

IMPLEMENTING PRIORITIZED REASONING IN LOGIC PROGRAMMING

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Abstract: Prioritized reasoning is an important extension of logic programming and is a powerful tool for expressing desiderata on the program solutions in order to establish the best ones. This paper discusses the implementation of the case of preference relation among atoms and introduces a system, called CHOPPER, realizing choice optimization recently proposed in (Caroprese et al., 2007). CHOPPER supports the ASO_{Ch} and ASO_{FCh} semantics based on the concept of “choice”, as a set of preference rules describing common choice options in different contexts, and the ASO semantics (Brewka et al., 2003), which evaluates each preference rule separately. This paper outlines the architecture of the system, discusses aspects of the choice identification strategies and of the feasibility of choice options. Moreover, the comparison of the proposed approach with the other implementation approaches proposed in the literature is provided.

1 INTRODUCTION

Prioritized reasoning is an important extension of logic programming, used in a large variety of AI problems. The most common form of preference consists in specifying preference conditions among rules (Brewka and Eiter, 1999; Delgrande et al., 2000; Delgrande et al., 2003; Gelfond and Son, 1997), whereas, some other proposals admit the expression of preference relations among atoms (Brewka et al., 2003; Sakama and Inoue, 2000; Wakaki et al., 2003). See (Delgrande et al., 2004) for a survey on this topic.

This work is a contribution to realizing prioritized reasoning in logic programming in the presence of preference conditions involving atoms. This topic has been investigated in (Brewka et al., 2003) and (Sakama and Inoue, 2000), proposing the ASO and PLP semantics respectively. The ASO semantics evaluates the degree of satisfaction of all preference rules to determine the preferred models and can establish a preference relation between each couple of models directly. On the contrary, the PLP semantics compares two models only on the basis of their common preferences, and needs to test the transitive property to derive the non directly visible preference relations; this lies in

a more complex implementation. A generalization of the ASO semantics which look beyond the each single preference rule, by considering preferences as a tool for choice representation, was recently proposed in (Caroprese et al., 2007).

The prioritized reasoning based on the choice optimization is illustrated by the following example.

Example 1 Consider the prioritized program $\langle P_1, \Phi_1 \rangle$. P_1 describes the possible menus by means of three rules r_1 , r_2 and r_3 , where \oplus states that exactly one of the head's atoms has to be taken in the model, and three constraints c_1 , c_2 and c_3 , i.e. rules with empty heads satisfied if the body is false (e.g. c_1 states that *beef* and *red* cannot be simultaneously present).

r_1 : fish \oplus beef \leftarrow	c_1 : \leftarrow beef, red
r_2 : red \oplus white \leftarrow	c_2 : \leftarrow beef, pie
r_3 : pie \oplus ice-cream \leftarrow	c_3 : \leftarrow fish, white
ρ_1 : white $>$ red \leftarrow fish	
ρ_2 : red $>$ white \leftarrow beef	
ρ_3 : pie $>$ ice-cream \leftarrow	

P_1 has three stable models $M_1 = \{\text{fish, red, pie}\}$, $M_2 = \{\text{fish, red, ice-cream}\}$ and $M_3 = \{\text{beef, white, ice-cream}\}$. The preference rules ρ_1 and ρ_2 specify that (i) the choice of drink (*white* or *red*) follows the selection of the main dish (*fish* or *beef*);

(ii) *white* is better than *red* in the presence of *fish*;
 (iii) *red* is better than *white* in the presence of *beef*;
 whereas ρ_3 denotes that *pie* is better than *ice-cream*.
 M_1 is better than M_2 as *pie* is preferred to *ice-cream*
 and they have the same choice option of drink. M_1
 is better than M_3 as they have the second best choice
 of the drink (in the presence of *fish* and *beef* respec-
 tively), whereas the choice of the *dessert* is satisfied
 better by M_1 . Thus M_1 is a preferred model.

