup \{\neg \varphi_1, \neg \varphi_2\},\$$

 $F' = \{n\}, S' = \emptyset$ . Furthermore we update the set S of sons of n to  $S = S \cup \{n\}$ . This rule will from now on be referred to as  $(\wedge 2)$ .

3.  $\varphi = \varphi_1 \vee \varphi_2$ . In this case we create two nodes  $n' = (\mathcal{F}(n'), F', S')$  and  $n'' = (\mathcal{F}(n''), F'', S'')$  as follows:

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\varphi_1 \lor \varphi_2\}) \cup \{\varphi_1\},\$$

 $F' = \{n\}, S' = \emptyset$  and

$$\mathcal{F}(n'') = (\mathcal{F}(n) \setminus \{\varphi_1 \lor \varphi_2\}) \cup \{\varphi_2\},\$$

 $F''\{n\} = S' = \emptyset$ . The set S of sons of n will be updated to  $S = S \cup \{n', n''\}$ . This rule will be referred to as  $(\vee \mathbf{1})$ .

4.  $\varphi = \neg (\varphi_1 \land \varphi_2)$ . In this case we create two nodes  $n' = (\mathcal{F}(n'), F', S')$  and  $n'' = (\mathcal{F}(n''), F'', S'')$  as follows:

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\neg (\varphi_1 \land \varphi_2)\}) \cup \{\neg \varphi_1\},\$$

 $F' = \{n\}, S' = \emptyset$  and

$$\mathcal{F}(n'') = (\mathcal{F}(n) \setminus \{\neg (\varphi_1 \land \varphi_2)\}) \cup \{\neg \varphi_2\},\$$

 $F''\{n\} = S' = \emptyset$ . The set S of sons of n will be updated to  $S = S \cup \{n', n''\}$ . This rule will be referred to as  $(\vee 2)$ .

5.  $\varphi = \varphi_1 \rightarrow \varphi_2$ . In this case two nodes  $n' = (\mathcal{F}(n'), F', S')$  and  $n'' = (\mathcal{F}(n''), F'', S'')$ will be constructed with

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\varphi_1 \to \varphi_2\}) \cup \{\neg \varphi_1\},\$$

 $F' = \{n\}, S' = \emptyset$  and

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\varphi_1 \to \varphi_2\}) \cup \{\varphi_2\},\$$

 $F' = \{n\}, S' = \emptyset$ . The set S will be updated to  $S = S \cup \{n', n''\}$ . This rule will be referred to as  $(\to \mathbf{1})$ .

6.  $\varphi = \neg (\varphi_1 \to \varphi_2)$ . Here a node  $n' = (\mathcal{F}(n'), F', S')$  will be constructed with

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\neg (\varphi_1 \to \varphi_2)\}) \cup \{\varphi_1, \neg \varphi_2\},\$$

 $F' = \{n\}$  and  $S' = \emptyset$ . S will be updated to  $S = S \cup \{n'\}$ . The rule will be referred to as  $(\rightarrow 2)$ .

7.  $\varphi = \varphi_1 \leftrightarrow \varphi_2$ . Here we create a node  $n' = (\mathcal{F}(n'), F', S')$  with

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\varphi_1 \leftrightarrow \varphi_2\}) \cup \{\varphi_1 \to \varphi_2, \varphi_2 \to \varphi_1\},\$$

 $F' = \{n\}$  and  $S' = \emptyset$ . S is updated to  $S = S \cup \{n'\}$ . The rule will be referred to

as  $(\leftrightarrow \mathbf{1})$ .

8.  $\varphi = \neg (\varphi_1 \leftrightarrow \varphi_2)$ . Here we create two nodes  $n' = (\mathcal{F}(n'), F', S')$  and  $n'' = (\mathcal{F}(n''), F'', S'')$  with

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\neg (\varphi_1 \leftrightarrow \varphi_2)\}) \cup \{\neg (\varphi_1 \rightarrow \varphi_2)\},\$$

 $F' = \{n\}, S' = \emptyset$  and

$$\mathcal{F}(n'') = (\mathcal{F}(n) \setminus \{\neg (\varphi_1 \leftrightarrow \varphi_2)\}) \cup \{\neg (\varphi_2 \to \varphi_1)\},\$$

 $F'' = \{n\}, S'' = \emptyset$ . Again S is updated to  $S = S \cup \{n', n''\}$ . The rule will be referred to as  $(\leftrightarrow 2)$ .

9.  $\varphi = \neg \neg \psi$ . Here we will create a node  $n' = (\mathcal{F}(n'), F', S')$  with

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\neg \neg \psi\}) \cup \{\psi\},\$$

 $F' = \{n\}$  and  $S' = \emptyset$ . S is updated to  $S = S \cup \{n'\}$  and the rule is referred to as  $(\neg -\mathbf{E})$ .

10.  $\varphi = \neg \forall \mathbf{X} \psi(\mathbf{X})$ . Here we will create a node  $n' = (\mathcal{F}(n'), F', S')$  with

$$\mathcal{F}(n') = \left(\mathcal{F}(n) \setminus \{\neg \forall \mathtt{X} \psi(\mathtt{X})\}\right) \cup \{\exists \mathtt{X} \neg \psi(\mathtt{X})\},\$$

 $F' = \{n\}$  and  $S' = \emptyset$ . S is updated to  $S = S \cup \{n'\}$  and the rule is referred to as  $(\mathbf{R} \forall)$ .

11.  $\varphi = \neg \exists \mathbf{X} \psi(\mathbf{X})$ . Here we will create a node  $n' = (\mathcal{F}(n'), F', S')$  with

$$\mathcal{F}(n') = \left(\mathcal{F}(n) \setminus \{\neg \exists \mathtt{X} \psi(\mathtt{X})\}\right) \cup \{\forall \mathtt{X} \neg \psi(\mathtt{X})\},\$$

 $F' = \{n\}$  and  $S' = \emptyset$ . S is updated to  $S = S \cup \{n'\}$  and the rule is referred to as  $(\mathbf{R}\exists)$ .

12.  $\varphi = \forall X \psi(X)$ . Here we create a new node  $n' = (\mathcal{F}(n'), F', S')$  with

$$\mathcal{F}(n') = \mathcal{F}(n) \cup \{\psi(t)\}$$

for some ground term t which can be built up from symbols occurring in  $\mathcal{F}(n)$ ,  $F' = \{n\}$  and  $S' = \emptyset$ . We update S to  $S = S \cup \{n'\}$  and refer to this rule as  $(\sigma - \mathbf{I_1})$ .

13.  $\varphi = \exists X \psi(X)$ . Here we create a new node  $n' = (\mathcal{F}(n'), F', S')$  with

$$\mathcal{F}(n') = \mathcal{F}(n) \cup \{\psi(\mathbf{X}_{\text{new}})\}$$

for some variable symbol  $\mathbf{X}_{\text{new}} \notin \text{VAR}(\mathcal{F}(n)), F' = \{n\} \text{ and } S' = \emptyset$ . We update S to  $S = S \cup \{n'\}$  and refer to this rule as  $(\sigma - \mathbf{I_2})$ .

14.  $\varphi = \forall \mathbf{X} \psi$  and  $\mathbf{X} \notin VAR(\psi)$ . Here we create a node  $n' = (\mathcal{F}(n'), F', S')$  with

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\forall X\psi\}) \cup \{\psi\},\$$

 $F' = \{n\}$  and  $S' = \emptyset$ . S is updated to  $S = S \cup \{n'\}$  and the rule is referred to as  $(\forall -\mathbf{E})$ .

15.  $\varphi = \exists \mathbf{X} \psi$  and  $\mathbf{X} \notin \operatorname{Var}(\psi)$ . Here we create a node  $n' = (\mathcal{F}(n'), F', S')$  with

$$\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\exists X\psi\}) \cup \{\psi\},\$$

 $F' = \{n\}$  and  $S' = \emptyset$ . S is updated to  $S = S \cup \{n'\}$  and the rule is referred to as  $(\exists -\mathbf{E})$ .

We can group the above rules as follows:  $(\wedge \mathbf{1})$  and  $(\wedge \mathbf{2})$  are called  $\wedge$ -type rules,  $(\vee \mathbf{1})$ 

and  $(\vee 2)$  are called  $\vee$ -type rules,  $(\rightarrow 1)$  and  $(\rightarrow 2)$  are called  $\rightarrow$ -type rules and  $(\leftrightarrow 1)$ and  $(\leftrightarrow 2)$  are called  $\leftrightarrow$ -type rules. Furthermore  $(\neg -\mathbf{E})$  is called the  $\neg$ -elimination rule,  $(\mathbf{R}\forall)$  and  $(\mathbf{R}\exists)$  are called rewrite rules,  $(\sigma-\mathbf{I_1})$  and  $(\sigma-\mathbf{I_2})$  are called substitution rules and  $(\forall-\mathbf{E})$  and  $(\exists-\mathbf{E})$  are called quantifier-elimination rules. The rules are summarized in Figure 5.2. Here the formula above the fraction stroke denotes the selected formula from  $\mathcal{F}(n)$  and the formulas below the fraction strokes denote the formulas created from this formula. The number of columns below the fraction stroke denotes the number of new nodes to be created. For example in

$$\frac{\varphi_1 \land \varphi_2}{\varphi_1}$$
$$\frac{\varphi_2}{\varphi_2}$$

one new node containing two new formulas has to be created.

#### Saturation Rules

In contrast to the expansion rules described above, the saturation rules which will be described deal with the temporal part of a formula rather than with the first order part. We will present two rules for each of the operators G, F, U and P. For the operator X no such rule will be given.

As for the expansion rules we will describe the rules separately by distinguishing the different possibilities of how a selected formula might look like. For the rest of this section assume that *i* is any *fixed* integer and  $n = (\mathcal{F}(n), F, S)$  is the node from which a formula  $\varphi$  is chosen. The saturation rules are given as follows:

1. If  $\varphi = X^i G \psi$ , then we create a new node  $n' = (\mathcal{F}(n'), F', S')$  with  $\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{X^i G \psi\}) \cup \{X^i \psi, X^{i+1} G \psi\}, F' = \{n\}$  and  $S' = \emptyset$ . S is updated to  $S = S \cup \{n\}$ . The rule is referred to as (G1).

$$\begin{array}{cccc} (\wedge \mathbf{1}) & \frac{\varphi_{1} \wedge \varphi_{2}}{\varphi_{1}} & (\wedge \mathbf{2}) & \frac{\neg (\varphi_{1} \vee \varphi_{2})}{\neg \varphi_{1}} \\ \varphi_{2} & \neg \varphi_{2} \end{array} \\ (\vee \mathbf{1}) & \frac{\varphi_{1} \vee \varphi_{2}}{\varphi_{1} \varphi_{2}} & (\vee \mathbf{2}) & \frac{\neg (\varphi_{1} \wedge \varphi_{2})}{\neg \varphi_{1} \neg \varphi_{2}} \\ (\rightarrow \mathbf{1}) & \frac{\varphi_{1} \rightarrow \varphi_{2}}{\neg \varphi_{1} \varphi_{2}} & (\rightarrow \mathbf{2}) & \frac{\neg (\varphi_{1} \rightarrow \varphi_{2})}{\varphi_{1}} \\ \varphi_{2} \rightarrow \varphi_{1} & & & \\ (\leftrightarrow \mathbf{1}) & \frac{\varphi_{1} \leftrightarrow \varphi_{2}}{\varphi_{2} \rightarrow \varphi_{2}} & (\leftrightarrow \mathbf{2}) & \frac{\neg (\varphi_{1} \rightarrow \varphi_{2})}{\neg (\varphi_{1} \rightarrow \varphi_{2}) \neg (\varphi_{2} \rightarrow \varphi_{1})} \\ & & & & \\ \varphi_{2} \rightarrow \varphi_{1} & & & \\ (\neg -\mathbf{E}) & \frac{\neg \neg \varphi}{\varphi} \\ (\forall -\mathbf{E}) & \frac{\forall \mathbf{X}\varphi}{\varphi} & \mathbf{X} \notin \mathrm{VAR}(\varphi) & (\exists -\mathbf{E}) & \frac{\exists \mathbf{X}\varphi}{\varphi} & \mathbf{X} \notin \mathrm{VAR}(\varphi) \\ & & & & \\ (\mathbf{R} \forall) & \frac{\neg \forall \mathbf{X}\varphi(\mathbf{X})}{\exists \mathbf{X} \neg \varphi(\mathbf{X})} & (\mathbf{R} \exists) & \frac{\neg \exists \mathbf{X}\varphi(\mathbf{X})}{\forall \mathbf{X} \neg \varphi(\mathbf{X})} \\ (\sigma -\mathbf{I}_{1}) & \frac{\forall \mathbf{X}\varphi(\mathbf{X})}{\varphi(t)} & \text{for a ground term } t \text{ which can be built up from symbols occurring in } \mathcal{F}(n) \end{array}$$

 $(\sigma - \mathbf{I_2}) \quad \frac{\exists \mathbf{X} \varphi(\mathbf{X})}{\varphi(\mathbf{X}_{\text{new}})} \quad \text{for any } new \text{ variable symbol } \mathbf{X}_{\text{new}}$ 

Figure 5.2.: Expansion Rules

- 2. If  $\varphi = \neg \mathsf{X}^i \mathsf{G} \psi$ , then we create two new nodes  $n' = (\mathcal{F}(n'), F', S')$  and  $n'' = (\mathcal{F}(n''), F'', S'')$  with  $\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\neg \mathsf{X}^i \mathsf{G} \psi\}) \cup \{\neg \mathsf{X}^i \psi\}, F' = \{n\}$  and  $S' = \emptyset$ and  $\mathcal{F}(n'') = (\mathcal{F}(n) \setminus \{\neg \mathsf{X}^i \mathsf{G} \psi\}) \cup \{\neg \mathsf{X}^{i+1} \mathsf{G} \psi\}, F'' = \{n\}$  and  $S'' = \emptyset$ . Additionally S is updated to  $S = S \cup \{n', n''\}$ . The rule will be called (**G2**).
- 3. If  $\varphi = \mathsf{X}^i \mathsf{F} \psi$ , then we create two new nodes  $n' = (\mathcal{F}(n'), F', S')$  and  $n'' = (\mathcal{F}(n''), F'', S'')$  with  $\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\mathsf{X}^i \mathsf{F} \psi\}) \cup \{\mathsf{X}^i \psi\}, F' = \{n\}$  and  $S' = \emptyset$ and  $\mathcal{F}(n'') = (\mathcal{F}(n) \setminus \{\mathsf{X}^i \mathsf{F} \psi\}) \cup \{\mathsf{X}^{i+1} \mathsf{F} \psi\}, F'' = \{n\}$  and  $S'' = \emptyset$ . Additionally S is updated to  $S = S \cup \{n', n''\}$ . The rule will be called (F1).
- 4. If  $\varphi = \neg \mathsf{X}^i \mathsf{F} \psi$ , then we create a new node  $n' = (\mathcal{F}(n'), F', S')$  with  $\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\neg \mathsf{X}^i \mathsf{F} \psi\}) \cup \{\neg \mathsf{X}^i \psi, \neg \mathsf{X}^{i+1} \mathsf{F} \psi\}, F' = \{n\} \text{ and } S' = \emptyset$ . S is updated to  $S = S \cup \{n\}$ . The rule is referred to as  $(\mathsf{F2})$ .
- 5. If  $\varphi = \mathsf{X}^{i}\psi_{1}\mathsf{U}\psi_{2}$ , then we create two new nodes  $n' = (\mathcal{F}(n'), F', S')$  and  $n'' = (\mathcal{F}(n''), F'', S'')$  with  $\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\mathsf{X}^{i}\psi_{1}\mathsf{U}\psi_{2}\}) \cup \{\mathsf{X}^{i}\psi_{2}\}, F' = \{n\}$  and  $S' = \emptyset$ and  $\mathcal{F}(n'') = (\mathcal{F}(n) \setminus \{\mathsf{X}^{i}\psi_{1}\mathsf{U}\psi_{2}\}) \cup \{\mathsf{X}^{i}\psi_{1} \wedge \mathsf{X}^{i+1}\psi_{1}\mathsf{U}\psi_{2}\}, F'' = \{n\}$  and  $S'' = \emptyset$ . Additionally S is updated to  $S = S \cup \{n', n''\}$ . The rule will be called (U1).
- 6. If  $\varphi = \neg \mathsf{X}^{i}\psi_{1}\mathsf{U}\psi_{2}$ , then we create a new node  $n' = (\mathcal{F}(n'), F', S')$  with  $\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\neg \mathsf{X}^{i}\psi_{1}\mathsf{U}\psi_{2}\}) \cup \{\neg \mathsf{X}^{i}\psi_{2}, \neg \mathsf{X}^{i}\psi_{1} \lor \neg \mathsf{X}^{i+1}\psi_{1}\mathsf{U}\psi_{2}\}, F' = \{n\} \text{ and } S' = \emptyset. S$  is updated to  $S = S \cup \{n\}$ . The rule is referred to as  $(\mathsf{U}\mathbf{2})$ .
- 7. If  $\varphi = \mathsf{X}^{i}\psi_{1}\mathsf{P}\psi_{2}$ , then we create a new node  $n' = (\mathcal{F}(n'), F', S')$  with  $\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\mathsf{X}^{i}\psi_{1}\mathsf{P}\psi_{2}\}) \cup \{\neg\mathsf{X}^{i}\psi_{2}, \mathsf{X}^{i}\psi_{1} \lor \mathsf{X}^{i+1}\psi_{1}\mathsf{P}\psi_{2}\}, F' = \{n\} \text{ and } S' = \emptyset. S \text{ is updated to } S = S \cup \{n\}.$  The rule is referred to as (P1).
- 8. If  $\varphi = \neg \mathsf{X}^{i}\psi_{1}\mathsf{P}\psi_{2}$ , then we create two new nodes  $n' = (\mathcal{F}(n'), F', S')$  and  $n'' = (\mathcal{F}(n''), F'', S'')$  with  $\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\neg \mathsf{X}^{i}\psi_{1}\mathsf{P}\psi_{2}\}) \cup \{\mathsf{X}^{i}\psi_{2}\}, F' = \{n\}$  and  $S' = \emptyset$  and  $\mathcal{F}(n'') = (\mathcal{F}(n) \setminus \{\neg \mathsf{X}^{i}\psi_{1}\mathsf{P}\psi_{2}\}) \cup \{\neg \mathsf{X}^{i}\psi_{1} \wedge \neg \mathsf{X}^{i+1}\psi_{1}\mathsf{P}\psi_{2}\}, F'' = \{n\}$  and

G-type

$$(G1) \quad \frac{\mathsf{X}^{i}\mathsf{G}\varphi}{\mathsf{X}^{i}\varphi} \qquad (G2) \quad \frac{\neg\mathsf{X}^{i}\mathsf{G}\varphi}{\neg\mathsf{X}^{i}\varphi \quad \neg\mathsf{X}^{i+1}\mathsf{G}\varphi}$$

F-type

(F1) 
$$\frac{\mathsf{X}^{i}\mathsf{F}\varphi}{\mathsf{X}^{i}\varphi \ \mathsf{F}^{i+1}\mathsf{F}\varphi}$$
 (F2)  $\frac{\neg\mathsf{X}^{i}\mathsf{F}\varphi}{\neg\mathsf{X}^{i}\varphi}$   
 $\neg\mathsf{X}^{i+1}\mathsf{F}\varphi$ 

U-type

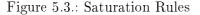
$$(U1) \quad \frac{\mathsf{X}^{i}\varphi_{1}\mathsf{U}\varphi_{2}}{\mathsf{X}^{i}\varphi_{2} \quad \mathsf{X}^{i}\varphi_{1}\wedge\mathsf{X}^{i+1}\varphi_{1}\mathsf{U}\varphi_{2}} \qquad (U2) \quad \frac{\neg\mathsf{X}^{i}\varphi_{1}\mathsf{U}\varphi_{2}}{\neg\mathsf{X}^{i}\varphi_{2}} \\ \neg\mathsf{X}^{i}\varphi_{1}\vee\neg\mathsf{X}^{i+1}\varphi_{1}\mathsf{U}\varphi_{2}$$

P-type

$$(P1) \quad \frac{\mathsf{X}^{i}\varphi_{1}\mathsf{P}\varphi_{2}}{\neg\mathsf{X}^{i}\varphi_{2}} \qquad (P2) \quad \frac{\neg\mathsf{X}^{i}\varphi_{1}\mathsf{P}\varphi_{2}}{\mathsf{X}^{i}\varphi_{1}\vee\mathsf{X}^{i+1}\varphi_{1}\mathsf{P}\varphi_{2}}$$

**Reduction Rule** 

$$(\mathbf{Red}) \quad \frac{\varphi}{\operatorname{Red}(\varphi)} \quad \text{if } \operatorname{Red}(\varphi) \neq \varphi$$



 $S'' = \emptyset$ . Additionally S is updated to  $S = S \cup \{n', n''\}$ . The rule will be called (P2).

Additionally we will have to use the following rule (called the *reduction rule*): If  $\varphi$  is a literal and  $\text{ReD}(\varphi) \neq \varphi$ , then we create a new node  $n' = (\mathcal{F}(n'), F', S')$  with  $\mathcal{F}(n') = (\mathcal{F}(n) \setminus \{\varphi\}) \cup \{\text{ReD}(\varphi)\}$ , set  $F' = \{n\}$  and  $S' = \emptyset$  and update S to  $S = S \cup \{n'\}$ . As for the expansion rules the saturation rules are summarized in Figure 5.3.

# **Termination Criteria**

In order to be useful the tableaux procedure needs some criteria to indicate when the expansion of a node can be aborted. This is a nontrivial task since in general (and in contrast to first order logic) there is *nearly always* a rule which may be applied to some

formula. To be more precise if n is a tableaux node and if there is at least one formula  $\varphi \in \mathcal{F}(n)$  such that  $\varphi$  contains one of the operators G, F, U and P, there is an infinite sequence  $(n_i)_{i=j_0}^{\infty}$  of tableaux nodes such that  $n_{j_0} = n$  and each  $n_{j_{i+1}}$  emerges from  $n_{j_i}$  by application of a saturation rule. Consider for example a node n such that  $G\varphi \in \mathcal{F}(n)$  for some formula  $\varphi$ . Then a node n' can be constructed such that  $\{\varphi, XG\varphi\} \subseteq \mathcal{F}(n')$  (by application of the rule (G1)). This rule can then be applied to n' yielding n'' such that  $\{X\varphi, XXG\varphi\} \subseteq \mathcal{F}(n'')$  and so forth. This example might illustrate the need of some more sophisticated criteria of when to abort the expansion of nodes.

We will adapt the concept of *closedness* of a node which is known from first order logic to include the temporal operators. In first order logic a node is called *closed* if there are literals  $\varphi_1, \varphi_2 \in \mathcal{F}(n)$  such that  $\varphi_2 = \neg \psi$  and  $\varphi_1$  and  $\varphi_2$  are unifiable. For first order logical literals this is adequate but for FOLTL-literals we need a more complicated criterion since the syntactical form of a literal is in general not unique (consider e.g.  $\varphi_1 = \operatorname{not}(Xp(X))$  and  $\varphi_2 = \operatorname{Xnot}(p(X))$  where  $\varphi_1 \neq \varphi_2$  but  $\varphi_1 \equiv \varphi_2$ ).

# Definition 5.3.3

Let n be a tableaux node. Then n is called *closed* if there are formulas  $\varphi_1$  and  $\varphi_2$  from  $\mathcal{F}(n)$  such that

- 1.  $\operatorname{Red}(\varphi_1)$  and  $\operatorname{Red}(\neg \varphi_2)$  are unifiable or
- 2. there is an  $i \ge 0$  such that  $\operatorname{RED}(\varphi_1) = \mathsf{X}^i \psi_1^{(1)} \mathsf{P} \psi_2^{(1)}$  and  $\operatorname{RED}(\varphi_2) = \mathsf{X}^i \psi^{(2)}$  and  $\operatorname{RED}(\psi_2^{(1)})$  and  $\operatorname{RED}(\psi^{(2)})$  are unifiable or
- 3. there is an  $i \ge 0$  such that  $\operatorname{RED}(\varphi_1) = \neg \mathsf{X}^i \psi_1^{(1)} \mathsf{U} \psi_2^{(1)}$  and  $\operatorname{RED}(\varphi_2) = \mathsf{X}^i \psi^{(2)}$  and  $\operatorname{RED}(\psi_2^{(1)})$  and  $\operatorname{RED}(\psi^{(2)})$  are unifiable.

A path  $\pi$  is called *closed* if it contains a closed node. Otherwise  $\pi$  is called *open*.

The definition of closedness presented here is more complicated then necessary. It

would suffice to define closedness solely by the first point of the definition. However, the two other points allow detection of closedness at an earlier point of time which might speed up the proof procedure.

Given the definition of closedness we are able to detect several nontrivial cases of closed nodes.

- 1. Assume that  $\{\neg Xp(X), \neg X \neg p(X)\} \subseteq \mathcal{F}(n)$ . Then  $\operatorname{Red}(\neg Xp(X)) = \neg Xp(X)$  and  $\operatorname{Red}(\neg \neg X \neg p(X)) = \neg Xp(X)$  and an mgu is given by  $\varepsilon$ . So *n* is closed.
- 2. Assume that  $\{\mathsf{GXF}\neg\mathsf{Gp}(X,a),\mathsf{FFXGGGp}(b,Y)\}\subseteq \mathcal{F}(n)$ . Then

$$\operatorname{Red}(\mathsf{GXF}\neg\mathsf{Gp}(\mathtt{X},\mathtt{a})) = \neg\mathsf{FXGp}(\mathtt{X},\mathtt{a})$$

and

$$\operatorname{Red}(\neg \mathsf{FFXGGGp}(\mathtt{b}, \mathtt{Y})) = \neg \mathsf{FXGp}(\mathtt{b}, \mathtt{Y}).$$

An mgu is given by  $\left\{\frac{\mathbf{X}}{\mathbf{b}}, \frac{\mathbf{Y}}{\mathbf{a}}\right\}$ . So *n* is closed.

3. Finally assume that  $\{XXFp(a)Pq(b), X\neg XG\neg p(X)\} \subseteq \mathcal{F}(n)$ . Here we have  $\operatorname{RED}(XXFp(a)Pq(b)) = XXFp(a)Pq(b)$  and  $\operatorname{RED}(X\neg XG\neg p(X)) = XXFp(X)$  which are unifiable with the mgu  $\{\frac{X}{a}\}$ . So *n* is closed.

Using this definition of closedness of nodes we can state the following criterion:

**Termination by Closedness** If n is a node such that n is closed. Then n can be skipped. Here the term *skipping* means that it is not necessary to try to apply expansion and/or saturation rules to n. As we will see, closedness of a node n corresponds to unsatisfiability of  $\mathcal{F}(n)$ . Consequently we will have to apply rules in such a way that every path starting from the initial node leads to a closed node.

The termination criterion from above can be seen as a kind of *success criterion*, that is it enables a proof procedure to determine if the actual node has to be expanded or not. In contrast we can also state a dual (but much weaker) criterion of when the *complete* construction can be aborted. In order to state this criterion we need some more definitions.

# Definition 5.3.4

Let  $n = (\Phi, F, S)$  be a tableaux node. n is called *disjunctively expanded* if only the rules (G1) and (F2) can be applied to formulas from  $\mathcal{F}(n)$  and no subformula of a formula in  $\mathcal{F}(n)$  contains U or P.

So a node is disjunctively expanded if every possible continuation of the path from the root node leading to this node does not contain a node which contains a formula which can be split due to the presence of a disjunction symbol.

In addition to the property of being disjunctively expanded we have to state a property of a path which denotes the fact that every possible instantiation has indeed been carried out.

## Definition 5.3.5

Let  $\pi = (n_0, \ldots, n_k)$  be a path in a tableaux. Then  $\pi$  is called *completely instantiated* if for every ground term t occurring in  $\mathcal{F}(n_0)$  and every node  $n_i$  on  $\pi$  such that there is a formula  $\forall \mathbf{X} \varphi(\mathbf{X}) \in \mathcal{F}(n_i)$  there is a node  $n_j$  on  $\pi$  with  $j \ge i$  and  $\varphi(t) \in \mathcal{F}(n_j)$ .

These two definitions enable us to state the following criterion for termination by failure.

**Termination by Failure** Let  $\pi = (n_0, ..., n)$  be a completely instantiated path and let n be disjunctively expanded. Then the tableaux construction may be aborted if there are no formulas  $\varphi_1, \varphi_2 \in \mathcal{F}(n)$  and no  $i \ge 0$  such that at least one of the following conditions is fulfilled:

- 1.  $\operatorname{RED}(\varphi_1) = \mathsf{X}^i \psi_1$ ,  $\operatorname{RED}(\varphi_2) = \neg \mathsf{X}^i \psi_2$  and  $\operatorname{RED}(\psi_1)$  and  $\operatorname{RED}(\psi_2)$  are unifiable,
- 2.  $\operatorname{ReD}(\varphi_1) = \neg X^i \psi_1$ ,  $\operatorname{ReD}(\varphi_2) = X^j G \psi_2$  for some  $j \leq i$  and  $\operatorname{ReD}(\psi_1)$  and  $\operatorname{ReD}(\psi_2)$  are unifiable or
- 3.  $\operatorname{RED}(\varphi_1) = \neg \mathsf{X}^i \psi_1$ ,  $\operatorname{RED}(\varphi_2) = \neg \mathsf{X}^j \mathsf{F} \psi_2$  for some  $j \leq i$  and  $\operatorname{RED}(\psi_1)$  and  $\operatorname{RED}(\psi_2)$  are unifiable.

This criterion is sound as the following lemma states.

#### Lemma 5.3.1

Let n be a node on an open path  $\pi$  in a such that the criterion Termination by Failure holds for n. Then  $\mathcal{F}(n)$  is satisfiable.

**Proof.** Let  $\mathcal{F}(n)$  be as required. Then every  $\varphi \in \mathcal{F}(n)$  is either a first order literal or a FOLTL-literal containing at most the operators X, G and F (if at all). Define the set  $\mathcal{J}_{\pi}$  as follows:

$$\mathcal{J}_{\pi} = \{ \varphi \in \mathcal{F}(n) \mid \text{VAR}(\varphi) = \emptyset, \text{Red}(\varphi) \text{ is positive } \}$$

Then  $\mathcal{J}_{\pi} \models \mathcal{F}(n)$  and every temporally closed superset of  $\mathcal{J}_{\pi}$  is a Herbrand-model of  $\mathcal{F}(n)$ .

For the sake of simplicity we will introduce another formal concept. A node n is said to *fail* if the criterion *Termination by Failure* can be applied to the path  $\pi_n$  leading from the root node to n.

#### The Tableaux–Procedure

Having defined the rules of interest and the termination criteria, the definition of a proof procedure based on these rules is almost immediate. Assume that  $P = \{P_1, \ldots, P_n\}$  is a PROLOG(+T)-program and assume that  $G = \psi_1 \wedge \cdots \wedge \psi_m$  is a PROLOG(+T)-goal. As always assume that the  $P_i$  have the form  $P_i = \varphi_i :-\psi_1^{(i)}, \ldots, \psi_{n_i}^{(i)}$ . We construct the following formula  $\varphi_{P,G}$  to be proven unsatisfiable:

$$\begin{aligned} \varphi_{P,G} &= \left(\bigwedge_{i=1}^{n} P_{i}\right) \wedge :-G \\ &\equiv \left(\bigwedge_{i=1}^{n} \varphi_{i} :-\psi_{1}^{(i)}, \dots, \psi_{n_{i}}^{(i)}\right) \wedge \left(:-\psi_{1}, \dots, \psi_{m}\right) \\ &\equiv \left(\bigwedge_{i=1}^{n} \psi_{1}^{(i)} \wedge \dots \wedge \psi_{n_{i}}^{(i)} \to \varphi_{i}\right) \wedge \left(\bigvee_{i=1}^{m} \neg \psi_{i}\right). \end{aligned}$$

We then construct the initial tableaux node  $n_0 = (\{\varphi_{P,G}\}, \emptyset, \emptyset)$  as described above and define two sets of nodes. The set N will consist of *all* nodes which have been constructed so far while the set U will contain all *unprocessed* nodes, that is nodes which have not yet been completely expanded. Initially we have  $N = U = \{n_0\}$ . We assume that U is realized as a queue, that is elements can only be taken from the front of U and put at the back of U. This ensures fairness of the node selection rule. See the discussion on page 89 for a treatment of the question of fairness regarding the selection rule for formulas. While U is not empty and no termination criterion is applicable we take the first node  $n_{\rm act}$  from U and search for the first formula and the first rule which can be applied to this formula. New nodes are then created according to the definition of the rule which has been applied. We formalize this algorithm in Algorithm 1. There we will use the following notations: since U is assumed to be a queue structure, the only accessible element is the element at the front of U. This element will be returned by the function U.first. Consequently the operation U.pop will remove the first element from U. Insertion of elements is only possible at the end of U. So if  $U = \{n_1, \ldots, n_k\}$  and  $N_{\text{new}} = \{n_{\text{new}}\}$  (respectively  $N_{\text{new}} = \{n_{\text{new}}^1, n_{\text{new}}^2\}$ ) is a set of newly created nodes, then  $U \cup N_{\text{new}} = \{n_1, \dots, n_k, n_{\text{new}}\} \text{ (respectively } U \cup N_{\text{new}} = \{n_1, \dots, n_k, n_{\text{new}}^1, n_{\text{new}}^2\}).$ 

# Algorithm 1 Tableaux algorithm for ground goals Input:

- PROLOG(+T)-program P
- PROLOG(+T)-ground goal :- $G = G_1, \ldots, G_n$

**Output**: yes iff  $P \models G_1 \land \cdots \land G_n$ 

1: construct  $\varphi_{P,G}$ 2:  $n_0 \leftarrow (\{\varphi_{P,G}\}, \emptyset, \emptyset)$ 3:  $N \leftarrow \{n_0\}, U \leftarrow \{n_0\}$ **Require:** U is realized as a queue 4: while  $U \neq \emptyset$  do  $n_{\mathrm{act}} \leftarrow U.\mathrm{first}$ 5:U.pop 6: 7: if  $\mathcal{F}(n_{\text{act}})$  is not closed then if n fails then 8: 9:return no 10:else select a formula  $\varphi \in \mathcal{F}(n_{\mathrm{act}})$  and a rule R applicable to  $\varphi$ 11: apply R to  $\varphi$ 12: $N_{\text{new}} \leftarrow (\text{set of}) \text{ node}(s) \text{ created by rule } R$ 13: $N \leftarrow N \cup N_{\text{new}}$ 14: $U \leftarrow U \cup N_{\text{new}}$ 15:end if 16:end if 17:18: end while 19: return yes

#### 5.3.2. Soundness and Completeness Issues

We will now address the topic of proving that the tableaux method described in the foregoing section is sound and refutation complete. The first part will be considered with soundness. Therefore we have to note that each node  $n = (\mathcal{F}(n), F, S)$  in a tableaux can be seen as the root of a tableaux starting at this node.

The proofs from this section closely follow the proofs for the nontemporal tableaux procedure as presented in [20]. Our contribution is the treatment of the temporal constructs.

# Lemma 5.3.2

Let n be any tableaux node which is the root of a closed tableaux. Then F(n) is unsatisfiable.

**Proof**. Let *h* denote the height of the tableaux  $\mathcal{T}$  rooted by *n*, that is *h* is the length of the longest path starting at *n*. We proceed by induction on *h*. First assume that h = 0. Then *n* is a leaf and since  $\mathcal{T}$  is closed, we can distinguish the following three cases:

- **Case 1** There is a pair  $\varphi_1, \varphi_2$  of literals in  $\mathcal{F}(n)$  such that  $\operatorname{RED}(\varphi_1)$  and  $\operatorname{RED}(\neg \varphi_2)$  are unifiable. So clearly  $\mathcal{F}(n)$  is unsatisfiable.
- **Case 2** There is  $i \ge 0$  such that  $\operatorname{RED}(\varphi_1) = \mathsf{X}^i \psi_1^{(1)} \mathsf{P} \psi_2^{(1)}$ ,  $\operatorname{RED}(\varphi_2) = \mathsf{X}^i \psi^{(2)}$  and  $\operatorname{RED}(\psi_2^{(1)})$  and  $\operatorname{RED}(\psi^{(2)})$  are unifiable. Let  $\sigma = \operatorname{mgu}(\operatorname{RED}(\varphi_1), \operatorname{RED}(\varphi_2))$  and  $\psi = \sigma(\psi^{(2)}) = \sigma(\psi_2^{(1)})$  be given. Then we have

$$\begin{split} \sigma(\operatorname{RED}(\varphi_{1})) &\wedge \sigma(\operatorname{RED}(\varphi_{2})) = \sigma(\mathsf{X}^{i}\psi_{1}^{(1)}\mathsf{P}\psi_{2}^{(1)}) \wedge \sigma(\mathsf{X}^{i}\psi^{(2)}) \\ &= \mathsf{X}^{i}\sigma(\psi_{1}^{(1)})\mathsf{P}\sigma(\psi_{2}^{(1)}) \wedge \mathsf{X}^{i}\sigma(\psi^{(2)}) \\ &\equiv \neg\mathsf{X}^{i}\sigma(\psi_{2}^{(1)}) \wedge (\mathsf{X}^{i}\sigma(\psi_{1}^{(1)}) \vee \mathsf{X}^{i+1}\sigma(\psi_{1}^{(1)})\mathsf{P}\sigma(\psi_{2}^{(1)})) \wedge \mathsf{X}^{i}\sigma(\psi^{(2)}) \\ &\equiv \neg\mathsf{X}^{i}\psi \wedge \mathsf{X}^{i}\psi \wedge (\mathsf{X}^{i}\sigma(\psi_{1}^{(1)}) \vee \mathsf{X}^{i+1}\sigma(\psi_{1}^{(1)})\mathsf{P}\sigma(\psi_{2}^{(1)})) \\ &\equiv \texttt{false} \wedge (\mathsf{X}^{i}\sigma(\psi_{1}^{(1)}) \vee \mathsf{X}^{i+1}\sigma(\psi_{1}^{(1)})\mathsf{P}\sigma(\psi_{2}^{(1)})) \\ &\equiv \texttt{false}. \end{split}$$

so clearly  $\mathcal{F}(n)$  is unsatisfiable.

**Case 3** There is  $i \ge 0$  such that  $\operatorname{RED}(\varphi_1) = \neg \mathsf{X}^i \psi_1^{(1)} \mathsf{U} \psi_2^{(1)}$ ,  $\operatorname{RED}(\varphi_2) = \mathsf{X}^i \psi^{(2)}$  and  $\operatorname{RED}(\psi_2^{(1)})$  and  $\operatorname{RED}(\psi^{(2)})$  are unifiable. Again let  $\sigma = \operatorname{mgu}(\operatorname{RED}(\varphi_1), \operatorname{RED}(\varphi_2))$  and  $\psi = \sigma(\psi^{(2)}) = \sigma(\psi_2^{(1)})$  be given. Then we have

$$\begin{split} &\sigma(\operatorname{Red}(\varphi_{1})) \wedge \sigma(\operatorname{Red}(\varphi_{2})) = \sigma(\neg \mathsf{X}^{i}\psi_{1}^{(1)} \, \mathsf{U}\psi_{2}^{(1)}) \wedge \sigma(\mathsf{X}^{i}\psi^{(2)}) \\ &= \neg \mathsf{X}^{i}\sigma(\psi_{1}^{(1)}) \, \mathsf{U}\sigma(\psi_{2}^{(1)}) \wedge \mathsf{X}^{i}\sigma(\psi^{(2)}) \\ &\equiv \neg \mathsf{X}^{i}\sigma(\psi_{2}^{(1)}) \wedge (\neg \mathsf{X}^{i}\sigma(\psi_{1}^{(1)}) \vee \neg \mathsf{X}^{i+1}\sigma(\psi_{1}^{(1)} \, \mathsf{U}\psi_{2}^{(1)})) \wedge \mathsf{X}^{i}\sigma(\psi^{(2)}) \\ &\equiv \neg \mathsf{X}^{i}\psi \wedge \mathsf{X}^{i}\psi \wedge (\neg \mathsf{X}^{i}\sigma(\psi_{1}^{(1)}) \vee \neg \mathsf{X}^{i+1}\sigma(\psi_{1}^{(1)} \, \mathsf{U}\psi_{2}^{(1)})) \\ &\equiv \text{false} \wedge (\neg \mathsf{X}^{i}\sigma(\psi_{1}^{(1)}) \vee \neg \mathsf{X}^{i+1}\sigma(\psi_{1}^{(1)} \, \mathsf{U}\psi_{2}^{(1)})) \\ &\equiv \text{false.} \end{split}$$

So  $\mathcal{F}(n)$  is unsatisfiable.

In every of the above cases we have shown that  $\mathcal{F}(n)$  is unsatisfiable. So the case that h = 0 is proved.

Now assume that h > 0. Then a rule has been applied to n yielding one or two successor nodes n' and n''. Since  $\mathcal{T}$  is closed, the tableaux  $\mathcal{T}'$  and  $\mathcal{T}''$  rooted by n' and n'' are also closed. Furthermore their height is h - 1 and so by induction  $\mathcal{F}(n')$  and  $\mathcal{F}(n'')$  are both unsatisfiable. We distinguish the following cases of how n might have been expanded.

**Case 1** Rule ( $\wedge$  1) has been applied. Then  $\mathcal{F}(n) = \{\varphi_1 \land \varphi_2\} \cup F'$  and  $\mathcal{F}(n') = F' \cup \{\varphi_1, \varphi_2\}$ . Since  $\mathcal{F}(n')$  is unsatisfiable we have  $\mathcal{J} \not\models F(n')$  for every  $\mathcal{J}$ . Fix one such  $\mathcal{J}$ . Since  $\mathcal{J} \not\models F(n')$  at least one of the formulas in  $\mathcal{F}(n')$  is not satisfied by  $\mathcal{J}$ . There are three possibilities:

**Case 1.1** There is  $\varphi_0 \in F'$  such that  $\mathcal{J} \not\models \varphi_0$ . This immediately gives  $\mathcal{J} \not\models \mathcal{F}(n)$ . **Case 1.2**  $\mathcal{J} \not\models \varphi_1$ . Then  $\mathcal{J} \not\models \varphi_1 \land \varphi_2$  and so  $\mathcal{J} \not\models \mathcal{F}(n)$ . **Case 1.3**  $\mathcal{J} \not\models \varphi_2$  This is analogous to the foregoing case.

So this proves that  $\mathcal{F}(n)$  is unsatisfiable.

- **Case 2** Rule ( $\vee$  1) has been applied. Then  $\mathcal{F}(n) = \{\varphi_1 \lor \varphi_2\} \cup F'$  and  $\mathcal{F}(n') = F' \cup \{\varphi_1\}$ ,  $\mathcal{F}(n'') = F' \cup \{\varphi_2\}$ . By induction both  $\mathcal{F}(n')$  and  $\mathcal{F}(n'')$  are unsatisfiable. Fix any  $\mathcal{J}$ . Then  $\mathcal{J} \not\models \mathcal{F}(n')$  and  $\mathcal{J} \not\models \mathcal{F}(n'')$ . We again distinguish two cases:
  - **Case 2.1** There is  $\varphi_0 \in F'$  such that  $\mathcal{J} \not\models \varphi_0$ . This case is identical to case 1.1.
  - **Case 2.2**  $\mathcal{J} \models \varphi_0$  for every  $\varphi_0 \in F'$ . Then the unsatisfiability of  $\mathcal{F}(n')$  and  $\mathcal{F}(n'')$  gives  $\mathcal{J} \not\models \varphi_1$  and  $\mathcal{J} \not\models \varphi_2$ . Therefore  $\mathcal{J} \not\models \varphi_1 \lor \varphi_2$  and so  $\mathcal{J} \not\models \mathcal{F}(n)$ .
- **Case 3** Rule  $(\neg -\mathbf{E})$  has been applied. Here  $\mathcal{F}(n') = \{\neg \neg \varphi\} \cup F'$  and  $\mathcal{F}(n') = F' \cup \{\varphi\}$ . Since  $\varphi \equiv \neg \neg \varphi$  we have  $\mathcal{J} \models \varphi$  if and only if  $\mathcal{J} \models \neg \neg \varphi$ . Since  $\mathcal{F}(n')$  is unsatisfiable F(n) is also unsatisfiable.
- **Case 4** A rewrite rule or the reduction rule has been applied. Then the claim is proved exactly as in the foregoing case.
- **Case 5** One of the rules  $(\forall -\mathbf{E})$  and  $(\exists -\mathbf{E})$  has been applied. This case is trivial since if  $\mathbf{X} \notin \operatorname{VAR}(\varphi)$  we have  $\forall \mathbf{X} \varphi \equiv \exists \mathbf{X} \varphi \equiv \varphi$ . So if n' is created from n by application of one of the above rules then  $\mathcal{F}(n')$  is unsatisfiable if and only if  $\mathcal{F}(n)$  is unsatisfiable.
- **Case 6** Rule  $(\sigma \mathbf{I_1})$  has been applied. Then  $\mathcal{F}(n) = \{\forall \mathbf{X} \varphi(\mathbf{X})\} \cup F'$  and  $\mathcal{F}(n') = F' \cup \{\varphi(t)\}$ . Again fix any  $\mathcal{J}$ . Then  $\mathcal{J} \not\models \mathcal{F}(n')$ . The possible cases are:
  - **Case 6.1**  $\mathcal{J} \not\models \varphi_0$  for some  $\varphi_0 \in F'$ . Then we have the situation from case 1.1 and case 2.1.
  - **Case 6.2**  $\mathcal{J} \models \varphi_0$  for every  $\varphi_0 \in F'$ . Then  $\mathcal{J} \not\models \varphi(t)$  that is  $\mathcal{J} \not\models \sigma(\varphi(x))$  for  $\sigma = \left\{\frac{\mathbf{X}}{t}\right\}$  and therefore  $\mathcal{J} \not\models \forall \mathbf{X} \varphi(\mathbf{X})$ . So  $\mathcal{F}(n)$  is unsatisfiable.

This proves the lemma.

The above lemma especially holds for the starting node of a tableaux so we have the following easy corollary.

#### Corollary 5.3.1 (Soundness)

Let  $\varphi$  be any FOLTL-formula. If the tableaux for  $\varphi$  is closed, then  $\varphi$  is unsatisfiable.

Although rather technical, the proof that the tableaux construction is sound is quite straightforward exploiting only basic proof techniques. Proving refutation completeness is much more tricky. The tableaux construction technique known from first order logic uses so called *Hintikka-sets* (see [66]) in order to prove that for an unsatisfiable formula the tableaux construction is indeed capable of constructing a closed tableaux. Hintikka-sets are also of great use for proving the refutation completeness of the PROLOG(+T) tableaux procedure as we will see now.

Intuitively a *Hintikka-set* is a set of formulas which is *semantically closed*, that if there is e.g. a formula  $\varphi_1 \wedge \varphi_2$  in the set, then so are  $\varphi_1$  and  $\varphi_2$ . The following definition formalizes this.

#### Definition 5.3.6 (Temporal Hintikka-Set)

Let P be a PROLOG(+T)-program and let S be a set of FOLTL-formulas such that every function- or predicate-symbol occurring in S also occurs in P. Then S is called a *(temporal) Hintikka-set* if for every FOLTL-formula  $\varphi$  the following holds for every  $i \ge 0$ :

1. if  $\varphi \in B_P^{\text{FoLTL}}$ , then  $\varphi \in S$  iff  $\{\psi \mid \psi \equiv \neg \varphi\} \cap S = \emptyset$ ,

2. if  $\varphi = \mathsf{X}^i (\varphi_1 \land \varphi_2)$ , then  $\varphi \in S$  implies  $\mathsf{X}^i \varphi_1 \in S$  and  $\mathsf{X}^i \varphi_2 \in S$ ,

3. if 
$$\varphi = \mathsf{X}^i (\varphi_1 \lor \varphi_2)$$
, then  $\varphi \in S$  implies  $\mathsf{X}^i \varphi_1 \in S$  or  $\mathsf{X}^i \varphi_2 \in S$ ,

- 4. if  $\varphi = \mathsf{X}^i \forall \mathtt{X} \psi(\mathtt{X})$ , then  $\varphi \in S$  implies  $\mathsf{X}^i \psi(t) \in S$  for every  $t \in U_P$ ,
- 5. if  $\varphi = \mathsf{X}^i \exists \mathsf{X} \psi(\mathsf{X})$ , then  $\varphi \in S$  implies  $\mathsf{X}^i \psi(t) \in S$  for some ground term t,
- 6. if  $\varphi = \mathsf{X}^i \mathsf{G} \psi$ , then  $\varphi \in S$  implies  $\mathsf{X}^{i+j} \psi \in S$  for every  $j \ge 0$ ,
- 7. if  $\varphi = \mathsf{X}^i \mathsf{F} \psi$ , then  $\varphi \in S$  implies  $\mathsf{X}^{i+j} \psi \in S$  for some  $j \ge 0$ ,
- 8. if  $\varphi = \neg \mathsf{X}^i \mathsf{G} \psi$ , then  $\varphi \in S$  implies there is  $j \ge 0$  such that  $\neg \mathsf{X}^{i+j} \psi \in S$  or  $\neg \mathsf{X}^{i+j}\mathsf{G}\psi \in S \text{ for every } j \ge 0,$
- 9. if  $\varphi = \neg \mathsf{X}^i \mathsf{F} \psi$ , then  $\varphi \in S$  implies  $\neg \mathsf{X}^{i+j} \psi \in S$  for every  $j \ge 0$ ,

10. if 
$$\varphi = \mathsf{X}^{i}\varphi_{1}\mathsf{U}\varphi_{2}$$
, then  $\varphi \in S$  implies  $\mathsf{X}^{i}\varphi_{2} \in S$  or  $\mathsf{X}^{i}\varphi_{1} \in S$  and  $\mathsf{X}^{i+1}\varphi_{1}\mathsf{U}\varphi_{2} \in S$ ,

- 11. if  $\varphi = \varphi_1 \mathsf{P} \varphi_2$ , then  $\varphi \in S$  implies  $\neg \mathsf{X}^i \varphi_2 \in S$  and  $\mathsf{X}^i \varphi_1 \in S$  or  $\mathsf{X}^{i+1} \varphi_1 \mathsf{P} \varphi_2 \in S$ ,
- 12. if φ = ¬X<sup>i</sup>φ<sub>1</sub>Uφ<sub>2</sub>, then φ ∈ S implies ¬X<sup>i</sup>φ<sub>2</sub> ∈ S and ¬X<sup>i</sup>φ<sub>1</sub> ∈ S or ¬X<sup>i+1</sup>φ<sub>1</sub>Uφ<sub>2</sub> ∈ S and
  13. if φ = ¬X<sup>i</sup>φ<sub>1</sub>Pφ<sub>2</sub>, then φ ∈ S implies X<sup>i</sup>φ<sub>2</sub> ∈ S or ¬X<sup>i</sup>φ<sub>1</sub> ∈ S and ¬X<sup>i+1</sup>φ<sub>1</sub>Pφ<sub>2</sub> ∈ S

If S is a Hintikka–set constructed from symbols occurring in a program P, then we will also say that S is a Hintikka-set with respect to (wrt.) P. The definition of Hintikka sets can easily be adapted to deal with the special form of PROLOG(+T)-rules by requiring  $\varphi := \psi_1, \ldots, \psi_n \in S$  if and only if there is *i* such that  $\psi_i \notin S$  or  $\varphi \in S$ .

