Intensional Sets in CLP

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Abstract. We propose a parametric introduction of intensionally defined sets into any $CLP(\mathcal{D})$ language. The result is a language $CLP(\{\mathcal{D}\})$, where constraints over sets of elements of $\mathcal{D}$ and over sets of sets of elements, and so on, can be expressed. The semantics of $CLP(\{\mathcal{D}\})$ is based on the semantics of logic programs with aggregates and the semantics of CLP over sets. We investigate the problem of constraint resolution in $CLP(\{\mathcal{D}\})$ and propose algorithms for constraints simplification.

Keywords. *Constraint Logic Programming, Sets, Aggregates.*

1 Introduction

The literature is rich of proposals aimed at developing declarative programming frameworks that incorporate different types of *set-based primitives* (e.g., [13, 2, 5, 6, 1]). These frameworks provide a high level of data abstraction, where complex algorithms can be encoded in a natural fashion, by directly using the popular language of set theory. These features make this type of languages particularly effective for modeling and rapid prototyping of algorithms.

A recognized downside of most of the existing languages embedding sets is the focus on extensional set constructions [6, 13, 15, 11] and/or the severe restrictions imposed on the use of *intensional set constructions* [2, 20]. *Intensionally* defined sets (a.k.a. *intensional sets*) are collections of elements where the membership is decided by properties instead of enumeration. There is significant evidence that the ability to handle general intensional sets can drastically simplify the development of solutions to complex problems, leading to more convenient languages and compact programs—e.g., [16] suggests that encodings of traditionally hard problems in this type of notations are more compact that SAT encodings.

In this work we propose a parametric introduction of intensionally defined sets into any $CLP(\mathcal{D})$ language—e.g., $\mathcal{D}$ can be $\mathbb{FD}$ for Finite Domains or $\mathbb{R}$ for Real numbers. Given a language $CLP(\mathcal{D})$ and its interpretation domain D, we define the domain $\mathcal{U}_D$, which is used to construct an intuitive interpretation for intensional sets and set-based constraints (Section 3). We define a new language, $CLP(\{\mathcal{D}\})$, where $\mathcal{D}$-constraints can be expressed, as well as arbitrarily nested extensional and intensional sets of elements over $\mathcal{D}$ and constraints over these entities. In this new framework the declarative style of constraint-based programming is enhanced by the availability of well-known set-based constructors. The development of a semantics for $CLP(\{\mathcal{D}\})$ introduces problems analogous to those studied in the context of logic programming with aggregates. In Section 2 we explain the relationships between aggregates and intensional sets.

The semantic characterization of $CLP(\{\mathcal{D}\})$ (Section 4) is provided as a generalization of Gelfond and Lifschitz stable model semantics; this allows us to provide a semantics to a larger class of programs than various previously proposed schemes (e.g., [2,9])—in particular those relying on the use of stratification. E.g., a simple program such as

$$\begin{array}{ll} q(1). & q(Z) \leftarrow Y = \{X \,:\, p(X)\} \wedge Z \in Y. \\ p(2). & p(Z) \leftarrow Y = \{X \,:\, q(X)\} \wedge Z \in Y. \end{array}$$

has the stable model $I = \{q(1), q(2), p(1), p(2)\}$, in spite of not being stratified.

In Section 5 we build on our previous research on constraint solving in presence of sets [6] to provide an incomplete solver for constraints of $CLP(\{\mathcal{D}\})$. The proposed solver simplifies constraints to a solved form. In particular, the main goal is to eliminate occurrences of intensional sets from the constraints *without* explicit enumeration of the sets. Some constraints between intensional sets are easily seen to be undecidable (e.g., $\{X : p(X)\} = \{X : q(X)\}$); this prevents us from developing a parametric and *complete* constraint solver. To address this issue we subdivide the constraint solving process in two phases. The first phase (*propagation*) consists of rewriting rules, that take advantage of the semantics of set operation to avoid the explicit computation of intensional sets. The second phase (*labeling*) forces the removal of intensional sets—via translation to formulae containing negation and/or explicit collection of solutions. The intuition is that, while propagation is transparently performed whenever possible, labeling should be explicitly requested—being a potentially unsafe and expensive step.

2 Related work

A number of proposals have been made to support the introduction of aggregate functions in deductive databases and logic programming. Among them we discuss [14,21,19,18,4], where an *aggregate subgoal* is a constraint of the form

$$E = \mathbb{F}(\{\!\!\{\, e[\bar{X}, \bar{Y}, \bar{Z}] \,:\, (\exists \bar{Z})\, p[\bar{X}, \bar{Y}, \bar{Z}] \,\}\!\!\}) \qquad (1)$$

whose intuitive semantics is: given values for the variables $\bar{Y}$ (grouping variables), collect in a multiset all the expressions involving $\bar{X}$ such that there are values for the variables $\bar{Z}$ (local variables) for which $p[\bar{X}, \bar{Y}, \bar{Z}]$ holds. The function $\mathbb{F}$ is finally applied to this multiset. The constraint aggregate (1) is written as: $E = \mathbb{F}e[\bar{X}, \bar{Y}, \bar{Z}] : p[\bar{X}, \bar{Y}, \bar{Z}]$ in [18], and as: group_by$(p[\bar{X}, \bar{Y}, \bar{Z}], [\bar{Y}], E = \mathbb{F}(e[\bar{X}, \bar{Y}, \bar{Z}]))$ in [14].

In this paper we consider a particular function $\mathbb{F}$: $\mathbb{F}(S)$ returns the set of all elements in the multiset S—i.e., it removes multiple occurrences of the same element. On one hand, this may appear as a simplification—there is no need to compute possibly complex functions. On the other hand, this leads to a number of complications. In particular, sets have to be introduced as first-class citizens of the language and, hence, they must be properly dealt with.

The work [18] provides a minimal model semantics for *monotonic* programs, i.e., programs for which the T_P operator is monotonic. However, monotonicity is in general undecidable, and the syntactic restrictions they impose to ensure it are rather strong. For instance, the program:

$$\begin{array}{lll} r(0). & r(1). & p(1,E) \leftarrow E = \text{MIN}\,\{X \,:\, r(X)\}\,. \\ s(1). & & p(1,E) \leftarrow E = \text{MIN}\,\{X \,:\, s(X)\}\,. \end{array}$$

is not accepted—the argument of a predicate defined in terms of an aggregate is required to depend functionally on the other arguments (cost-consistency). Moreover, $\mathbb{F}$ can be only a simple function of the elements of the aggregate (such as SUM, MIN, MAX).