Observe that the ASO semantics, which considers
 each preference rule separately, cannot establish the
 preference ordering between M_1 and M_3 and derives
 that both M_1 and M_3 are preferred models. In more
 details, it derives that M_1 is better than M_3 w.r.t. ρ_2
 and M_3 is better than M_1 w.r.t. ρ_1 , thus they cannot be
 compared. \square

This paper discusses the implementation of choice op-
 timization in logic programming and presents a sys-
 tem, called CHOPPER, realizing this kind of priori-
 tized reasoning.

2 PRELIMINARIES

An *Answer Set Optimization* program, ASO, is a pair
 $\langle \mathcal{P}, \Phi \rangle$, where \mathcal{P} is called *Generating Program* and Φ
 is called *Preference Program*; Φ consists of a finite
 set of rules of the form $C_1 > \dots > C_k \leftarrow \text{body}$, where
body is a conjunction of literals, i.e. atoms or negation
 of atoms, and C_i s are *boolean combinations* of literals.
 Intuitively, a preference rule $\rho \in \Phi$ above described
 introduces a preference order among C_1, \dots, C_k : C_i
 is preferred to C_j , for $i < j$ and $i, j \in [1..k]$. Thus, the set
 of preferences Φ determines a preference ordering on
 the answer sets described by \mathcal{P} .

Let $\Phi = \{r_1, \dots, r_n\}$ be a preference program and S
 be an answer set, then S induces a *satisfaction vector*
 $V_S = [v_S(r_1), \dots, v_S(r_n)]$ where: a) $v_S(r_j) = I$, if r_j is *Ir-*
relevant to S , i.e. (i) the body of r_j is not satisfied in S
 or (ii) the body of r_j is satisfied, but none of the C_i s
 is satisfied in S . b) $v_S(r_j) = \min\{i \mid S \models C_i\}$, otherwise.
 The satisfaction vectors are used to compare the answer
 sets under the assumption that I is equal to 1 (i.e.,
 $v_{S_j}(r_i) = I$ is equivalent to $v_{S_j}(r_i) = 1$).

Let S_1 and S_2 be two answer sets, then (i) $S_1 \geq S_2$ if
 $V_{S_1} \leq V_{S_2}$, i.e. if $v_{S_1}(r_i) \leq v_{S_2}(r_i)$ for every $i \in [1..n]$;
 (ii) $S_1 > S_2$ if $V_{S_1} < V_{S_2}$, i.e. if $V_{S_1} \leq V_{S_2}$ and for some
 $i \in [1..n]$ $v_{S_1}(r_i) < v_{S_2}(r_i)$.

A set of literals S is an *preferred (optimal) model* of
 an ASO program $\langle \mathcal{P}, \Phi \rangle$ if S is an answer set of \mathcal{P} and
 there is no answer set S' of \mathcal{P} such then $S' > S$.

Given an ASO program $\langle \mathcal{P}, \Phi \rangle$, its computational
 complexity is one level above the complexity of the
 generating program \mathcal{P} (Brewka et al., 2003).

ASO_{ch} Semantics. The generalization of ASO se-
 mantics, which captures the intuition depicted in the
 Example 1, is called the ASO_{ch} semantics. Diffe-
 rently from the ASO semantics, the ASO_{ch} seman-
 tics looks beyond the single preference rules, consi-
 dering them as a tool for choice representation: head
 atoms correspond to the *choice options*, whereas the
 body of preference rule specifies the *choice context*,
 i.e. the decisions which have to precede this choice.
 The ASO_{ch} semantics admits only atoms or disjunc-
 tion of atoms in the head of preference rules. If C_i is a
 disjunction of atoms, all atoms in C_i are equally good
 choice options; atoms in C_i are preferred to the atoms
 in C_j , for $i < j$ and $i, j \in [1..k]$. The evaluation stra-
 tegy consists in the identification of choices and in the
 comparison of choices instead of single rules in order
 to select preferred models.

More specifically, the partition of Φ into a *set of*
choices (subset of preference rules), denoted by
 $\text{Ch}(\Phi)$, is performed following the *choice identifica-*
tion strategy: two preference rules ρ_1 and ρ_2 *define*
the same choice Ch , denoted by $\rho_1, \rho_2 \in \text{Ch}$, if (i) ρ_1
 and ρ_2 have at least one common atom in their heads;
 and (ii) $\exists \rho_3$ such that $\rho_1, \rho_3 \in \text{Ch}$ and $\rho_3, \rho_2 \in \text{Ch}$.

