# A Specialization Technique for Deriving Deterministic Constraint Logic Programs and Its Application to Pattern Matching \*

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**Abstract.** When specializing programs one can increase efficiency by reducing nondeterminism. We consider constraint logic programs and we propose a technique which, by making use of a new transformation rule called clause splitting, allows us to generate efficient, specialized programs which are deterministic. We have applied our technique to the specialization of pattern matching programs.

## 1 Introduction

Programs are often written in a parametric form so that one can reuse them in different contexts. When one reuses parametric programs, one may want to transform those programs for taking advantage of the contexts of use and, indeed, by doing so, often program efficiency is improved. This program transformation is usually called *program specialization* [18] and it can be performed by using well established techniques such as *partial evaluation* [4, 12, 18, 20, 22].

Various program specialization methods have been proposed in the literature for different programming languages. In this paper we consider a program specialization method for constraint logic programming (CLP) and we use the rules + strategies transformation approach. This approach was first suggested by Burstall-Darlington for functional languages [3] and later applied to logic languages by Tamaki-Sato [25]. Our method increases program efficiency by deriving deterministic, specialized programs starting from nondeterministic, general programs.

Our specialization method makes use of a set of rules for transforming constraint logic programs which are an extension of the ones presented in [6, 8, 23]. This set includes extensions of the familiar unfolding and folding rules, and an extra rule, called *clause splitting*, which generalizes the *case splitting* rule presented in [23]. Given a clause  $H \leftarrow Body$  and a constraint c, by the clause

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splitting rule we generate the two clauses:  $H \leftarrow c \land Body$  and  $H \leftarrow \neg c \land Body$ . By using these clauses which have mutual exclusive bodies, in what follows we show how we can derive efficient programs with reduced nondeterminism. The correctness of the derived programs follows from the fact that the transformation rules preserve the least model semantics [17].

Our specialization method is realized by an automatic strategy which guides the application of the transformation rules. This strategy is an enhancement of the one presented in [23] and includes a specific treatment of constraints. Our enhanced strategy consists of the following steps: (i) the introduction of an initial definition, corresponding to the goal w.r.t. which we want to specialize the initial program, (ii) the execution of some unfolding steps and constraint manipulations, and (iii) the execution of some folding steps. If in order to perform these folding steps we have to introduce new definitions, we do so, and we continue the specialization process by executing unfolding, constraint manipulations, and folding steps starting from each of these new definitions. On the contrary, if the folding steps do not require the introduction of new definitions, we terminate the specialization process. We will see this strategy in action in Section 5.

This paper is an extended version of [9]. In particular, in Section 6 we will describe some experimental results obtained by implementing our specialization method on the MAP transformation system [7].

## 2 An Introductory Example: Specialization of a Constrained Matching Program

In this Section we present an example of program specialization using the rules + strategies approach. Starting from a nondeterministic, general program which specifies a pattern matcher on strings, we derive a deterministic, specialized pattern matcher for a given pattern. In this example we define a general matching relation between strings which is expressed as a constraint logic program. Our derivation generalizes the derivations of the Knuth-Morris-Pratt matcher [19] which were presented, among others, in [10-12, 15, 23, 24]. As in the case of that matcher, we derive a program which behaves like a deterministic finite automaton with transitions labelled by constraints, rather than symbols of the strings. We improve over the derivations of specialized pattern matchers presented in [10-12, 15, 24] because we start from a *nondeterministic* specification of the matcher, while in those papers the initial programs are deterministic. As already mentioned, the improvement over [23] is that we now deal with a general pattern matcher presented as a constraint logic program.

In our example we define a matching relation  $leq\_match(P, S)$  between a pattern  $P = [p_1, \ldots, p_n]$  and a string S, which holds iff in S there exists a substring  $Q = [q_1, \ldots, q_n]$  and for all  $i = 1, \ldots, n$ , we have that  $p_i \leq q_i$ . The following CLP program  $Leq\_Match$  can be taken as the specification of our general pattern matching problem:

- 1.  $leq_match(P, S) \leftarrow app(B, C, S) \land app(A, Q, B) \land leq(P, Q)$
- 2.  $app([], Ys, Ys) \leftarrow$
- 3.  $app([X|Xs], Ys, [X|Zs]) \leftarrow app(Xs, Ys, Zs)$
- 4.  $leq([],[]) \leftarrow$

5.  $leq([X|Xs], [Y|Ys]) \leftarrow X \leq Y \land leq(Xs, Ys)$ 

where *app* denotes the list concatenation. Now let us suppose that we want to specialize this general program w.r.t. the pattern P = [1, 0, 2]. We start off by introducing the following definition:

6.  $leq_match_{sp}(S) \leftarrow leq_match([1, 0, 2], S)$ 

Clauses 1–6 constitute the initial program  $P_0$  from which we begin our program specialization process. We generate a sequence of programs, each of which is derived from the previous one by applying a transformation rule (Section 4) according to the Determinization Strategy (Section 5). As indicated in Section 5, we will get the following final program  $Leq_Match_{sp}$ :

9.  $leq\_match_{sp}(S) \leftarrow new1(S)$ 16.  $new1([X|Xs]) \leftarrow 1 \le X \land new2(Xs)$ 17.  $new1([X|Xs]) \leftarrow 1 > X \land new1(Xs)$ 18.  $new2([X|Xs]) \leftarrow 1 \le X \land new3(Xs)$ 19.  $new2([X|Xs]) \leftarrow 0 \le X \land 1 > X \land new4(Xs)$ 20.  $new2([X|Xs]) \leftarrow 0 > X \land new1(Xs)$ 21.  $new3([X|Xs]) \leftarrow 2 \le X \land new5(Xs)$ 22.  $new3([X|Xs]) \leftarrow 1 \le X \land 2 > X \land new4(Xs)$ 23.  $new3([X|Xs]) \leftarrow 0 > X \land new1(Xs)$ 24.  $new3([X|Xs]) \leftarrow 0 > X \land new1(Xs)$ 25.  $new4([X|Xs]) \leftarrow 0 > X \land new6(Xs)$ 26.  $new4([X|Xs]) \leftarrow 1 \le X \land 2 > X \land new2(Xs)$ 27.  $new4([X|Xs]) \leftarrow 1 > X \land new1(Xs)$ 28.  $new5([X|Xs]) \leftarrow$ 29.  $new6([X|Xs]) \leftarrow$ 

This final program is deterministic in the sense that at most one clause can be applied during the evaluation of every ground goal. As in the case of the Knuth-Morris-Pratt matcher, the efficiency of this final program is very high because it behaves like a deterministic finite automaton.