In [14] the authors introduce aggregate subgoals and they investigate both the 3-valued Well-Founded Semantics and the 2-valued Stable Model Semantics for programs containing aggregations. In [19] the authors investigate the problem of checking satisfiability for programs with aggregate subgoals when the function $\mathbb{F}$ is a simple *SQL aggregate function*. The well-founded and stable model semantics for logic programs with aggregates has been extended through the use of approximation theory in [4].

The work that comes closer in spirit to what we propose here is [21], where Van Gelder provides a treatment of aggregates based on the capability of expressing a collection (findall) of answers to a predicate. The idea is that the function $\mathcal{F}$ can be easily programmed on top of this aggregate capability. Similarly to [5, 2], Van Gelder shows how to program findall using negation. The definition of set-grouping can exploit the following intended semantics of intensional sets [8]:

$$\begin{aligned} E = \{X \,:\, p(X)\} &\leftrightarrow \forall X(X \in E \rightarrow p(X)) \wedge \forall X(p(X) \rightarrow X \in E) \\ &\leftrightarrow \forall X(X \in E \rightarrow p(X)) \wedge \neg\exists X(X \notin E \wedge p(X)). \end{aligned} \tag{2}$$

The first subformula can be computed (for finite sets) using a recursively defined predicate—this corresponds to what authors have called *restricted universal quantification* [6, 15]—while the second subformula can be expressed as a negated predicate, whose single clause definition has $X \notin E \wedge p(X)$ as its body. Semantics is therefore reduced to the semantics of Prolog programs with negation. In [21] well-founded semantics is employed to handle negation in this framework, while in [8] a general form of constructive negation is used. This approach is very general. Nevertheless, the direct transformation approach used to handle aggregations via negation has drawbacks. Semantic characterization of negation in logic programming is fairly involved and it is not clear how it reflects on the corresponding meaning attributed to the aggregate constructs. Furthermore, if the desire is to allow a general use of sets as first-class citizens of the language, then one needs to explore the interactions between sets and negation—which are not straightforward [7]. Moreover, investigating the semantics of aggregation through translation to other constructs hampers the development of implementation techniques directly targeted to aggregation—especially implementations based on constraint solving and delaying techniques.

Our investigation builds on the existing work dedicated to the development of a *CLP* language over sets. The language $CLP(\mathcal{SET})$ [6] is an instance of the *CLP* framework, whose constraint domain is that of *hereditarily finite sets*.

The language $CLP(\mathcal{SET})$ allows sets to be *nested* and *partially specified*, e.g., set elements can contain unbound variables and sets can be only partially enumerated. $CLP(\mathcal{SET})$ provides a collection of primitive constraint predicates sufficient to cover all the basic set-theoretic operations. In [6] we presented a complete constraint solver capable of deciding the satisfiability of arbitrary conjunctions of these primitive set constraints. Intensional sets are allowed, but they are rewritten [8] according to the technique sketched in formula (2). Moreover, $CLP(\mathcal{SET})$ does not interoperate with other constraints solvers.

3 Syntax

Following the notation of [12], let $\mathcal{D}$ be an arbitrary constraint domain and $\mathcal{C}_\mathcal{D}$ be the class of *admissible constraints* for $\mathcal{D}$. We assume that $\mathcal{C}_\mathcal{D}$ is closed under negation. The language we propose has a signature $\Sigma = \langle \Pi_\mathcal{S}, \Pi_\mathcal{D}, \Pi_\mathcal{P}, \mathcal{F}, \mathcal{V}\rangle$ ($\mathcal{S}$ stands for *Set*, $\mathcal{D}$ for *Domain*, and $\mathcal{P}$ for *Program*). Intuitively, $\Pi_\mathcal{S}$ provides constraint predicates to handle sets, $\Pi_\mathcal{D}$ provides the constraint predicates inherited from the underlying constraint domain $\mathcal{D}$, while $\Pi_\mathcal{P}$ contains the user-defined predicates. In particular, we assume that $\Pi_\mathcal{S}$ contains $\in, \cup_3, \cap_3, \subseteq, ||$ (these are the basic set predicates used in [6]). We also assume that $=$ is present in $\Pi_\mathcal{S} \cap \Pi_\mathcal{D}$. Furthermore, $\mathcal{F} = \mathcal{F}_\mathcal{S} \cup \mathcal{F}_\mathcal{D} \cup \mathcal{F}_\mathcal{P}$, where $\mathcal{F}_\mathcal{S}$ contains function symbols used to create set terms, $\mathcal{F}_\mathcal{D}$ contains the function symbols provided by the language of $\mathcal{D}$, and $\mathcal{F}_\mathcal{P}$ contains free function symbols. In particular $\emptyset$ and the binary set-constructor $\{\cdot \mid \cdot\}$ [6] are expected to be in $\mathcal{F}_\mathcal{S}$. The term $\{s \mid t\}$ denotes the set $\{s\} \cup t$. $\mathcal{F}$ and $\Pi_\mathcal{S}$ allow us to write terms and constraints regarding extensionally defined finite sets, such as $\emptyset, \{a, b\}, \{\emptyset, \{a, b\}\}$. The set $\mathcal{V}$ contains a countable number of variables, separated in the three sorts D, P, and S, described below.

Definition 1. *We allow three sorts* D, P, *and* S *for* terms.

- *For* $\mathcal{X} \in \{\mathcal{D}, \mathcal{P}, \mathcal{S}\}$, *constant symbols from* $\mathcal{F}_\mathcal{X}$ *are terms of sort* X.
- *We assume that function symbols from* $\mathcal{F}_\mathcal{D}$ *of arity* n *have the sort* $\mathsf{D}^n \longrightarrow \mathsf{D}$.
- *The sort of* $\{\cdot \mid \cdot\}$ *is* $(\mathsf{D} \cup \mathsf{P} \cup \mathsf{S}) \times \mathsf{S} \longrightarrow \mathsf{S}$.
- *The sort of a function symbol from* $\mathcal{F}_\mathcal{P}$ *of arity* n *is:* $(\mathsf{D} \cup \mathsf{P} \cup \mathsf{S})^n \longrightarrow \mathsf{P}$.

Definition 2. Atoms *are defined as follows:*

- *If* $p \in \Pi_\mathcal{P}$ *with arity* n *and* $t_1, \dots, t_n$ *are terms (of any sort), then* $p(t_1, \dots, t_n)$ *is a* $\mathcal{P}$*-atom.*
- *If* $p \in \Pi_\mathcal{D}$ *with arity* n *and* $t_1, \dots, t_n$ *are terms of sort* D, *then* $p(t_1, \dots, t_n)$ *is a domain constraint atom (or simply a* $\mathcal{D}$*-atom).*
- *If* $p \in \Pi_\mathcal{S}$ *with arity* n *and* $t_1, \dots, t_n$ *are terms of sort* S, *then* $p(t_1, \dots, t_n)$ *is a* $\mathcal{SET}$-constraint atom. *The only exception is represented by the symbol* $\in$*: if* t *is a term of any sort and* s *is a term of sort* S, *then* $t \in s$ *is a* $\mathcal{SET}$*-constraint atom, as well.*
- *If* $E \in \mathcal{V}$, $t_1, \dots, t_n$ *are terms and* $p \in \Pi_\mathcal{P}$ *with arity* n, *then* $\{X : p(t_1, \dots, t_n)\}$ *is an* intesional set *and* $E = \{X : p(t_1, \dots, t_n)\}$ *is an* aggregate constraint atom. *The sort of* E *is* S.

