

# Verifying CTL Properties\* of Infinite State Systems by Specializing Constraint Logic Programs

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(Extended Abstract)

## 1 Introduction

The goal of automated verification of systems is the definition and the implementation of logical frameworks which allow one: (i) to formally specify these systems, and (ii) to prove their properties in an automatic way. These logical frameworks require formalisms both for the description of the systems and the description of their properties.

In this paper we assume that a system makes transitions from states to states and the evolution of a system can be formalized using a *computation tree*, which is defined as follows. Given a system  $S$  and its initial state  $s_0$ , the root of the computation tree for  $S$  is  $s_0$ , and every node  $s_i$  of the computation tree for  $S$  has a child node  $s_j$  iff there exists in  $S$  a transition from state  $s_i$  to state  $s_j$ . The state  $s_j$  is called a *successor* state of  $s_i$ . We assume that in every system for every state  $s_i$  there exists at least one successor state. Notice that the set of all states of a system may be finite or infinite.

We also assume that the description of the properties of any system can be done in the *Computation Tree Logic* formalism [4] (CTL for short). CTL formulas describe properties of computation trees, and they are built using: (i) atomic state properties, (ii) logical connectives:  $\neg, \wedge, \vee$ , (iii) quantifiers over paths:  $A$  ('for all paths') and  $E$  ('for some path'), and (iv) quantifiers along paths:  $G$  ('for all states on the path'),  $F$  ('for some state on the path'). CTL formulas are very expressive, and in particular, one may use them to describe the so called *safety* and *liveness* properties. Given a CTL formula  $\varphi$  and state  $s$ , the semantics of CTL defines the satisfaction relation  $s \models \varphi$  which holds whenever  $\varphi$  is true in  $s$  [4].

In this paper we will present a method for verifying CTL properties of possibly infinite state systems by using Constraint Logic Programming [7] (CLP, for short) and Program Specialization. Our method is applicable to a large class of concurrent systems, like those described by [14].

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Constraint Logic Programming extends standard logic programming by incorporating mechanisms for solving constraints over some given constraint domains by using domain specific efficient algorithms. The following are constraint domains which are usually considered: (i) the domain of inequations over real or rational numbers, (ii) the domain of boolean formulas, and (iii) every finite domain. CLP programs turn out to be very suitable for modelling infinite state systems because an infinite set of states can be described by constraints over the state space.

For our purpose of verifying CTL properties of systems which may require the use of negated formulas, we will consider CLP programs with *locally stratified negation* [1]. We assume that the semantics of a program  $P$  in this class is provided by the unique perfect model, denoted by  $M(P)$ , which coincides with the unique stable model, and the total well-founded model. The reader may refer to [1,7,11] for the definition of these models and other notions of logic programming or constraint logic programming which we do not recall here.

Program specialization is a program transformation technique whose goal is the adaptation of a program to the context where it is used. In [5] we have developed a general framework for the automatic specialization of constraint logic programs over a generic constraint domain  $\mathcal{D}$ . The specialization also improves the computational properties w.r.t. a given class of goals, while it preserves the *least  $\mathcal{D}$ -model* [8]. In this paper we use an extension of that framework to deal with locally stratified CLP programs.

Our specialization technique is *correct* w.r.t. the perfect model semantics in the sense that, given a locally stratified CLP program  $P$  and an atom  $A$  whose predicate is defined in  $P$ , and given a program  $Q$  which is a specialization of  $P$  w.r.t.  $A$ , for every ground instance  $A_g$  of  $A$ ,

$$A_g \in M(P) \quad \text{iff} \quad A_g \in M(Q). \quad (1)$$

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**Verification Method.** Our method for verifying whether or not a system  $S$  in its initial state  $s_0$ , satisfies a CTL property  $\varphi$ , consists of the following two steps.

*Step 1.* We introduce a CLP program  $P_S$  which defines a binary predicate *sat* such that  $s_0 \models \varphi$  iff  $\text{sat}(s_0, \varphi) \in M(P_S)$ . We assume that  $s_0$  and  $\varphi$  are ground terms.

*Step 2.* We introduce a new 0-ary predicate  $f$  defined by the clause  $f \leftarrow \text{sat}(s_0, \varphi)$  and thus,  $\text{sat}(s_0, \varphi) \in M(P_S)$  iff  $f \in M(P_S \cup \{f \leftarrow \text{sat}(s_0, \varphi)\})$ . We then apply our program specialization technique and transform the program  $P_S \cup \{f \leftarrow \text{sat}(s_0, \varphi)\}$  into a specialized program  $P_f$ . By the correctness of program specialization, stated by the equivalence (1) above, we have that  $f \in M(P_S \cup \{f \leftarrow \text{sat}(s_0, \varphi)\})$  iff  $f \in M(P_f)$ .

