# Proof Logging for Preprocessing/Presolving in MaxSAT and 0-1 Integer Linear Programming

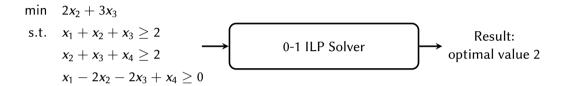
Andy Oertel
Lund University & University of Copenhagen

WHOOPS '25

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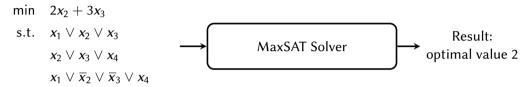
Based on work together with Jeremias Berg, Ambros Gleixner, Alexander Hoen, Hannes Ihalainen, Matti Järvisalo, Magnus O. Myreen, Jakob Nordström, and Yong Kiam Tan





- Specialization of mixed integer programming (MIP)
- ▶ Input: 0-1 integer linear program (or pseudo-Boolean formula)
  - Integer linear objective function and collection of integer linear inequalities/constraints
  - ▶ Variables with domain {0,1}
- ► Output:
  - Optimal value of objective subject to satisfying all inequalities

# Maximum Satisfiability (MaxSAT)



- Optimization variant of SAT problem
- ► Input: MaxSAT problem
  - ► Integer linear objective function and collection of clauses
  - ► Variables with domain {0,1}
- ► Output:
  - Optimal value of objective subject to satisfying all clauses

# Maximum Satisfiability (MaxSAT)

min 
$$2x_2 + 3x_3$$
  
s.t.  $x_1 + x_2 + x_3 \ge 1$   
 $x_2 + x_3 + x_4 \ge 1$ 

$$x_1 + \overline{x}_2 + \overline{x}_3 + x_4 \ge 1$$
Result: optimal value 2

- Optimization variant of SAT problem
- ► Input: MaxSAT problem
  - ► Integer linear objective function and collection of clauses
  - ► Variables with domain {0,1}
- Output:
  - Optimal value of objective subject to satisfying all clauses
  - Specialization of 0-1 ILP

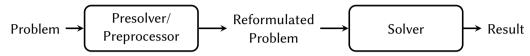


#### So far:

Introduction

Problem directly given to solver

# Idea of Presolving/Preprocessing



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Problem directly given to solver

### Typical workflow:

- Problem reformulated before it is given to core solver
- Known as presolving in the MIP community
- Known as preprocessing in the MaxSAT community
- Can be tightly integrated with solver or independent tool

# Idea of Presolving/Preprocessing



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Problem directly given to solver

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### Important for state-of-the-art performance!

# Importance of Presolving in MIP

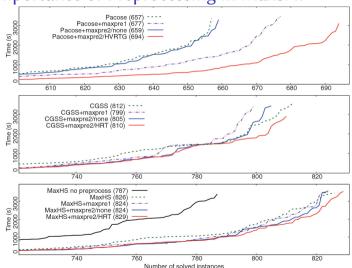
▶ Performance analysis of presolve reductions in MIP [ABG<sup>+</sup>20]

		default	disabled presolving			
bracket	#instances	#timeout	#timeout	#faster	#slower	×slower
all	3047	547	1035	255	1755	3.36
$\geq 0~{\sf sec}$	2511	16	504	255	1755	4.52
$\geq$ 1 sec	1944	16	504	210	1634	6.60
$\geq$ 10 sec	1575	16	504	141	1380	9.05
$\geq$ 100 sec	1099	16	504	86	983	12.36
≥ 1000 sec	692	16	504	34	643	19.48

Presolving is one of the most important techniques in mixed-integer programming!

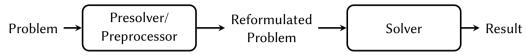
## Importance of Preprocessing in MaxSAT

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 Performance analysis for MaxSAT preprocessing with MaxPre [IB]22]

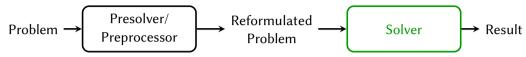
Preprocessing improves performance significantly for many MaxSAT solvers!



