Model Generation with Boolean Constraints

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Model Generation with Boolean Constraints

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Abstract: We present a simple method for eliminating redundant searches in model generation. The method employs Boolean Constraints which are conjunctions of ground instances of clauses having participated in proofs. Boolean Constraints work as sets of lemmas with which duplicate subproofs and irrelevant model extensions can be eliminated. The method has been tentatively implemented on a constraint logic programming system. We evaluated effects of the method by proving some typical problems taken from the TPTP problem library.

Keywords: Theorem proving, Constraint solver, Folding-up, Proof condensation

1. Introduction

The model generation procedure tries to construct Herbrand models for a given clause set and determines its satisfiability. It maintains a set $M$ of ground atoms called a model candidate, finds violated clauses that are not satisfied under $M$, then extends $M$ to satisfy them, and repeats this process until a model is found or all model candidates are rejected.

There are two types of redundancy in model generation: One is that the same subproof tree may be generated at several descendant nodes after a case-splitting occurs. Another is caused by unnecessary model candidate extensions with irrelevant clauses. We embedded both folding-up 3) and proof condensation 5) into model generation for eliminating these redundancies by analyzing dependency in a proof 2). The embedded function examines the structure of proof in order to append a solved subproof-tree to an open branch.

This paper presents yet another method to eliminate the redundancies on the basis of semantical information. If the current model candidate conflicts with a set of instances of clauses that have participated in model generation so far, we can reject the model candidate without further exploration. We call the set a Boolean Constraint. It is worth noting that the Boolean Constraint consists of only ground instances of clauses and all atoms in model candidates are ground. Therefore, a conflict test is essentially propositional theorem proving.

In this work, we utilize a constraint solver 7) on Boolean expressions for the test, though we could utilize model generation itself or other proving methods in principle. The main reason for utilizing the constraint solver is that it can compute a simple (canonical) form of the Boolean Constraint which is incrementally updated as the proof progresses. Since the constraint solver reduces the Boolean Constraint as simple as possible, it can detect the conflict efficiently.

2. Model Generation

Throughout this paper, a clause $\neg A_1 \lor \cdots \lor \neg A_n \lor B_1 \lor \cdots \lor B_m$ is represented in implicational form: $A_1 \land \cdots \land A_n \rightarrow B_1 \lor \cdots \lor B_m$, where $A_i$ (1 $\leq$ $i$ $\leq$ $n$) and $B_j$ (1 $\leq$ $j$ $\leq$ $m$) are atoms; the left hand side of “$\rightarrow$” is said to be the antecedent; and the right hand side of “$\rightarrow$” the consequent.

A clause is said to be positive if its antecedent is $\top$ ($n = 0$), and negative if its consequent is $\bot$ ($m = 0$); otherwise it is mixed ($n \neq 0, m \neq 0$). A clause is said to be violated under a set $M$ of ground atoms if the following condition holds with some ground substitution $\sigma$: $\forall i(1 \leq i \leq n)A_i \sigma \in M \land \forall j(1 \leq j \leq m)B_j \sigma \not\in M$.

A model generation proof procedure is sketched in Fig. 1. Given a set $S$ of clauses, $MG$ tries to construct a model by extending the current model candidate $M$ so as to satisfy violated clauses under $M$ (model extension). When a negative clause is violated under $M$, $MG$ rejects $M$ because there is no way of extending $M$ (model rejection). If no clause is violated under $M$, we conclude $M$ is a model of $S$, that is, $S$ is satisfiable (model finding).

Consider the following set of clauses $S1$:

\begin{align*}
C1 : & T \rightarrow p(a) \lor p(c) & C2 : & p(a) \rightarrow q(b) \\
C3 : & p(X) \land q(Y) \rightarrow r(X, Y) \lor r(X, X) \lor r(Y, X) & C4 : & p(X) \land q(Y) \rightarrow r(s(X), Y) \lor r(X, X) \lor r(Y, X) \\
C5 : & p(X) \land r(s(X), Y) \rightarrow r(Y, s(X))
\end{align*}
procedure \textsc{MGTP} (S) : Res; /* Input(S):Clause set, Output(Res):satisfiability of S */
return(\textsc{MG}(\emptyset));

procedure \textsc{MG} (M) : Res; /* Input(M): Model candidate */
1. (Model rejection) If a negative clause \( A_1 \land \ldots \land A_n \rightarrow \bot \in S \) is violated under \( M \) with a ground substitution \( \sigma \), \textbf{return unsatisfiable};
2. (Model extension) If a positive or mixed clause \( A_1 \land \ldots \land A_n \rightarrow B_1 \lor \ldots \lor B_m \in S \) is violated under \( M \) with a ground substitution \( \sigma \),
   \textbf{for} \( (i = 1; i \leq m; i++) \) \{
   if \((\textsc{MG}(M \cup \{ B_i \sigma \}) = \text{satisfiable}) \textbf{return satisfiable};
   \}
3. (Model finding) If neither 1 nor 2 is applicable, \textbf{return satisfiable};