Putting Step 1 and Step 2 together, we have that  $s_0 \models \varphi$  iff  $f \in M(P_f)$ , and we can check whether or not  $s_0 \models \varphi$  as follows: (i) if the unit clause  $f \leftarrow$  occurs in  $P_f$  then  $s_0 \models \varphi$ , and (ii) if no clause with head  $f$  occurs in  $P_f$  (that is,  $f$  has an empty definition in  $P_f$ ) then it is not the case that  $s_0 \models \varphi$ .

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The structure of our paper is as follows. In Section 2 we show how Step 1 of our verification method can be realized by encoding CTL properties of a finite or infinite state system as a constraint logic program with locally stratified negation. In Section 3 we describe the program specialization technique which we use to realize Step 2 of our verification method. In Section 4 we show how our method works for the verification of a safety and a liveness property of the bakery protocol for mutual exclusion [9]. Finally, in Section 5 we compare our work with related verification techniques described in the literature.

## 2 Encoding CTL Properties as Constraint Logic Programs

Given a system  $S$ , its initial state  $s_0$ , and a CTL property  $\varphi$ , Step 1 of our verification method is realized by providing the recursive definition of the relation  $s_0 \models \varphi$  as a locally stratified program  $P_S$ . Step 1 can be performed in an automatic way for a very large class of concurrent systems, namely those which are state transition systems with *enabling conditions* and *actions* [14] with conditions and actions which can be expressed by constraints over the values of *state variables*.

The following simple example will clarify the reader's ideas. This example will be used throughout this paper to illustrate our approach. Let  $S_0$  be the system whose set of states is a subset of  $\{a, b\} \times \mathbf{Z}$ , where  $\mathbf{Z}$  is the set of integers. Let the initial state of  $S_0$  be the pair  $\langle a, 0 \rangle$ , and let us assume that the transitions between states are the following ones:

$$\begin{aligned} & \langle a, 1 \rangle \rightarrow \langle b, 1 \rangle \\ \forall y \in \mathbf{Z}. & \langle a, y \rangle \rightarrow \langle a, y+2 \rangle \\ \forall y \in \mathbf{Z}. & \langle b, y \rangle \rightarrow \langle b, y-1 \rangle \end{aligned}$$

The system  $S_0$  can be depicted as shown in Figure 1.

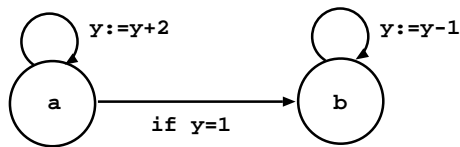


Figure 1. The system  $S_0$  whose set of states is  $\{a, b\} \times \mathbf{Z}$ .

We want to verify that the property that starting from the initial state  $\langle a, 0 \rangle$ , the system  $S_0$  never reaches a state  $\langle x, y \rangle$  with  $y < 0$ , for some  $x \in \{a, b\}$ . This property is expressed by the satisfaction relation  $\langle a, 0 \rangle \models \neg EF \text{ neg}$  which asserts that in the initial state  $\langle a, 0 \rangle$  the CTL formula  $\neg EF \text{ neg}$  is true, where  $\text{neg}$  is the atomic state property which holds in a state  $\langle x, y \rangle$  iff  $y < 0$ .

The satisfaction relation  $s \models \varphi$  between the state  $s$  and the CTL formula  $\varphi$ , is recursively defined as follows:

$$\begin{aligned} s \models p & \text{ iff } p \text{ is an atomic state property and } p \text{ holds in } s \\ s \models \neg \varphi & \text{ iff it is not the case that } s \models \varphi \end{aligned}$$

$s \models EF \varphi$  iff either  $s \models \varphi$   
or  $\exists$  a state  $s_1$  such that (i)  $\exists$  a transition from  $s$  to  $s_1$  and  
(ii)  $s_1 \models EF \varphi$

For the system  $S_0$  the relation  $s \models \varphi$  can be encoded by the following locally stratified program  $P_{S_0}$  written a Prolog-like syntax, where the state  $\langle x, y \rangle$  is denoted by  $s(X, Y)$ :

```

sat(s(X,Y),neg) :- Y<0.
sat(S,not(F)) :- \+ sat(S,F).
sat(S,ef(F)) :- sat(S,F).
sat(s(a,Y),ef(F)) :- Y1=Y+2, sat(s(a,Y1),ef(F)).
sat(s(a,Y),ef(F)) :- Y=1, sat(s(b,Y),ef(F)).
sat(s(b,Y),ef(F)) :- Y1=Y-1, sat(s(b,Y1),ef(F)).

