Removing Unnecessary Variables from Horn Clause Verification Conditions

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Verification conditions (VCs) are logical formulas whose satisfiability guarantees program correctness. We consider VCs in the form of constrained Horn clauses (CHC) which are automatically generated from the encoding of (an interpreter of) the operational semantics of the programming language. VCs are derived through program specialization based on the unfold/fold transformation rules and, as it often happens when specializing interpreters, they contain unnecessary variables, that is, variables which are not instrumental to the correctness proofs of the programs under verification. In this paper we adapt to the CHC setting some of the techniques that were developed for removing unnecessary variables from logic programs, and we show that, in some cases, the application of these techniques increases the effectiveness of Horn clause solvers when proving program correctness.

1 Introduction

Correctness of an imperative program \( P \) can be verified by: first, (i) generating verification conditions (VCs, for short) for the program \( P \) and the considered property, and then, (ii) using SMT solvers for checking the satisfiability of the VCs.

In this paper we consider VCs which are automatically derived by applying program specialization to a constrained Horn clause encoding of the operational semantics of the programming language. (In this paper we will use the notions of Constrained Horn Clauses (CHC) and constraint logic programs (CLP) interchangeably.) Program specialization is based on the application of semantics preserving unfold/fold transformation rules, guided by a strategy, called the VCG strategy, which has been specifically designed for VCs generation (see [3] for a detailed presentation). Other notable applications of CLP program specialization to the analysis of imperative or object-oriented programs can be found in [1, 9].

Given an imperative program \( P \) and a safety property, we introduce a CLP program \( I \), which defines the nullary predicate unsafe such that \( P \) is safe if and only if the atom unsafe is not derivable from \( I \) or, equivalently, unsafe does not belong to the least model of \( I \), denoted \( \mathcal{M}(I) \).

The VCG strategy works by performing the so-called removal of the interpreter, that is, it removes the level of interpretation which is present in the initial CLP program \( I \), where commands are encoded as CLP clauses and there are references to the operational semantics of the imperative programming language. The output of the VCG strategy is a program \( I_{sp} \) such that unsafe \( \notin \mathcal{M}(I) \) iff unsafe \( \notin \mathcal{M}(I_{sp}) \). Moreover, due to the absence of the interpretative level, the test of whether or not unsafe belongs to \( \mathcal{M}(I_{sp}) \) is often easier than the test of whether or not unsafe belongs to \( \mathcal{M}(I) \).
The specialization-based approach for generating VCs is parametric with respect to: (i) the imperative program $P$, (ii) the operational semantics of the imperative language in which the program $P$ is written, (iii) the property to be proved, and (iv) the logic used for specifying the property of interest (in this case, the reachability of an unsafe state).

One of the most significant advantages of this approach is that it enables the design of widely applicable VC generators for programs written in different programming languages, and for different operational semantics of languages with the same syntax, by making small modifications only [3].

2 Removing Unnecessary Variables

It is well known that program specialization and transformation techniques often produce clauses with more arguments than those that are actually needed [7, 10, 6]. Thus, it is not surprising to observe that such a side-effect also occurs when generating VCs via program specialization. Indeed, it is often the case that some of the variables occurring in the CLP program $I_p$, which is generated by the VCG strategy, are not actually needed to check whether or not $\text{unsafe} \in M(I_p)$. Avoiding those unnecessary variables, and thus deriving predicates with smaller arity, can increase the effectiveness and the efficiency of applying Horn clause solvers, and proving program correctness.

Now we present two transformation techniques which allow us to reduce the number of arguments of the predicates used in the VCs. These techniques extend to the case of CLP programs similar techniques that have been developed for logic programs [7, 10]. The first technique is a transformation strategy, called the Non-Linking variable Removal strategy (or the NLR strategy, for short) that removes variables occurring as arguments of an atom in the body of a clause, but that do not occur elsewhere in the clause. The second technique, called the constrained FAR algorithm (or the cFAR algorithm, for short), is a generalization of a liveness analysis, and removes arguments that are not actually used during program execution.

1. Non-Linking variable Removal Strategy. First we consider the NLR strategy, whose objective is to remove the non-linking variables, which are defined as follows.

**Definition 1 (Linking variables [10])** Let $C$ be the clause $H :- c, L, B, R$, where: $c$ is a constraint, $L$ and $R$ are (possibly empty) conjunctions of atoms, and $B$ is an atom. The set of linking variables of $B$ in $C$, denoted by $\text{linkvars}(B, C)$, is $\text{vars}(B) \cap \text{vars}(\{H, c, L, R\})$. The set of non-linking variables of $B$ in $C$ is $\text{vars}(B) - \text{linkvars}(B, C)$.

