Certified Symmetry and Dominance Breaking for Combinatorial Optimisation

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MIAO

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Joint AAAI '22 paper with Bart Bogaerts, Stephan Gocht, and Ciaran McCreesh

Combinatorial Solving and Optimisation

- Revolution last couple of decades in combinatorial solvers for
 - Boolean satisfiability (SAT) solving [BHvMW21]
 - Satisfiability modulo theories (SMT) solving [BHvMW21]
 - Constraint programming (CP) [RvBW06]
 - Mixed integer linear programming (MIP) [AW13, BR07]
- Solve NP problems (or worse) very successfully in practice!
- Except solvers are sometimes wrong... (Even best commercial ones)
 [BLB10, CKSW13, AGJ+18, GSD19, GS19]
- Software testing doesn't suffice to resolve this problem
- Formal verification techniques cannot deal with level of complexity of modern solvers

Design certifying algorithms [ABM+11, MMNS11] that

- not only solve problem but also
- do proof logging to certify that solution is correct

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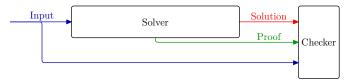


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- Feed input + solution + proof to proof checker

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Workflow:

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- Feed input + solution + proof to proof checker
- Verify that proof checker says solution is correct

Yet Another SAT Success Story

Many proof logging formats for SAT solving using CNF clausal format:

- DRAT [HHW13a, HHW13b, WHH14]
- GRIT [CMS17]
- LRAT [CHH+17]
- ...

Well established — required in main track of SAT competitions

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But efficient proof logging has remained out of reach for stronger paradigms

And, in fact, even for some advanced SAT solving techniques:

- cardinality reasoning
- Gaussian elimination
- symmetry handling

Clausal Proof Logging Approaches

Cardinality and pseudo-Boolean reasoning [SB06, BBH22]

Evaluated on fairly specific crafted benchmarks More challenging and/or real-world benchmarks would be valuable

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Problems with proof logging overhead and proof file size

Symmetry handling [HHW15, TD20]

No fully general method for symmetry breaking (i.e., adding constraints to remove symmetric solutions)

Method for symmetric learning (i.e., adding symmetric versions of derived constraints) not compatible with SAT preprocessing

Our Work: Efficient Proof Logging for Symmetry Breaking

Paper Certified Symmetry and Dominance Breaking for Combinatorial Optimisation at AAAI '22 [BGMN22]:

Implementation in proof checker VeriPB [Ver]

- First general & efficient proof logging method for symmetry breaking
- Supports also pseudo-Boolean reasoning and Gaussian elimination
- Based on 0-1 integer linear constraints instead of clauses
- Uses cutting planes method [CCT87] with additional rules

Outline of Presentation

What I hope to cover in the rest of this presentation:

- Basics of proof logging with 0-1 linear constraints
- New rule for symmetry and dominance breaking
- Application to symmetry breaking for SAT (and some other problems)
- Some future research directions

0-1 Integer Linear (a.k.a. Pseudo-Boolean) Constraints

Pseudo-Boolean (PB) constraints are 0-1 integer linear constraints

$$C \doteq \sum_{i} a_{i} \ell_{i} \geq A$$

- $a_i, A \in \mathbb{Z}$
- literals ℓ_i : x_i or \overline{x}_i (where $x_i + \overline{x}_i = 1$)
- variables x_i take values 0 = false or 1 = true

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Pseudo-Boolean formulas $F \doteq \bigwedge_{i=1}^m C_i$ are conjunctions of pseudo-Boolean constraints

Some Types of Pseudo-Boolean Constraints

Clauses

$$x \vee \overline{y} \vee z \quad \Leftrightarrow \quad x + \overline{y} + z \ge 1$$

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General pseudo-Boolean constraints

$$x_1 + 2\overline{x}_2 + 3x_3 + 4\overline{x}_4 + 5x_5 \ge 7$$

$$\label{eq:linear_combination} \begin{array}{l} \textbf{Literal axioms} \ \hline \\ \hline \\ \ell_i \geq 0 \\ \\ \textbf{Linear combination} \ \hline \\ \frac{\sum_i a_i \ell_i \geq A}{\sum_i (c_A a_i + c_B b_i) \ell_i \geq c_A A + c_B B} \\ \hline \\ \hline \\ \textbf{Division} \ \hline \\ \frac{\sum_i ca_i \ell_i \geq A}{\sum_i a_i \ell_i \geq \lceil A/c \rceil} \\ \hline \\ [c \in \mathbb{N}^+] \\ \hline \end{array}$$

$$2x + 4y + 2z + w \ge 5 \qquad 2x + y + w \ge 2$$
 Lin comb

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$$\text{Lin comb } \frac{2x+4y+2z+w \geq 5}{(2x+4y+2z+w)+2 \cdot (2x+y+w) \geq 5+2 \cdot 2}$$