```

Thus, we have that:

$\langle a, 0 \rangle \models \neg EF \text{ neg}$  iff  $\text{sat}(s(a,0), \text{not}(\text{ef}(\text{neg}))) \in M(P_{S_0})$ .

### 3 Verification of CTL Properties via Program Specialization

Step 2 of our verification method is realized by using an automatic program specialization technique which is derived from the one described in [5] and it is a particular case of the *program transformation* technique based on *rules* and *strategies*. The transformation rules ensure the correctness of the specialized program w.r.t. the given initial program. Thus, given a system  $S$  with initial state  $s_0$ , and a CTL property  $\varphi$ , after Step 1 and Step 2, whereby we introduce the predicate  $f$ , and we derive the specialized program  $P_f$ , we have that:

$s_0 \models \varphi$  iff  $f \in M(P_f)$ .

The specialization strategy guides the application of the rules with the aim of deriving a program  $P_f$  where the definition of  $f$  is either (i) the unit clause  $f \leftarrow$  or (ii) it is the empty definition. As already mentioned, in case (i)  $s_0 \models \varphi$  holds, while in case (ii)  $s_0 \models \varphi$  does not hold.

We will show that our specialization strategy terminates for all initial programs  $P_S \cup \{f \leftarrow \text{sat}(s_0, \varphi)\}$ , but due to the undecidability of CTL for infinite state systems, in some cases our specialization strategy may produce a program  $P_f$  in which the definition of  $f$  is neither the unit clause  $f \leftarrow$  nor the empty definition.

However, we have that our strategy is complete for finite state systems.

#### 3.1 The Rules for Program Specialization

Now we introduce some notions which we use below for describing our program specialization technique.

We assume that every atom is *pure*, that is, it is of the form  $p(X_1, \dots, X_m)$ , where  $X_1, \dots, X_m$  are distinct variables. A *constrained atom* is the conjunction of a constraint and an atom. *Goals* are (possibly empty) conjunctions of atoms.

A *constrained goal* is the conjunction of a constraint and a goal. Conjunction is commutative and, thus, the order of constraints and atoms in the body of a clause is immaterial. The empty conjunction (of constraints or atoms) is *true*. *Clauses* are of the form  $H \leftarrow c, G$ .

The assumption that all atoms are pure is not restrictive because, for instance, any clause  $\gamma$  with occurrences of non-pure atoms, can be transformed into a clause  $\delta$  with occurrences of pure atoms only, such that  $\delta$  is equivalent to  $\gamma$  w.r.t. the least  $\mathcal{D}$ -model semantics. This can be done by adding suitable equality constraints. For example, the clause:  $p(X+1) \leftarrow X \leq 0, r(X-1)$  can be transformed into the equivalent clause:  $p(Y) \leftarrow X \leq 0, r(Z), Y = X+1, Z = X-1$ , where all atoms are pure.

We define the set of *useless predicates* of a program  $P$  to be the maximal set  $U$  of predicates occurring in  $P$  such that the predicate  $p$  is in  $U$  iff the body of each clause defining  $p$  in  $P$  contains a positive literal whose predicate is in  $U$ .

We say that the atom  $A$  is *failed* in a program  $P$  iff  $A$  does not unify with the head of any clause in  $P$ . We say that  $A$  is *valid* in a program  $P$  iff  $P$  contains a unit clause whose head has the predicate symbol of  $A$ .

The process of *specializing* a given program  $P$  whereby deriving program  $P_s$ , can be formalized as a sequence  $P_0, \dots, P_n$  of programs, called a *transformation sequence*, where  $P_0 = P$ ,  $P_n = P_s$  and, for  $k = 0, \dots, n-1$ , program  $P_{k+1}$  is obtained from program  $P_k$ , called the current program, by applying one of the transformation rules are listed below. These rules are an extension of the rules presented in [5] to the case of CLP programs with locally stratified negation.

**R1. Constrained Atomic Definition.** Introduce a new predicate defined by a *definition* clause:  $newp(X_1, \dots, X_n) \leftarrow c, A$  where  $c, A$  is a constrained atom.

**R2. Positive Unfolding.** Replace clause  $H \leftarrow c, G_1, A, G_2$  where  $A$  is an atom, by the set of clauses  $\{H \leftarrow c, A = A_j, c_j, G_1, G_j, G_2 \mid j = 1, \dots, m\}$ , where  $\{A_j \leftarrow c_j, G_j \mid j = 1, \dots, m\}$  is the set of *all* renamed apart clauses in the current program such that the atoms  $A$  and  $A_j$  have the same predicate symbol.