Before presenting the NLR strategy, we see it in action in an example. Let us consider the C program $P$ in Figure [1]. We want to verify the Hoare triple $\{x \geq 0\} P \{y \leq 0\}$. By applying the VCG strategy, we get the set of clauses $P_1$ in Figure [1] where $\text{unsafe}$ holds iff the Hoare triple is not valid. In $P_1$ the non-linking variables have been underlined. Then, by applying the NLR strategy, we get the set of clauses $P_2$ without non-linking variables (see Figure [2]). $P_1$ and $P_2$ are equivalent with respect to the query $\text{unsafe}$, in the sense that $\text{unsafe} \in M(P_1)$ iff $\text{unsafe} \in M(P_2)$.

In particular, NLR replaces the predicates $\text{newp}_1$ and $\text{newp}_2$, which are called with the non-linking variables $X_2$, $Y_1$, and $Y_2$ (see clauses [1] and [2] of $P_1$ in Figure [1]), with two new predicates $\text{newp}_3$ and $\text{newp}_4$, respectively, whose arguments are linking variables only. Note that the removal of the two arguments $Y_1$ and $X_2$ of $\text{newp}_1$, which are the non-linking variables in clause [1], determines in clause [2] the removal of the two arguments $Y_1$ and $X_2$, which are linking variables of $\text{newp}_2$. Thus, from $\text{newp}_2$ with six arguments in clause [2] by removing also the non-linking variable $Y_2$, we get the predicate $\text{newp}_4$ in clauses [3] and [4] of program $P_2$ with three arguments only (see Figure [2]).
The C program

```c
int x, y;
void main() {
    int z = x + 1;
    while (z <= 9) {
        z = z + 1;
    }
    y = z;
}
```

The C program $P$ is the set of Verification Conditions VCs obtained by applying the VCG strategy starting from the C program $P$, the initial condition $x \geq 0$ and the error property $y \leq 0$.

The NLR strategy consists in a repeated application of the unfolding, definition introduction and folding transformation rules [4].

We assume that the input of NLR is any CLP program $Prog$. To keep the notation simple, we will identify a tuple of variables with the set of variables occurring in it. The union of two tuples is constructed by erasing duplicate elements.

During the execution the NLR strategy maintains in a set $Defs$ all the definitions that have been introduced so far. Every definition clause in $Defs$ is unfolded with respect to the leftmost atom in its body, thereby producing a set of clauses $S$. Then every clause in $S$ is folded (repeatedly, with respect all atoms in its body) by using either definitions that already occur in $Defs$ or new definitions that are introduced in $Defs$ for performing those folding steps.

The peculiarity of the NLR strategy lies in the careful management of the set of variables occurring in the head of the definition clauses.

Let $C$ be a clause in $S$ of the form: $H : - c, L, B, R$, where the predicate symbol of $B$ occurs in $Prog$. If $C$ cannot be folded w.r.t. the atom $B$ using any clause in $Defs$, then we have to introduce a new definition clause as explained below. First, we consider a definition whose head contains only the linking variables of the atom $B$ in the clause $C$.

Let $F$ be $\text{newp}(P) : - B$, where $\text{newp}$ is a predicate symbol not occurring in the set $Prog \cupDefs$, and $P$ is the set linkvars($B, C$) of the linking variables of $B$ in $C$.

If the set $Defs$ contains a clause $D$ of the form $\text{newq}(Q) : - S$ such that, for some renaming substitution $\vartheta$, $B\vartheta = S$, then we replace the clause $D$ in $Defs$ with the clause $\text{newp}(L) : - B$, where $L = P \vartheta \cup Q$. Otherwise, we introduce the definition clause $F$ and we add it to $Defs$.

The introduction of the definition $F$ might seem to be the best choice in the sense that it contains exactly the head variables which are actually needed for folding the clause $C$. However, (variants of) $B$ can also occur in some other clauses to be folded. Thus, if we directly introduce definitions whose heads contain linking variables only, we run the risk of generating several definitions with the same atom in the body and different sets of variables in the head (modulo renaming).

In order to keep the number of definitions low (and this will often improve the ability of proving program correctness), instead of introducing multiple definitions containing the same atom in the body, by applying the NLR strategy, we merge them in a single definition whose set of head variables is the union of the head variables occurring in the merged definitions (modulo renaming).