The above definition is suitable for dealing with finite paths of tableaux nodes (as we will see soon). However since paths might also be of infinite length, we need some more definitions. In particular we need a concept of *maximality* of infinite paths. As before the formalisms used in the sequel closely follow [20].

# Definition 5.3.7

Let  $n = (\mathcal{F}(n), F, S)$  be a tableaux node, let  $\varphi \in \mathcal{F}(n)$  be a formula and let **R** be a tableaux rule which can be applied to  $\varphi$ . Then  $\mathbf{R}(\varphi)$  denotes the set of new formulas created by the application of **R**.

Having introduced the operator  $\mathbf{R}$  we can characterize paths as in tableaux to be maximal whether they are of finite length or of infinite length.

## Definition 5.3.8

Let  $\pi$  be an open path. Then  $\pi$  is called *maximal* if

- $\pi$  is finite and no more rules can be applied to  $\mathcal{F}(\pi)$  or
- $\pi = (n_0, n_1, \dots, n_i, \dots)$  is infinite and for every  $i \ge 0$ , every  $\varphi \in \mathcal{F}(n_i)$  and every rule **R** which can be applied to  $\varphi$  there is a node  $n_j$  (j > i) such that  $\mathbf{R}(\varphi) \subseteq \mathcal{F}(n_j)$ .

Maximality of an open path will now turn out to be the key concept for proving refutation completeness.

#### Lemma 5.3.3

Let  $\pi$  be a maximal open path of tableaux nodes constructed from symbols occurring in a program P. Then  $\mathcal{F}(\pi)$  is a (temporal) Hintikka–set wrt P.

**Proof**. First assume that  $\pi$  is finite. Since  $\pi$  is open and maximal, no rules can be applied to the formulas from  $\mathcal{F}(\pi)$ . In particular, no element of  $\mathcal{F}(\pi)$  contains one of the operators G, F, U or P. Henceforth if we assume that  $\mathcal{F}(\pi)$  is not a Hintikka–set, one of the following cases has to occur:

**Case 1** There is a literal  $\varphi \in \mathcal{F}(\pi)$  such that  $\psi \in \mathcal{F}(\pi)$  for some  $\psi$  with  $\psi \equiv \neg \psi$ . But then  $\pi$  is closed which contradicts the assumptions on  $\pi$ .

- **Case 2** There is  $\varphi = \varphi_1 \land \varphi_2 \in \mathcal{F}(\pi)$  such that  $\varphi_1 \notin \mathcal{F}(\pi)$  or  $\varphi_2 \notin \mathcal{F}(\pi)$ . But in this case the rule  $(\land \mathbf{1})$  can be applied to  $\varphi$  which contradicts the maximality of  $\pi$ .
- **Case 3** There is  $\varphi = \varphi_1 \lor \varphi_2 \in \mathcal{F}(\pi)$  such that  $\varphi_1 \notin \mathcal{F}(\pi)$  and  $\varphi_2 \notin \mathcal{F}(\pi)$ . In this case  $(\lor \mathbf{1})$  can be applied which again contradicts the maximality of  $\pi$ .
- **Case 4** There is  $\varphi = \forall \mathbf{X} \psi(\mathbf{X}) \in \mathcal{F}(\pi)$  and  $t \in U_P$  such that  $\psi(t) \notin \mathcal{F}(\pi)$ . Then  $(\sigma \mathbf{I_1})$  can be applied to  $\varphi$  and therefore  $\pi$  is not maximal.
- **Case 5** There is  $\varphi = \exists \mathbf{X} \psi(\mathbf{X}) \in \mathcal{F}(\pi)$  and  $\psi(t) \notin \mathcal{F}(\pi)$  for every t. Then  $(\sigma \mathbf{I_2})$  can be applied to  $\varphi$  and therefore  $\pi$  is not maximal.

So in the case of a finite path  $\pi$  the claim is true. Now assume that  $\pi$  is of infinite length. Then we have more cases to distinguish. The first five cases are identical to the cases from above. So we will only have to consider the cases in which the operators G, F, U and P are involved. Assume that the assumptions from above are fulfilled, that is assume that  $\pi$  is maximal and open and assume that  $\mathcal{F}(\pi)$  is not a Hintikka–set, i.e. assume that the conclusion of the implications from the definition of temporal Hintikka–sets are violated.

**Case 6** There is  $X^i G \psi \in \mathcal{F}(\pi)$  such that  $X^{i+1} \psi \notin \mathcal{F}(\pi)$  for some j. Then let  $j_0$  be the minimal such j, that is  $j_0 = \min \{j \mid X^{i+j} \psi \notin \mathcal{F}(\pi)\}$ . Since  $j_0$  is minimal, we have  $X^{j_0-1}\psi \in \mathcal{F}(\pi)$  and  $X^{i+j_0-2}G\psi \in \mathcal{F}(\pi)$ . This implies  $X^{i+j_0-1}G\psi \in \mathcal{F}(\pi)$  since  $\pi$  is maximal (otherwise (G1) could be applied). Again maximality now yields  $X^{i+j_0}G\psi \in \mathcal{F}(\pi)$  and therefore  $X^{i+j_0}\psi \in \mathcal{F}(\pi)$  which is a contradiction.

**Case 7** Let  $X^i F \psi$  be in  $\mathcal{F}(\pi)$ .

a) Assume that  $X^{i+j}\psi \notin \mathcal{F}(\pi)$  for every  $j \ge 0$ . Since  $\pi$  is maximal, every possible application of the rule  $(\lor \mathbf{1})$  has been carried out. Furthermore the path  $\pi$  corresponds to the path which contains the *right one* of the new formulas

created by  $(\vee \mathbf{1})$  (since instead there would be a minimal value  $j_0$  such that  $\mathsf{X}^{i+j_0}\psi \in \mathcal{F}(\pi)$ ). This yields  $\mathsf{X}^{i+j}\mathsf{F}\psi \in \mathcal{F}(\pi)$  for every  $j \geq 0$ .

- b) Now assume that there is  $j \ge 0$  such that  $X^{i+j}F\psi \notin \mathcal{F}(\pi)$ . As in case 6 we chose the minimal value of all these j's, namely  $j_0 = \min \{j \mid X^{i+j}F\psi \notin \mathcal{F}(\pi)\}$ . So  $X^{i+j_0}F\psi \notin \mathcal{F}(\pi)$  and  $X^{i+j_0-1}F\psi \in \mathcal{F}(\pi)$ . By maximality of  $\pi$  the application of  $(\vee \mathbf{1})$  has been carried out and with  $X^{i+j_0}F\psi \notin \mathcal{F}(\pi)$  we have  $X^{i+j_0-1}\psi \in \mathcal{F}(\pi)$ .
- **Case 8** Let  $\neg X^i G \psi$  be in  $\mathcal{F}(\pi)$ . We have the following cases:
  - a) for each  $j \ge 0$  it holds that  $\neg X^{i+j} \psi \notin \mathcal{F}(\pi)$ . Then in particular we have  $\neg X^i \psi \notin \mathcal{F}(\pi)$ . But since  $\pi$  is maximal we have  $\neg X^{i+1} \mathsf{G} \psi \in \mathcal{F}(\pi)$  since otherwise (**G2**) could be applied and  $\pi$  is not maximal.
  - b) there is  $j \ge 0$  such that  $\neg X^{i+j} G \psi \notin \mathcal{F}(\pi)$ . As before we chose the minimal value of all these j's:  $j_0 = \min \{j \mid \neg X^{i+j} G \psi \notin \mathcal{F}(\pi)\}$ . Then  $\neg X^{i+j_0} G \psi \notin \mathcal{F}(\pi)$ . Furthermore since  $\neg X^i G \psi \in \mathcal{F}(\pi)$  we have  $j_0 \ge 1$ . Since  $j_0$  is minimal we have  $\neg X^{i+j_0-1} G \psi \in \mathcal{F}(\pi)$  and the maximality of  $\pi$  yields  $\neg X^{i+j_0-1} \psi \in \mathcal{F}(\pi)$ .
- **Case 9** Let  $\neg X^i F \psi$  be in  $\mathcal{F}(\pi)$  and assume that there is  $j \ge 0$  such that  $X^{i+j} \psi \in \mathcal{F}(\pi)$ . Then we immediately have that there is a node *n* occurring on  $\pi$  which is closed. This contradicts the assumption that  $\pi$  is open.
- **Case 10** Let  $X^i \psi_1 \cup \psi_2$  be in  $\mathcal{F}(\pi)$ . If  $X^i \psi_2 \in \mathcal{F}(\pi)$  then the case is clear. Now assume that  $X^i \psi_2 \notin \mathcal{F}(\pi)$ . Since  $\pi$  is maximal we have that the rule (U1) which is applicable has indeed been applied and creates the formula  $\varphi' = X^i \psi_1 \wedge X^{i+1} \psi_1 \cup \psi_2$  to which ( $\wedge$ 1) can be applied. By maximality of  $\pi$  we have  $\{X^i \psi_1, X^{i+1} \psi_1 \cup \psi_2\} \subseteq \mathcal{F}(\pi)$ .
- **Case 11** Let  $X^i \psi_1 P \psi_2$  be in  $\mathcal{F}(\pi)$ . Then we immediately have  $\neg X^i \psi_2 \in \mathcal{F}(\pi)$  by maximality of  $\pi$ . Now if  $X^i \psi_1 \in \mathcal{F}(\pi)$ , the case is clear. So assume that  $X^i \psi_1 \notin \mathcal{F}(\pi)$ .

Then we have  $X^{i+1}\psi_1 P \psi_2 \in \mathcal{F}(\pi)$  since otherwise  $\pi$  is not maximal.

- **Case 12** Let  $\neg X^i \psi_1 \cup \psi_2$  be in  $\mathcal{F}(\pi)$ . By maximality of  $\pi$  we have  $\{\neg X^i \psi_2, \neg X^i \psi_1 \lor \neg X^{i+1} \psi_1 \cup \psi_2\} \subseteq \mathcal{F}(\pi)$  and therefore the application of  $(\lor 1)$  yields the desired result.
- **Case 13** Let  $\neg X^i \psi_1 P \psi_2$  be in  $\mathcal{F}(\pi)$ . By application of (P2) we have the following possibilities:
  - a)  $X^i \psi_2 \in \mathcal{F}(\pi)$ . Then the case is clear.
  - b)  $X^i\psi_2 \notin \mathcal{F}(\pi)$ . Then (P2) creates the formula  $\neg X^i\psi_1 \land \neg X^{i+1}\psi_1\mathsf{P}\psi_2$  and an application of  $(\land \mathbf{1})$  yields  $\{\neg X^i\psi_1, \neg X^{i+1}\psi_1\mathsf{P}\psi_2\} \subseteq \mathcal{F}(\pi)$ .

Therefore  $\mathcal{F}(\pi)$  is a temporal Hintikka–set wrt. P and the lemma is proved.

As in first order logic, we are able to construct models for certain kinds of temporal Hintikka–sets.

#### Lemma 5.3.4

Let S be a temporal Hintikka-set such that  $S = \mathcal{F}(\pi)$  for a maximal open path. Then S is satisfiable.

**Proof.** The claim is immediately by considering any interpretation which satisfies every ground *atom* from  $\mathcal{F}(\pi)$ .

Combining these two lemmas we have the following theorem.

# Theorem 5.3.1 (Refutation-Completeness)

Let P be a PROLOG(+T)-program and let  $G = \psi_1 \wedge \cdots \wedge \psi_n$ . be a goal. If  $P \models \psi_1 \wedge \cdots \wedge \psi_n$ , then the tableaux rooted with  $(\{\varphi_{P,G}\}, \emptyset, \emptyset)$  is closed.

**Proof.** We have  $P \models \psi_1 \land \cdots \land \psi_n$  iff  $P \cup \{G\} \models \Box$  iff  $\varphi_{P,G}$  is unsatisfiable. Now assume that the tableaux rooted with  $(\{\varphi_{P,G}\}, \emptyset, \emptyset)$  is not closed. Then there is a maximal open path  $\pi$  in this tableaux. We then have that  $\mathcal{F}(\pi)$  is a temporal Hintikka–set wrt. P and

This result states the most desirable property of PROLOG(+T) and its inference mechanism. We have therefore shown that PROLOG(+T) is indeed an adequate programming language for the fragment of first order temporal logic under consideration. So we can proceed by treating the lattice properties of PROLOG(+T)-objects in order to justify our treatment of the refinement operations in the following chapters.

# The Lattice Structure of Prolog(+T) objects

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6.3.2. Least Generalizations

We will now show how the concept of *subsumption* can be generalized from first order formulas to PROLOG(+T)-objects. The main part of this generalization will be the integration of the temporal operators X, G, F, U and P, so the complicated part is section 6.2 where it will be shown that the lattice properties of the subsumption ordering carry over from first order logic literals to FOLTL-literals. In contrast the results from section 6.3 will be nearly identical to results from first order ILP.

# 6.1. Subsumption

During the construction of programs which satisfy a specification given by sets  $\mathcal{E}^+ \subseteq B_P^{\text{FoLTL}}$  and  $\mathcal{E}^- \subseteq B_P^{\text{FoLTL}}$  it might be necessary to specialize and/or generalize certain objects. In general, specialization and generalization should be related to the process of *largening* and *shrinking* the set of logical consequences (that is modifying a program P in order to yield a program P' which implies *more* – in the case of generalization – respectively *less* – in the case of specialization – than the original program P). However, the logical consequence relation  $\models$  is undecidable and therefore one needs another ordering which is on the one hand decidable (that is it can be implemented on a computer) and on the other side closely related to logical consequence. Subsumption has turned out useful for this purpose (see [77] for a discussion of the difference between subsumption and implication in First Order Logic).

Informally, subsumption models the assumption that some object is *more general* than another one in the way that the more general object implies *more* than the less general  $one^{1}$ . Formally, subsumption between literals is defined as follows.

# Definition 6.1.1 (Subsumption for atoms, Plotkin [133])

Let  $\varphi_1$  and  $\varphi_2$  be literals from FOLTL. Then  $\varphi_1 \succcurlyeq \varphi_2$  if and only if there is a substitution  $\theta$  such that  $\theta(\varphi_1) = \varphi_2$ .

As one might expect, we will write  $\varphi_1 \preccurlyeq \varphi_2$  if  $\varphi_2 \succcurlyeq \varphi_1$ ,  $\varphi_1 \succ \varphi_2$  if  $\varphi_1 \succcurlyeq \varphi_2$  and not  $\varphi_2 \succcurlyeq \varphi_1, \varphi_1 \prec \varphi_2$  if  $\varphi_2 \succ \varphi_1$  and  $\varphi_1 \approx \varphi_2$  if  $\varphi_1 \succcurlyeq \varphi_2$  and  $\varphi_2 \succcurlyeq \varphi_1$ . Additionally we will write  $\varphi_1 \nvDash \varphi_2$  denoting that  $\varphi_1 \succcurlyeq \varphi_2$  does not hold.

It is easily seen that  $\succeq$  is reflexive and transitive. But it is not a partial ordering on the set of *all* FOLTL-literals since it is not anti-symmetric. Consider the literals  $\varphi_1 = p(x)$ 

<sup>&</sup>lt;sup>1</sup>This is due to the fact that the less general object can be constructed by instantiating the more general one.

and  $\varphi_2 = p(y)$ . Then for  $\theta_1 = \left\{\frac{x}{y}\right\}$  and  $\theta_2 = \left\{\frac{y}{x}\right\}$  we have  $\theta_1(\varphi_1) = \varphi_2$  and  $\theta_2(\varphi_2) = \varphi_1$ and therefore  $\varphi_1 \succcurlyeq \varphi_2$  and  $\varphi_2 \succcurlyeq \varphi_1$  but  $\varphi_1 \neq \varphi_2$ . However,  $\varphi_1$  and  $\varphi_2$  are variants.

We will use the ordering  $\succeq$  in order to define a *quasi-order*  $\succeq_s$  on literals which yields a lattice-structure. Therefore we define

- false  $\succeq_s \varphi$  for every FOLTL-literal  $\varphi$ ,
- $\varphi \succeq_s$  true for every FoLTL-literal  $\varphi$  and
- $\varphi_1 \succeq_s \varphi_2$  for  $\varphi_1, \varphi_2 \notin \{\texttt{true}, \texttt{false}\}$  if and only if  $\varphi_1 \succcurlyeq \varphi_2$ .

The notations  $\leq_s$ ,  $\succ_s$ ,  $\prec_s$  and  $\approx_s$  are defined as expected.

In the following section we will see how the ordering  $\succeq_s$  yields a lattice structure in the set of all FOLTL-literals thus extending a well-known result from first order logic (see [133] and [134]).

# 6.2. The Lattice Structure of Literals

# 6.2.1. Generalizations of Terms

In order to present operations for computing least generalizations and greatest specializations of literals, we have to review some operations operating on terms. Recall that a *unifier* for two literals  $\varphi_1$  and  $\varphi_2$  is a substitution  $\theta$  such that  $\theta(\varphi_1) = \theta(\varphi_2)$ . The process of *unification* can be reversed by constructing from  $\varphi_1$  and  $\varphi_2$  both a literal  $\varphi$ and substitutions  $\theta_1$  and  $\theta_2$  such that  $\theta_1(\varphi) = \varphi_1$  and  $\theta_2(\varphi) = \varphi_2$ . Figure 6.1 illustrates the situation.

In [133] it has been shown that it is always possible to construct least generalizations and greatest specializations of terms. We will review the algorithm for constructing least generalizations here since we will need it later when computing least generalizations of literals.

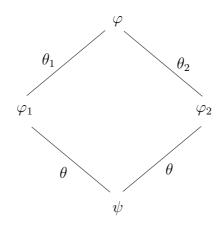


Figure 6.1.: Unification vs. Anti-unification

# Algorithm 2 Anti–unification for terms

**Input**: terms  $t_1$  and  $t_2$ **Output**: term  $\overline{t}$  such that  $\overline{t}$  is a least generalization of  $t_1$  and  $t_2$ 3:  $i \leftarrow 0$ **Require:**  $\{z_i \mid i \in \mathbb{N}\}$  is a set of variables not occurring in  $t_1$  and  $t_2$ 4: if  $t'_1 = t'_2$  then return  $\bar{t} = t'_1$ 5:6: **else** 7:  $p \leftarrow \text{leftmost position at which } t'_1 \text{ and } t'_2 \text{ differ}$  $s \leftarrow t'_1|_p$ 8: 9:  $t \leftarrow t'_2|_p$ if there is  $j \in \{1, \ldots, i\}$  such that  $\theta_1(z_j) = s$  and  $\theta_2(z_j) = t$  then 10:replace  $t'_1|_p$  by  $z_j$ 11:replace  $t'_2|_p$  by  $z_j$ 12:else 13: $i \leftarrow i + 1$ 14:replace  $t'_1|_p$  by  $z_i$ 15:replace  $t'_2|_p$  by  $z_i$ 16: $\begin{aligned} \theta_1 &= \theta_1 \circ \left\{ \frac{z_i}{s} \right\} \\ \theta_2 &= \theta_2 \circ \left\{ \frac{z_i}{t} \right\} \end{aligned}$ 17:18:goto 4 19:end if 20:21: end if

Algorithm 2 indeed produces a least generalization of the two input terms. This result can for example be found in [133] or [126]. Since this least generalization is a generalization with respect to subsumption we will denote the term returned by Algorithm 2 as

$$LGS(t_1, t_2).$$

In later sections we will also denote any least generalization of literals  $\varphi_1$  and  $\varphi_2$  as  $LGS(\varphi_1, \varphi_2)$  and least generalizations of clauses  $C_1$  and  $C_2$  as  $LGS(C_1, C_2)$  since this will not cause any confusion.

# 6.2.2. Generalizations and Specializations of Literals

We will now apply the results from the last section in order to prove that the set of literals from FOLTL is a lattice ordered by the subsumption ordering  $\succeq_s$ . For formal reasons the set of all FOLTL-literals will from now on be denoted as  $\mathcal{L}^{\text{FOLTL}}$ . The proof will not be that difficult but rather long due to the different cases which have to be distinguished.

# Theorem 6.2.1

 $(\mathcal{L}^{\text{FoLtl}}, \succeq_s)$  is a lattice.

**Proof**. In order to prove the theorem we will show that both a least generalization and a greatest specialization of two given literals  $\varphi_1$  and  $\varphi_2$  exists in  $\mathcal{L}^{\text{FoLTL}}$ .

**Specialization** Let  $\varphi_1, \varphi_2 \in \mathcal{L}^{\text{FoLTL}}$  be given. Assume without loss of generality that  $\varphi_1$  and  $\varphi_2$  have no variables in common, that is  $\varphi_1$  and  $\varphi_2$  are standardized apart.

1. if  $\varphi_1 =$ true or  $\varphi_2 =$ true, then

$$GSS(\varphi_1, \varphi_2) = \texttt{true},$$

2. a) if  $\varphi_1 = \texttt{false}$ , then

 $GSS(\varphi_1, \varphi_2) = \varphi_2$  and

b) if  $\varphi_2 = \texttt{false}$ , then

 $\mathrm{GSS}(\varphi_1,\varphi_2)=\varphi_1,$ 

3. if  $\varphi_1$  and  $\varphi_2$  are unifiable with  $\sigma = \text{mgu}(\varphi_1, \varphi_2)$ , then

$$GSS(\varphi_1, \varphi_2) = \sigma(\varphi_1)$$
 and

4. if  $\varphi_1$  and  $\varphi_2$  are *not* unifiable, then

$$GSS(\varphi_1,\varphi_2) =$$
true.

It is obvious that  $GSS(\varphi_1, \varphi_2)$  is always a specialization of the original literals  $\varphi_1$ and  $\varphi_2$ . It remains to show that it is indeed a greatest specialization. The only two nontrivial cases are that both  $\varphi_1 \notin \{\texttt{true}, \texttt{false}\}$  and  $\varphi_2 \notin \{\texttt{true}, \texttt{false}\}$ . Consider the two cases:

**Case 1**  $\varphi_1$  and  $\varphi_2$  are not unifiable. Then there cannot exist any literal  $\psi$  such that  $\psi \neq \texttt{true}$  and  $\varphi_1 \succeq_s \psi$  and  $\varphi_2 \succeq_s \psi$  since then there would be substitutions  $\theta_1$  and  $\theta_2$  such that  $\theta_1(\varphi_1) = \psi$  and  $\theta_2(\varphi_2) = \psi$ . Since  $\varphi_1$  and  $\varphi_2$  are standardized apart this would give

$$\theta_1(\varphi_1) = (\theta_1 \circ \theta_2)(\varphi_1) = \psi$$
  
$$\theta_2(\varphi_2) = (\theta_1 \circ \theta_2)(\varphi_2) = \psi$$

So  $\theta_1 \circ \theta_2$  is a unifier for  $\varphi_1$  and  $\varphi_2$  which is a contradiction to the assumption that  $\varphi_1$  and  $\varphi_2$  are not unifiable. So the claim is proved.

**Case 2**  $\varphi_1$  and  $\varphi_2$  are unifiable. Then there is  $\sigma = \text{mgu}(\varphi_1, \varphi_2)$ . if  $\sigma(\varphi_1)$  is not a greatest specialization, then there is  $\psi$  and substitutions  $\theta_1, \theta_2$  such that

 $\theta_1(\varphi_1) = \psi$ ,  $\theta_2(\varphi_2) = \psi$  and  $\psi \succ_s \sigma(\varphi_1)$ . But then there would be  $\gamma \neq \varepsilon$  such that  $\gamma(\psi) = \sigma(\varphi_1)$  and therefore  $\sigma$  would not be most general, which is a contradiction. So the claim is proved.

**Generalization** Let  $\varphi_1, \varphi_2 \in \mathcal{L}^{\text{FoLtL}}$  be given.

- 1. if  $\varphi_1$  is positive and  $\varphi_2$  is negative or if  $\varphi_1$  is negative and  $\varphi_2$  is positive, then  $LGS(\varphi_1, \varphi_2) := true.$
- 2. if  $\varphi_1 = \texttt{false}$  or  $\varphi_2 = \texttt{false}$ , then  $LGS(\varphi_1, \varphi_2) = \texttt{false}$ . Similarly if  $\varphi_1 = \texttt{true}$ , then  $LGS(\varphi_1, \varphi_2) = \varphi_2$  and if  $\varphi_2 = \texttt{true}$ , then  $LGS(\varphi_1, \varphi_2) = \varphi_1$ .
- 3. if both  $\varphi_1$  and  $\varphi_2$  are negative, then assume that

$$\varphi_1 = \neg \psi_1$$
 and  $\varphi_2 = \neg \psi_2$ 

and define

$$LGS(\varphi_1, \varphi_2) = \neg LGS(\psi_1, \psi_2)$$

- 4. if both  $\varphi_1$  and  $\varphi_2$  are positive, then
  - a) if  $\varphi_1 = p(t_1, \dots, t_n)$  and  $\varphi_2 = p(t'_1, \dots, t'_n)$  for some p with  $\alpha(p) = n$  and terms  $t_1, \dots, t_n, t'_1, \dots, t'_n$ , then

$$LGS(\varphi_1, \varphi_2) = p(LGS(t_1, t'_1), \dots, LGS(t_n, t'_n)),$$

b) if  $\varphi_1 = p(t_1, \dots, t_n)$  and  $\varphi_2 = q(t'_1, \dots, t'_m)$  for p, q with  $\alpha(p) = n, \alpha(q) = m$  and  $t_1, \dots, t_n, t'_1, \dots, t'_m$  such that  $p \neq q$ , then

$$LGS(\varphi_1, \varphi_2) = false,$$

c) if  $\varphi_1 = \mathsf{X}\psi_1$  and  $\varphi_2 = \mathsf{X}\psi_2$ , then

$$LGS(\varphi_1, \varphi_2) = XLGS(\psi_1, \psi_2),$$

d) if 
$$\varphi_1 = \mathsf{G}\psi_1$$
 and  $\varphi_2 = \mathsf{G}\psi_2$ , then

$$LGS(\varphi_1, \varphi_2) = GLGS(\psi_1, \psi_2),$$

e) if  $\varphi_1 = \mathsf{F}\psi_1$  and  $\varphi_2 = \mathsf{F}\psi_2$ , then

$$LGS(\varphi_1, \varphi_2) = FLGS(\psi_1, \psi_2),$$

f) if 
$$\varphi_1 = \psi_1^{(1)} \mathsf{U} \psi_2^{(1)}$$
 and  $\varphi_2 = \psi_1^{(2)} \mathsf{U} \psi_2^{(2)}$ , then

$$LGS(\varphi_1, \varphi_2) = LGS(\psi_1^{(1)}, \psi_1^{(2)}) ULGS(\psi_2^{(1)}, \psi_2^{(2)}),$$

g) if 
$$\varphi_1 = \psi_1^{(1)} \mathsf{P} \psi_2^{(1)}$$
 and  $\varphi_2 = \psi_1^{(2)} \mathsf{P} \psi_2^{(2)}$ , then

LGS
$$(\varphi_1, \varphi_2) = LGS(\psi_1^{(1)}, \psi_1^{(2)}) PLGS(\psi_2^{(1)}, \psi_2^{(2)})$$
 and

h) in all other cases:

$$LGS(\varphi_1, \varphi_2) = false.$$

Again we will distinguish several cases.

- Case 1 Case 1 from the above list occurs. Then the claim is trivial.
- **Case 2** Case 2 occurs. Then the claim is due to the definition of  $\succeq_s$ .
- **Case 3** We now proceed by induction on the structure of the literals. First assume that both  $\varphi_1$  and  $\varphi_2$  are nontemporal. Then the claim is due to results from [126] regarding the lattice structure of first order logic literals. Similarly we can

prove the case in which both literals are negative by exploiting the assumption that the algorithm is correct for the subliterals under consideration. Now assume that  $\varphi_1$  and  $\varphi_2$  are of the form described in points c), d) and e). Then we can exploit the induction hypothesis for the literals  $\psi_1$  and  $\psi_2$  and the claim is immediate. Similarly we can treat the points f) and g). Finally in case h) no nontrivial least generalization can exist.

So the theorem is proved.

The techniques from the proof of the above theorem are summarized in Algorithms 3 and 4.

Algorithm 3 Greatest Specialization of FOLTL-literals

Input: literals  $\varphi_1, \varphi_2$ Output:  $GSS(\varphi_1, \varphi_2)$ 1: if  $\varphi_1$  and  $\varphi_2$  are unifiable then 2:  $\sigma \leftarrow mgu(\varphi_1, \varphi_2)$ 3: return  $\sigma(\varphi_1)$ 4: else 5: return true 6: end if

#### Example 6.2.1

- 1. Consider  $\varphi_1 = \mathsf{GF}p(\mathbf{X}, f(a))$  and  $\varphi_2 = \mathsf{GF}q(a)$ . Then  $\mathrm{LGS}(\varphi_1, \varphi_2) = \texttt{false}$  and  $\mathrm{GSS}(\varphi_1, \varphi_2) = \texttt{true}$ .
- 2. Now consider  $\varphi_1 = p(\mathbf{X}, \mathbf{X})$  and p(f(a), b). Here we have  $LGS(\varphi_1, \varphi_2) = p(\mathbf{Z}_1, \mathbf{Z}_2)$ and  $GSS(\varphi_1, \varphi_2) = true$ .
- 3. Finally consider  $\varphi_1 = \mathsf{GX}p(A, X)$  and  $\varphi_2 = \mathsf{GX}p(Y, b)$ . Then  $\mathrm{LGS}(\varphi_1) = \mathsf{GX}p(Z_1, Z_2)$ and  $\mathrm{GSS}(\varphi_1, \varphi_2) = \mathsf{GX}p(a, b)$ .

Algorithm 4 Least Generalization of FOLTL-literals

```
Input: literals \varphi_1, \varphi_2
Output: LGS(\varphi_1, \varphi_2)
  1: if \varphi_1 = p(t_1, \ldots, t_n) and \varphi_2 = p(t'_1, \ldots, t'_n) for some p and t_1, \ldots, t_n, t'_1, \ldots, t'_n then
           return p(LGS(t_1, t'_1), \dots, LGS(t_n, t'_n))
  2:
  3: end if
  4: if \varphi_1 = \neg \psi_1 and \varphi_2 = \neg \psi_2 then
           return \neg LGS(\psi_1, \psi_2)
  5:
  6: end if
  7: if \varphi_1 = X\psi_1 and \varphi_2 = X\psi_2 then
           return XLGS(\psi_1, \psi_2)
  8:
  9: end if
10: if \varphi_1 = \mathsf{G}\psi_1 and \varphi_2 = \mathsf{G}\psi_2 then
11:
           return GLGS(\psi_1, \psi_2)
12: end if
13: if \varphi_1 = \mathsf{F}\psi_1 and \varphi_2 = \mathsf{F}\psi_2 then
           return FLGS(\psi_1, \psi_2)
14:
15: end if
15: end if

16: if \varphi_1 = \psi_1^{(1)} \cup \psi_2^{(1)} and \varphi_2 = \psi_1^{(2)} \cup \psi_2^{(2)} then

17: return LGS(\psi_1^{(1)}, \psi_1^{(2)}) \cup LGS(\psi_2^{(1)}, \psi_2^{(2)})
18: end if

19: if \varphi_1 = \psi_1^{(1)} \mathsf{P} \psi_2^{(1)} and \varphi_2 = \psi_1^{(2)} \mathsf{P} \psi_2^{(2)} then

20: return \mathrm{LGS}(\psi_1^{(1)}, \psi_1^{(2)}) \mathsf{PLGS}(\psi_2^{(1)}, \psi_2^{(2)})
21: end if
22: return false
```

# 6.3. The Lattice Structure of Rules

Algorithms 3 and 4 from the last section can now be used in order to compute least generalizations and greatest specializations of clauses, that is least generalizations and greatest specializations of PROLOG(+T)-rules. For this purpose we will adapt a proof for the existence of least generalizations and greatest specializations of first order logic clauses which might for example be found in [126]. The subsumption ordering for rules is defined as follows.

# Definition 6.3.1 (Subsumption for Rules, Plotkin [133])

Let  $C_1$  and  $C_2$  be PROLOG(+T)-rules (represented as sets of literals). Then  $C_1 \succeq_s C_2$ if and only if there is a substitution  $\theta$  such that  $\theta(C_1) \subseteq C_2$ .

The symbols  $\succ_s, \preceq_s$  and  $\prec_s$  are then defined as usual.

# 6.3.1. Greatest Specializations

As in the last section, the simpler part is the computation of greatest specializations of clauses. Therefore we will adapt a technique described in [126]. We will see that the greatest specialization of two PROLOG(+T)-rules is in general not unique. This is due to the fact that we allow negated atoms in the tails of rules.

Assume that

$$C_1 = \varphi_1 :- \psi_1^{(1)}, \dots, \psi_{n_1}^{(1)}$$
 and  
 $C_2 = \varphi_2 :- \psi_1^{(2)}, \dots, \psi_{n_2}^{(2)}$ 

are given. We identify  $C_1$  and  $C_2$  with the sets of literals involved in these rules, that is

we concentrate on

$$S_{C_1} = \left\{ \neg \psi_1^{(1)}, \dots, \neg \psi_{n_1}^{(1)}, \varphi_1 \right\} \text{ and}$$
  
$$S_{C_2} = \left\{ \neg \psi_1^{(2)}, \dots, \neg \psi_{n_2}^{(2)}, \varphi_2 \right\}$$

as the objects of interest. We now build the set  $C = S_{C_1} \cup S_{C_2}$  and construct a set of PROLOG(+T)-rules from this set each of which is a greatest specialization of  $C_1$  and  $C_2$ . But in order to do this we define a helper function TAIL which will be useful in the definition.

Let  $T = \{\gamma_1, \ldots, \gamma_k\}$  be a set of FOLTL-literals. Then the string TAIL(T) is defined as

$$TAIL(T) = \neg \gamma_1, \ldots, \neg \gamma_k,$$

where  $\neg \neg \gamma_i$  will be identified with  $\gamma_i$  for  $i = 1, \ldots, k$ .

Now assume that  $S = \{l_1, \ldots, l_{n_1+n_2+2}\}$ . We then build the set L consisting of all PROLOG(+T)-rules which can be constructed using the following scheme: if  $l_i$  is positive, then L contains the rule

$$l_i :- \operatorname{TAIL}(S \setminus \{l_i\})$$

Algorithm 5 illustrates this scheme.

The properties of Algorithm 5 are summarized in the following theorem.

#### Theorem 6.3.1

Let  $C_1$  and  $C_2$  be PROLOG(+T)-rules. Then Algorithm 5 computes a set of rules each of which is a greatest specialization of  $C_1$  and  $C_2$ .

**Proof.** Let  $C_1$  and  $C_2$  be given. Without loss of generality we can assume that  $C_1$ and  $C_2$  are standardized apart. Let C be any rule computed by Algorithm 5. Then for  $S = S_{C_1} \cup S_{C_2}$  we have  $S_{C_1} \subseteq S$  and  $S_{C_2} \subseteq S$  so both  $C_1 \succeq_s C$  and  $C_2 \succeq_s C$  holds, that is C is a specialization under subsumption of  $C_1$  and  $C_2$ . Now consider any rule Algorithm 5 Greatest Specialization of PROLOG(+T)-rules **Input**: PROLOG(+T)-rules  $C_1, C_2$ **Output**: set of greatest specializations of  $C_1, C_2$ **Require:**  $C_1 = \varphi_1 := \psi_1^{(1)}, \dots, \psi_{n_1}^{(1)}, C_2 = \varphi_1 := \psi_1^{(2)}, \dots, \psi_{n_2}^{(2)}$ 1:  $L \leftarrow \emptyset$ 2:  $S_{C_1} \leftarrow \left\{ \neg \psi_1^{(1)}, \dots, \neg \psi_{n_1}^{(1)}, \varphi_1 \right\}$ 3:  $S_{C_2} \leftarrow \left\{ \neg \psi_1^{(2)}, \dots, \neg \psi_{n_2}^{(2)}, \varphi_2 \right\}$ 4:  $S \leftarrow S_{C_1} \cup S_{S_2}$ **Require:**  $S = \{l_1, ..., l_o\}$ 5: for i = 1, ..., o do if  $l_i$  is positive then 6:  $L \leftarrow L \cup \{l_i := \operatorname{TAIL}(S \setminus \{l_i\})\}$ 7: end if 8: 9: end for 10: return L

 $\overline{C}$  such that  $C_1 \succeq_s \overline{C}$  and  $C_2 \succeq_s \overline{C}$ . Then there are substitutions  $\theta_1$  and  $\theta_2$  such that  $\theta_1(C_1) \subseteq \overline{C}$  and  $\theta_2(C_2) \subseteq \overline{C}$  and  $\theta_1$  and  $\theta_2$  only replace variables in  $C_1$  and  $C_2$  (since  $C_1$  and  $C_2$  are standardized apart). Define  $\theta = \theta_1 \cup \theta_2$ . Then

$$\begin{aligned} \theta\left(S_{C}\right) &= \theta\left(S_{C_{1}} \cup S_{C_{2}}\right) \\ &= \theta\left(S_{C_{1}}\right) \cup \theta\left(S_{C_{2}}\right) \\ &= \theta_{1}\left(S_{C_{1}}\right) \cup \theta_{2}\left(S_{C_{2}}\right) \\ &\subseteq S_{\bar{C}}. \end{aligned}$$

So  $C \succeq_s \overline{C}$ . Since C was chosen arbitrary from the set of rules computed by Algorithm 5 the claim is proved.

Although the syntactical form of a greatest specialization of PROLOG(+T)-rules is not uniquely determined we can adapt the notation  $GSS(C_1, C_2)$ . This is justified since for every  $C^{(1)}, C^{(2)}$  computed by Algorithm 5 given the inputs  $C_1$  and  $C_2$  we have  $S_{C^{(1)}} =$   $S_{C^{(2)}} = S_{C_1} \cup S_{C_2}$  and therefore  $C^{(1)} \equiv C^{(2)}$ . Consequently

$$GSS(C_1, C_2)$$

will from now on be used in order to denote any of the rules computed by Algorithm 5.

We will now close this section by presenting a simple example of how Algorithm 5 creates greatest specializations of PROLOG(+T)-rules.

#### Example 6.3.1

Consider the rules

$$C_1 = \mathsf{Gp}(\mathsf{a}) := \mathsf{XXp}(\mathtt{X}_1). \text{ and}$$
$$C_2 = \mathsf{XFp}(\mathtt{X}_2) := \mathsf{not}(\mathtt{r}(\mathtt{X}_2)), \mathsf{Gp}(\mathsf{b}).$$

Then we have

$$S_{C_1} = \{ \operatorname{not}(\mathsf{X}\mathsf{X}\mathsf{p}(\mathtt{X}_1)), \mathsf{Gp}(\mathtt{a}) \} \text{ and}$$
$$S_{C_2} = \{ \operatorname{r}(\mathtt{X}_2), \operatorname{not}(\mathsf{Gp}(\mathtt{b})), \mathsf{X}\mathsf{Fq}(\mathtt{X}_2) \}$$

and therefore

$$S = \{ not(XXp(X)), Gp(a), r(X), not(Gp(b)), XFq(X) \}$$

The rules generated by Algorithm 3 are the following:

1. 
$$Gp(a) := XXp(X_1), not(r(X_2)), not(Gp(b)), not(XFq(X_2)).$$

- 2.  $r(X_2) := XXp(X_1), not(Gp(a)), Gp(b), not(XFq(X_2))$ . and
- 3.  $XFq(X_2) := XXp(X_1), not(Gp(a)), not(r(X_2)), Gp(b).$

#### 6.3.2. Least Generalizations

What remains to be established is the existence of least generalizations of PRoLOG(+T)rules under subsumption. For first order literals the concept of *compatibility* has been used for the proof of the existence of least generalizations of clauses. Two first order literals  $\varphi_1$  and  $\varphi_2$  are considered *compatible* if they are either both positive or both negative and if they start with the same predicate symbol. For FoLTL-literals the situation is slightly more complicated. Intuitively we should consider literals compatible if they are either both positive or both negative and if they contain the same temporal operators in the same order. Formally we will present a technique which constructs for a literal  $\varphi$  a tree TREE( $\varphi$ ) from which a tuple TEMP( $\varphi$ ) of *words* built up from the operators involved in the literal  $\varphi$  can be extracted. Two literals  $\varphi_1$  and  $\varphi_2$  are then considered *compatible* if they yield identical sets TEMP( $\varphi_1$ ) and TEMP( $\varphi_2$ ).

Given a literal  $\varphi$  we will now show how to construct a labeled graph  $\text{TREE}(\varphi)$ . Recall from chapter 5.2 that a labeled graph is a tuple T = (V, E, l) consisting of a finite set Vcontaining the vertices or nodes, a set  $E \subseteq V \times V$  containing the edges and a mapping l. Here l has the form  $l : V \to \{U, P, X, G, F, \neg\} \cup P \cup T$ . We will partition the set Vinto three sets  $V_t$ ,  $V_p$  and  $V_f$  containing so called *temporal nodes*, predicate nodes and function nodes, that is  $V = V_t \cup V_P \cup V_f$ . The construction of  $\text{TREE}(\varphi)$  is now given by induction on the form of  $\varphi$ .

**Case 1**  $\varphi = p(t_1, \ldots, t_n) \in B_P$  is a first order atom. Then we set

TREE
$$(\varphi) = (\{v_0, v_1, \dots, v_n\}, \{(v_0, v_i) \mid i = 1, \dots, n\}, l),$$

where l is defined by

$$l(v_0) = \mathbf{p}$$
 and  
 $l(v_i) = t_i$  for  $i > 0$ 

and

$$V_t = \emptyset,$$
  

$$V_f = \{v_1, \dots, v_n\} \text{ and }$$
  

$$V_p = \{v_0\}.$$

**Case 2**  $\varphi = \operatorname{not}(\psi)$  for some  $\operatorname{PROLOG}(+T)$ -literal  $\psi$ . Then assume that  $\operatorname{TREE}(\psi) = (\bar{V}, \bar{E}, \bar{l})$  is given. If  $\bar{V} = \left\{ \bar{v}_1, \dots, \bar{v}_{|\bar{V}|} \right\} = \bar{V}_t \stackrel{.}{\cup} \bar{V}_p \stackrel{.}{\cup} \bar{V}_f$  then we define

$$\begin{array}{lll} V &=& \left\{ v_0, \bar{v}_1, \dots, \bar{v}_{|\bar{V}|} \right\} \text{ for some } v_0 \not\in \bar{V}, \\ V_t &=& \bar{V}_t, \\ V_f &=& \bar{V}_f, \\ V_p &=& \bar{V}_p \cup \{v_0\}, \\ E &=& \bar{E} \cup \left\{ (v_0, \bar{v}) \mid \bar{v} \in \bar{V} \text{ such that } (\bar{v}, \bar{v}) \not\in \bar{E} \text{ for each } \bar{v} \in \bar{V} \right\}, \\ l(v_0) &=& \operatorname{\textbf{not}} \operatorname{and} \\ l(v) &=& \bar{l}(v) \text{ for each } v \neq v_0. \end{array}$$

**Case 3**  $\varphi = \oplus \psi$  for some PROLOG(+T)-literal  $\psi$  and some  $\oplus \in \{X, G, F\}$ . Then assume that  $\text{TREE}(\psi) = (\bar{V}, \bar{E}, \bar{l})$  is given. If  $\bar{V} = \{\bar{v}_1, \dots, \bar{v}_{|\bar{V}|}\} = \bar{V}_t \stackrel{.}{\cup} \bar{V}_p \stackrel{.}{\cup} \bar{V}_f$ , then define

$$\begin{split} V &= \left\{ v_0, \bar{v}_1, \dots, \bar{v}_{|\bar{V}|} \right\} \text{ for some } v_0 \not\in \bar{V}, \\ V_t &= \bar{V}_t, \\ V_f &= \bar{V}_f, \\ V_p &= \bar{V}_p \cup \{v_0\}, \\ E &= \bar{E} \cup \left\{ (v_0, \bar{v}) \mid \bar{v} \in \bar{V} \text{ such that } (\bar{v}, \bar{v}) \notin \bar{E} \text{ for each } \bar{v} \in \bar{V} \right\}, \end{split}$$

 $l(v_0) = \oplus$  and  $l(v) = \overline{l}(v)$  for each  $v \neq v_0$ .

**Case 4**  $\varphi = \psi_1 \oplus \psi_2$  for PROLOG(+T)-literals  $\psi_1$  and  $\psi_2$  and some  $\oplus \in \{\mathsf{U},\mathsf{P}\}$ . Assume that  $\mathrm{TREE}(\psi_1) = (\bar{V}_1, \bar{E}_1, \bar{l}_1)$  with  $\bar{V}_1 = \bar{V}_{1,t} \stackrel{.}{\cup} \bar{V}_{1,p} \stackrel{.}{\cup} \bar{V}_{1,f}$  and  $\mathrm{TREE}(\psi_2) = (\bar{V}_2, \bar{E}_2, \bar{l}_2)$  with  $\bar{V}_2 = \bar{V}_{2,t} \stackrel{.}{\cup} \bar{V}_{2,p} \stackrel{.}{\cup} \bar{V}_{2,f}$  are given such that  $\bar{V}_1 \cap \bar{V}_2 = \emptyset$ . Choose some new  $v_0 \notin \bar{V}_1 \cup \bar{V}_2$  and set

$$\begin{split} V &= \{v_0\} \cup \bar{V}_1 \cup \bar{V}_2, \\ V_t &= \bar{V}_{1,t} \cup \bar{V}_{2,t} \cup \{v_0\}, \\ V_p &= \bar{V}_{1,p} \cup \bar{V}_{2,p}, \\ V_f &= \bar{V}_{1,f} \cup \bar{V}_{2,f}, \\ E &= \bar{E}_1 \cup \bar{E}_2 \cup \left\{ (v_0, v) \mid \begin{array}{l} v \in \bar{V}_1 \cup \bar{V}_2 \text{ such that } (\bar{v}, v) \notin \bar{E}_1 \cup \bar{E}_1 \\ \text{ for each } \bar{v} \in \bar{V}_1 \cup \bar{V}_2 \end{array} \right\}, \\ l(v_0) &= \oplus, \\ l(v) &= \bar{l}_1(v) \text{ for } v \in \bar{V}_1 \text{ and} \\ l(v) &= \bar{l}_2(v) \text{ for } v \in \bar{V}_2. \end{split}$$

#### Example 6.3.2

- 1. Assume that  $\varphi_1 = \mathbf{q}(\mathbf{a}) \mathsf{U}\mathbf{p}(\mathbf{b})$  and  $\varphi_2 = \mathbf{q}(\mathbf{c}) \mathsf{U}\mathbf{p}(\mathbf{a})$ . The graphs  $\mathsf{TREE}(\varphi_1)$  and  $\mathsf{TREE}(\varphi_2)$  are depicted in Figures 6.2 and 6.3 where the nodes from  $V_f$  are drawn as squares while all other nodes are drawn as circles.
- 2. Now assume that  $\varphi_3 = \mathbf{q}(\mathbf{c})\mathsf{UXp}(\mathbf{a})$ . Then  $\mathrm{TREE}(\varphi_3)$  is as depicted in Figure 6.4.

Now recall that a *path* from a node  $v_1$  to a node  $v_2$  in a (labeled) graph G = (V, E, l) is defined as follows:

1. either  $(v_1, v_2) \in E$  or

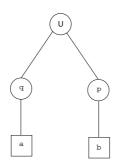


Figure 6.2.: TREE(q(a)Up(b))

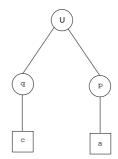


Figure 6.3.: TREE(q(c)Up(a))

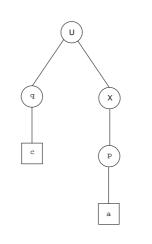


Figure 6.4.: TREE(q(c)UXp(a))

2. there is  $v \in V$  such that  $(v_1, v) \in E$  and there is a path from v to  $v_2$  in G.

Paths can be described very naturally by giving the sequence of nodes on this path. We will from now on describe paths as  $\pi = (v_1, \ldots, v_n)$  and call  $\pi$  a path of *length* n. Having constructed TREE( $\varphi$ ) from the literal  $\varphi$ , we can extract the information TEMP( $\varphi$ ) as follows: assume that the nodes from  $V_f$  are numbered in ascending order from *left to* right. If  $V_f = \{v_1, \ldots, v_k\}$  then TEMP( $\varphi$ ) =  $(s_1, \ldots, s_k)$  where  $s_i$  is defined as follows:

- 1. Let  $\pi = (v_0, \ldots, v_{k_i}, v_i)$  be the uniquely determined path from  $v_0$  to  $v_i$  and
- 2.  $s_i = l(v_0) \circ \cdots \circ l(v_{k_i})$  where  $\circ$  denotes the concatenation of words.

## Example 6.3.3

Again consider the literals from Example 6.3.2. We then have

$$T EMP(\varphi_1) = (Uq, Up),$$
  

$$T EMP(\varphi_2) = (Uq, Up) \text{ and}$$
  

$$T EMP(\varphi_3) = (Uq, UXp).$$

The construction of  $\text{TEMP}(\varphi)$  from a given literal  $\varphi$  allows the extension of the concept of *compatibility* from first order logic.

# Definition 6.3.2

Let  $\varphi_1$  and  $\varphi_2$  be PROLOG(+T)-literals.  $\varphi_1$  and  $\varphi_2$  are called *compatible* if  $\text{TEMP}(\varphi_1) = \text{TEMP}(\varphi_2)$ .

Intuitively  $\varphi_1$  and  $\varphi_2$  are assumed to be compatible if they only differ in their subterms. So compatibility is a criterion for the existence of a nontrivial least generalization of two literals.

#### Example 6.3.4

Again consider the literals  $\varphi_1$ ,  $\varphi_2$  and  $\varphi_3$  from Example 6.3.2. From Example 6.3.3 it is clear that  $\text{TEMP}(\varphi_1) = \text{TEMP}(\varphi_2) = (Uq, Up)$ . So  $\varphi_1$  and  $\varphi_2$  are compatible. On the other hand  $\text{TEMP}(\varphi_3) \neq \text{TEMP}(\varphi_1)$ . So  $\varphi_1$  and  $\varphi_3$  are not compatible.