#### Goal:

Introduction

Certification for whole solving workflow



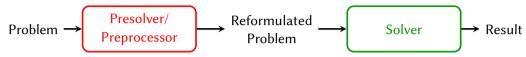
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Introduction

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#### Good news:

- ► Certification for some MIP solving algorithms using VIPR [CGS17]
- ► Certification for MaxSAT using VERIPB [VDB22, Van23, BBN<sup>+</sup>23, BBN<sup>+</sup>24]



#### Goal:

Introduction

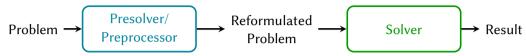
Certification for whole solving workflow

#### Good news:

- ► Certification for some MIP solving algorithms using VIPR [CGS17]
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#### Problem:

Also need to certify problem reformulations in presolver/preprocessor



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### This talk:

- Certification of 0-1 ILP presolving and MaxSAT preprocessing
- Formally verified end-to-end verification framework for problem reformulations

### Outline

- 1. Proof Logging for Preprocessing/Presolving
- 2. Example
- 3. Formal Verification
- 4. Experiments

### **Preliminaries**

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

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

- 0-1 integer linear constraint is same as PB constraint
- ► Equality constraint: syntactic sugar for 2 inequalities

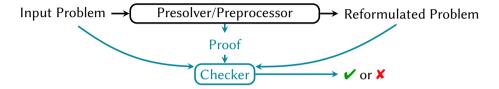
$$3x_1 + 2x_2 + 5\overline{x}_3 = 5$$
  $\longrightarrow$   $3x_1 + 2x_2 + 5\overline{x}_3 \ge 5$   $3x_1 + 2x_2 + 5\overline{x}_3 \le 5$ 

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

$$x_1 \vee \overline{x}_2 \vee \overline{x}_3 \iff x_1 + \overline{x}_2 + \overline{x}_3 \geq 1$$

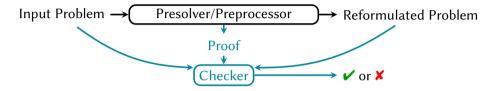
MaxSAT is special case of 0-1 ILP

### **Proof Invariants**



Step-by-step modify optimization problem preserving optimal value

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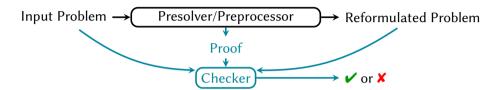


Step-by-step modify optimization problem preserving optimal value

#### Two sets of constraints needed:

- ightharpoonup Core set  $\mathcal{C}$  guarantee:
  - ightharpoonup Current problem (min f', s.t. C) has same optimal value as input problem (min f, s.t. F)
- ightharpoonup Derived set  $\mathcal{D}$  of constraints are all constraints derived by rules
  - ▶ Any solution to C can be extended to a solution of  $C \cup D$
  - Constraints can be moved from derived to core set

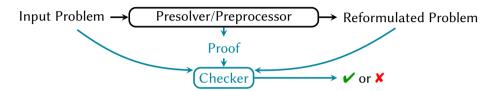
# **Certifying Problem Reformulations**



### How to certify presolving/preprocessing?

- Represent each reformulation using proof steps
- ► Soundness of proof system guarantees that optimal value does not change
- Check that core set and objective at end of proof match output problem

# **Certifying Problem Reformulations**



### How to certify presolving/preprocessing?

- Represent each reformulation using proof steps
- Soundness of proof system guarantees that optimal value does not change
- Check that core set and objective at end of proof match output problem

#### Guarantee:

Input problem has same optimal value as output problem of presolver/preprocessor

# Cutting Planes Proof System [CCT87]

### Rules that preserve set of solutions:

► Literal axiom

Literal 
$$x \overline{x \ge 0}$$
 Literal  $\overline{x} \overline{\overline{x} \ge 0}$ 

Addition

Addition 
$$\frac{x_1 + 2\overline{x}_2 + 2\overline{x}_3 \ge 3}{x_1 + 3\overline{x}_2 + x_3 \ge 4}$$