**R3. Negative Unfolding.** Let  $\gamma$  be the clause  $H \leftarrow c, G_1, \neg A, G_2$ . If  $A$  is failed in the current program then replace  $\gamma$  by the clause  $H \leftarrow c, G_1, G_2$ . If  $A$  is valid in the current program then remove  $\gamma$ .

**R4. Constrained Atomic Folding.** Replace clause  $A \leftarrow c, G_1, L, G_2$ , by clause  $A \leftarrow c, G_1, L', G_2$ , where  $L'$  is  $newp(X_1, \dots, X_n)$  if  $L$  is  $B$ , or  $\neg newp(X_1, \dots, X_n)$  if  $L$  is  $\neg B$ , provided that there exists a renamed apart definition clause  $newp(X_1, \dots, X_n) \leftarrow d, B$  such that  $\mathcal{D} \models c \rightarrow d$ .

**R5. Removal of Clauses with Unsatisfiable Body.** Remove clause  $A \leftarrow c, G$  if the constraint  $c$  is unsatisfiable.

**R6. Removal of Useless Clauses.** Remove all clauses whose head predicate is useless in the current program.

**R7. Removal of Subsumed Clauses.** If the current program contains a unit clause  $p(X_1, \dots, X_n) \leftarrow$ , then remove all clauses whose head predicate symbol is  $p$ .

**R8. Contextual Constraint Replacement.** Given a set  $\mathbf{C}$  of constrained atoms, replace clause  $p(X_1, \dots, X_n) \leftarrow c_1, G$  by  $p(X_1, \dots, X_n) \leftarrow c_2, G$ , if for

some constraint  $c_2$ , we have that for every constrained atom  $c, p(Y_1, \dots, Y_n)$  in  $\mathbf{C}$ ,  $\mathcal{D} \models (c, X_1 = Y_1, \dots, X_n = Y_n) \rightarrow (c_1 \leftrightarrow c_2)$ , that is, in the constraint domain  $\mathcal{D}$  if  $c, X_1 = Y_1, \dots, X_n = Y_n$  holds then the constraints  $c_1$  and  $c_2$  are equivalent.

### 3.2 The Specialization Strategy

The specialization process is performed according to a strategy which guides the application of the rules R1-R8 above. Our specialization strategy is parameterized by: (i) a function *solve* for solving constraints over the constraint domain  $\mathcal{D}$ , (ii) an *unfolding function* for controlling the unfolding process, (iii) a clause *generalization function* for controlling the introduction of new predicate definitions. Once these parameters have been chosen, our strategy can be applied in a fully automatic way.

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The specialization strategy is divided into three phases.

*Phase A.* We consider the program  $P_S \cup \{f \leftarrow \text{sat}(s_0, \varphi)\}$  and we iterate the procedures *Unfold-Replace* and *Define-Fold* as we now explain. During the *Unfold-Replace* procedure we unfold the program to be specialized by using the given unfolding function, and we solve the constraints in the derived clauses by using the given function *solve*. We then apply the *Define-Fold* procedure and we fold the clauses we have derived. For folding we make use of already available definitions and, possibly, some new definitions introduced by using the clause generalization function. Phase A terminates with output program  $P_A$  when no new definitions need to be introduced for performing the folding steps.

*Phase B.* We consider program  $P_A$  and, by applying the contextual constraint replacement rule, from each clause defining a predicate, say  $p$ , we remove the constraints which hold before the execution of the clause. These constraints are determined by computing the least upper bound of the set of constraints which occur in the clauses containing a call of  $p$  (see [5] for details). Under suitable conditions, that we discuss below, the output of Phase B is a program  $P_B$  which, by construction, admits a finite stratification so that  $P_B = S_1 \cup \dots \cup S_n$ .

*Phase C.* We consider the program  $P_B$  and by working bottom-up on the strata  $S_1, \dots, S_n$ , we simplify the definition of every predicate  $p$  with the aim of deriving either the unit clause  $p(\dots) \leftarrow$  or the empty definition. During this phase we apply the following rules: (i) positive and negative unfolding, (ii) removal of useless and subsumed clauses, and (iii) contextual constraint replacement.