A $\mathcal{S}$-atom *is either an* $\mathcal{SET}$*-constraint or an aggregate constraint atom.*

Observe that the aggregate constraint atom is exactly the constraint aggregate (1) where $\mathbb{F}$ is the function that removes duplicates from a multiset and the expression e is simply X. It is immediate to see that limiting our attention to this form of aggregation does not lead to any loss of generality. In particular, the variables indicated with $\bar{Y}$ in (1) can appear as arguments of $p(t_1 \dots, t_n)$, while the local (existentially quantified) variables can be directly placed within the program rules defining p.

Definition 3. *A* $\mathcal{D}$-admissible constraint *is any formula belonging to the class of constraints* $\mathcal{C}_{\mathcal{D}}$.

A $\mathcal{S}$-admissible constraint *is a propositional combination of* $\mathcal{S}$*-atoms. For the sake of simplicity we do not allow the use of negation applied to aggregate constraint atoms. Let us denote with* $\mathcal{C}_{\mathcal{S}}$ *the class of* $\mathcal{S}$*-admissible constraints.*

A $\{\mathcal{D}\}$-admissible constraint *is an arbitrary propositional composition of* $\mathcal{D}$*-admissible and* $\mathcal{S}$*-admissible constraints. We will denote with* $\mathcal{C}_{\{\mathcal{D}\}}$ *the class of all* $\{\mathcal{D}\}$*-admissible constraints.*

Example 1. Assume that $\mathcal{D} = \mathbb{FD}$ is the constraint domain for *finite domain constraints*. Thus, constraints such as $X \text{ in } 1..5, Y \text{ in } 2..4, X < Y$ belong to $\mathcal{C}_{\mathcal{D}}$. An example of $\{D\}$-admissible constraint is:

$$A \text{ in } 1..2, B \text{ in } 2..3, E = \{X : p(A, X, B)\}, K = \{A, E\}.$$

Definition 4. *A* $CLP(\{\mathcal{D}\})$ rule *is a formula* $H \leftarrow c_{\mathcal{S}}, c_{\mathcal{D}} | B_1, \dots, B_n$ *where*

- $c_{\mathcal{D}}$ *is a* $\mathcal{D}$*-admissible constraint,*
- $c_{\mathcal{S}}$ *is a* $\mathcal{S}$*-admissible constraint, and*
- H *is a* $\mathcal{P}$*-atom and* $B_1, \dots, B_n$ *are* $\mathcal{P}$*-literals (i.e., positive or negated* $\mathcal{P}$*-atoms).*

A $CLP(\{\mathcal{D}\})$ *program* P *is a finite collection of* $CLP(\{\mathcal{D}\})$ *rules.*

Example 2. Assume again that $\mathcal{D} = \mathbb{FD}$. The following is a $CLP(\{\mathcal{D}\})$ program:

$$p(A, X, B) \leftarrow A < X, X < B.$$
$$q(K) \leftarrow A \text{ in } 1..2, B \text{ in } 2..3, E = \{X : p(A, X, B)\}, K = \{A, E\} \,|\, p(A, Z, B).$$

Let us observe that even if we accept any propositional combination, in our examples we will deal with conjunctions of $\{\mathcal{D}\}$-atomic constraints or negated atomic $\{\mathcal{D}\}$-constraints. Disjunctions can be removed by introducing new rules.

4 Semantics

In this section we provide the semantics for the language $CLP(\{\mathcal{D}\})$. In particular, we propose an interpretation of the aggregate constraint atoms based on a variation of stable model semantics [10]. A three-valued well-founded semantics can be also derived from the ideas provided in [21].

Let $\mathcal{D}$ be the initial domain constraint, D its interpretation domain, and $I_{\mathcal{D}}$ its interpretation function. We define the domain $\mathcal{U}_D = \bigcup_{i \geq 0} \mathcal{U}_i$ where:

$$\begin{cases} \mathcal{U}_0 = D \\ \mathcal{U}_{i+1} = \mathcal{U}_i \cup \wp(\mathcal{U}_i) \cup \\ \qquad\qquad \{c_{f(a_1,\dots,a_n)} \,:\, a_1 \in \mathcal{U}_i, \dots, a_n \in \mathcal{U}_i, f \in \mathcal{F_P}\} \end{cases}$$

where for all f and $a_1, \dots, a_n$, $c_{f(a_1,\dots,a_n)}$ is a new object in the domain, different from all other objects. Observe that the Herbrand universe for $\mathcal{F_P}$ and D are both contained in $\mathcal{U}_D$.

We will use the following partial order on $\mathcal{U}_D$: given two elements $a, b \in \mathcal{U}_D$: $a \le b$ if and only if (1) $a, b \in D$ and $a = b$, or (2) a and b are sets and $a \subseteq b$, or (3) $a = c_{f(a_1,\dots,a_n)}$ and $b = c_{g(b_1,\dots,b_n)}$ and $f \equiv g, a_1 \le b_1, \dots, a_n \le b_n$.

Stable model semantics for logic programs is defined for *ground* programs. In a CLP context, the notion of grounding has to be properly modified to accommodate for the properties of the selected interpretation domain.