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Toy example:

$$\begin{array}{c} 2x+4y+2z+w\geq 5 & 2x+y+w\geq 2 \\ \hline \\ \text{Lin comb} & \frac{6x+6y+2z+3w\geq 9}{} \\ \hline \\ \text{Div } & \frac{6x+6y+3w\geq 7}{} \\ \hline \\ 2x+2y+w\geq 3 \end{array}$$

(See [BN21] for more details about cutting planes)

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- Generalize reverse unit propagation (RUP) rule [GN03, Van08] to PB constraints — just convenient shorthand for derivation
- Also need extension rule (analogue of RAT [JHB12]) to deal with, e.g., preprocessing

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C is redundant with respect to F if and only if there is a substitution ω (mapping variables to truth values or literals), called a witness, for which

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- Proof sketch for interesting direction: If α satisfies F but falsifies C, then $\alpha \circ \omega$ satisfies $F \wedge C$
- Implication should be efficiently verifiable every $D \in (F \wedge C) \upharpoonright_{\omega}$ should follow from $F \wedge \neg C$ by, e.g.,
 - weakening (addition of literal axioms $\ell_i \geq 0$)
 - 2 reverse unit propagation (RUP)
 - explicit derivation presented in proof log

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- Pre- and inprocessing [GN21] (since redundance rule subsumes RAT)
- Pseudo-Boolean reasoning (by design)
- Gaussian elimination [GN21]
- Subgraph problems [GMN20, GMM+20]
- Pseudo-Boolean solving via translation to CNF [GMNO22]
- Core-guided MaxSAT solving [VDB22, BBN⁺23]
- Constraint programming [EGMN20, GMN22]
- This talk: Symmetry and dominance breaking [BGMN22]

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More info in tutorial Combinatorial solving with provably correct results (https://youtu.be/s_5BIi4I22w) on pseudo-Boolean proof logging

The Challenge of Symmetries

(Syntactic) symmetry: substitution σ preserving F ($F \upharpoonright_{\sigma} \doteq F$)

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Note that $\sum_i w_i \ell_i < \sum_i w_i \cdot \alpha(\ell_i)$ means $\sum_i w_i \ell_i \le -1 + \sum_i w_i \cdot \alpha(\ell_i)$

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Redundance-based strengthening, optimisation version

Add constraint C to formula F if exists witness substitution ω such that

$$F \wedge \neg C \models (F \wedge C) \upharpoonright_{\omega} \wedge f \upharpoonright_{\omega} \leq f$$

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- If so, don't need to show that $C \upharpoonright_{\omega}$ implied!

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- **7** . . .
- **1** Can't go on forever, so finally reach α' satisfying $F \wedge C$

Strength of Dominance Rule

Dominance-based strengthening (stronger, still simplified)

If C_1,C_2,\ldots,C_{m-1} have been derived from F (maybe using dominance), then can derive also C_m if exists witness substitution ω such that

$$F \wedge \bigwedge_{i=1}^{m-1} C_i \wedge \neg C_m \models F \upharpoonright_{\omega} \wedge f \upharpoonright_{\omega} < f$$

Only consider original formula — no need to show that any $C_i
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Now why is this sound?

- Same inductive proof as before, but nested
- \bullet Or pick α satisfying F and minimizing f and argue by contradiction

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Dominance-based strengthening (stronger, still simplified)

If C_1,C_2,\ldots,C_{m-1} have been derived from F (maybe using dominance), then can derive also C_m if exists witness substitution ω such that

$$F \wedge \bigwedge_{i=1}^{m-1} C_i \wedge \neg C_m \models F \upharpoonright_{\omega} \wedge f \upharpoonright_{\omega} < f$$

Only consider original formula — no need to show that any $C_i \upharpoonright_{\omega}$ implied!

Now why is **this** sound?