## 3 Preliminaries

In this section we recall some basic notions of constraint logic programming. For notions not defined here the reader may refer to [1, 17, 21].

Syntax of Constraint Logic Programs. We consider a first order language  $\mathcal{L}$  generated by an infinite set Vars of variables, a set Funct of function symbols with arity, and a set Pred of predicate symbols with arity. We assume that Pred is the union of two disjoint sets: (i) the set Pred<sub>c</sub> of constraint predicate symbols, including true, false, and the equality symbol =, and (ii) the set Pred<sub>u</sub> of user defined predicate symbols. Terms and formulas of  $\mathcal{L}$  are constructed from

the element in *Vars, Funct*, and *Pred*, by means of connectives  $(\neg, \land, \lor)$  and quantifiers  $(\forall, \exists)$ , as usually done in first order logic.

Given a sequence of terms or formulas  $e_1, \ldots, e_n$  (with n > 0), the set of variables occurring in that sequence is denoted by  $vars(e_1, \ldots, e_n)$ . Given a formula  $\varphi$ , the set of the *free variables* in  $\varphi$  is denoted by  $FV(\varphi)$ . A term or a formula is ground iff it contains no variable. Given a set  $X = \{X_1, \ldots, X_n\}$  of variables, by  $\forall X \varphi$  we denote the formula  $\forall X_1 \ldots \forall X_n \varphi$ . By  $\forall (\varphi)$  we denote the *universal closure* of  $\varphi$ , that is, the formula  $\forall X \varphi$ , where  $FV(\varphi) = X$ . Analogously, by  $\exists (\varphi)$  we denote the *existential closure* of  $\varphi$ .

A primitive constraint is an atomic formula  $p(t_1, \ldots, t_n)$  where p is a predicate symbol in  $Pred_c$  and  $t_1, \ldots, t_n$  are terms. The set C of constraints, ranged over by  $c, d, \ldots$ , is the smallest set of formulas of  $\mathcal{L}$  which contains all primitive constraints and it is closed w.r.t. all connectives and quantifiers.

An atom A is an atomic formula  $p(t_1, \ldots, t_n)$  where p is an element of  $Pred_u$ and  $t_1, \ldots, t_n$  are terms. A goal G is the conjunction of  $m (\geq 0)$  atoms. A constrained goal  $c \wedge G$  is the conjunction of a constraint and a goal. The empty conjunction of constraints or atoms is identified with true.

A clause  $\gamma$  is a formula of the form  $H \leftarrow c \wedge G$ , where: (i) H is an atom, called the *head* of  $\gamma$ , and (ii)  $c \wedge G$  is a constrained goal, called the *body* of  $\gamma$ . Clauses of the form  $H \leftarrow c$  are called *constrained facts*. Clauses of the form  $H \leftarrow true$  are also written as  $H \leftarrow$ .

A constraint logic program (or program, for short) is a finite set of clauses. (Here we do not allow for negated atoms in the bodies of clauses.)

Given a program P, we say that a predicate p depends on a predicate q iff either there exists in P a clause whose head predicate is p and whose body contains an occurrence of q or there exists a predicate r such that p depends on r and r depends on q.

Given two atoms  $p(t_1, \ldots, t_n)$  and  $p(u_1, \ldots, u_n)$ , we denote by  $p(t_1, \ldots, t_n) = p(u_1, \ldots, u_n)$  the conjunction of the constraints:  $t_1 = u_1 \land \ldots \land t_n = u_n$ .

A variable renaming is a bijective mapping from Vars to Vars. The application of a variable renaming  $\rho$  to a formula  $\varphi$  returns the formula  $\rho(\varphi)$ , called a variant of  $\varphi$ , obtained by replacing each (bound or free) occurrence of X in  $\varphi$ by the variable  $\rho(X)$ . A renamed apart clause is a variant of a clause such that all its (bound or free) variables do not occur elsewhere.

We will feel free to apply to clauses the following two transformations which, as the reader may verify, preserve program semantics (see below):

(1) application of variable renamings, and

(2) replacement of a clause of the form  $H \leftarrow X = t \land c \land G$ , where  $X \notin vars(t)$ , by the clause  $(H \leftarrow c \land G) \{X/t\}$ , and vice versa.

**Least**  $\mathcal{D}$ -model Semantics. We assume that we are given an interpretation  $\mathcal{D}$  for the constraints in  $\mathcal{C}$ . Let D be the carrier of  $\mathcal{D}$ .  $\mathcal{D}$  assigns a subset of  $D^n$  to each *n*-ary constraint predicate symbol in  $Pred_c$ . In particular,  $\mathcal{D}$  assigns the whole carrier D to *true*, the empty set to *false*, and the identity over D to the equality symbol =.

A  $\mathcal{D}$ -interpretation is an interpretation for the formulas of  $\mathcal{L}$  which extends the interpretation  $\mathcal{D}$ . In particular, a  $\mathcal{D}$ -interpretation assigns a subset of  $D^n$  to each *n*-ary user defined predicate symbol in  $Pred_u$ . Thus, a  $\mathcal{D}$ -interpretation is isomorphic to a subset of the set  $\mathcal{B}_{\mathcal{D}}$  where:

 $\mathcal{B}_{\mathcal{D}} = \{ p(d_1, \dots, d_n) \mid p \in Pred_u \text{ and } (d_1, \dots, d_n) \in D^n \}$ 

A  $\mathcal{D}$ -model of a program P is a  $\mathcal{D}$ -interpretation I such that  $I \models \forall(P)$ . It can be shown that for every CLP program P there exists a *least*  $\mathcal{D}$ -model (w.r.t. set inclusion), denoted by  $lm(P, \mathcal{D})$  [17].