Definition 5. *(Pre-interpretation) Let t be a term and $\sigma : vars(t) \mapsto \mathcal{U}_D$ be a valuation function that maps variables of sort* D *to elements in D, variables of sort* S *to elements in $\mathcal{U}_D \setminus \mathcal{U}_0$, and variables of sort* P *to elements in $\mathcal{U}_D$. Furthermore, let R be the map that associates to each element $f \in \mathcal{F}_P$ of arity k a k-ary function from ${\mathcal{U}_D}^k$ to $\mathcal{U}_D$. R is called the* base *of the pre-interpretation. The pre-interpretation $t^{R,\sigma}$ of a term t w.r.t. R, σ is defined as follows:*

- *If t is a variable, then $t^{R,\sigma}$ is $\sigma(t)$.*
- *If t is a constant of sort* D, *then $t^{R,\sigma} = t^{\mathcal{D}}$ (i.e., the standard interpretation of the constant in the constraint domain $\mathcal{D}$).*
- *If t is $f(t_1, \dots, t_k)$ and $f \in \mathcal{F}_D$, then $t^{R,\sigma} = f^{\mathcal{D}}(t_1^{R,\sigma}, \dots, t_k^{R,\sigma})$, where $f^{\mathcal{D}}$ is the standard interpretation of f in $\mathcal{D}$.*
- *If t is $f(t_1, \dots, t_k)$ and $f \in \mathcal{F_S}$, then $t^{R,\sigma} = f^{\mathcal{S}}(t_1^{R,\sigma}, \dots, t_k^{R,\sigma})$, where $f^{\mathcal{S}}$ is the standard set-theoretic interpretation of f. In particular,*
  - *If t is $\emptyset$, then $t^{R,\sigma}$ is the empty set.*
  - *If t is $\{a \,|\, b\}$, then $t^{R,\sigma}$ is the set $\{a^{R,\sigma}\} \cup b^{R,\sigma}$.*
- *If t is a constant in $\mathcal{F_P}$ then $t^{R,\sigma}$ is simply t^R.*
- *If t is $f(t_1, \dots, t_k)$ and $f \in \mathcal{F_P}$ then $t^{R,\sigma} = f^R(t_1^{R,\sigma}, \dots, t_k^{R,\sigma})$.*

In the rest of this discussion we will assume that the base R is fixed and in particular, that it interprets $f \in \mathcal{F_P}$ using the $c_{f(\cdot,\dots,\cdot)}$ introduced in $\mathcal{U}_D$.

Definition 6. *(Grounding) Given an atom A and a pre-interpretation R, σ for the terms in A (where σ is defined for all variables in A), we define the notion of* grounding *of A w.r.t. R, σ as follows:*

- *If $A = p(t_1, \dots, t_n)$ is a $\mathcal{D}$-constraint atom, the grounding of $p(t_1, \dots, t_n)$ w.r.t. R, σ is* `true` *if $\mathcal{D} \models p(t_1^{R,\sigma}, \dots, t_n^{R,\sigma})$,* `false` *otherwise.*
- *If $A = p(t_1, \dots, t_n)$ is a $\mathcal{SET}$-constraint atom, then the grounding of the atom w.r.t. R, σ is the atom* `true` *if $p^{\mathcal{U}_D}(t_1^{R,\sigma}, \dots, t_n^{R,\sigma})$ is true (where $p^{\mathcal{U}_D}$ is the traditional set-theoretic interpretation of the predicate p on the domain $\mathcal{U}_D$), the atom* `false` *otherwise. In particular, given R and σ:*
  - *$\cup_3(s_1, s_2, s_3)$ is pre-interpreted as $s_3^{R,\sigma} = s_1^{R,\sigma} \cup s_2^{R,\sigma}$*

- *$s \in t$ is pre-interpreted as $s^{R,\sigma} \in t^{R,\sigma}$*
- *$s \subseteq t$ is pre-interpreted as $s^{R,\sigma} \subseteq t^{R,\sigma}$*
- *$s||t$ is pre-interpreted as $s^{R,\sigma} \cap t^{R,\sigma} = \emptyset$*
- *$\cap_3(s_1, s_2, s_3)$ is pre-interpreted as $s_3^{R,\sigma} = s_1^{R,\sigma} \cap s_2^{R,\sigma}$*

– *If $A = p(t_1, \ldots, t_n)$ is a $\mathcal{P}$-atom, then its grounding w.r.t. R, σ is the object $p(t_1^{R,\sigma}, \ldots, t_n^{R,\sigma})$.*
– *If A is an atom of the form $E = \{X : B\}$ and $B[X/X']$ is B with X renamed to X' (X' a new variable), then its grounding w.r.t. R, σ is the equality $E^{R,\sigma} = \{X' : B[X/X']^{R,\sigma}\}$.*

Given an atom A and a pre-interpretation base R, a grounding of A w.r.t. R is a grounding of A w.r.t. R, σ for an arbitrary σ such that R, σ is a pre-interpretation for A. The notion of grounding can be extended to rules and to programs.

Let $I_D = \langle D, (\cdot)^D \rangle$ be the standard interpretation for the constraint domain $\mathcal{D}$.

Definition 7. *An interpretation I is a pair $\langle \mathcal{U}_D, (\cdot)^I \rangle$, where the interpretation function $(\cdot)^I$ is defined as follows:*

– *I coincides with $I_\mathcal{D}$ on the interpretation of atoms built using $\mathcal{F}_\mathcal{D}$ and $\Pi_\mathcal{D}$.*
– *$\in, \subseteq, \cup_3$, and the other symbols in $\Pi_\mathcal{S}$ are interpreted in $\mathcal{U}_D$ according to their standard set-theoretical meaning.*
– *$=$ is interpreted as the identity over $\mathcal{U}_D$.*
– *$(\cdot)^I$ interprets each predicate symbol in $\Pi_\mathcal{P}$ as a predicate over $\mathcal{U}_D$.*
– *for each grounding R, σ of $E = \{X : B\}$, $(\cdot)^I$ interprets $(E = \{X : B\})^{R,\sigma}$ to `true` if $E^{R,\sigma}$ is equal to the set $\{X' : (B[X/X']^{R,\sigma})^I\}$ (where X' is a new variable), to `false` otherwise.*

If the variable X does not occur in B, then the semantics of $\{X : B\}$ is the empty set if there is no σ s.t. $(B^{R,\sigma})^I$ is true; it is the universe $\mathcal{U}_D$ otherwise.

Example 3. Let A in $1..2$ and B in $2..3$ be two $\mathbb{FD}$ variables, R be an arbitrary base and $\sigma = \{A/2, B/3, E/\{2\}\}$ be a valuation function. The pre-interpretation $t^{R,\sigma}$ of the term $t \equiv \{X : p(A, X, B)\}$ is $\{2\} = \{X' : p(2, X', 3)\}$.

Let I and J be two interpretations on $\mathcal{U}_D$. We say that $I \preceq J$ if for each atom $P \equiv p(t_1, \ldots, t_n)$ and for each grounding R, σ of P there exists a an atom $Q \equiv p(s_1, \ldots, s_n)$ and a grounding R, θ of Q such that $(P^{R,\sigma})^I \to (Q^{R,\theta})^J$ and $(t_i^{R,\sigma})^I \leq (s_i^{R,\theta})^J$ for $i = 1, \ldots, n$. Given an atom P and a grounding R, σ of P, we define an interpretation I to be a *model* of $P^{R,\sigma}$ if $(P^{R,\sigma})^I$ is true. We denote this fact with $I \models P^{R,\sigma}$. Similarly, we can extend the definition of $\models$ to conjunctions of atoms. We define an interpretation I to be a *model* of a grounded rule *head* $\leftarrow c_\mathcal{S}, c_\mathcal{D} \mid$ *body* if $I \models c_\mathcal{S} \wedge c_\mathcal{D} \wedge$ *body* implies $I \models$ *head*. I is a model of a rule if it is a model of each grounding of the rule.