The ASO_{ch} semantics presumes that prioritized pro-
 grams are *well-formed*, i.e. each set of preference
 rules ρ_1, \dots, ρ_k defining a choice is specified over al-
 ternative contexts. This hypothesis ensures that in
 each model M at most one of the different contexts
 of a choice is satisfied, i.e. \forall model M , \forall Ch , there is
 at most one preference rule $\rho \in \text{Ch}$, denoted as *active*
in M , such that *body*(ρ) is satisfied in M .

After constructing the set of choices $\text{Ch}(\Phi)$, the pre-
 ferred stable models of $\langle \mathcal{P}, \Phi \rangle$ are computed by as-
 sociating to each model M of \mathcal{P} a satisfaction vec-
 tor reporting the degree of satisfaction of each choice
 $\text{Ch} \in \Phi$. The evaluation strategy of ASO_{ch} semantics
 is illustrated by the following example.

Example 2 Consider the prioritized program
 $\langle \mathcal{P}_1, \Phi_1 \rangle$ reported in Example 1. The preference
 rules $\{\rho_1, \rho_2\} \in \Phi_1$, having common atoms in their
 heads, define the same choice, say Ch_{dr} , whereas
 the last preference rule ρ_3 defines another choice,
 say Ch_{d} . Consequently, the set of preference rules
 $\{\rho_1, \rho_2, \rho_3\} \in \Phi_1$ is partitioned into two different
 choices: $\text{Ch}_{\text{dr}} = \{\rho_1, \rho_2\}$ and $\text{Ch}_{\text{d}} = \{\rho_3\}$ that models
 the choice of drink and dessert respectively. The
 choice satisfaction vectors are $V_{M_1} = [2, 1]$, $V_{M_2} = [2, 2]$,
 $V_{M_3} = [2, 2]$, consequently, M_1 is the preferred model
 owing to the dessert choice. Note that ASO semantics
 gives $V_{M_1} = [2, I, 1]$, $V_{M_2} = [2, I, 2]$, $V_{M_3} = [I, 2, 2]$ and
 returns M_1 and M_3 as preferred models. It does
 not admit the comparison of M_1 and M_3 owing to
 opposite satisfaction degrees of ρ_1 and ρ_2 . \square

Given a prioritized program $\langle \mathcal{P}, \Phi \rangle$, the ASO_{Ch} semantics coincides with the ASO semantics if $\nexists \rho_1, \rho_2 \in \Phi$ having at least one common atom in their heads.

The computational complexity of ASO_{Ch} semantics is not increased w.r.t. the ASO semantics. In fact, the choice identification can be done in polynomial time; while given a model M the evaluation of its choice satisfaction vector V_M can be performed by (i) identifying the active rule for each choice, which can be done in polynomial time; and (ii) establishing the satisfaction degrees of the active rules as in the case of ASO semantics.

ASO_{FCh} Semantics. Given a choice Ch and a model M , the ASO_{Ch} semantics evaluates M on the basis of the selected option in the active preference rule $\rho \in Ch$. Observe that the head atoms of ρ describe the possible options of Ch and the preference among them, without taken into account their *feasibility*, i.e. the really possibility of selecting these options during the choice. This property depends on the constraints present in logic program and is determined by the *choice context* in a given model, i.e. by the set of atoms whose selection precede this choice. An alternative to the ASO_{Ch} semantics evaluating the choices on the basis on their really possible options is called ASO_{FCh} semantics.