Multiplication

Multiply by 2 
$$\frac{x_1 + 2\overline{x}_2 \ge 3}{2x_1 + 4\overline{x}_2 \ge 6}$$

Division (and rounding up)

Divide by 2 
$$\frac{2x_1 + 2\bar{x}_2 + 4x_3 \ge 5}{x_1 + \bar{x}_2 + 2x_3 \ge \lceil 2.5 \rceil}$$

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$$\frac{2x_1 + 2\bar{x}_2 + 4x_3 \ge 5}{x_1 + \bar{x}_2 + 2x_3 \ge 3}$$

# **Cutting Planes: Example**

#### Proof tree:

$$\begin{array}{c} \text{Literal $\overline{x}_2$} \\ \text{Multiply by 2} \\ \text{Addition} \\ \hline \\ \text{Divide by 3} \\ \hline \\ \hline \\ \hline \\ x_1 + \overline{x}_3 \geq 1 \\ \hline \end{array}$$

### **VERIPB** syntax:

$$po1 \sim x2 2 * 42 + 3 d$$
;

# Redundance-Based Strengthening

- ► Cutting planes rules preserve set of solutions
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### Redundance-based strengthening [BT19, GN21]

- $\triangleright$  Requires substitution  $\omega$  (mapping variables to truth values or literals)
- lacktriangle We can introduce C with respect to constraints  $\mathcal{C} \cup \mathcal{D}$  and objective f if

$$\mathcal{C} \cup \mathcal{D} \cup \{\neg C\} \vdash \mathcal{C} \cup \mathcal{D} \cup \mathcal{C} \upharpoonright_{\omega} \cup \{f \ge f \upharpoonright_{\omega}\}$$

- $\triangleright \omega$  has to be given explicitly
- Implication should be efficiently checkable:
  - Obvious to proof checker
  - Or explicitly by cutting planes proof

# Redundance-Based Strengthening: Example

min 
$$x_1 + x_2$$
  
s.t.  $x_1 + 2x_2 + 3\overline{x}_3 \ge 3$  (42)  $\xrightarrow{\text{min }} x_1 + x_2$   
s.t.  $x_1 + 2x_2 + 3\overline{x}_3 \ge 3$  (42)  $\overline{x}_3 \ge 1$  (43)

### **VERIPB** syntax:

red 1 
$$\sim x3 >= 1 : x3 -> 0 :$$

 $ightharpoonup \overline{x}_3 \ge 1$  derived by redundance-based strengthening with  $\{\overline{x}_3 \mapsto 0\}$ 

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### **VERIPB** syntax:

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$$\mathcal{C} \cup \mathcal{D} \cup \{\neg C\} \vdash (\mathcal{C} \cup \mathcal{D} \cup C) \upharpoonright_{\omega} \cup \{f \geq f \upharpoonright_{\omega}\}$$

- All implications are trivial:
  - For constraint (42),  $(x_1 + 2x_2 + 3\overline{x}_3 \ge 3)$  is  $x_1 + 2x_2 \ge 0$
  - For derived constraint,  $(\bar{x}_3 \ge 1) \upharpoonright_{\omega}$  is  $0 \ge 0$
  - For objective condition  $f \ge f \upharpoonright_{\omega}$ ,  $x_1 + x_2 \ge (x_1 + x_2) \upharpoonright_{\omega}$  is  $x_1 + x_2 \ge x_1 + x_2$

# Usage of Strengthening Rules

### Strengthening useful for:

- Basic symmetry breaking
- ► Without-loss-of-generality reasoning
- Introducing extension variables

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- So-called dominance-based strengthening rule for advanced symmetry breaking
- See [BGMN23] for details

# Usage of Strengthening Rules

### Strengthening useful for:

- ► Basic symmetry breaking
- ► Without-loss-of-generality reasoning
- Introducing extension variables

### Additional strengthening rule:

- So-called dominance-based strengthening rule for advanced symmetry breaking
- ► See [BGMN23] for details
- ... or next talk by Markus Anders about "Proof logging for symmetry breaking"

### Problem:

- Deleting constraints arbitrarily is unsound, as
  - Introduce better than optimal solutions
  - Even remove all solutions (when combined with dominance-based strengthening)
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### Deletion