- Same inductive proof as before, but nested
- \bullet Or pick α satisfying F and minimizing f and argue by contradiction

Further extensions:

- Define dominance rule w.r.t. order independent of objective function
- Switch between different orders in same proof
- See [BGMN22] for details

Strategy for SAT Symmetry Breaking

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 Derive CNF encoding of lex-leader constraints from PB constraint (in same spirit as [GMNO22])

$$\begin{array}{ll} y_0 & \overline{y}_j \vee \overline{\sigma(x_j)} \vee x_j \\ \overline{y}_{j-1} \vee \overline{x}_j \vee \sigma(x_j) & y_j \vee \overline{y}_{j-1} \vee \overline{x}_j \\ \overline{y}_j \vee y_{j-1} & y_j \vee \overline{y}_{j-1} \vee \sigma(x_j) \end{array}$$

Theorem

 $C_{\sigma} \doteq f \leq f \! \upharpoonright_{\! \sigma}$ can be derived from F using dominance with witness σ

$$F \wedge \neg C_{\sigma} \models F \upharpoonright_{\sigma} \wedge f \upharpoonright_{\sigma} < f$$

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Comparison to DRAT-style proofs

Redundance-based strengthening can be used analogously to [HHW15]

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Comparison to DRAT-style proofs

Redundance-based strengthening can be used analogously to [HHW15]

- ullet but only guaranteed to work for breaking single symmetry σ
- if σ is involution (i.e., its own inverse)
- not known how to deal with symmetries that are complex or interact

Breaking symmetries with the dominance rule

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$$F \wedge C_{\sigma} \wedge \neg C_{\tau} \models F \upharpoonright_{\tau} \wedge f \upharpoonright_{\tau} < f$$

Why does it work?

- Witness need not satisfy all derived constraints
- Sufficient to just produce "better" assignment

Applied Symmetry Breaking for SAT Solving

- Break symmetries with BREAKID and get proofs for symmetry breaking clauses
- 2 Concatenate with CDCL solver proof (DRAT rewritten in VERIPB format)

Applied Symmetry Breaking for SAT Solving

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Short dictionary for DRAT-to-VeriPB translations

DRAT	VeriPB
1	x1
-2	~x2
1 -2 3 0	1 x1 1 ~x2 1 x3 >= 1 ;
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Applied Symmetry Breaking for SAT Solving

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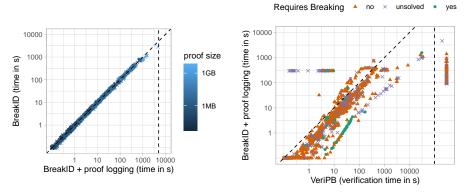
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(Or use KISSAT fork with VERIPB proof logging at https://gitlab.com/MIAOresearch/tools-and-utilities/kissat_fork)

Experimental Evaluation

- Evaluated on SAT competition benchmarks
- BREAKID [DBBD16, Bre] used to find and break symmetries

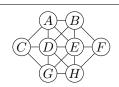


- proof logging overhead negligible
- verification at most 20 times slower than solving for 95% of instances

Symmetry Breaking for Constraint Programming

Crystal Maze puzzle

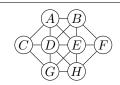
Place numbers 1 to 8 without repetition Adjacent circles mustn't have consecutive numbers



Symmetry Breaking for Constraint Programming

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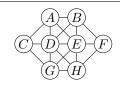
Without loss of generality:

- A < G (horizontal mirror symmetry)
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- $A \le 4$ (value symmetry)

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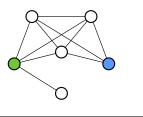
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- $A \le 4$ (value symmetry)

Technical challenge: integer-valued variables See [GMN22] for more detailed discussion

Dominance Breaking for Maximum Clique Solving

Maximum clique solving

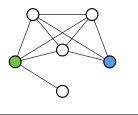
Find largest fully connected component



Dominance Breaking for Maximum Clique Solving

Maximum clique solving

Find largest fully connected component



Lazy global domination [MP16]

Only consider green and not blue vertex (since every neighbour of blue is also neighbour of green)

Technical challenge: vertex domination detected only lazily during search Dominance rule (rather than redundance rule) really helpful here

Performance and reliability of pseudo-Boolean proof logging

- Trim proof while verifying (as in DRAT-TRIM [HHW13a])
- Compress proof file using binary format
- Design formally verified proof checker (work in progress [BMM+23])

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Proof logging for other combinatorial problems and techniques

- Symmetric learning and recycling (substitution) of subproofs
- MaxSAT solving and PB optimization (work in progress [VDB22, BBN⁺23])
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And more...

• Lots of challenging problems and interesting ideas

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And more...

- Lots of challenging problems and interesting ideas
- We're hiring! Talk to me to join the proof logging revolution! ©

Summing up

- Combinatorial solving and optimization is a true success story
- But ensuring correctness is a crucial, and not yet satisfactorily addressed, concern
- Certifying solvers producing machine-verifiable proofs of correctness seems like most promising approach
- Cutting planes reasoning with pseudo-Boolean constraints seems to hit a sweet spot between simplicity and expressivity
- This work: Efficient proof logging for symmetry and dominance breaking using cutting planes with extensions

Summing up

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Thank you for your attention!

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