**Operational Semantics.** Let  $\mathcal{D}$  be the given interpretation for the constraints in  $\mathcal{C}$ . In order to define the operational semantics of constraint logic programs, we assume that there is a computable total function  $solve: \mathcal{C} \times \mathcal{P}_{fin}(Vars) \to \mathcal{C}$ , where  $\mathcal{P}_{fin}(Vars)$  is the set of all finite subsets of *Vars*. The function solve can be used for simplifying constraints in  $\mathcal{C}$ . We assume that solve is sound w.r.t. constraint equivalence, that is, for all constraints  $c_1, c_2 \in \mathcal{C}$  and for every finite set X of variables, if  $solve(c_1, X) = c_2$  then  $\mathcal{D} \models \forall X((\exists Y c_1) \leftrightarrow (\exists Z c_2))$ , where Y = $FV(c_1) - X$  and  $Z = FV(c_2) - X$ .

We also assume that *solve* is *complete* w.r.t. satisfiability, in the sense that, for any constraint c,

(i)  $solve(c, \emptyset) = true$  iff c is satisfiable, i.e.,  $\mathcal{D} \models \exists (c)$ , and

(ii)  $solve(c, \emptyset) = false$  iff c is unsatisfiable, i.e.,  $\mathcal{D} \models \neg \exists (c)$ .

The totality and the soundness of the *solve* function guarantee the correctness of the transformation strategy (see Section 5). The assumption that *solve* is complete w.r.t. satisfiability is fulfilled by many classes of constraints considered in practice, such as: (quantified) boolean formulas, equations over the Herbrand universe, and equations over the reals. This assumption guarantees that constraint satisfiability tests, which are required in our transformation method, are decidable. Moreover, the completeness w.r.t. satisfiability guarantees that for any constraints  $c_1$  and  $c_2$ , by evaluating  $solve(\forall (c_1 \rightarrow c_2), \emptyset)$  we can check whether or not  $\mathcal{D} \models \forall (c_1 \rightarrow c_2)$  holds.

Now we define the operational semantics of a CLP program P by introducing a derivability relation  $\mapsto_P$  between constrained goals as follows.

 $\begin{array}{ccc} c \land A \land G & \mapsto_P & c \land A = H_1 \land c_1 \land G_1 \land G \\ \text{iff} \end{array}$ 

 $H_1 \leftarrow c_1 \wedge G_1$  is a renamed apart clause of P and  $c \wedge A = H_1 \wedge c_1$  is satisfiable. The relation  $\mapsto_P^*$  is the reflexive and transitive closure of  $\mapsto_P$ . We say that the constrained goal  $c \wedge G$  succeeds in P iff  $c \wedge G \mapsto_P^* d$  for some satisfiable constraint d.

### 4 Rules for Transforming CLP Programs

The process of transforming a given program P thereby deriving a program Q, can be formalized as a sequence  $P_0, \ldots, P_n$  of programs, called a *transformation* sequence, where  $P_0 = P$ ,  $P_n = Q$  and, for  $k = 0, \ldots, n-1$ , program  $P_{k+1}$  is obtained from program  $P_k$  by applying one of the following transformation rules.

### R1. Definition. We introduce a set of clauses

 $\delta_1: newp(X_1, \ldots, X_h) \leftarrow c_1 \wedge G_1$ 

 $\delta_m : newp(X_1, \dots, X_h) \leftarrow c_m \wedge G_m$ 

where: (i) *newp* is a predicate symbol not occurring in  $P_0, \ldots, P_k$ ,

(ii)  $\{X_1,\ldots,X_h\}\subseteq FV(c_1\wedge G_1,\ldots,c_m\wedge G_m)$ , and

(iii) the predicates occurring in  $G_1, \ldots, G_m$  occur also in  $P_0$ .

We derive the new program  $P_{k+1} = P_k \cup \{\delta_1, \ldots, \delta_m\}$ . For  $i \ge 0$ ,  $Defs_i$  is the set of clauses introduced by the definition rule during the transformation sequence  $P_0, \ldots, P_i$ . In particular,  $Defs_0 = \emptyset$ .

**R2. Unfolding.** Let  $\gamma$ :  $H \leftarrow c \land G' \land A \land G''$  be a renamed apart clause of  $P_k$ . By unfolding  $\gamma$  w.r.t. A we derive the set of clauses

$$\begin{split} \varGamma: \; \{H \leftarrow c \land A = H_1 \land c_1 \land G' \land G_1 \land G'' \mid \\ & H_1 \leftarrow c_1 \land G_1 \text{ is a clause in } P_k \text{ and} \\ & c \land A = H_1 \land c_1 \text{ is satisfiable} \} \end{split}$$

and the new program  $P_{k+1} = (P_k - \{\gamma\}) \cup \Gamma$ .

R3. Folding. Let

$$\begin{array}{c} \gamma_1: H \leftarrow c \wedge c_1 \vartheta \wedge G' \wedge G_1 \vartheta \wedge G'' \\ \dots \\ \gamma_m: H \leftarrow c \wedge c_m \vartheta \wedge G' \wedge G_m \vartheta \wedge G'' \end{array}$$

be m(>0) clauses in  $P_k$  and let *newp* be a predicate such that

 $\delta_1 : newp(X_1, \dots, X_h) \leftarrow c_1 \wedge G_1$  $\dots$  $\delta_m : newp(X_1, \dots, X_h) \leftarrow c_m \wedge G_m$ 

are the clauses in  $Defs_k$  which have newp as head predicate. Suppose that, for  $i = 1, \ldots, m$  and for every variable  $X \in (FV(c_i \wedge G_i) - \{X_1, \ldots, X_h\})$ , we have that: (i)  $X\vartheta$  is a variable not occurring in (H, c, G', G''), and (ii) for every variable  $Y \in (FV(c_i \wedge G_i)) - \{X\})$ ,  $X\vartheta$  does not occur in  $Y\vartheta$ . By folding  $\gamma_1, \ldots, \gamma_m$  using  $\delta_1, \ldots, \delta_m$  we derive the clause

$$\eta: \quad H \leftarrow c \wedge G' \wedge newp(X_1, \dots, X_h)\vartheta \wedge G''$$

and the new program  $P_{k+1} = (P_k - \{\gamma_1, \dots, \gamma_m\}) \cup \{\eta\}.$ 

**R4. Clause Removal.** Let  $\gamma$  be a clause in  $P_k$ . We derive the new program  $P_{k+1} = P_k - \{\gamma\}$  if one of the following cases occurs:

(Unsatisfiable Constraint)  $\gamma$  is the clause  $H \leftarrow c \land G$  and c is unsatisfiable, that is,  $\mathcal{D} \models \neg \exists (c)$ ;

(Subsumed Clause)  $\gamma$  is the clause  $(H \leftarrow c_1 \land G_1)\vartheta$  and there exists a clause in  $P_k - \{\gamma\}$  of the form  $H \leftarrow c_2 \land G_2$  such that  $\mathcal{D} \models \forall (c_1 \rightarrow \exists X c_2)$ , where  $X = FV(c_2) - vars(H, G_2)$  and  $G_2$  is a subconjunction of  $G_1$ .