Example 4. A model for the $CLP(\{\mathbb{FD}\})$ program of the Example 2 is

$$I = \{p(a, b, c) : a, b, c \in \mathsf{inf..sup}, a < b < c\} \cup \{q(\{1, \{2\}\})\}.$$

In this case D is the set of integer numbers between inf and sup.

4.1 Stable Model Semantics

Let P be a $CLP(\{\mathcal{D}\})$ program and let I be an interpretation (built on the pre-interpretation base R). Let P' contain all possible R, σ groundings of the rules in P. Following the principle used in [10], we define the intensional stable model transformation $G(P, I)$. The transformation is achieved in two steps, i.e., $G(P, I) = G_{set}(G_{neg}(P, I), I)$. The first transformation, G_{neg} is defined as follows: for each rule in P'

$$head \leftarrow c_{\mathcal{S}}, c_{\mathcal{D}} \mid A_1, \ldots, A_n, \neg B_1, \ldots, \neg B_m$$

if for all B_i, $1 \leq i \leq m$ we have that $I \models \neg B_i$, then the rule grounding

$$head \leftarrow c_{\mathcal{S}}, c_{\mathcal{D}} \mid A_1, \ldots, A_n$$

is added to $G_{neg}(P, I)$ (otherwise the rule is erased). Observe that $G_{neg}(P, I)$ is the grounding of a program without negation.

The transformation G_{set} is defined as follows; for each rule $head \leftarrow c_{\mathcal{S}}, c_{\mathcal{D}} \mid A_1, \ldots, A_n$ in $G_{neg}(P, I)$ (i.e., a grounding of a program rule without negation), if for all atoms A in $c_{\mathcal{S}}$ it holds that $I \models A$ then the rule

$$head \leftarrow c_{\mathcal{D}} \mid A_1, \ldots, A_n$$

is added to $G_{set}(G_{neg}(P, I), I)$ (otherwise the rule is erased). $G_{set}(G_{neg}(P, I), I)$ is the grounding of a program with neither negation nor set atoms.

Definition 8. *I is a* stable model *of P if I is the least $\mathcal{U}_D$-model of the program $G_{set}(G_{neg}(P, I), I)$.*

Example 5. Consider the program: $r(1).p(X) \leftarrow X = \{Y : r(Y)\}$. and let us consider $G_{set}(P, I)$ for $I = \{r(1), p(\{1\})\}$. With $X = \emptyset$ the constraint aggregate becomes: $\emptyset = \{1\}$ which is an atom false in its interpretation in $\mathcal{U}_D$. This grounding is removed. The only true grounded clause is: $p(\{1\}) \leftarrow \{1\} = \{Y : r(Y)\}$. Thus, $G_{set}(G_{neg}(P, I), I) = \{r(1), p(\{1\})\}$. I is a stable model of P.

Proposition 1. *Let P be a $CLP(\{\mathcal{D}\})$ program.*

1. *if I be a stable model of P, then I is a model of P.*
2. *if I is the unique stable model of P, then I is the $\preceq$-minimal model of P.*

We cannot claim minimality in general. Consider:

$$\begin{array}{lll} r(1). & q(1). & \\ r(2). & q(2) \leftarrow Z = \{X : r(X)\}, p(Z). & p(Y) \leftarrow Y = \{X : q(X)\}. \end{array}$$

This program has two stable models: $I_1 = \{q(1), q(2), p(\{1,2\}), r(1), r(2)\}$ and $I_2 = \{q(1), p(1), r(1), r(2)\}$. Observe that $I_1 \preceq I_2$. The fact that non-minimal models can be stable models is a common problem in the use of stable model semantics for handling of aggregates (without restrictions on aggregations) [14]. Similar problems are present for the limited set aggregations described in [17].

Example 6. Consider the program

$$\begin{array}{lrcl} r(1). & p(X) & \leftarrow & q(X), s(X). \\ s(\{1\}). & q(X) & \leftarrow & X = \{Y : r(Y)\}. \end{array}$$

This program has the unique stable model: $\{r(1), s(\{1\}), q(\{1\}), p(\{1\})\}$.

Example 7. Simple recursive constructions may lead to infinite intensional sets. For example, consider the program

$$p(1). \qquad p(X) \leftarrow X = \{Y\}, p(Y).$$
$$q(X) \leftarrow X = \{Y : p(Y)\}.$$

The stable model contains: $\{p(1), p(\{1\}), p(\{\{1\}\}), \ldots\}$. Additionally, it should contain $q(\{1, \{1\}, \{\{1\}\}, \ldots\})$.

Example 8. Intensional definitions may also lead to situations that cannot be justified according to our treatment of intensional sets. E.g., the program

$$p(a). \qquad p(S) \leftarrow S = \{X : p(X)\}.$$

does not admit any stable model.

5 Constraint solver

A $\{\mathcal{D}\}$-constraint (Def. 3) is a propositional combination of $\mathcal{D}$ and $\mathcal{S}$ constraints. In the language $CLP(\{\mathcal{D}\})$ we recognize different classes of constraints:

- *$\mathcal{D}$-constraints.* We assume that this class is decidable, and we denote with $\mathcal{SAT}_{\mathcal{D}}$ the procedure used to solve this class of constraints.
- *Extensional $\mathcal{S}$-constraints.* these are $\mathcal{S}$-constraints that do not involve any occurrence of intensional sets. This class of constraints is decidable and it corresponds to the class of constraints supported by the language $CLP(\mathcal{SET})$. In [6] an effective procedure to solve constraints in this class is proposed—we denote with $SAT_{\mathcal{SET}}$ such procedure. Traditional equality and disequality constraints between terms of the sort P are also treated in this procedure.
- *Intensional $\mathcal{S}$-constraints.* These are $\mathcal{S}$-constraints that contain occurrences of intensional sets (aggregate constraints). A procedure to handle this type of constraints is described in Sect. 5.1; we refer to this procedure as $SAT_{\mathcal{S}}$.