Example 3 Consider the prioritized program $\langle \mathcal{P}_3, \Phi_3 \rangle$, describing different menus and the preferences among drinks and desserts:

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r1 : fish ⊕ beef ←          c1 : ← beef, pie
r2 : red ⊕ white ⊕ beer ←   c2 : ← fish, ice-cream
r3 : pie ⊕ ice-cream ←
ρ1 : white > red > beer ← fish
ρ2 : red ∨ beer > white ← beef
ρ3 : pie > ice-cream ← beer

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The generating program \mathcal{P}_3 has six stable models: $M_1 = \{\text{fish, white, pie}\}$, $M_2 = \{\text{fish, red, pie}\}$, $M_3 = \{\text{fish, beer, pie}\}$, $M_4 = \{\text{beef, white, ice-cream}\}$, $M_5 = \{\text{beef, red, ice-cream}\}$ and $M_6 = \{\text{beef, beer, ice-cream}\}$. Φ_3 defines the set of choices $Ch = \{Ch_{dr}, Ch_d\}$, where $Ch_{dr} = \{\rho_1, \rho_2\}$ and $Ch_d = \{\rho_3\}$ describe the choices of drinks and desserts respectively.

In order to establish the set of preferred solutions the ASO_{Ch} semantics constructs the choice satisfaction vector for each model: $V_{M_1} = [1, I]$, $V_{M_2} = [2, I]$, $V_{M_3} = [3, 1]$, $V_{M_4} = [2, I]$, $V_{M_5} = [1, I]$, $V_{M_6} = [1, 2]$ and obtain M_1 and M_5 as a result. The ASO_{FCh} semantics takes into account the feasibility of choice options and substitutes $V_{M_6} = [1, 2]$ by $V'_{M_6} = [1, 1]$ following the intuition that its option of the dessert choice (*ice-cream*) is the unique (and, consequently, best) possible option. Thus M_6 is also a preferred model. \square

The implementation of the ASO_{FCh} semantics requires the test of the feasibility of choice options, which can be performed in a polynomial time. Thus, the computational complexity of ASO_{FCh} semantics is not increased w.r.t. the ASO_{Ch} semantics (Caroprese et al., 2007).

3 THE CHOPPER SYSTEM

CHOPPER (Choice Optimizer for Prioritized Reasoning) is an answer set optimization system realizing prioritized reasoning based on the choice evaluation.

The system prototype has been developed on top of the well-known DLV prover (Leone et al., 2002) by using Java 2 Platform. In particular, the answer set evaluation is performed with the DLV system, whereas the prioritized reasoning and the user interface are realized by means of personalized java procedures. We point out that besides DVL several other deductive systems based on stable model semantics have been proposed in the literature (Smodels, DeRes) (Cholewinski et al., 1996; Syrjanen and Niemela, 2001). The choice of DLV is orthogonal w.r.t. the development of the prototype, since CHOPPER can exploit any other deductive systems proposed in the literature.

Using of CHOPPER. CHOPPER receives in input the prioritized program and the specification of the semantics to be applied and extracts and returns the preferred stable models as a result. The system can be used by means of a *User Interface - UI* - which allows to specify (i) the prioritized program - $\langle \mathcal{P}, \Phi \rangle$, (ii) the semantics - *CS* - chosen among the semantics discussed in the paper and allows to visualize the obtained result, i.e. the set of preferred stable models - *PSM*.

The running of CHOPPER for the program $\langle \mathcal{P}_3, \Phi_3 \rangle$ under the ASO_{FCh} semantics is provided in Figure 1. In more details, the prioritized program is provided to the system by means of two different textual files, written following the DLV syntax, containing the generating program and the preference program, respectively. Note that in the preference program the system admits the symbol “>” for expressing the preference order among choice options.

CHOPPER also offers the possibility of comparing two different semantics by evidencing the set of models common to both the semantics. The result of comparison of ASO_{Ch} and ASO_{FCh} for the program $\langle \mathcal{P}_3, \Phi_3 \rangle$ is provided in Figure 2. The symbol “*”

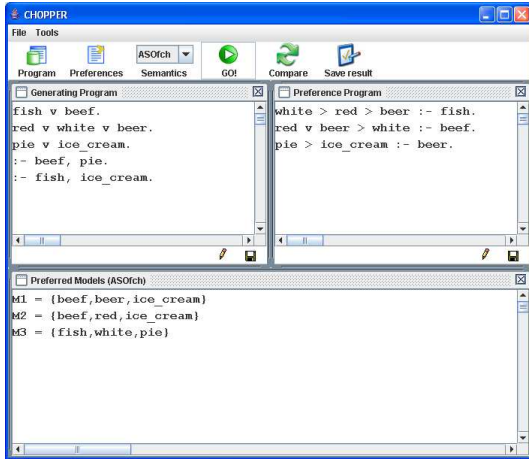


Figure 1: The UI of CHOPPER.

marks the first model provided by the ASO_{Fch} semantics and states that this model is not provided by the ASO_{Ch} semantics.