The concept of compatibility of literals will now be used in order to compute least generalizations of PROLOG(+T)-rules. Essentially the procedure is identical to a procedure which is known from first order ILP. Assume that

$$C_1 = \varphi_1 := -\psi_1^{(1)}, \dots, \psi_{n_1}^{(1)}$$
 and  
 $C_2 = \varphi_2 := -\psi_1^{(2)}, \dots, \psi_{n_2}^{(2)}$ .

are two PROLOG(+T)-rules. We will again work with the sets  $S_{C_1}$  and  $S_{C_2}$  of literals which represent these rules. So assume that

$$S_{C_1} = \left\{ \neg \psi_1^{(1)}, \dots, \neg \psi_{n_1}^{(1)}, \varphi_1 \right\} = \left\{ \chi_0^{(1)}, \chi_1^{(1)}, \dots, \chi_{n_1}^{(1)} \right\} \text{ and}$$
  

$$S_{C_2} = \left\{ \neg \psi_1^{(2)}, \dots, \neg \psi_{n_2}^{(2)}, \varphi_2 \right\} = \left\{ \chi_0^{(2)}, \chi_1^{(2)}, \dots, \chi_{n_2}^{(2)} \right\},$$

where  $\chi_0^{(i)} = \varphi_i$  and  $\chi_j^{(i)} = \neg \psi_j^{(i)}$  for i = 1, 2 and j > 0. Let

$$Sel = \left\{ \left( \chi_{i_0}^{(1)}, \chi_{j_0}^{(2)} \right), \dots, \left( \chi_{i_k}^{(1)}, \chi_{j_k}^{(2)} \right) \right\}$$

be the set of all pairs of compatible literals from  $S_{C_1}$  and  $S_{C_2}$ . If  $Sel = \emptyset$  or if Sel does not contain at least one pair of *positive* literals, then we define

$$LGS(C_1, C_2) = \{ \texttt{false} \}.$$

This is reasonable since

1. if there are no compatible literals, then no rule C and no pair of substitutions

 $\sigma_1, \sigma_2 \neq \varepsilon$  such that  $\sigma(S_C) \subseteq S_{C_1}$  and  $\sigma_2(S_C) \subseteq S_{C_2}$  can exist (assuming that  $S_C \neq \emptyset$ ) and

2. if there is no pair of compatible atoms, then there is no literal which can be generalized in order to yield the *head* of the generalized rule.

We will now concentrate on the case that  $Sel \neq \emptyset$ , that is there is at least one pair of compatible literals. In order to compute the set of least generalizations of  $C_1$  and  $C_2$  we first adapt a technique presented in [133] which transforms literals to terms. Therefore assume that sig =  $(\mathcal{X}, F, P, \alpha)$  is the signature from which  $C_1$  and  $C_2$  are built. We extend this signature to sig<sub>ext</sub> =  $(\mathcal{X}, F_{ext}, P_{ext}, \alpha_{ext})$  as follows:

- $F_{\text{ext}} = F \cup \{ \mathbf{f}_{p_1}, \dots, \mathbf{f}_{p_{|P|}} \} \cup \{ \mathbf{f}_{\text{not}} \} \cup \{ \mathbf{f}_{\text{next}}, \mathbf{f}_{\text{always}}, \mathbf{f}_{\text{finally}}, \mathbf{f}_{\text{until}}, \mathbf{f}_{\text{precedes}} \} \text{ assuming}$ that  $P = \{ \mathbf{p}_1, \dots, \mathbf{p}_{|P|} \},$
- $P_{\text{ext}} = P \cup \{p_{\text{new}}\}$  for some symbol  $p_{\text{new}} \notin P$  and
- •

$$\begin{aligned} \alpha_{\text{ext}}(\mathbf{f}) &= \alpha(\mathbf{f}) \text{ for } \mathbf{f} \in F, \\ \alpha_{\text{ext}}\left(\mathbf{f}_{\mathbf{p}_{i}}\right) &= \alpha(\mathbf{p}_{i}) \text{ for } i = 1, \dots, |P|, \\ \alpha_{\text{ext}}(\mathbf{p}_{\text{new}}) &= k, \\ \alpha_{\text{ext}}(\mathbf{f}_{\text{not}}) &= \alpha_{\text{ext}}(\mathbf{f}_{\text{next}}) = \alpha_{\text{ext}}(\mathbf{f}_{\text{always}}) = \alpha_{\text{ext}}(\mathbf{f}_{\text{finally}}) = 1 \text{ and} \\ \alpha_{\text{ext}}(\mathbf{f}_{\text{until}}) &= \alpha_{\text{ext}}(\mathbf{f}_{\text{precedes}}) = 2. \end{aligned}$$

Using the signature  $\operatorname{sig}_{ext}$  we define an operation  $\operatorname{TERM} : \mathcal{L}_t(\operatorname{sig}) \to \mathcal{T}(\operatorname{sig}_{ext})$  which maps literals to terms. TERM is defined inductively as follows:

- if  $\varphi = \mathbf{p}(t_1, \dots, t_n)$  for some symbol  $\mathbf{p} \in P$  and  $t_1, \dots, t_n \in \mathcal{T}(\text{sig})$ , then  $\text{TERM}(\varphi) = \mathbf{f}_{\mathbf{p}}(t_1, \dots, t_n)$ ,
- if  $\varphi = \operatorname{not}(\psi)$  for some literal  $\psi \in \mathcal{L}_t(\operatorname{sig})$ , then  $\operatorname{TERM}(\varphi) = \mathbf{f}_{\operatorname{not}}(\operatorname{TERM}(\psi))$ ,

• - if 
$$\varphi = X\psi$$
 for some  $\psi \in \mathcal{L}_t(\operatorname{sig})$ , then  $\operatorname{TERM}(\varphi) = \mathbf{f}_{\operatorname{next}}(\operatorname{TERM}(\psi))$ ,  
- if  $\varphi = \mathsf{G}\psi$  for some  $\psi \in \mathcal{L}_t(\operatorname{sig})$ , then  $\operatorname{TERM}(\varphi) = \mathbf{f}_{\operatorname{always}}(\operatorname{TERM}(\psi))$  and  
- if  $\varphi = \mathsf{F}\psi$  for some  $\psi \in \mathcal{L}_t(\operatorname{sig})$ , then  $\operatorname{TERM}(\varphi) = \mathbf{f}_{\operatorname{finally}}(\operatorname{TERM}(\psi))$ ,

• - if 
$$\varphi = \psi_1 \mathsf{U} \psi_2$$
 for  $\psi_1, \psi_2 \in \mathcal{L}_t(\mathrm{sig})$ , then

$$\operatorname{Term}(\varphi) = \mathbf{f}_{\mathrm{until}}(\operatorname{Term}(\psi_1), \operatorname{Term}(\psi_2))$$

and

- if 
$$\varphi = \psi_1 \mathsf{P} \psi_2$$
 for  $\psi_1, \psi_2 \in \mathcal{L}_t(\mathrm{sig})$ , then

$$\mathrm{TERM}(\varphi) = \mathbf{f}_{\mathrm{precedes}}(\mathrm{TERM}(\psi_1), \mathrm{TERM}(\psi_2))$$

Now we construct

$$\psi_{1} = \mathbf{p}_{\text{new}} \left( \text{TERM} \left( \chi_{i_{0}}^{(1)} \right), \dots, \text{TERM} \left( \chi_{i_{k}}^{(1)} \right) \right) \text{ and}$$
  
$$\psi_{2} = \mathbf{p}_{\text{new}} \left( \text{TERM} \left( \chi_{j_{0}}^{(2)} \right), \dots, \text{TERM} \left( \chi_{j_{k}}^{(2)} \right) \right)$$

and compute

$$\psi = \mathrm{LGS}(\psi_1, \psi_2) =: \mathsf{p}_{\mathrm{new}}(t_1, \dots, t_k)$$

The set of generalized literals can now be extracted from the sequence  $t_1, \ldots, t_k$  of terms using the transformation LITERAL :  $\mathcal{T}(sig_{ext}) \to \mathcal{L}_t(sig)$  defined as follows:

- if  $t = \mathbf{f}_{\mathbf{p}}(t_1, \dots, t_n)$  for some  $\mathbf{p} \in P$  and  $t_1, \dots, t_n \in \mathcal{T}(\text{sig})$ , then  $\text{LITERAL}(t) = \mathbf{p}(t_1, \dots, t_n)$ ,
- if  $t = \mathbf{f}_{not}(\bar{t})$  for some  $\bar{t} \in \mathcal{T}(sig_{ext})$ , then  $LITERAL(t) = not(LITERAL(\bar{t}))$ ,
- - if  $t = \mathbf{f}_{next}(\bar{t})$  for some  $\bar{t} \in \mathcal{T}(sig_{ext})$ , then LITERAL $(t) = XLITERAL(\bar{t})$ , - if  $t = \mathbf{f}_{always}(\bar{t})$  for some  $\bar{t} \in \mathcal{T}(sig_{ext})$ , then LITERAL $(t) = GLITERAL(\bar{t})$  and

- if  $t = \mathbf{f}_{\text{finally}}(\bar{t})$  for some  $\bar{t} \in \mathcal{T}(\text{sig}_{\text{ext}})$ , then  $\text{LITERAL}(t) = \text{FLITERAL}(\bar{t})$ ,

• - if  $t = \mathbf{f}_{until}(t_1, t_2)$  for  $t_1, t_2 \in \mathcal{T}(sig_{ext})$ , then

$$LITERAL(t) = LITERAL(t_1)ULITERAL(t_2)$$

and

- if 
$$t = \mathbf{f}_{\text{precedes}}(t_1, t_2)$$
 for  $t_1, t_2 \in \mathcal{T}(\text{sig}_{\text{ext}})$ , then

$$LITERAL(t) = LITERAL(t_1)PLITERAL(t_2).$$

It is immediately clear from the definition of the operations TERM and LITERAL that we have for every literal  $\varphi$  and every term t:

LITERAL(TERM(
$$\varphi$$
)) =  $\varphi$  and  
TERM(LITERAL( $t$ )) =  $t$ 

Having constructed  $\psi = \mathbf{p}_{\text{new}}(t_1, \dots, t_k)$  as described above, we extract the generalized literals by computing the following set:

$$S_C = \{ \text{LITERAL}(t_1), \dots, \text{LITERAL}(t_k) \} = \{ \chi_1, \dots, \chi_k \}.$$

By assumption we have that there is at least one positive literal  $\chi_i$ . So the set of possible rules which can be extracted from  $S_C$  is nonempty. We will again use the operation TAIL defined on page 126. The algorithm to construct the set or rules from  $S_C$  is described in Algorithm 6.

#### Theorem 6.3.2

Let  $C_1$  and  $C_2$  be PROLOG(+T)-rules and let C be any PROLOG(+T)-rule from the set computed by Algorithm 6 given inputs  $C_1$  and  $C_2$ . Then C is a least generalization

Algorithm 6 Least Generalization of PROLOG(+T)-rules Input: PROLOG(+T)-rules  $C_1, C_2$ **Output**: set of least generalizations of  $C_1$  and  $C_2$ **Require:**  $C_1 = \varphi_1 := \psi_1^{(1)}, \dots, \psi_{n_1}^{(1)}, C_2 = \varphi_1 := \psi_1^{(2)}, \dots, \psi_{n_2}^{(2)}$  $1: \ L \gets \emptyset$  $1: D \leftarrow \psi$  $2: S_{C_1} \leftarrow \left\{ \neg \psi_1^{(1)}, \dots, \neg \psi_{n_1}^{(1)}, \varphi_1 \right\}$ 3:  $S_{C_2} \leftarrow \left\{ \neg \psi_1^{(2)}, \dots, \neg \psi_{n_2}^{(2)}, \varphi_2 \right\}$ 4:  $S \leftarrow S_{C_1} \cup S_{S_2}$ **Require:**  $S = \{l_1, ..., l_o\}$ 5: CO  $\leftarrow \emptyset$ 6: for i = 1, ..., o do for j = i + 1, ..., o do 7: 8: if  $i \neq j$  then if  $\text{TEMP}(l_i) = \text{TEMP}(l_i)$  then 9:  $Co \leftarrow Co \cup \{(l_i, l_j)\}$ 10:end if 11:12:end if end for 13:14: **end for Require:** Co = { $(l_{i_1}, l_{j_1}), \ldots, (l_{i_k}, l_{j_k})$ } 15:  $S_C \leftarrow \emptyset$ **Require:**  $p_{new}$  is some new predicate symbol with arity k16:  $\psi_1 \leftarrow \mathsf{p}_{\text{new}} \left( \text{TERM} \left( l_{i_1} \right), \dots, \text{TERM} \left( l_{i_k} \right) \right)$ 17:  $\psi_2 \leftarrow \mathsf{p}_{\text{new}} \left( \operatorname{Term} \left( l_{j_1} \right), \dots, \operatorname{Term} \left( l_{j_1} \right) \right)$ 18:  $\psi \leftarrow \text{LGS}(\psi_1, \psi_2)$ **Require:**  $\psi = p_{new}(t_1, \ldots, t_k)$ 19:  $S_C \leftarrow \{ \text{LITERAL}(t_1), \dots, \text{LITERAL}(t_k) \}$ **Require:**  $S_C = \{l_1, ..., l_k\}$ 20: for i = 1, ..., k do 21:if  $l_i$  is positive then  $L \leftarrow L \cup \{l_i := \text{TAIL}(S_C \setminus \{l_i\})\}$ 22:end if 23:24: end for 25: if  $L \neq \emptyset$  then return L26:27: else return {false} 28:29: end if

under subsumption of  $C_1$  and  $C_2$ .

**Proof**. Let  $C_1$  and  $C_2$  be given as

$$C_1 = \varphi_1 := -\psi_1^{(1)}, \dots, \psi_{n_1}^{(1)}$$
 and  
 $C_2 = \varphi_2 := -\psi_1^{(2)}, \dots, \psi_{n_2}^{(2)}.$ 

We first show that every clause C computed by Algorithm 6 given input  $C_1$  and  $C_2$  is a generalization of  $C_1$  and  $C_2$ . Let S be the set returned by Algorithm 6. If  $S = \{\texttt{false}\}$ , then the claim is immediate by definition of the ordering  $\succeq_s$ . Now assume that  $S \neq \emptyset$ . If C is any element from S, then C has been constructed from a literal  $\mathbf{p}_C(t_1, \ldots, t_n)$  for some predicate symbol  $\mathbf{p}$ , some  $n \in \mathbb{N}$  and terms  $t_1, \ldots, t_n$ . This literal has been constructed from  $C_1$  and  $C_2$  by application of Algorithm 4. So  $p_C(t_1, \ldots, t_n)$  is indeed a least generalization of the two literals constructed from  $C_1$  and  $C_2$ . This gives the claim of the theorem.

As in the case of computing sets of greatest specializations we have that each element in a set returned by Algorithm 6 is semantically equivalent to the remaining elements in this set. So we will again denote any PROLOG(T)-rule computed by Algorithm 6 given inputs  $C_1$  and  $C_2$  as

$$LGS(C_1, C_2).$$

#### Example 6.3.5

Consider the following three PROLOG(+T)-rules:

$$\begin{array}{lll} C_1 &=& \mathsf{Gp}(\mathtt{a}) :-\mathsf{Fq}(\mathtt{X})., \\ C_2 &=& \mathsf{p}(\mathtt{b}) :-\mathsf{not}(\mathtt{q}(\mathtt{X}))\mathsf{Ur}(\mathtt{X}), \mathsf{Fq}(\mathtt{a}). \text{ and} \\ C_3 &=& \mathsf{p}(\mathtt{a}) :-\mathsf{not}(\mathtt{q}(\mathtt{a}))\mathsf{Ur}(\mathtt{c}). \end{array}$$

We have  $LGS(C_1, C_2) = LGS(C_1, C_3) = \{ \texttt{false} \}$ . The computation of  $LGS(C_2, C_3)$  is

carried out as follows: We have  $Co = \{(p(a), p(a)), (q(X)Ur(X), q(a)Ur(c))\}$ . So  $p_{new}$  with arity 2 is added to the signature. Then the following two literals are created:

$$\begin{split} \psi_1 &= p_{\text{new}} \left( \mathbf{f}_{\mathbf{p}}(\mathbf{a}), \mathbf{f}_{\text{until}}(\mathbf{f}_{\mathbf{q}}(\mathbf{X}), \mathbf{f}_{\mathbf{r}}(\mathbf{X})) \right) \text{ and} \\ \psi_2 &= p_{\text{new}} \left( \mathbf{f}_{\mathbf{p}}(\mathbf{a}), \mathbf{f}_{\text{until}}(\mathbf{f}_{\mathbf{q}}(\mathbf{a}), \mathbf{f}_{\mathbf{r}}(\mathbf{c})) \right). \end{split}$$

The least generalization of these literals is  $p_{\text{new}}(f_p(\mathbf{a}), f_{\text{until}}(f_q(Z_1), f_r(Z_2)))$ . This gives  $S_C = \{p(\mathbf{a}), q(Z_1) \cup r(Z_2)\}$  and therefore the set C of least generalizations is

$$C = \{ \mathsf{p}(\mathsf{a}) := \mathsf{not}(\mathsf{q}(\mathsf{Z}_1)\mathsf{Ur}(\mathsf{Z}_2)), \mathsf{q}(\mathsf{Z}_1)\mathsf{Ur}(\mathsf{Z}_2) := \mathsf{not}(\mathsf{p}(\mathsf{a})). \}.$$

The results from this section and the section before are summarized in the following theorem.

#### Theorem 6.3.3

Let  $\mathcal{R}$  be the set of all PROLOG(+T)-rules. Then  $(\mathcal{R} \cup \{\{\texttt{false}\}\}, \succeq_s)$  is a lattice.

This theorem enables systems to refine programs in order to fit their specifications without taking hazards of *overgeneralization* or *overspecialization*. It will turn out important in the following chapter where we will define refinement operators for PROLOG(+T)rules.

# Refinement Operators for Prolog(+T)-programs

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We will now consider the problem of *refining* literals and rules. In contrast to the computation of least generalizations and greatest specializations as described in the last chapter, only one object is involved now. For objects from first order logic several refinement operators have been described and studied (see [98] and [126]). We will see that the refinement operators for literals can be directly used in order to refine PROLOG(+T)-literals while the refinement operators for first order clauses have to be extended in order to be useful for the computation of refinements of PROLOG(+T)-rules. But this extension causes the number of refinements to grow very fast. So it is necessary to control the

refinement process in order to avoid computational intractability. How to achieve this will be the topic of chapter 7.3.

The use of refinement operators in order to construct more special resp. more general objects from given ones dates back to 1981 when Shapiro introduced the model inference framework (see [145] and [146]). The approach posed there has gained a great interest in refinement operators although it has been shown by van der Laag (see [159]) that Shapiro's operator is (in contrast to Shapiro's arguments which shows that his proof is incorrect) not complete. However, in [159] and [98] it has been shown that complete refinement operators indeed exist.<sup>1</sup> [127]

Further research on refinement operators has pointed out several conditions for the existence of complete refinement operators (see [162], [161], [160], [127] and [163]). Additionally refinement operators for *theories* have been introduced (see e.g. [15]). Refinement operators for theories work on sets of clauses rather than on single clauses. This approach may yield smaller hypothesis-programs since the application of such refinement operators can be combined with techniques such as clause-deletion.

### 7.1. Refinement Operators for Prolog(+T)-Literals

We will now briefly resume some refinement operators for literals which have been described for first order logic atoms in [126]. The extension of these operators to PRO-LOG(+T)-literals is obvious and their properties carry over to PROLOG(+T). Therefore assume that the signature under consideration is sig =  $(\mathcal{X}, F, P, \alpha)$  with  $F = \{\mathbf{f}_1, \ldots, \mathbf{f}_n\}$ .

**Downward Refinement** Let  $\varphi \in \mathcal{L}_t(\operatorname{sig})$  be given with  $\operatorname{VaR}(\varphi) = \{Z_1, \ldots, Z_k\}$  and let  $X_1^{(1)}, \ldots, X_{\alpha(f_1)}^{(1)}, \ldots, X_1^{(n)}, \ldots, X_{\alpha(f_n)}^{(n)}$  be a sequence of pairwise distinct variables such that  $X_i^{(j)} \notin \operatorname{VaR}(\varphi)$  for all i, j. The downward refinement operator  $\Theta_d^{\mathcal{L}} : \mathcal{L}_t(\operatorname{sig}) \to \mathbb{C}$ 

<sup>&</sup>lt;sup>1</sup>The problem with Shapiro's incomplete refinement operator is simply due to the fact that it requires clauses to be reduced (see [133] and [78]). Relaxing this requirement yields on the one hand a larger search space but on the other hand it yields a complete operator.

1.	$GXFp\left(\mathtt{f}\left(\mathtt{X}_{1}^{(1)}\right)\right)Up\left(\mathtt{g}\left(\mathtt{f}\left(\mathtt{f}\left(\mathtt{X}_{1}^{(1)}\right)\right),\mathtt{Y}\right)\right)$
2.	$GXFp\left(g\left(\mathtt{X}_{1}^{(2)}, \mathtt{X}_{2}^{(2)}\right)\right) Up\left(g\left(\mathtt{f}\left(g\left(\mathtt{X}_{1}^{(2)}, \mathtt{X}_{2}^{(2)}\right)\right), \mathtt{Y}\right)\right)$
3.	$GXFp\left(\mathtt{X}\right)Up\left(\mathtt{g}\left(\mathtt{f}\left(\mathtt{X}\right),\mathtt{f}\left(\mathtt{X}_{1}^{\left(1\right)}\right)\right)\right)$
4.	$GXFp(\mathtt{X})  Up\left(g\left(\mathtt{f}(\mathtt{X}), g\left(\mathtt{X}_{1}^{(2)}, \mathtt{X}_{2}^{(2)}\right)\right)\right)$
5.	$GXFp(\mathtt{a})Up(\mathtt{g}(\mathtt{f}(\mathtt{a}),\mathtt{Y})),GXFp(\mathtt{X})Up(\mathtt{g}(\mathtt{f}(\mathtt{X}),\mathtt{a}))$
6.	$GXFp(\mathtt{Y})Up(\mathtt{g}(\mathtt{f}(\mathtt{Y}),\mathtt{Y})),GXFp(\mathtt{X})Up(\mathtt{g}(\mathtt{f}(\mathtt{X}),\mathtt{X}))$

Table 7.2.: Set of downward refinements of  $\varphi = \mathsf{GXFp}(\mathtt{X})\mathsf{Up}(\mathtt{g}(\mathtt{f}(\mathtt{X}), \mathtt{Y}))$ 

 $2^{\mathcal{L}_t(\text{sig})}$  is now defined as follows:

$$\Theta_{d}^{\mathcal{L}}(\varphi) = \left\{ \varphi \left\{ \frac{\mathsf{Z}_{i}}{\mathsf{f}_{j}\left(\mathsf{X}_{1}^{(j)}, \dots, \mathsf{X}_{\alpha(\mathsf{f}_{j})}^{(j)}\right)} \right\} \mid i = 1, \dots, k, j = 1, \dots, n \right\}$$
(7.1)

$$\cup \left\{\varphi\left\{\frac{\mathbf{Z}_j}{\mathbf{Z}_i}\right\} \mid i = 1, \dots, k, j = 1, \dots, k, i \neq j\right\}.$$
(7.2)

The set from (7.1) creates the literals which emerge from the original literal by replacing *one* variable by all possible instantiations of function symbols with (new) variables (note that replacement of variables with constant symbols is just a special case of this case) while the set from (7.2) replaces variables with other variables occurring in the original expression. The procedure carried out by the application of  $\Theta_d^{\mathcal{L}}$  is summarized in Algorithm 7.

#### Example 7.1.1

Assume that  $F = \{\mathbf{f}, \mathbf{g}, \mathbf{a}\}$  and  $P = \{\mathbf{p}\}$  with  $\alpha(\mathbf{p}) = \alpha(\mathbf{g}) = 2$ ,  $\alpha(\mathbf{f}) = 1$  and  $\alpha(\mathbf{a}) = 0$ . If  $\varphi = \mathsf{GXFp}(\mathbf{X})\mathsf{Up}(\mathbf{g}(\mathbf{f}(\mathbf{X}), \mathbf{Y}))$ , then  $\operatorname{VAR}(\varphi) = \{\mathbf{X}, \mathbf{Y}\}$ . The new variables to be introduced are  $\mathbf{X}_1^{(1)}, \mathbf{X}_1^{(2)}$  and  $\mathbf{X}_2^{(2)}$ . The result of  $\Theta_d^{\mathcal{L}}$  given input  $\varphi$  is summarized in Table 7.2.

It is easy to see that

1.  $\Theta_d^{\mathcal{L}}$  is ideal and

2.  $\left|\Theta_d^{\mathcal{L}}(\varphi)\right| \leq |\operatorname{Var}(\varphi)| \cdot (|F| + |\operatorname{Var}(\varphi)| - 1)$  for every  $\varphi \in \mathcal{L}_t(\operatorname{sig})$ .

Algorithm 7 Downward–Refinement of PROLOG(+T)–literals

**Input**: PROLOG(+T)-literal  $\varphi$  built from sig = ( $\mathcal{X}, F, P, \alpha$ ) **Output**: set of specialized literals **Require:**  $VAR(\varphi) = \{X_1, \dots, X_k\}$ **Require:**  $F = \{\mathbf{f}_1, \dots, \mathbf{f}_n\}$  **Require:**  $\mathbf{X}_1^{(1)}, \dots, \mathbf{X}_{\alpha(\mathbf{f}_1)}^{(1)}, \dots, \mathbf{X}_1^{(n)}, \dots, \mathbf{X}_{\alpha(\mathbf{f}_n)}^{(n)}$  is a sequence of pairwise distinct variables from  $\mathcal{X} \setminus \operatorname{VAR}(\varphi)$ 1:  $Ref \leftarrow \emptyset$ 2: for i = 1, ..., k do for  $j = 1, \ldots, n$  do 3:  $\sigma \leftarrow \left\{ \frac{\mathtt{X}_i}{\mathtt{f}_j \left( \mathtt{X}_1^{(j)}, ..., \mathtt{X}_{\alpha(\mathtt{f}_j)}^{(j)} \right)} \right\}$ 4:  $Ref \leftarrow Ref \cup \{\sigma(a)\}$ 5: end for 6: 7: end for 8: for i = 1, ..., k do for  $j = 1, \ldots, k$  do 9: if  $i \neq j$  then 10: $\sigma \leftarrow \left\{ \frac{\mathbf{x}_i}{\mathbf{x}_j} \right\}$ Ref  $\leftarrow$  Ref  $\cup \{ \sigma(\varphi) \}$ 11: 12:13:end if end for 14:15: end for 16: return Ref

Upward Refinement The dual case of downward refinement is upward refinement. The upward refinement operator  $\Theta_u^{\mathcal{L}}$  for PROLOG(+T)-literals which we will present now is (as the operator  $\Theta_d^{\mathcal{L}}$  is) an extension of an ideal refinement operator for first order logic literals. In order to present the refinement operator we need some more formal concepts.

First we will define the mapping TERMS :  $\mathcal{L}_t(\text{sig}) \cup \mathcal{T}(\text{sig}) \rightarrow 2^{\mathcal{T}(\text{sig})}$  which returns all terms which occur in a term respectively in a literal:

1. if  $t = \mathbf{X} \in \mathcal{X}$ , then  $\operatorname{Terms}(t) = \{\mathbf{X}\},\$ 

- 2. if  $t = \mathbf{f}(t_1, \dots, t_n)$  for some function symbol  $\mathbf{f}$  with arity n and terms  $t_1, \dots, t_n$ , then  $\operatorname{TERMS}(t) = \{t\} \cup \bigcup_{i=1}^n \operatorname{TERMS}(t_i)$ ,
- 3. if  $\varphi = \mathbf{p}(t_1, \dots, t_n)$  for some predicate symbol  $\mathbf{p}$  with arity n and terms  $t_1, \dots, t_n$ , then  $\operatorname{TERMS}(\varphi) = \bigcup_{i=1}^n \operatorname{TERMS}(t_i)$ ,
- 4. if  $\varphi = \operatorname{not}(\psi)$  for some  $\psi \in \mathcal{L}_t(\operatorname{sig})$ , then  $\operatorname{TERMS}(\varphi) = \operatorname{TERMS}(\psi)$ ,
- 5. if  $\varphi = \oplus \psi$  for some  $\psi \in \mathcal{L}_t(\text{sig})$  and some  $\oplus \in \{X, G, F\}$ , then  $\text{TERMS}(\varphi) = \text{TERMS}(\psi)$  and
- 6. if  $\varphi = \psi_1 \oplus \psi_2$  for  $\psi_1, \psi_2 \in \mathcal{L}_t(\text{sig})$  and  $\oplus \in \{\mathsf{U}, \mathsf{P}\}$ , then  $\operatorname{TERMS}(\varphi) = \operatorname{TERMS}(\psi_1) \cup \operatorname{TERMS}(\psi_2)$ .

We will call a term  $t \in \mathcal{T}(\text{sig})$  simple if  $t = \mathbf{f}(\mathbf{X}_1, \dots, \mathbf{X}_n)$  for a function symbol  $\mathbf{f}$ with arity n and variables  $\mathbf{X}_1, \dots, \mathbf{X}_n$  such that  $\mathbf{X}_i \neq \mathbf{X}_j$  for  $i \neq j$ .

Now let  $o_1$  and  $o_2$  be any objects (terms or literals). The set of all *occurrences* of  $o_1$  in  $o_2$  is defined as

$$OCC(o_1, o_2) = \{ p \in POS(o_2) \mid o_1 \mid_p = o_2 \}.$$

An occurrence  $p_1$  of an object  $o_1$  is said to be *inside* an occurrence  $p_2$  of another object  $o_2$  if there is  $p \in \mathbb{N}^*$  such that  $p_2p = p_1$ .

Recall that for a literal  $\varphi$ ,  $p \in \text{Pos}(\varphi)$  and a term t the literal  $\varphi[t]_p$  is defined as the literal which emerges from  $\varphi$  by replacing the term at position p with t. Similarly for  $p_1, \ldots, p_k \in \text{Pos}(\varphi)$  and k > 1, the literal  $\varphi[t]_{p_1,\ldots,p_k}$  emerges from  $\varphi$ by replacing the terms at positions  $p_1, \ldots, p_k$  with t.

These concepts enable the definition of the upward refinement operator  $\Theta_u^{\mathcal{L}}$  for FoLTL-literals.

$$\Theta_{u}^{\mathcal{L}}(\varphi) = \begin{cases} \emptyset \neq \{p_{1}, \dots, p_{k}\} = \operatorname{OCC}(t, \varphi) \text{ for every simple } t \\ \varphi[\mathbf{Z}]_{p_{1},\dots,p_{k}} \mid \text{ such that for every } \mathbf{X} \in \operatorname{VAR}(t) \text{ every } p \in \operatorname{OCC}(\mathbf{X}, t) \\ \text{ is inside one of the } p_{j} \text{ and } \mathbf{Z} \notin \operatorname{VAR}(\varphi) \end{cases} \\ =:S_{1} \\ \cup \underbrace{\begin{cases} \varphi[\mathbf{Z}]_{p_{1},\dots,p_{k}} \mid & \text{ for every } \mathbf{a} \in F \text{ such that } \alpha(\mathbf{a}) = 0 \text{ and every} \\ \emptyset \neq \{p_{1},\dots,p_{k}\} \subseteq \operatorname{OCC}(\mathbf{a},\varphi) \text{ and some } \mathbf{Z} \notin \operatorname{VAR}(\varphi) \end{cases}}_{=:S_{2}} \\ \cup \underbrace{\begin{cases} \varphi[\mathbf{Z}]_{p_{1},\dots,p_{k}} \mid & \text{ for every } \mathbf{X} \in \operatorname{VAR}(\varphi), \text{ every set} \\ \emptyset \neq \{p_{1},\dots,p_{k}\} \subset \operatorname{OCC}(\mathbf{X},\varphi) \text{ and some } \mathbf{Z} \notin \operatorname{VAR}(\varphi) \end{cases}}_{=:S_{3}} \end{cases}} \end{cases}$$

The estimation of the number of elements in  $\Theta_u^{\mathcal{L}}(\varphi)$  is not that easy and we are only able to present a very weak estimation. We have

$$\begin{split} |S_1| &\leq |\operatorname{Terms}(\varphi)|, \\ |S_2| &\leq |F| \cdot \left(2^{|\operatorname{Pos}(\varphi)|} - 1\right) \text{ and} \\ |S_3| &\leq |\operatorname{Var}(\varphi)| \cdot \left(2^{|\operatorname{Pos}(\varphi)|} - 2\right), \end{split}$$

which gives

$$\begin{aligned} \left| \Theta_{u}^{\mathcal{L}}(\varphi) \right| &\leq |S_{1}| + |S_{2}| + |S_{3}| \\ &\leq |\operatorname{Terms}(\varphi)| + |F| \cdot \left( 2^{|\operatorname{Pos}(\varphi)|} - 1 \right) + |\operatorname{Var}(\varphi)| \cdot \left( 2^{|\operatorname{Pos}(\varphi)|} - 2 \right). \end{aligned}$$

This estimation is not very precise as the following example shows. But it is not yet clear how a better estimation might be derived from the definition of  $\Theta_u^{\mathcal{L}}$  without taking the structure of the involved terms into account.

#### Example 7.1.2

Consider the signature sig =  $(\mathcal{X}, \{\mathbf{f}, \mathbf{a}\}, \{\mathbf{p}\}, \alpha)$  with  $\alpha(\mathbf{a}) = 0$ ,  $\alpha(\mathbf{f}) = 2$  and  $\alpha(\mathbf{p}) = 3$  and the literal  $\varphi = \mathsf{FGp}(\mathbf{X}_1, \mathbf{f}(\mathbf{a}, \mathbf{f}(\mathbf{X}_1, \mathbf{f}(\mathbf{a}, \mathbf{a}(\mathbf{a}, \mathbf{a})))), \mathbf{f}(\mathbf{X}_1, \mathbf{X}_2))$ . Then the only simple term in  $\mathsf{TERMS}(\varphi)$  is  $\mathbf{f}(\mathbf{X}_2, \mathbf{X}_3)$  at position p = 113. So  $\Theta_u^{\mathcal{L}}(\varphi)$  contains the literal  $\mathsf{FGp}(\mathbf{X}_1, \mathbf{f}(\mathbf{a}, \mathbf{f}(\mathbf{a}, \mathbf{f}(\mathbf{a}, \mathbf{a}(\mathbf{a}, \mathbf{a})))), \mathbf{Z})$ . Since  $\mathsf{VAR}(\varphi) = \{\mathbf{X}_1, \mathbf{X}_2, \mathbf{X}_3\}$  we have that  $\mathsf{OCC}(\mathbf{X}_1, \varphi) = \{111\}$ ,  $\mathsf{OCC}(\mathbf{X}_2, \varphi) = \{1131\}$  and  $\mathsf{OCC}(\mathbf{X}_3, \varphi) = \{1132\}$ . None of these sets has a nonempty proper subset so  $S_3 = \emptyset$  in this case. Furthermore  $\mathsf{OCC}(\mathbf{a}, \varphi) = \{1121, 11221, 112221, 1122221, 1122222\}$ , so there are 31 possible literals which might be added by  $\Theta_u^{\mathcal{L}}$ . So the overall size of  $\Theta_u^{\mathcal{L}}(\varphi)$  is 32 while the above estimation yields

$$\begin{aligned} |\Theta_{u}^{\mathcal{L}}(\varphi)| &\leq |S_{1}| + |S_{2}| + |S_{3}| \\ &\leq |\mathrm{T}_{\mathrm{ERMS}}(\varphi)| + |F| \cdot \left(2^{|\mathrm{Pos}(\varphi)|} - 1\right) + |\mathrm{VAR}(\varphi)| \cdot \left(2^{|\mathrm{Pos}(\varphi)|} - 2\right) \\ &= 9 + 2 \cdot \left(2^{16} - 1\right) + 3 \cdot \left(2^{16} - 2\right) \\ &= 327681 \end{aligned}$$

The complete set of refined literals is listed in Table 7.3.

Since the original refinement operator for first order logic literals is ideal,  $\Theta_u^{\mathcal{L}}$  is also ideal. The procedure for computing  $\Theta_u^{\mathcal{L}}(\varphi)$  is summarized in Algorithm 8.

#### 7.2. Refinement Operators for Rules

We will now present adapted versions of *classical* refinement operators for sets of literals, i.e. for PROLOG(+T)-rules. As in the case of refining literals, there is in general more than one refinement of an input rule. So we will have a set of rules as the result of a refinement operation. Each of these rules is a set from which we may construct one or more rules each of which is a refinement of the original rule.

$\begin{array}{ll} 1. & FGp(X_1, f(a, f(a, f(a, f(a, a))), Z) \\ \hline 2. & FGp(X_1, f(a, f(a, f(a, f(a, Z)))), f(X_1, X_2)) \\ \hline 3. & FGp(X_1, f(a, f(a, f(a, f(Z, Z)))), f(X_1, X_2)) \\ \hline 4. & FGp(X_1, f(a, f(a, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 5. & FGp(X_1, f(a, f(a, f(Z, f(a, Z)))), f(X_1, X_2)) \\ \hline 6. & FGp(X_1, f(a, f(a, f(Z, f(a, Z)))), f(X_1, X_2)) \\ \hline 7. & FGp(X_1, f(a, f(a, f(Z, f(Z, Z)))), f(X_1, X_2)) \\ \hline 8. & FGp(X_1, f(a, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 9. & FGp(X_1, f(a, f(Z, f(a, f(a, Z)))), f(X_1, X_2)) \\ \hline 10. & FGp(X_1, f(a, f(Z, f(a, f(a, Z)))), f(X_1, X_2)) \\ \hline 11. & FGp(X_1, f(a, f(Z, f(a, f(Z, Z)))), f(X_1, X_2)) \\ \hline 12. & FGp(X_1, f(a, f(Z, f(a, f(Z, Z)))), f(X_1, X_2)) \\ \hline 13. & FGp(X_1, f(a, f(Z, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 14. & FGp(X_1, f(a, f(Z, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 15. & FGp(X_1, f(a, f(Z, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 15. & FGp(X_1, f(A, f(Z, f(a, f(a, a, f(a, f($		
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	1.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{a})))), \mathtt{Z})$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	2.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{Z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	3.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	4.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{Z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	5.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
8. $FGp(X_1, f(a, f(Z, f(Z, Z)))), f(X_1, X_2))$ 9. $FGp(X_1, f(a, f(Z, f(a, f(Z, Z)))), f(X_1, X_2))$ 10. $FGp(X_1, f(a, f(Z, f(a, f(Z, A)))), f(X_1, X_2))$ 11. $FGp(X_1, f(a, f(Z, f(a, f(Z, A)))), f(X_1, X_2))$ 12. $FGp(X_1, f(a, f(Z, f(Z, f(a, A)))), f(X_1, X_2))$ 13. $FGp(X_1, f(a, f(Z, f(Z, f(a, A)))), f(X_1, X_2))$ 14. $FGp(X_1, f(a, f(Z, f(Z, f(Z, A)))), f(X_1, X_2))$ 15. $FGp(X_1, f(a, f(Z, f(Z, f(Z, A)))), f(X_1, X_2))$ 16. $FGp(X_1, f(a, f(Z, f(Z, f(Z, A)))), f(X_1, X_2))$ 17. $FGp(X_1, f(Z, f(a, f(a, f(a, A)))), f(X_1, X_2))$ 18. $FGp(X_1, f(Z, f(a, f(a, f(A, A)))), f(X_1, X_2))$ 19. $FGp(X_1, f(Z, f(a, f(a, f(Z, A)))), f(X_1, X_2))$ 20. $FGp(X_1, f(Z, f(a, f(A, f(Z, A)))), f(X_1, X_2))$ 21. $FGp(X_1, f(Z, f(a, f(A, f(Z, A)))), f(X_1, X_2))$ 22. $FGp(X_1, f(Z, f(a, f(A, f(Z, A)))), f(X_1, X_2))$ 23. $FGp(X_1, f(Z, f(a, f(A, f(Z, A)))), f(X_1, X_2))$ 24. $FGp(X_1, f(Z, f(a, f(A, f(Z, A)))), f(X_1, X_2))$ 25. $FGp(X_1, f(Z, f(a, f(Z, f(A, A)))), f(X_1, X_2))$ 26. $FGp(X_1, f(Z, f(A, f(Z, f(A, A)))), f(X_1, X_2))$ 27. $FGp(X_1, f(Z, f(A, f(Z, f(A, A)))), f(X_1, X_2))$ 28. $FGp(X_1, f(Z, f(A, f(Z, f(Z, A)))), f(X_1, X_2))$ 29. $FGp(X_1, f(Z, f(A, f(Z, f(Z, A)))), f(X_1, X_2))$ 20. $FGp(X_1, f(Z, f(A, f(Z, f(Z, A)))), f(X_1, X_2))$ 21. $FGp(X_1, f(Z, f(A, f(Z, f(Z, A)))), f(X_1, X_2))$ 23. $FGp(X_1, f(Z, f(A, f(Z, f(Z, A)))), f(X_1, X_2))$ 24. $FGp(X_1, f(Z, f(A, f(Z, f(Z, A)))), f(X_1, X_2))$ 25. $FGp(X_1, f(Z, f(A, f(Z, f(Z, A)))), f(X_1, X_2))$ 26. $FGp(X_1, f(Z, f(A, f(Z, f(Z, A)))), f(X_1, X_2))$ 27. $FGp(X_1, f(Z, f(A, f(Z, f(Z, A)))), f(X_1, X_2))$ 28. $FGp(X_1, f(Z, f(Z, f(Z, f(Z, A)))), f(X_1, X_2))$ 29. $FGp(X_1, f(Z, f(Z, f(Z, f(Z, A)))), f(X_1, X_2))$ 30. $FGp(X_1, f(Z, f(Z, f(Z, f(Z, A)))), f(X_1, X_2))$ 31. $FGp(X_1, f(Z, f(Z, f(Z, f(Z, A)))), f(X_1, X_2))$	6.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{Z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
9. $FGp(X_1, f(a, f(Z, f(a, f(a, a)))), f(X_1, X_2))$ 10. $FGp(X_1, f(a, f(Z, f(a, f(z, a)))), f(X_1, X_2))$ 11. $FGp(X_1, f(a, f(Z, f(a, f(Z, a)))), f(X_1, X_2))$ 12. $FGp(X_1, f(a, f(Z, f(a, f(Z, a)))), f(X_1, X_2))$ 13. $FGp(X_1, f(a, f(Z, f(Z, a(a, Z)))), f(X_1, X_2))$ 14. $FGp(X_1, f(a, f(Z, f(Z, f(Z, a)))), f(X_1, X_2))$ 15. $FGp(X_1, f(a, f(Z, f(Z, f(Z, a)))), f(X_1, X_2))$ 16. $FGp(X_1, f(a, f(Z, f(a, f(a, f(a, a)))), f(X_1, X_2))$ 17. $FGp(X_1, f(Z, f(a, f(a, f(a, a)))), f(X_1, X_2))$ 18. $FGp(X_1, f(Z, f(a, f(a, f(a, a)))), f(X_1, X_2))$ 19. $FGp(X_1, f(Z, f(a, f(a, f(Z, a)))), f(X_1, X_2))$ 20. $FGp(X_1, f(Z, f(a, f(a, f(Z, a)))), f(X_1, X_2))$ 21. $FGp(X_1, f(Z, f(a, f(a, f(Z, a)))), f(X_1, X_2))$ 22. $FGp(X_1, f(Z, f(a, f(a, f(Z, a)))), f(X_1, X_2))$ 23. $FGp(X_1, f(Z, f(a, f(a, f(Z, a)))), f(X_1, X_2))$ 24. $FGp(X_1, f(Z, f(a, f(a, f(Z, f(a, a)))), f(X_1, X_2)))$ 25. $FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2))$ 26. $FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2))$ 27. $FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2))$ 28. $FGp(X_1, f(Z, f(a, f(Z, f(Z, a)))), f(X_1, X_2))$ 29. $FGp(X_1, f(Z, f(a, f(Z, f(Z, a)))), f(X_1, X_2))$ 20. $FGp(X_1, f(Z, f(a, f(Z, f(Z, a)))), f(X_1, X_2))$ 30. $FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2))$ 31. $FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2))$	7.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	8.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{Z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	9.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	10.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{Z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	11.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	12.	
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	13.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{c c} 16. & FGp(X_1, f(a, f(Z, f(Z, f(Z, Z)))), f(X_1, X_2)) \\ \hline 17. & FGp(X_1, f(Z, f(a, f(a, f(a, a)))), f(X_1, X_2)) \\ \hline 18. & FGp(X_1, f(Z, f(a, f(a, f(a, a)))), f(X_1, X_2)) \\ \hline 19. & FGp(X_1, f(Z, f(a, f(a, f(Z, a)))), f(X_1, X_2)) \\ \hline 20. & FGp(X_1, f(Z, f(a, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 21. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 22. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 23. & FGp(X_1, f(Z, f(a, f(Z, f(a, 2)))), f(X_1, X_2)) \\ \hline 24. & FGp(X_1, f(Z, f(a, f(a, f(Z, a)))), f(X_1, X_2)) \\ \hline 25. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 26. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 27. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 28. & FGp(X_1, f(Z, f(a, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 29. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 30. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f$	14.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{a}(\mathtt{a}, \mathtt{Z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	15.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	16.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{Z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{r c c c c c c c c c c c c c c c c c c c$	17.	$FGp(\mathtt{X}_1,\mathtt{f}(\mathtt{Z},\mathtt{f}(\mathtt{a},\mathtt{f}(\mathtt{a},\mathtt{f}(\mathtt{a},\mathtt{a})))),\mathtt{f}(\mathtt{X}_1,\mathtt{X}_2))$
$\begin{array}{c ccccc} 20. & FGp(X_1, f(Z, f(a, f(a, f(Z, Z)))), f(X_1, X_2)) \\ \hline 21. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 22. & FGp(X_1, f(Z, f(a, f(Z, f(a, Z)))), f(X_1, X_2)) \\ \hline 23. & FGp(X_1, f(Z, f(a, f(a, f(Z, a)))), f(X_1, X_2)) \\ \hline 24. & FGp(X_1, f(Z, f(a, f(a, f(Z, f(a)))), f(X_1, X_2)) \\ \hline 25. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 26. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline 27. & FGp(X_1, f(Z, f(a, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 28. & FGp(X_1, f(Z, f(a, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 29. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 30. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline 31. & FGp(X_1, f(Z, f(Z, $	18.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{c c} 21. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline \\ 22. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline \\ 23. & FGp(X_1, f(Z, f(a, f(a, f(z, f(a, z)))), f(X_1, X_2)) \\ \hline \\ 24. & FGp(X_1, f(Z, f(a, f(a, f(Z, a)))), f(X_1, X_2)) \\ \hline \\ 25. & FGp(X_1, f(Z, f(a, f(z, f(a, a)))), f(X_1, X_2)) \\ \hline \\ 26. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline \\ 27. & FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline \\ 28. & FGp(X_1, f(Z, f(a, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \hline \\ 29. & FGp(X_1, f(Z, f(Z, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline \\ 30. & FGp(X_1, f(Z, f(Z, f(Z, f(a, a)))), f(X_1, X_2)) \\ \hline \\ 31. & FGp(X_1, f(Z, f(Z, f(Z, f(Z, a, a)))), f(X_1, X_2)) \\ \hline \\ \end{array}$	19.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	20.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{Z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{c} 23.  FGp(X_1, f(Z, f(a, f(a, f(Z, a)))), f(X_1, X_2)) \\ 24.  FGp(X_1, f(Z, f(a, f(a, f(Z, z)))), f(X_1, X_2)) \\ 25.  FGp(X_1, f(Z, f(a, f(Z, f(a, a)))), f(X_1, X_2)) \\ 26.  FGp(X_1, f(Z, f(a, f(Z, f(a, z)))), f(X_1, X_2)) \\ 27.  FGp(X_1, f(Z, f(a, f(Z, f(Z, a)))), f(X_1, X_2)) \\ 28.  FGp(X_1, f(Z, f(a, f(Z, f(Z, z)))), f(X_1, X_2)) \\ 29.  FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ 30.  FGp(X_1, f(Z, f(Z, f(Z, f(Z, a, z)))), f(X_1, X_2)) \\ 31.  FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2)) \\ \end{array}$	21.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	22.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{Z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	23.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	24.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{Z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{c} 27.  FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2)) \\ 28.  FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2)) \\ 29.  FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2)) \\ 30.  FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2)) \\ 31.  FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2)) \\ \end{array}$	25.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	26.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{Z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$\overline{27.}$	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
30. $FGp(X_1, f(Z, f(Z, f(Z, f(a, z)))), f(X_1, X_2))$ 31. $FGp(X_1, f(Z, f(Z, f(Z, f(Z, a)))), f(X_1, X_2))$	28.	$FGp(\mathtt{X}_1,\mathtt{f}(\mathtt{Z},\mathtt{f}(\mathtt{a},\mathtt{f}(\mathtt{Z},\mathtt{f}(\mathtt{Z},\mathtt{Z})))),\mathtt{f}(\mathtt{X}_1,\mathtt{X}_2))$
31. $FGp(X_1, \mathbf{f}(Z, \mathbf{f}(Z, \mathbf{f}(Z, \mathbf{f}(Z, \mathbf{a})))), \mathbf{f}(X_1, X_2))$	29.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
	30.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{a}, \mathtt{z})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
32. $FGp(X_1, f(Z, f(Z, f(Z, f(Z, Z)))), f(X_1, X_2))$	31.	$FGp(\mathtt{X}_1, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{f}(\mathtt{Z}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$
	32.	$FGp(\mathtt{X}_1,\mathtt{f}(\mathtt{Z},\mathtt{f}(\mathtt{Z},\mathtt{f}(\mathtt{Z},\mathtt{f}(\mathtt{Z},\mathtt{Z})))),\mathtt{f}(\mathtt{X}_1,\mathtt{X}_2))$

 $\text{Table 7.3.: Set of upward refinements for } \varphi = \mathsf{FGp}(\mathtt{X}_1, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{f}(\mathtt{a}, \mathtt{a})))), \mathtt{f}(\mathtt{X}_1, \mathtt{X}_2))$ 

#### Algorithm 8 Upward–Refinement of PROLOG(+T)–literals

```
Input: PROLOG(+T)-literal \varphi
Output: set of generalized literals
Require: VAR(C) = {X_1, \ldots, X_k}
Require: Z \in \mathcal{X} \setminus \{X_1, \ldots, X_k\}
Require: TERMS(\varphi) = \{t_1, \ldots, t_n\}
 1: Ref \leftarrow \emptyset
 2: for i = 1, ..., n do
        if t_i is simple then
 3:
                     VAR(t_i) = \{Z_1, \dots, Z_l\}
Require:
 4:
           for o = 1, \ldots, l do
              if each p \in OCC(\mathbf{Z}_o, t_i) is inside of one element from OCC(t_i, \varphi) then
 5:
Require:
                            \{p_1,\ldots,p_m\} = \operatorname{OCC}(t_i,\varphi)
                  Ref \leftarrow Ref \cup \{\varphi[\mathbf{Z}]_{p_1,\dots,p_m}\}
 6:
              end if
 7:
           end for
 8:
        end if
 9:
10: end for
Require: F = \{f_1, \ldots, f_{|F|}\}
11: for i = 1, ..., |F| do
12:
        if \alpha(\mathbf{f}_i) = 0 then
           for each \emptyset \neq \{p_1, \ldots, p_m\} \subseteq OCC(f_i, \varphi) do
13:
               Ref \leftarrow Ref \cup \{\varphi[\mathbf{Z}]_{p_1,\ldots,p_m}\}
14:
15:
           end for
16:
        end if
17: end for
18: for i = 1, ..., k do
        for each \emptyset \neq \{p_1, \ldots, p_m\} \subset OCC(X_i, \varphi) do
19:
20:
           Ref \leftarrow Ref \cup \{\varphi[\mathbf{Z}]p_1, \ldots, p_m\}
        end for
21:
22: end for
23: return Ref
```

In the following sections we will sometimes refer to certain refinement operators which have been introduced for first order logic literals respectively clauses. A general description of refinement operators can be found in [98]. In particular we will refer to the operators  $\rho_{\mathcal{A}}$  (downward refinement of atoms),  $\delta_{\mathcal{A}}$  (upward refinement of atoms),  $\rho_l$ (downward refinement of clauses) and  $\delta_u$  (upward refinement of clauses) from [126].