#### Problem:

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  - Even remove all solutions (when combined with dominance-based strengthening)
- Deletion needs to be restricted

#### Solution:

- Constraint C can only be deleted if
  - ightharpoonup C in derived set  $\mathcal{D}$
  - ightharpoonup C rederivable by redundance-based strengthening from core set C without using C

$$(\mathcal{C} \setminus \{C\}) \cup \{\neg C\} \vdash ((\mathcal{C} \setminus \{C\}) \cup C) \upharpoonright_{\omega} \cup \{f \geq f \upharpoonright_{\omega}\}$$

# Deletion: Example

del id 42 :

- ▶ Deletion of constraint 42 with empty substitution
- Deleted constraint implied by propagation

$$(\mathcal{C} \setminus \{C\}) \cup \{\neg C\} \vdash ((\mathcal{C} \setminus \{C\}) \cup C) \cup \{f \ge f\}$$

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$$(\mathcal{C} \setminus \{C\}) \cup \{\neg C\} \vdash ((\mathcal{C} \setminus \{C\}) \cup C) \cup \{f \ge f\}$$

► Also with explicit subproof (proof by contradiction)

```
del id 42 : subproof pol -1 43 3 * + ; Qued : -1 ; Add negated constraint \overline{x}_1 + 2\overline{x}_2 + 3x_3 \ge 4 Cutting planes proof resulting in \overline{x}_1 + 2\overline{x}_2 \ge 4 Previous constraint is contradiction
```

# Objective Update Rule

#### Effect:

- Allows objective function change from  $f_{old}$  to  $f_{new}$
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#### Check:

- Equality  $f_{old} = f_{new}$  trivial or explicit cutting planes proof
- Only core constraints can be used for this check
  - ▶ Deriving  $f_{old} \le f_{new}$  from the derived set can introduce better than optimal solutions

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  - ▶ Deriving  $f_{old} \le f_{new}$  from the derived set can introduce better than optimal solutions

### Objective update specification options:

- 1. Specify new objective  $f_{new}$ 
  - ► Good if change is large or new objective small
- 2. Specify difference between new and old objective  $f_{new} f_{old}$ 
  - Good for small objective changes

### Without objective update:

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### Example:

min 
$$x_1 + x_2$$
  
s.t.  $x_1 + x_2 + \overline{x}_3 + \overline{x}_4 \ge 2$   
 $x_1 + x_2 + \overline{x}_3 + \overline{x}_4 \le 2$   
 $\overline{x}_3 + \overline{x}_4 \ge 1$ 

### Without objective update:

Redundance-based strengthening becomes more complicated to impossible

### Example:

min 
$$x_1 + x_2$$
  
s.t.  $x_1 + x_2 = x_3 + x_4$   
 $\bar{x}_3 + \bar{x}_4 \ge 1$ 

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s.t.  $x_1 + x_2 = x_3 + x_4$   
 $\overline{x}_3 + \overline{x}_4 \ge 1$ 

▶ Deletion of  $x_1 + x_2 + \overline{x}_3 + \overline{x}_4 \ge 2$  with substitution  $\{x_1 \mapsto 1\}$ 

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- ▶ Deletion of  $x_1 + x_2 + \overline{x}_3 + \overline{x}_4 \ge 2$  with substitution  $\{x_1 \mapsto 1\}$
- ▶ If objective unchanged, then  $x_1 + x_2 \ge 1 + x_2$  ( $f \ge f \upharpoonright_{\omega}$ ) has to be shown
- ▶ This is not required if objective is updated, as  $x_3 + x_4 \ge x_3 + x_4$  ( $f \ge f \upharpoonright_{\omega}$ ) is trivial

## Objective Update: Example

min 
$$2x_1 + 3\bar{x}_2$$
  
s.t.  $3\bar{x}_2 + 2\bar{x}_3 > 3$  (1)

$$3x_2+2x_3\geq 2 \qquad (2)$$

min 
$$2x_1 + 2x_3 + 1$$

s.t. 
$$3\bar{x}_2 + 2\bar{x}_3 \ge 3$$
 (1)

$$3x_2 + 2x_3 \ge 2$$
 (2)