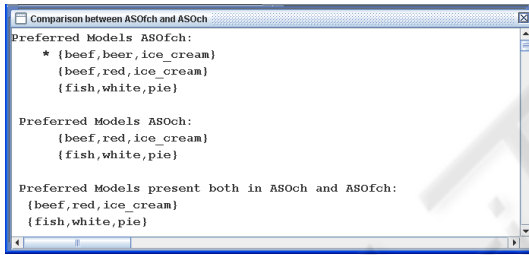


Figure 2: Comparison result.

CHOPPER Architecture. The overall architecture of the CHOPPER prototype is reported in Figure 3. The core of the system consists in two main blocks: the *Analyzer Block* and the *Evaluator Block* described below.

The *Analyzer Block* is constituted by two different modules: the \mathcal{P} -analyzer and the Φ -analyzer.

- The Φ -analyzer takes in input the set of preference rules Φ and the chosen semantics CS , and computes $Ch(\Phi)$, i.e. the partition of Φ into the set of choices by following the choice identification strategy.
- The \mathcal{P} -analyzer receives in input the generating program \mathcal{P} and additional information gained from Φ -analyzer, and invokes the DLV prover (Leone et al., 2002) in order to obtain the stable models of the program \mathcal{P} , $SM(\mathcal{P})$.
- As an additional task, in the case of the ASO_{Fch}

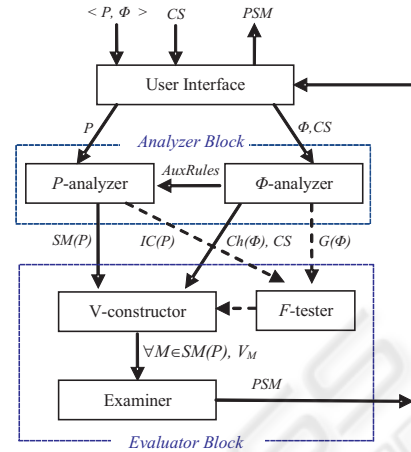


Figure 3: The architecture of CHOPPER.

semantics, the Φ -analyzer establishes the precedence relation among choices, while the \mathcal{P} -analyzer extracts the set of constraints of \mathcal{P} , $IC(\mathcal{P})$, necessary for testing the feasibility of choice options.

The *Evaluator Block* is responsible for PSM discovering and is constituted by three modules: the *F-tester*, the *V-constructor* and the *Examiner*.

- The *F-tester* is able to test the feasibility of choice options.
- The *V-constructor* receives in input $SM(\mathcal{P})$, CS and $Ch(\Phi)$, and constructs the (feasible) choice satisfaction vector V_M for each stable model $M \in SM(\mathcal{P})$.
- The *Examiner* module, performs the comparison of the choice satisfaction vectors so that establishing the set of preferred stable models PSM . The result is provided to the UI responsible of user's communication.

4 IMPLEMENTATION

In the following we detail the implementation of the main modules constituting the *Analyzer Block* and the *Evaluator Block*.

4.1 Φ -analyzer

The Φ -analyzer examines the structure of preference rules in order to establish the partition of Φ into the set of choices $Ch(\Phi)$ and to determine the order among choices. Moreover, it produces the set of auxiliary

rules necessary to simplify the construction of the choice satisfaction vector without increasing the computational complexity of \mathcal{P} .