In order to accomplish the goal of resolving constraints in $CLP(\{\mathcal{D}\})$, we develop a constraint solver, called $\mathsf{Solve}_{\{\mathcal{D}\}}$. This solver repeatedly simplifies the constraint until no further simplifications are possible. The overall structure is shown in Fig. 1. Since constraint solvers can be non-deterministic, with $c' = SAT_{\mathcal{X}}(c)$ we mean *one* of the possible non-deterministic solutions returned. Thus, $\mathsf{Solve}_{\{\mathcal{D}\}}(C)$ is a non-deterministic procedure. In addition, for a program P we also introduce another function, called $\mathsf{Solvegoal}_P$. The predicate $\mathsf{Solvegoal}_P$

$\mathsf{Solve}_{\{\mathcal{D}\}}(C)$:
repeat
  select c **in** C;
  if c *is a $\mathcal{D}$-constraint* **then** $c' = SAT_{\mathcal{D}}(c)$;
  if c *is an Extensional $\mathcal{S}$-constraint* **then** $c' = SAT_{\mathcal{SET}}(c)$;
  if c *is an Intensional $\mathcal{S}$-constraint* **then** $c' = SAT_{\mathcal{S}}(c)$;
  replace c **by** c' **in** C;
until *no rewriting is possible;*

Fig. 1. Overall structure of $\mathsf{Solve}_{\{\mathcal{D}\}}$

applied to a $\mathcal{P}$-atom or formula means that the solving of its argument is delayed at the end of the constraint solving and it will be done by the general (constraint) resolution procedure. Let us observe that the non-deterministic result c of $\mathsf{Solve}_{\{\mathcal{D}\}}(C)$ is the conjunction of a constraint with possibly some atoms $\mathsf{Solvegoal}_P(G)$. At this time, the only requirement we impose is that: $\mathsf{Solvegoal}_P(G) \Leftrightarrow G$. The intuition is that the $\mathsf{Solvegoal}_P$ will encode the language mechanisms required to support the explicit removal of intensional sets—e.g., through the use of negation (as in Sect. 2).

In this section we develop constraint solvers for intensional constraints, showing the generality of some rewriting rules (propagation) and the difficulties introduced by other rewriting rules (labeling). We also discuss the cooperation between the three solvers. We allow intensional terms to occur freely as terms in programs—it is simple to transform a program into the flat form of Def. 4.

5.1 Propagation Procedures

In this subsection we present some rewriting rules for $\mathcal{S}$-constraints of $CLP(\{\mathcal{D}\})$ that can be easily implemented starting from any initial domain $\mathcal{D}$; these rules allow us to deal finitely with intensional sets (without any restriction on the finiteness of the intensional sets). The application of these rewriting rules to a $\mathcal{S}$-constraint C leads to a disjunction of constraints that is equisatisfiable to C. However, the rewriting rules are incomplete, since constraints obtained might be unsatisfiable, but unsatisfiability could be not detected right away. Thus, the behavior is not dissimilar from that of incomplete constraint solvers used in other CLP systems—e.g., arc and bound consistency employed in $\mathbb{FD}$-solvers (e.g., in SICStus Prolog) are insufficient to detect unsatisfiability of a constraint such as

$$X \mathsf{~in~} 1..2, Y \mathsf{~in~} 1..2, Z \mathsf{~in~} 1..2, X \neq Y, X \neq Z, Y \neq Z.$$

A complete constraint solver can be called into play by the user with a procedure analogous to the $\mathsf{labeling}$ used in $CLP(\mathbb{FD})$. We discuss this enhanced capability in the following subsection 5.2.

Some propagation rewriting rules are presented in Fig. 2. These rules are meant to complement the rewriting rules that have been proposed in our previous works to handle constraints over *extensionally defined sets*—due to lack of space we omit reproducing these rewriting rules [6, 3]. We give only the rewriting rules for some of the predicates involved, in particular for those assumed primitive in [6] and $\subseteq$. No rules are needed for $\not\emptyset_3$ since the unique rule in [6] applies to intensional sets as well.

All the propagation rules presented in Fig. 2 do not actually compute the intensional sets—i.e., they do not force the explicit enumeration of the elements of the intensional set; thus, they can work without any assumption on the finiteness of these sets. Moreover, they do not introduce negation, but negated constraint literals that are treated as constraints—this can be seen, for example, in case =-2 in Fig. 2. In addition, some of the propagation rules are non-deterministic—see, for example, the case $\neq$-3. The presence of non-determinism leads to a family of formulas (consisting of constraints and $\mathsf{Solvegoal}_P$ atoms) that are returned at the end of the processing, whose disjunction is equisatisfiable to the initial constraint C. In the rewriting rules we also omit the explicit description of the steps used to verify violation of the sorts of the predicates (with the exception, for the sake of clarity, of the =-1 and ≠-1 cases).

We use some syntactic sugars in the rewriting rules. We make use of the notation $\{X : \varphi_1 \vee \varphi_2\}$ to represent $\{X : p(\bar{Y})\}$, where $\{\bar{Y}\} = FV(\varphi_1) \cup FV(\varphi_2)$ and p is defined as: $p(\bar{Y}) \leftarrow \varphi_1$. $p(\bar{Y}) \leftarrow \varphi_2$. Another syntactic sugar is $\mathsf{less}(X, Y, Z)$, defined as

$$\forall X, Y, Z(\mathsf{less}(X, Y, Z) \leftrightarrow Y = \{X \mid Z\} \wedge X \notin Z).$$

namely, Z is the set Y without the element X. Further rewriting rules for propagation can be easily defined; for instance,

$$\cap_3(\{X : \varphi_1\}, \{X : \varphi_2\}, s) \mapsto s = \{X : \varphi_1 \wedge \varphi_2\}$$

Other rewriting rules can be considered if we accept the use of negation. Negation is allowed in $CLP(\{\mathcal{D}\})$ programs. However, it is clear that it might introduce new problems. In particular, it introduces requirements on the capabilities of the $\mathcal{D}$ constraint solver to handle more difficult constraints. Observe that the rules in Fig. 3 produce a negated version of the property employed to construct the intensional set. It is important to observe that the interaction between the constraint solvers may actually facilitate the handling of negated constraints—e.g., by grounding its argument and thus making it easier to solve. Although the rules for $\notin$- and $||$-constraints make use of negation, the intensional sets (that could be infinite) do not need to be explicitly generated (using the process described in equation (2) of Sect. 2).

Correctness and completeness of the rewriting rules are immediate consequences of the semantics of the set-based operators involved. These results can be formally proved in the theory *Set* [6], a minimal set theory that deals with $\emptyset, \{\cdot \mid \cdot\}, =, \in, \cup_3, ||$ (and also with $\cap_3$ and $\subseteq$, that can be easily defined in terms of the previous ones). The theory has to be extended by adding the well-known *comprehension scheme* of the ZF set theory

$$\forall s \forall y\, (y \in \{X \in s : \varphi[X]\} \leftrightarrow (y \in s \wedge \varphi[X/y])) \text{ for any f.o.f. } \varphi\,.$$

Condition '$X \in s$'—introduced by Zermelo in 1908—is used to overcome Russell's famous paradox (pick φ as $X \notin X$). In the syntax for intensional sets this condition is not required. We assume that $X \in \mathcal{U}_D$.