The algorithm for Phase C is as follows.

```

 $P_f := \emptyset$ 
for  $i := 1, \dots, n$  do
  repeat
     $S'_i := S_i$ ;
    Apply to  $S_i$ , as long as possible, the rule for removing subsumed clauses;
    Apply to  $S_i$ , as long as possible, the negative unfolding rule and the positive
    unfolding rule w.r.t. valid and failed atoms in the current program;

```

Replace all clauses in  $S_i$  of the form  $H \leftarrow c$  by the clause  $H \leftarrow$  provided that  $\mathcal{D} \models \forall(\exists Y c)$  where  $Y$  is the set of variables in  $c$  which do not occur in  $H$

**until**  $S' = S_i$   
 Remove the useless clauses of  $S_i$ ;  
 $P_f := P_f \cup S_i$ ;  
**end-for**

Finally, (i) if the unit clause  $f \leftarrow$  belongs to  $P_f$ , we conclude that  $\varphi$  is true in  $S$ , and otherwise, (ii) if there are no clauses in  $P_f$  defining  $f$ , we conclude that  $\varphi$  is false in  $S$ . If neither case (i) nor case (ii) occurs, we cannot conclude that  $\varphi$  is true in  $S$  and we cannot conclude that  $\varphi$  is false in  $S$ .

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Now we illustrate how our specialization strategy works by applying it to program  $P_{S_0}$  described in Section 2. We start by introducing a new definition clause  $\delta$

`property :- X=0, sat(s(a,X), not(ef(neg)))`.

which encodes the property to be verified. The specialization of program  $P_{S_0}$  w.r.t. the atom `property` proceeds as follows. The output of Phase A is the following program  $P_A$ :

`property :- A=0, \+(newsat1(A)).`  
`newsat1(A) :- A=0, newsat2(A).`  
`newsat1(A) :- B=2, A=0, newsat3(B).`  
`newsat3(A) :- A>1, newsat4(A).`  
`newsat3(A) :- B=2+A, A>1, newsat3(B).`

During Phase B we apply the contextual constraint replacement rule and we obtain the following program  $P_B$ :

(1) `property :- A=0, \+ newsat1(A).`  
 (2) `newsat1(A) :- newsat2(A).`  
 (3) `newsat1(A) :- B=2, newsat3(B).`  
 (4) `newsat3(A) :- newsat4(A).`  
 (5) `newsat3(A) :- B=2+A, newsat3(B).`

We start Phase C by computing a minimal stratification of the program  $P_B$ . We get:  $P_B = S_1 \cup S_2$ , where  $S_1 = \{2, 3, 4, 5\}$  and  $S_2 = \{1\}$ . Then, we start processing the lowest stratum  $S_1$ .

*Stratum  $S_1$ .* Since there is no clause defining `newsat2` and `newsat4`, we apply the positive unfolding rule to clauses 2 and 4 and we remove them from  $S_1$ . We get:  $S_1 = \{3, 5\}$ . Predicates `newsat1` and `newsat3` are useless in  $S_1$ , so we remove their definitions and we obtain  $S_1 = \emptyset$ . The iteration on stratum  $S_1$  ends with  $P_f = \emptyset$ .

*Stratum  $S_2$ .* The current program contains no clause defining `newsat1` and this allows us to apply the negative unfolding rule to clause 1 obtaining  $S_2 = \{\text{property} \text{ :- } A=0.\}$ . We solve the constraint in the body of the only clause of  $S_2$  w.r.t. the variables in its head, and we obtain  $S_2 = \{\text{property}.\}$ . The output of Phase C of our specialization strategy is  $P_f = \{\text{property}.\}$  which gives us a proof that `property`  $\neg EF$  `neg` is true in  $S_0$ .

Let us now briefly comment on our three-phase strategy.

The output program  $P_A$  of Phase A of the specialization strategy admits a finite stratification, if we choose a clause generalization function which does not generalize the second argument of *sat* encoding a CTL formula. Indeed, by construction, each clause of  $P_A$  with a negative literal in the body, will be of the form:

$$newp(\dots) \leftarrow \dots, \neg newq(\dots), \dots$$

and the predicates *newp* and *newq* are defined by clauses of the form:

$$\begin{aligned} newp(\dots) &\leftarrow sat(\dots, \varphi_p) \\ newq(\dots) &\leftarrow sat(\dots, \varphi_q) \end{aligned}$$

where: (i)  $\varphi_p$  and  $\varphi_q$  are ground terms encoding CTL formulas, and (ii)  $\varphi_q$  is a proper subformula of  $\varphi_p$ .

Since the contextual constraint replacement rule preserves stratification, also the output program  $P_B$  of Phase B admits a finite stratification.

Our specialization strategy terminates if the following facts hold: (i) the *Unfold-Replace* procedure terminates, and (ii) the set of predicate definitions introduced by the generalization function is finite. Notice that Phases B and C always terminate. In the example presented in the next section we will provide an unfolding function and a generalization function for which Phase A of our specialization strategy terminates.