#### 7.2.1. Downward Refinement

The basic idea of *downward refinement* of PROLOG(+T)-rules is first to consider the set  $S_C$  induced by a rule C and then to add certain literals to  $S_C$ . Additionally we will replace variables with terms and variables with other variables from the original rule as in the case of the operator  $\Theta_d^{\mathcal{L}}$ .

The original downward refinement operator presented in [98] only treats first order clauses. So in order to construct refinements of PROLOG(+T)-rules we have to deal with the temporal operators. This will be done as follows:

- whenever a literal  $\psi$  is contained in the original set  $S_C$ , the set of refinements contains  $S_C \cup \{\oplus \psi\}$  and  $S_C \cup \{\mathsf{not}(\oplus \psi)\}$  for  $\oplus \in \{\mathsf{X}, \mathsf{G}, \mathsf{F}\}$  and
- whenever two literals ψ<sub>1</sub> and ψ<sub>2</sub> are contained in S<sub>C</sub>, the set of refinements contains both S<sub>C</sub> ∪ {ψ<sub>1</sub> ⊕ ψ<sub>2</sub>} and S<sub>C</sub> ∪ {not(ψ<sub>1</sub> ⊕ ψ<sub>2</sub>)} for ⊕ ∈ {U, P}.

From the resulting set of literals we will extract those rules which can be written using a head literal which is positive.

The operator  $\Theta_d^{\mathcal{R}}: 2^{\mathcal{L}_t(\text{sig})} \to 2^{2^{\mathcal{L}_t(\text{sig})}}$  is therefore defined as follows:

$$\Theta_{d}^{\mathcal{R}}(S_{C}) = \left\{ S_{C} \left\{ \frac{\mathsf{Z}}{\mathsf{f}(\mathsf{X}_{1},\ldots,\mathsf{X}_{l})} \right\} \mid \begin{array}{c} \mathsf{X}_{1},\ldots,\mathsf{X}_{l} \notin \mathsf{VAR}(C), \\ \mathsf{X}_{i} \neq \mathsf{X}_{j} \text{ for } i \neq j \end{array} \right\}$$
(7.3)

$$\cup \left\{ S_C \left\{ \frac{\mathsf{Z}}{\mathsf{X}} \right\} \mid \mathsf{X}, \mathsf{Z} \in \operatorname{Var}(C), \mathsf{X} \neq \mathsf{Z} \right\}$$
(7.4)

$$\cup \left\{ S_C \cup \{ \mathbf{p}(\mathbf{X}_1, \dots, \mathbf{X}_l) \} \mid \begin{array}{c} \mathbf{p} \in P, \alpha(\mathbf{p}) = l, \mathbf{X}_1, \dots, \mathbf{X}_l \notin \operatorname{VAR}(C) \\ \mathbf{X}_i \neq \mathbf{X}_j \text{ for } i \neq j \end{array} \right\}$$
(7.5)

$$\cup \left\{ S_C \cup \{ \operatorname{not}(\operatorname{p}(X_1, \dots, X_l)) \} \mid X_1, \dots, X_l \notin \operatorname{VAR}(C) \\ X_i \neq X_j \text{ for } i \neq j \right\}$$
(7.6)

$$\cup \{S_C \cup \{\mathsf{X}\psi\} \mid \psi \in S_C\} \cup \{S_C \cup \{\mathsf{not}(\mathsf{X}\psi)\} \mid \psi \in S_C\}$$
(7.7)

$$\cup \{S_C \cup \{\mathsf{G}\psi\} \mid \psi \in S_C\} \cup \{S_C \cup \{\mathsf{not}(\mathsf{G}\psi)\} \mid \psi \in S_C\}$$
(7.8)

$$\cup \{S_C \cup \{\mathsf{F}\psi\} \mid \psi \in S_C\} \cup \{S_C \cup \{\mathsf{not}(\mathsf{F}\psi)\} \mid \psi \in S_C\}$$
(7.9)

$$\cup \{S_C \cup \{\psi_1 \cup \psi_2\} \mid \psi_1, \psi_2 \in S_C\}$$
(7.10)

$$\cup \{S_C \cup \{ \mathsf{not}(\psi_1 \cup \psi_2) \} \mid \psi_1, \psi_2 \in S_C \}$$
(7.11)

$$\cup \{S_C \cup \{\psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}$$
(7.12)

$$\cup \{S_C \cup \{\mathsf{not}(\psi_1 \mathsf{P}\psi_2)\} \mid \psi_1, \psi_2 \in S_C\}.$$
(7.13)

The line (7.3) generates all sets of literals which emerge from the original set by instantiating variables with terms. This construction is an obvious extension of the construction from (7.1). Similarly the second line (7.4) adds sets of literals in which single variables have been replaced by other variables from the original set. The lines (7.5) and (7.6)add new literals to the original set of literals. Finally in lines (7.7) to (7.12) temporal literals built up from literals of the original set are added as described above. Obviously the resulting set is subsumed by the original set. Furthermore the literals which have been added are general enough to be instantiated to more special literals. The complete procedure is summarized in Algorithm 9.

We illustrate the upward refinement of PROLOG(+T)-rules in the following example.

#### Example 7.2.1

Let sig =  $(\mathcal{X}, F, P, \alpha)$  with  $F = \{f, g, a\}, P = \{p, q\}$  and  $\alpha(f) = 1, \alpha(g) = 2, \alpha(a) = 0$ ,

 $\alpha(\mathbf{p}) = 1$  and  $\alpha(\mathbf{q}) = 2$  be given. If  $C = \mathsf{Gp}(X_1) :- \mathsf{Fq}(X_1, \mathbf{f}(X_2)), \mathbf{q}(\mathbf{a}, \mathbf{a})\mathsf{Up}(\mathbf{a})$ , then the set of refined rules is as listed in Tables 7.5, 7.7 and 7.9.

#### Theorem 7.2.1

 $\Theta_d^{\mathcal{R}}$  is locally finite and complete.

**Proof.** Locally finiteness is immediately by definition of  $\Theta_d^{\mathcal{R}}$ . For the completeness the proof relies on the completeness of the restriction of  $\Theta_d^{\mathcal{R}}$  to first order logic clauses. This has been shown in [126]. Since every possible most general (see section 7.2.2 for a formal definition of most general literals) temporal literal is added to the set of refinements (lines 17-22) the completeness carries over to PROLOG(+T)-rules. 

#### 7.2.2. Upward Refinement

As having done for PROLOG(+T)-literals we will now describe how to refine PRO-LOG(+T)-rules upward. We will see that the definition of an upward refinement operator for first order logic clauses does not need to be changed. This is due to the fact that upward refinement of rules is in some sense easier than downward refinement since an upward refinement operator does not need to capture all possible cases of temporal literals which might be added. Therefore the upward refinement operator  $\Theta_u^{\mathcal{R}}$  to be introduced is more tractable than the operator  $\Theta_d^{\mathcal{R}}$ . But before we need some more definitions.

#### Definition 7.2.1

Let sig =  $(\mathcal{X}, F, P, \alpha)$  be a signature, let  $\varphi \in \mathcal{L}_t(\text{sig})$  be a literal and let C be a PROLOG(+T)-rule over sig. Then  $\varphi$  is called *most general* with respect to C if  $TERMS(\varphi) = \{\mathbf{X}_1, \dots, \mathbf{X}_n\} \subseteq \mathcal{X}$  and 1.  $\varphi = \mathbf{p}(\mathbf{X}_1, \dots, \mathbf{X}_n)$  for some  $\mathbf{p} \in P$  with  $\alpha(\mathbf{p}) = n$  and  $\{\mathbf{X}_1, \dots, \mathbf{X}_n\} \cap VAR(C) = \emptyset$ 

or

1.	$Gp(\mathtt{a}) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}).$
2.	$\frac{Gp(a):=Fq(X_1,f(a)),q(a,a)Up(a)}{Gp(X_1):-Fq(X_1,f(a)),q(a,a)Up(a)}.$
3.	$\frac{Gp(II_1) \cdot Iq(II_1,I(a)),q(a,a)Op(a))}{Gp(f(\bar{X}_1)) :-Fq(f(\bar{X}_1),f(X_2)),q(a,a)Up(q).}$
4.	$\frac{Gp(X_1) :- Fq(X_1, f(f(\bar{X}_1))), q(a, a)Up(a).}{Gp(X_1) :- Fq(X_1, f(f(\bar{X}_1))), q(a, a)Up(a).}$
5.	Fr( <b>1</b> ,
6.	$\frac{F(X_1) - F(X_1, f(g(\bar{X}_1, \bar{X}_2)))}{Gp(X_1) - Fq(X_1, f(g(\bar{X}_1, \bar{X}_2))), q(a, a)Up(a).}$
7.	$Gp(X_1) := Fq(X_1, f(X_1)), q(a, a)Up(a).$
8.	$Gp(\mathtt{X}_2) := Fq(\mathtt{X}_2, \mathtt{f}(\mathtt{X}_2)), q(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}).$
9.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), q(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), not(p(\bar{\mathtt{X}}_1)).$
10.	$\mathtt{p}(\mathtt{X}_1):-\mathtt{not}(\mathtt{Gp}(\mathtt{X}_1)), \mathtt{Fq}(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)), \mathtt{g}(\mathtt{a},\mathtt{a}) \mathtt{Up}(\mathtt{a}).$
11.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), \mathtt{not}(\mathtt{q}(\bar{\mathtt{X}}_1, \bar{\mathtt{X}}_2)).$
12.	$q(\overline{\mathtt{X}}_1,\overline{\mathtt{X}}_2):-\mathtt{not}(Gp(\mathtt{X}_1)),Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),q(\mathtt{a},\mathtt{a})Up(\mathtt{a}).$
13.	$Gp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),q(\mathtt{a},\mathtt{a})Up(\mathtt{a}),p(\overline{\mathtt{X}}_1).$
14.	$Gp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),q(\mathtt{a},\mathtt{a})Up(\mathtt{a}),q(\bar{\mathtt{X}}_1,\bar{\mathtt{X}}_2).$
15.	$Gp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(Xnot(Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)))).$
16.	$Xnot(Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2))):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(Gp(\mathtt{X}_1)).$
17.	$Gp(\mathtt{X}_1):=Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}), \mathtt{not}(Xnot(\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}))).$
18.	$Xnot(q(\mathtt{a},\mathtt{a})Up(\mathtt{a})):=Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),q(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(Gp(\mathtt{X}_1)).$
19.	$Gp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(XGp(\mathtt{X}_1)).$
20.	$XGp(\mathtt{X}_1)):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(Gp(\mathtt{X}_1)).$
21.	$Gp(\mathtt{X}_1):=Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}), Xnot(Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2))).$
22.	$Gp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),Xnot(\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a})).$
23.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), XGp(\mathtt{X}_1).$
24.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), \mathtt{not}(Gnot(Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)))).$
25.	$Gnot(Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2))):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(Gp(\mathtt{X}_1)).$
26.	$Gp(X_1) := Fq(X_1, f(X_2)), q(a, a)Up(a), not(Gnot(q(a, a)Up(a))).$
27.	$Gnot(q(a, a)Up(a)) := Fq(X_1, f(X_2)), q(a, a)Up(a), not(Gp(X_1)).$
28.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), \mathtt{not}(GGp(\mathtt{X}_1)).$
29.	$GGp(X_1)) := Fq(X_1, f(X_2)), q(a, a) Up(a), not(Gp(X_1)).$
30.	$\frac{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), Gnot(Fq(X_1, f(X_2))).}{Gnot(r(X_1, f(X_2)), r(a, a)Up(a), Gnot(Fq(X_1, f(X_2))).}$
31.	$\frac{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), Gnot(q(a, a)Up(a)).}{Gnot(q(a, a), Up(a)) :-Fq(x, a) :Fq(x, a$
32. 33.	$\frac{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), GGp(X_1).}{Gp(X_1) : Fq(X_1, f(X_1)), q(a, a)Up(a), pot(Fpot(Fq(X_1, f(X_1))))}$
34.	$ \begin{array}{l} Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), \mathtt{not}(Fnot(Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)))). \\ \\ Fnot(Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2))) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), \mathtt{not}(Gp(\mathtt{X}_1)). \end{array} $
35.	$Fig(X_1) := Fq(X_1, f(X_2)) := fq(X_1, f(X_2)), q(a, a)Up(a), not(Fnot(q(a, a)Up(a))).$
36.	$\frac{Op(X_1) - Pq(X_1, Pq(X_2)), q(a, a) Op(a), Hot(Phot(q(a, a) Op(a))).}{Fnot(q(a, a) Up(a)) :- Fq(X_1, f(X_2)), q(a, a) Up(a), not(Gp(X_1)).}$
37.	$\frac{Fr_{A}(A_1,A_2)F(A_1)}{Gp(X_1):-Fq(X_1,f(X_2)),q(a,a)Up(a),not(FGp(X_1)).$
38.	$FGp(X_1): -Fq(X_1, f(X_2)), q(a, a) Up(a), not(Gp(X_1)).$
39.	$\frac{F_{P}(X_{1})}{F_{P}(X_{1})} = F_{q}(X_{1}, \mathfrak{f}(X_{2})), q(a, a) U_{P}(a), F_{OO}(F_{q}(X_{1}, \mathfrak{f}(X_{2}))).$
40.	$\frac{Gp(X_1) := Pq(X_1, f(X_2)), q(a, a) Op(a), Flot(q(a, a) Up(a)).}{Gp(X_1) := Fq(X_1, f(X_2)), q(a, a) Up(a), Flot(q(a, a) Up(a)).}$
-0.	-r(-1) = -1(-1) - (-2)/(-1) + (-2)/(-1)/(-1) + (-2)/(-1)/(-1) + (-2)/(-1)/(-1) + (-2)/(-1)/(-1) + (-2)/(-1)/(-1) + (-2)/(-1)/(-1)/(-1)/(-1)/(-1)/(-1)/(-1)/(-1

Table 7.5.: Set of downward refinements of  $C = \mathsf{Gp}(X_1) : -\mathsf{Fq}(X_1, f(X_2)), q(a, a)\mathsf{Up}(a).$  (part 1)

41.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a})Up(\mathtt{a}), FGp(\mathtt{X}_1).$
42.	$\frac{Gp(X_1) :- Fq(X_1, f(X_2)), q(a, a)Op(a), rot(Gp(X_1)UGp(X_1)).}{Gp(X_1) :- Fq(X_1, f(X_2)), q(a, a)Up(a), not(Gp(X_1)UGp(X_1)).}$
43.	$\frac{Gp(X_1) + q(X_1, q(X_2)), q(X, g) Gp(X_1) Gp(X_1) Gp(X_1)}{Gp(X_1) UGp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), not(Gp(X_1)).$
44.	$\frac{Gp(X_1) \circ Gp(X_1) + Fq(X_1, f(X_2)), q(a, a)Up(a), not(Gp(X_1)Unot(Fq(X_1, f(X_2))))}{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), not(Gp(X_1)Unot(Fq(X_1, f(X_2)))).$
45.	$\frac{Gp(X_1) + q(X_1, f(X_2))}{Gp(X_1)Unot(Fq(X_1, f(X_2))) :-Fq(X_1, f(X_2)), q(a, a)Up(a), not(Gp(X_1)).$
46.	$\frac{Gp(X_1) \circ Ir(V_1(A_1,I(X_2)),q(a,a)Up(a),not(Gp(X_1)Unot(q(a,a)Up(a))).}{Gp(X_1) :-Fq(X_1,f(X_2)),q(a,a)Up(a),not(Gp(X_1)Unot(q(a,a)Up(a))).}$
47.	$\frac{G_{P}(X_{1}) \cdots G_{q}(X_{1}) - (X_{2})}{G_{P}(X_{1}) - O(q(a, a) \cup p(a))} = -Fq(X_{1}, f(X_{2})), q(a, a) \cup p(a), not(Gp(X_{1})).$
48.	$\frac{G_{F}(X_1) \circ F(G_1) \circ F(G_2)}{G_{F}(X_1) \circ F(X_2), q(a, a) U_{F}(a), not(not(F_{q}(X_1, f(X_2))) U_{G}(G_{F}(X_1)).$
49.	$Gp(X_1) := Fq(X_1, f(X_2)), q(a, a)Up(a), not(not(Fq(X_1, f(X_2)))Unot(Fq(X_1, f(X_2)))).$
50.	$\frac{1}{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), not(not(Fq(X_1, f(X_2)))Unot(q(a, a)Up(a))).}{Unot(q(a, a)Up(a)), ut(not(rd(X_1, f(X_2)))Unot(q(a, a)Up(a))).}$
51.	$\frac{1}{\operatorname{Gp}(X_1):-\operatorname{Fq}(X_1,f(X_2)),q(a,a)\operatorname{Up}(a),\operatorname{not}(\operatorname{not}(q(a,a)\operatorname{Up}(a)\operatorname{UGp}(X_1)).}$
52.	$ \frac{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), not(not(q(a, a)Up(a))Unot(Fq(X_1, f(X_2)))). }{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), not(not(q(a, a)Up(a))Unot(Fq(X_1, f(X_2)))). } $
53.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), \mathtt{not}(\mathtt{not}(\mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a})) Unot(\mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}))).$
54.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), q(\mathtt{a}, \mathtt{a})Up(\mathtt{a}), Gp(\mathtt{X}_1)UGp(\mathtt{X}_1).$
55.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), Gp(\mathtt{X}_1) Unot(Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2))).$
56.	$Gp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),Gp(\mathtt{X}_1)Unot(\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a})).$
57.	$Gp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)))UGp(\mathtt{X}_1).$
58.	$Gp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)))Unot(Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2))).$
59.	$Gp(\mathtt{X}_1):=Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}), \mathtt{not}(Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)))Unot(\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a})).$
60.	$Gp(\mathtt{X}_1):=Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a},\mathtt{a}) Up(\mathtt{a}), \mathtt{not}(\mathtt{q}(\mathtt{a},\mathtt{a}) Up(\mathtt{a}) UGp(\mathtt{X}_1).$
61.	$Gp(\mathtt{X}_1):=Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}), \mathtt{not}(\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}))Unot(Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2))).$
62.	$Gp(\mathtt{X}_1):=Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a},\mathtt{a}) Up(\mathtt{a}), \mathtt{not}(\mathtt{q}(\mathtt{a},\mathtt{a}) Up(\mathtt{a})) Unot(\mathtt{q}(\mathtt{a},\mathtt{a}) Up(\mathtt{a})).$
63.	$Gp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(Gp(\mathtt{X}_1)PGp(\mathtt{X}_1)).$
64.	$Gp(\mathtt{X}_1)PGp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(Gp(\mathtt{X}_1)).$
65.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), \mathtt{not}(Gp(\mathtt{X}_1)Pnot(Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)))).$
66.	$Gp(\mathtt{X}_1)Pnot(Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2))):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(Gp(\mathtt{X}_1)).$
67.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), \mathtt{not}(Gp(\mathtt{X}_1) Pnot(\mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}))).$
68.	$Gp(\mathtt{X}_1)Pnot(q(\mathtt{a},\mathtt{a})Up(\mathtt{a})):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),q(\mathtt{a},\mathtt{a})Up(\mathtt{a}),not(Gp(\mathtt{X}_1)).$
<u>69</u> .	$Gp(X_1) := Fq(X_1, \mathfrak{f}(X_2)), q(a, a)Up(a), not(not(Fq(X_1, \mathfrak{f}(X_2)))PGp(X_1)).$
70.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), q(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), not(not(Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2))) Pnot(Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)))).$
71.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), q(\mathtt{a}, \mathtt{a})Up(\mathtt{a}), not(not(Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)))Pnot(q(\mathtt{a}, \mathtt{a})Up(\mathtt{a}))).$
72.	$\frac{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), not(not(q(a, a)Up(a)PGp(X_1)).}{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), not(not(q(a, a)Up(a)))$
73.	$\frac{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), not(not(q(a, a)Up(a))Pnot(Fq(X_1, f(X_2)))).}{Gp(X_1) :-Fq(X_1, f(X_2)) :-Fq(X_1, f($
74.	$\frac{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), not(not(q(a, a)Up(a))Pnot(q(a, a)Up(a))).}{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), not(not(q(a, a)Up(a)))Pnot(q(a, a)Up(a))).}$
75. 76.	$\frac{Gp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), Gp(X_1)PGp(X_1).}{Cp(X_1) :-Fq(X_1, f(X_2)), q(a, a)Up(a), Cp(X_1)PGp(X_1).}$
70. 77.	$\frac{Gp(\mathtt{X}_1) :- Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), Gp(\mathtt{X}_1) Pnot(Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2))).}{Gp(\mathtt{X}_1) :- Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), Gp(\mathtt{X}_1) Pnot(\mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a})).}$
78.	$\frac{Gp(X_1) :- Fq(X_1, f(X_2)), q(a, a)Op(a), Gp(X_1) Fnot(q(a, a)Op(a)).}{Gp(X_1) :- Fq(X_1, f(X_2)), q(a, a)Up(a), not(Fq(X_1, f(X_2)))PGp(X_1).}$
78. 79.	$\frac{Gp(X_1) := Pq(X_1, f(X_2)), q(a, a)Op(a), not(Pq(X_1, f(X_2)))Pop(X_1).}{Gp(X_1) := Fq(X_1, f(X_2)), q(a, a)Up(a), not(Fq(X_1, f(X_2)))Poot(Fq(X_1, f(X_2))).}$
80.	$\frac{Gp(X_1) :- Fq(X_1, f(X_2)), q(a, a)Op(a), not(Fq(X_1, f(X_2)))Fnot(Fq(X_1, f(X_2)))}{Gp(X_1) :- Fq(X_1, f(X_2)), q(a, a)Up(a), not(Fq(X_1, f(X_2)))Pnot(q(a, a)Up(a)).}$
00.	$OP(\mathbf{x}_1) \cdot OP(\mathbf{x}_1, I(\mathbf{x}_2)), q(\mathbf{a}, \mathbf{a}) OP(\mathbf{a}), mod(I(q(\mathbf{x}_1, I(\mathbf{x}_2)))) mod(q(\mathbf{a}, \mathbf{a}) OP(\mathbf{a})).$

Table 7.7.: Set of downward refinements of  $C = \mathsf{Gp}(X_1) : -\mathsf{Fq}(X_1, \mathtt{f}(X_2)), \mathtt{q}(\mathtt{a}, \mathtt{a})\mathsf{Up}(\mathtt{a}).$  (part 1)

**Input**: PROLOG(+T)-rule  $C = \varphi := \psi_1, \ldots, \psi_n$ . **Output**: set of specialized rules **Require:**  $C = \varphi := \psi_1, \ldots, \psi_n$ . 1:  $S_C \leftarrow \{\varphi, \neg \psi_1, \ldots, \neg \psi_n\}$ 2:  $Ref \leftarrow \emptyset$  {set of sets of literals} **Require:**  $S_C = \{\chi_0, \ldots, \chi_n\}, \text{ VAR}(C) = \{\mathtt{X}_1, \ldots, \mathtt{X}_k\}, \text{ sig} = (\mathcal{X}, F, P, \alpha)$ **Require:**  $F = \{f_1, ..., f_{|F|}\}$ **Require:**  $\overline{X}_1, \overline{X}_2, \ldots, \overline{X}_i, \ldots$  new variables, pairwise distinct 3: for i = 1, ..., k do for j = 1, ..., |F| do 4:  $Ref \leftarrow Ref \cup \left\{ S_C \left\{ \frac{\mathbf{x}_i}{\mathbf{f}_j \left( \overline{\mathbf{x}}_1, \dots, \overline{\mathbf{x}}_{\alpha(\mathbf{f}_i)} \right)} \right\} \right\}$ 5:end for 6: 7: end for 8: for i = 1, ..., k do for  $j = 1, \ldots, k$  do 9: 10:if  $i \neq j$  then  $Ref \leftarrow Ref \cup \left\{ S_C \left\{ \frac{\mathbf{x}_i}{\mathbf{x}_i} \right\} \right\}$ 11: 12:end if end for 13:14: **end for Require:**  $P = \{p_1, \ldots, p_m\}$ 15: for i = 1, ..., m do  $Ref \leftarrow Ref \cup \left\{ S_C \cup \left\{ \mathtt{p}_i\left(\overline{\mathtt{X}}_1, \dots, \overline{\mathtt{X}}_{\alpha(\mathtt{p}_i)}\right) \right\} \right\} \cup \left\{ S_C \cup \left\{ \mathtt{not}\left( \mathtt{p}_i\left(\overline{\mathtt{X}}_1, \dots, \overline{\mathtt{X}}_{\alpha(\mathtt{p}_i)}\right) \right) \right\} \right\}$ 16: $Ref \leftarrow Ref \cup \{ \mathsf{X}\psi \mid \psi \in S_C \} \cup \{ \mathsf{not}(\mathsf{X}\psi) \mid \psi \in S_C \}$ 17:18: $Ref \leftarrow Ref \cup \{\mathsf{G}\psi \mid \psi \in S_C\} \cup \{\mathsf{not}(\mathsf{G}\psi) \mid \psi \in S_C\}$  $Ref \leftarrow Ref \cup \{\mathsf{F}\psi \mid \psi \in S_C\} \cup \{\mathsf{not}(\mathsf{F}\psi) \mid \psi \in S_C\}$ 19:20: $Ref \leftarrow Ref \cup \{\psi_1 \mathsf{U}\psi_2 \mid \psi_1, \psi_2 \in S_C\} \cup \{\mathsf{not}(\psi_1 \mathsf{U}\psi_2) \mid \psi_1, \psi_2 \in S_C\}$ 21: $Ref \leftarrow Ref \cup \{\psi_1 \mathsf{P}\psi_2 \mid \psi_1, \psi_2 \in S_C\} \cup \{\mathsf{not}(\psi_1 \mathsf{P}\psi_2) \mid \psi_1, \psi_2 \in S_C\}$ 22: end for **Require:**  $Ref = \{S_1, ..., S_o\}$ 23:  $R \leftarrow \emptyset$  {set of rules} 24: for i = 1, ..., o do **Require:**  $S_i = \left\{ l_1^{(i)}, \dots, l_{n_i}^{(i)} \right\}$ for  $j = 1, ..., n_i$  do 25:26: if  $l_i$  is positive then 27:28:end if end for 29:30: end for 31: return R

81.	$Gp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a})PGp(\mathtt{X}_1).$
82.	$Gp(\mathtt{X}_1) := Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a}), \mathtt{not}(\mathtt{q}(\mathtt{a}, \mathtt{a}) Up(\mathtt{a})) Pnot(Fq(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2))).$
83.	$Gp(\mathtt{X}_1):-Fq(\mathtt{X}_1,\mathtt{f}(\mathtt{X}_2)),\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}),\mathtt{not}(\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a}))Pnot(\mathtt{q}(\mathtt{a},\mathtt{a})Up(\mathtt{a})).$

Table 7.9.: Set of downward refinements of  $C = \mathsf{Gp}(X_1) : -\mathsf{Fq}(X_1, \mathsf{f}(X_2)), \mathsf{q}(\mathsf{a}, \mathsf{a})\mathsf{Up}(\mathsf{a}).$ (part 3)

- 2.  $\varphi = \operatorname{not}(p(X_1, \dots, X_n))$  for some  $p \in P$  with  $\alpha(p) = n$  and  $\{X_1, \dots, X_n\} \cap$
- $\operatorname{VAR}(C) = \emptyset$  or 3.  $\varphi = \bigoplus \psi$  for  $\bigoplus \in \{X, G, F\}$  and some most general literal  $\psi$  or 4.  $\varphi = \psi_1 \oplus \psi_2$  for  $\oplus \in \{U, P\}$  and most general literals  $\psi_1, \psi_2$  such that  $\operatorname{VAR}(\psi_1) \cap \operatorname{VAR}(\psi_2) = \emptyset$ .

It will be necessary to consider rules which may contain literals more than once. In this case, treating rules as sets of literals is not adequate. Therefore we will introduce sequences of literals which are computed by *duplicating* certain literals from the underlying set.

# Definition 7.2.2 (Duplication of Literals, e.g. [126])

Let  $S_C = \{\varphi_1, \ldots, \varphi_n\}$  be a set of PROLOG(+T)-literals and let  $t \in TERMS(S_C)$  be a term. Then the sequence  $dup(S_C, t)$  is defined as

$$dup(S_C, t) = (\underbrace{\varphi_1, \dots, \varphi_1}_{2^{|\operatorname{Occ}(t,\varphi_1)|} \text{ times}}, \dots, \underbrace{\varphi_n, \dots, \varphi_n}_{2^{|\operatorname{Occ}(t,\varphi_n)|} \text{ times}}).$$

Positions of such tuples of literals are defined as expected:

$$\operatorname{Pos}(\varphi_1,\ldots,\varphi_n) = \bigcup_{i=1}^n \{ip \mid p \in \operatorname{Pos}(\varphi_i)\}.$$

Similarly for  $p_1, \ldots, p_k \in Pos(\varphi_1, \ldots, \varphi_n)$  and  $t \in \mathcal{T}(sig)$  we define

$$(\varphi_1,\ldots,\varphi_n)[t]_{p_1,\ldots,p_k} = \left(\varphi_1[t]_{p_{i_0}^{(1)},\ldots,p_{i_{m_1}}^{(1)}},\ldots,\varphi_n[t]_{p_{i_0}^{(n)},\ldots,p_{i_{m_n}}^{(n)}}\right),$$

where for each j the set  $\left\{ jp_{i_0^{(j)}}^{(j)}, \dots, jp_{i_{m_j}^{(j)}}^{(j)} \right\} \subseteq \operatorname{Pos}(\varphi_1, \dots, \varphi_n)$  is maximal.

Finally for a sequence  $(\varphi_1, \ldots, \varphi_n)$  consisting of n (not necessarily distinct) literals from  $\mathcal{L}_t(\text{sig})$  we define

SET
$$(\varphi_1, \ldots, \varphi_n) = \bigcup_{i=1}^n \{\varphi_i\}.$$

Using these definitions the extension of the refinement operator for first order logic clauses to PROLOG(+T)-rules is straightforward. As in the case of downward refinement we will work on sets of literals instead of rules.

$$\begin{split} \Theta_{u}^{\mathcal{R}}(S_{C}) &= \begin{cases} t \in \mathrm{TERMS}(C), t = \mathbf{f} \left( \mathbf{X}_{1}, \dots, \mathbf{X}_{\alpha(\mathbf{f})} \right) \text{ simple} \\ S_{C}[\mathbf{Z}]_{p_{1},\dots,p_{k}} \mid \{p_{1},\dots,p_{k}\} = \mathrm{OCC}(t,C) \text{ and every } p \in \mathrm{OCC}(\mathbf{X}_{i},C) \\ & \text{ is inside of one } p_{i} \text{ for } i = 1,\dots,\alpha(\mathbf{f}), \mathbf{Z} \notin \mathrm{VAR}(C) \end{cases} \\ \cup & \left\{ \mathrm{SET}(dup(S_{C},\mathbf{a})[\mathbf{Z}]_{p_{1},\dots,p_{k}}) \mid \begin{array}{c} \mathbf{a} \in F, \alpha(\mathbf{a}) = 0, \mathbf{Z} \notin \mathrm{VAR}(C) \\ \emptyset \neq \{p_{1},\dots,p_{k}\} \subseteq \mathrm{OCC}(\mathbf{a},dup(S_{C},\mathbf{a})) \end{cases} \right\} \\ \cup & \left\{ \mathrm{SET}(dup(S_{C},\mathbf{X})[\mathbf{Z}]_{p_{1},\dots,p_{k}}) \mid \begin{array}{c} \mathbf{X} \in \mathrm{VAR}(C), \mathbf{Z} \notin \mathrm{VAR}(C) \\ \emptyset \neq \{p_{1},\dots,p_{k}\} \subset \mathrm{OCC}(\mathbf{X},dup(S_{C},\mathbf{X})), \end{array} \right\} \\ \cup & \left\{ S_{C} \setminus \{\varphi\} \mid \varphi \in S_{C} \text{ is most general wrt. } C \right\}. \end{split}$$

Given C we can therefore compute  $S_C$  and then apply  $\Theta_u^{\mathcal{R}}$  to C. From the resulting set the extraction of a set of generalized rules is then carried out as usual. The procedure for computing the set of generalized rules is summarized in Algorithm 10.

#### Theorem 7.2.2

 $\Theta_u^{\mathcal{R}}$  is locally finite and complete.

Algorithm 10 Upward–Refinement of PROLOG(+T)–rules

**Input**: PROLOG(+T)-rule  $C = \varphi := \psi_1, \ldots, \psi_n$ . Output: set of generalized rules **Require:**  $C = \varphi := \psi_1, \ldots, \psi_n$ . 1:  $S_C \leftarrow \{\varphi, \neg \psi_1, \ldots, \neg \psi_n\}$ 2:  $Ref \leftarrow \emptyset$  {set of sets of literals} **Require:**  $S_C = \{\chi_0, \ldots, \chi_n\}, \ \mathsf{Z} \notin \operatorname{Var}(C)$ **Require:**  $F = \{f_1, ..., f_{|F|}\}, \text{TERMS}(C) = \{t_1, ..., t_m\}$ 3: for i = 1, ..., m do 4: if  $t_i$  is simple then **Require:**  $\operatorname{VaR}(t_i) = \{\mathsf{Z}_1, \ldots, \mathsf{Z}_l\}$ 5:for o = 1, ..., l do if each  $p \in OCC(X_o, C)$  is inside a  $p' \in OCC(t_i, C)$  then 6:**Require:**  $\{p_1,\ldots,p_k\}$ OCC $(t_i,C)$ 7:  $Ref \leftarrow Ref \cup \{S_C[\mathbf{Z}]_{p_1,\ldots,p_k}\}$ end if 8: 9: end for 10:end if 11: end for 12: for i = 1, ..., |F| do if  $\alpha(\mathbf{f}_i) = 0$  then 13:for each  $\emptyset \neq \{p_1, \ldots, p_k\} \subseteq OCC(\mathbf{f}_i, dup(S_C, \mathbf{f}_i))$  do 14: $Ref \leftarrow Ref \cup \{ SET(dup(S_C, f_i)[Z]_{p_1, \dots, p_k}) \}$ 15:16:end for 17:end if 18: end for **Require:**  $VAR(C) = \{X_1, \ldots, X_l\}$ 19: for i = 1, ..., l do 20:for each  $\emptyset \neq \{p_1, \ldots, p_k\} \subset OCC(X_i, dup(S_C, X_i))$  do 21: $Ref \leftarrow Ref \cup \{ SET(dup(S_C, X_i)[Z]_{p_1, \dots, p_k}) \}$ 22:end for 23: end for 24: for i = 0, ..., n do 25:if  $\chi_i$  is most general wrt. C then  $Ref \leftarrow Ref \cup \{S_C \setminus \{\chi_i\}\}$ 26:end if 27:28: end for **Require:**  $Ref = \{S_1, ..., S_{|S|}\}$ 29:  $R \leftarrow \emptyset$  {set of rules} 30: for i = 1, ..., |S| do **Require:**  $S_i = \left\{ l_1^{(i)}, \dots, l_{n_i}^{(i)} \right\}$ 31: for  $j = 1, \dots, n_i$  do 32: if  $l_i$  is positive then  $R \leftarrow R \cup \{l_j : -\mathrm{TAIL}(S_i \setminus \{l_j\}).\}$ 33: 34:end if end for 35:36: end for 37: return R

**Proof.** As for the operator  $\Theta_d^{\mathcal{R}}$ , locally finiteness is immediately clear from the definition of  $\Theta_u^{\mathcal{R}}$ . For the completeness the argumentation from [126] can be directly adapted to  $\Theta_u^{\mathcal{R}}$ . So the theorem is proved.

## 7.3. Refinement Strategies

Up to now we have defined operators for refining PROLOG(+T)-literals and -rules both upward and downward. However, it is not yet clear how these operators should be applied. For example, exhaustive application of  $\Theta_d^{\mathcal{R}}$  to any clause is not applicable as we have seen in Example 7.2.1. The size of the set computed by Algorithm 9 is given as stated in the following lemma<sup>2</sup>.

#### Lemma 7.3.1

Let C be a PROLOG(+T)-rule built over sig =  $(\mathcal{X}, F, P, \alpha)$ . Then

$$|\Theta_d^{\mathcal{R}}(C)| \le |\operatorname{Var}(C)| (|F| + |\operatorname{Var}(C)| - 1) + 6|S_C| + 4|S_C|^2 + 2|P|.$$

**Proof**. We have

$$\begin{aligned} |\Theta_{d}^{\mathcal{R}}(C)| &\leq \left| \begin{cases} \mathbf{Z} \in \operatorname{VAR}(C), \mathbf{f} \in F, \alpha(\mathbf{f}) = l, \\ S_{C} \left\{ \frac{\mathbf{Z}}{\mathbf{f}(\mathbf{X}_{1}, \dots, \mathbf{X}_{l})} \right\} \mid \mathbf{X}_{1}, \dots, \mathbf{X}_{l} \notin \operatorname{VAR}(C), \\ \mathbf{X}_{i} \neq \mathbf{X}_{j} \text{ for } i \neq j \end{cases} \right\} \\ &+ \left| \left\{ S_{C} \left\{ \frac{\mathbf{Z}}{\mathbf{X}} \right\} \mid \mathbf{X}, \mathbf{Z} \in \operatorname{VAR}(C), \mathbf{X} \neq \mathbf{Z} \right\} \right| \\ &+ \left| \begin{cases} S_{C} \cup \left\{ \mathbf{p}(\mathbf{X}_{1}, \dots, \mathbf{X}_{l}) \right\} \mid \mathbf{p} \in P, \alpha(\mathbf{p}) = l, \mathbf{X}_{1}, \dots, \mathbf{X}_{l} \notin \operatorname{VAR}(C) \\ \mathbf{X}_{i} \neq \mathbf{X}_{j} \text{ for } i \neq j \end{cases} \right\} \right| \\ &+ \left| \begin{cases} S_{C} \cup \left\{ \operatorname{not}(\mathbf{p}(\mathbf{X}_{1}, \dots, \mathbf{X}_{l})) \right\} \mid \mathbf{p} \in P, \alpha(\mathbf{p}) = l, \mathbf{X}_{1}, \dots, \mathbf{X}_{l} \notin \operatorname{VAR}(C) \\ \mathbf{X}_{i} \neq \mathbf{X}_{j} \text{ for } i \neq j \end{cases} \right\} \right| \end{aligned}$$

<sup>2</sup>We will restrict ourselves on the case of downward refinement since upward refinement often yields smaller sets of rules.

$$\begin{split} &+ |\{S_C \cup \{\mathsf{X}\psi\} \mid \psi \in S_C\}| + |\{S_C \cup \{\mathsf{not}(\mathsf{X}\psi)\} \mid \psi \in S_C\}| \\ &+ |\{S_C \cup \{\mathsf{G}\psi\} \mid \psi \in S_C\}| + |\{S_C \cup \{\mathsf{not}(\mathsf{G}\psi)\} \mid \psi \in S_C\}| \\ &+ |\{S_C \cup \{\mathsf{F}\psi\} \mid \psi \in S_C\}| + |\{S_C \cup \{\mathsf{not}(\mathsf{F}\psi)\} \mid \psi \in S_C\}| \\ &+ |\{S_C \cup \{\psi_1 \cup \psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \cup \psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup \{\neg \psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| \\ &+ |\{S_C \cup$$

by definition of  $\Theta_d^{\mathcal{R}}$ . Furthermore

• 
$$\left| \begin{cases} Z \in \operatorname{VAR}(C), \mathbf{f} \in F, \alpha(\mathbf{f}) = l, \\ S_C \left\{ \frac{Z}{\mathbf{f}(\mathbf{X}_1, \dots, \mathbf{X}_l)} \right\} \mid \begin{array}{l} \mathbf{X}_1, \dots, \mathbf{X}_l \notin \operatorname{VAR}(C), \\ \mathbf{X}_i \neq \mathbf{X}_j \text{ for } i \neq j \end{cases} \right\} \right| \leq |\operatorname{VAR}(C)| \cdot |F|, \\ \left| \left\{ S_C \left\{ \frac{Z}{\mathbf{X}} \right\} \mid \mathbf{X}, \mathbf{Z} \in \operatorname{VAR}(C), \mathbf{X} \neq \mathbf{Z} \right\} \right| \leq |\operatorname{VAR}(C)| \cdot (|\operatorname{VAR}(C)| - 1), \\ \left| \left\{ S_C \cup \left\{ \mathbf{p}(\mathbf{X}_1, \dots, \mathbf{X}_l) \right\} \mid \begin{array}{l} \mathbf{p} \in P, \alpha(\mathbf{p}) = l, \mathbf{X}_1, \dots, \mathbf{X}_l \notin \operatorname{VAR}(C) \\ \mathbf{X}_i \neq \mathbf{X}_j \text{ for } i \neq j \end{cases} \right\} \right| \leq |P| \text{ and} \\ \left| \left\{ S_C \cup \left\{ \operatorname{not}(\mathbf{p}(\mathbf{X}_1, \dots, \mathbf{X}_l)) \right\} \mid \begin{array}{l} \mathbf{p} \in P, \alpha(\mathbf{p}) = l, \mathbf{X}_1, \dots, \mathbf{X}_l \notin \operatorname{VAR}(C) \\ \mathbf{X}_i \neq \mathbf{X}_j \text{ for } i \neq j \end{cases} \right\} \right| \leq |P|, \\ \\ \end{bmatrix}$$

 $|\{S_C \cup \{\mathsf{X}\psi\} \mid \psi \in S_C\}| \leq |S_C|,$ 

$$\begin{split} |\{S_C \cup \{\operatorname{not}(\mathsf{X}\psi)\} \mid \psi \in S_C\}| &\leq |S_C|, \\ |\{S_C \cup \{\mathsf{G}\psi\} \mid \psi \in S_C\}| &\leq |S_C|, \\ |\{S_C \cup \{\operatorname{not}(\mathsf{G}\psi)\} \mid \psi \in S_C\}| &\leq |S_C|, \\ |\{S_C \cup \{\mathsf{F}\psi\} \mid \psi \in S_C\}| &\leq |S_C| \text{ and } \\ |\{S_C \cup \{\operatorname{not}(\mathsf{F}\psi)\} \mid \psi \in S_C\}| &\leq |S_C|, \end{split}$$

$$\begin{split} |\{S_C \cup \{\psi_1 \cup \psi_2\} \mid \psi_1, \psi_2 \in S_C\}| &\leq |S_C|^2, \\ |\{S_C \cup \{\mathsf{not}(\psi_1 \cup \psi_2)\} \mid \psi_1, \psi_2 \in S_C\}| &\leq |S_C|^2, \\ |\{S_C \cup \{\psi_1 \mathsf{P}\psi_2\} \mid \psi_1, \psi_2 \in S_C\}| &\leq |S_C|^2 \text{ and} \\ |\{S_C \cup \{\mathsf{not}(\psi_1 \mathsf{P}\psi_2)\} \mid \psi_1, \psi_2 \in S_C\}| &\leq |S_C|^2. \end{split}$$

Combining these inequalities we have

$$\begin{aligned} |\Theta_d^{\mathcal{R}}(C)| &\leq |\operatorname{VAR}(C)| \cdot |F| + |\operatorname{VAR}(C)| \cdot (|\operatorname{VAR}(C)| - 1) + 6|S_C| + 4|S_C|^2 + 2|P| \\ &= |\operatorname{VAR}(C)| \cdot (|F| + |\operatorname{VAR}(C)| - 1) + 6|S_C| + 4|S_C|^2 + 2|P| \end{aligned}$$

as claimed.

•

We will now see how refinement steps can be carried out *without* constructing the maximum number of rules.

#### 7.3.1. Elimination of Variants

Recall that two PROLOG(+T)-rules  $C_1$  and  $C_2$  are variants if there are substitutions  $\theta_1$ and  $\theta_2$  such that  $\theta_1(C_1) = C_2$  and  $\theta_2(C_2) = C_1$ . Since in our setting all rules in programs are considered universally closed, only one (of possibly arbitrary many) variants must be included in the set of refined rules. In the example from the last section, one of the rules 7 and 8 may be dropped.

In general, the presence of variants in the set of refinements is due to the construction of the loop from line 8 to line 14 in Algorithm 9. If we change the j-loop to range from i+1 to k and drop the **if**-condition (which now has no effect), the variants are not added.

In the original algorithm the number of rules added in the loop is given as  $k^2 - k$ . If the modification is added, only  $\sum_{i=1}^{k} \sum_{j=i+1}^{k} 1$  rules are added. So the difference in the size of the *original* set of refined rules and the *modified* set is given as

$$k^{2} - k - \left(k^{2} - \frac{k(k+1)}{2}\right) = \frac{k(k+1)}{2} - k$$
$$= \frac{k^{2} + k - 2k}{2}$$
$$= \frac{k^{2} - k}{2}.$$

Setting k = |VAR(C)| and using Lemma 7.3.1 we have

$$\begin{aligned} \left| \Theta_{d}^{\mathcal{R}}(C) \right| &\leq |\operatorname{VAR}(C)| \cdot (|F| + |\operatorname{VAR}(C)| - 1) + 6|S_{C}| + 4|S_{C}|^{2} \\ &+ 2|P| - \frac{|\operatorname{VAR}(C)|^{2} - |\operatorname{VAR}(C)|}{2} \\ &= |\operatorname{VAR}(C)| \cdot \left( |F| + \frac{|\operatorname{VAR}(C)|}{2} - \frac{1}{2} \right) + 6|S_{C}| + 4|S_{C}|^{2} + 2|P| \end{aligned}$$

where  $\Theta_d^{\mathcal{R}}$  from now on denotes the refinement operator which implements the above strategy.

The growth of the set of refinements (depending on the number of variables in the rule

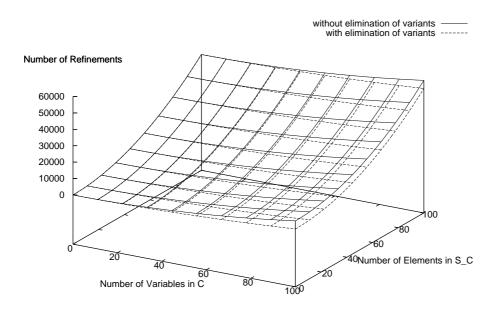


Figure 7.1.: Growth rate of refined rules with and without elimination of variants

to be refined and the size of the original rule) is depicted in Figure 7.1 using |F| = 3 and |P| = 2.

#### 7.3.2. Restriction to reduced Rules

We have introduced the concept of *reduced literals* in order to keep the representation of a literal canonical, so that we can assume that each literal from  $\mathcal{L}_t(\text{sig})$  has a certain form. Similarly a rule is reduced if every literal in this rule is reduced. Restricting ourselves to the construction of reduced rules during the refinement of PROLOG(+T)rules guarantees canonicity.

#### Example 7.3.1

Again consider rule

$$C = \mathsf{Gp}(\mathtt{X}_1) :- \mathsf{Fq}(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) \mathsf{Up}(\mathtt{a}).$$

Then the set of refinements contains among others the rules

$$C_1 = \mathsf{Gp}(X_1) := \mathsf{Fq}(X_1, \mathtt{f}(X_2)), \mathtt{q}(\mathtt{a}, \mathtt{a})\mathsf{Up}(\mathtt{a}), \mathtt{not}(\mathsf{GGp}(X_1)). \text{ and}$$
$$C_2 = \mathsf{GGp}(X_1) := \mathsf{Fq}(X_1, \mathtt{f}(X_2)), \mathtt{q}(\mathtt{a}, \mathtt{a})\mathsf{Up}(\mathtt{a}), \mathtt{not}(\mathsf{Gp}(X_1)).$$

These rules are syntactically different. However, reduction of the literals involved yields:

$$\begin{aligned} \operatorname{RED}(C_1) &= \operatorname{\mathsf{Gp}}(\mathtt{X}_1) :- \operatorname{\mathsf{Fq}}(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) \operatorname{\mathsf{Up}}(\mathtt{a}), \mathtt{not}(\operatorname{\mathsf{Gp}}(\mathtt{X}_1)). \text{ and} \\ \\ \operatorname{RED}(C_2) &= \operatorname{\mathsf{Gp}}(\mathtt{X}_1) :- \operatorname{\mathsf{Fq}}(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) \operatorname{\mathsf{Up}}(\mathtt{a}), \mathtt{not}(\operatorname{\mathsf{Gp}}(\mathtt{X}_1)). \end{aligned}$$

Since  $\operatorname{Red}(C_1) = \operatorname{Red}(C_2)$  one of the refinement steps can be skipped.

But the example from above yields even more possible improvements: both rules can be skipped. This is simply due to

$$\begin{aligned} \operatorname{RED}(C_1) &= \operatorname{Gp}(X_1) := \operatorname{Fq}(X_1, \operatorname{f}(X_2)), \operatorname{q}(a, a) \operatorname{Up}(a), \operatorname{not}(\operatorname{Gp}(X_1)). \\ &\equiv \operatorname{Gp}(X_1) \lor \operatorname{not}(\operatorname{Fq}(X_1, \operatorname{f}(X_2))) \lor \operatorname{not}(\operatorname{q}(a, a)) \operatorname{Up}(a) \lor \operatorname{not}(\operatorname{not}(\operatorname{Gp}(X_1))) \\ &\equiv \operatorname{Gp}(X_1) \lor \operatorname{not}(\operatorname{Fq}(X_1, \operatorname{f}(X_2))) \lor \operatorname{q}(a, a) \operatorname{Up}(a) \\ &\equiv \operatorname{Gp}(X_1) := \operatorname{Fq}(X_1, \operatorname{f}(X_2)), \operatorname{q}(a, a) \operatorname{Up}(a). \\ &= C. \end{aligned}$$

However, estimating the number of rules which can be skipped by restriction to reduced literals depends on the structure of the literals involved in the refined rule. So in general we are not able to give an estimation of the reduction of  $|\Theta_d^{\mathcal{R}}(C)|$ .

#### 7.3.3. Elimination of Tautologies

Tautologies can be considered as rules which do not have any effect on the provability resp. non-provability of goals. Therefore they do not have to be constructed. A sufficient criterion is stated in the following lemma.

#### Lemma 7.3.2

Let C be any PROLOG(+T)-rule built over the signature sig. If there is  $\varphi \in \mathcal{L}_t(sig)$  such that  $\{\varphi, \mathsf{not}(\varphi)\} \subseteq S_C$ , then C is a tautology.