**R5.** Constraint Replacement. Let  $\gamma_1 : H \leftarrow c_1 \wedge G$  be a clause in  $P_k$ . Suppose that for some constraint  $c_2$ , we have that:  $\mathcal{D} \models \forall (\exists Y c_1 \leftrightarrow \exists Z c_2)$  where: (i)  $Y = FV(c_1) - vars(H,G)$ , and (ii)  $Z = FV(c_2) - vars(H,G)$ . In particular, we may take  $c_2 = solve(c_1, vars(H,G))$ . Then we derive the clause

 $\gamma_2: H \leftarrow c_2 \wedge G$ 

and the new program  $P_{k+1} = (P_k - \{\gamma_1\}) \cup \{\gamma_2\}.$ 

R6. Clause Fusion. Let

 $\gamma_1: \quad H \leftarrow c \wedge G \qquad \qquad \gamma_2: \quad H \leftarrow d \wedge G$ 

be clauses in  $P_k$ . Then we derive the clause

 $\gamma: H \leftarrow (c \lor d) \land G$ 

and the new program  $P_{k+1} = (P_k - \{\gamma_1, \gamma_2\}) \cup \{\gamma\}.$ 

### R7. Clause Splitting. Let

 $\gamma: H \leftarrow (c \lor d) \land G$ 

be a clause in  $P_k$ . Then we derive the clauses

 $\gamma_1: \ H \leftarrow c \wedge G \qquad \qquad \gamma_2: \ H \leftarrow d \wedge G$ 

and the new program  $P_{k+1} = (P_k - \{\gamma\}) \cup \{\gamma_1, \gamma_2\}.$ 

The following result ensures the correctness of the transformation rules w.r.t. the least model semantics.

**Theorem 1.** Let  $P_0, \ldots, P_n$  be a transformation sequence. Suppose that, for every  $k \in \{0, \ldots, n-1\}$  such that  $P_{k+1}$  is derived by folding clauses  $\gamma_1, \ldots, \gamma_m$ in  $P_k$  using clauses  $\delta_1, \ldots, \delta_m$  in Defs<sub>k</sub>, one of the following conditions holds: (1) for  $i = 1, \ldots, m$ , clause  $\delta_i$  is unfolded during the construction of  $P_0, \ldots, P_n$ ; (2) the head predicate of  $\delta_1, \ldots, \delta_m$  does not depend on itself in  $P_n$ . Then  $lm(P_0 \cup Defs_n, \mathcal{D}) = lm(P_n, \mathcal{D})$ .

The rules listed above are an extension of the rules for transforming logic programs and constraint logic programs presented in [2, 6, 8, 14, 23, 25]. In particular, the folding rules considered in [2, 6, 8, 25] allow us to fold only one clause at a time, while by using our rule R3 we can fold  $m (\geq 1)$  clauses simultaneously. Our rule R3 is an adaptation to the case of CLP programs of the folding rules considered in [14, 23]. Our clause splitting rule R7 generalizes to constraint logic programs the case splitting rule for logic programs presented in [23]. The folding and clause splitting rule play a crucial role in the strategy for deriving deterministic programs presented in the next section.

## 5 A Strategy for Deriving Deterministic Specialized Programs

In this section we present the *Determinization Strategy* for guiding the application of the transformation rules. By applying this strategy we can derive deterministic, specialized programs starting from nondeterministic, general programs.

#### 5.1 Determinism and Modes

Let us first introduce the following definition.

**Definition 1.** A program P is deterministic w.r.t. a constrained atom  $c_0 \wedge A_0$ iff for all constrained goals  $c \wedge A \wedge G$  such that  $c_0 \wedge A_0 \mapsto_P^* c \wedge A \wedge G$ , there exists at most one clause  $\gamma$  in P with a renamed apart variant  $H_1 \leftarrow c_1 \wedge G_1$ such that the constraint  $c \wedge A = H_1 \wedge c_1$  is satisfiable.

Given a constrained atom, the determinism of a program may depend on whether or not the variables in the atom are grounded by the constraint [17]. Recall that a variable X is said to be grounded by a constraint c iff  $\mathcal{D} \models \exists Y \forall Z \ (c \rightarrow X = Y)$ , where Y is a new variable and  $Z = FV(c) \cup \{X\}$  (i.e., there is at most one value for X which makes c satisfiable). For instance, the following program over integers:

 $p(X, Y) \leftarrow X = 0 \land Y = 0$  $p(X, Y) \leftarrow X > 0 \land Y = 1$ 

is deterministic w.r.t. the constrained atom  $X = 1 \land p(X, Y)$  (where X is grounded by X = 1), while it is not deterministic w.r.t. the constrained atom  $X \leq 1 \land p(X, Y)$  (where X is not grounded by  $X \leq 1$ ). For this reason we now introduce the notion of *mode* which provides information about the groundness of the variables occurring in constrained atoms.

A mode M is a set of expressions of the form  $p(m_1, \ldots, m_h)$  such that: (i) p is a user defined predicate, (ii) for each p there exists at most one expression  $p(m_1, \ldots, m_h)$ , and (iii) for  $i = 1, \ldots, h$ ,  $m_i$  is either + (meaning that every variable in the *i*-th argument of p is grounded by some constraint) or ? (meaning that the *i*-th argument of p is any term). An expression of the form  $p(m_1, \ldots, m_h)$  is called a mode for the predicate p. A mode M is a mode for a program P iff there exists in M a mode for each user defined predicate occurring in P.

Given an atom  $p(t_1, \ldots, t_h)$  and a mode M with the element  $p(m_1, \ldots, m_h)$ , (1) for  $i = 1, \ldots, h$ , the term  $t_i$  is said to be an *input argument* of p (relative to M) iff  $m_i$  is +, and (2) a variable of  $p(t_1, \ldots, t_h)$  which occurs in an input argument of p, is said to be an *input variable* of  $p(t_1, \ldots, t_h)$ .