- Constraint (1) says  $3\overline{x}_2 \ge 2x_3 + 1$
- Constraint (2) says  $3\overline{x}_2 \leq 2x_3 + 1$

Updating to a new objective:

► Change objective to  $2x_1 + 2x_3 + 1$ 

obju diff 
$$-3 \sim x2 2 x3 1$$
;

- ► Change objective with difference  $f_{new} f_{old} = 2x_3 + 1 3\overline{x_2}$
- New objective is  $f_{new} = f_{old} + 2x_3 + 1 3\overline{x_2}$

Proof starts with a header specifying the format version: pseudo-Boolean proof version 3.0

# Example: Start of Proof

Proof starts with a header specifying the format version:

pseudo-Boolean proof version 3.0

Input problem is loaded, e.g., from OPB file, and IDs assigned to initial constraints:

$$1 \ x1 \ 1 \ x2 \ 1 \ \sim x3 \ 1 \ \sim x4 >= 3$$
;

$$1 \sim x1 \ 1 \sim x2 \ 1 \ x3 \ 1 \ x4 >= 1 :$$

$$1 \sim x1 \ 1 \ x5 >= 1$$
:

min 
$$x_1 + x_2$$

s.t. 
$$x_1 + x_2 + \overline{x}_3 + \overline{x}_4 \ge 3$$
 (1)

$$\bar{x}_1 + \bar{x}_2 + x_3 + x_4 \ge 1$$
 (2)

$$\overline{x}_1 + x_5 \ge 1 \tag{3}$$

## Example (1/4): Substitution

- Constraints (1) and (2) say that  $x_1 = \overline{x}_2 + x_3 + x_4$
- $\triangleright$  Substitute  $x_1$  in constraint (3) using this equality

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#### Certification:

- Constraints (1) and (2) say that  $x_1 = \overline{x}_2 + x_3 + x_4$
- ▶ Substitute  $x_1$  in constraint (3) using this equality

#### Certification:

Add up constraints (1) and (3) to derive (4)

Move constraint (4) to core set

Add negation of constraint (3) to get  $x_1 + \overline{x}_5 \ge 2$ 

Add (2) to previous constraint to get  $\bar{x}_2 + x_3 + x_4 + \bar{x}_5 \ge 2$ Add (4) to previous constraint to get  $0 \ge 1$ 

End subproof to delete constraint (3)

# Example (2/4): Objective Function Update

min 
$$x_1 + x_2$$
  
s.t.  $x_1 + x_2 + \overline{x}_3 + \overline{x}_4 \ge 3$  (1)  
 $\overline{x}_1 + \overline{x}_2 + x_3 + x_4 \ge 1$  (2)  
 $x_2 + \overline{x}_3 + \overline{x}_4 + x_5 \ge 3$  (4)

min 
$$x_3 + x_4 + 1$$

s.t. 
$$x_1 + x_2 + \overline{x}_3 + \overline{x}_4 \ge 3$$
 (1)

$$\bar{x}_1 + \bar{x}_2 + x_3 + x_4 \ge 1$$
 (2)

$$x_2 + \overline{x}_3 + \overline{x}_4 + x_5 \ge 3$$
 (4)

- ► Change objective from  $x_1 + x_2$  to  $x_3 + x_4 + 1$  using constraints (1) and (2)
  - Constraint (1) says  $x_1 + x_2 \ge x_3 + x_4 + 1$
  - Constraint (2) says  $x_1 + x_2 \le x_3 + x_4 + 1$

# Example (2/4): Objective Function Update

- ► Change objective from  $x_1 + x_2$  to  $x_3 + x_4 + 1$  using constraints (1) and (2)
  - Constraint (1) says  $x_1 + x_2 \ge x_3 + x_4 + 1$
  - Constraint (2) says  $x_1 + x_2 \le x_3 + x_4 + 1$

#### Certification:

obju new 1 x3 1 x4 1; Change objective to  $x_3 + x_4 + 1$  using (1) and (2)