**Proof**. immediately from

$$C \equiv S_C$$

$$= \{\varphi, \operatorname{not}(\varphi), \psi_1, \dots, \psi_k\} \text{ for } \varphi, \psi_1, \dots, \psi_k \in \mathcal{L}_t(\operatorname{sig})$$

$$\equiv \varphi \lor \operatorname{not}(\varphi) \lor \psi_1 \lor \dots \lor \psi_k$$

$$\equiv \operatorname{true} \lor \psi_1 \dots \lor \psi_k$$

$$\equiv \operatorname{true.}$$

#### Example 7.3.2

Let the rule

$$C = \mathsf{Gp}(\mathtt{X}) :- \mathtt{p}(\mathtt{X}), \mathsf{Fq}(\mathtt{X}, \mathtt{X}).$$

be given. Then  $\{\mathsf{Gp}(X) := p(X), \mathsf{Fq}(X, X), \mathsf{Gp}(X)\} \subseteq \Theta_d^{\mathcal{R}}(C)$  which is a tautology and can therefore be skipped.

In general, testing a rule C for being a tautology using the approach sketched above can be accomplished in time  $\mathcal{O}(|S_C|^2)$ . But similarly as in the foregoing section we cannot estimate the number of rules which might be skipped by tautology-elimination without taking the structure of C into account.

#### 7.3.4. Premises vs. Conclusions

The downward refinement operator  $\Theta_d^{\mathcal{R}}$  adds positive as well as negative literals to the set of refinements. Of course, every rule which is generated in this way is a downward refinement of the original rule. But one can rely on the following point of view: the conclusions of rules should be known in advance, therefore it is better to adjust the premises.

In the definition of  $\Theta_d^{\mathcal{R}}$  this is modeled by only adding negative literals during the execution of the loop in lines 15–22. The number of literals added by only adding negative literals is then

$$|P| + 3|S_C| + 2|S_C|^2$$
,

so the overall number of refinements is

$$\left|\Theta_{d}^{\mathcal{R}}(C)\right| \le \left|\operatorname{Var}(C)\right| \cdot \left(|F| + \frac{\left|\operatorname{Var}(C)\right|}{2} - \frac{1}{2}\right) + 3|S_{C}| + 2|S_{C}|^{2} + |P|$$

Figure 7.2 compares the introduced strategy with the strategy of elimination of variants described above.

However, by adding negative literals only we lose the completeness of the operator. This can be easily seen as follows: consider the rules

$$\begin{array}{lll} C_1 &=& \mathsf{Gp}(\mathtt{X}_1) :- \mathsf{Fq}(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) \mathsf{Up}(\mathtt{a}). \mbox{ and} \\ \\ C_2 &=& \mathsf{Gp}(\mathtt{X}_1) :- \mathsf{Fq}(\mathtt{X}_1, \mathtt{f}(\mathtt{X}_2)), \mathtt{q}(\mathtt{a}, \mathtt{a}) \mathsf{Up}(\mathtt{a}), \mathtt{not}(\mathtt{q}(\mathtt{X}_1, \mathtt{X}_1)) \end{array}$$

Since only negative literals are added in the process of refining, every rule C contained in  $(\Theta_d^{\mathcal{R}})^n(C)$  for any n only may contain other *positive* premises. So no rule  $C' \approx_s C_2$ will be constructed.

The above results regarding techniques for the restriction of the number of refinements complete our study of refinement operators for PROLOG(+T)-objects. We have seen that refinement operators for nontemporal logic programming languages can be natu-

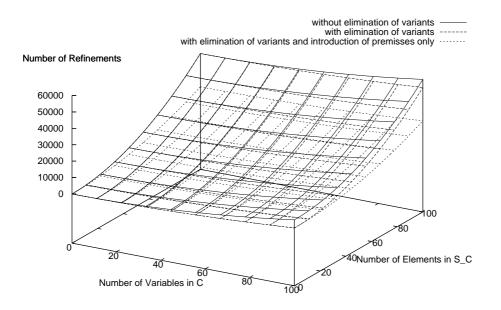


Figure 7.2.: Growth Rate by adding Premises only

rally generalized in order to include mechanisms for refining objects containing temporal operators. Therefore refinement operators for PROLOG(+T)-objects could be easily derived by adapting several well-known techniques from the field of first order logic programming.

What remains to be studied is the complexity of the search for a correct program given sets  $\mathcal{E}^+$  and  $\mathcal{E}^-$  of examples. This topic will be attacked in the following chapter.

# 8. Identifiability of

# Prolog(+T)-programs

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The operators from the last chapters always create programs which are at every point of time correct with respect to the examples which have been presented so far. But in order to be of great use it is necessary to be able to ensure that other examples, that is examples which have not (yet) been seen are classified correct. Consider two sets  $\mathcal{E}^+$  and  $\mathcal{E}^-$  of examples and a program P constructed by the algorithms from the last chapters giving these examples as inputs. Furthermore assume that e is any ground atom from  $B_P^{\text{FoLTL}} \setminus (\mathcal{E}^+ \cup \mathcal{E}^-)$ . In order to be good, P should classify e correct. Formally let  $P_{\text{cor}}$ be the correct program from which the examples from  $\mathcal{E}^+$  and  $\mathcal{E}^-$  are derived, that is let  $P_{\text{cor}}$  be the program to be identified. Then we want P to have the following properties:

- if  $P_{\text{cor}} \models e$ , then  $P \models e$  and
- if  $P_{\text{cor}} \not\models e$ , then  $P \not\models e$ .

In other words

$$\{e \mid P_{\operatorname{cor}} \models e\} = \{e \mid P \models e\}.$$

In general this cannot be reached. Therefore we will adapt a model of *learning* which allows some errors with a certain probability. The issue of identifying logic programs in the PAC-setting has been studied in depth in [36] and [37]. In general not every program is identifiable. But for certain subsets identifiability can be ensured.

### 8.1. PAC–Learning

PAC-learning is a model of learning which has been introduced by Vailant (see [157]). Any algorithm which has to solve the learning problem is evaluated with respect to two parameters  $\varepsilon$  and  $\delta$  which specify a limit for the *difference* between the concept to be learned and the actual hypothesis ( $\varepsilon$ ) and the probability that the actual difference is greater than this level ( $\delta$ ).

In order to give formal definitions of PAC-learnability we need some more concepts. Single objects to be identified will be referred to as *concepts*, the sets of all such objects will be called *concept classes*. Each concept class is defined over some set X. Formally a concept class over X is a set  $C \subseteq 2^X$ . So a concept is an element  $C \in C$ , that is a set  $C \in 2^X$ .

In our case of learning PROLOG(+T)-programs we assume that a signature sig =  $(\mathcal{X}, F, P, \alpha)$  is given. Then we define X to consists of all sets of PROLOG(+T)-literals which contain at least one positive literal, that is

$$X := \{ C \subseteq \mathcal{L}_t(\operatorname{sig}) \mid |\operatorname{Pos}(C)| \ge 1 \}.$$

Concepts are PROLOG(+T)-programs, that is sets  $P \subseteq X$  and concept classes are sets of PROLOG(+T)-programs, that is sets  $\mathcal{C} \subseteq 2^X$ . In order to identify a particular concept an algorithm has to process some kind of examples. In [157] and [63] the examples are considered to be elements from X. However, it will be more adequate for us to consider only ground atoms as examples. So an *example* is an element  $\varphi$  from the set { $\psi \in \mathcal{A}_t(\text{sig}) \mid \text{VAR}(\varphi) = \emptyset$ }. A classified example for a concept P is a tuple  $\langle \varphi, v \rangle$  consisting of an example  $\varphi$  and  $v \in \{0, 1\}$ . If v = 1 we call the example positive and assume that  $P \models \varphi$  and in the case that v = 0 we call it negative and therefore assume that  $P \not\models \varphi$ . For the sake of readability we will from now on use the following notational convention: let P be a PROLOG(+T)-program and let  $\varphi$  be an example. Then

$$P(\varphi) = 1 \quad \Leftrightarrow \quad P \models \varphi \text{ and}$$
$$P(\varphi) = 0 \quad \Leftrightarrow \quad P \not\models \varphi.$$

A sample for P is a sequence  $S_P = (\langle \varphi_1, P(\varphi_1) \rangle, \dots, \langle \varphi_n, P(\varphi_n) \rangle)$  of classified examples. The number n will be called the *length* of the sample or the sample complexity.

Let P and P' be PROLOG(+T)-programs and let  $S_P = (\langle \varphi_1, P(\varphi_1) \rangle, \dots, \langle \varphi_n, P(\varphi_n) \rangle)$ be a sample of length n for P. P' is called *consistent* with respect to  $S_P$  or  $S_P$ -consistent if for every i it holds that  $P'(\varphi_i) = P(\varphi_i)$ , that is a consistent program classifies every example exactly in the same way as the program from which the examples are derived does.

We assume that the examples which are presented to the algorithms are chosen with respect to a fixed probability distribution D on the set of all ground atoms. Also we will sometimes write  $\mathbf{Pr}_D$  instead of D or simply  $\mathbf{Pr}$  if D is clear from the context.

Now let  $\mathcal{S}(X, \mathcal{C})$  be the set of all samples which can be constructed for concepts from  $\mathcal{C}$  if the examples are chosen with respect to D and let  $\mathcal{C}$  and  $\mathcal{H}$  be concept classes. A *learning algorithm* (or a *learner*) is a total function<sup>1</sup>  $\mathcal{A}_{\mathcal{C},\mathcal{H}} : \mathcal{S}(X,\mathcal{C}) \to \mathcal{H}$ . That is a

<sup>&</sup>lt;sup>1</sup>That is, a function which is defined for all possible inputs.

learning algorithm is presented a sample of some length and constructs a hypothesis from the example which it has seen by processing the sample. In our case we will have  $\mathcal{H} = \mathcal{C}$ , so we do not have to distinguish between different representation languages.

As one might expect, a hypothesis computed by some learning algorithm  $\mathcal{A}_{\mathcal{C},\mathcal{H}}$  is called consistent with respect to a sample  $S_P = (\langle \varphi_i, P(\varphi_i) \rangle)_{i=1}^n$  if for every  $i \in \{1, \ldots, n\}$  it holds that

$$\left(\mathcal{A}_{\mathcal{C},\mathcal{H}}(S_P)\right)(\varphi_i) = P(\varphi_i).$$

 $\mathcal{A}_{\mathcal{C},\mathcal{H}}$  is called *consistent* if every hypothesis computed from a given sample is consistent with respect to this sample.

The quality of a hypothesis is measured by analyzing the probability that a randomly chosen example is contained in the symmetric difference between the *correct concept* and the *computed concept*. So the *error* of a program P' with respect to a program P is given as

$$\operatorname{error}(P, P') = D\left(P\Delta P'\right) = D\left((P \setminus P') \cup (P' \setminus P)\right).$$

The error of a hypothesis computed by a learning algorithm is usually given as a parameter called  $\varepsilon$  and one is interested in upper bounds for the probability that a learning algorithm or a classifier induces hypotheses such that the error of these hypotheses exceeds the value of  $\varepsilon$ . Several attempts have been carried out in order to analyze such errors. In 1971 Vapnik and Chervonenkis have shown (see [166]) that classifiers can be constructed which have an error ration bounded from above by  $4s(\mathcal{C}, 2n)e^{-\frac{n\varepsilon^2}{8}}$  for a given value or  $\varepsilon$  where  $s(\mathcal{C}, 2n)$  denotes the relative amount of samples of size 2n which can be drawn from the concept class  $\mathcal{C}$ . Devroye and Wagner (see [46]) have extended the results from [166] in order to derive distribution-free upper bounds for the error in the case of half-planes. They show that classifiers can be constructed with an error rate bounded by  $4(1+2^d n_i^d)e^{-\frac{n_i\varepsilon^2}{8}}$  for i = 1, 2 where d denotes the size of the samples under consideration while the problem domain is divided into classes  $\mathcal{C}_1$  and  $\mathcal{C}_2$ . Furthermore they extended

their results to concept classes with higher dimension yielding similar results.

Furthermore some attempts for analyzing the learning task in the domain of identifying boolean functions have been carried out. In [131] Pearl derives upper and lower bounds for the size of the value s(n, c) denoting the relative frequency of samples of size n derived from a concept class representing a boolean function which can be built up with at most c binary gates by proving that

$$s(n,c) \ge \begin{cases} 2^{n_0 - n} & \text{if } n \ge n_0 \\ 1 & \text{else} \end{cases}$$

and

$$s(n,c) \le 2^{\log_2|F_c|-n}$$

where  $n_0$  denotes the maximum value of n such that a presented sample  $e_n$  can be embedded in a boolean function f using at most c logical gates (where  $|F_c|$  denotes the total number of boolean functions containing at most c gates). Using these bounds Pearl proves that classifiers can be constructed which have an error ration bounded from above by

$$\left(\sqrt{2\ln 2 \cdot c(2 + \log_2 c)} + \frac{1}{\sqrt{2\ln 2 \cdot c(2 + \log_2 c)}}\right) \frac{2}{\sqrt{n}}$$

Finally Devroye (see [45]) derives upper bounds for the error of classifiers for both the case of finite and infinite concept classes. For finite classes C he proves that the error can be bounded from above by  $2|C|e^{-2m\varepsilon^2}$  (using samples of size m) and that the expected error is given by

$$\sqrt{\frac{\log(2|\mathcal{C}|}{2m}} + \frac{1}{\sqrt{8m\log(2|\mathcal{C}|)}}$$

For infinite concept classes they show that the error can be bounded from above by

 $cs(\mathcal{C},m^2)e^{-2m\varepsilon^2}$  and that the expected error is given by

$$\sqrt{\frac{\log(4e^8s(\mathcal{C},m^2))}{2m}} + \frac{1}{\sqrt{8m\log(4e^8s(\mathcal{C},m^2))}}$$

Having presented all necessary prerequisites we can now formally define the concept of *PAC-identifiability*. For our definition we will only slightly change the notations from [22].

Definition 8.1.1 (PAC-Learnability, Valiant [157] and Blumer et al. [22]) Let X be some set and let  $\mathcal C$  and  $\mathcal H$  be concept classes over X.  $\mathcal C$  is called  $\mathit{PAC}$ *learnable* using  $\mathcal{H}$  if and only if there is a learning algorithm  $\mathcal{A}_{\mathcal{C},\mathcal{H}}$  and a function  $m: \mathbb{R}^2 \to \mathbb{R}$  such that for every probability distribution D, every  $P \in \mathcal{C}$  and every values of  $0 < \varepsilon, \delta < 1$  it holds that:

- *A*<sub>C,H</sub> is presented some sample *S<sub>P</sub>* of length [*m*(ε, δ)],
   *A*<sub>C,H</sub>(*S<sub>P</sub>*) is put out and
   **Pr**(error(*A*<sub>C,H</sub>(*S<sub>C</sub>*, *P*)) ≥ ε) = **Pr**(*D*((*A*<sub>C,H</sub>(*S<sub>P</sub>*)) Δ*P*) ≥ ε) < δ.</li>

The last point from the above definition is usually referred to as the PAC-criterion or the PAC-criterion with respect to  $\varepsilon$  and  $\delta$ .

So  $\mathcal{C}$  is PAC-learnable if there exists an algorithm which regardless of the underlying distribution (which determines how the examples are chosen) only needs to process a finite set of examples in order to keep the difference between the hypothesis and the correct program small ( $\leq \varepsilon$ ) with a high probability ( $< \delta$ ).

The PAC concept has been introduced in [157] for the domain of learning boolean functions. Often PAC-identifiability is referred to as an abbreviation to *polynomial time* PAC-identifiability where a further restriction is put on the learning algorithm  $\mathcal{A}_{\mathcal{C},\mathcal{H}}$ , namely that its runtime is bounded from above by some suitable polynomial. So in this case the term C is PAC-identifiable means that C is polynomial time PAC-identifiable. Consequently there might be concept classes which are PAC-identifiable in our concept which does not include this requirement for the runtime of learning algorithms but which are not polynomial time PAC-identifiable. In [158] and [132] L.G. Valiant addresses this problem deriving several classes of relatively simply structured boolean functions which are not PAC-identifiable if polynomial runtime is required by a learning algorithm.

In order to characterize the complexity of learning single concepts from a concept class C the Vapnik-Chervonenkis-Dimension has been proven to be an adequate parameter. Intuitively the Vapnik-Chervonenkis-Dimension characterizes the difficulty of how to distinguish between different objects from C. This intuition will now be made formally clear.

## Definition 8.1.2 (Blumer et al. [22])

Let  $\mathcal{C}$  be a concept class over some set X and let  $T \in \mathcal{C}$  be a concept. Then

$$\Pi_{\mathcal{C}}(T) = \{ C \cap T \mid C \in \mathcal{C} \} \,.$$

A set T of cardinality k is said to be *shattered* by C if  $|\Pi_{\mathcal{C}}(T)| = 2^k$ , that is if  $\Pi_{\mathcal{C}}(T) = 2^T$ . So the sets C which are shattered by C can be seen as the most difficult concepts from C. The Vapnik–Chervonenkis–Dimension of a concept class C is now defined to be the maximum size of a concept which is shattered by C.

#### Definition 8.1.3 (VC-Dimension, Blumer et al. [22])

Let  $\mathcal{C}$  be a concept class over some set X. The Vapnik-Chervonenkis-Dimension of  $\mathcal{C}$  is defined as

VCDIM(
$$\mathcal{C}$$
) = max  $\left\{ k \mid \text{ there is a } T \in \mathcal{C} \text{ with } |T| = k \text{ and } |\Pi_{\mathcal{C}}(T)| = 2^k \right\}$ 

 $= \max\left\{ |T| \mid T \in \mathcal{C} \text{ is shattered by } \mathcal{C} \right\}.$ 

If no such k exists we will write  $VCDIM(\mathcal{C}) = \infty$ .  $\mathcal{C}$  is said to have unbounded VC-Dimension in this case.

The link between the VC–Dimension and the learnability of a concept class C is given by the following theorem (see [22]).

#### Theorem 8.1.1 (Blumer et al. [22])

Let  $\mathcal{C}$  be a concept class over some set X. Then  $\mathcal{C}$  is PAC-learnable if and only if  $\operatorname{VCDiM}(\mathcal{C}) < \infty$ .

The following theorem gives a possibility to estimate the length of the sample needed in order to identify the target concept given fixed values of  $\varepsilon$  and  $\delta$ .

#### Theorem 8.1.2 (Blumer et al. [22])

Let  $\mathcal{C}$  be a concept class over some set X such that  $1 \leq \text{VCDIM}(\mathcal{C}) < \infty$  and let  $0 < \varepsilon \leq \frac{1}{2}$ and  $0 < \delta < 1$  be given. Then every consistent learning algorithm for  $\mathcal{C}$  using  $\mathcal{C}$  needs to process at most

$$\left\lceil \max\left\{\frac{4}{\varepsilon}\ln\frac{4}{\delta}, \ln\frac{8\text{VCDIM}(\mathcal{C})}{\varepsilon}\ln\frac{13}{\varepsilon}\right\} \right\rceil$$

examples in order to ensure the PAC-criterion with respect to  $\varepsilon$  and  $\delta^2$ .

In general estimating the VC–Dimension of some concept class C is a very difficult task. But in the case of finite concept classes, the VC–Dimension is bounded by the logarithm of the size of the class. This is the tenor of the following lemma.

#### Lemma 8.1.1 (Fischer [63])

Let  $\mathcal{C}$  be a concept class over some set X. If  $\mathcal{C}$  is finite, then

 $\mathrm{VCDIM}(\mathcal{C}) \leq \log_2 |\mathcal{C}|.$ 

<sup>&</sup>lt;sup>2</sup>In general one has to take the VC-Dimension VCDIM( $\mathcal{H}$ ) of the target concept class into account. But since we require  $\mathcal{C} = \mathcal{H}$  this makes the analysis of the learning problem a bit easier.

In the following sections, Lemma 8.1.1 will be used in order to derive some upper bounds for the VC–Dimensions of several classes of PROLOG(+T)–programs.

# 8.2. Learnability and Non–Learnability of selected classes of Prolog(+T)–programs

#### 8.2.1. The general case

We will now derive upper and lower bounds for the VC–Dimension of some classes of PROLOG(+T)–programs. Therefore we will extend some techniques presented recently in [11].

#### Definition 8.2.1

Let c, t, l and o be nonnegative integers. The class  $\mathcal{P}^{\leq c, t, l, o}$  is defined as the set of all PROLOG(+T)-programs P with the following properties:

- 1. P consists of at most c rules,
- 2. each rule in P consists of at most l literals,
- 3. each literal in a rule in P does not contain more than t distinct terms and
- 4. each literal in a rule in P does not contain more than o temporal operators.

Assume that a fixed signature sig =  $(\mathcal{X}, F, P, \alpha)$  is given. We will from now on use the following abbreviations:

- 1. f := |F|,
- 2. p := |P| and
- 3.  $a := \max \{ \alpha(\sigma) \mid \sigma \in F \cup P \}.$

Programs will be encoded as strings over the binary alphabet  $\Sigma = \{0, 1\}$ . We will see that for fixed values of c, t, l and o, the number  $|\mathcal{P}^{\leq c, t, l, o}|$  is finite. The VC–Dimension can then be estimated using Lemma 8.1.1. We will assume that each literal in any program is reduced. This is no restriction since we have seen that each literal has a reduced normal form which can be effectively computed.

First we will review some of the results from [11]. There it is shown that a term containing a arguments and at most t distinct subterms can be encoded using  $\log_2 f + a \log_2 t$  bits. Consequently a set of t (distinct) terms can be encoded using not more than  $t(\log_2 f + a \log_2 t)$  bits. Since f and a are constant we have  $t(\log_2 f + a \log_2 t) = \mathcal{O}(t \log_2 t)$ .

We will now fix a numbering for the symbols from P and the temporal operators. Let  $P_{\text{ext}}$  denote the set  $P \cup \{X, F, G, U, P\}$  and assume that the set P is *ordered* as follows:  $P = \{p_0, \ldots, p_{|P|-1}\}$ . The symbols  $p_i$  will be mapped to bin(i) where bin(i) denotes the string representing the binary representation of i. Furthermore we fix the following mapping:

$$\begin{array}{rcl} \mathsf{X} & \mapsto & \mathrm{bin}(|P|+1), \\ \mathsf{F} & \mapsto & \mathrm{bin}(|P|+2), \\ \mathsf{G} & \mapsto & \mathrm{bin}(|P|+3), \\ \mathsf{U} & \mapsto & \mathrm{bin}(|P|+4) \ \mathrm{and} \\ \mathsf{P} & \mapsto & \mathrm{bin}(|P|+5), \end{array}$$

where the strings might be padded with zeros on the left side in order to obtain strings of equal length.

#### Example 8.2.1

The set  $P = \{p_0\}$  containing only a single predicate symbol yields the following mapping:

$$p_0 \mapsto 000,$$

$$X \mapsto 001,$$

$$F \mapsto 010,$$

$$G \mapsto 011,$$

$$U \mapsto 100 \text{ and}$$

$$P \mapsto 101.$$

Consequently any PROLOG(+T)-literal containing at most t distinct terms and at most o temporal operators can be encoded using at most

$$o + 1 + 2\left(\left\lceil \log_2(p+5)\right\rceil + \left\lceil \log_2 p \right\rceil + \left\lceil a \log_2 t \right\rceil\right)\right)$$

bits. Therefore a rule consisting of at most l such literals can be encoded using at most

$$l\left(o+1+2\left(\left\lceil \log_2(p+5)\right\rceil+\left\lceil \log_2 p\right\rceil+\left\lceil a\log_2 t\right\rceil\right)\right)$$

bits and a program P containing at most c such rules can be encoded using at most

$$cl\left(o+1+2\left(\left\lceil \log_2(p+5)\right\rceil+\left\lceil \log_2 p\right\rceil+\left\lceil a\log_2 t\right\rceil\right)\right)$$

bits. This gives:

$$cl (o + 1 + 2 (\lceil \log_2(p+5) \rceil + \lceil \log_2 p \rceil + \lceil a \log_2 t \rceil))$$
  
=  $clo + cl + 2cl (\lceil \log_2(p+5) \rceil + \lceil \log_2 p \rceil + \lceil a \log_2 t \rceil)$ 

$$= cl(o+1) + 2cl(\lceil \log_2(p+5) \rceil + \lceil \log_2 p \rceil + \lceil a \log_2 t \rceil))$$
  
=  $\mathcal{O}(cl(o+1) + 2cl(\log_2(p+5) + \log_2 p + a \log_2 t)))$   
=  $\mathcal{O}(cl(o+1) + 2cl(\log_2((p^2 + 5p)t^a))).$ 

With this number of bits we can encode at most  $2^{\mathcal{O}(cl(o+1)+2cl(\log_2(p^2+5p)t^a)))}$  different bitstrings, that is we have

$$|\mathcal{P}^{\leq c,t,l,o}| = 2^{\mathcal{O}\left(cl(o+1)+2cl\left(\log_2\left((p^2+5p)t^a\right)\right)\right)} < \infty$$

and therefore

VCDIM 
$$\left(\mathcal{P}^{\leq c,t,l,o}\right) = \mathcal{O}\left(cl(o+1) + 2cl\left(\log_2\left((p^2 + 5p)t^a\right)\right)\right)$$

using Lemma 8.1.1.

On the other hand the best case is given if no temporal operators are involved. [11] then gives the following estimation:

VCDIM 
$$\left(\mathcal{P}^{\leq c,t,l,o}\right) = \Omega\left(cl+ct\right).$$

The results are summarized in the following theorem.

#### Theorem 8.2.1

Let c, t, l and o be fixed, nonnegative integers. Then

$$\text{VCDIM}\left(\mathcal{P}^{\leq c,t,l,o}\right) = \mathcal{O}\left(c + (o+1) + 2cl\left(\log_2\left((p^2 + 5p)t^a\right)\right)\right) \text{ and}$$
$$\text{VCDIM}\left(\mathcal{P}^{\leq c,t,l,o}\right) = \Omega\left(cl + ct\right).$$

Using these equalities we can estimate the number of examples needed in order to ensure the PAC-criterion given fixed values of  $\varepsilon$  and  $\delta$ .

#### Theorem 8.2.2

Let  $c, t, l, o \ge 0, 0 < \varepsilon \le \frac{1}{2}$  and  $0 < \delta < 1$  be fixed. Then every learning algorithm  $\mathcal{A}$  needs at most

$$\max\left\{\frac{4}{\varepsilon}\ln\frac{4}{\delta}, \frac{8\mathcal{O}\left(cl(o+1)+2cl\left(\log_2\left((p^2+5p)t^a\right)\right)\right)}{\varepsilon}\ln\frac{13}{\varepsilon}\right\}$$

examples in order to ensure the PAC-criterion.

#### Example 8.2.2

We conclude this section by illustrating the results for the number of examples. Let o = 10, l = 30, c = 200, t = 50, p = 7 and a = 17. Then the equation from Theorem 8.2.2 can be simplified to:

$$\max \left\{ \frac{4}{\varepsilon} \ln \frac{4}{\delta}, \frac{8 \text{VCDIM} \left( \mathcal{P}^{\leq c, t, l, o} \right)}{\varepsilon} \ln \frac{13}{\varepsilon} \right\}$$

$$= \max \left\{ \frac{4}{\varepsilon} \ln \frac{4}{\delta}, \frac{8 \mathcal{O} \left( cl(o+1) + 2cl \left( \log_2 \left( (p^2 + 5p)t^a \right) \right) \right)}{\varepsilon} \ln \frac{13}{\varepsilon} \right\}$$

$$\approx \max \left\{ \frac{4}{\varepsilon} \ln \frac{4}{\delta}, \frac{8 \left( cl(o+1) + 2cl \left( \log_2 \left( (p^2 + 5p)t^a \right) \right) \right)}{\varepsilon} \ln \frac{13}{\varepsilon} \right\}$$

$$\approx \max \left\{ \left\lceil \frac{4}{\varepsilon} \ln \frac{4}{\delta} \right\rceil, \left\lceil \frac{8 \left( 200 \cdot 30(10+1) + 2 \cdot 200 \cdot 30 \left( \log_2 \left( (7^2 + 5 \cdot 7)50^{17} \right) \right) \right)}{\varepsilon} \ln \frac{13}{\varepsilon} \right\rceil \right\}$$

$$= \max \left\{ \left\lceil \frac{4}{\varepsilon} \ln \frac{4}{\delta} \right\rceil, \left\lceil \frac{528000 + 96000 \left( \log_2 \left( (7^2 + 5 \cdot 7)50^{17} \right) \right)}{\varepsilon} \ln \frac{13}{\varepsilon} \right\rceil \right\}$$

$$\approx \max \left\{ \left\lceil \frac{4}{\varepsilon} \ln \frac{4}{\delta} \right\rceil, \left\lceil \frac{528000 + 69000 \cdot 96}{\varepsilon} \ln \frac{13}{\varepsilon} \right\rceil \right\}$$

$$= \max \left\{ \left\lceil \frac{4}{\varepsilon} \ln \frac{4}{\delta} \right\rceil, \left\lceil \frac{9744000}{\varepsilon} \ln \frac{13}{\varepsilon} \right\rceil \right\}.$$

Note that the approximation given above is quite weak since the omission of the symbol  $\mathcal{O}$  may result in omitting quite large constants.

Figure 8.1 illustrates the number of examples for c, t, l, o, p and a as above and variable values of  $\varepsilon$  and  $\delta$ .

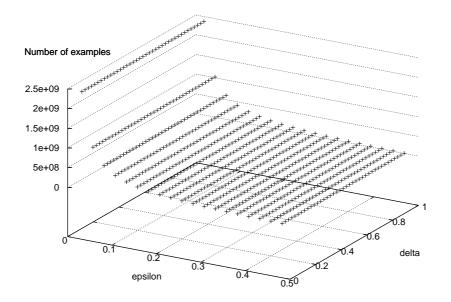


Figure 8.1.: Number of examples given fixed values for c, t, l, o, p and a with  $\varepsilon$  ranging from 0 to  $\frac{1}{2}$  and  $\delta$  ranging from 0 to 1.

#### 8.2.2. Programs with syntactical restrictions

We will now see how restricting the form of the involved rules in programs can lower the VC-dimension of a class of PROLOG(+T)-programs and therefore make it easier to identify these programs by presenting positive and negative examples. Therefore we will study two classes of programs which have already been pointed out in [11]: constrained programs and range-restricted programs.

Definition 8.2.2 (Syntactical Restrictions, Arias and Khardon [11])

Let  $C = \varphi := \psi_1, \dots, \psi_n$ . be a  $\operatorname{PROLOG}(+T)$ -rule. C is called • range-restricted if  $\operatorname{TERMS}(\varphi) \subseteq \bigcup_{i=1}^n \operatorname{TERMS}(\psi_i)$  and • constrained if  $\bigcup_{i=1}^n \operatorname{TERMS}(\psi_i) \subseteq \operatorname{TERMS}(\varphi)$ .

Consequently a PROLOG(+T)-program  $P = \{P_1, \ldots, P_k\}$  is called *range-restricted* (resp. constrained) if every  $P_i$  is range-restricted (resp. constrained).

Fixing nonnegative integers c, t, l and o, the definition of the classes  $\mathcal{P}_{\text{con}}^{\leq c,t,l,o}$  and  $\mathcal{P}_{\mathrm{rr}}^{\leq c,t,l,o}$  is as one might expect:

- $\mathcal{P}_{\text{con}}^{\leq c,t,l,o} = \{ P \in \mathcal{P}^{\leq c,t,l,o} \mid P \text{ is constrained} \}$  and
- $\mathcal{P}_{\mathrm{rr}}^{\leq c,t,l,o} = \{ P \in \mathcal{P}^{\leq c,t,l,o} \mid P \text{ is range-restricted} \}.$

We will now study how the values of VCDIM  $\left(\mathcal{P}_{con}^{\leq c,t,l,o}\right)$  and VCDIM  $\left(\mathcal{P}_{rr}^{\leq c,t,l,o}\right)$  can be estimated using the results from the foregoing section.

#### The VC–Dimension of constrained Prolog(+T)–programs

Let  $P \in \mathcal{P}_{con}^{\leq c,t,l,o}$  be given. Assume that  $P = \{P_1, \ldots, P_k\}$  for some  $k \leq c$  and

$$P_i = \varphi_i :-\psi_1^{(i)}, \dots, \psi_{n_i}^{(i)}.$$

Since  $P \in \mathcal{P}_{\text{con}}^{\leq c,t,l,o}$  we have

$$\bigcup_{j=1}^{n_i} \operatorname{Terms}\left(\psi_j^{(i)}\right) \subseteq \operatorname{Terms}(\varphi_i)$$

for i = 1, ..., k. Therefore only the terms in the heads of the rules in P have to be encoded. This can be achieved by using at most  $o+1+2(\lceil \log_2(p+5) \rceil + \lceil \log_2 p \rceil + \lceil a \log_2 t \rceil)$  bits. The literals from  $\left\{\psi_1^{(j)}, \ldots, \psi_{n_j}^{(j)}\right\}$  can be encoded by only encoding the negation symbols, the predicate symbol(s) and the temporal operators involved. Consequently this can be achieved by using at most

$$o + 1 + 2\left(\left\lceil \log_2(p+5)\right\rceil + \left\lceil \log_2 p\right\rceil\right)$$

bits per literal. Therefore the tail of a rule containing at most l literals can be encoded by using at most

$$(l-1)(o+1+2(\lceil \log_2(p+5) \rceil + \lceil \log_2 p \rceil))$$

bits. So a complete rule can be encoded using at most

 $o+1+2\left(\left\lceil \log_2(p+5)\right\rceil + \left\lceil \log_2 p\right\rceil + \left\lceil a \log_2 t\right\rceil\right) + (l-1)\left(o+1+2\left(\left\lceil \log_2(p+5)\right\rceil + \left\lceil \log_2 p\right\rceil\right)\right)$ 

bits. We have

$$\begin{split} o+1+2\left(\lceil \log_2(p+5)\rceil + \lceil \log_2 p\rceil + \lceil a \log_2 t\rceil\right) \\ +(l-1)\left(o+1+2\left(\lceil \log_2(p+5)\rceil + \lceil \log_2 p\rceil\right)\right) \\ = & o+1+2\lceil \log_2(p+5)\rceil + 2\log_2 p\rceil + 2\lceil a \log_2 t\rceil + (l-1)(o+1) \\ & +2(l-1)\lceil \log_2(p+5)\rceil + 2(l-1)\lceil \log_2 p\rceil \\ = & l(o+1)+2l\lceil \log_2(p+5)\rceil + 2l\lceil \log_2 p\rceil + 2\lceil a \log_2 t\rceil \\ = & l(o+1)+2\left(l\left(\lceil \log_2(p+5)\rceil + \lceil \log_2 p\rceil\right) + \lceil a \log_2 t\rceil\right). \end{split}$$

So any program  $P \in \mathcal{P}_{con}^{\leq c,t,l,o}$  can be encoded using at most

$$cl(o+1) + 2c\left(l\left(\lceil \log_2(p+5) \rceil + \lceil \log_2 p \rceil\right) + \lceil a \log_2 t \rceil\right)$$

bist and therefore the VC-dimension of  $\mathcal{P}_{\text{con}}^{\leq c,t,l,o}$  can be estimated as stated in the following theorem.

#### Theorem 8.2.3

Let c, t, l and o be nonnegative integers. Then

VCDIM 
$$\left(\mathcal{P}_{\text{con}}^{\leq c,t,l,o}\right) = \mathcal{O}\left(cl(o+1) + 2c\log_2\left((p^2+5p)^lt^a\right)\right).$$

**Proof**. The claim is due to

$$cl(o+1) + 2c \left( l \left( \lceil \log_2(p+5) + \lceil \log_2 p \rceil \right) + \lceil a \log_2 t \rceil \right) \right)$$
  
=  $\mathcal{O} \left( cl(o+1) + 2c \left( l \log_2(p+2+5p) + a \log_2 t \right) \right)$   
=  $\mathcal{O} \left( cl(o+1) + 2c \left( \log_2(p^2+5p)^l + \log_2 t^a \right) \right)$   
=  $\mathcal{O} \left( cl(o+1) + 2c \log_2 \left( (p^2+5p)^l t^a \right) \right)$ 

and an application of Lemma 8.1.1.

#### The VC-Dimension of range-restricted Prolog(+T)-programs

In some sense this situation is similar to the situation from the last section. Assume that  $P = \{P_1, \ldots, P_k\} \in \mathcal{P}_{rr}^{\leq c,t,l,o}$  is given such that  $P_i = \varphi_i : -\psi_1^{(i)}, \ldots, \psi_{n_i}(i)$ . Similarly as in the case of constrained programs it suffices to encode the terms from  $\{\psi_1^{(i)}, \ldots, \psi_{n_i}^{(i)}\}$ . Since  $\left|\{\psi_1^{(i)}, \ldots, \psi_{n_i}^{(i)}\}\right| \leq l-1$  the complete tail of  $P_i$  can be encoded using at most

$$(l-1)(o+1+2(\lceil \log_2(p+5) \rceil + \lceil \log_2 p \rceil + \lceil a \log_2 t \rceil))$$

bits while the remaining literal  $\varphi_i$  can be encoded by skipping the involved terms using at most

$$o + 1 + 2(\lceil \log_2(p+5) \rceil + \lceil \log_2 p \rceil)$$

bits. So the complete rule  $P_i$  can be encoded with

 $o+1+2\left(\left\lceil \log_2(p+5)\right\rceil + \left\lceil \log_2 p\right\rceil\right) + (l-1)\left(o+1+2\left(\left\lceil \log_2(p+5)\right\rceil + \left\lceil \log_2 p\right\rceil + \left\lceil a\log_2 t\right\rceil\right)\right)$ 

bits. We have

$$\begin{split} & o+1+2\left(\lceil \log_2(p+5)\rceil+\lceil \log_2 p\rceil\right)\\ & +(l-1)\left(o+1+2\left(\lceil \log_2(p+5)\rceil+\lceil \log_2 p\rceil+\lceil a\log_2 t\rceil\right)\right)\\ & = o+1+2\lceil \log_2(p+5)\rceil+2\lceil \log_2 p\rceil+(l-1)(o+1)+2(l-1)\lceil \log_2(p+5)\rceil\\ & +2(l-1)\lceil \log_2 p\rceil+2(l-1)\lceil a\log_2 t\rceil\\ & = l(o+1)+2\left(l\left(\lceil \log_2(p+5)\rceil+\lceil \log_2 p\rceil\right)+(l-1)\lceil a\log_2 t\rceil\right). \end{split}$$

Therefore P can be encoded using at most

$$cl(o+1) + 2c\left(l\left(\left\lceil \log_2(p+5)\right\rceil + \left\lceil \log_2 p\right\rceil\right) + (l-1)\left\lceil a\log_2 t\right\rceil\right)$$

bits and we can estimate the VC-dimension of  $\mathcal{P}_{rr}^{\leq c,t,l,o}$  as follows.

#### Theorem 8.2.4

Let c, t, l and o be nonnegative integers. Then

$$\operatorname{VCDIM}\left(\mathcal{P}_{\operatorname{rr}}^{\leq c,t,l,o}\right) = \mathcal{O}\left(cl(o+1) + 2c\log_2\left((p^2 + 5p)^l t^{a+l-1}\right)\right).$$

**Proof**. The claim is due to

$$cl(o+1) + 2c\left(l\left(\left\lceil \log_2(p+5)\right\rceil + \left\lceil \log_2 p \right\rceil\right) + (l-1)\left\lceil a \log_2 t \right\rceil\right)$$

$$= \mathcal{O}\left(cl(o+1) + 2c\left(l\left(\log_2(p+5) + \log_2 p\right) + (l-1)a\log_2 t\right)\right) \\ = \mathcal{O}\left(cl(o+1) + 2c\left(l\left(\log_2(p^2 + 5p)\right) + (l-1)a\log_2 t\right)\right) \\ = \mathcal{O}\left(cl(o+1) + 2c\left(\log_2(p^2 + 5p)^l + \log_2 t^{a+l-1}\right)\right) \\ = \mathcal{O}\left(cl(o+1) + 2c\log_2\left((p^2 + 5p)^l t^{a+l-1}\right)\right)$$

and Lemma 8.1.1.

Having obtained the above results on the complexity of the identification task for PRO-LOG(+T)-programs our treatment of first order inductive temporal logic programming is complete. We have seen the following:

- It is reasonable to study programs written in PROLOG(+T) since this language is both powerful (it contains the full first order fragment of horn clause programs) and still relatively tractable (due to the syntactic limitation to rules),
- PROLOG(+T) is equipped with a well defined semantics given by temporally closed sets of ground atoms and
- PROLOG(+T) allows specialization and generalization of concepts by application of refinement operators.

All these points make clear that PROLOG(+T) is a suitable language for the specification of reactive systems using first order temporal logic. In contrast a restriction to propositional temporal logic results in the language LTL which is decidable for satisfiability and which is not limited to formulas in clause form. Although LTL is consequently less expressive than PROLOG(+T) the two properties mentioned before justify studying the identification problem in LTL. This will therefore be the topic of the following part of this thesis.

Part III.

# Propositional Inductive Temporal Logic Programming

# 9. Preliminaries

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Having completed our treatment of first order inductive temporal logic programming we will now restrict ourselves on propositional logic languages. On the one hand this will be a limitation since we do not allow reasoning about functions and predicates with arity greater than 0 but on the other hand we will be able to use much more syntactically complex statements since we will not be limited to statements in clause form any longer. The language of interest will be LTL as introduced in chapter 2.3.1.

In this chapter we will describe the necessary preliminaries from the theory of propositional (linear) temporal logic which will be used in the sequel. First we will introduce the concept of Büchi-automata which are automata over infinite sequences of symbols (*infinite words* or  $\omega$ -words). After this we will see that every LTL-formula  $\varphi$  can be translated into a Büchi automaton which has a nonempty accepted language if and only if the formula  $\varphi$  from which is has been constructed is satisfiable.

## 9.1. Finite Automata on infinite Objects

We will now introduce notations which allow us to generalize the theory of formal languages consisting of finite words to such languages which consist of infinite words, that is words which can be seen as an infinite sequence of elements (called the *letters* of the word). Therefore assume for the rest of this chapter that  $\Sigma$  is a finite set of symbols, called the *alphabet*.

## Definition 9.1.1 ( $\omega$ -word, e.g. Lothaire [106])

An *infinite word* (or  $\omega$ -word) over  $\Sigma$  is a mapping  $w : \mathbb{N} \to \Sigma$ . The set of all infinite words over  $\Sigma$  is denoted as  $\Sigma^{\omega}$ . Every set  $L \subseteq \Sigma^{\omega}$  is called an  $\omega$ -language over  $\Sigma$ .

 $\omega$ -languages are a natural way to extend the theory of formal languages to infinite sequences of letters. In many practical applications  $\omega$ -languages are used in order to model infinite sequences of actions performed by nonterminating (reactive) systems. The set of *all* possible behaviors of such a system is described in terms of an  $\omega$ -language and it is then checked if a property  $\varphi$  holds in this system simply by checking if the  $\omega$ -language which is recognized by the product structure which emerges from the model of the system and the negation of the formula, is empty. How to achieve this, will the subject of the rest of this chapter.

In order to recognize  $\omega$ -languages, the theory of finite automata has been extended by adding structures which allow acceptance of infinite words. This leads to the theory of  $\omega$ -automata as described by Büchi (see [26]), Street (see [155]), Muller (see [123]) and Rabin (see [138]). We will describe  $\omega$ -automata in a similar was as defined by Büchi since this type of automata is used in model checking to characterize the set of models of LTL-formulas.

#### Definition 9.1.2 (Büchi-automaton, e.g. Wolper [172])

Let  $\Sigma$  be an alphabet. A nondeterministic Büchi-automaton over  $\Sigma$  is a tuple

$$\mathcal{A} = (\Sigma, \text{STATES}, \delta, S_0, S_f),$$

where

- STATES is a finite set (the set of *states*),
- $\delta$  : STATES  $\times \Sigma \to 2^{\text{STATES}}$  is the transition relation,
- $S_0 \subseteq \text{STATES}$  is the set of *initial states* and
- $S_f \subseteq$  STATES is the set of *accepting* or *final states*.

Note that the formal definition of such a Büchi automaton does only slightly differ from the definition of a finite state automaton accepting finite words. In order to accept infinite words over  $\Sigma$  the acceptance condition of  $\mathcal{A}$  has to be modified since it is clear that the acceptance condition for finite words, requiring that the consuming the last symbol leads into a final state, cannot be applied (since there is no *last symbol*). So assume that a Büchi automaton  $\mathcal{A} = (\Sigma, \text{STATES}, \delta, S_0, S_f)$ , an  $\omega$ -word w and a mapping  $\gamma : \mathbb{N} \to \text{STATES}$  are given. We will denote the set of all states occurring infinitely often in the sequence labeled by  $\gamma$  as  $S^{\infty}(\gamma)$ . Formally:

$$S^{\infty}(\gamma) = \{s \in \text{STATES} \mid || \{i \mid \gamma(i) = s\} || = \infty \}.$$

We call  $\gamma$  adequate for  $(\mathcal{A}, w)$  (or  $(\mathcal{A}, w)$ -adequate) if and only if  $\gamma$  satisfies the following properties:

- 1.  $\gamma(0) \in S_0$ ,
- 2. for every  $i \ge 0$  it holds that  $\gamma(i+1) \in \delta(\gamma(i), w(i))$  and
- 3.  $S^{\infty}(\gamma) \cap S_f \neq \emptyset$ .

A word  $w \in \Sigma^{\omega}$  is said to be *accepted by*  $\mathcal{A}$  if and only if there is a mapping  $\gamma$  such that  $\gamma$  is adequate for  $(\mathcal{A}, w)$ . Similarly a subset L of  $\Sigma^{\omega}$  (that is, an  $\omega$ -language) is accepted by  $\mathcal{A}$  if and only if there is a mapping  $\gamma$  which is such that  $\gamma$  is  $(\mathcal{A}, w)$ -adequate for every  $w \in L$ .

It has proven useful to define a slight modification of Büchi–automata in which the last component of the tuple is not a set of states but a set of sets of states. This leads to generalized Büchi–automata. If

$$\mathcal{A} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$$

is such a generalized Büchi-automaton and  $w \in \Sigma^{\omega}$  is an infinite word over  $\Sigma$ , then  $\gamma : \mathbb{N} \to \text{STATES}$  is called *adequate for*  $(\mathcal{A}, w)$  (or  $(\mathcal{A}, w)$ -adequate) if any only if

- 1.  $\gamma(0) \in S_0$ ,
- 2. for every  $i \ge 0$  it holds that  $\gamma(i+1) \in \delta(\gamma(i), w(i))$  and
- 3. if  $\mathcal{F} = \{F_1, \ldots, F_n\}$ , then  $S^{\infty}(\gamma) \cap F_i \neq \emptyset$  for  $i = 1, \ldots, n$ .

Again we will call w accepted by  $\mathcal{A}$  if there is  $\gamma$  such that  $\gamma$  is  $(\mathcal{A}, w)$ -adequate and  $L \subseteq \Sigma^{\omega}$  is called *accepted by*  $\mathcal{A}$  if there is  $\gamma$  such that  $\gamma$  is  $(\mathcal{A}, w)$ -adequate for every  $w \in L$ .

In both cases, that is if  $\mathcal{A}$  is a Büchi–automaton or if  $\mathcal{A}$  is a generalized Büchi– automaton, the language  $L(\mathcal{A})$  is defined as the set of all  $\omega$ -words which are accepted by  $\mathcal{A}$ .  $L \subseteq \Sigma^{\omega}$  is called *(generalized) Büchi–acceptable* if and only if there is a (generalized) Büchi–automaton  $\mathcal{A}$  such that  $L = L(\mathcal{A})$ . From [172] we have the following lemma.

#### Lemma 9.1.1 (e.g. Wolper [172])

Let  $L \subseteq \Sigma^{\omega}$  be any  $\omega$ -language. Then L is Büchi-acceptable if and only if L is generalized Büchi-acceptable.

**Proof.** The only-if part is immediately: if  $\mathcal{A} = (\Sigma, \text{STATES}, \delta, S_0, S_f)$  is a Büchiautomaton, then we define  $\overline{\mathcal{A}} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  with  $\mathcal{F} = \{\{s_1, \ldots, s_n\}\}$  for  $S_f = \{s_1, \ldots, s_n\}$ . Then  $\overline{\mathcal{A}}$  is a generalized Büchi-automaton and it is straightforward to show that  $w \in L(\mathcal{A})$  if and only if  $w \in L(\overline{\mathcal{A}})$  for every  $w \in \Sigma^{\omega}$ . For the if-part assume that  $\mathcal{A} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  with  $\mathcal{F} = \{F_1, \ldots, F_n\} \subseteq 2^{\text{STATES}}$  is given. We then define  $\overline{\mathcal{A}} = (\Sigma, \text{STATES}', \delta', S'_0, S_f)$  as follows:

- STATES' = { $(s, i) | s \in \text{STATES}, i = 1, ..., n$ },
- $S'_0 = \{(s,1) \mid s \in S_0\},\$
- for all  $s,t \in \text{STATES}$ , for all  $i,j \in \{1,\ldots,j\}$  and each  $\sigma \in \Sigma$  define  $(t,i) \in \delta'((s,j),a)$  if and only if  $t \in \delta(s,a)$  and i = j if  $s \notin F_j$  respectively i = j + 1mod k if  $s \in F_j$  and
- $S_f = \{(s,1) \mid s \in F_1\}.$

It is now straightforward to prove that  $w \in L(\mathcal{A})$  if and only if  $w \in L(\bar{\mathcal{A}})$  for every  $w \in \Sigma^{\omega}$ . So the lemma is proved.

Since Büchi–acceptance and generalized Büchi–acceptance are equivalent, it is sufficient to concentrate on algorithms which construct generalized Büchi–automata and apply the construction from the proof of Lemma 9.1.1 to the resulting automaton.

The usefulness of nondeterministic Büchi–automata also comes from the fact that they are closed under every boolean operation. We have the following properties.

Theorem 9.1.1 (e.g. Clarke et al. [35] and Sistla et al. [149])

1. If  $L_1 \subseteq \Sigma^{\omega}$  and  $L_2 \subseteq \Sigma^{\omega}$  are Büchi-acceptable languages, then so are

a)  $L_1 \cap L_2$  and

b)  $L_1 \cup L_2$ 

and

2. if  $L \subseteq \Sigma^{\omega}$  is Büchi–acceptable, then so is  $\overline{L} = \Sigma^{\omega} \setminus L$ .

We will not prove this theorem but we refer to section 9.2.2 for a construction of the *product* of two Büchi–automata which results in an automaton which accepts the intersection of the languages accepted by the original automata.

# 9.2. Automata Constructions for Propositional Temporal Logic Formulas

The reason for studying Büchi–automata stems from the fact that from each LTL–formula one can construct an automaton which accepts exactly the sequences of states which are models of this formula. So we will now study a language which differs from the language PROLOG(+T) studied in the last part in two ways:

- 1. it is a propositional logic based temporal logic language and
- 2. it is not limited to sets of clauses which contain at least one positive literal.