## 4 Verification of the Bakery Protocol

The bakery protocol describes the behavior of a system consisting of two processes,  $A$  and  $B$ , which run in parallel and try to access a shared resource. The state  $s_A$  of process  $A$  is represented by a pair  $\langle c_A, a \rangle$  where  $c_A$  is an element of the set  $\{think, wait, use\}$  of *control states*, and  $a$  is a counter which takes as value a non negative rational number. The possible transitions for process  $A$  are the following (see also Figure 2):

- (1) if the current state is  $\langle think, a \rangle$ , then the next state is  $\langle wait, b + 1 \rangle$ ,
- (2) if the current state is  $\langle wait, a \rangle$  and, either  $a < b$  or  $b = 0$ , then the next state is  $\langle use, a \rangle$ ,
- (3) if the current state is  $\langle use, a \rangle$ , then the next state is  $\langle think, 0 \rangle$ .

Analogously for process  $B$ , by interchanging  $a$  and  $b$ .

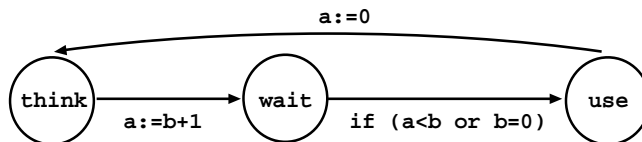


Figure 2. Transitions of the process  $A$ .

The generic state  $s$  of the system consisting of the two processes  $A$  and  $B$  together, is represented by the 4-tuple  $\langle c_A, a, c_B, b \rangle$ , which is initially set to  $s_0 = \langle think, 0, think, 0 \rangle$ .



This system has an infinite number of states, because counters may increase in an unbounded way, as the following sequence of transitions illustrates:

$\langle think, 0, think, 0 \rangle \rightarrow \langle wait, 1, think, 0 \rangle \rightarrow \langle \underline{wait, 1, wait, 2} \rangle \rightarrow \langle use, 1, wait, 2 \rangle \rightarrow \langle think, 0, wait, 2 \rangle \rightarrow \langle think, 0, use, 2 \rangle \rightarrow \langle wait, 3, use, 2 \rangle \rightarrow \langle wait, 3, think, 0 \rangle \rightarrow \langle \underline{wait, 3, wait, 4} \rangle \rightarrow \dots$

We have applied our specialization method to the verification of two properties of the bakery protocol: (i) the mutual exclusion property, and (ii) the starvation freedom property. The mutual exclusion property says that ‘the system will never reach a state where both processes are using the shared resource’. It is a *safety* property in the sense that during the evolution of the system ‘something (bad) may never happen’. The starvation freedom property says that ‘if a process wants to use a resource then it will eventually get it’. It is a *liveness* property in the sense that during the evolution of a system ‘something (good) eventually happens’.

The mutual exclusion property can be expressed by the CTL formula  $s_0 \models \neg EF(unsafe)$ , where *unsafe* is an atomic state property which holds iff both processes are in control state *use*. The starvation freedom property for a process can be expressed by the CTL formula  $s_0 \models AG(wait \rightarrow AF(use))$  which is equivalent to  $s_0 \models \neg EF(wait \wedge \neg AF(use))$ , where *wait* and *use* are atomic state properties which hold iff the process is in control state *wait* and *use*, respectively.

Our two-step verification method works as follows.

Step 1. We introduce a CLP program  $P_{bakery}$ , which is shown in Appendix, such that for the mutual exclusion property we have:  $s_0 \models \neg EF(unsafe)$  iff  $\text{sat}(s_0, \text{not}(\text{ef}(\text{unsafe}))) \in M(P_{bakery})$ , and for the starvation freedom property we have:  $s_0 \models \neg EF(wait \wedge \neg AF(use))$  iff  $\text{sat}(s_0, \text{not}(\text{ef}(\text{and}(\text{wait}, \text{not}(\text{af}(\text{use})))))) \in M(P_{bakery})$ .

Step 2. We choose the parameters of our strategy as follows. (i) The *solve* function is the  $\text{clp}(q,r)$  solver of [6] for simplifying conjunctions of linear equations and inequations over real numbers. (ii) The unfolding function takes a definition of the form  $\text{newp}(X_1, \dots, X_n) \leftarrow c, A$  and it unfolds it w.r.t.  $A$ . Then it unfolds the derived clauses w.r.t. the atoms of the form  $\text{sat}(S, F)$ , where either  $F$  is an atomic state property or the outermost connective of  $F$  is one of the following: *and*, *or*, *not*. (iii) The generalization function takes a clause  $H \leftarrow c, A$  and it generalizes it to the clause  $H \leftarrow d, A$ , where the constraint  $d$  is defined as follows starting from the constraint  $c$ . Let  $\mathcal{L}(c)$  be the set of constraints whose generic element  $r$  is defined as follows:

$$r ::= \text{true} \mid t = t \mid t > t \mid t \geq t \mid r \wedge r$$

where  $t \in \text{vars}(c) \cup \{0\}$ . The constraint  $d$  is the least constraint of  $\mathcal{L}(c)$  (in the implication ordering) which is entailed by  $c$ . Notice that  $\text{vars}(d) \subseteq \text{vars}(c)$ .

