

Proof Logging for Projected Enumeration (and Counting?) Problems in VERIPB

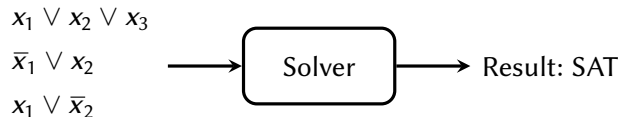
Ciaran McCreesh, Jakob Nordström, **Andy Oertel**, and Yong Kiam Tan

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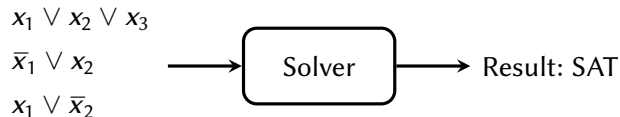
Combinatorial Solving



Combinatorial problem:

- ▶ Collection of constraint to be satisfied (e.g., clauses)
- ▶ Variables take discrete values (e.g., Boolean with domain $\{0, 1\}$)

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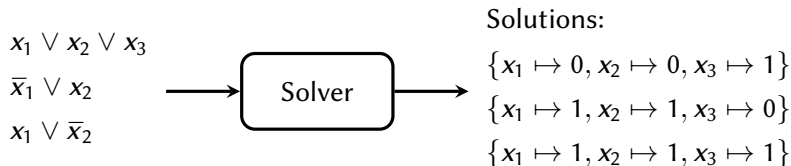
Decision problem:

- ▶ Constraints satisfiable or unsatisfiable?

Optimization problem:

- ▶ Best solution(s) with respect to objective function

Solution Enumeration



Enumeration problem:

- ▶ List solutions
- ▶ **Complete enumeration:** List **all** solutions
- ▶ Enumeration in SAT community known as **AlSAT** [TS16]
- ▶ Enumeration used for, e.g., computer algebra and graph theory [AMV22, TTT06]

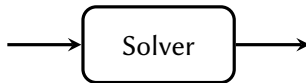
Projected Solution Enumeration

preserved: $\{x_1, x_2\}$

$x_1 \vee x_2 \vee x_3$

$\bar{x}_1 \vee x_2$

$x_1 \vee \bar{x}_2$



Solutions:

$\{x_1 \mapsto 0, x_2 \mapsto 0, x_3 \mapsto 1\}$

$\{x_1 \mapsto 1, x_2 \mapsto 1, x_3 \mapsto 0\}$

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Projected enumeration problem:

- ▶ Only interested in unique solutions to **preserved variables**
- ▶ Different names for preserved set in literature:
 - ▶ projection [FHH21], priority [ACMS15] or data [BNAH25] variables

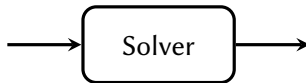
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Projected enumeration problem:

- ▶ Only interested in unique solutions to **preserved variables**
- ▶ Different names for preserved set in literature:
 - ▶ projection [FHH21], priority [ACMS15] or data [BNAH25] variables
- ▶ Solutions are equivalent if they are equivalent with respect to the preserved variables

- ▶ **Example:** $\begin{array}{l} \{x_1 \mapsto 1, x_2 \mapsto 1, x_3 \mapsto 0\} \\ \{x_1 \mapsto 1, x_2 \mapsto 1, x_3 \mapsto 1\} \end{array} \xrightarrow{\text{project to}} \{x_1 \mapsto 1, x_2 \mapsto 1\}$

Correctness of Solver Result

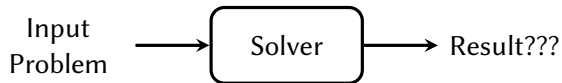
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How to guarantee correct results?

- ▶ **Testing**: no correctness guarantees
- ▶ **Formal verification**: does not scale to modern solvers
- ▶ **Proof logging** (our approach):

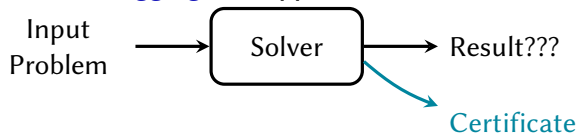


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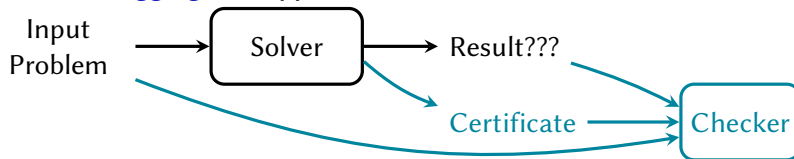
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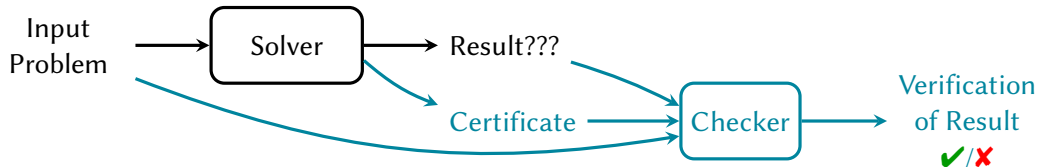
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Workflow for proof logging:

1. Solver outputs certificate for correctness of result
2. Independent checker verifies result w.r.t. input using certificate
3. Checker confirms correctness of result or that *something* is wrong

Proof Logging for (Projected) Enumeration

Proof logging exists for

- ▶ decision problems (e.g., [ZM03, HHW13, BGMN23])
- ▶ optimization problems (e.g., [CGS17, GMM⁺24])
- ▶ and to some extent for counting problems (e.g., [FHR22, BNAH25])

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This work:

- ▶ First proof logging for (projected) enumeration
- ▶ Solution count preserving reformulation proofs
- ▶ Formal guarantees for enumeration and count preserving reformulations

Proof Preliminaries

- ▶ **Boolean variable x :** with domain 0 (false) and 1 (true)
- ▶ **Literal:** x or negation $\bar{x} = 1 - x$
- ▶ **Proof lines:** Pseudo-Boolean (PB) constraint (integer linear inequality over literals)

$$3x_1 + 2x_2 + 5\bar{x}_3 \geq 5$$

- ▶ **Clause:** disjunction of literals / at-least-one constraint

$$x \vee \bar{x}_2 \vee \bar{x}_3 \iff x + \bar{x}_2 + \bar{x}_3 \geq 1$$

- ▶ **Reification:** $x \iff C$ for variable x and constraint $C \doteq \sum_i a_i \ell_i \geq A$ is constraints

$$x \Rightarrow C \doteq A\bar{x} + \sum_i a_i \ell_i \geq A$$

$$x \Leftarrow C \doteq (\sum_i a_i - A + 1)x + \sum_i a_i \bar{\ell}_i \geq \sum_i a_i - A + 1$$

(Projected) Enumeration Proofs in VERIPB

Idea:

- ▶ **Enumeration**: log solutions and disallow them to be logged again
- ▶ **Goal** (complete enumeration): prove that there are no more solutions
 - ▶ Same as showing unsatisfiability (by deriving contradiction)

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Remaining problems:

- ▶ Preventing logging a solution twice (blocking)
- ▶ Deriving constraint in proof can remove solutions
- ▶ Deleting constraint in proof can add solutions

Blocking (Projected) Solutions

Blocking clause:

- **Example:** If preserved set $\{x_1, x_2\}$ and solution $\{x_1 \mapsto 1, x_2 \mapsto 0, x_3 \mapsto 1\}$, then solution is blocked by

$$\bar{x}_1 \vee x_2$$

Blocking (Projected) Solutions

Blocking clause:

- **Example:** If preserved set $\{x_1, x_2\}$ and solution $\{x_1 \mapsto 1, x_2 \mapsto 0, x_3 \mapsto 1\}$, then solution is blocked by

$$\bar{x}_1 \vee x_2$$

- **In general:** Given preserved set P and solution σ , the clause

$$\bigvee_{\{p \in P: \sigma(p)=0\}} p \vee \bigvee_{\{p \in P: \sigma(p)=1\}} \bar{p}$$

excludes all solutions assigning the same values to the variables in P as σ

- New **solx** rule for logging and blocking solution

Deriving Constraints

Danger:

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Strengthening rules:

- ▶ Derived constraint satisfied by at least one but not all solutions to current constraints
- ▶ **Example:** $\{x_1 \vee x_2 \vee x_3, \quad \bar{x}_1 \vee x_2, \quad x_1 \vee \bar{x}_2\}$
 - ▶ W.l.o.g. $x_3 \mapsto 1$, strengthening allows us to derive clause x_3
- ▶ **Strengthening can remove solution that have not been enumerated!**

Strengthening in More Detail

Redundance-based strengthening [GN21]:

- **Substitution:** Mapping variables to literals/values (e.g., $\omega' = \{x_1 \mapsto 1, x_2 \mapsto \bar{x}_3\}$)

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- ▶ **Substitution**: Mapping variables to literals/values (e.g., $\omega' = \{x_1 \mapsto 1, x_2 \mapsto \bar{x}_3\}$)
- ▶ Constraint C can be derived from constraints F given substitution ω and showing

$$F \cup \{\neg C\} \models (F \cup \{C\})|_{\omega}$$

- ▶ If solution σ satisfies F but falsifies C , then $\sigma \circ \omega$ satisfies $F \cup \{C\}$
- ▶ E.g., $\sigma' = \{x_1 \mapsto 0, x_2 \mapsto 1, x_3 \mapsto 1\}$, then $\sigma' \circ \omega' = \{x_1 \mapsto 1, x_2 \mapsto \bar{x}_3 \mapsto 0, x_3 \mapsto 1\}$

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- ▶ **Observation**: Only assignment to variables in domain of substitution are changed
- ▶ If ω only changes non-preserved variables, $\sigma \circ \omega$ and σ **project to same solution**
- ▶ Domain of substitution can not contain preserved variables

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- ▶ **Observation**: Only assignment to variables in domain of substitution are changed
- ▶ If ω only changes non-preserved variables, $\sigma \circ \omega$ and σ **project to same solution**
- ▶ Domain of substitution can not contain preserved variables
- ▶ **Dominance-based strengthening** and **deletion** in VERIPB works similarly
- ▶ **Key takeaway**: Only substitutions need to be restricted