In fact we will study the full language LTL from now in. Therefore assume that a finite set X of *proposition symbols* is given. We will refer to elements of X by writing  $p, q, \ldots$  sometimes using indexes. *Formulas* of the language LTL are defined as in chapter 2.3. Note that in contrast to FOLTL we do not deal with the operator P here but instead we use the operator R.

#### 9.2.1. A Modified Formal Automata–Model

In this section we will describe a slight modification of the concept of Büchi–automata which is needed in order to allow the manipulation of such automata during the process of formula refinement as described later. In *classical* construction procedures for automata from LTL-formulas (see e.g. [172]) states are labeled with formulas from a set of subformulas of the original formula  $\varphi$ . This set of subformulas is usually referred to as the *closure* of  $\varphi$ . The problem which we face here is that certain formulas might appear more than once (and therefore at different positions) in  $\varphi$ . Consider for example the formula  $\varphi = \mathsf{GF}((\mathsf{F}p \to q) \lor (r \to \mathsf{F}p))$ . As a subformula, the formula  $\psi = \mathsf{F}p$  is treated as a single element. But for refining the formula  $\varphi$  it can make a difference if the refined formula is for example  $\varphi_1 = \mathsf{GF}((\mathsf{X}p \to q) \lor (r \to \mathsf{F}p))$  or  $\varphi_2 = \mathsf{GF}((\mathsf{F}p \to q) \lor (r \to \mathsf{X}p))$ . So we need some model which allows to store additional information regarding the positions at which certain formulas occur.

For the construction we will assume that every LTL-formula is in negation-normalform, that is it contains only the operators X, U and R, the connectives  $\land$  and  $\lor$  together with the constant symbols **true** and **false** and negations only occur in front of propositional symbols. In order to obtain the negation-normal-form NNF( $\varphi$ ) of a formula  $\varphi$ we will have to exploit semantical identities (de-Morgan's laws) and properties of the temporal operators. In particular we will need the following equivalences:

$$F\varphi \equiv trueU\varphi$$
 and  
 $G\varphi \equiv falseR\varphi$ .

and

$$\neg (\varphi_1 \mathsf{U} \varphi_2) \equiv (\varphi_1) \mathsf{R} (\varphi_2) \text{ and}$$
$$\neg (\varphi_1 \mathsf{R} \varphi_2) \equiv (\varphi_1) \mathsf{U} (\varphi_2).$$

### Example 9.2.1

For  $\varphi = \neg(p \to \mathsf{X}q)\mathsf{UF}r$  we have

$$\begin{split} \operatorname{NNF}(\varphi) &= \operatorname{NNF}(\neg(p \to \mathsf{X}q)\mathsf{UF}r) \\ &= \operatorname{NNF}(\neg(p \to \mathsf{X}q))\mathsf{UNNF}(\mathsf{F}r) \\ &= (\operatorname{NNF}(p) \land \operatorname{NNF}(\neg\mathsf{X}q)) \operatorname{UNNF}(\mathsf{trueU}r) \\ &= (p \land \mathsf{XNNF}(\neg q)) \operatorname{U}(\operatorname{NNF}(\mathsf{true})\mathsf{UNNF}(r)) \\ &= (p \land \mathsf{X}\neg q) \operatorname{U}(\mathsf{trueU}r). \end{split}$$

So assume without loss of generality that  $\varphi$  is in negation-normal-form. The *closure* of  $\varphi$  is defined to be the set  $CLOSURE(\varphi)$  of all pairs  $(\psi, p)$  where  $\psi$  is an LTL-formula and  $p \in \mathbb{N}^*$  such that  $CLOSURE(\varphi)$  satisfies the following conditions:

- $(\varphi, \varepsilon) \in \text{CLOSURE}(\varphi),$
- if  $(\varphi_1 \wedge \cdots \wedge \varphi_n, p) \in \text{CLOSURE}(\varphi)$ , then  $(\varphi_1, p1), \ldots, (\varphi_n, pn) \in \text{CLOSURE}(\varphi)$ ,
- if  $(\varphi_1 \lor \cdots \lor \varphi_n, p) \in \text{CLOSURE}(\varphi)$ , then  $(\varphi_1, p1), \ldots, (\varphi_n, pn) \in \text{CLOSURE}(\varphi)$ ,
- if  $(X\psi, p) \in CLOSURE(\varphi)$ , then  $(\psi, p1) \in CLOSURE(\varphi)$ ,
- if  $(\varphi_1 \cup \varphi_2, p) \in \text{CLOSURE}(\varphi)$ , then  $(\varphi_1, p1), (\varphi_2, p2) \in \text{CLOSURE}(\varphi)$  and
- if  $(\varphi_1 \mathsf{R} \varphi_2, p) \in \text{CLOSURE}(\varphi)$ , then  $(\varphi_1, p1), (\varphi_2, p2) \in \text{CLOSURE}(\varphi)$ .

So  $CLOSURE(\varphi)$  contains all the (not necessarily proper) subformulas of  $\varphi$  together with their positions on  $\varphi$ .

The states of the generalized Büchi–automaton  $\mathcal{A}_{\varphi}$  are now defined as certain subsets of  $\text{CLOSURE}(\varphi)$  which satisfy several semantical constraints.

#### 9.2.2. A primitive Construction

We will now see how the automaton  $\mathcal{A}_{\varphi}$  can be constructed from  $\varphi$  if  $\varphi$  is in negation– normal–form.

The *alphabet* of the automaton will be the set of subsets of symbols occurring in the original formula. The automaton will be a generalized Büchi–automaton, so it will have a set of sets of accepting states. Assume that  $\varphi$  is a formula which contains exactly the symbols of some set X. Then the alphabet of the automaton  $\mathcal{A}_{\varphi}$  is  $\Sigma = 2^X$ .

The set STATES of states is now given as the elements s from  $\text{SEQ}(\text{CLOSURE}(\varphi)) \times 2^{\mathbb{N}^* 1}$ such that  $s = (\Phi, \text{POS}) = \left(\underbrace{\{\varphi_1, \dots, \varphi_n\}}_{=\Phi}, \underbrace{\{p_1, \dots, p_m\}}_{=\text{POS}}\right)$  has the following properties:

- 1. false  $\not\in \Phi$ ,
- 2. n = m,
- 3. for each  $i: (\varphi_i, p_i) \in \text{CLOSURE}(\varphi)$ ,
- 4. for each *i*: if  $\varphi_i = \varphi_1^{(i)} \land \varphi_2^{(i)}$ , then  $\left(\varphi_1^{(i)}, p_i 1\right) \in s$  and  $\left(\varphi_2^{(i)}, p_i 2\right) \in s$ ,
- 5. for each *i*: if  $\varphi_i = \varphi_1^{(i)} \lor \varphi_2^{(i)}$ , then  $\left(\varphi_1^{(i)}, p_i 1\right) \in s$  or  $\left(\varphi_2^{(i)}, p_i 2\right) \in s$  and
- 6. for each *i*: if  $\varphi \in X \cup \{\texttt{true}\}$ , then  $\{p \in \text{Pos} \mid \varphi|_p = \varphi_i\} = \text{Occ}(\varphi_i, \varphi)$ .

Here  $(\varphi, p) \in s$  denotes the fact that there is  $j \in \{1, ..., n\}$  such that  $\varphi_j = \varphi$  and  $p_j = p$ .

What remains to be defined are the transition relation  $\delta$ , the set  $S_0$  and the set  $\mathcal{F}$  of accepting sets of states.  $\delta$  has the form

$$\delta : \text{STATES} \times 2^P \to 2^{\text{States}}.$$

Now let  $s_1, s_2 \in \text{STATES}$  be given such that  $s_1 = (\Phi, \text{Pos}) = (\{\varphi_1, \dots, \varphi_n\}, \{p_1, \dots, p_n\})$ 

<sup>&</sup>lt;sup>1</sup>Here SEQ(CLOSURE( $\varphi$ )) denotes the set of all sequences of elements from CLOSURE( $\varphi$ ).

and  $s_2 = (\Phi', \text{Pos}') = (\{\varphi'_1, \dots, \varphi'_m\}, \{p'_1, \dots, p'_m\})$ . Furthermore let  $\sigma \in \Sigma$  be fixed. Then  $s_2 \in \delta(s_1, \sigma)$  iff

- 1. for each  $p \in X$ :
  - a)  $p \in \Phi$  implies  $p \in \sigma$  and
  - b)  $\neg p \in \Phi$  implies  $p \notin \sigma$ ,
- 2. for each  $\psi = X\bar{\varphi} \in \Phi$  and each  $p_i \in \text{Pos such that } \varphi_i = \psi$  there is  $j_i \in \{1, \dots, m\}$ such that  $\varphi'_{j_i} = \bar{\varphi}$  and  $p'_{j_i} = p_i 1$ ,
- 3. for each  $\psi = \bar{\varphi_1} \cup \bar{\varphi_2} \in \Phi$  and each  $p_i \in \text{Pos such that } \varphi_i = \psi$  there is either  $j_i \in \{1, \ldots, n\}$  such that  $\varphi'_{j_i} = \bar{\varphi_2}$  or there are  $j_{i_1} \in \{1, \ldots, n\}, j_{i_2} \in \{1, \ldots, m\}$  such that  $\varphi_{j_{i_1}} = \bar{\varphi_1}$  and  $\varphi'_{j_{i_2}} = X\psi = X\bar{\varphi_1} \cup \bar{\varphi_2}$  and
- 4. for each  $\psi = \bar{\varphi_1} R \bar{\varphi_2} \in \Phi$  and each  $p_i \in Pos$  such that  $\varphi_i = \psi$  there is  $j_{i_1} \in \{1, \ldots, n\}$  such that  $\varphi_{j_{i_1}} = \bar{\varphi_2}$  and there is either  $j_{i_2} \in \{1, \ldots, n\}$  such that  $\varphi_{j_{i_2}} = \bar{\varphi_1}$  or there is  $j_{i_2} \in \{1, \ldots, m\}$  such that  $\varphi'_{j_{i_2}} = X\psi = X\bar{\varphi_1}R\bar{\varphi_2}$ .

The set  $S_0$  is defined as the set of all states such that the original formula  $\varphi$  is contained in its  $\Phi$ -component, that is

$$S_0 = \{s = (\Phi, \operatorname{Pos}) \in \operatorname{States} \mid \varphi \in \Phi\}.$$

To define the set  $\mathcal{F}$  of sets of accepting states we define the concept of *eventualities*. Eventualities are formulas which are needed in order to guarantee that given a formula  $\varphi_1 \cup \varphi_2$ , the formula  $\varphi_2$  is indeed fulfilled at some point of time. So if  $(\varphi_1 \cup \varphi_2, p) \in$  $CLOSURE(\varphi)$ , then  $e(\varphi_2) = \varphi_1 \cup \varphi_2$  is called an *eventuality*. Now assume that  $CLOSURE(\varphi)$ contains exactly the eventualities  $(e_1(\varphi_1), p_1), \ldots, (e_k(\varphi_k), p_k)$ . Then we define  $\mathcal{F} =$  $\{F_1, \ldots, F_k\}$  with

$$F_i = \{ s = (\Phi, \operatorname{Pos}) \in \operatorname{STATES} \mid \{ e_i(\varphi_i), \varphi_i \} \subseteq \Phi \text{ or } e_i(\varphi_i) \notin \Phi \}.$$

The construction of  $\mathcal{A}_{\varphi}$  is now complete. This construction can be extended from single formulas to sets of formulas in the obvious way. Let  $\Phi = \{\varphi_1, \ldots, \varphi_n\}$  be any set of LTL-formulas. Then

$$\mathcal{A}_{\Phi} = \mathcal{A}_{arphi_1} imes \cdots imes \mathcal{A}_{arphi_n}$$

where  $\times$  is a (slightly more complicated) generalization of the *product construction* from finite automata to Büchi–automata.

This construction can be carried out as follows: first we construct the set STATES from the sets of states of the input automata and after this we extract the remaining components from this set. How to achieve this is described in section 9.2.4.

Now let  $\mathcal{A}_{\varphi_1} = (\Sigma, \text{STATES}_1, \delta_1, S_{0,1}, \mathcal{F}_1)$  and  $\mathcal{A}_{\varphi_2} = (\Sigma, \text{STATES}_2, \delta_2, S_{0,2}, \mathcal{F}_2)$  be Büchi-automata representing LTL-formulas  $\varphi_1$  and  $\varphi_2$ . In order to respect the positions of the original formulas in the *new* formula  $\varphi_1 \wedge \varphi_2$  we have to change the positions in the original states from STATES<sub>1</sub> and STATES<sub>2</sub>. Therefore we replace STATES<sub>1</sub> and STATES<sub>2</sub> by

$$\bigcup_{s \in \text{States}_1} \left( \Phi(s), \bigcup_{p \in \text{Pos}(s)} 1p \right)$$

and

$$\bigcup_{s \in \text{States}_2} \left( \Phi(s), \bigcup_{p \in \text{Pos}(s)} 2p \right).$$

Furthermore assume that

STATES<sub>1</sub> = 
$$\left\{ s_1^{(1)}, \dots, s_{n_1}^{(1)} \right\}$$
 and  
STATES<sub>2</sub> =  $\left\{ s_1^{(2)}, \dots, s_{n_2}^{(2)} \right\}$ 

and

$$S_{0,1} = \left\{ s_{i_1}^{(1)}, \dots, s_{i_k}^{(1)} \right\}$$
 and

$$S_{0,2} = \left\{ s_{j_1}^{(2)}, \dots, s_{2_k}^{(2)} \right\}$$

Then  $\mathcal{A} = \mathcal{A}_{\varphi_1} \times \mathcal{A}_{\varphi_2}$  is the Büchi–automaton  $\mathcal{A} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  which has a set STATES of states constructed as follows:

$$S_{0} = \bigcup_{k_{1}=1}^{k} \bigcup_{k_{2}=1}^{l} \left\{ \begin{pmatrix} \Phi\left(s_{i_{k_{1}}}^{(1)}\right) \cup \Phi\left(s_{j_{k_{2}}}^{(2)}\right) \cup \{\varphi_{1} \land \varphi_{2}\}, \\ \operatorname{Pos}\left(s_{i_{k_{1}}}^{(1)}\right) \cup \operatorname{Pos}\left(s_{j_{k_{2}}}^{(2)}\right) \cup \{\varepsilon\} \end{pmatrix} \right\},$$

$$S = \bigcup_{i=1}^{n_{1}} \bigcup_{j=1}^{n_{2}} \left\{ \left( \Phi\left(s_{i}^{(1)}\right) \cup \Phi\left(s_{j}^{(2)}\right), \operatorname{Pos}\left(s_{i}^{(1)}\right) \cup \operatorname{Pos}\left(s_{j}^{(2)}\right) \right) \right\} \text{ and}$$
STATES =  $S_{0} \cup S$ ,

while the remaining components of the automaton (i.e. the transition relation  $\delta$  and the acceptance component  $\mathcal{F}$ ) have to be extracted from STATES (see the algorithms in section 9.2.4 for details).

This construction yields an automaton which accepts the language  $L(\mathcal{A}_{\varphi_1}) \cap L(\mathcal{A}_{\varphi_2})$ . Furthermore we have

$$|\text{STATES}| = |\text{STATES}_1| \cdot |\text{STATES}_2| + |S_{0,1}| \cdot |S_{0,2}|$$
 and  
 $|S_0| = |S_{0,1}| \cdot |S_{0,2}|.$ 

In order to prove that for a set  $\Phi$  of LTL-formulas and an LTL-formula  $\varphi$  the relation  $\Phi \models \varphi$  holds, we proceed as follows:

- 1. For  $\Phi = \{\varphi_1, \ldots, \varphi_n\}$  we construct  $\mathcal{A}_{\Phi} = \mathcal{A}_{\varphi_1} \times \cdots \times \mathcal{A}_{\varphi_n}^2$ ,
- 2. construct  $\mathcal{A}_{\neg\varphi}$ ,
- 3. construct  $\mathcal{A} = \mathcal{A}_{\Phi} \times \mathcal{A}_{\neg \varphi}$  and

<sup>&</sup>lt;sup>2</sup>Note that it is also possible to construct  $\mathcal{A}_{\Phi} = \mathcal{A}_{\bigwedge_{i=1}^{n} \varphi_{i}}$  directly (i.e. without using the product construction).

4. check if  $L(\mathcal{A}) = \emptyset$ .

For some of the algorithms which we will introduce in a later chapter we will also need a method for constructing the *union* of two Büchi–automata. Again assume that  $\varphi_1$  and  $\varphi_2$  are given and assume that  $\mathcal{A}_{\varphi_1}$  and  $\mathcal{A}_{\varphi_2}$  are as above. Then the Büchi– automaton  $\mathcal{A}_{\varphi_1} || \mathcal{A}_{\varphi_2}$  representing  $\varphi_1 \vee \varphi_2^3$  can be constructed by first modifying STATES<sub>1</sub> and STATES<sub>2</sub> as described above and then constructing the set STATES by

$$S_{0} = \bigcup_{k_{1}=1}^{k} \bigcup_{k_{2}=1}^{l} \left\{ \left( \Phi\left(s_{i_{k_{1}}}^{(1)}\right) \cup \Phi\left(s_{j_{k_{2}}}^{(2)}\right) \cup \{\varphi_{1} \land \varphi_{2}\}, \operatorname{Pos}\left(s_{i_{k_{1}}}^{(1)}\right) \cup \operatorname{Pos}\left(s_{j_{k_{2}}}^{(2)}\right) \cup \{\varepsilon\} \right) \right\}$$

and

$$STATES = S_0 \cup STATES_1 \cup STATES_2.$$

The automaton  $\mathcal{A}_{\varphi_1} || \mathcal{A}_{\varphi_2}$  is then given as  $(\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  (again with  $\delta$  and  $\mathcal{F}$  extracted from STATES). It is easily seen that this construction is sound. Furthermore we have

$$|STATES| = |S_{0,1}| \cdot |S_{0,2}| + |STATES_1| + |STATES_2|$$

#### 9.2.3. An Overview over improved Constructions

The primitive construction presented in the last section always yields an automaton whose state set is of size exponential in the length of the input formula. We will therefore give an overview over several optimization techniques which allow the construction of *smaller* automata.

**Removing Transitions** A transition which is not explicitly needed (and which is therefore *redundant*) in the automaton can be deleted. Such transitions can be identified as follows: if there is  $s \in \text{STATES}$ ,  $\sigma \in \Sigma$ ,  $s_1 = (\Phi_1, \text{Pos}_1) \in \text{STATES}$  and  $s_2 = (\Phi_2, \text{Pos}_2) \in \text{STATES}$  such that

<sup>&</sup>lt;sup>3</sup>That is:  $\mathcal{A}_{\varphi_1} || \mathcal{A}_{\varphi_2} = \mathcal{A}_{\varphi_1 \vee \varphi_2}.$ 

- 1.  $s_1 \in \delta(s, \sigma), s_2 \in \delta(s, \sigma)$  and
- 2. for every  $\varphi_1 \mathsf{U} \varphi_2 \in \Phi_1$ :  $\varphi_1 \mathsf{U} \varphi_2 \in \Phi_2$  and  $\varphi_2 \in \Phi_1$  implies  $\varphi_2 \in \Phi_2$ ,

then the transition from s to  $s_2$  can be deleted from the automaton (see [172] for a justification of this optimization).

- Eliminating equivalent states By identifying sets of formulas which are in some sense equivalent it is possible to lower the number of states. Consider for example the state s given by  $s = (\{\varphi_1, \varphi_2, \varphi_2 \land \varphi_2\}, \text{Pos})$ . This state has the property that  $\Phi \models \varphi$  iff  $\Phi' = \Phi \setminus \{\varphi_1 \land \varphi_2\} = \{\varphi_1, \varphi_2\} \models \varphi$  for every LTL-formula  $\varphi$ . This allows us to remove the original state s and replace it with some state  $s' = (\Phi', \text{Pos}')$ . In [76], [172] and several other papers the following improvements have been discussed:
  - 1.  $\{\varphi_1, \varphi_2, \varphi_2 \land \varphi_2\} \rightarrow \{\varphi_1, \varphi_2\},\$
  - 2.  $\{\varphi_1, \varphi_1 \lor \varphi_2\} \to \{\varphi_1\},\$
  - 3.  $\{\varphi_2, \varphi_1 \lor \varphi_2\} \to \{\varphi_2\}$  and
  - 4.  $\{\varphi_2, \varphi_1 \mathsf{U}\varphi_2\} \to \{\varphi_1\}.$

Several other simplification techniques may be applicable. We do not give a more in depth-treatment here since we do not need all these techniques in the sequel.

A number of approaches has been introduced for the construction of Büchi–automata from LTL–formulas. Probably the most straightforward and simple construction (which is the basis for our automaton model) can for example be seen in [172] although its origin comes from [168] and [167] continuing the work originally started by Büchi in [26]. The early constructions of Büchi–automata for LTL–formulas were of exponential size in the size of the original formula. More sophisticated constructions have been developed in [76], [39], [150] and [72]. [17] and [60] introduce similar approaches for the problem of LTL model checking which do not construct Büchi–automata directly but which use similar techniques and results. Besides constructing automata several other approaches for checking LTL-formulas for satisfiability resp. unsatisfiability have been presented. An early paper by Venkatesh (see [169]) describes the construction of a normal form for LTL-formulas which is then given to a resolution-style theorem proving procedure as an input. Similarly Fisher (see [64]) introduces another normal form called *separated normal form* for formulas including future and past operators. Again the theorem proving procedure is based on resolution. Another treatment of this procedure can be found in a paper by Dixon, Fisher and Peim (see [65]). Dixon (see [48] and [47]) and Dixon and Fisher (see [49]) also addressed the topic of speeding up theorem proving procedures in order to improve the satisfiability tests.

Another approach for LTL is presented by Felty (see [61]). Here the calculus for checking formulas is based on the sequent calculus originally introduced by Gentzen (see e.g. [75] or [89]). A further paper extends this sequent calculus from LTL to the modal logic S4.3 (see [62]).

The third popular approach is based on tableaux-style techniques similar to our techniques from chapter 5.3. This technique has been developed by Manna and Wolper (see [171] and [111]) as well as by Lichtenstein, Pnueli and Zuck (see [102] and [103]). Good surveys of tableaux techniques have been presented by Emerson (see [57]) and Reynolds and Dixon (see [139]).

All these approaches have their own powers and weaknesses. But for our purposes the automata-based approach seems to be the most promising one as it allows the generalization and specialization of given formulas from their representing automata as we will see in chapter 10.

#### 9.2.4. Some Complexity Results

In order to estimate the complexity of the refinement procedures to be introduced in the following chapter we will present some results regarding the complexity of some *basic*  operations on Büchi–automata. Assume that STATES is a set of sates which has been constructed. We will sketch the complexity of extracting  $\delta$ ,  $S_0$  and  $\mathcal{F}$  from STATES.

#### **Extracting the Transition Relation**

Obviously the complexity of the extraction of  $\delta$  from STATES has to depend on the number of states, i.e. |STATES| and the number of elements in the alphabet of the automaton, i.e.  $|\Sigma| = |2^X| = 2^{|X|}$ . Algorithm 11 is a straightforward implementation of the definition of the transition relation for Büchi-automata. For notational simplicity we will use the following abbreviation. If  $s = (\Phi, \text{Pos})$  is a state, then  $\Phi(s)$  will denote the set of formulas stored in s, i.e.  $\Phi(s) = \Phi$ . Furthermore assume that  $\delta(s, \sigma) = \emptyset$  holds as an initial condition.

We will now give a detailed analysis of the runtime of Algorithm 11. Therefore assume that  $n_{\max}$  denotes the maximum number of formulas stored in any element of STATES, that is  $n_{\max} = \max \{ |\Phi| \mid \Phi = \Phi(s), s \in \text{STATES} \}$ . Furthermore we will assume that each check of the form  $\varphi \in \Phi(s)$  is atomic, i.e. is can be performed in *one* computation step and checks performed in conditions are performed one-by-one, that is a check which involves *n* subchecks requires *n* computation steps.

The part of Algorithm 11 between line 4 and line 11 can then be preformed in

$$T_1(|\text{STATES}|, |\Sigma|) \leq 6|P|$$
  
=  $6 \log_2 |\Sigma|$ 

computation steps.

Similarly we have that the part between lines 12 and 32 can be performed in

$$T_2(|\text{STATES}|, |\Sigma|) \leq n_{\max} \cdot (1 + 3n_{\max} + 3 + 4)$$
$$= n_{\max} \cdot (8 + 3n_{\max})$$

Algorithm 11 Extraction of $\delta$ from $\Sigma$ and STATES	
<b>Input</b> : set STATES of states and alphabet $\Sigma = 2^X$	
<b>Output</b> : transition relation $\delta$	
1: for each $s_1 \in \text{STATES}$ do	
2: for each $s_2 \in \text{STATES}$ do	
3: for each $\sigma \in \Sigma$ do	
4: for each $p \in P$ do	
5: if $p \in \sigma$ and $p \in \Phi(s_1)$ then	
6: $\delta(s_1, \sigma) \leftarrow \delta(s_1, \sigma) \cup \{s_2\}$	
7: end if	
8: if $p \notin \sigma$ and $\neg p \in \Phi(s_1)$ then	
9: $\delta(s_1, \sigma) \leftarrow \delta(s_1, \sigma) \cup \{s_2\}$	
10: end if	
11: end for	
Require: $s_1 = (\{\varphi_1, \dots, \varphi_n\}, \{p_1, \dots, p_n\})$	
<b>Require:</b> $s_2 = (\{\varphi'_1, \dots, \varphi'_m\}, \{p'_1, \dots, p'_m\})$	
12: for $i = 1, \dots, n$ do	
13: if $\varphi_i = X \psi$ then	
14: for $j = 1, \dots, m$ do	
15: if $\varphi'_j = \psi$ and $p'_j = p_i 1$ then	
16: $\delta(s_1, \sigma) \leftarrow \delta(s_1, \sigma) \cup \{s_2\}$	
17:     end if       18:     end for	
19: end if 20: if $\varphi_i = \overline{\varphi_1} \cup \overline{\varphi_2}$ then	
21: If $\varphi_1 = \varphi_1 \oplus \varphi_2$ then 21: if $\varphi_2 \in \Phi(s_1)$ or $\varphi_1 \in \Phi(s_1)$ and $X\psi \in \Phi(s_2)$ then	
21: If $\varphi_2 \in \varphi(s_1)$ of $\varphi_1 \in \varphi(s_1)$ and $\forall \varphi \in \varphi(s_2)$ then 22: $\delta(s_1, \sigma) \leftarrow \delta(s_1, \sigma) \cup \{s_2\}$	
22: $0(s_1, 0) + 0(s_1, 0) = [s_2]$ 23: end if	
24: end if	
25: <b>if</b> $\varphi_i = \overline{\varphi_1} R \overline{\varphi_2}$ then	
26: <b>if</b> $\overline{\varphi_2} \in \Phi(s_1)$ then	
27: if $\varphi_1 \in \Phi(s_1)$ or $X\psi \in \Phi(s_2)$ then	
28: $\delta(s_1, \sigma) \leftarrow \delta(s_1, \sigma) \cup \{s_2\}$	
29: end if	
30: end if	
31: end if	
32: end for	
33: end for	
34: end for	
35: end for	
36: return $\delta$	

$$= 8n_{\max} + 3n_{\max}^2$$

0

steps.

The complexity of the complete algorithm can therefore be estimated as follows:

$$T(|\text{STATES}|, |\Sigma|) \leq |\text{STATES}|^2 \cdot |\Sigma| \cdot (T_1(|\text{STATES}|, |\Sigma|) + T_2(|\text{STATES}|, |\Sigma|))$$
  
$$= |\text{STATES}|^2 \cdot |\Sigma| \cdot (6 \log_2 |\Sigma| + 8n_{\max} + 3n_{\max}^2)$$
  
$$\in \mathcal{O}\left(n_{\max}^2 \cdot |\text{STATES}|^2 \cdot |\Sigma| \cdot \log_2 |\Sigma|\right).$$

So we have the following theorem.

#### Theorem 9.2.1

Let STATES and  $\Sigma$  be given. Then  $\delta$ : STATES  $\times \Sigma \to 2^{\text{STATES}}$  can be constructed in time

$$\mathcal{O}\left(n_{\max}^2 \cdot |\mathrm{STATES}|^2 \cdot |\Sigma| \cdot \log_2 |\Sigma|\right).$$

#### **Extracting the Initial States**

Extracting the initial states from STATES is the simplest task. By definition every  $s = (\Phi, \text{Pos}) \in S_0$  is such that  $\varphi \in \Phi$ . Then a simple linear search strategy can check if s is indeed contained in  $S_0$ . The runtime of such a check is bounded from above by  $n_{\text{max}}$ . Consequently checking every  $s \in \text{STATES}$  can be done in time  $\mathcal{O}(n_{\text{max}} \cdot |\text{STATES}|)$ .

#### Theorem 9.2.2

Let STATES be given. Then  $S_0 \subseteq$  STATES can be extracted in time  $\mathcal{O}(n_{\max} \cdot |\text{STATES}|)$ .

#### **Extracting the Acceptance Component**

For the extraction of the acceptance component  $\mathcal{F}$  it is necessary to *collect* the eventualities which are included in the states. This can be accomplished in  $n_{\max} \cdot |\text{STATES}|$  steps yielding a set EV of pairs  $(e_i(\varphi_i), p_i)$ . The size of EV is bounded from above by  $n_{\max} \cdot |\text{STATES}|$ . Following the definition of the acceptance component we can compute  $\mathcal{F}$  using Algorithm 12.

#### Algorithm 12 Extracting the Acceptance Component $\mathcal{F}$ from STATES

**Input**: set STATES of states **Output**: acceptance component  $\mathcal{F}$ 1:  $\mathcal{F} \leftarrow \emptyset$ 2: compute Ev (as described) 3: for each  $(e_i(\varphi_i), p_i) \in \text{Ev do}$  $F \leftarrow \emptyset$ 4:for each  $s \in \text{STATES}$  do 5:if  $e_i(\varphi_i) \in \Phi(s)$  and  $\varphi \in \Phi(s)$  or  $e_i(\varphi_i) \notin \Phi(s)$  then 6:  $F \leftarrow F \cup \{s\}$ 7: end if 8: 9: end for  $\mathcal{F} \leftarrow \mathcal{F} \cup \{F\}$ 10:11: end for 12: return  $\mathcal{F}$ 

The time complexity of Algorithm 12 can be estimated as follows:

$$T(|\text{STATES}|) \leq 1 + n_{\max} \cdot |\text{STATES}| + |\text{Ev}| \cdot (1 + |\text{STATES}| \cdot (2 + 1) + 1) + 1$$
  
$$= 2 + n_{\max} \cdot |\text{STATES}| + |\text{Ev}| \cdot (2 + 3 \cdot |\text{STATES}|)$$
  
$$\leq 2 + n_{\max} \cdot |\text{STATES}| + n_{\max} \cdot |\text{STATES}| \cdot (2 + 3 \cdot |\text{STATES}|)$$
  
$$= 2 + 3n_{\max} \cdot |\text{STATES}| + 3n_{\max} \cdot |\text{STATES}|^2$$
  
$$\in \mathcal{O}\left(n_{\max} \cdot |\text{STATES}|^2\right).$$

So we have the following theorem.

#### Theorem 9.2.3

Let STATES be given. Then the acceptance component  $\mathcal{F}$  can be extracted from STATES in time  $\mathcal{O}(n_{\max} \cdot |\text{STATES}|^2)$ .

#### 9.2.5. Checking Language–Emptiness

We have already mentioned that it will be necessary to check the emptiness of the languages accepted by Büchi–automata. In this section we will see how this can be achieved.

Let G = (V, E) be a directed graph and let  $V' \subseteq V$  be a nonempty set of vertices from V and let E' be a nonempty set of edges from E. Then the subgraph G' = (V', E') is called *maximal strongly connected* if

- for each pair  $n_1, n_2$  of vertices from V' it holds that  $n_2$  is reachable from  $n_1$  and  $n_1$  is reachable from  $n_2$  and
- V' is maximal wrt. ⊆, that is there is no v ∈ V \ V' such that (V' ∪ {v}, E') is strongly connected.

The set of all strongly connected subgraphs of G is called the set of maximal strongly connected components of G. The maximal strongly connected components of G form a partition of G into disjoint subsets. A maximal strongly connected component (V', E')is called non-trivial if either |V'| > 1 or  $V' = \{v\}$  for some v and  $(v, v) \in E'$ .

The maximal strongly connected components of a graph can be computed in time  $\mathcal{O}(|V|)$  (see [156] and [74]).

The link to our problem is given as follows: Obviously every extended Büchi-automaton  $\mathcal{A} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  induces a directed Graph  $G_{\mathcal{A}} = (\text{STATES}, E)$  where for every pair  $s_1, s_2 \in \text{STATES}$  the edge  $(s_1, s_2)$  belongs to E if there is an element  $\sigma \in \Sigma$  such that  $s_2 \in \delta(s_1, \sigma)$ . For checking that  $L(\mathcal{A}) = \emptyset$  it suffices to compute the maximal strongly connected components of  $G_{\mathcal{A}}$  and check if there is a maximal strongly connected component S which is reachable from some initial state  $s_0 \in S_0$  such that  $S \cap F \neq \emptyset$  for every  $F \in \mathcal{F}$ . So checking logical implication of some property  $\varphi$  in a system given by an LTL-program P can be reduced to the computation of the maximal strongly connected components of  $\mathcal{A}_P \times \mathcal{A}_{\neg \varphi}$  and a simple containment check.

# 10. Automata Manipulations

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10.5. The Identification Process

This chapter will deal with the procedures which are necessary in order to refine LTL– programs. As we have seen in the last chapter, we are always able to construct a generalized Büchi–automaton from a temporal logic formula which has a nonempty accepted language if and only if the formula from which the automaton had been constructed is satisfiable. The construction relies on the set of subformulas occurring in the original formula. So a state is labeled with a set of formulas together with their positions. Now assume that a modified Büchi–automaton  $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  constructed from a LTL–formula  $\varphi$  is given. By definition we have  $\varphi \in \Phi(s)$  for every  $s \in S_0$ . Now assume further that any LTL–formula e is given (an example). Then this example can be a positive one or a negative one. In either case we can construct the representing automaton  $\mathcal{A}_{\neg e}$ .

- 1. If e is a positive example, then a modification of the model under consideration has to be carried out if  $L(\mathcal{A}_{\varphi} \times \mathcal{A}_{\neg e}) \neq \emptyset$  since in this case the positive example e is not implied by the model under consideration.
- 2. If e is a negative example, then consequently we have to refine the model if  $L(\mathcal{A}_{\varphi} \times \mathcal{A}_{\neg e}) = \emptyset$  since in this case the negative example e is implied.

In either case we will have to modify the states of the representing automaton  $\mathcal{A}_{\varphi}$  in such a way that the modified model is compatible with the new example e.

#### 10.1. Implication as an Ordering

In the case of FOLTL, we have used the concept of subsumption as the basis for generalization and specialization operations. In LTL this would not be a good choice for two reasons:

- 1. LTL is a propositional temporal logic language, so the concept of substitutions would not make any sense since there are no variable symbols to substitute and
- PROLOG(+T)-objects are essentially clauses. The results regarding greatest specializations and least generalizations are only concerned with such clauses. In LTL the objects are not limited to clauses and so we cannot hope to apply the techniques established in chapter 6.

So we see that subsumption is not applicable here. However, since LTL is propositional, we know that the logical consequence relation  $\models$  is decidable. This enables us to use a finer relation than the subsumption ordering  $\geq$ , namely the *implication ordering*.

#### Definition 10.1.1

Let  $\varphi_1$  and  $\varphi_2$  be LTL formulas. Then the ordering  $\succeq$  is defined as  $\varphi_1 \succeq \varphi_2$  if and only if  $\varphi_1 \models \varphi_2$ .

The notations  $\precsim$ ,  $\succ$ ,  $\prec$  are defined in the usual way. In the case of  $\varphi_1 \succeq \varphi_1$  and  $\varphi_2 \succeq \varphi_1$  we will not introduce a new symbol since in this case  $\varphi_1 \equiv \varphi_2$ .

#### 10.2. Upward Refinement

We will now show how LTL-formulas can be refined upwards, that is we will see how we can construct a formula  $\psi$  from a given formula  $\varphi$  such that  $\psi \succsim \varphi$  where  $\psi$  is in some sense *minimal*. This concept of minimality will be made precise now.

#### Definition 10.2.1

Let  $\varphi$  be an LTL-formula. An LTL-formula  $\psi$  is called a *minimal upward refinement* wrt.  $\succeq$  of  $\varphi$  if 1.  $\psi \succeq \varphi$  and

2. there is no LTL-formula  $\psi'$  such that  $\psi \succ \psi' \succ \varphi$ .

Our concept of minimal refinements is identical to the concept of *covers* which is a well-known concept from the theory of ILP (see [126] for example). In other words, a formula  $\psi$  is a minimal upward refinement wrt.  $\succsim$  of a formula  $\varphi$  if and only if  $\psi$  is an upward cover of  $\varphi$  (with respect to the ordering  $\geq$ ). In particular all properties of covers also hold for minimal refinements.

We will now show how upward covers can be constructed. Therefore assume that  $\varphi$ is a fixed LTL-formula in which propositional variables from the set  $X = \{p_1, \ldots, p_m\}$  occur. The construction of refinements of  $\varphi$  depends on the question if there are temporal operators involved in  $\varphi$ .

#### 10.2.1. Formulas without Temporal Operators

Assume that  $\varphi$  is a purely propositional logic formula. The idea of how to refine  $\varphi$  is to construct a formula  $\psi$  which has nearly the same models as  $\varphi$ . To obtain such a formula we need a special kind of formulas, namely so called maximal minterms.

#### Definition 10.2.2

Let X be a set of propositional variable symbols and let  $\varphi$  be a formula in which exactly the variables from  $X = \{p_1, \ldots, p_m\}$  occur. Then  $\varphi$  is called a *maximal minterm* if and only if

1. 
$$\varphi = \bigwedge_{i=1}^{|X|} \psi_i$$

2. for every  $i \in \{1, ..., k\}$  it holds that  $\psi_i \in X$  or  $\psi_i \in \{\neg p \mid p \in X\}$  and 3. there is no pair  $i_0, i_1$  such that  $\psi_{i_0} \equiv \neg \psi_1$ .

The set of all maximal minterms containing variables from X will be denoted as  $\operatorname{MinTerms}(X)$ 

#### Theorem 10.2.1

Let  $\varphi$  be a propositional logic formula containing variables from a finite set X of propositional symbols. If  $\varphi$  is satisfiable, then for every  $\chi \in MINTERMS(X)$  with  $\varphi \not\models \neg \chi$  a minimal upward refinement of  $\varphi$  is given by  $\varphi \wedge \neg \chi$ .

**Proof.** Assume that  $\varphi$  and X are given as required. First we observe that  $\varphi \wedge \neg \chi \models \varphi$ for every  $\chi \in MINTERMS(X)$ , that is  $\varphi \wedge \neg \chi \succeq \varphi$  holds. Now assume that  $\chi$  is chosen such that  $\varphi \not\models \neg \chi$ . Assume that  $\varphi \land \neg \chi \not\succ \varphi$ , that is assume that  $\varphi \succeq \varphi \land \neg \chi$ . Then  $MD(\varphi = MD(\varphi \land \neg \chi) = MD(\varphi) \cap MD(\neg \chi)$  or equivalently  $MD(\neg \chi) \supseteq MD(\varphi)$ . But this gives  $\varphi \models \neg \chi$  contradicting the assumptions on  $\varphi$  and  $\chi$ .

It remains to prove that the formula  $\varphi \wedge \neg \chi$  is indeed a minimal refinement. Assume that this is *not* the case, that is assume that there is a formula  $\alpha$  such that  $\varphi \wedge \neg \chi \succ \alpha \succ \varphi$ . Then we have

- 1.  $MD(\varphi \land \neg \chi) \subset MD(\alpha) \subset MD(\varphi)$  and
- 2.  $|MD(\neg \chi)| = 2^{|X|} 1$  since  $\chi \in MINTERMS(X)$  and therefore  $|MD(\chi)| = 1$ .

This gives  $|MD(\varphi \wedge \neg \chi)| \leq |MD(\alpha)| \leq |MD(\varphi)|$  since X is assumed to be finite. We can distinguish two cases:

- **Case 1**  $\varphi \models \neg \chi$ . Then  $|MD(\varphi \land \neg \chi)| = |MD(\varphi)|$  and therefore  $|MD(\alpha)| = |MD(\varphi)|$ which gives  $\alpha \equiv \varphi$  and in particular  $\alpha \not\succ \varphi$  which is a contradiction.
- **Case 2**  $\varphi \not\models \neg \chi$ . Then we have  $|MD(\varphi \land \neg \chi)| = |MD(\varphi)| 1 < |MD(\varphi)|$ . But since  $MD(\alpha) \subset MD(\varphi)$  and  $MD(\varphi \land \neg \chi) \subset MD(\alpha)$  this gives either  $|MD(\alpha)| = |MD(\varphi)| 1 = |MD(\varphi \land \neg \chi)|$  which yields  $\alpha \equiv \varphi \land \neg \chi$  or  $|MD(\alpha)| = |MD(\varphi)|$  which gives  $\alpha \equiv \varphi$ . In the former case we have  $\varphi \land \neg \chi \not\succeq \alpha$  and in the latter case we have  $\alpha \not\succ \varphi$ . So both cases yield a contradiction.

Since every case yields a contradiction, such a formula  $\varphi$  cannot exist and the claim is proved.

For the sake of simplicity we will introduce a special mapping  $\Psi_u$  containing all upward refinements of a formula  $\varphi$ , that is

$$\Psi_u(\varphi) = \{\varphi \land \neg \chi \mid \chi \in \operatorname{MinTerms}(X)\}.$$

#### Example 10.2.1

Let  $X = \{p_1, p_2, p_3\}$  and  $\varphi = p_1 \rightarrow (p_2 \rightarrow p_3)$  be given. Then

$$MINTERMS(X) = \begin{cases} \neg p_1 \land \neg p_2 \land \neg p_3, \\ \neg p_1 \land \neg p_2 \land p_3, \\ \neg p_1 \land p_2 \land \neg p_3, \\ \neg p_1 \land p_2 \land \neg p_3, \\ p_1 \land \neg p_2 \land \neg p_3, \\ p_1 \land \neg p_2 \land \neg p_3, \\ p_1 \land p_2 \land \neg p_3, \\ p_1 \land p_2 \land \neg p_3, \\ p_1 \land p_2 \land p_3, \end{cases}$$

and

$$\Psi_{u}(\varphi) = \{(p_{1} \to (p_{2} \to p_{3})) \land \neg \chi \mid \chi \in \operatorname{MINTERMS}(X)\} \\
= \begin{cases}
(p_{1} \to (p_{2} \to p_{3})) \land (p_{1} \lor p_{2} \lor p_{3}), (p_{1} \to (p_{2} \to p_{3})) \land (p_{1} \lor p_{2} \lor \neg p_{3}) \\
(p_{1} \to (p_{2} \to p_{3})) \land (p_{1} \lor \neg p_{2} \lor p_{3}), (p_{1} \to (p_{2} \to p_{3})) \land (p_{1} \lor \neg p_{2} \lor \neg p_{3}) \\
(p_{1} \to (p_{2} \to p_{3})) \land (\neg p_{1} \lor p_{2} \lor p_{3}), (p_{1} \to (p_{2} \to p_{3})) \land (\neg p_{1} \lor p_{2} \lor \neg p_{3}) \\
(p_{1} \to (p_{2} \to p_{3})) \land (\neg p_{1} \lor \neg p_{2} \lor p_{3}), (p_{1} \to (p_{2} \to p_{3})) \land (\neg p_{1} \lor \neg p_{2} \lor \neg p_{3})
\end{cases}$$

#### 10.2.2. Formulas with Temporal Operators

In the foregoing section we have described how to refine purely propositional LTLformulas. There is one problem with this approach: consider the formula  $\varphi = p_1 \wedge p_2$ . The refinement procedures for propositional formulas construct formulas such as  $(p_1 \wedge p_2) \wedge (\neg p_1 \vee \neg p_2), (p_1 \wedge p_2) \wedge (\neg p_1 \vee p_2), \ldots$  but if the formula to be identified is for example  $\psi = \mathsf{G}p_1 \wedge \mathsf{G}p_2$ , then this formula will never be constructed. In order to overcome this limitation we will have to describe how to *temporally generalize* formulas. For the case of upward refinement, which will be discussed in this section, consider the following formula  $\varphi = \mathsf{F}p$  for some propositional variable symbol  $p \in X$ . Obviously a temporal interpretation  $\mathcal{J} = (s_0, s_1, ...)$  is a model of  $\varphi$  if there is an index j such that  $\mathcal{J}^j \models \varphi$ (or equivalently if  $\mathcal{J} \models \mathsf{X}^j \varphi$ ). So we have that  $p \models \mathsf{F}p$ . More general  $\mathsf{X}^j p \models \mathsf{F}p$  for every  $j \ge 0$ , that is  $\mathsf{X}^j p \succeq \mathsf{F}p$  for every  $j \in \mathbb{N}$ . So an upward refinement of any formula  $\mathsf{F}\psi$ can be constructed as  $\mathsf{X}^j \psi$ . This approach can be continued: obviously  $\mathsf{G}\psi \models \mathsf{X}^j\psi$ , that is  $\mathsf{G}\psi \succeq \mathsf{X}^j\psi \succeq \mathsf{F}\psi$ . We will now examine how this idea of temporal refinement can be combined with the refinement procedure described for propositional formulas.

Therefore assume that any LTL-formula  $\varphi$  is given. For the formulas  $\varphi_1$  and  $\varphi_2$  we will require that  $\psi \succeq \varphi_1$  respectively  $\psi \succeq \varphi_2$  for every formula  $\psi \in \Gamma_u(\varphi_1)$  respectively  $\psi \in \Gamma_u(\varphi_2)$ . Then the set  $\Gamma_u(\varphi)$  is defined as follows:

$$\begin{split} \Gamma_{u}(\varphi) &= \Psi_{u}(\varphi) \\ &\cup \left\{ \varphi[\mathsf{X}^{i}\chi]_{p} \mid p \in \operatorname{Pos}(\varphi), \varphi|_{p} = \mathsf{F}\chi \text{ for some } i \geq 0 \right\} \\ &\cup \left\{ \varphi[\mathsf{G}\chi]_{p} \mid p \in \operatorname{Pos}(\varphi), \varphi|_{p} = \mathsf{X}^{i}\chi \text{ for some } i \geq 0 \right\} \\ &\cup \left\{ \varphi[\bar{\varphi}_{1} \star \varphi_{2}]_{p}, \varphi[\varphi_{1} \star \bar{\varphi}_{2}]_{p} \mid \stackrel{p \in \operatorname{Pos}(\varphi), \varphi|_{p} = \varphi_{1} \star \varphi_{2}, \bar{\varphi}_{1} \in \Gamma_{u}(\varphi_{1}), \right\} \\ &\cup \left\{ \varphi[\bar{\varphi}_{1} \star \varphi_{2}]_{p}, \varphi[\varphi_{1} \star \bar{\varphi}_{2}]_{p} \mid \stackrel{p \in \operatorname{Pos}(\varphi), \varphi|_{p} = \varphi_{1} \star \varphi_{2}, \bar{\varphi}_{1} \in \Gamma_{u}(\varphi_{1}), \right\} . \end{split}$$

For proving the properties of the formulas from  $\Gamma_u(\varphi)$  we will exploit the following observation:

#### Lemma 10.2.1 (Replacement-Lemma, Upward-Version)

Let  $\varphi$  and  $\psi$  be LTL-formulas, let  $p \in \text{Pos}(\varphi)$  be a position in  $\varphi$ . If  $\psi \succeq \varphi|_p$ , then  $\varphi[\psi]_p \succeq \varphi$ .

**Proof.** Assume that  $\varphi|_p \succeq \psi$ , that is  $\varphi|_p \models \psi$ . Let  $\mathcal{J} \in MD(\varphi)$  be a fixed model of  $\varphi$ . We will prove that  $\mathcal{J} \models \varphi[\psi]_p$ .

**Case 1** If  $\mathcal{J} \models \varphi|_p$ , then  $\mathcal{J} \models \psi$  since  $\varphi|_p \succeq \psi$ . But in this case we also have  $\mathcal{J} \models \varphi[\psi]_p$ since  $\mathcal{J}(\varphi|_p) = \mathcal{J}(\psi)$ .

**Case 2** If  $\mathcal{J} \not\models \varphi|_p$ , then we have to distinguish how  $\mathcal{J}$  evaluates  $\psi$ . If  $\mathcal{J} \not\models \psi$ , then

the claim is obvious (using the same argumentation as in the foregoing case). So assume that  $\mathcal{J} \models \psi$ . Then we have two subcases:

**Case 2.1** If  $\varphi|_p$  is positive in  $\varphi$ , then the claim is immediately.

**Case 2.2** If  $\varphi|_p$  is negative in  $\varphi$ , then we have  $\varphi|_p = \neg q$  for some propositional variable symbol q since  $\varphi$  is assumed to be in negation normal form. But the assumption  $\varphi|_p \succeq \psi$  yields  $\psi \in \{\texttt{true}, \neg q\}$  and so the claim follows.

This proves the lemma.

It is now straightforward to prove that every formula contained in the set  $\Gamma_u(\varphi)$  is a generalization of  $\varphi$  with respect to the ordering  $\succeq$ .

#### **Theorem 10.2.2**

For every LTL-formula  $\varphi$  and every  $\psi \in \Gamma_u(\varphi)$  it holds that  $\psi \succeq \varphi$ .

**Proof.** If  $\psi \in \Psi_u(\varphi)$ , then the claim is due to Theorem 10.2.1. Otherwise the replacement– lemma can be applied. 

### 10.3. Downward Refinement

Dually to the construction of minimal upward refinements we can construct maximal downward refinements of a formula  $\varphi$ . As we might expect, a maximal downward refinement is defined as follows.

#### Definition 10.3.1

Let φ be an LTL-formula. An LTL-formula ψ is called a maximal downward refinement wrt. ≿ of φ if
1. φ ≿ ψ and
2. there is no LTL-formula ψ' such that φ ≻ ψ' ≻ ψ.

As in the case of upward refinement we will distinguish between formulas with and without temporal operators. We will see that the concepts developed in the foregoing section can again be adapted.

#### 10.3.1. Formulas without Temporal Operators

Recall that a minimal upward refinement of a formula  $\varphi$  which does not contain temporal operators is given as  $\varphi \wedge \neg \chi$  for some maximal minterm  $\chi$ . The philosophy was as follows: since  $\chi$  is a minterm,  $\chi$  has exactly one model. So  $\neg \chi$  has  $2^{|X|} - 1$  models where X is the set of variables under consideration. Building the conjunction of  $\varphi$  and a formula which has as many models as possible without being a tautology (i.e.  $\neg \chi$ ) yields a more general formula which is a minimal upward refinement. The dual aspect of *removing a model* is *adding a model*. So for building a maximal upward refinement of  $\varphi$  we will have to construct a disjunction of  $\varphi$  and a formula which has exactly one model, i.e. a minterm.