## Example (3/4): Delete Substitution Constraints (1/2)

min 
$$x_3 + x_4 + 1$$
  
s.t.  $x_1 + x_2 + \overline{x}_3 + \overline{x}_4 \ge 3$  (1)  $x_1 + \overline{x}_2 + x_3 + x_4 \ge 1$  (2)  $x_2 + \overline{x}_3 + \overline{x}_4 + x_5 \ge 3$  (4)  $x_2 + \overline{x}_3 + \overline{x}_4 + x_5 \ge 3$  (4)

Constraints (1) and (2) can be deleted, as they define  $x_1 = \overline{x}_2 + x_3 + x_4$ 

# Example (3/4): Delete Substitution Constraints (1/2)

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s.t.  $x_1 + x_2 + \overline{x}_3 + \overline{x}_4 \ge 3$  (1)  $x_1 + \overline{x}_2 + x_3 + x_4 \ge 1$  (2)  $x_2 + \overline{x}_3 + \overline{x}_4 + x_5 \ge 3$  (4)  $x_2 + \overline{x}_3 + \overline{x}_4 + x_5 \ge 3$  (4)

 $\triangleright$  Constraints (1) and (2) can be deleted, as they define  $x_1 = \overline{x}_2 + x_3 + x_4$ 

#### Certification:

del id 1 : x1 -> 1 ; Negation of (1) is 
$$\overline{x}_1 + \overline{x}_2 + x_3 + x_4 \ge 2$$
, hence: For (1),  $x_2 + \overline{x}_3 + \overline{x}_4 \ge 2$  is implied by (4); For (2),  $\overline{x}_2 + x_3 + x_4 \ge 1$  is implied by negated constraint; For (4), substitution does not change the constraint; For  $f \ge f \upharpoonright_{\{x_1 \mapsto 1\}}$ , is trivial, since  $x_1$  not in objective

## Example (3/4): Delete Substitution Constraints (2/2)

Now also delete constraint (2)

#### Certification:

del id 2 : x1 -> 0 ; Delete (2) using substitution 
$$\{x_1 \mapsto 0\}$$
: For (2),  $\bar{x}_2 + x_3 + x_4 \ge 0$  is trivial; (4) and  $f \ge f \upharpoonright_{\{x_1 \mapsto 0\}}$  are again trivial

# Example (4/4): Duality-Based Fixing

min 
$$x_3 + x_4 + 1$$
 min  $x_3 + x_4 + 1$   
s.t.  $x_2 + \overline{x}_3 + \overline{x}_4 + x_5 \ge 3$  (4) s.t.  $\emptyset$ 

- $\triangleright$  W.l.o.g.  $x_2/x_5$  can be fixed to 1, as
  - $ightharpoonup x_2/x_5$  only appear positive in constraints with non-negative coefficient
  - b objective coefficients for  $x_2/x_5$  are 0
- $\triangleright$  W.l.o.g.  $x_3/x_4$  can be fixed to 0, as
  - $x_3/x_4$  only appear negative in constraints with non-negative coefficient
  - $ightharpoonup x_3/x_4$  only appear positive in the objective with non-negative coefficient

# Example (4/4): Duality-Based Fixing

min 
$$x_3 + x_4 + 1$$
 min  $x_3 + x_4 + 1$  s.t.  $x_2 + \bar{x}_3 + \bar{x}_4 + x_5 \ge 3$  (4) s.t.  $\emptyset$ 

- $\triangleright$  W.l.o.g.  $x_2/x_5$  can be fixed to 1, as
  - $\triangleright$   $x_2/x_5$  only appear positive in constraints with non-negative coefficient
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#### Certification:

del id 4 : x2 -> 1 x3 -> 0 Delete (4) using 
$$\omega = \{x_2 \mapsto 1, x_3 \mapsto 0, x_4 \mapsto 0, x_5 \mapsto 1\},$$
 as  $(x_2 + \overline{x}_3 + \overline{x}_4 + x_5 \ge 3) \upharpoonright_{\omega}$  is  $0 \ge -1$  and  $f \ge f \upharpoonright_{\omega}$  is  $x_3 + x_4 \ge -1$ , which are both trivial