With $c \in \mathsf{Solve}_{\{\mathcal{D}\}}(C)$ we mean one of the possible non-deterministic solutions returned by $\mathsf{Solve}_{\{\mathcal{D}\}}(C)$. Each of them is a conjunction of $\{\mathcal{D}\}$-constraints and $\mathsf{Solvegoal}_P$ atoms. In the second result, the introduction of negation requires to call into play the completion $\mathsf{comp}(P)$ of a program P.

Proposition 2. *(Non-negative Simplification) Let $\mathcal{C}_\mathcal{D}$ be a decidable class of $\mathcal{D}$-constraints; let $\mathsf{Solve}_{\{\mathcal{D}\}}$ be defined as in Fig. 1, with $SAT_\mathcal{S}$ composed of the rules in Fig. 2. Let C be a $CLP(\{\mathcal{D}\})$ constraint and P be a $CLP(\{\mathcal{D}\})$ program. Then $Set, \mathcal{D}, P \models \left(C \Leftrightarrow \bigvee_{c \in \mathsf{Solve}_{\{\mathcal{D}\}}(C)} c\right)$.*

Proposition 3. *(Negative Simplification) Let $\mathcal{C}_\mathcal{D}$ be a decidable class of $\mathcal{D}$-constraints; let $\mathsf{Solve}_{\{\mathcal{D}\}}$ be defined as in Fig. 1, with $SAT_\mathcal{S}$ composed of the rules in Fig. 2–3. Let C be a $CLP(\{\mathcal{D}\})$ constraint and P be a $CLP(\{\mathcal{D}\})$ program. Then $Set, \mathcal{D}, \mathsf{comp}(P) \models \left(C \Leftrightarrow \bigvee_{c \in \mathsf{Solve}_{\{\mathcal{D}\}}(C)} c\right)$.*

=-constraints	
=-1.	$\left.\begin{array}{r}\{X : \varphi\} = t \\ \mathsf{sort}(t) \neq \mathsf{S}\end{array}\right\} \mapsto \texttt{false}$
=-2.	$\{X : \varphi\} = \{t \mid s\} \mapsto \begin{array}{l}(i)\ t \in \{X : \varphi\} \wedge \{X : (\varphi \wedge X \neq t)\} = s \\ (ii)\ t \in \{X : \varphi\} \wedge \{X : \varphi\} = s\end{array}$

≠-constraints	
≠-1.	$\left.\begin{array}{r}\{X : \varphi\} \neq t \\ \mathtt{sort}(t) \neq \mathsf{S}\end{array}\right\} \mapsto \texttt{true}$
≠-2.	$\{X : \varphi\} \neq \emptyset \mapsto \mathsf{Solvegoal}_P(\exists X \varphi)$
≠-3.	$\{X : \varphi_1\} \neq \{X : \varphi_2\} \mapsto \begin{array}{l}(i)\ \ Z \in \{X : \varphi_1\} \wedge Z \notin \{X : \varphi_2\} \\ (ii)\ Z \in \{X : \varphi_2\} \wedge Z \notin \{X : \varphi_1\}\end{array}$
≠-4.	$\{X : \varphi\} \neq \{s \mid t\} \mapsto \begin{array}{l}(i)\ \ Z \in \{X : \varphi\} \wedge Z \notin \{s \mid t\} \\ (ii)\ Z \in \{s \mid t\} \wedge Z \notin \{X : \varphi\}\end{array}$

∈-constraints	
∈-1.	$t \in \{X : \varphi\} \mapsto \mathsf{Solvegoal}_P(\varphi[X/t])$

$\cup_3$-constraints	
$\cup_3$-1.	$\cup_3(\{X : \varphi\}, \emptyset, s) \mapsto s = \{X : \varphi\}$
$\cup_3$-2.	$\cup_3(\{X : \varphi\}, s, \emptyset) \mapsto \emptyset = \{X : \varphi\} \wedge \emptyset = s$
$\cup_3$-3.	$\cup_3(\{s \mid t\}, r, \{X : \varphi\}) \mapsto \begin{array}{l}s \in \{X : \varphi\} \wedge \mathsf{less}(s, \{s \mid t\}, N) \wedge \\ \cup_3(N, r, \{X : \varphi \wedge X \neq s\})\end{array}$
$\cup_3$-4.	$\cup_3(\{X : \varphi_1\}, \{X : \varphi_2\}, s) \mapsto s = \{X : \varphi_1 \vee \varphi_2\}$
$\cup_3$-5.	$\cup_3(\{X : \varphi_1\}, r, \{X : \varphi_2\}) \mapsto \begin{array}{r}r \subseteq \{X : \varphi_2\} \wedge \{X : \varphi_1\} \subseteq \{X : \varphi_2\} \wedge \\ \{X : \varphi_2\} \subseteq \{X : \varphi_1 \vee X \in r\}\end{array}$
$\cup_3$-6.	$\cup_3(\{X : \varphi\}, r, \{s \mid t\}) \mapsto$ $\mathsf{less}(s, \{s \mid t\}, N) \wedge$ $(i)\ \ s \in \{X : \varphi\} \wedge s \notin r \wedge \cup_3(\{X : \varphi \wedge X \neq s\}, r, N)$ $(ii)\ \ s \in \{X : \varphi\} \wedge \mathsf{less}(s, r, N_1) \wedge \cup_3(\{X : \varphi \wedge X \neq s\}, N_1, N)$ $(iii)\ \ s \notin \{X : \varphi\} \wedge \mathsf{less}(s, r, N_1) \wedge \cup_3(\{X : \varphi\}, N_1, N)$

$\parallel$-constraints	
$\parallel$-1.	$\emptyset \parallel \{X : \varphi\} \mapsto \texttt{true}$

$\nparallel$-constraints	
$\nparallel$-1.	$\emptyset \nparallel \{X : \varphi\} \mapsto \texttt{false}$
$\nparallel$-2.	$\{X : \varphi\} \nparallel \{s \mid t\} \mapsto \begin{array}{l}(i)\ \ \mathsf{Solvegoal}_P(\varphi[X/s]) \\ (ii)\ \mathsf{less}(s, \{s \mid t\}, N)\ \wedge \{X : \varphi\} \nparallel N\end{array}$
$\nparallel$-3.	$\{X : \varphi\} \nparallel \{X : \psi\} \mapsto \mathsf{Solvegoal}_P(\exists X(\varphi \wedge \psi))$