Choice identification. The choice identification process performed by the Φ -*analyzer* depends on the selected semantics and is based on the choice identification strategy. In the case of ASO semantics each rule is associated with a different choice. In the case of ASO_{Ch} or ASO_{FCh} semantics the rules, having at least one common atom in their heads, are assigned to the same choice. The algorithm, performing this task, is presented in Figure 4. It takes in input a set of preference rules Φ and returns the partition of Φ into the set of choices $\text{Ch}(\Phi)$.

Algorithm 1 Choice Identification
Input: A set of preference rules: Φ ;
Output: A set of choices: $\text{Ch}(\Phi)$;
begin
 $\text{Ch}(\Phi) = \{\}$;
while (Φ is not empty) {
 $\rho = \text{extractFirstElement}(\Phi)$;
 $\text{Ch} = \{\rho\}$; $\Phi = \Phi - \{\rho\}$;
 $\text{update} = \text{true}$;
while ($\text{update} == \text{true}$) {
 $\text{update} = \text{false}$;
for each $\rho' \in \Phi$
for each $\rho \in \text{Ch}$
if ($\text{sameChoice}(\rho, \rho')$) {
 $\Phi = \Phi - \{\rho'\}$; $\text{Ch} = \text{Ch} + \{\rho'\}$;
 $\text{update} = \text{true}$; }
} $\text{Ch}(\Phi) = \text{Ch}(\Phi) + \text{Ch}$;
} **return** $\text{Ch}(\Phi)$;
end.

Figure 4: Choice Identification Algorithm.

At each step of the external iteration process the algorithm selects the first preference rule ρ from Φ and identifies and extracts all preference rules from Φ defining the same choice of ρ , called Ch . The function *sameChoice* receives in input two preference rules ρ and ρ' and returns *true* if they have at least one common atom in their heads. In the positive case, ρ' is added to Ch . The transitive property is supported by the nested *while*-block, which ensures that all preference rules defining the same choice are considered together. For the sake of simplicity of presentation, the optimization details, regarding the implementation of the transitive property are not reported in the above algorithm.

Choice ordering. The relationship among choices can be established by means of the *Choice depen-*

ency graph described below.

Definition 1 *Choice dependency graph:* Given a set of choices $\Phi = \{\text{Ch}_1, \dots, \text{Ch}_n\}$, its *dependency graph* $G(\Phi)$ is an oriented graph $\langle N, E \rangle$, where N denotes the set of nodes associated to choices, i.e. $N = \{\text{Ch}_1, \dots, \text{Ch}_n\}$, and E denotes the set of edges, $E = \{e(\text{Ch}_i, \text{Ch}_j) \mid \exists \rho_2 \in \text{Ch}_j, \rho_1 \in \text{Ch}_i \text{ and } \exists \text{ atom } A, \text{ s.t. } A \in \text{body}(\rho_2) \wedge A \in \text{head}(\rho_1)\}$. \square

The implementation of $G(\Phi)$ is really simple as its nodes correspond to the choices, whereas the presence of the edge $e(\text{Ch}_i, \text{Ch}_j)$ can be tested by considering the choice options of Ch_i and the positive body atoms, present in each $\rho \in \text{Ch}_j$.

Definition 2 Given a set of choices $\Phi = \{\text{Ch}_1, \dots, \text{Ch}_n\}$ and its *dependency graph* $G(\Phi)$, we say that the choice Ch_i *precedes* Ch_j , denoted by $\text{Ch}_i \prec \text{Ch}_j$, if there is a path from the vertex Ch_i to the vertex Ch_j , whereas there is no path from Ch_j to Ch_i . \square

Previous definition allows the computation of the set of choices precedent to a choice Ch : Ch' precedes Ch if there is a path from the vertex Ch' to the vertex Ch ; moreover, it takes into account the intuition that we cannot establish a precedence relation between two choices if they are involved in a cycle.

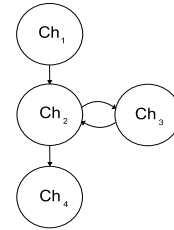


Figure 5: Choice Dependency Graph.