# **Example: Concluding Proof**

We claim at the end of the proof that resulting formula has same optimal value:

```
output EQUIOPTIMAL FILE;
```

- ► Check that core set and objective is syntactically equivalent to reformulated problem
- ► FILE means that reformulated problem is given in additional file

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There is no conclusion regarding solving the instance:

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conclusion NONE;
```

End the proof:

```
end pseudo-Boolean proof;
```

## **Example: Full Proof**

```
reformulation.proof
pseudo-Boolean proof version 3.0
pol 1 3 + ;
core id 4:
del id 3 : : subproof
   pol -1 2 + ;
   pol -1 4 + :
qed: -1;
obju new 1 x3 1 x4 1;
del id 1 : x1 -> 1 :
del id 2 : x1 -> 0 :
del id 4 : x2 \rightarrow 1 \ x3 \rightarrow 0 \ x4 \rightarrow 0 \ x5 \rightarrow 1 :
output EQUIOPTIMAL FILE ;
conclusion NONE;
end pseudo-Boolean proof;
```

### input.opb

```
min: 1 x1 1 x2;

1 x1 1 x2 1 ~x3 1 ~x4 >= 3;

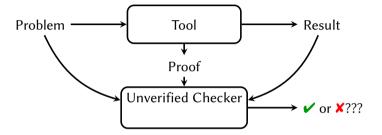
1 ~x1 1 ~x2 1 x3 1 x4 >= 1;