**Definition 2.** Let P be a program and M be a mode for P. We say that a constrained atom  $c \land p(t_1, \ldots, t_h)$  satisfies M iff  $p(m_1, \ldots, m_h) \in M$  and for  $i=1,\ldots,h$ , if  $m_i$  is + then every variable in  $t_i$  is grounded by c. We say that P satisfies M iff for each constrained atom  $c_0 \land A_0$  which satisfies M, and for each constrained goal  $c \land A \land G$  such that  $c_0 \land A_0 \mapsto_P^* c \land A \land G$ , we have that  $c \land A$  satisfies M.

Often the property that a program satisfies a mode can be automatically verified by abstract interpretation methods [13].

**Definition 3.** We say that a program P is deterministic w.r.t. a mode M iff P is deterministic w.r.t. every constrained atom  $c_0 \wedge A_0$  which satisfies M.

Now we give a sufficient condition which ensures that a program is deterministic w.r.t. a mode. We need the following definition.

**Definition 4.** Let us consider the following two clauses without variables in common:

 $\gamma_1: p(t_1, \dots, t_h, u_1, \dots, u_k) \leftarrow c_1 \wedge G_1$  $\gamma_2: p(v_1, \dots, v_h, w_1, \dots, w_k) \leftarrow c_2 \wedge G_2$ 

where p is a k-ary predicate whose first h arguments are input arguments relative to a given mode M. We say that  $\gamma_1$  and  $\gamma_2$  are mutually exclusive w.r.t. M iff  $\mathcal{D} \models \neg \exists (t_1 = v_1 \land \ldots \land t_h = v_h \land c_1 \land c_2).$ 

**Proposition 1.** Let P be a program and M be a mode for P. If P satisfies M and the clauses of P are pairwise mutually exclusive w.r.t. M, then P is deterministic w.r.t. M.

### 5.2 The Determinization Strategy

Our Determinization Strategy is based upon the following three subsidiary strategies: (i) Unfold-Simplify, which uses the unfolding, clause removal, and constraint replacement rules, (ii) Partition, which uses the clause removal, constraint replacement, clause fusion, and clause splitting rules, and (iii) Define-Fold, which uses the definition and folding rules.

Let us consider an initial program P, a mode M for P, and a constrained atom  $c \wedge p(t_1, \ldots, t_h)$ , with  $FV(c) \subseteq vars(t_1, \ldots, t_h)$ . In order to specialize Pw.r.t.  $c \wedge p(t_1, \ldots, t_h)$ , we introduce, by the definition rule, the clause

 $\delta_{sp}: p_{sp}(X_1, \dots, X_r) \leftarrow c \wedge p(t_1, \dots, t_h)$ 

where  $X_1, \ldots, X_r$  are the distinct variables occurring in  $p(t_1, \ldots, t_h)$ . The mode  $p_{sp}(m_1, \ldots, m_r)$  for the predicate  $p_{sp}$  is the following: for  $j = 1, \ldots, r, m_j$  is + iff  $X_j$  is an input variable of  $p(t_1, \ldots, t_h)$  relative to M. We assume that P satisfies M and thus, the program  $P \cup \{\delta_{sp}\}$  satisfies  $M \cup \{p_{sp}(m_1, \ldots, m_r)\}$ .

Our Determinization Strategy is an iterative procedure that at each iteration manipulates the following three sets of clauses: (1) Defs, which is the set of clauses introduced so far by the definition rule, (2) Cls, which is the set of clauses to be transformed during the current iteration, and (3)  $P_{sp}$ , which is the specialized program derived so far. Initially, both Defs and Cls consist of the single clause  $\delta_{sp}$ . From the set Cls a new set of deterministic clauses is derived by applying the transformation rules according to the Unfold-Simplify, Partition, and Define-Fold subsidiary strategies. This new set of deterministic clauses is added to  $P_{sp}$ . During each iteration, in order to derive deterministic clauses, we may need to introduce new predicates, whose defining clauses are stored in the set NewDefs. At the end of each iteration NewDefs is added to Defs, and the value of the set Cls is updated to NewDefs. The transformation strategy terminates when Cls =  $\emptyset$ , that is, when no new predicate is introduced during an iteration.

The following definition is needed for presenting the *Unfold-Simplify* subsidiary strategy. **Definition 5.** Let  $H \leftarrow c \land G' \land A \land G''$  be a clause in a program P and let M be a mode for P. We say that A is a consumer atom iff for every renamed apart clause  $H_1 \leftarrow c_1 \land G_1$  in P, we have that one of the following three conditions holds:

(i)  $G_1$  is the empty conjunction;

(ii)  $c \wedge A = H_1 \wedge c_1$  is unsatisfiable;

(iii)  $\mathcal{D} \models \forall (c \rightarrow \exists Y (A = H_1))$  where  $Y = \{X \in FV (A = H_1) \mid X \text{ is not an input variable of } A \text{ relative to } M\}$ .

During the Unfold-Simplify subsidiary strategy we unfold w.r.t. consumer atoms. In particular, when Condition (iii) of Definition 5 holds, we unfold w.r.t. atoms whose input arguments are instances of the corresponding arguments in the heads of the clauses of P.

### **Determinization Strategy**

**Input**: A program P, a mode M for P such that P satisfies M, and a clause  $\delta_{sp}$ :  $p_{sp}(X_1, \ldots, X_r) \leftarrow c \wedge p(t_1, \ldots, t_h)$ 

**Output**: A specialized program  $P_{sp}$  and a mode  $M_{sp}$  for  $P_{sp}$ .

Initialize: Defs :=  $\{\delta_{sp}\}$ ; Cls :=  $\{\delta_{sp}\}$ ;  $P_{sp} := \emptyset$ ;  $M_{sp} := \{p_{sp}(m_1, \dots, m_r)\}$ ; while  $Cls \neq \emptyset$  do

(1) Unfold-Simplify:

 $UnfCls := \{\eta \mid \eta \text{ is a constrained fact in } Cls \text{ or } \eta \text{ is derived by unfolding a clause in } Cls w.r.t. the leftmost atom in its body};$ 

while there exists a clause  $\gamma$  in UnfCls whose body has a leftmost consumer atom  ${\bf do}$ 