For instance, the generalization of the clause:

`new1(A,B) :- A=1, B=2, sat(s(wait,A,wait,B), ef(unsafe))`

is the clause:

`new2(A,B) :- A>0, B>A, sat(s(wait,A,wait,B), ef(unsafe)).`

The definition of the set  $\mathcal{L}(c)$  of constraints used by the clause generalization function is based upon the transitions which can be made by the system. This choice can easily be automated because this set is constructed by using the constants and the constraint predicate symbols occurring in the transitions.

We have verified the mutual exclusion and the starvation freedom properties by specializing  $P_{bakery}$  w.r.t. the atoms `safe` and `starvfree`, respectively, which are defined as follows:

```
safe :- A=0, B=0,
        sat(s(think,A,think,B), not(ef(unsafe))).
starvfree :- A=0, B=0,
            sat(s(think,A,think,B), not(ef(and(wait,not(af(use)))))).
```

The whole verification process was performed automatically by using the MAP transformation system (available at <http://www.iasi.rm.cnr.it/~fioravan>).

## 5 Related Work and Conclusions

During the last years many logic-based techniques have been developed for automatically verifying properties of systems, the most successful of them being model checking [2]. The success of model checking is also due to the use of Binary Decision Diagrams which provide a very compact symbolic representation of a possibly very large, but finite, set of states. In order to overcome this finiteness restriction, some efforts have recently been devoted for the incorporation into model checking of abstraction and deduction techniques [15].

Recent papers also demonstrate the usefulness of logic programming and constraint logic programming as a basis for the verification of finite or infinite state systems. In particular, in [13] the authors present XMC, a model checking system implemented in the tabulation-based logic programming language XSB. XMC can verify  $\mu$ -calculus properties of finite state transitions systems expressed in a CCS-like language, with performances comparable to that of state-of-the-art model checkers. In [3] model checking using constraint logic programming is described and some CTL operators of infinite state concurrent systems are expressed in terms of least and greatest fixed points of the CLP programs used to describe the systems. In [10] the authors define a method based on partial deduction of logic programs, augmented with abstract interpretation, for solving coverability problems of infinite state Petri nets. In [12] a finite state local model checker is defined for CTL, by using CLP with finite domains, extended with constructive negation and tabled resolution.

This paper presents some preliminary results obtained when applying to the verification of infinite state systems an extension of the techniques developed in [5] for specializing constraint logic programs. We performed some experiments on a simplified version of the bakery algorithm [9] which is a protocol for mutual exclusion between two processes. We proved that this protocol ensures both mutual exclusion and starvation freedom.

We believe that the integration of CLP as modeling language, and program specialization as inference system, can provide a very flexible tool for the verification of infinite state systems. Indeed, constraints provide a natural representation for infinite sets of values (e.g.,  $X \geq 0$  describes the infinite set of non-negative real numbers), and the declarativeness of logic programming makes it easy to model a large variety of systems and properties.

Future work on the application of specialization of CLP program for the verification of infinite state systems will include: experimentation with different choices of constraint domains, unfolding functions, and generalization operators. We will also plan to experiment with different classes of systems and properties.