\textbf{Fig.1} Model generation procedure.

\begin{align*}
   C6 : & \quad r(s(X),Y) \land r(Y,s(X)) \rightarrow r(X,X) \\
   C7 : & \quad r(X,X) \rightarrow r(s(X),X) \lor r(X,s(X)) \\
   C8 : & \quad p(X) \land q(Y) \land r(Y,X) \rightarrow r(X,X) \\
   C9 : & \quad r(s(X),X) \rightarrow \bot \\
   C10 : & \quad r(X,s(X)) \rightarrow \bot \\
   C11 : & \quad p(c) \rightarrow \bot
\end{align*}

\textbf{Figure} 2 shows a proof-tree for \( S_1 \). The inner nodes of a proof-tree except the root node are labeled with atoms used for model extension. A branch or a path from the root to a node corresponds to a model candidate. A leaf labeled with \( \bot \) indicates that the corresponding model candidate has been rejected. \( S_1 \) is unsatisfiable because all leaves of its proof-tree are labeled with \( \bot \).

The procedure \textsc{MG} in \textbf{Fig. 1} can be proved sound and complete in the sense that \textsc{MG} examines only models containing the model candidate \( M^4 \).

\textbf{Theorem 1} Let \( S \) be a set of clauses and \( M \) be a set of ground atoms. Then \textsc{MG}(M) return unsatisfiable if and only if there is no model containing \( M \).

Let \( BC \) be a set of ground instances of violated clauses in \( S \) that have been used for model rejection and extension. If \( BC \cup M \) is unsatisfiable, \( S \cup M \) is unsatisfiable. In this case, according to Theorem 1, we can reject \( M \) without further proving. This rejection mechanism can reduce search spaces by orders of magnitude. \textbf{Figure 3} shows a model generation procedure in which the rejection mechanism is embedded. The framed parts are embedded ones. We call the procedure \textit{model generation with Boolean Constraints} because \( BC \) is essentially the conjunction of propositional clauses and can be treated as a Boolean expression.

Initially, the set \( BC \) is set to the empty set ((1)). \( BC \) is updated whenever ground instances of clauses are used for model extension or model rejection ((3),(4)). \( BC \) is used for model rejection prior to performing normal model rejection and extension ((2)). This rejection works as folding-up to eliminate duplicates subproofs. \( BC \) is also used for model rejection testing whenever each extension \( \textsc{MG}(M \cup \{ B_i \sigma \}) \) is finished ((5)). This rejection test works as proof condensation to avoid unnecessary model extensions.

\textbf{Figure} 4 shows a proof tree for \( S_1 \) obtained by model generation with Boolean Constraints. The mark * indicates a branch pruned by operation (2), while the mark \( \times \) indicates that by operation (5).

\( BC \) becomes \( BC_1 = \{ C1, C2, C3_1, C4_1, C5_1, C6_1, C7_2, C9_2, C10_2 \} \) after the second branch from the left has been rejected where \( \sigma_1 = \{ X \leftarrow a, Y \leftarrow b \} \) and \( \sigma_2 = \{ X \leftarrow a \} \). Then, the next model candidate \( M_1 \) to be solved is \{\( p(a), q(b), r(a,b), r(a,a) \). However, since \( BC_1 \cup M_1 \) is unsatisfiable, \( M_1 \) is rejected. After the model extension under \( r(b,a) \) with clause \( C8_2 \).
has been performed, $BC$ becomes $BC_2 = BC_1 \cup \{C8a2\}$. The corresponding model candidate $M_2$ is $\{p(a), q(b), r(a, b), r(b, a), r(a, a)\}$. In this case, $BC_2 \cup M_2$ is unsatisfiable as well, so that $M_2$ is rejected.