The NLR strategy terminates when all clauses in $Defs$ have been unfolded and no new definition need to be introduced for folding.
1'. unsafe:- X1>=0, Y2=<0, newp3(X1,Y2).
2'. newp3(X1,Z2):- Z1=X1+1, newp4(X1,Z1,Z2).
3'. newp4(X1,Z1,Z2):- Z1=<9, Z3=Z1+1, newp4(X1,Z3,Z2).
4'. newp4(X1,Z1,Z1):- Z1>=10.

P2: Verification Conditions obtained by NLR

P3: Verification Conditions obtained by cFAR

Figure 2: Program P2 and Program P3 are the Verification Conditions VCs obtained by applying the NLR strategy and the cFAR algorithm, respectively.

Theorem 1 (Termination and Correctness of the NLR Strategy) Given any CLP program Prog, the NLR strategy terminates and produces a CLP program Prog’ such that unsafe ∈ $\mathcal{M}(Prog)$ holds iff unsafe ∈ $\mathcal{M}(Prog’)$ holds.

2. Constrained FAR Algorithm (cFAR). Now we present an extension to constraint logic programs of the FAR algorithm presented in [7] for removing redundant arguments from logic programs. The objective of the FAR algorithm is to remove arguments that are not actually used during every computation of the program at hand. Indeed, it has been shown in [5] that the FAR algorithm can be seen as a generalization of the liveness analysis.

In Figure 2 we show the effect of applying the cFAR algorithm to the CLP program P2 obtained by the NLR strategy. The output of the algorithm is the CLP program P3. Note that in program P3 the predicate symbol newp4 denotes a different relation with respect to the one in program P2, because in P3 it has arity 2 and not 3.

In order to define the constrained FAR algorithm we need to introduce some preliminary notions, some of which have been adapted from [7].

Definition 2 (Erasure, Erased Atom, Erased Clause, Erased Program) (i) An erasure is a set of pairs each of which is of the form (p,k), where p is a predicate symbol of arity n and 1 ≤ k ≤ n.
(ii) Given an erasure E and an atom A whose predicate symbol is p, the erased atom A|E is obtained by dropping all the arguments that occur at position k, for some (p,k) ∈ E.
(iii) Given an erasure E and a clause C (respectively, a CLP program Prog), the erased clause C|E (respectively, the erased program Prog|E) is obtained by replacing all atoms A in C (respectively, in Prog) by A|E.

In order to avoid the risk of collisions between predicate symbols after erasing some arguments, we assume that Prog does not contain identical predicate symbols with different arity.

Obviously, we are interested in removing redundant arguments without altering the semantics of the original program, in the sense captured by the following definition.

Definition 3 (Correctness of Erasure) An erasure E is correct for a program Prog if, for all atoms A we have that: A ∈ $\mathcal{M}(Prog)$ iff A|E ∈ $\mathcal{M}(Prog|E)$.

Since we are dealing with constraint logic programs, the notion of multiple occurrences of a variable which is used in the original formulation of FAR [7], needs to be generalized as follows. In this paper we assume that a constraint is a conjunction of $h$ ($\geq 0$) atomic constraints in the theory $\mathcal{A}$ of linear integer arithmetics and integer arrays.
Definition 4 (Variable Constrained to Another Variable) Given two variables \( X \) and \( Y \) and a constraint \( c \) of the form \( c_1 \land \ldots \land c_h \), we say that \( X \) is constrained to \( Y \) (in \( c \)) if there exists \( c_j \), with \( 1 \leq j \leq h \), such that either (i) \( \{X,Y\} \subseteq \text{vars}(c_j) \), or (ii) there exists a variable \( Z \) such that (ii.1) in \( \{X,Z\} \subseteq \text{vars}(c_j) \) and (ii.2) \( Z \) is constrained to \( Y \) (in \( c \)).

We are now ready to introduce the notion of safe erasure that will be used during the application of the constrained FAR algorithm.

Definition 5 (Safe Erasure) Given a program \( \text{Prog} \), an erasure \( E \) is safe if, for all \( (p,k) \in E \) and clauses \( H : \neg c, G \) in \( \text{Prog} \), where \( H \) is of the form \( p(X_1,\ldots,X_n) \) and \( c \) is of the form \( c_1 \land \ldots \land c_h \), we have that: (i) \( X_k \) is a variable and \( \mathcal{A} \models \forall X_k \exists Y_1,\ldots,Y_m. c \), with \( \{Y_1,\ldots,Y_m\} = \text{vars}(c) - \{X_k\} \), (ii) \( X_k \) is not constrained to any other variable occurring in \( H \), and (iii) \( X_k \) is not constrained to any variable occurring in \( G \mid E \).