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## Appendix

The CLP program  $P_{bakery}$ .

```
sat(s(u,A,u,B),unsafe).
sat(s(w,A,B,C),wait).
sat(s(u,A,B,C),use).

sat(A,or(B,C)) :- sat(A,B).
sat(A,or(B,C)) :- sat(A,C).
sat(A,and(B,C)) :- sat(A,B), sat(A,C).
sat(A,not(B)) :- \+ sat(A,B).

sat(A,ef(B)) :- sat(A,B).
sat(s(t,A,S,B),ef(C)) :- D=B+1, A>=0, B>=0, sat(s(w,D,S,B),ef(C)).
sat(s(w,A,S,B),ef(C)) :- A<B, A>=0, sat(s(u,A,S,B),ef(C)).
sat(s(w,A,S,B),ef(C)) :- B=0, A>=0, sat(s(u,A,S,B),ef(C)).
sat(s(u,A,S,B),ef(C)) :- D=0, A>=0, B>=0, sat(s(t,D,S,B),ef(C)).
sat(s(S,A,t,B),ef(C)) :- D=A+1, A>=0, sat(s(S,A,w,D),ef(C)).
sat(s(S,A,w,B),ef(C)) :- B<A, B>=0, sat(s(S,A,u,B),ef(C)).
sat(s(S,A,w,B),ef(C)) :- A=0, B>=0, sat(s(S,A,u,B),ef(C)).
sat(s(S,A,u,B),ef(C)) :- D=0, B>=0, A>=0, sat(s(S,A,t,D),ef(C)).

sat(A,af(B)) :- sat(A,B).
sat(s(t,T1,t,T2),af(P)) :- T3=T2+1, T4=T1+1,
    sat(s(w,T3,t,T2),af(P)), sat(s(t,T1,w,T4),af(P)).
sat(s(t,T1,u,T2),af(P)) :- T3=T2+1, T4=0,
    sat(s(w,T3,u,T2),af(P)), sat(s(t,T1,t,T4),af(P)).
sat(s(t,T1,w,T2),af(P)) :- T3=T2+1, T2<T1,
    sat(s(w,T3,w,T2),af(P)), sat(s(t,T1,u,T2),af(P)).
sat(s(t,T1,w,T2),af(P)) :- T3=T2+1, T1=0,
    sat(s(w,T3,w,T2),af(P)), sat(s(t,T1,u,T2),af(P)).
sat(s(t,T1,w,T2),af(P)) :- T3=T2+1, T1>0, T1<T2,
    sat(s(w,T3,w,T2),af(P)).
sat(s(u,T1,u,T2),af(P)) :- T3=0, T4=0,
    sat(s(t,T3,u,T2),af(P)), sat(s(u,T1,t,T4),af(P)).
sat(s(u,T1,w,T2),af(P)) :- T3=0, T2<T1,
    sat(s(t,T3,w,T2),af(P)), sat(s(u,T1,u,T2),af(P)).
sat(s(u,T1,w,T2),af(P)) :- T3=0, T1=0,
    sat(s(t,T3,w,T2),af(P)), sat(s(u,T1,u,T2),af(P)).
sat(s(u,T1,w,T2),af(P)) :- T3=0, T1>0, T1<T2,
    sat(s(t,T3,w,T2),af(P)).
sat(s(w,T1,w,T2),af(P)) :- T1<T2, T1=0,
    sat(s(u,T1,w,T2),af(P)), sat(s(w,T1,u,T2),af(P)).
sat(s(w,T1,w,T2),af(P)) :- T1<T2, T1>0,
    sat(s(u,T1,w,T2),af(P)).
sat(s(w,T1,w,T2),af(P)) :- T2=0, T1=0,
    sat(s(u,T1,w,T2),af(P)), sat(s(w,T1,u,T2),af(P)).
sat(s(u,T2,t,T1),af(P)) :- T3=T2+1, T4=0,
    sat(s(u,T2,w,T3),af(P)), sat(s(t,T4,t,T1),af(P)).
```

```

sat(s(w,T2,t,T1),af(P)) :- T3=T2+1, T2<T1,
    sat(s(w,T2,w,T3),af(P)), sat(s(u,T2,t,T1),af(P)).
sat(s(w,T2,t,T1),af(P)) :- T3=T2+1, T1=0,
    sat(s(w,T2,w,T3),af(P)), sat(s(u,T2,t,T1),af(P)).
sat(s(w,T2,t,T1),af(P)) :- T3=T2+1, T1>0, T1=<T2,
    sat(s(w,T2,w,T3),af(P)).
sat(s(u,T2,u,T1),af(P)) :- T3=0, T4=0,
    sat(s(u,T2,t,T3),af(P)), sat(s(t,T4,u,T1),af(P)).
sat(s(w,T2,u,T1),af(P)) :- T3=0, T2<T1,
    sat(s(w,T2,t,T3),af(P)), sat(s(u,T2,u,T1),af(P)).
sat(s(w,T2,u,T1),af(P)) :- T3=0, T1=0,
    sat(s(w,T2,t,T3),af(P)), sat(s(u,T2,u,T1),af(P)).
sat(s(w,T2,u,T1),af(P)) :- T3=0, T1>0, T1=<T2,
    sat(s(w,T2,t,T3),af(P)).
sat(s(w,T2,w,T1),af(P)) :- T1<T2, T1=0,
    sat(s(w,T2,u,T1),af(P)),
    sat(s(u,T2,w,T1),af(P)).
sat(s(w,T2,w,T1),af(P)) :- T1<T2, T1>0,
    sat(s(w,T2,u,T1),af(P)).
sat(s(w,T2,w,T1),af(P)) :- T2=0, T1=0,
    sat(s(w,T2,u,T1),af(P)), sat(s(u,T2,w,T1),af(P)).

```