⊆-constraints	
⊆-1.	$\emptyset \subseteq \{X : \varphi\} \mapsto \texttt{true}$
⊆-2.	$\{X : \varphi\} \subseteq \emptyset \mapsto \emptyset = \{X : \varphi\}$
⊆-3.	$\{s \mid t\} \subseteq \{X : \varphi\} \mapsto s \in \{X : \varphi\} \wedge t \subseteq \{X : \varphi\}$
⊆-4.	$\{X : \varphi\} \subseteq \{s \mid t\} \mapsto \begin{array}{l}(i)\ \ s \in \{X : \varphi\} \wedge \mathsf{less}(s, \{s \mid t\}, N) \wedge \\ \qquad\qquad \{X : \varphi \wedge X \neq s\} \subseteq N \\ (ii)\ s \notin \{X : \varphi\} \wedge \mathsf{less}(s, \{s \mid t\}, N) \wedge \\ \qquad\qquad \{X : \varphi\} \subseteq N\end{array}$

Fig. 2. Propagation Rewriting Rules

$\notin$-constraints	
$\notin$-1.	$t \notin \{X : \varphi\} \mapsto \mathsf{Solvegoal}_P(\neg\varphi[X/t])$

$\|\|$-constraints	
$\|\|$-2.	$\{X : \varphi\} \,\|\|\, \{X : \psi\} \mapsto \mathsf{Solvegoal}_P\,(\exists X(\varphi \wedge \neg\psi) \vee \exists X(\neg\varphi \wedge \psi))\,) \vee$ $\{X : \varphi\} = \emptyset \wedge \{X : \psi\} = \emptyset$
$\|\|$-3.	$\{X : \varphi\} \,\|\| \{s \mid t\} \mapsto \mathsf{Solvegoal}_P(\neg\varphi[X/s]) \wedge \mathsf{less}(s, \{s \mid t\}, N) \wedge \{X : \varphi\} \not\|\, N$

Fig. 3. Propagation with negation

5.2 Labeling

As mentioned in the previous subsection, the propagation rules allow us, given an original constraint C, to determine a disjunction of constraints that is equi-satisfiable to C. Each constraint belonging to the disjunction is "simpler" than C—e.g., it may contain fewer occurrences of intensional constraints. On the other hand, these rewriting rules do not constitute a complete solver—in particular, there is no guarantee that unsatisfiable constraints are reduced to `false` and tautologies are reduced to `true`. This situation makes the rewriting procedure weaker than, e.g., the procedures used in $CLP(\mathcal{SET})$, where each set constraint can be always reduced to `false` or to a satisfiable constraint in solved form.

In order to approximate a similar behavior in the context of $CLP(\{\mathcal{D}\})$ it is necessary to force the removal of intensional sets from the constraint. This process can be seen as a sort of *labeling* of the variables in the constraint. More in detail, intensional sets must be expanded (and, possibly, $\mathcal{D}$-constraints may have to be subjected to a similar transformation). We show in Fig. 4 how the labeling can be accomplished in the various cases left. See also the formula (2) for computing a single intensional sets. Observe that the right-hand side of the transformation can be encoded either through the use of negation or by introducing an explicit construct to collect solutions to a goal (e.g., `findall`). The additional rewriting rules satisfy the following properties:

- the new rules cannot guarantee completeness, as the problem we are trying to solve is undecidable in general (e.g., intensional sets build on properties with an infinite number of solutions);
- the rules rely on the ability to handle negated computations.

$=$-constraints	
$=$-3. $\{X : \varphi\} = \emptyset$	$\mapsto \mathsf{Solvegoal}_P(\neg(\exists X\varphi))$
$=$-4. $\{X : \varphi_1\} = \{X : \varphi_2\}$	$\mapsto \mathsf{Solvegoal}_P(\forall X(\varphi_1 \leftrightarrow \varphi_2))$

$\subseteq$-constraints	
$\subseteq$-5. $\{X : \varphi_1\} \subseteq \{X : \varphi_2\}$	$\mapsto \mathsf{Solvegoal}_P(\forall X(\varphi_1 \rightarrow \varphi_2))$

Fig. 4. Labeling using negation

5.3 Further Considerations

Negation handling in this context deserves special attention. Negation as failure in the context of a language with sets has been studied only for programs that are stratified and meet restrictive allowedness requirements to avoid floundering [2]. Constructive negation in the context of Constraint Logic Programming with Sets has been studied in [8]. However, the class of programs that can be dealt with successfully does not enlarge significantly the class of those that can be dealt with negation as failure and stratification.

We are currently investigating how these mechanisms can be employed in the context of *Answer Set Programming (ASP)* [16]. For instance, the negative reductions required in $\notin$-1 and $||$-2 of Fig. 3 can be encoded as ASP rules:

$\notin$-1.	$\leftarrow \varphi[t].$	
$\|\|$-2.	$p_1 \mid p_2 \mid p_3 \leftarrow$	
	$q_1 \leftarrow dom_\varphi(X), dom_\psi(X), \varphi, not\ \psi.$	
	$q_2 \leftarrow dom_\varphi(X), dom_\psi(X), \psi, not\ \varphi.$	
	$\leftarrow p_1, not\ q_1.$	$\leftarrow p_2, not\ q_2.$
	$\leftarrow p_3, q_1.$	$\leftarrow p_3, q_2.$

where p_1, p_2, p_3, q_1, q_2 are brand new atoms. The predicates dom_φ and dom_ψ are domain predicates as required by `smodels` [17]. Similar constructions can be employed to handle rules $=$-3, $=$-4, and $\subseteq$-5:

$=$-3.	$p \leftarrow dom_\varphi(X), \varphi.$
	$\leftarrow p.$
$=$-4.	$\leftarrow \varphi_1, not\ \varphi_2.$
	$\leftarrow \varphi_2, not\ \varphi_1.$
$\subseteq$-5.	$\leftarrow \varphi_1, not\ \varphi_2.$

6 Conclusions

In this paper we presented preliminary ideas on how to extend any $CLP(\mathcal{D})$ language with set-based primitives and constraints. The novelty of the framework is not only the presence of intensional sets but the ability to develop (extensional and intensional) sets on top of arbitrary constraint domains $\mathcal{D}$. We developed a syntactic and semantics specification of the new language (called $CLP(\{\mathcal{D}\})$). We also developed rewriting algorithms to simplify constraints containing intensional sets—possibly relying on the use of negation.

In the immediate future we plan to effectively implement the technique at least for some largely used constraint domains, such as finite domain constraint. A preliminary result in this direction is [3] where the $CLP(\mathcal{SET})$ constraint solver is integrated with the $CLP(\mathbb{FD})$ constraint solver of SICStus Prolog. In this preliminary work intensional set constraints are allowed but currently solved via explicit enumeration.

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