#### Theorem 10.3.1

Let  $\varphi$  be a propositional logic formula containing variables from a finite set X of propositional symbols. If  $\varphi$  is satisfiable but no tautology, then for every  $\chi \in \text{MINTERMS}(X)$ with  $\chi \not\models \varphi$  a maximal downward refinement of  $\varphi$  is given by  $\varphi \lor \chi$ .

**Proof.** Let  $\varphi$  be any formula containing variables from X. First we will again note that for every  $\chi \in \text{MINTERMS}(X)$  it holds that  $\varphi \models \varphi \lor \chi$ , so  $\varphi \succeq \varphi \lor \chi$ . Now assume that  $\chi \not\models \varphi$ . Then we have  $MD(\chi) \not\subseteq MD(\varphi)$  and therefore  $MD(\varphi \lor \chi) = MD(\varphi) \cup MD(\chi) \neq$  $MD(\varphi)$  which (together with  $MD(\varphi) \subseteq MD(\varphi \lor \chi)$ ) gives  $MD(\varphi) \subset MD(\varphi \lor \chi)$ , i.e.  $\varphi \succ \varphi \lor \chi$ .

Now assume that  $\varphi \lor \chi$  is not a maximal downward refinement of  $\varphi$ . Let  $\alpha$  be a formula such that  $\varphi \succ \alpha \succ \varphi \lor \chi$ . Then we have  $MD(\varphi) \subset MD(\alpha) \subset MD(\varphi \lor \chi)$  and therefore  $|MD(\varphi)| < |MD(\alpha)| < |MD(\varphi \lor \chi)|$  since  $MD(\varphi)$ ,  $MD(\alpha)$  and  $MD(\chi)$  are finite. Since  $\chi \in \text{MINTERMS}(X)$  we have  $|MD(\chi)| = 1$ , so either  $|MD(\alpha)| = |MD(\varphi)|$  or  $|MD(\alpha)| = |MD(\varphi)| + 1$ . This is a contradiction, so the claim is proved.

Again we will *collect* all downward refinements of  $\varphi$  as follows:

$$\Psi_d(\varphi) = \{ \varphi \lor \chi \mid \chi \in \operatorname{MinTerms}(X) \}.$$

#### Example 10.3.1

Consider the set  $X = \{p_1, p_2, p_3\}$  and the formula  $\varphi = p_1 \rightarrow (p_2 \rightarrow p_3)$  from Example 10.2.1. Here we have

$$\Psi_{d}(\varphi) = \{ \varphi \lor \chi \mid \chi \in \text{MINTERMS}(X) \}$$

$$= \begin{cases} (p_{1} \rightarrow (p_{2} \rightarrow p_{3})) \land (\neg p_{1} \land \neg p_{2} \land \neg p_{3}), \\ (p_{1} \rightarrow (p_{2} \rightarrow p_{3})) \land (\neg p_{1} \land \neg p_{2} \land p_{3}), \\ (p_{1} \rightarrow (p_{2} \rightarrow p_{3})) \land (\neg p_{1} \land p_{2} \land \neg p_{3}), \\ (p_{1} \rightarrow (p_{2} \rightarrow p_{3})) \land (\neg p_{1} \land p_{2} \land p_{3}), \\ (p_{1} \rightarrow (p_{2} \rightarrow p_{3})) \land (p_{1} \land \neg p_{2} \land \neg p_{3}), \\ (p_{1} \rightarrow (p_{2} \rightarrow p_{3})) \land (p_{1} \land p_{2} \land \neg p_{3}), \\ (p_{1} \rightarrow (p_{2} \rightarrow p_{3})) \land (p_{1} \land p_{2} \land \neg p_{3}), \\ (p_{1} \rightarrow (p_{2} \rightarrow p_{3})) \land (p_{1} \land p_{2} \land \neg p_{3}), \\ (p_{1} \rightarrow (p_{2} \rightarrow p_{3})) \land (p_{1} \land p_{2} \land \neg p_{3}), \\ (p_{1} \rightarrow (p_{2} \rightarrow p_{3})) \land (p_{1} \land p_{2} \land \neg p_{3}), \end{cases}$$

#### 10.3.2. Formulas with Temporal Operators

In full analogy to the upward case we will define a set  $\Gamma_d(\varphi)$  for an LTL-formula  $\varphi$  as follows:

$$\begin{split} \Gamma_{d}(\varphi) &= \Psi_{d}(\varphi) \\ \cup & \left\{ \varphi[\mathsf{F}\chi]_{p} \mid p \in \operatorname{Pos}(\varphi), \varphi|_{p} = \mathsf{X}^{i}\chi \text{ for some } i \geq 0 \right\} \\ \cup & \left\{ \varphi[\mathsf{X}^{i}\chi]_{p} \mid p \in \operatorname{Pos}(\varphi), \varphi|_{p} = \mathsf{G}\chi \text{ for every } i \geq 0 \right\} \\ \cup & \left\{ \varphi[\bar{\varphi}_{1} \star \varphi_{2}]_{p}, \varphi[\varphi_{1} \star \bar{\varphi}_{2}]_{p} \mid \stackrel{p \in \operatorname{Pos}(\varphi), \varphi|_{p} = \varphi_{1} \star \varphi_{2}, \bar{\varphi}_{1} \in \Gamma_{d}(\varphi_{1}), \\ \varphi[\bar{\varphi}_{2} \in \Gamma_{d}(\varphi_{2}), \star \in \{\mathsf{U},\mathsf{R}\} \right\}. \end{split}$$

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Again we assume that the formulas  $\bar{\varphi}_1$  and  $\bar{\varphi}_2$  from  $\Gamma_d(\varphi_1)$  and  $\Gamma_d(\varphi_2)$  satisfy the properties  $\varphi_1 \succeq \bar{\varphi}_1$  and  $\varphi_2 \succeq \bar{\varphi}_2$ . The properties of  $\Gamma_d(\varphi)$  are proved as in the upward case.

#### Lemma 10.3.1 (Replacement-Lemma, Downward-Version)

Let  $\varphi$  and  $\psi$  be LTL-formulas, let  $p \in \text{Pos}(\varphi)$  be a position in  $\varphi$ . If  $\psi \succeq \varphi|_p$ , then  $\varphi \succeq \varphi[\psi]_p$ .

**Proof**. analogous to the upward case.

So we have that the formulas from  $\Gamma_d(\varphi)$  are downward refinements of  $\varphi$ .

**Theorem 10.3.2** 

For every LTL-formula  $\varphi$  and every  $\psi \in \Gamma_u(\varphi)$  it holds that  $\varphi \succeq \psi$ .

**Proof**. analogous to the upward case.

In the following section we will show how the choice of a formula from  $\Gamma_u(\varphi)$  or  $\Gamma_d(\varphi)$ can be implemented as a manipulation of the representing automaton  $\mathcal{A}_{\varphi}$ .

# 10.4. Modifying Automata by Application of Refinement Operations

#### 10.4.1. Upward Refinement

We will now present algorithms which allow upward-refinement of given Büchi-automata by manipulating the set of states (and updating the transition relation, the set of initial state and the acceptance component). As pointed out in the foregoing section we can distinguish between refinement by application of propositional formulas and refinement by application of temporal formulas.

All algorithms will only differ in the way the new states are constructed. So we will place our attention on the procedures to construct new states from given ones. For the rest of this section we will assume that  $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  is given.

#### **Propositional Refinement**

Assume that  $\operatorname{VAR}(\varphi) = \{p_1, \ldots, p_{|\operatorname{VAR}(\varphi)|}\}$  and that  $\chi = \bigwedge_{i=1}^{|\operatorname{VAR}(\varphi)|} l_i \in \operatorname{MINTERMS}(\operatorname{VAR}(\varphi))$ for  $l_i \in \{p_i, \neg p_i\}$  are given. Then the construction of  $\mathcal{A}_{\varphi \wedge \neg \chi}$  is simple: First we have to construct  $\operatorname{NNF}(\neg \chi)$ . After that we construct  $\mathcal{A}_{\operatorname{NNF}(\neg \chi)}$  and return  $\mathcal{A}_{\varphi \wedge \neg \chi} = \mathcal{A}_{\varphi} \times \mathcal{A}_{\operatorname{NNF}(\neg \chi)}$ . Since  $\chi = \bigwedge_{i=1}^{|\operatorname{VAR}(\varphi)|} l_i$  we have  $\alpha = \operatorname{NNF}(\neg \chi) = \operatorname{NNF}\left(\neg \bigwedge_{i=1}^{|\operatorname{VAR}(\varphi)|} l_i\right) =$  $\operatorname{NNF}\left(\bigvee_{i=1}^{|\operatorname{VAR}(\varphi)|} \neg l_i\right) = \bigvee_{i=1}^{|\operatorname{VAR}(\varphi)|} \neg l_i$ . Algorithm 13 gives the implementation of the strategy described here.

### Algorithm 13 Propositional Upward Refinement Input:

- Büchi-automaton  $\mathcal{A} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$
- $\chi \in MINTERMS(VAR(\varphi)).$

**Output**:  $\mathcal{A}_{\varphi \wedge \neg \chi}$ .

- 1: compute  $\alpha = \text{NNF}(\neg \chi)$
- 2: compute  $\mathcal{A}_{\alpha}$
- 3: return  $\mathcal{A}_{arphi} imes \mathcal{A}_{lpha}$

By soundness of the product operation  $\times$  we have the following theorem.

#### Theorem 10.4.1

Let  $\varphi$  be an LTL-formula, let  $\mathcal{A}_{\varphi}$  and let  $\chi \in \text{MINTERMS}(\varphi)$  be given. Then Algorithm 13 returns  $\mathcal{A}_{\varphi \wedge \neg \chi}$ .

#### **Temporal Refinement**

The next step is now to present a procedure which allows the introduction of temporal operators. We will see that for each of the rewritten formulas from  $\Gamma_u(\varphi) \setminus \Psi_u(\varphi)$ . So assume that any such  $\psi \in \Gamma_u(\varphi) \setminus \Psi_u(\varphi)$  is chosen. We will proceed by distinction of the form of  $\psi$ .

**Case 1**  $\psi = \varphi[X^i\chi]_p$  for some  $p \in Pos(\varphi)$  such that  $\varphi|_p = F\chi$ , some LTL-formula  $\chi$  and some  $i \ge 0$ . By definition of the set of states of a Büchi-automaton, for every state

s such that  $X^i \chi \in \Phi(s)$  there has to be at least one s' such that  $X^{i-1} \in \Phi(s')$ . Our construction will have to take this into account.

- **Case 2**  $\psi = \varphi[\mathsf{G}\chi]_p$  for some  $p \in \operatorname{Pos}(\varphi)$  such that  $\varphi|_p = \mathsf{X}^i\chi$ . This case is simpler in the sense that some states might be deleted and so the resulting automaton might be smaller than the original one.
- **Case 3** If  $\psi = \varphi[\bar{\varphi}_1 \star \varphi_2]_p$  or  $\psi = \varphi[\varphi_1 \star \bar{\varphi}_2]_p$  for  $\star \in \{\mathsf{U},\mathsf{R}\}$  and  $p \in \operatorname{Pos}(\varphi)$  with  $\varphi|_p = \varphi_1 \star \varphi_2$ , then states from  $\mathcal{A}_{\psi}$  emerge from states of  $\mathcal{A}_{\varphi}$  and states from  $\mathcal{A}_{\bar{\varphi}_1}$  (respectively from  $\mathcal{A}_{\bar{\varphi}_2}$ ) be *merging* the sets of formulas stored in these states. We will see below how this can be achieved.

We will now examine the cases sketched above in more detail.

- **Case 1** We will first present a method which is suitable for the situation described in case 1. Assume that  $p \in \text{Pos}(\varphi)$  is given such that  $\varphi|_p = \text{trueU}\chi \equiv \mathsf{F}\chi$  for some LTLformula  $\chi$  and assume that  $i \ge 0$  is some fixed integer. If  $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$ then we have to process every  $s = (\{\varphi_1, \ldots, \varphi_n\}, \{p_1, \ldots, p_n\}) \in \text{STATES}$  and check if the following cases occur:
  - **Case 1.1** There is  $i_0 \in \{1, ..., n\}$  such that  $\varphi_{i_0} = \text{trueU}\chi$  and  $p_i = p$ . In this case we have to construct states  $\bar{s}_0, ..., \bar{s}_i$  as follows:

$$\bar{s}_j = (\{\varphi_1, \dots, \varphi_{i_0-1}, \mathsf{X}^j \chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_{i_0-1}, p_{i_0} \mathsf{1}^{i-j}, p_{i_0+1}, \dots, p_n\})$$

Here the term  $1^{i-j}$  denotes a sequence of i-j occurrences of the letter 1.

**Case 1.2** Case 1.1 is not fulfilled but there is some  $i_0 \in \{1, ..., n\}$  such that  $p_{i_0} < p$ and **trueU** $\chi \sqsubset \varphi_{i_0}$ . In this case the formula to be manipulated is a proper subformula of the formula  $\varphi_{i_0}$  and therefore it has to be replaced. Let  $\bar{p}$  be from  $\mathbb{N}^*$  such that  $p_{i_0}\bar{p} = p$ . Then we have to construct the state

$$\bar{s} = (\{\varphi_1, \dots, \varphi_{i_0-1}, \varphi_{i_0}[\mathsf{X}^i \chi]_{\bar{p}}, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_n\}).$$

**Case 1.3** In the case that neither of the above cases is fulfilled we only add the state s to the set of new states.

The method described here is summarized in Algorithm 14.

- **Case 2** In contrast to the method presented in the foregoing case constructing the set of states of the automaton  $\mathcal{A}_{\varphi[\mathsf{G}\chi]_p}$  given some  $p \in \operatorname{Pos}(\varphi)$  such that  $\varphi|_p = \mathsf{X}^i\chi$ for some  $i \geq 0$  might yield a smaller set of states since the states which have been built in order to guarantee that  $\chi$  holds can be deleted. Again let  $\mathcal{A}_{\varphi} =$  $(\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  be given. If  $s = (\{\varphi_1, \ldots, \varphi_n\}, \{p_1, \ldots, p_n\})$  is the state which is actually processed, we have to distinguish the following cases:
  - **Case 2.1** If there is  $i_0 \in \{1, ..., n\}$  such that  $\varphi_{i_0} = X^i \chi$  and  $p_{i_0} = p$ , then we can construct *two* new states:

$$\bar{s}_1 = (\{\varphi_1, \dots, \varphi_{i_0-1}, \texttt{falseR}\chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_n\}) \text{ and }$$
$$\bar{s}_2 = (\{\varphi_1, \dots, \varphi_{i_0-1}, \chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_{i_0-1}, p_{i_0}1, p_{i_0+1}, \dots, p_n\}).$$

These states are added to the set of new states. Following the construction of these states we can identify the states which might be deleted. These are the states  $s' \neq s$  such that

$$s' = (\{\varphi_1, \dots, \varphi_{i_0-1}, \mathsf{X}^{i-j}\chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_{i_0-1}, p_{i_0}1^j, p_{i_0+1}, \dots, p_n\}).$$

Each such state is marked as to be deleted.

**Case 2.2** Case 2.1 is not fulfilled but there is  $i_0 \in \{1, ..., n\}$  such that  $p_{i_0} < p$ and  $X^i \chi$  is a proper subformula of  $\varphi_{i_0}$ . Here we will identify  $\bar{p} \in \mathbb{N}^*$  such that  $p_{i_0}\bar{p} = p$  and construct the state

$$\bar{s} = (\{\varphi_1, \dots, \varphi_{i_0-1}, \varphi_{i_0}[\texttt{falseR}\chi]_{\bar{p}}, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_n\}).$$

**Case 2.3** If neither case 2.1 or case 2.2 is fulfilled, we only add s to the set of new states.

The above construction is formalized in Algorithm 15.

- **Case 3** Let us now assume that  $\psi = \varphi[\bar{\varphi}_1 \star \varphi_2]_p$  for some  $\bar{\varphi}_1 \in \Gamma_u(\varphi_1)$ , some  $\star \in \{\mathsf{U},\mathsf{R}\}$ and some  $p \in \operatorname{Pos}(\varphi)$  such that  $\varphi|_p = \varphi_1 \star \varphi_2$ . Let  $\mathcal{A}_{\bar{\varphi}_1} = (\Sigma', \operatorname{STATES}', \delta', S'_0, \mathcal{F}')$ be given. Our construction will be divided into three parts.
  - **Step 1** First we will *rename* the positions from the states of  $\mathcal{A}_{\bar{\varphi_1}}$  in order to match the positions in  $\psi$ . That is if  $s' = (\Phi(s'), \{p_1, \dots, p_n\}) \in \text{STATES'}$  is given, then s has to be changed to  $(\Phi(s'), \{p_1p_1, \dots, p_1p_n)$ .
  - **Step 2** We will then process every  $s = \left(\left\{\varphi_1^{(1)}, \dots, \varphi_{n_1}^{(1)}\right\}, \left\{p_1^{(1)}, \dots, p_{n_1}^{(1)}\right\}\right) \in \text{STATES}$  as follows:
    - 1. for every  $i_0 \in \{1, ..., n_1\}$  such that  $p_{i_0}^{(1)} = p$  we have to carry out an *explicit replacement*, that is we replace s by

$$\bar{s} = \left( \left\{ \varphi_1^{(1)}, \dots, \varphi_{i_0-1}^{(1)}, \varphi_{i_0}^{(1)}[\bar{\varphi}_1]_1, \varphi_{i_0+1}^{(1)}, \dots, \varphi_{n_1}^{(1)} \right\}, \left\{ p_1^{(1)}, \dots, p_{n_1}^{(1)} \right\} \right),$$

2. for every  $i_0 \in \{1, \ldots, n_1\}$  such that  $p_{i_0}^{(1)} < p$  and  $\varphi$  is a proper subformula of  $\varphi_{i_0}^{(1)}$  we replace s by

$$\bar{s} = \left( \left\{ \varphi_1^{(1)}, \dots, \varphi_{i_0-1}^{(1)}, \varphi_{i_0}^{(1)}[\bar{\varphi}_1]_{\bar{p}1}, \varphi_{i_0+1}^{(1)}, \dots, \varphi_{n_1}^{(1)} \right\}, \left\{ p_1^{(1)}, \dots, p_{n_1}^{(1)} \right\} \right)$$

where  $\bar{p} \in \mathbb{N}^*$  is such that  $p_{i_0}^{(1)}\bar{p} = p$  and

3. for every  $i_0 \in \{1, \ldots, n_1\}$  such that  $p < p_{i_0}$  we delete the formula  $\varphi_{i_0}^{(1)}$ and the associated position  $p_{i_0}^{(1)}$ . So s is replaced by

$$\bar{s} = \left( \left\{ \varphi_1^{(1)}, \dots, \varphi_{i_0-1}^{(1)}, \varphi_{i_0+1}^{(1)}, \dots, \varphi_{n_1}^{(1)} \right\}, \left\{ p_1^{(1)}, \dots, p_{i_0-1}^{(1)}, p_{i_0+1}^{(1)}, \dots, p_{n_1}^{(1)} \right\} \right).$$

**Step 3** Assume that after having completed step 2, the state s has the form

$$s = \left( \left\{ \varphi_{j_1}^{(1)}, \dots, \varphi_{j_k}^{(1)} \right\}, \left\{ p_{j_1}^{(1)}, \dots, p_{j_k}^{(1)} \right\} \right).$$

Then for each  $s' = \left(\left\{\varphi_1^{(2)}, \dots, \varphi_{n_2}^{(2)}\right\}, \left\{p_1^{(2)}, \dots, p_{n_2}^{(2)}\right\}\right) \in \text{STATES'}$  we perform the following actions:

- 1.  $\Phi_{\text{new}} = \left\{ \varphi_{j_1}^{(1)}, \dots, \varphi_{j_k}^{(1)}, \varphi_1^{(2)}, \dots, \varphi_{n_2}^{(2)} \right\},\$ 2.  $\text{Pos}_{\text{new}} = \left\{ p_{j_1}^{(1)}, \dots, p_{j_k}^{(1)}, p_1^{(2)}, \dots, p_{n_2}^{(2)} \right\}$  and
- 3. create the state  $s_{\text{new}} = (\Phi_{\text{new}}, \text{Pos}_{\text{new}})$  and add it to NEWSTATES.

Setting  $\Sigma_{\text{new}} = \Sigma \cup \Sigma'$  we can then construct the new transition relation  $\delta'$ , the new set  $S_{0,\text{new}}$  of initial states and the new acceptance component  $\mathcal{F}'$  as usual and the automaton construction is completed. The algorithm for this construction is given in Algorithm 16.

**Case 4** The final case is given by the situation in which  $\psi = \varphi[\varphi_1 \star \overline{\varphi}_2]_p$  for some  $\overline{\varphi}_2 \in \Gamma_u(\varphi_2)$ , some  $\star \in \{\mathsf{U},\mathsf{R}\}$  and some  $p \in \operatorname{Pos}(\varphi)$  such that  $\varphi|_p = \varphi_1 \star \varphi_2$ . The construction is then carried out in full analogy to the construction from case 3. So we will only present the algorithm which is depicted in Algorithm 17.

As in the case of propositional upward refinement the above algorithms for constructing the set of new states of the resulting automaton representing the refined formula can be combined with the standard approaches for extracting the initial states, the transition **Algorithm 14** Temporal Upward Refinement: Constructing new States for  $\mathsf{F} \mapsto \mathsf{X}^i$ Input:

- $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$
- $p \in \operatorname{Pos}(\varphi)$  such that  $\varphi|_p = \operatorname{trueU} \chi \equiv \mathsf{F} \chi$
- $i \ge 0$

**Output**: set NEWSTATES of states of  $\mathcal{A}_{\varphi[X^i\chi]_n}$ 

1: NEWSTATES  $\leftarrow \{(\emptyset, \emptyset)\}$ 2: for each  $s \in \text{STATES}$  do **Require:**  $s = (\{\varphi_1, ..., \varphi_n\}, \{p_1, ..., p_n\})$ if there is  $i_0$  such that  $\varphi_{i_0} = \text{trueU}\chi$  and  $p_{i_0} = p$  then 3:for j = 0, ..., i do 4: $\bar{s} \leftarrow (\{\varphi_1, \dots, \varphi_{i_0-1}, \mathsf{X}^j \chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_{i_0-1}, p_{i_0} 1^{i-j}, p_{i_0+1}, \dots, p_n\})$ 5:NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{\bar{s}\}$ 6: end for 7: else if there is  $i_0$  such that trueU $\chi \sqsubset \varphi_{i_0}$  and  $p_{i_0} < p$  then 8:  $\bar{p} \leftarrow \text{element from } \mathbb{N}^* \text{ such that } p_{i_0}\bar{p}$ 9: = p $\overline{s}$  $\leftarrow$  $(\{\varphi_1,\ldots,\varphi_{i_0-1},\varphi_{i_0}[\mathsf{X}^i\chi]_{\bar{p}},\varphi_{i_0+1},\ldots,\varphi_n\},\{p_1,\ldots,p_n\}))$ NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{\bar{s}\}$ 10:else 11:NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{s\}$ 12:end if 13:14: end for

#### **Algorithm 15** Temporal Upward Refinement: Constructing new States for $X^i \mapsto G$ Input:

- $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$
- $p \in \operatorname{Pos}(\varphi)$  such that  $\varphi|_p = \mathsf{X}^i \chi$

**Output**: set NEWSTATES of states of  $\mathcal{A}_{\varphi[\mathsf{G}_{\chi}]_{p}}$ 

1: mark each  $s \in \text{STATES}$  as not to be deleted 2: NEWSTATES  $\leftarrow \{(\emptyset, \emptyset)\}$ 3: for each  $s \in \text{STATES}$  which is marked as not to be deleted do **Require:**  $s = (\{\varphi_1, \dots, \varphi_n\}, \{p_1, \dots, p_n\})$ if there is  $i_0$  such that  $\varphi_{i_0} = \mathsf{X}^i \chi$  and  $p_{i_0} = p$  then 4:  $\bar{s}_1 \leftarrow (\{\varphi_1, \dots, \varphi_{i_0-1}, \texttt{falseR}\chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_n\})$ 5:  $\bar{s}_2 \leftarrow (\{\varphi_1, \dots, \varphi_{i_0-1}, \chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_{i_0-1}, p_{i_0}, p_{i_0+1}, \dots, p_n\})$ 6: NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{\bar{s}_1, \bar{s}_2\}$ 7: for each  $s' \in \text{STATES } \mathbf{do}$ 8:  $s' = (\{\varphi'_1, \dots, \varphi'_m\}, \{p'_1, \dots, p'_m\})$ **Require:** if n = m then 9: if  $\{\varphi'_1, \ldots, \varphi'_m\} = \{\varphi_1, \ldots, \varphi_{i_0-1}, \mathsf{X}^{i-j}\chi, \varphi_{i_0+1}, \ldots, \varphi_n\}$  for some  $j \geq 0$ 10:then if  $\{p'_1, \ldots, p'_m\} = \{p_1, \ldots, p_{i_0-1}, p_{i_0}1^j, p_{i_0+1}, \ldots, p_n\}$  then 11: mark s' as to be deleted 12:end if 13:end if 14:end if 15:end for 16:else if there if  $i_0$  such that  $p_{i_0} < p$  and  $X^i \chi \sqsubset \varphi_{i_0}$  then 17: $\bar{p} \leftarrow \text{element from } \mathbb{N}^* \text{ such that } p_{i_0}\bar{p} = p$ 18:19: $\bar{s} \leftarrow (\{\varphi_1, \dots, \varphi_{i_0-1}, \varphi_{i_0} | \texttt{falseR}\chi|_{\bar{p}}, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_n\})$ NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{\bar{s}\}$ 20:21:else 22:NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{s\}$ end if 23:24: end for 25: return NEWSTATES

Algorithm 16 Temporal Upward Refinement: Constructing new States for Replacement of Eventualities (replacing the first component)

Input:

- $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$
- $p \in \operatorname{Pos}(\varphi)$  such that  $\varphi|_p = \varphi_1 \star \varphi_2$  for  $\star \in \{\mathsf{U},\mathsf{R}\}$
- $\mathcal{A}_{\overline{\varphi}} = (\Sigma', \text{STATES}', \delta', S'_0, \mathcal{F}')$  for some  $\overline{\varphi} \in \Gamma_u(\varphi_1)$

**Output**: set NEWSTATES of states of  $\mathcal{A}_{\varphi[\bar{\varphi}\star\varphi_2]_p}$ 

1: for each  $s' \in \text{STATES}'$  do **Require:**  $s' = (\Phi(s'), \{p'_1, \dots, p'_n\})$ 2:  $s \leftarrow (\Phi(s'), \{pp'_1, \dots, pp'_n\})$ 3: end for 4: NEWSTATES  $\leftarrow \{(\emptyset, \emptyset)\}$ 5: for each  $s \in \text{STATES do}$ Require:  $s = \left( \left\{ \varphi_1^{(1)}, \dots, \varphi_{n_1}^{(1)} \right\}, \left\{ p_1^{(1)}, \dots, p_{n_1}^{(1)} \right\} \right)$ **Require:** for  $i_0 = 1, ..., n_1$  do if  $p_{i_0}^{(1)} \le p$  and  $\varphi_1$  is a subformula of  $\varphi_{i_0}^{(1)}$  then 6: 7:  $\bar{p} \leftarrow \text{element from } \mathbb{N}^* \text{ such that } p_{i_0}^{(1)} \bar{p} = p \\ s \leftarrow \left( \left\{ \varphi_1^{(1)}, \dots, \varphi_{i_0-1}^{(1)}, \varphi_{i_0}^{(1)} \left[ \overline{\varphi} \right]_{\bar{p}1}, \varphi_{i_0+1}^{(1)}, \dots, \varphi_{n_1}^{(1)} \right\}, \left\{ p_1^{(1)}, \dots, p_{n_1}^{(1)} \right\} \right)$ 8: 9: else if  $p < p_{i_0}^{(1)}$  then  $s \leftarrow \left( \left\{ \varphi_1^{(1)}, \dots, \varphi_{i_0-1}^{(1)}, \varphi_{i_0+1}^{(1)}, \dots, \varphi_{n_1}^{(1)} \right\}, \left\{ p_1^{(1)}, \dots, p_{i_0-1}^{(1)}, p_{i_0+1}^{(1)}, \dots, p_{n_1}^{(1)} \right\} \right)$ end if 10:11: 12:13:end for tire:  $s = \left( \{\varphi_{j_1}^{(1)}, \dots, \varphi_{j_k}^{(1)}\}, \{p_{j_1}^{(1)}, \dots, p_{j_k}^{(1)}\} \right)$ for each  $s' \in \text{STATES'}$  do tire:  $s' = \left( \left\{\varphi_1^{(2)}, \dots, \varphi_{n_2}^{(2)}\right\}, \left\{p_1^{(2)}, \dots, p_{n_2}^{(2)}\right\} \right)$ **Require:** 14:**Require:**  $\Phi_{\text{new}} \leftarrow \left\{ \varphi_{j_1}^{(1)}, \dots, \varphi_{j_k}^{(1)}, \varphi_1^{(2)}, \dots, \varphi_{n_2}^{(2)} \right\}$  $\text{Pos}_{\text{new}} \leftarrow \left\{ p_{j_1}^{(1)}, \dots, p_{j_k}^{(1)}, p_1^{(2)}, \dots, p_{n_2}^{(2)} \right\}$  $\text{NewStates} \leftarrow \text{NewStates} \cup \left\{ (\Phi_{\text{new}}, \text{Pos}_{\text{new}}) \right\}$ 15:16:17:end for 18:19: end for 20: return NEWSTATES

# Algorithm 17 Temporal Upward Refinement: Constructing new States for Replacement of Eventualities (replacing the second component)

Input:

- $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$
- $p \in Pos(\varphi)$  such that  $\varphi|_p = \varphi_1 \star \varphi_2$  for  $\star \in \{\mathsf{U},\mathsf{R}\}$
- $\mathcal{A}_{\overline{\varphi}} = (\Sigma', \text{STATES}', \delta', S'_0, \mathcal{F}')$  for some  $\overline{\varphi} \in \Gamma_u(\varphi_1)$

**Output**: set NEWSTATES of states of  $\mathcal{A}_{\varphi[\varphi_1 \star \bar{\varphi}]_p}$ 

1: for each  $s' \in \text{STATES}'$  do **Require:**  $s' = (\Phi(s'), \{p'_1, \dots, p'_n\})$  $s \leftarrow (\Phi(s'), \{pp'_1, \dots, pp'_n\})$ 2: 3: end for 4: NEWSTATES  $\leftarrow \{(\emptyset, \emptyset)\}$ 5: for each  $s \in \text{STATES do}$ Require:  $s = \left( \left\{ \varphi_1^{(1)}, \dots, \varphi_{n_1}^{(1)} \right\}, \left\{ p_1^{(1)}, \dots, p_{n_1}^{(1)} \right\} \right)$ for  $i_0 = 1, ..., n_1$  do if  $p_{i_0}^{(1)} \leq p$  and  $\varphi_1$  is a subformula of  $\varphi_{i_0}^{(1)}$  then 6: 7:  $\bar{p} \leftarrow \text{element from } \mathbb{N}^* \text{ such that } p_{i_0}^{(1)} \bar{p} = p \\ s \leftarrow \left( \left\{ \varphi_1^{(1)}, \dots, \varphi_{i_0-1}^{(1)}, \varphi_{i_0}^{(1)} \left[ \overline{\varphi} \right]_{\bar{p}2}, \varphi_{i_0+1}^{(1)}, \dots, \varphi_{n_1}^{(1)} \right\}, \left\{ p_1^{(1)}, \dots, p_{n_1}^{(1)} \right\} \right)$ 8: 9: else if  $p < p_{i_0}^{(1)}$  then  $s \leftarrow \left( \left\{ \varphi_1^{(1)}, \dots, \varphi_{i_0-1}^{(1)}, \varphi_{i_0+1}^{(1)}, \dots, \varphi_{n_1}^{(1)} \right\}, \left\{ p_1^{(1)}, \dots, p_{i_0-1}^{(1)}, p_{i_0+1}^{(1)}, \dots, p_{n_1}^{(1)} \right\} \right)$ end if 10:11: 12:13:end for  $\begin{aligned} \text{for each } s' &= \left( \{\varphi_{j_1}^{(1)}, \dots, \varphi_{j_k}^{(1)}\}, \{p_{j_1}^{(1)}, \dots, p_{j_k}^{(1)}\} \right) \\ \text{for each } s' &\in \text{STATES' do} \\ \text{lire:} \qquad s' &= \left( \left\{\varphi_1^{(2)}, \dots, \varphi_{n_2}^{(2)}\right\}, \left\{p_1^{(2)}, \dots, p_{n_2}^{(2)}\right\} \right) \end{aligned}$ **Require:** 14:**Require:**  $\Phi_{\text{new}} \leftarrow \left\{ \varphi_{j_1}^{(1)}, \dots, \varphi_{j_k}^{(1)}, \varphi_1^{(2)}, \dots, \varphi_{n_2}^{(2)} \right\}$  $\text{Pos}_{\text{new}} \leftarrow \left\{ p_{j_1}^{(1)}, \dots, p_{j_k}^{(1)}, p_1^{(2)}, \dots, p_{n_2}^{(2)} \right\}$  $\text{NewStates} \leftarrow \text{NewStates} \cup \left\{ (\Phi_{\text{new}}, \text{Pos}_{\text{new}}) \right\}$ 15:16:17: 18:end for 19: end for 20: return NEWSTATES

function and the acceptance component of the resulting automaton. A procedure of how to achieve this is straightforward to implement. Therefore we will not give it in detail but will merely concentrate on the following theorem.

#### Theorem 10.4.2

Let  $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  be an automaton representing an LTL-formula  $\varphi$  and let  $\psi \in \Gamma_u(\varphi)$  be an upward-refinement of  $\varphi$ . Then the sets of states computed by Algorithms 14, 15, 16 and 17 are correct.

**Proof.** Let an LTL-formula  $\varphi$  be given and assume that some  $\psi \in \Gamma_u(\varphi)$  is chosen. Assume that  $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  is given (and is constructed *correctly*). Let STATES' be the set of states which has been constructed by application of one of the algorithms presented above and let  $\text{STATES}_{cor}$  be the *correct* set of states of  $\mathcal{A}_{\psi}$ . We will have to prove that  $\text{STATES}' = \text{STATES}_{cor}$ . The direction  $\subseteq$  is simple: Let  $s = (\Phi(s), \text{Pos})$ be any state from STATES'. By assumption that  $\mathcal{A}_{\varphi}$  is constructed correctly we have that **false**  $\notin \Phi(s)$  and since the algorithms presented above do not introduce conjunctions or disjunctions we have that if  $\varphi_1 \wedge \varphi_2 \in \Phi(s)$  ( $\varphi_1 \vee \varphi_2 \in \Phi(s)$ ) then  $\varphi_1 \in \Phi(s)$  and (or)  $\varphi_2 \in \Phi(s)$ . The correctness of the positions stored in Pos is immediate. So the direction  $\subseteq$  is proved.

Now assume that  $s = (\Phi(s), \text{Pos}) = (\{\varphi_1, \dots, \varphi_n\}, \{p_1, \dots, p_n\})$  is a state in STATES<sub>cor</sub> If  $p \notin \text{Pos}$ , then we trivially have that  $s \in \text{STATES'}$  since then  $s \in \text{STATES}$ . If  $p \in \text{Pos}$ , say  $p = p_j$  then the claim is proven by showing that all necessary dependencies have been constructed. These dependencies have the form  $X^i \chi$ , that is if they are constructed then  $X^{i-1}\chi, \dots, X^1\chi, X^0\chi$  have to be constructed and included. But this is done by the algorithms so  $s \in \text{STATES'}$  and the direction  $\supseteq$  is proved.

Since the extraction of the remaining components of  $\mathcal{A}_{\psi}$  is more or less trivial, we have that the refinement procedures described are correct.

#### 10.4.2. Downward Refinement

#### **Propositional Refinement**

As in the case of propositional upward refinement we can define an algorithm for constructing the refined formula  $\varphi \lor \chi$  from a formula  $\varphi$ , the automaton  $\mathcal{A}_{\varphi}$  and some  $\chi \in \text{MINTERMS}(\text{VAR}(\varphi))$  by first constructing the automaton  $\mathcal{A}_{\chi}$  and then constructing the automaton  $\mathcal{A}_{\varphi\lor\chi} = \mathcal{A}_{\varphi} ||\mathcal{A}_{\chi}$ . Since the operation || is sound we immediately have the soundness of the refinement procedure. The procedure itself is given in Algorithm 18

#### **Theorem 10.4.3**

Let  $\varphi$  be an LTL-formula, let  $\mathcal{A}_{\varphi}$  and let  $\chi \in \text{MINTERMS}(\varphi)$  be given. Then Algorithm 18 returns  $\mathcal{A}_{\varphi \lor \chi}$ .

#### Algorithm 18 Propositional Downward Refinement Input:

- Büchi-automaton  $\mathcal{A} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$
- $\chi \in MINTERMS(VAR(\varphi)).$

**Output**:  $\mathcal{A}_{\varphi \lor \chi}$ . 1: compute  $\alpha = \text{NNF}(\chi)$ 

- 2: compute  $\mathcal{A}_{\alpha}$
- 3: return  $\mathcal{A}_{\varphi} || \mathcal{A}_{\alpha}$

#### **Temporal Refinement**

The case of refining LTL-formulas by adding respectively changing temporal operators has been described for the case of upward refinement in chapter 10.4.1. Downward refinement is more or less dual to upward refinement as we will see soon. Again we can assume that some LTL-formula  $\varphi$  and some element  $\psi \in \Gamma_d(\varphi) \setminus \Psi_d(\varphi)$  are fixed. As in chapter 10.4.1 we have to distinguish the following cases for  $\psi$ :

**Case 1**  $\psi = \varphi[X^i\chi]_p$  for some  $p \in Pos(\varphi)$  such that  $\varphi|_p = G\chi$  and some  $i \ge 0$ . This case can be seen as the inversion of case 2 from chapter 10.4.1 where certain states had been marked as to be deleted. Consequently the states which are deleted there have to be introduced here. So assume that any element s = $(\{\varphi_1, \ldots, \varphi_n\}, \{p_1, \ldots, p_n\}) \in \text{STATES}$  is given. We can identify the following subcases:

**Case 1.1** If there is  $i_0 \in \{1, ..., n\}$  such that  $\varphi_{i_0} = \texttt{falseR}\chi$  and  $p_{i_0} = p$ , then we have to construct i + 1 states  $\bar{s}_0, ..., \bar{s}_i$  as follows:

$$\bar{s}_j = (\{\varphi_1, \dots, \varphi_{i_0-1}, \mathsf{X}^j \chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_{i_0-1}, p_{i_0} 1^{i-j}, p_{i_0+1}, \dots, p_n).$$

**Case 1.2** Case 1.1 is not fulfilled but there is  $i_0 \in \{1, ..., n\}$  such that  $p_{i_0} < p$ and falseR $\chi$  is a proper subformula of  $\varphi_{i_0}$ . Then let  $\bar{p} \in \mathbb{N}^*$  be such that  $p_{i_0}\bar{p} = p$ . We construct the state

$$\bar{s} = (\{\varphi_1, \dots, \varphi_{i_0-1}, \varphi_{i_0}[\mathsf{X}^i\chi]_{\bar{p}}, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_n\}).$$

**Case 1.3** If neither of the above cases is fulfilled we add the original state s to NEWSTATES.

The complete procedure is given in Algorithm 19.

- **Case 2** By analogy to the case of constructing  $\mathcal{A}_{\varphi[\mathsf{G}\chi]_p}$  where  $\mathsf{X}^i\chi$  had been replaced by  $\texttt{falseR}\chi$  we can construct  $\mathcal{A}_{\varphi[\mathsf{F}\chi]_p}$  from  $\mathcal{A}_{\varphi}$ ,  $i \geq 0$  and some  $p \in \operatorname{Pos}(\varphi)$  where  $\varphi|_p = \mathsf{X}^i\chi$ . If any  $s = (\{\varphi_1, \ldots, \varphi_n\}, \{p_1, \ldots, p_n\}) \in \text{STATES}$  is processed, then we distinguish the following cases:
  - **Case 2.1** There is  $i_0 \in \{1, ..., n\}$  such that  $\varphi_{i_0} = X^i \chi$  and  $p_{i_0} = p$ . Then we construct the following states:

$$\bar{s}_1 = (\{\varphi_1, \dots, \varphi_{i_0-1}, \texttt{trueU}\chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_n\}) \text{ and }$$

$$\bar{s}_2 = (\{\varphi_1, \dots, \varphi_{i_0-1}, \chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_{i_0-1}, p_{i_0}, p_{i_0+1}, \dots, p_n\}).$$

After  $\bar{s}_1$  and  $\bar{s}_2$  have been added to NEWSTATES, we mark each  $s' \neq s$  as to be deleted (see chapter 10.4.1) which have the form

$$s' = (\{\varphi_1, \dots, \varphi_{i_0-1}, \mathsf{X}^{i-j}\chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_{i_0-1}, p_{i_0}1^j, p_{i_0+1}, \dots, p_n\})$$

for some  $j \ge 0$ .

**Case 2.2** If case 2.1 is not fulfilled but there is  $i_0 \in \{1, ..., n\}$  such that  $p_{i_0} < p$ and  $X^i \chi$  is a proper subformula of  $\varphi_{i_0}$  we identify  $\bar{p} \in \mathbb{N}^*$  such that  $p_{i_0}\bar{p} = p$ and construct

$$\bar{s} = (\{\varphi_1, \dots, \varphi_{i_0-1}, \varphi_{i_0}[\texttt{trueU}\chi]_{\bar{p}}, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_n\})$$

and add  $\bar{s}$  to NewStates.

**Case 2.3** Neither of the above cases is fulfilled. then s is added to the set NEW-STATES.

The complete procedure is summarized in Algorithm 20.

- **Case 3** If the refined formula is  $\psi = \varphi[\bar{\varphi}_1 \star \varphi_2]_p$  for some  $p \in \operatorname{Pos}(\varphi)$  such that  $\varphi|_p = \varphi_1 \star \varphi_2$ , some  $\star \in \{U, R\}$  and some  $\bar{\varphi}_1 \in \Gamma_d(\varphi_1)$ , then we can re-use Algorithm 16 without any changes since this algorithm only refers to  $\bar{\varphi}_1$  and not to membership of  $\bar{\varphi}_1$  in  $\Gamma_u(\varphi_1)$  or  $\Gamma_d(\varphi_2)$ .
- **Case 4** By analogy to the foregoing case we can re-use Algorithm 17 in order to compute the set of states for  $\mathcal{A}_{\psi}$  in the case that  $\psi = \varphi[\varphi_1 \star \bar{\varphi}_2]_p$  for  $p \in \operatorname{Pos}(\varphi_2)$  with  $\varphi|_p = \varphi_1 \star \varphi_2, \star \in \{\mathsf{U},\mathsf{R}\}$  and  $\bar{\varphi}_2 \in \Gamma_d(\varphi_2)$ .

So all possible cases for the elements  $\psi \in \Gamma_d(\varphi)$  are covered.

**Algorithm 19** Temporal Downward Refinement: Constructing new States for  $G \mapsto X^i$ Input:

- $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$
- $p \in \text{Pos}(\varphi)$  such that  $\varphi|_p = \texttt{falseR}\chi \equiv \mathsf{G}\chi$
- $i \ge 0$

**Output**: set NEWSTATES of states of  $\mathcal{A}_{\varphi[X^i\chi]_n}$ 

1: NEWSTATES  $\leftarrow \{(\emptyset, \emptyset)\}$ 2: for each  $s \in \text{STATES}$  do **Require:**  $s = (\{\varphi_1, ..., \varphi_n\}, \{p_1, ..., p_n\})$ if there is  $i_0$  such that  $\varphi_{i_0} = \texttt{falseR}\chi$  and  $p_{i_0} = p$  then 3:for j = 0, ..., i do 4: $\bar{s} \leftarrow (\{\varphi_1, \dots, \varphi_{i_0-1}, \mathsf{X}^j \chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_{i_0-1}, p_{i_0} 1^{i-j}, p_{i_0+1}, \dots, p_n\})$ 5:NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{\bar{s}\}$ 6: end for 7:else if there is  $i_0$  such that  $\texttt{falseR}\chi \sqsubset \varphi_{i_0}$  and  $p_{i_0} < p$  then 8:  $\bar{p} \leftarrow \text{element from } \mathbb{N}^* \text{ such that } p_{i_0}\bar{p}$ 9: = p $\overline{s}$  $(\{\varphi_1,\ldots,\varphi_{i_0-1},\varphi_{i_0}[\mathsf{X}^i\chi]_{\bar{p}},\varphi_{i_0+1},\ldots,\varphi_n\},\{p_1,\ldots,p_n\}))$ NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{\bar{s}\}$ 10:else 11:NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{s\}$ 12:end if 13:14: end for

#### **Algorithm 20** Temporal Downward Refinement: Constructing new States for $X^i \mapsto F$ Input:

- $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$
- $p \in \operatorname{Pos}(\varphi)$  such that  $\varphi|_p = \mathsf{X}^i \chi$

**Output**: set NEWSTATES of states of  $\mathcal{A}_{\varphi[\mathsf{F}_{\chi}]_{p}}$ 

1: mark each  $s \in \text{STATES}$  as not to be deleted 2: NEWSTATES  $\leftarrow \{(\emptyset, \emptyset)\}$ 3: for each  $s \in \text{STATES}$  which is marked as not to be deleted do **Require:**  $s = (\{\varphi_1, \dots, \varphi_n\}, \{p_1, \dots, p_n\})$ if there is  $i_0$  such that  $\varphi_{i_0} = \mathsf{X}^i \chi$  and  $p_{i_0} = p$  then 4:  $\bar{s}_1 \leftarrow (\{\varphi_1, \dots, \varphi_{i_0-1}, \texttt{trueU}\chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_n\})$ 5:  $\bar{s}_2 \leftarrow (\{\varphi_1, \dots, \varphi_{i_0-1}, \chi, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_{i_0-1}, p_{i_0}, p_{i_0+1}, \dots, p_n\})$ 6: NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{\bar{s}_1, \bar{s}_2\}$ 7: for each  $s' \in \text{STATES } \mathbf{do}$ 8:  $s' = (\{\varphi'_1, \dots, \varphi'_m\}, \{p'_1, \dots, p'_m\})$ **Require:** if n = m then 9: if  $\{\varphi'_1, \ldots, \varphi'_m\} = \{\varphi_1, \ldots, \varphi_{i_0-1}, \mathsf{X}^{i-j}\chi, \varphi_{i_0+1}, \ldots, \varphi_n\}$  for some  $j \geq 0$ 10:then if  $\{p'_1, \ldots, p'_m\} = \{p_1, \ldots, p_{i_0-1}, p_{i_0}, p_{i_0+1}, \ldots, p_n\}$  then 11: mark s' as to be deleted 12:end if 13:end if 14:end if 15:end for 16:else if there if  $i_0$  such that  $p_{i_0} < p$  and  $X^i \chi \sqsubset \varphi_{i_0}$  then 17: $\bar{p} \leftarrow \text{element from } \mathbb{N}^* \text{ such that } p_{i_0}\bar{p} = p$ 18:19: $\bar{s} \leftarrow (\{\varphi_1, \dots, \varphi_{i_0-1}, \varphi_{i_0}[\texttt{trueU}\chi]_{\bar{p}}, \varphi_{i_0+1}, \dots, \varphi_n\}, \{p_1, \dots, p_n\})$ NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{\bar{s}\}$ 20:21:else 22:NEWSTATES  $\leftarrow$  NEWSTATES  $\cup \{s\}$ end if 23:24: end for 25: return NEWSTATES

By analogy to the situation of constructing the states for the upward-refined automata, we are now able to present the following theorem.

#### Theorem 10.4.4

Let  $\mathcal{A}_{\varphi} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  be an automaton representing an LTL-formula  $\varphi$  and let  $\psi \in \Gamma_d(\varphi)$  be an upward-refinement of  $\varphi$ . Then the sets of states computed by Algorithms 19 and 20, are correct.

**Proof**. Exactly as in the case of upward refinement.

#### 10.4.3. Extracting Formulas

The final step in the process of refining LTL-formulas by application of the automata manipulations introduced in this chapter is to extract the refined program which has been *learned*. This is a simple task due to the properties of the automata construction used to build the representing automata.

Recall that the set of initial states  $S_0$  of an automaton  $\mathcal{A}_{\varphi}$  representing  $\varphi$  has been defined as the set of all states s such that  $\varphi \in \Phi(s)$ . So after having completed the refinement process it suffices to extract this formula from one of the initial states of the resulting automaton.

## Definition 10.4.1

Let  $\varphi_1$  and  $\varphi_2$  be LTL-formulas. Then  $\varphi_1 \sqsubseteq \varphi_2$  if there is a position  $p \in \text{Pos}(\varphi_2)$ such that  $\varphi_2|_p = \varphi_1$ . Furthermore we will write  $\varphi_1 \sqsubset \varphi_2$  if  $\varphi_1 \sqsubseteq \varphi_2$  and  $p \neq \varepsilon$ .

In other words,  $\varphi_1 \sqsubseteq \varphi_2$  ( $\varphi_1 \sqsubset \varphi_2$ ) if and only if  $\varphi_1$  is a (proper) subformula of  $\varphi_2$ . Using the notation  $\sqsubseteq$  we can define the *maximum* of a state s.

**Definition 10.4.2** Let  $\mathcal{A} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  be an extended Büchi–automaton and let  $s \in \text{STATES}$ .

Then a formula  $\varphi \in \Phi(s)$  is called a *maximum* of s if there is no formula  $\psi \in \Phi(s)$  such that  $\varphi \sqsubset \psi$ .

Obviously for every  $s \in \text{STATES}$  there is exactly one maximum. This formula will be denoted as  $\max(s)$ . This enables the extraction of a formula from a Büchi-automaton simply by extracting the maximum  $\max(s_0)$  for any initial state.

#### **Theorem 10.4.5**

Let  $\mathcal{A} = (\Sigma, \text{STATES}, \delta, S_0, \mathcal{F})$  be an extended Büchi-automaton and let  $s_0 \in S_0$  be any initial state of  $\mathcal{A}$ . Then if  $\mathcal{A} = \mathcal{A}_{\varphi}$ , then  $\varphi = \max(s_0)$ .

**Proof**. Immediately by definition of  $\mathcal{A}_{\varphi}$ .

Since we have defined our version of Büchi-automata in such a way that not only the formulas but also the positions of these formulas in the original formula from which the automaton had been constructed are stored, we can extract the maximum of any  $s \in S_0$  simply by searching for the position  $\varepsilon$  in the position-component of s. So let s = $(\{\varphi_1, \ldots, \varphi_n\}, \{p_1, \ldots, p_n\})$  be given and let  $i_0$  be such that  $p_{i_0} = \varepsilon$ . Then  $\varphi_{i_0} = \max(s)$ . So searching for the maximum of s can be achieved in time  $\mathcal{O}(|\Phi(s)|)$ .