Similarly to what has been done in [7], it can be shown that if an erasure \( E \) is safe then it is also correct.

The cFAR algorithm takes as input a CLP program \( \text{Prog} \), computes a correct erasure \( E \), and produces as output the program \( \text{Prog} \mid E \). The algorithm starts by initializing the current erasure \( E \) to the full erasure, that is, the set of all erasures \( (p,k) \) where \( p \) is a predicate of arity \( n \) occurring in \( \text{Prog} \) and \( 1 \leq k \leq n \). Then, the algorithm proceeds by removing from \( E \) all erasures that are not safe, until a fixpoint is reached.

The cFAR algorithm terminates and preserves the least-model semantics, as stated by the following theorem.

Theorem 2 (Termination and Correctness of the cFAR Algorithm) Given any CLP program \( \text{Prog} \), the cFAR algorithm terminates and produces a CLP program \( \text{Prog} \mid E \) such that \( \text{unsafe} \in \mathcal{M}(\text{Prog}) \) iff \( \text{unsafe} \in \mathcal{M}(\text{Prog} \mid E) \).

Finally, we would like to note that, even if the objectives of the NLR and cFAR transformations are similar, they work in a different way. While NLR starts from the predicate unsafe and proceeds by unfolding in a goal directed fashion, cFAR is goal independent. It can be shown that, in general, the two transformations have incomparable effects.

3 Experimental evaluation

We have used the VeriMAP transformation and verification system [2] for evaluating the techniques presented in this paper. We have considered 320 verification problems for C programs (227 of which are safe and the remaining 93 are unsafe). We have applied the VCG strategy for generating the Verification Conditions VCs using a multi-step semantics [3]. The C programs and the VCs we have generated are available at: http://map.uniroma2.it/vcgen. Then, we have checked the satisfiability of the VCs by giving them as input to the Z3 Horn solver [8] (with the PDR engine). Finally, we have applied the NLR and cFAR transformations presented in Section 2 to evaluate the effect of these transformations in terms of efficiency and efficacy.

Improving effectiveness of solving. In Table 1 we show the experimental results obtained by using VeriMAP and Z3. Column ‘VCG; Z3’ reports the results obtained by applying the VCG strategy and then the Z3 solver. Column ‘VCG; NLR; Z3’ reports the results obtained, for the problems not solved by ‘VCG; Z3’, by applying VCG, followed by the NLR transformation, and then Z3. Column ‘VCG; NLR; cFAR; Z3’ reports the results obtained, for the problems not solved by ‘VCG; NLR; Z3’, by
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Table 1: Verification results obtained by using Z3 on the output generated by applying VCG and the auxiliary transformations NLR and cFAR. The timeout limit time is five minutes. Times are in seconds.

Applying VCG, followed by NLR, then cFAR, and finally Z3. Lines \(t_{\text{VCG}}\), \(t_{\text{NLR}}\), and \(t_{\text{cFAR}}\) report the time taken by the execution of the VCG, NLR, and cFAR transformations, respectively, to produce the verification conditions for which Z3 was able to return the correct answers (that is, to show the satisfiability or the unsatisfiability of the clauses). Line \(st\) reports the time taken by Z3 to produce the correct answers.

The NLR transformation enables Z3 to prove 7 additional verification problems. In particular, it allows Z3 to prove the program `ntdrvsimpl-cdaudio_simpl1_unsafe1.c`, which is the biggest program in the benchmark set (2.1 KLOC). Concerning the time required for executing the NLR transformation in this example, we want to point out that this program takes 91% of the total NLR time \(t_{\text{NLR}}\), that is 53.04 seconds. Therefore, the remaining 6 programs only require 5.35 seconds to be transformed. The cFAR transformation allows Z3 to prove 9 additional verification problems. In this case, about 89% of the total cFAR time \(t_{\text{cFAR}}\), that is, 271.62 seconds, is required for specializing two programs, namely `ntdrvsimpl-diskperf_simpl1_safeil.c` (98.82 seconds) and `ntdrvsimpl-floppy_simpl3_safeil.c` (172.80 seconds) whose size is about 1 KLOC each.

4 Conclusions

In this paper we have shown that the effectiveness of Horn clause solvers at proving the satisfiability of VCs can be improved by the use of program transformations that remove unnecessary variables.

As future work, we would like to investigate in more depth how the structure of the VCs influences the heuristics adopted by Horn solvers. Hopefully, this would allow us to tune the VCG strategy for generating VCs that are easier to be proved.

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References