On the other hand, $BC_2 \cup \{p(a), q(b)\}$, that is, $BC_2 \cup M_3$ is unsatisfiable. Therefore, the exploration of $r(a, a)$ and $r(b, a)$ below $q(b)$ can be eliminated. Thus, we obtain a proof-tree which has 12 inner nodes while the normal proof-tree shown in Fig. 2 has 23 inner nodes.

### 3. Implementation

The method is implemented on top of a constraint logic programming system B-Prolog\(^7\) which supports constraint solvers over trees, Boolean, finite-domains and sets. We manipulate a set $BC$ of ground instances of clauses through the constraint solver. When updating $BC$ (Fig. 3(3) (4)), we tell $(-A_1\sigma \lor \ldots \lor -A_n\sigma) = TRUE$ or $(-A_1\sigma \lor \ldots \lor -A_n\sigma \lor B_1\sigma \lor \ldots \lor B_m\sigma) = TRUE$ to the
constraint solver. On the other hand, when testing whether a conflict occurs (Fig. 3 (2) (5)), we ask the constraint solver “Is $A = \text{TRUE}$ possible for all $A \in M$?” If they become all TRUE, $BC \cup M$ is satisfiable, otherwise, it is unsatisfiable.

Table 1 compares the proving performance on several typical problems taken from the TPTP library. The problems were run on a SUN Ultra 60 (450MHz, 1GB, Solaris2.7) workstation with a time limit of 10 minutes and a space limit of 240MB. All problems exhibit the pruning effect of Boolean Constraint. Especially, all the 12 problems are solved with Boolean Constraints although 8 of 12 could not be solved without Boolean Constraints.

### Table 1: Comparison of experimental results.

<table>
<thead>
<tr>
<th>Problem</th>
<th>B.C.</th>
<th>Time (seconds)</th>
<th>Proof Tree</th>
</tr>
</thead>
<tbody>
<tr>
<td>PUZ 010-1</td>
<td>no</td>
<td>261.55</td>
<td>#B 2466</td>
</tr>
<tr>
<td>PUZ 015-2.006</td>
<td>yes</td>
<td>125.46</td>
<td>#N 17895</td>
</tr>
<tr>
<td>PUZ 018-1</td>
<td>no</td>
<td>0.50</td>
<td>#B 113</td>
</tr>
<tr>
<td>PUZ 018-2</td>
<td>yes</td>
<td>33.71</td>
<td>#N 343</td>
</tr>
<tr>
<td>PUZ 028-5</td>
<td>no</td>
<td>111.38</td>
<td>#B 272</td>
</tr>
<tr>
<td>PUZ 028-6</td>
<td>yes</td>
<td>9.05</td>
<td>#N 1147</td>
</tr>
<tr>
<td>CIV 008-1.002</td>
<td>no</td>
<td>0.63</td>
<td>#B 5</td>
</tr>
<tr>
<td>MSC 007-2.005</td>
<td>yes</td>
<td>124.49</td>
<td>#N 171899</td>
</tr>
<tr>
<td>GEO 013-3</td>
<td>no</td>
<td>55.63</td>
<td>#N 122</td>
</tr>
<tr>
<td>GEO 033-3</td>
<td>yes</td>
<td>44.75</td>
<td>#N 90</td>
</tr>
<tr>
<td>GEO 051-3</td>
<td>no</td>
<td>2.24</td>
<td>#B 1</td>
</tr>
<tr>
<td>SYN 009-1</td>
<td>yes</td>
<td>162.63</td>
<td>#N 29526</td>
</tr>
</tbody>
</table>

PUZ018-2 is satisfiable.

Others are unsatisfiable.

T.O.: Time Out (10 minutes)
B.C.: Boolean Constraint
#B: No. of branches
#N: No. of nodes

4. Future Work

In the current implementation, the most time consuming task is the conflict test (Fig. 3 (2) (5)). The cost of this task may be reduced by using Binary Decision Diagrams (BDDs). We are considering two approaches using BDD: One is to replace the constraint solver with BDD. Another is to use a BDD for representing a proof tree of model generation.

In the latter, all model candidates are simultaneously represented as the paths ending with a truth node in a BDD. With this representation, model candidates conflicting with the Boolean Constraint are automatically eliminated by standard BDD functions. Thus, the conflict test can be ignored. However, an implementation of the latter approach is more difficult than that of the former because BDD may create more model candidates than the model generation procedure, and it would be necessary to select a minimal one for efficiency. We are now developing a prototype for the former.

### References

1) R. Hähnle. BDDs for Representation of Model Candidates in MGTP. a private talk at Kyushu University, September 1998.