Example 4 Consider the Choice dependency graph presented in Figure 5. There are two nodes Ch_2 and Ch_3 involved in a cycle, thus we cannot establish the precedence relation between the corresponding choices. On the other hand, we can deduce, that $\text{Ch}_1 \prec \text{Ch}_2$, $\text{Ch}_1 \prec \text{Ch}_3$, $\text{Ch}_1 \prec \text{Ch}_4$, $\text{Ch}_2 \prec \text{Ch}_4$ and $\text{Ch}_3 \prec \text{Ch}_4$. \square

4.2 V-constructor and F-tester

The *V-constructor* and the *F-tester* modules collaborate in order to construct the choice satisfaction vector for each stable model of the generating program.

Choice Satisfaction Vector. The *V-constructor* receives in input $\mathcal{SM}(\mathcal{P})$, \mathcal{CS} and $\text{Ch}(\Phi)$, and constructs the choice satisfaction vector V_M for each sta-

ble model $M \in \mathcal{SM}(\mathcal{P})$ following the chosen semantics \mathcal{CS} . More specifically, given a model M , $V_M = [[v(\text{Ch}_1), \dots, v(\text{Ch}_n)]]$ reports the satisfaction degree of each choice $\text{Ch}_i \in \text{Ch}(\Phi)$.

In the case of ASO_{Ch} semantics each choice Ch_i is described by the set of preference rules, but at most one of them is active in M and $v(\text{Ch}_i)$ reports its satisfaction degree. The cases than (i) no preference rule is active in M or (ii) there exists an active rule, but no choice option described by it is present in M , are considered as the best cases, i.e. $v(\text{Ch}_i) = 1$.

In the case of the ASO semantics each choice is described by a unique preference rule, thus V_M coincides with the satisfaction vector proposed in (Brewka et al., 2003).

If the ASO_{FCh} is chosen, V_M has to be constructed by taking into account the feasibility of choice options. Consider, for instance, the choice Ch_i and suppose the choice option d -th to be present in M , whereas k options preceding it are not feasible in M , then $v(\text{Ch}_i) = d - k$. The *V-Constructor* performs this task by interacting with the *F-tester* module, able to analyze the feasibility of choice options.

Feasibility Testing. The *F-tester* module is in charge of verifying the feasibility of the choice options present in a preference rules. The goodness of a model depends on the degree of satisfaction of a set of preference rules; anyhow, the measure of the degree of satisfaction of a preference rule, say ρ , active w.r.t. a model M should be evaluated by only considering the set of choice options feasible in M . In order to better depict this intuition, let's examine the program $\langle \mathcal{P}_3, \Phi_3 \rangle$ and the model M_6 reported in Example 3. ρ_3 is active in M_6 and as the choice of *pie* is not feasible in M_6 , *ice-cream* becomes the best choice option as it is the only feasible one.

In order to capture this behavior the *F-tester* module computes the context of a choice Ch in a model M as the subset of atoms in M whose selection precedes Ch . Intuitively, this subset of atoms can be individuated (i) by analyzing the choice dependency graph in order to establish the set of choices preceding Ch , (ii) by identifying the atoms selected during these choices, or whose selection precede these choices.

More formally, given a set of choices $\text{Ch}(\Phi)$, the *context of a choice* Ch w.r.t. a model M , denoted by $\text{ctx}_M(\text{Ch})$, is the conjunction $\bigwedge_{\rho' \in \Omega(\text{Ch}, M)} (\text{body}(\rho') \wedge \text{best_head}(\rho')) \wedge \text{body}(\rho)$, where $\Omega(\text{Ch}, M) = \{\rho' \mid \exists \text{Ch}' \in \Phi \text{ s.t. } \text{Ch}' \prec \text{Ch} \wedge \rho' \in \text{Ch}' \wedge M \models \text{body}(\rho')\}$, $\rho \in \text{Ch}$ is s.t. $M \models \text{body}(\rho)$ and $\text{best_head}(\rho')$ is the best choice option of ρ' belonging to M .