 $UnfCls := (UnfCls - \{\gamma\}) \cup \{\eta \mid \eta \text{ is derived by unfolding } \gamma \text{ w.r.t. the leftmost}$ consumer atom in its body };

$$\begin{aligned} \textit{UnfCls} &:= \{ H \leftarrow \widetilde{c} \land G \mid \textit{there exists } H \leftarrow c \land G \textit{ in } \textit{UnfCls such that:} \\ &(\textit{i}) \ \widetilde{c} = \textit{solve}(c, \textit{vars}(H, G)), \\ &(\textit{ii}) \ c \textit{ is satisfiable, and} \\ &(\textit{iii}) \ H \leftarrow c \land G \textit{ is not subsumed by any other clause in } \textit{UnfCls} . \end{aligned}$$

(2) *Partition*: We apply the clause removal, constraint replacement, clause fusion, and clause splitting rules, and we derive from *UnfCls* a set *PartCls* of clauses which is the union of disjoint subsets, called *packets*, such that the following two properties hold.

(i) Each packet is a set of clauses of the form:

 $H \leftarrow c \wedge d_1 \wedge G_1$  $\dots$  $H \leftarrow c \wedge d_m \wedge G_m$ 

In particular, if for i = 1, ..., m,  $G_i$  is the empty conjunction, then by clause fusion we derive a packet consisting of one constrained fact only.

(ii) Any two clauses belonging to different packets are mutually exclusive w.r.t. mode  ${\cal M}_{sp}.$ 

(3) Define-Fold: Let CFacts be the union of the packets in PartCls consisting of constrained facts only, and let NonCFacts be the union of all other packets. Let NewDefs be a (possibly empty) set of new clauses introduced by the definition rule such that each packet in NonCFacts can be folded by using clauses in Defs ∪ NewDefs of the form:

$$newp(X_1,\ldots,X_r) \leftarrow d_1 \wedge G_1$$

 $newp(X_1,\ldots,X_r) \leftarrow d_m \wedge G_m$ 

thereby deriving from each packet a single clause of the form:

 $H \leftarrow c \land newp(X_1, \ldots, X_r)$ 

When we introduce *NewDefs* and perform folding, we also make sure that Condition (1) or (2) of Theorem 1 holds.

For each new predicate *newp* defined in *NewDefs*, we add to  $M_{sp}$  the mode  $newp(m_1, \ldots, m_r)$  defined as follows: for  $i = \ldots r$ ,  $m_i = +$  iff  $X_i$  is either an input variable of H or an input variable of the leftmost atom of one of the goals  $G_1, \ldots, G_m$ .

Let FldCls be the set of clauses derived by folding the packets in NonCFacts. (4)  $Defs := Defs \cup NewDefs; Cls := NewDefs; P_{sp} := P_{sp} \cup CFacts \cup FldCls$ 

#### end-while

As a consequence of Theorem 1, if the Determinization Strategy terminates, then the specialized program  $P_{sp}$  is equivalent to the initial program P in the sense indicated by the following theorem.

**Theorem 2 (Correctness of the Determinization Strategy).** Let P be a program, M be a mode for P such that P satisfies M, and  $\delta_{sp}$  be a clause of the form  $p_{sp}(X_1, \ldots, X_r) \leftarrow c \wedge p(t_1, \ldots, t_h)$ . Let  $P_{sp}$  be the specialized program returned by the Determinization Strategy. Then, for all substitutions  $\vartheta = \{X_1/d_1, \ldots, X_r/d_r\}$ , where  $d_1, \ldots, d_r$  are ground terms:

 $lm(P, \mathcal{D}) \models (c \land p(t_1, \ldots, t_h))\vartheta$  iff  $lm(P_{sp}, \mathcal{D}) \models (p_{sp}(X_1, \ldots, X_r))\vartheta$ .

We also have that the program derived by the Determinization Strategy is deterministic, as stated by the following theorem.

**Theorem 3 (Determinism).** Let P be a program, M be a mode for P such that P satisfies M, and  $\delta_{sp}$  be a clause of the form  $p_{sp}(X_1, \ldots, X_r) \leftarrow c \wedge p(t_1, \ldots, t_h)$ . Let  $P_{sp}$  be the specialized program and let  $M_{sp}$  be the mode for  $P_{sp}$  returned by the Determinization Strategy. Then,  $P_{sp}$  is deterministic w.r.t.  $M_{sp}$ .

*Proof.* By construction,  $P_{sp}$  satisfies  $M_{sp}$  and its clauses are pairwise mutually exclusive w.r.t.  $M_{sp}$ . Thus, by Proposition 1 we get the thesis.

The Determinization Strategy can be applied in a fully automatic way and, indeed, it has been implemented on the MAP transformation system [7]. We do not describe in this paper the implementation details. In Section 6 we describe specializing various CLP pattern matching programs by using our MAP system.

Now we show the Determinization Strategy in action on the matching example of Section 2. This will explain how the specialized program  $Leq_Match_{sp}$  has been automatically derived. We are given the program consisting of clauses 1–5, the mode  $M = \{leq_match(+,+), app(?,?,+), leq(+,+)\}$ , and  $\delta_{sp} = \text{clause } 6$ . Thus, initially,  $Defs = Cls = \{\text{clause } 6\}$  and  $M_{sp} = \{leq_match_{sp}(+)\}$ . Since  $Cls \neq \emptyset$ , we execute the body of the while-loop and we unfold clause 6 w.r.t.  $leq_match([1,0,2],S)$  and we get:

7.  $leq_match_{sp}(S) \leftarrow app(B, C, S) \land app(A, Q, B) \land leq([1, 0, 2], Q)$ 

Clause 7 is a packet in itself and in order to fold it, we introduce the following definition:

8.  $new1(S) \leftarrow app(B, C, S) \land app(A, Q, B) \land leq([1, 0, 2], Q)$ 

and then we fold clause 7, thereby getting:

9.  $leq\_match_{sp}(S) \leftarrow new1(S)$ 

Now  $Defs = \{\text{clause 6}, \text{clause 8}\}, Cls = \{\text{clause 8}\}, \text{ and } M_{sp} = \{leq\_match_{sp}(+), new1(+)\}.$  Since  $Cls \neq \emptyset$ , we execute once more the body of the while-loop and we unfold clause 8 w.r.t. the atoms *app* and *leq*. We get:

10.  $new1([X|Xs]) \leftarrow 1 \leq X \land app(Q, C, Xs) \land leq([0, 2], Q)$ 

11.  $new1([X|Xs]) \leftarrow app(B, C, Xs) \land app(A, Q, B) \land leq([1, 0, 2], Q)$ 

Since clause 10 and 11 are not mutually exclusive w.r.t.  $M_{sp}$ , we apply the clause splitting rule to clause 11, thereby getting:

12.  $new1([X|Xs]) \leftarrow 1 \leq X \land app(B, C, Xs) \land app(A, Q, B) \land leq([1, 0, 2], Q)$ 

13.  $new1([X|Xs]) \leftarrow 1 > X \land app(B, C, Xs) \land app(A, Q, B) \land leq([1, 0, 2], Q)$ 

We have two packets: (i) {clause 10, clause 12} and (ii) {clause 13}. In order to fold the first packet we introduce the following definition:

14.  $new2(Xs) \leftarrow app(Q, C, Xs) \land leq([0, 2], Q)$ 

15.  $new2(Xs) \leftarrow app(B, C, Xs) \land app(A, Q, B) \land leq([1, 0, 2], Q)$ 

We fold clauses 10 and 12 by using clauses 14 and 15, and we fold clause 13 by using clause 8. We get the following mutually exclusive clauses:

16. 
$$new1([X|Xs]) \leftarrow 1 \le X \land new2(Xs)$$
  
17.  $new1([X|Xs]) \leftarrow 1 > X \land new1(Xs)$ 

Now  $Defs = \{$ clause 6, clause 8, clause 14, clause 15 $\}$ ,  $Cls = \{$ clause 14, clause 15 $\}$ , and  $M_{sp} = \{$ leq\_match<sub>sp</sub>(+), new1(+), new2(+) $\}$ . Since  $Cls \neq \emptyset$ , the derivation continues by executing again the body of the while-loop. Thus, we unfold the clauses 14 and 15. We will not give all the details of the derivation here. We eventually get the specialized, deterministic program  $Leq_Match_{sp}$  of Section 2.

The termination of our Determinization Strategy depends on the finiteness of: (i) the unfolding subsidiary strategy, and (ii) the set of definitions which are introduced for performing folding steps. In particular, for ensuring termination it may be necessary to consider suitable generalizations of the bodies of the clauses to be folded (see, for instance, the techniques presented in [5, 8, 12, 18, 20, 26]).

However, the generalization technique required when applying the Determinization Strategy is more sophisticated than the ones proposed for the case of partial evaluation of (constraint) logic programs [8, 12, 20]. This is due to the fact that the new definitions which are introduced during the Determinization Strategy are more complex than the ones introduced during partial evaluation, because the latter ones essentially consist of one clause whose body is a single (constrained) atom. We leave it for further investigation the issue of designing suitable generalization techniques for our Determinization Strategy.

### 6 More Examples of Constrained Matching

We have applied our Determinization Strategy for specializing several nondeterministic matching programs. By using our MAP system [7], we automatically performed the specialization of the *Leq\_Match* program illustrated in Section 2 and the specialization of the programs listed below.

**Near Matching.** Given a pattern  $P = [p_1, \ldots, p_n]$ , a real number K, and a string S, *near\_match*(P, K, S) holds iff there exists a sublist  $Q = [q_1, \ldots, q_n]$  of S such that for  $i = 1, \ldots, n$ , we have that  $|p_i - q_i| \leq K$ . The initial program *Near\_Match* is the following one:

 $\begin{array}{l} near\_match(P,K,S) \leftarrow app(B,C,S) \land app(A,Q,B) \land near(P,K,Q) \\ near([],K,[]) \leftarrow \\ near([X|Xs],K,[Y|Ys]) \leftarrow X \geq Y \land X - Y \leq K \land near(Xs,K,Ys) \\ near([X|Xs],K,[Y|Ys]) \leftarrow X < Y \land Y - X \leq K \land near(Xs,K,Ys) \end{array}$ 

together with the usual clauses for the predicate *app*. By specializing program  $Near_Match$  w.r.t. the goal  $near_match([2, 0], 2, S)$  we obtain the following deterministic program:

 $\begin{array}{l} near\_match_{sp}(S) \leftarrow new1(S) \\ new1([X|Xs]) \leftarrow X \ge 0 \land X \le 2 \land new2(Xs) \\ new1([X|Xs]) \leftarrow X < 0 \land new1(Xs) \\ new1([X|Xs]) \leftarrow X > 2 \land X \le 4 \land new2(Xs) \\ new1([X|Xs]) \leftarrow X > 4 \land new1(Xs) \\ new2([X|Xs]) \leftarrow X > 0 \land X \le 2 \land new3(Xs) \\ new2([X|Xs]) \leftarrow X \ge 0 \land new3(Xs) \\ new2([X|Xs]) \leftarrow X \ge -2 \land X < 0 \land new4(Xs) \\ new2([X|Xs]) \leftarrow X < -2 \land new1(Xs) \\ new2([X|Xs]) \leftarrow X > 4 \land new1(Xs) \\ new2([X|Xs]) \leftarrow X > 4 \land new1(Xs) \\ new3(Xs) \leftarrow \\ new4(Xs) \leftarrow \end{array}$ 

**Near Multimatching.** Given a list Ps of patterns, a real number K, and a string S,  $near\_mmatch(Ps, K, S)$  holds iff there exists a pattern P in Ps such that  $near\_match(P, K, S)$  holds. The initial program  $Near\_Multimatch$  is the following one:

 $near\_mmatch([P|Ps], K, S) \leftarrow near\_match(P, K, S)$  $near\_mmatch([P|Ps], K, S) \leftarrow near\_mmatch(Ps, K, S)$  together with the clauses needed for the predicate  $near\_match$  (see the above program  $Near\_Match$ ). By specializing the program  $Near\_Multimatch$  w.r.t. the goal  $near\_mmatch([[1,1],[1,2]],1,S)$  we obtain the following deterministic program:

 $\begin{array}{l} near\_mmatch_{sp}(S) \leftarrow new1(S) \\ new1([X|Xs]) \leftarrow X \geq 0 \land X \leq 1 \land new2(Xs) \\ new1([X|Xs]) \leftarrow X < 0 \land new1(Xs) \\ new1([X|Xs]) \leftarrow X > 1 \land X \leq 2 \land new2(Xs) \\ new1([X|Xs]) \leftarrow X > 2 \land new1(Xs) \\ new2([X|Xs]) \leftarrow X > 1 \land X \leq 2 \land new3(Xs) \\ new2([X|Xs]) \leftarrow X > 1 \land X \leq 2 \land new3(Xs) \\ new2([X|Xs]) \leftarrow X \geq 0 \land X < 1 \land new3(Xs) \\ new2([X|Xs]) \leftarrow X \geq 0 \land X < 1 \land new3(Xs) \\ new2([X|Xs]) \leftarrow X > 2 \land X \leq 3 \land new5(Xs) \\ new2([X|Xs]) \leftarrow X > 3 \land new1(Xs) \\ new3(Xs) \leftarrow \\ new4(Xs) \leftarrow \\ new5(Xs) \leftarrow \end{array}$ 

**Regular Expressions of Intervals.** A regular expression of intervals is an extension of the standard notion of regular expression where symbols are replaced by intervals of real numbers. Given a regular expression E of intervals and a string S of reals,  $in\_language(E, S)$  holds iff S belongs to the language denoted by E. The initial program  $Int\_Reg\_Expr$  is the following one:

 $\begin{array}{l} in\_language(E,S) \leftarrow string(S) \land accepts(E,S) \\ accepts(lt(X),[Y]) \leftarrow Y < X \\ accepts(gt(X),[Y]) \leftarrow Y > X \\ accepts(int(X_1,X_2),[Y]) \leftarrow Y > X_1 \land Y < X_2 \\ accepts(E_1 \cdot E_2,S) \leftarrow app(S_1,S_2,S) \land accepts(E_1,S_1) \land accepts(E_2,S_2) \\ accepts(E_1 + E_2,S) \leftarrow accepts(E_1,S) \\ accepts(E_1 + E_2,S) \leftarrow accepts(E_2,S) \\ accepts(E^*,[]) \leftarrow \\ accepts(E^*,S) \leftarrow ne\_app(S_1,S_2,S) \land accepts(E,S_1) \land accepts(E^*,S_2) \\ ne\_app([X],Ys,[X|Ys]) \leftarrow \\ ne\_app([X]Xs],Ys,[X|Zs]) \leftarrow ne\_app(Xs,Ys,Zs) \\ string([]) \leftarrow \\ string([X|Xs]) \leftarrow string(Xs) \end{array}$ 

where: (i) *app* is the usual *append* predicate on lists, and (ii)  $ne\_app(Xs, Ys, Zs)$  holds iff the concatenation of the non-empty list Xs and the list Ys is the list Zs. By specializing program  $Int\_Reg\_Expr$  w.r.t. the goal  $in\_language(int(0,2)^* + int(1,3)^*), S)$  we obtain the following deterministic program:

 $\begin{array}{l} in\_language_{sp}(S) \leftarrow new1(S) \\ new1([]) \leftarrow \\ new1([X|Xs]) \leftarrow X > 1 \land X < 2 \land new2(Xs) \\ new1([X|Xs]) \leftarrow X \geq 2 \land X < 3 \land new3(Xs) \\ new1([X|Xs]) \leftarrow X > 0 \land X \leq 1 \land new4(Xs) \end{array}$ 

 $\begin{array}{l} new2([]) \leftarrow \\ new2([X|Xs]) \leftarrow X > 1 \land X < 2 \land new2(Xs) \\ new2([X|Xs]) \leftarrow X \ge 2 \land X < 3 \land new3(Xs) \\ new2([X|Xs]) \leftarrow X > 0 \land X \le 1 \land new4(Xs) \\ new3([]) \leftarrow \\ new3([X|Xs]) \leftarrow X > 1 \land X < 3 \land new3(Xs) \\ new4([]) \leftarrow \\ new4([X|Xs]) \leftarrow X > 0 \land X < 2 \land new4(Xs) \end{array}$ 

**Experimental Results.** Table 1 shows the speedups achieved by applying our Determinization Strategy to the program of Section 2 and the programs of this section. Our experiments have been performed on a PentiumIII, 900MHz, running Linux 2.4.17, SICStus 3.8.5, and Holzbaur's clp(q,r) solver [16]. All speedups have been measured for input strings where the matching pattern occurs at the hundredth position. On the second, fifth, and sixth row of the Speedup column, we have reported a range of values, instead of a single value, because we have obtained different speedups for different input strings. Higher speedups have been obtained in the case of input strings for which the initial, nondeterministic program requires more backtracking for finding an occurrence of the matching pattern.

Program	Specialization Goal	Speedup
$Leq\_Match$	$leq_match([1, 0, 2], S)$	2.74
Near_Match	$near\_match([2,0],2,S)$	37.6 - 56.1
		45.7
		45.1
$Int\_Reg\_Expr$	$in\_language(int(0,2)^* + int(1,3)^*,S)$	1.2 - 7
Int_Reg_Expr	$in\_language(((gt(1) \cdot gt(1)^*) + (gt(1) \cdot gt(1)^* \cdot lt(0))), S))$	1.3 - 320

Table 1. Speedups achieved by the Determinization Strategy

## 7 Conclusions

We have introduced a new transformation rule, called *clause splitting*, which allows us to *reason by cases* and can be used for reducing nondeterminism when specializing constrained logic programs. We have also extended the folding rule for constrained logic programming by allowing folding of several clauses in a single step, and we have presented a strategy, called *Determinization Strategy*, for the automatic application of the transformation rules with the objective of deriving deterministic, specialized programs.

Our transformation technique preserves the least  $\mathcal{D}$ -model semantics. One could slightly modify the transformation rules and the transformation strategy if different semantics are to be preserved.

The Determinization Strategy is an extension to constraint logic programs of the strategy presented in [23]. Our strategy is also an enhancement of *conjunctive partial deduction* [5], in that we allow new predicates to be defined in terms of *disjunctions of conjunctions* of constrained atoms. We have used our strategy for specializing constrained matching algorithms and we have derived programs which are highly efficient because, as in the case of the Knuth-Morris-Pratt matcher, they correspond to deterministic finite automata. Their transitions, however, are labelled by constraints, rather than symbols.

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