#### 10.5. The Identification Process

During the foregoing sections we have developed techniques which are necessary in order to identify LTL-programs from positive and negative examples. So this final section of this chapter will deal with the topic of how to combine these techniques in order to derive an identification procedure. In order to implement an identification procedure we have to ensure that at every point of time there is a uniquely determined continuation of the refinement process. In other words we have to ensure that our method is deterministic.

So assume that there is a (heuristic) function  $\mathfrak{h}$  selecting a type of refinement step to be carried out. Depending on the outcome of the computation carried out by  $\mathfrak{h}$  we have to call one of the algorithms developed in the foregoing sections. Formally  $\mathfrak{h}$  has to return the following information:

- a number of an algorithm to be called and
- the input data for this algorithm.

Considering the different signatures which Algorithms 13 to 20 have, the following data might be necessary:

- a formula  $\chi \in MINTERMS(VAR(\varphi))$  as input for Algorithms 13 and 18,
- a position  $p \in Pos(\varphi)$ ,
- a value  $i \ge 0$  for Algorithms 14 and 19,
- an index  $j \in \{1, 2\}$  for Algorithms 16 and 17 selecting which component has to be replaced and
- an automaton  $\mathcal{A}_{\overline{\varphi}}$  as input for Algorithms 16 and 17.

So if  $\mathfrak{Aut}$  denotes the set of *all* Büchi–automata constructed in the way developed in this and the foregoing chapter,  $\mathfrak{h}$  should have the following signature:

$$\mathfrak{h}: \mathfrak{Aut} \to \mathbb{Z}_{10} \times \operatorname{MinTerms}(\operatorname{Var}(\varphi)) \times \operatorname{Pos}(\varphi) \times \mathbb{N} \times \{1, 2\}.$$

For the sake of formal clearness we assume that the components of  $\mathfrak{h}(\mathcal{A})$  given any automaton  $\mathcal{A} \in \mathfrak{Aut}$  can be accessed by application of a simple projection. That is the projections

$$\begin{array}{lll} (\cdot)_{1} & : & \mathbb{Z}_{10} \times \operatorname{MINTerms}(\operatorname{Var}(\varphi)) \times \operatorname{Pos}(\varphi) \times \mathbb{N} \times \{1,2\} \to \mathbb{Z}_{10}, \\ (\cdot)_{2} & : & \mathbb{Z}_{10} \times \operatorname{MINTerms}(\operatorname{Var}(\varphi)) \times \operatorname{Pos}(\varphi) \times \mathbb{N} \times \{1,2\} \to \operatorname{MINTerms}(\operatorname{Var}(\varphi)), \\ (\cdot)_{3} & : & \mathbb{Z}_{10} \times \operatorname{MINTerms}(\operatorname{Var}(\varphi)) \times \operatorname{Pos}(\varphi) \times \mathbb{N} \times \{1,2\} \to \operatorname{Pos}(\varphi), \end{array}$$

 $(\cdot)_4$ :  $\mathbb{Z}_{10} \times \operatorname{MinTerms}(\operatorname{Var}(\varphi)) \times \operatorname{Pos}(\varphi) \times \mathbb{N} \times \{1, 2\} \to \mathbb{N}$  and

$$(\cdot)_5 : \mathbb{Z}_{10} \times \text{MINTERMS}(\text{VAR}(\varphi)) \times \text{Pos}(\varphi) \times \mathbb{N} \times \{1, 2\} \rightarrow \{1, 2\}$$

are defined.

Additionally we assume that another function  $\overline{\mathfrak{h}} : \mathbb{Z}_2 \times \{1,2\} \times \operatorname{Pos}(\varphi) \times \mathfrak{Aut} \to \mathfrak{Aut}$ is given which returns an automaton representing the formula to be inserted during a run of Algorithms 16 and 17. The first argument of  $\overline{\mathfrak{h}}$  denotes the direction which the refinement step has to take (0 for upward refinement and 1 for downward refinement) while the second argument denotes the component to be replaced in the formula at the position given by the third component.

So the identification process can be sketched as follows:

- **Step 0** Given input  $\varphi$ , construct the automaton  $\mathcal{A} = \mathcal{A}_{\varphi}$ .
- **Step 1** Now assume that  $\mathcal{E}^+ = \{\varphi_1, \dots, \varphi_k\}$ . For *i* ranging from 1 to *k* perform the following loop.

**Step 1.1** Construct  $\mathcal{A}_{NNF}(\neg \varphi_i)$ ,

- **Step 1.2** construct  $\mathcal{A} \times \mathcal{A}_{NNF(\neg \varphi_i)}$  and
- **Step 1.3** as long as  $L(\mathcal{A} \times \mathcal{A}_{NNF(\neg \varphi_i)}) \neq \emptyset$  do
  - **Step 1.3.1** compute  $\mathfrak{h}(\mathcal{A}) = ((\mathfrak{h}(\mathcal{A}))_1, \dots, (\mathfrak{h}(\mathcal{A}))_5)$  and
  - **Step 1.3.2** depending on the value of  $(\mathfrak{h}(\mathcal{A}))_1$  perform the following actions:
    - if (\$\mathcal{h}(\mathcal{A})\$)\_1 = 0\$, then replace \$\mathcal{A}\$ by the result of Algorithm 13 given inputs \$\mathcal{A}\$ and \$(\$\mathcal{h}(\mathcal{A})\$)\_2\$,
    - 2. if  $(\mathfrak{h}(\mathcal{A}))_1 = 1$ , then

      - b) extract the remaining components:

- i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{V_{\text{AR}}(\varphi)}$ ,
- ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
- iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

- c) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,
- 3. if  $(\mathfrak{h}(\mathcal{A}))_1 = 2$ , then
  - a) compute the set NEWSTATES by application of Algorithm 15 given inputs  $\mathcal{A}$  and  $(\mathfrak{h}(\mathcal{A}))_3$ ,
  - b) extract the remaining components:
    - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{\text{VAR}(\varphi)}$ ,
    - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
    - iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

and then

- c) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,
- 4. if  $(\mathfrak{h}(\mathcal{A}))_1 = 3$ , then
  - a) compute the automaton  $\bar{\mathcal{A}} = \bar{\mathfrak{h}}(\mathcal{A}, 0, 1, (\mathfrak{h}(\mathcal{A}))_3),$
  - b) compute the set NEWSTATES by application of Algorithm 16 given inputs  $\mathcal{A}$ ,  $(\mathfrak{h}(\mathcal{A}))_3$  and  $\overline{\mathcal{A}}$ ,
  - c) extract the remaining components:

- i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{\text{VAR}(\varphi)}$ ,
- ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
- iii. extract  $\mathcal{F}_{\text{new}}$  by application of Algorithm 12 given input NEW-STATES

- d) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,
- 5. if  $(\mathfrak{h}(\mathcal{A}))_1 = 4$ , then
  - a) compute the automaton  $\bar{\mathcal{A}} = \bar{\mathfrak{h}}(\mathcal{A}, 0, 2, (\mathfrak{h}(\mathcal{A}))_3),$
  - b) compute the set NEWSTATES by application of Algorithm 17 given inputs  $\mathcal{A}$ ,  $(\mathfrak{h}(\mathcal{A}))_3$  and  $\overline{\mathcal{A}}$ ,
  - c) extract the remaining components:
    - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{V_{\text{AR}}(\varphi)}$ ,
    - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
    - iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

and then

- d) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,
- if (\$\mathcal{h}(\mathcal{A})\$)\_1 = 5\$, then replace \$\mathcal{A}\$ by the result of Algorithm 18 given inputs \$\mathcal{A}\$ and \$(\$\mathcal{h}(\mathcal{A})\$)\_2\$,
- 7. if  $(\mathfrak{h}(\mathcal{A}))_1 = 6$ , then
  - a) compute the set NEWSTATES by application of Algorithm 19 given inputs A, (h(A))<sub>3</sub> and (h(A))<sub>4</sub>,

- b) extract the remaining components:
  - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{\text{VAR}(\varphi)}$ ,
  - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
  - iii. extract  $\mathcal{F}_{\text{new}}$  by application of Algorithm 12 given input NEW-STATES

- c) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,
- 8. if  $(\mathfrak{h}(\mathcal{A}))_1 = 7$ , then
  - a) compute the set NEWSTATES by application of Algorithm 20 given inputs A and (h(A))<sub>3</sub>,
  - b) extract the remaining components:
    - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{V_{\text{AR}}(\varphi)}$ ,
    - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
    - iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

and then

c) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,

9. if  $(\mathfrak{h}(\mathcal{A}))_1 = 8$ , then

- a) compute the automaton  $\bar{\mathcal{A}} = \bar{\mathfrak{h}}(\mathcal{A}, 1, 1, (\mathfrak{h}(\mathcal{A}))_3),$
- b) compute the set NEWSTATES by application of Algorithm 16 given inputs  $\mathcal{A}$ ,  $(\mathfrak{h}(\mathcal{A}))_3$  and  $\overline{\mathcal{A}}$ ,
- c) extract the remaining components:

- i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{\text{VAR}(\varphi)}$ ,
- ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
- iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

d) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}}$ ),

and

- 10. if  $(\mathfrak{h}(\mathcal{A}))_1 = 9$ , then
  - a) compute the automaton  $\bar{\mathcal{A}} = \bar{\mathfrak{h}}(\mathcal{A}, 1, 2, (\mathfrak{h}(\mathcal{A}))_3),$
  - b) compute the set NEWSTATES by application of Algorithm 17 given inputs  $\mathcal{A}$ ,  $(\mathfrak{h}(\mathcal{A}))_3$  and  $\overline{\mathcal{A}}$ ,
  - c) extract the remaining components:
    - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{\text{VAR}(\varphi)}$ ,
    - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
    - iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

and then

- d) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}}$ ).
- **Step 2** Now assume that  $\mathcal{E}^- = \{\varphi_1, \dots, \varphi_l\}$ . For *i* ranging from 1 to *l* perform the following loop.

**Step 2.1** Construct  $\mathcal{A}_{NNF(\neg \varphi_i)}$ ,

**Step 2.2** construct  $\mathcal{A} \times \mathcal{A}_{NNF(\neg \varphi_i)}$  and

**Step 2.3** as long as  $L(\mathcal{A} \times \mathcal{A}_{NNF}(\neg \varphi_i)) = \emptyset$  do

**Step 2.3.1** compute  $\mathfrak{h}(\mathcal{A}) = ((\mathfrak{h}(\mathcal{A}))_1, \dots, (\mathfrak{h}(\mathcal{A}))_5)$  and

- **Step 2.3.2** depending on the value of  $(\mathfrak{h}(\mathcal{A}))_1$  perform the following actions:
  - if (\$\mathbf{h}(\mathcal{A})\$)\_1 = 0\$, then replace \$\mathcal{A}\$ by the result of Algorithm 13 given inputs \$\mathcal{A}\$ and \$(\$\mathbf{h}(\mathcal{A})\$)\_2\$,
  - 2. if  $(\mathfrak{h}(\mathcal{A}))_1 = 1$ , then
    - a) compute the set NEWSTATES by application of Algorithm 14 given inputs  $\mathcal{A}$ ,  $(\mathfrak{h}(\mathcal{A}))_3$  and  $(\mathfrak{h}(\mathcal{A}))_4$ ,
    - b) extract the remaining components:
      - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{\text{VAR}(\varphi)}$ ,
      - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
      - iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

and then

- c) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,
- 3. if  $(\mathfrak{h}(\mathcal{A}))_1 = 2$ , then
  - a) compute the set NEWSTATES by application of Algorithm 15 given inputs A and (h(A))<sub>3</sub>,
  - b) extract the remaining components:
    - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{\text{Var}(\varphi)}$ ,
    - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and

iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

and then

- c) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,
- 4. if  $(\mathfrak{h}(\mathcal{A}))_1 = 3$ , then
  - a) compute the automaton  $\bar{\mathcal{A}} = \bar{\mathfrak{h}}(\mathcal{A}, 0, 1, (\mathfrak{h}(\mathcal{A}))_3),$
  - b) compute the set NEWSTATES by application of Algorithm 16 given inputs  $\mathcal{A}$ ,  $(\mathfrak{h}(\mathcal{A}))_3$  and  $\overline{\mathcal{A}}$ ,
  - c) extract the remaining components:
    - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{V_{\text{AR}}(\varphi)}$ ,
    - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
    - iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

and then

- d) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,
- 5. if  $(\mathfrak{h}(\mathcal{A}))_1 = 4$ , then
  - a) compute the automaton  $\bar{\mathcal{A}} = \bar{\mathfrak{h}}(\mathcal{A}, 0, 2, (\mathfrak{h}(\mathcal{A}))_3),$
  - b) compute the set NEWSTATES by application of Algorithm 17 given inputs  $\mathcal{A}$ ,  $(\mathfrak{h}(\mathcal{A}))_3$  and  $\overline{\mathcal{A}}$ ,
  - c) extract the remaining components:
    - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{\text{Var}(\varphi)}$ ,
    - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and

iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

and then

- d) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,
- if (\$\mathcal{h}(\mathcal{A})\$)\_1 = 5\$, then replace \$\mathcal{A}\$ by the result of Algorithm 18 given inputs \$\mathcal{A}\$ and \$(\$\mathcal{h}(\mathcal{A})\$)\_2\$,
- 7. if  $(\mathfrak{h}(\mathcal{A}))_1 = 6$ , then
  - a) compute the set NEWSTATES by application of Algorithm 19 given inputs A, (h(A))<sub>3</sub> and (h(A))<sub>4</sub>,
  - b) extract the remaining components:
    - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{\text{VAR}(\varphi)}$ ,
    - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
    - iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

and then

- c) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,
- 8. if  $(\mathfrak{h}(\mathcal{A}))_1 = 7$ , then
  - a) compute the set NEWSTATES by application of Algorithm 20 given inputs A and (h(A))<sub>3</sub>,
  - b) extract the remaining components:
    - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{\text{VAR}(\varphi)}$ ,
    - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and

iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

and then

- c) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,
- 9. if  $(\mathfrak{h}(\mathcal{A}))_1 = 8$ , then
  - a) compute the automaton  $\bar{\mathcal{A}} = \bar{\mathfrak{h}}(\mathcal{A}, 1, 1, (\mathfrak{h}(\mathcal{A}))_3),$
  - b) compute the set NEWSTATES by application of Algorithm 16 given inputs  $\mathcal{A}$ ,  $(\mathfrak{h}(\mathcal{A}))_3$  and  $\overline{\mathcal{A}}$ ,
  - c) extract the remaining components:
    - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{V_{\text{AR}}(\varphi)}$ ,
    - ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
    - iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES

and then

d) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}})$ ,

and

- 10. if  $(\mathfrak{h}(\mathcal{A}))_1 = 9$ , then
  - a) compute the automaton  $\bar{\mathcal{A}} = \bar{\mathfrak{h}}(\mathcal{A}, 1, 2, (\mathfrak{h}(\mathcal{A}))_3),$
  - b) compute the set NEWSTATES by application of Algorithm 17 given inputs  $\mathcal{A}$ ,  $(\mathfrak{h}(\mathcal{A}))_3$  and  $\overline{\mathcal{A}}$ ,
  - c) extract the remaining components:
    - i. extract  $\delta_{\text{new}}$  by application of Algorithm 11 given inputs NEW-STATES and  $\Sigma = 2^{V_{\text{AR}}(\varphi)}$ ,

- ii. extract  $S_{0,\text{new}}$  from NEWSTATES as described before Theorem 9.2.2 and
- iii. extract  $\mathcal{F}_{new}$  by application of Algorithm 12 given input NEW-STATES
- and then
- d) replace  $\mathcal{A}$  by (NEWSTATES,  $\delta_{\text{new}}, S_{0,\text{new}}, \mathcal{F}_{\text{new}}$ ).

The strategy described above is then an implementation of the identification process of LTL-formulas from sets  $\mathcal{E}^+$  and  $\mathcal{E}^-$  of positive and negative examples. As soon as the identification process stops we therefore extract the learned formula from the set of initial states of the resulting automaton as described in the proof of Theorem 10.4.5.

The definition of the identification process has shown that LTL-programs can be learned from sets  $\mathcal{E}^+$  and  $\mathcal{E}^-$  by application of our algorithms. What remains to be examined is (as in the case of PROLOG(+T)-programs) the complexity of the identification task. The following chapter will therefore attack this problem.

## 11. Identifiability of Ltl-programs

#### Contents

11.1. General Notations
11.2. Upper Bounds for the VC–Dimension of Classes of LTL–
Programs
11.2.1. General LTL–Programs
11.2.2. The deterministic Sublogic of LTL

This final chapter will deal with a similar problem as chapter 8.2, namely the problem of estimating the complexity of the identification process of LTL-programs. In chapter 8.2 we have extended some techniques from [11] in order to derive upper bounds for the VC-dimension of several classes of PROLOG(+T)-programs. We will proceed similarly for LTL-programs.

## 11.1. General Notations

We will first identify the objects to be encoded and then show how many bits will be needed in order to encode these objects. It is immediately clear that the propositional symbols from X and the propositional constants **true** and **false** have to be encoded. This yields |X| + 2 distinct objects. Additionally we will choose the symbols "(", "," and ")" (i.e. the brackets and the comma-symbol), the logical connectives  $\land$ ,  $\lor$ ,  $\rightarrow$ ,  $\leftrightarrow$  and  $\neg$ and the temporal operators X, G, F, U and R. Assuming that  $X = \{p_1, \ldots, p_n\}$  this gives

Symbol	Encoding	Symbol	Encoding
(	bin(n+3)	Х	bin(n+11)
)	bin(n+4)	F	$\sin(n+12)$
,	bin(n+5)	G	bin(n+13)
$\wedge$	bin(n+6)	U	bin(n+14)
V	bin(n+7)	R	bin(n+15)
$\rightarrow$	bin(n+8)		
$\leftrightarrow$	bin(n+9)		
7	bin(n+10)		

Table 11.1.: Encodings for logical connectives from LTL

a total of n + 15 distinct objects. For the sake of formal clearance we will assume that all formulas are represented in prefix-notation, that is the 2-ary connectives  $\land$ ,  $\lor$ ,  $\rightarrow$  and  $\leftrightarrow$  as well as the 2-ary temporal operators U and R are seen as 2-ary function symbols and the remaining logical and temporal operators are treated as unary function symbols. For example if  $\varphi = p_1 \land X p_2 U G p_3$ , then  $\varphi$  is treated as the string  $\land (p_1, U(X(p_2), G(p_3)))$ .

The encoding *cod* which we will use is based on the value *n* of elements occurring in *X*. If  $p_i \in X$  is any propositional constant, then  $cod(p_i) = bin(i)$  where bin(i) denotes the binary representation of the integer *i*. Additionally we define cod(true) = bin(n+1)and cod(talse) = bin(n+2). The remaining symbols can be encoded in an arbitrary but fixed way. We will choose the encoding from Table 11.1.

The encoding from Table 11.1 has to be changed in the way that the encoding of symbols with strictly less than |bin(n+15)| symbols are padded with zeros from the left side. This yields equal length for every encoded symbol. For example if  $X = \{p_1, p_2, p_3\}$ , then n = 3 and n + 15 = 18. So we have  $cod(p_1) = 00001$  and  $cod(\mathsf{R}) = 10010$ .

Obviously every symbol to be encoded can be represented as a string from  $\{0,1\}^*$  of length  $\lceil \log_2(n+15) \rceil$ . Since the encoding is padded with zeros every formula will be encoded as a string from  $\{0,1\}^*$  which has a length which is a multiple of this value.

Having defined the encoding of the symbols used in LTL-formulas it remains to show how composite formulas are encoded. But this is straightforward: Let any LTL-formula  $\varphi$  be given.

**Case 1** if  $\varphi = \varphi_1 \oplus \varphi_2$  for  $\oplus \in \{\land, \lor, \rightarrow, \leftrightarrow, \mathsf{U}, \mathsf{R}\}$ , then

$$cod(\varphi) = cod(\oplus)cod("(")cod(\varphi_1)cod(",")cod(\varphi_2)cod(")")$$

and

**Case 2** if  $\varphi = \oplus \psi$  for  $\oplus \in \{\neg, X, G, F\}$ , then

$$cod(\varphi) = cod(\oplus)cod(\psi)$$

## Example 11.1.1

Let  $\varphi = \mathsf{G}(p_1 \to (\mathsf{X}p_2 \to \neg p_3))$  be given. Due to the assumption from above we will treat  $\varphi$  as  $\mathsf{G} \to (p_1, \to (\mathsf{X}p_2, \neg p_3))$ . We have  $X = \{p_1, p_2, p_3\}$  that is n = 3 and therefore n + 15 = 18. Consequently the symbols are encoded using  $\lceil \log_2(18) \rceil = 5$  bits. The encoding of the relevant symbols is therefore:  $cod(p_1) = 00001$ ,  $cod(p_2) = 00010$ ,  $cod(p_3) = 00011$ ,  $cod("(") = 00100, cod(")") = 00110, cod(",") = 00101, cod((\to)) = 01011, cod((\neg)) = 01101$ ,  $cod((\neg)) = 01000$ .

$$cod(\varphi) = cod(\mathsf{G}(p_1 \to (\mathsf{X}p_2 \to \neg p_3)))$$
  
=  $cod(\mathsf{G}(\to (p_1, \to (\mathsf{X}(p_2), \neg (p_3)))))$   
=  $cod(\mathsf{G})cod("(")cod(\to) \dots$   
=  $10000 \ 00100 \ 01011 \ 00100 \ 00001 \ 00101 \ 01011 \ 01000 \ 01110 \ 00110 \ \dots$   
 $\dots \ 00010 \ 00110 \ 00110 \ 00101 \ 01010 \ 00011 \ 00110 \ 00110 \ \dots$   
 $\dots \ 00110 \ 00110 \ 00110 \ \dots$ 

Some results for the value of VCDIM(C) for classes C consisting of certain propositional logic formulas have been presented before. Early studies by Natarajan (see [124]) merely deal with polynomial time PAC-learnability. Several newer papers also deal with estimations as well as with exact bounds for the VC-dimension of classes of boolean formulas. For example in [135] it is shown that the class  $MON_n$  of monotone boolean formulas consisting of n propositional variables has  $VCDIM(MON_n) = \binom{n}{\lfloor \frac{n}{2} \rfloor}$ . Further approaches for the problem of learning propositional formulas which do not use the notion of VC-dimension are for example presented in [141] and [80].

As in chapter 8 we will now proceed by first deriving estimations for an upper bound for the VC-dimension of syntactically unrestricted LTL-formulas and then by studying a more restricted language given by the so-called *deterministic sublogic* LTL<sup>det</sup> of LTL.

# 11.2. Upper Bounds for the VC–Dimension of Classes of Ltl–Programs

### 11.2.1. General Ltl–Programs

We will now derive upper bounds for the VC–Dimension of certain classes LTL–programs. For this purpose we define the following: for given values of n, c and t the class  $LTL^{n,c,t}$ denotes the set of all LTL–formulas containing at most n distinct propositional variables, at most c logical connectives and at most t temporal operators.

For the rest of this section assume that n, c and t are fixed nonnegative integers and assume that a formula  $\varphi \in LTL^{n,c,t}$  is chosen. We will define a measuring-function  $\|\cdot\|: LTL^{n,c,t} \to \mathbb{N}$  mapping formulas to integers as follows:  $\|\varphi\| = |cod(\varphi)|$ , that is  $\|\varphi\|$ denotes the number of binary digits in the representation of  $cod(\varphi)$ .

We will now derive an upper bound for the value VCDIM(LTL<sup>*n,c,t*</sup>) by presenting a function  $l : \mathbb{N}^3 \to \mathbb{N}$  such that  $\|\varphi\| \leq l(n,c,t)$  for any values of n, c and t and every  $\varphi \in \text{LTL}^{n,c,t}$ . First we have to recall that the propositional constants **true** and **false** and the propositional variables  $p_i \in X$  can be encoded using  $\lceil \log_2(n+15) \rceil$  bits, that is  $\|\varphi\| = \lceil \log_2(n+15) \rceil$  for  $\varphi \in X \cup \{\texttt{true}, \texttt{false}\}$ . Furthermore the connectives and the temporal operators can be encoded with the same number of bits. So we will have to identify the *worst case*, that is the case in which  $\varphi \in \text{LTL}^{n,c,t}$  has maximum length. Clearly in this case *exactly* t binary temporal operators have to be involved in  $\varphi$ . Similarly we need c binary connectives in order to reach the maximum length. Each of these operators and connectives is encoded using  $\lceil \log_2(n+15) \rceil$  bits. Additionally we need  $3\lceil \log_2(n+15) \rceil$  more bits in order to encode the brackets and the commas. Furthermore in this case we have t + c + 1 subformulas  $\psi_j$  without occurrences of any temporal operator or connective, that is  $\psi_j \in X \cup \{\texttt{true}, \texttt{false}\}$  for  $j = 1, \ldots, t + c + 1$ . This gives a total number of

$$4(t+c)\lceil \log_2(n+15) \rceil + (t+c+1)\lceil \log_2(n+15) \rceil$$
  
=  $(5t+5c+1)\lceil \log_2(n+15) \rceil$ 

bits as the maximum value of  $\|\varphi\|$  for any  $\varphi \in LTL^{n,c,t}$ . But with this number of bits we can encode at most  $2^{(5t+5c+1)\lceil \log_2(n+15)\rceil}$  different formulas, that is

$$|\operatorname{LTL}^{n,c,t}| \le 2^{(5t+5c+1)\lceil \log_2(n+15)\rceil}$$

and so the following theorem can be proved.

### Theorem 11.2.1

Let n, c and t be fixed nonnegative integers. Then

VCDIM 
$$(LTL^{n,c,t}) = \mathcal{O}\left((5t+5c+1)\log_2(n+15)\right).$$

**Proof**. Immediately by the above estimation for the size of  $LTL^{n,c,t}$  and Lemma 8.1.1.

## 11.2.2. The deterministic Sublogic of Ltl

In the foregoing section we have derived an upper bound for the value of the VC– dimension of structured classes of arbitrary LTL–formulas. Here we will see that a particular subset of LTL, namely the set of all *deterministic* formulas as introduced by [109] can be identified using only marginally more bits in the encoding of formulas.

Roughly speaking the language  $LTL^{det}$  of *deterministic* LTL-formulas consists of all these elements from LTL in which the usage of the *nondeterministic* operators  $\lor$ , U and R is guarded by some propositional variable symbol p. Formally  $LTL^{det}$  is defined as follows:

- 1. true and false are in LTL<sup>det</sup>,
- 2. every  $p \in X$  is in LTL<sup>det</sup> and
- 3. for all  $\varphi_1, \varphi_2 \in LTL^{det}$  and each  $p \in X \ \varphi_1 \land \varphi_2, \ X\varphi_1, \ (p \land \varphi_1) \lor (\neg p \lor \varphi_2),$  $(p \land \varphi_1) \cup (\neg p \lor \varphi_2)$  and  $(p \land \varphi_1) \mathsf{R}(\neg p \lor \varphi_2)$  are in  $LTL^{det}$ .

As before the set  $LTL^{n,c,t,det}$  denotes the set of all formulas  $\varphi \in LTL^{det}$  containing at most n distinct elements from X, at most c connectives and at most t temporal operators.

Since  $LTL^{det}$ -formulas are syntactically more complex than general LTL-formulas, we can ask if this does change the value of VCDIM ( $LTL^{n,c,t,det}$ ). Below we will see that this is *not* the case.

Of course, the language  $LTL^{det}$  is less expressive than LTL. But in [109] it is shown that  $LTL^{det}$ -formulas have the property that their negation can be represented by a 1-weak Büchi-automaton (a certain type of Büchi-automaton which is equipped with partial ordering on the set STATES which is compatible with the relation  $\delta$ ) which has a set of states of size linear in the size (that is the length) of  $\varphi$ .

In order to derive an upper bound for the VC-dimension of  $LTL^{det}$  we will again rewrite formulas in prefix notation and change the arity of the nondeterministic symbols  $\lor$ , U and R to 3 as follows: assume that  $\oplus \in \{\lor, U, R\}$ ,  $p \in X$  and  $\varphi_1, \varphi_2$  are chosen. Then  $\varphi = (p \land \varphi_1) \oplus (\neg p \land \varphi_2)$  will be rewritten to  $\oplus (p, \varphi_1, \varphi_2)$ . Using the encoding scheme from the foregoing section we can encode these formulas by setting  $cod(\varphi) = cod(\oplus)cod("(")cod(p)cod(",")cod(\varphi_1)cod(",")cod(\varphi_2)cod(")")$ .

Since the operators  $\mathsf{F}$  and  $\mathsf{G}$  do not have to be encoded, we can now encode every symbol occurring in a formula from  $\mathrm{LTL}^{n,c,t,\mathrm{det}}$  using at most  $\lceil \log_2(n+13) \rceil$  bits. So we clearly have for arbitrary formulas  $\varphi \in \mathrm{LTL}^{n,c,t,\mathrm{det}}$ :

1. If  $\varphi \in X \cup \{\texttt{true}, \texttt{false}\}$ , then

$$\|\varphi\| = \lceil \log_2(n+13) \rceil,$$

2. if  $\varphi = \mathsf{X}\psi$ , then

$$\|\varphi\| = \left\lceil \log_2(n+13) \right\rceil + \|\psi\|,$$

3. if  $\varphi = \varphi_1 \wedge \varphi_2$ , then

$$\|\varphi\| = 4\lceil \log_2(n+13)\rceil + \sum_{i=1}^2 \|\varphi_i\|$$

and

4. if  $\varphi = (p \land \varphi_1) \oplus (\neg p \land \varphi_2)$  for some  $\oplus \in \{\lor, \mathsf{U}, \mathsf{R}\}$ , then

$$\|\varphi\| = 6\lceil \log_2(n+13) + \sum_{i=1}^2 \|\varphi_i\|.$$

The next step is to determine the maximum number of positions in  $\phi$  such that an element from  $X \cup \{\texttt{true}, \texttt{false}\}$  occurs at this position. Obviously for t = 0 we have at most c + 1 such positions. In the case that t > 0 we can distinguish the following cases:

1. If  $\varphi = \mathsf{X}\psi$ , then

$$|\{p \in \operatorname{Pos}(\varphi) \mid \varphi|_p \in X \cup \{\texttt{true}, \texttt{false}\}\}|$$

$$= |\{p \in \operatorname{Pos}(\varphi) \mid \psi|_p \in X \cup \{\texttt{true}, \texttt{false}\}\}|$$

and

2. if 
$$\varphi = (p \land \varphi_1) \oplus (\neg p \land \varphi_2)$$
 for  $\oplus \in \{\mathsf{U},\mathsf{R}\}$ , then

$$\begin{split} &|\{p\in\operatorname{Pos}(\varphi)\mid\varphi|_p\in X\cup\{\texttt{true},\texttt{false}\}\}|\\ &= 2+\sum_{i=1}^2|\{p\in\operatorname{Pos}(\varphi_i)\mid\psi|_p\in X\cup\{\texttt{true},\texttt{false}\}\}|\,. \end{split}$$

Clearly we have the maximum value of such positions for if  $\varphi$  contains t binary temporal operators. It is obvious that in this case we have

$$|\{p \in \operatorname{Pos}(\varphi) \mid \varphi|_p \in X \cup \{\texttt{true}, \texttt{false}\}\}| \le 4t.$$

But since t such binary operators consume 3t logical connectives there must be  $k \ge 0$ such that  $c \le 3t + k$ . So additional we have 3t + k + 1 more such positions. So in total we have 4t + 3t + k + 1 = 7t + k + 1 such positions. So the elements from  $X \cup \{\texttt{true}, \texttt{false}\}$ which occur in  $\varphi$  can be encoded using at most

$$(7t + k + 1) \lceil \log_2(n + 13) \rceil$$

bits.

The total value of  $\|\varphi\|$  can be estimated using the following two parameters which determine the number of positions with *deterministic* connectives and the number of positions with *nondeterministic* connectives:

$$DET(\varphi) = |\{p \in POS(\varphi) \mid \varphi|_p = \neg \psi \text{ or } \varphi|_p = \varphi_1 \land \varphi_2 \text{ or } \varphi|_p = \mathsf{X}\psi\}| \text{ and}$$
$$NDET(\varphi) = |\{p \in POS(\varphi) \mid \varphi|_p = (p \land \varphi_1) \oplus (\neg p \land \varphi_2) \text{ for some } \oplus \in \{\lor, \mathsf{U}, \mathsf{R}\}\}|.$$

Since each of the c connectives in  $\varphi$  is either deterministic or nondeterministic we have

$$\begin{aligned} \|\varphi\| &\leq (7t+k+1)\lceil \log_2(n+13)\rceil + 4\mathrm{DET}(\varphi)\lceil \log_2(n+13)\rceil + 6\mathrm{NDET}(\varphi)\lceil \log_2(n+13)\rceil \\ &= (7t+k+1+4\mathrm{DET}(\varphi)+6\mathrm{NDET}(\varphi))\lceil \log_2(n+13)\rceil. \end{aligned}$$

Using the estimations  $Det(\varphi) \leq c$  and  $NDet(\varphi) \leq t + c$  we have

$$\|\varphi\| \leq (7t + k + t + 4c + 6(c + t))\lceil \log_2(n + 13) \rceil$$
  
=  $(13t + 10c + k + 1)\lceil \log_2(n + 13) \rceil.$ 

Finally we have  $k \leq c$  and therefore

$$\begin{aligned} \|\varphi\| &\leq (13t + 10c + c + 1) \lceil \log_2(n + 13) \rceil \\ &= (13t + 11c + 1) \lceil \log_2(n + 13) \rceil. \end{aligned}$$

So any formula  $\varphi \in LTL^{n,c,t,det}$  can be encoded using at most  $(13t+11c+1)\lceil \log_2(n+13) \rceil$  bits and therefore

$$|\operatorname{LTL}^{n,c,t,\operatorname{det}}| \le 2^{(13t+11c+1)\lceil \log_2(n+13)\rceil}.$$

So the following theorem is proved.

### Theorem 11.2.2

Let n, c and t be fixed nonnegative integers. Then

VCDIM 
$$(LTL^{n,c,t,det}) = \mathcal{O}\left((13t + 11c + 1)\log_2(n + 13)\right)$$

As we have already mentioned in the beginning of this section the value VCDIM  $(LTL^{n,c,t,det})$  is not significantly greater than the value of VCDIM  $(LTL^{n,c,t})$ .

The results from the theorems derived in this chapter complete our study on the learnability of LTL-programs from positive and negative examples. We have seen that LTL-programs can be identified by applying manipulations to the set of states of certain automata (the representing automata) of actual hypotheses. By application of upward and downward refinement procedures we were able to define an identification process for LTL-formulas. The following last chapter of this thesis summarizes the results obtained during our studies and points out open problems and directions for future research.

## 12. Conclusions

This short final chapter will summarize the topics addressed in this thesis and point out some aspects for future research activities.

## 12.1. Summary of the Topics

We have addressed the problem of *learning* temporal logic programs written in some temporal logic programming language from positive and negative examples. Therefore we have structured the theory into two parts:

First order Inductive Temporal Logic Programming In the case of first order languages we have developed the programming language PROLOG(+T) which is a rule-based, PROLOG-style programming languages which is equipped with the temporal operators X, G, F, U and P. We have described how to prove goals from PROLOG(+T)programs and have seen how the semantics of PROLOG(+T)-programs can be characterized. While the former topic can be solved using a modified and extended *tableaux* procedure, the latter can be solved by extending the theory of *Herbrand*-*Interpretations* and *Herbrand*-*Models* which are a standard tool from the field of first order Logic Programming.

Following the definition and description of the programming language of interest we have seen how specialization and generalization operations can be carried out in order to modify the programs in the case that they contradict the examples. The last point of the first part dealt with the question of how to analyze the complexity of the learning task. We have seen that the class of PROLOG(+T)-programs can be structured depending on certain syntactic parameters in such a way that it enables the derivation of upper bounds for the VC-Dimension of these classes.

Propositional Inductive Temporal Logic Programming While the techniques developed for first order temporal programming languages face the problem that first order logic in general and first order temporal logic in particular is undecidable, the restriction to propositional temporal logic languages yields decidable satisfiability and implication problems. Therefore we have studied how LTL can be used in order to solve the problem of learning certain temporal logic formulas from positive and negative examples. We have recalled that LTL-formulas can be translated into nondeterministic Büchi-automata and that these automata can be refined in order to fit specifications imposed by the sets of examples. Furthermore we have seen that the generality ordering chosen in the case of first order languages (i.e. the subsumption ordering) cannot be applied in the case of LTL since the objects under consideration are not necessarily clauses. But since propositional temporal programming languages have decidable satisfiability problems we could refine programs with respect to the implication ordering which is finer than subsumption.

As in the case of first order temporal logic we have also studied the complexity of identifying LTL-formulas (resp. LTL-programs) from positive and negative examples by deriving upper bounds for the VC-dimension of certain classes of such formulas.

## 12.2. Perspectives for Future Research

Three major perspectives can or should be pointed out: Integration of *constraints* (and integration of constraint solving techniques into the theorem proving procedure), the definition of a syntactically less restricted first order language (i.e. relaxing the condition that the objects of PROLOG(+T)-programs are essentially clauses) and studying other, perhaps more expressive propositional temporal languages, such as CTL, CTL<sup>\*</sup>, the  $\mu$ -calculus or some of the sublogics of the foregoing languages.

- Integration of Constraints Constraints are constructs which model relations between certain objects such that the relations have a fixed interpretation. Such relation symbols with fixed interpretation can be the equality symbol = or some comparison symbol such as  $\leq$ , <, > or  $\geq$ . Of course these symbols are well suited for reasoning about arithmetic relations and consequently many constraints are arithmetic constraints. The integration of constraints into PROLOG(+T) would make it necessary to extend the theorem proving procedure in order to integrate constraint solving resp. constraint satisfaction routines (see [84] or [85] for a survey of constraint programming techniques) and techniques for synthesizing logic programs containing constraints (see [144])
- Extending the syntax of Prolog(+T) Another perspective for future research might be the extension of PRolog(+T) in such a way that the program statements need not to be clauses anymore. This would lead us to the full first order linear time temporal logic FoLTL. Dealing with FoLTL objects might cause several problems. On the one hand we can not hope for efficient refinement procedures anymore since the objects under consideration are not necessarily clauses, so subsumption does not make any sense for general FoLTL-formulas. On the other hand the theorem proving task would become much more complicated (see [3] for a description of temporal logic theorem proving in the case of nonclausal languages).

- Other propositional temporal languages The restriction to propositional temporal programming languages has (as we have seen in the third part of this thesis) the advantage that satisfiability and implication problems can be decided (in contrast to PROLOG(+T) which contains the full first order predicate logic and which is therefore undecidable). But we have only studied *one* of all possible propositional temporal languages, namely the *linear* time temporal logic LTL. Here the term *linear* means that the language contains only such temporal quantifiers which allow reasoning about *one* possible continuation of the actual point of time under consideration. Branching time Logic (as we will see below) contains other quantifiers which are capable of modeling questions dealing with *all* continuations resp. *some* continuations and which can in these sense be seen as temporal versions of the universal and existential quantifier.
  - Branching Time Logic As we have already noted above, Branching time temporal logic allows reasoning about more than one continuation of the actual point of time. CTL (another prominent specification language which implements the concept of branching time) is equipped with the quantifiers E and A which model the circumstance that there exists a computation path resp. that for all computation paths the quantified formula has to hold (see [34] or [58] for a formal description of CTL). Consequently the term CTL stands for Computation Tree Logic since the set of possible computations can be seen as trees. CTL-formulas can be translated into Alternating Tree Automata (see [21]) which could also be studied and extended in order to allow refinement operations.
  - Mixed Logic Combining Linear and Branching Time Temporal Logics one gets the logic CTL<sup>\*</sup> which is more expressive than both CTL and LTL. Formulas from CTL<sup>\*</sup> can be translated into Street tree Automata (see e.g. [59]). This logic could also be studied in order to learn CTL<sup>\*</sup> construct from examples.

**Fixpoint Logic** From the theoretical point of view the  $\mu$ -calculus  $\mathcal{L}_{\mu}$  is perhaps

the most interesting propositional temporal language. It integrates temporal operators and fixpoint operators (see [96]). It is more expressive than LTL, CTL and CTL\* (see [143]).

All these logics are essentially propositional logics and therefore they are decidable for satisfiability. Consequently they should be studied in order to characterize the complexity of identifying concepts from positive and negative examples.

## A. Formal Description of the Programming Languages

#### Contents

A.1. Syntax of $PROLOG(+T)$
A.1.1. Terms
A.1.2. Atoms and Literals
A.1.3. Rules
A.1.4. General $PROLOG(+T)$ -Objects
A.2. Syntax of LTL

For the sake of completeness and in order to make it easier to develop parsers and interpreters for the languages used throughout this thesis we will now give grammars for these languages. The grammars will be presented in an EBNF-like syntax, i.e. each grammar will consist of a set of rules with one nonterminal symbol on the left hand side and more or less arbitrary right hand sides. These grammars can be easily converted into a form which is suitable for tools generating compilers (e.g. **bison**, see [68]).

## A.1. Syntax of Prolog(+T)

By definition we have several different types of objects which have to be generated by the grammar. These are terms, atoms and literals and rules (including facts as special cases).

Consequently we have to present rules which are capable to define all these objects.

## A.1.1. Terms

Terms have been defined to be constructs of the following form:

- Variable Terms, i.e. constructs such as t = x for symbols  $x \in \mathcal{X}$ ,
- Strings representing Integers from Z,
- Function Terms, i.e. constructs such as  $t = f(t_1, \ldots, t_n)$  or
- List Terms, that is t = [] or  $t = [t_1, \ldots, t_n]$  or  $t = [\overline{t}|t_1, \ldots, t_n]$ .

Consequently the first production step for the generation of a term must chose which type of term has to be created.

 $Term ::= Variable-symbol \mid Function-Term \mid List-Term \mid Number$ 

### Variable–Terms and Integers

Depending on the type of term to be generated we have to give rules for generating each such type. Therefore we make the following convention:

- a variable symbol has to start with an upper case letter and
- any other symbol (i.e. a function or a predicate symbol) has to start with a lower case letter.

The simplest objects to be created are variable terms and integers.

Variable–symbol	::=	$'_'   Upper-Case-Letter  $
		Upper-Case-Letter Variable-Suffix
Upper-Case-Letter	::=	(A,  , B,  , C,  , D,  , E,  , L,  , C,  , H,  , I,  , I,  , I,  , K,  , F,  , M,
		'N'  'O'  'P'  'Q'  'R'  'S'  'T'  'U'  'V'  'W'  'X'  'Y'  'Z'

Lower-Case-Letter	::=	'a'  'b'  'c'  'd'  'e'  'f'  'g'  'h'  'i'  'j'  'k'  'l'  'm'
		'n'  'o'  'p'  'q'  'r'  's'  't'  'u'  'v'  'w'  'x'  'y'  'z'
Letter	::=	$Upper-Case-Letter \mid Lower-Case-Letter$
Digit	::=	(0,   (1,   (2,   (3,   (4,   (5,   (6,   (7,   (8,   (9, )))))))))
Special– $Symbol$	::=	$Letter \mid Digit \mid `_'$
Variable–Suffix	::=	Special–Symbol   Special–Symbol Special–Symbol
Number	::=	$Positive-Number \   \ Negative-Number$
Positive - Number	::=	Digit   Digit Positive–Number
Negative-Number	::=	'-' Positive–Number

### **Function**-Terms

The next more complicated constructs are function terms, i.e. terms of the form  $t = f(t_1, \ldots, t_n)$  with  $n = \alpha(f)$  and terms  $t_1, \ldots, t_n$  or t = f for constant symbols f (i.e. such symbols f with  $\alpha(f) = 0$ ).

Function_Term		Function-Name
	—	
		Function–Name Left–Delimiter
		Argument–List Right–Delimiter
Left-Delimiter	::=	"("
Right-Delimiter	::=	·);
Function-Name	::=	$Lower-Case-Letter \   \ Lower-Case-Letter \ Function-Suffix$
Function–Suffix	::=	Variable–Suffix
Argument-List	::=	Term   Term ',' Argument–List

### List-Terms

List terms are the most complicated terms to generate since they might have several different forms. A term t is a list term if it is t = [] (i.e. t is the empty list) or if t consists

of an enumeration of elements (i.e.  $t = [t_1, \ldots, t_n]$  for terms  $t_1, \ldots, t_n$ ) or if t is given by its head and its tail. In this case we can distinguish the cases that  $t = [\bar{t}|t_1, \ldots, t_n]$  and  $t = [\bar{t}|t']$  where  $\bar{t}$  denotes the head and  $t_1, \ldots, t_n$  is an enumeration of the elements in the tail or t' is a term representing the whole tail of t.

List-Term	::=	Left–List–Delimiter Right–List–Delimiter
		Left–List–Delimiter List–Arguments
		Right-List-Delimiter
Left-List-Delimiter	::=	·[]
Right–List–Delimiter	::=	·],
List-Arguments	::=	Argument–List
		Term ' ' Argument–List
		Term ' ' Term

## A.1.2. Atoms and Literals

Atoms and literals are either temporal or nontemporal ones. Nontemporal atoms and literals are then either one of the constants **true** and **false** or built using predicate symbols. In this case they have the form  $p(t_1, \ldots, t_n)$  or  $not(p(t_1, \ldots, t_n))$ .

Atom	::=	'true'   'false'   Relational-Atom				
Relational–Atom	::=	Predicate–Name				
		Predicate–Name Left–Delimiter				
		Argument–List Right–Delimiter				
Predicate-Name	::=	$Lower-Case-Letter \   \ Lower-Case-Letter \ Predicate-Suffix$				
Predicate–Suffix	::=	Variable–Suffix				

Literal ::= Atom | 'not' Left-Delimiter Literal Right-Delimiter

In order to integrate the temporal operators we generalize the concept of literals to general literals. General literals are then divided into two classes, namely temporal literals and nontemporal literals.

General–Literal	::=	Literal   Temporal–Literal
		'not' Left-Delimiter General-Literal
		Right–Delimiter
Temporal–Literal	::=	Unary–Temporal–Literal
		Binary–Temporal–Literal
Unary–Temporal–Literal	::=	Unary–Temporal–Connective General–Literal
Binary–Temporal–Literal	::=	General–Literal
		Binary–Temporal–Connective General–Literal
Unary–Temporal–Connective	::=	'X'   'F'   'G'
Binary–Temporal–Connective	::=	'U'   'P'

## A.1.3. Rules

Rules are created using the productions for literals. Each rule is either a fact or a definite rule, that is a rule with a nonempty tail.

Rule	::=	$Fact \mid Definite-Rule$
Fact	::=	General–Literal End–Delimiter
Definite-Rule	::=	General–Literal Implication–Symbol
		List–Of–Literals End–Delimiter
End-Delimiter	::=	
Implication-Symbol	::=	·:'
List-Of-Literals	::=	General–Literal   General–Literal ',' List–Of–Literals

## A.1.4. General Prolog(+T)-Objects

Since every PROLOG(+T)-object is either a term or a formula, we add a rule produce these two types of objects. Each formula is then given as a literal (a general literal) or a rule. Prolog(+T)-Object ::= Term | Formula Formula ::= General-Literal | Rule

These rules complete the syntax of PROLOG(+T). In order to be well formed an object needs

- to be parsed and then
- to be checked if the symbols used in the object are compatible with the constraints given by the signature.

## A.2. Syntax of Ltl

The syntax of LTL is very simple, since there are nearly no constraints on the form of a formula. Consequently a grammar which generates the set of all LTL-formulas can be extracted directly from the definition of the language LTL (see page 20).

Formulas from LTL are either *atomic* formulas (i.e. the constants **true** and **false** or proposition symbols) or composite formulas. Composite formulas are built using unary or binary connectives which can be either propositional or temporal ones.

LTL–Formula	::=	$LTL-Atom \mid LTL-Composite-Formula$
LTL-Atom	::=	'true'   'false'   LTL-Proposition-Symbol
LTL–Composite–Formula	::=	LTL–Unary   LTL–Binary

For the sake of simplicity we will introduce names for the symbols used in the rules to generate the formulas, in particular we will introduce names for the brackets and the connectives.

Left–Delimiter	::=	"( '
Right- $Delimiter$	::=	')'
LTL-Unary-Propositional-Connective	::=	<b>!</b> ?

LTL-Unary-Temporal-Connective	::=	'X'   'G'   'F'
LTL-Binary-Propositional-Connective	::=	$`+`\mid`*`\mid`\to`\mid`\leftrightarrow`$
LTL–Binary–Temporal–Connective	::=	'U'   'R'

Proposition symbols are used in order to build atomic formulas of LTL. They have to start with a letter (no matter if it is an upper case letter or a lower case letter) followed by a (possibly empty) string of arbitrary symbols. Such strings are generated using the rule below which has the symbol *symbol-suffix* on its left hand side.

LTL–Proposition–Symbol	::=	Nondigit   Nondigit Symbol–Suffix
Nondigit	::=	'a'  'b'  'c'  'd'  'e'  'f'  'g'  'h'  'i'  'j'  'k'  'l'  'm'  'n'
		'o'  'p'  'q'  'r'  's'  't'  'u'  'v'  'w'  'x'  'y'  'z'
		'A'  'B'  'C'  'D'  'E'  'F'  'G'  'H'  'I'  'J'  'K'  'L'
		'M'   'N'  'O'  'P'  'Q'  'R'  'S'  'T'  'U'  'V'  'W'
		'X'  'Y'  'Z'
Digit	::=	`0'   `1'   `2'   `3'   `4'   `5'   `6'   `7'   `8'   `9'
Symbol–Suffix	::=	''   Digit Symbol–Suffix   Nondigit Symbol–Suffix
		`_' Symbol–Suffix

As we have already mentioned above, composite LTL-formulas can be built up using unary and binary connectives which can be either propositional ones or temporal ones. The rules for building such formulas are given below. First we have to present two rules for building general formulas using binary and unary connectives.

LTL–Unary	::=	$LTL-Negation \mid LTL-Unary-Temporal-Formula$
LTL–Binary	::=	LTL-Binary-Propositional-Formula
		LTL-Binary-Temporal-Formula

Having chosen the arity of the connective rules for generating the formulas using connectives with the chosen arity have to be applied.

LTL–Negation	::=	LTL-Unary-Propositional-Connective
		LTL–Formula
LTL–Unary–Temporal–Formula	::=	LTL–Unary–Temporal–Symbol
		LTL–Formula
LTL–Binary–Propositional–Formula	::=	Left–Delimiter LTL–Formula
		LTL-Binary-Propositional-Connective
		LTL–Formula Right–Delimiter
LTL–Binary–Temporal–Formula	::=	Left–Delimiter LTL–Formula
		LTL–Binary–Temporal–Connective
		LTL–Formula Right–Delimiter

Since there are no further restrictions on the syntax of LTL-formulas the grammar for LTL is now complete.

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