Example 5 Consider the prioritized program $\langle \mathcal{P}_3, \Phi_3 \rangle$ from the Example 3. Notice that *beer* is

the option of the drink choice and that each model M containing *beer* performs this choice following either the preference rule ρ_1 or ρ_2 , i.e. before *beer*, M has selected either *fish* or *beef*. Let's consider the model M_6 : ρ_3 is active in M_6 as it contains *beer*, and *beer* follows the selection of *beef*. Thus the choice described by ρ_3 works in the presence (context) of $\{\text{beef}, \text{beer}\}$. \square

Given a model M and a choice Ch , the module is therefore in charge of testing the feasibility of the options expressed in Ch by considering the constraints of the generating program and the context of Ch in M .

Definition 3 *Feasibility of choice options:* Given a prioritized program $\langle \mathcal{P}, \Phi \rangle$ defining the set of choices, a stable model M of \mathcal{P} , a choice $\text{Ch} \in \text{Ch}(\Phi)$ and a preference rule $\rho \in \text{Ch}$, then $A_i \in \text{head}(\rho)$ is a *feasible* choice option w.r.t. M iff $A_i \wedge \text{ctx}_M(\text{Ch}) \models \text{IC}(\mathcal{P})$ and is *unfeasible* otherwise. \square

The test of the feasibility is performed by the *F-tester* module by constructing for each tested choice option a logic program which is satisfied if the tested option is feasible and fails otherwise. This program contains (i) the atoms of the choice context and a tested atom (choice option) as facts and (ii) the set of constraint $\text{IC}(\mathcal{P})$ of the generating program \mathcal{P} . The *F-tester* invokes the DLV prover in order to solve this program and if the program does not have stable models deduces that the tested option is unfeasible, or deduces that it is feasible otherwise.

5 OTHER APPROACHES

The computation of preferred models proposed in (Brewka et al., 2003) and (Wakaki et al., 2003) for ASO semantics and PLP semantics respectively is based on the use of a *tester program*¹. In more details, given a prioritized logic program $\langle \mathcal{P}, \Phi \rangle$, the tester program takes in input a candidate solution S of \mathcal{P} and looks for the other solutions of \mathcal{P} (strictly) preferred w.r.t. S . This latter task is performed by generating the solutions of \mathcal{P} and by comparing them with S .

The ASO semantics permits a direct comparison of two solutions, thus the tester program can be used as follows. The computation starts with an arbitrary solution S of \mathcal{P} . If the tester fails, S is an optimal solution and the computational process stops. Otherwise, a strictly better solution S_1 is discovered and the tester

¹A similar technique was also implemented in (Janhunen et al., 2000; Brewka, 2002; Sakama and Inoue, 2000).

is run with S_1 as input. This process continues until an optimal solution is reached.

On the contrary, \mathcal{PLP} semantics introduces the transitive property of preference relation, thus admits the preference relation between two solutions which are not directly comparable. Consequently, in this case two-step procedure is needed: in the first step all the direct preference relations among solutions have to be established; then the transitive relations can be discovered and the final conclusion can be derived. Wakaki et. al in (Wakaki et al., 2003) implement the direct comparisons by testing each answer set of \mathcal{P} with a tester program; and then create an auxiliary logic program which extracts all preferred solutions on the basis of the preference relations generated at the previous step and those discovered by using the transitive property.

In the general case, both the approaches, previously described, need to perform more calls to the tester program which is in charge of computing the set of solution of \mathcal{P} . The approach adopted in CHOPPER aims to avoid the redundant computation of the set of solution of \mathcal{P} , performed during each call to the tester program. CHOPPER uses once the logic prover to find the set of solutions of the problem, and realizes the prioritized reasoning by means of personalized comparison procedures.

6 CONCLUSION

In this paper the implementation of prioritized reasoning in logic programming has been discussed. In particular, the case of preference relation among atoms has been investigated and a system, called CHOPPER, has been described. This system realizes choice optimization in logic programming by implementing the ASO_{Ch} and ASO_{FCh} semantics recently proposed in (Caroprese et al., 2007), and supports the ASO semantics (Brewka et al., 2003). In this paper the architecture of the system has been presented and aspects of the choice identification strategies and of the feasibility of choice options has been discussed. Moreover, the comparison of the proposed approach with the other implementation approaches proposed in the literature has been provided.

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