Projected Enumeration Example

Preserved set: $\{x_1, x_2\}$

Constraints:

$$x_1 \vee x_2 \vee x_3$$

$$\bar{x}_1 \vee x_2$$

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Enumerated solutions:

proof.pbp

pseudo-Boolean proof version 3.0

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red 1 x3 >= 1 : x3 -> 1;
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proof.pbp

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% W.l.o.g. x3 is 1 with substitution x3 -> 1
red 1 x3 >= 1 : x3 -> 1;
% Log first solution x1 -> 0 and x2 -> 0
solx ~x1 ~x2;
```

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% No more solutions with x1 -> 0
rup 1 x1 >= 1;
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% Log second solution x1 -> 1 and x2 -> 1
solx x1 x2;
% No more solutions
rup >= 1;
output NONE;
conclusion ENUMERATION_COMPLETE 2;
end pseudo-Boolean proof;
```

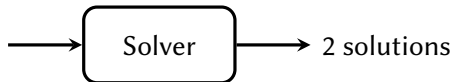

(Projected) Model/Solution Counting

preserved: $\{x_1, x_2\}$

$x_1 \vee x_2 \vee x_3$

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(Projected) counting:

- Give the number of (projected) solutions

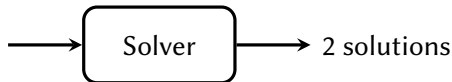
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(Projected) counting:

- ▶ Give the number of (projected) solutions

Idea for certification:

- ▶ Inspiration from CPOG framework [BNAH25, BTH25]
- ▶ Reformulate problem into “easy-to-count” form preserving (projected) solution count
- ▶ Reformulation proofs already supported by VERIPB (e.g., prove equisatisfiable)

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Rule:

- ▶ Add/remove x w.r.t. constraint C over other preserved variables
- ▶ Show that $x \Leftrightarrow C$ holds
- ▶ If C is satisfied, then x is 1
- ▶ If C is falsified, then x is 0

Counting Example

Original problem

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% Define variable s1 <==> x1 + x2 >= 2
red 2 ~s1 1 x1 1 x2 >= 2 : s1 0;
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% Add s1 to preserved with s1 <==> x1 + x2 >= 2
preserved_add s1 1 x1 1 x2 >= 2;
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preserved_add s1 1 x1 1 x2 >= 2;
% Remove x1 and x2 from preserved
preserved_rm x1 1 s1 1 x2 >= 2;
preserved_rm x2 1 s1 >= 1;
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preserved_rm x2 1 s1 >= 1;
% Delete definitions of s1 using solutions
del id 3 : x1 1 x2 1; del id 4 : x1 0 x2 0;
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% Delete definitions of s1 using solutions
del id 3 : x1 1 x2 1; del id 4 : x1 0 x2 0;
% Delete original constraint with one solution
del id 1 : x1 0 x2 0; del id 2 : x1 0 x2 0;
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Counting Example

Original problem

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Constraints:

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Reformulated problem

Preserved set: $\{s_1\}$

Constraints: $\emptyset$

Number of solutions: 2

proof.pbp

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del id 3 : x1 1 x2 1; del id 4 : x1 0 x2 0;
% Delete original constraint with one solution
del id 1 : x1 0 x2 0; del id 2 : x1 0 x2 0;
% Output has same projected solution count
output EQUIENUMERABLE FILE; ...
```

Formally Verified Checking

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Trust base:

1. Higher-order logic (HOL) definitions of parser and problems
 - ▶ kept as simple as possible, easy to check
2. HOL model of CakeML environment and correspondence to real system
 - ▶ has been validated extensively
3. HOL4 theorem prover: its logic, implementation, and execution environment
 - ▶ separated and trustworthy kernel checks every logical inference

Experimental Results

Maximal clique enumeration [TTT06]:

- ▶ Adapted implementation by Gocht et al. [GMM⁺20]
 - ▶ no guarantees about enumeration, no formal verification
- ▶ Guarantee complete enumeration
- ▶ Formally verified maximal clique enumeration frontend for CAKEPB
- ▶ One instance out of memory for CAKEPB compared to Gocht et al.

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Enumeration in Glasgow Constraint Solver [GMN22]:

- ▶ Projected enumeration of pseudo-Boolean encoding of constraint problem
- ▶ Proofs checked by CAKEPB
- ▶ Verified implementation on thousands of randomly generated problems

Conclusions & Future Work

Background:

- ▶ Proof logging for decision and optimization problems well-established
- ▶ No proof logging for enumeration

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This Work:

- ▶ Proof logging of (projected) enumeration problems in the VERIPB
- ▶ Problem reformulations guaranteeing same (projected) solution count
- ▶ Formally verified checker for enumeration and count preserving reformulations

Conclusions & Future Work

Background:

- ▶ Proof logging for decision and optimization problems well-established
- ▶ No proof logging for enumeration

This Work:

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Future work:

- ▶ Enumeration of solution cubes (all solutions of cube satisfy constraints)
- ▶ Enumeration of optimal solutions
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Thank you for your attention!

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