1 ~x1 1 x5 >= 1;
```

#### output.opb

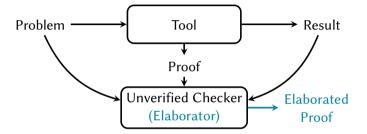
```
min: 1 x3 1 x4 1;
```

# Formally Verified Proof Checking



How can we trust the checker?

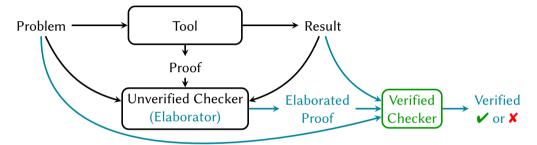
# Formally Verified Proof Checking



#### How can we trust the checker?

- 1. Tool generates proof, which contains syntactic sugar for easy logging
- 2. Unverified proof checker elaborates syntactic sugar to simpler elaborated proof

# Formally Verified Proof Checking



#### How can we trust the checker?

- 1. Tool generates proof, which contains syntactic sugar for easy logging
- 2. Unverified proof checker elaborates syntactic sugar to simpler elaborated proof
- 3. Elaborated proof checked by formally verified checker

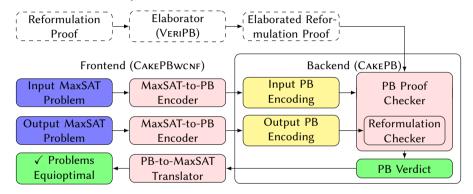
### Formal Verification Trust Base

#### What we have to trust:

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

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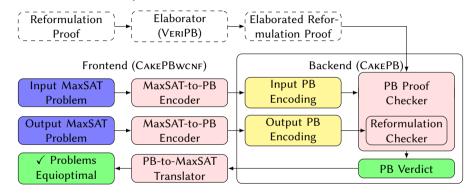
## Framework for Formally Verified Reformulation Checkers



- Example workflow for checking MaxSAT preprocessing proofs
- But easy to adapt checker to new problem domain, e.g., 0-1 ILP, graph problems, ...

Formal Verification

## Framework for Formally Verified Reformulation Checkers



- ► Example workflow for checking MaxSAT preprocessing proofs
- ▶ But easy to adapt checker to new problem domain, e.g., 0-1 ILP, graph problems, ...
- ► Talk on CakePB tomorrow 9:00 by Yong Kiam Tan

# Experimental Setup 0-1 ILP Presolving [HOGN24]

#### Tools:

- Added pseudo-Boolean proof logging to presolver PAPILO<sup>1</sup>
  - ► All presolve reductions applied to 0-1 ILPs in PAPILO covered
- Proof checked using proof checker VERIPB<sup>2</sup>

<sup>1</sup>https://github.com/scipopt/papilo

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#### Benchmarks:

- ▶ PB competition 2016 instances [Pse16]
- MIPLIB17 instances translated to OPB format [Dev20]

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# Proof Logging Overhead in PAPILO

Test set size		default [s]	w/proof log [s]	relative
PB16-dec	1397	0.06	0.06	1.00
MIPLIB01-dec	291	0.42	0.43	1.02
PB16-opt	531	0.65	0.66	1.02
MIPLIB01-opt	142	0.33	0.35	1.06

- ► Additional time required to write proof is very small
- ► For 99% of instances less than 0.001s per reduction for certification

## Certificate Checking Performance

			PaPILO time [s]		VeriPB	relative time w.r.t.	
test set	size	verified	default	w/proof log	time [s]	default	w/proof log
PB-dec	1397	1397	0.06	0.06	0.88	14.67	14.67
MIPLIB-dec	291	267	0.42	0.43	9.64	22.85	22.42
PB-opt	531	520	0.65	0.66	10.44	16.06	15.82
MIPLIB-opt	142	139	0.33	0.35	5.25	15.91	15.00

- ► Most instances verified within 10 000s timeout
- Overhead can be explained by PAPILO having more context than VERIPB
- PAPILO parallelizes some tasks, VeriPB works only sequentially
- ► Old version of VeriPB used, as new version<sup>3</sup> does not yet support required features

<sup>&</sup>lt;sup>3</sup>https://gitlab.com/MIAOresearch/software/pboxide

# Experimental Setup MaxSAT Preprocessing [IOT+24]

#### Tools:

- ► Added pseudo-Boolean proof logging to MaxSAT preprocessor MaxPre<sup>4</sup>
  - ► All techniques in MaxPre covered
- Proofs elaborated by VeriPB<sup>2</sup>
- Elaborated proofs checked by formally verified checker CAKEPB<sup>5</sup>

<sup>4</sup>https://bitbucket.org/coreo-group/maxpre2

<sup>&</sup>lt;sup>2</sup>https://gitlab.com/MIAOresearch/software/VeriPB

<sup>&</sup>lt;sup>5</sup>https://gitlab.com/MIAOresearch/software/cakepb

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#### Benchmarks:

MaxSAT evaluation 2023 instances [Max23]

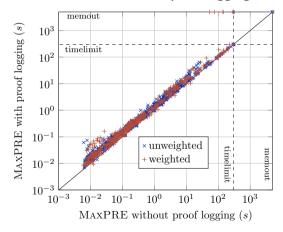
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## Proof Logging Overhead in MaxPRE

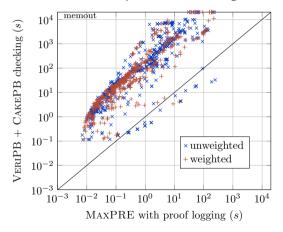
### MAXPRE with and without proof logging:



- ► 46% slower with proof logging
- Larger overhead than for 0-1 ILP presolving
- ► Bottleneck: Renaming of variables required for MaxSAT file format

## Certificate Checking Performance

### MAXPRE vs. formally verified checking:



- ▶ 92% of instances checked
- Renaming of variables also bottleneck
- ► Elaboration with VeriPB 6.7× slower than CakePB
- Old version of VERIPB used
  - ► New VeriPB version should improve performance significantly
  - CAKEPB has been also been improved

### **Conclusion & Future Directions**

### Summary:

- ► Proof logging for reformulating optimization problems is possible with VERIPB
  - Can justify preprocessing/presolving techniques
  - Rules preserve optimal value
  - Deleting constraints requires care
  - Also special rule for updating objectives required
- Proof logging for standalone reformulation tools
- Formally verified end-to-end verification for problem reformulations

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#### Future research directions:

- ▶ Proof logging for MIP presolving (integer variables, rational coefficients) [DEGH23]
- Generalize reformulation proofs to enumeration and counting problems

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## Thank you for your attention!

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