Solving and Verifying the Boolean Pythagorean Triples Problem via Cube-and-Conquer

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Joint work with
Oliver Kullmann and Victor W. Marek

50 Years of the Hales-Jewett Theorem Bellingham, Washington May 6, 2016

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Introduction

Schur's Theorem [Schur 1917]

Can the set of natural numbers $\{1, 2, ...\}$ be partitioned into k parts with no part containing a, b, and c such that a+b=c? Otherwise, what is the smallest finite counter-example?

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Theorem (Schur's Theorem)

For each k > 0, there exists a number S(k), known as Schur number k, such that [1, S(k)] can be partitioned into k parts with no part containing a, b, and c such that a + b = c, while this is impossible for [1, S(k)+1].

$$S(1) = 1$$
, $S(2) = 4$, $S(3) = 13$, $S(4) = 44$ [Baumert 1965], $160 \le S(5) \le 315$ [Exoo 1994, Fredricksen 1979].

Schur's Theorem on Squares

Can the set of squares $\{1^2, 2^2, 3^2, \dots\}$ be partitioned into k parts with no part containing a, b, and c such that a + b = c?

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The case k = 2 is already very difficult to determine:

```
\begin{array}{l} \{1^2, 2^2, 3^2, 4^2, 6^2, 7^2, 8^2, 9^2, 11^2, 12^2, 13^2, 14^2, 16^2, 17^2, 18^2, 19^2, \dots\} \\ \{5^2, 10^2, 15^2, 20^2, \dots\} \end{array}
```

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$$\{5^2, 10^2, 15^2, 20^2, \dots \}$$

Partitioning the first thousand squares is easy (even manually).

A computer program can partition the first several thousands squares ($\{1^2,\ldots,7664^2\}$) [Cooper and Overstreet 2015].

Can the infinite set of squares be partitioned into two parts?

Pythagorean Triples Problem [Graham]

The Boolean Pythagorean Triples problem is a reformulation of Schur's theorem on squares restricted to two parts:

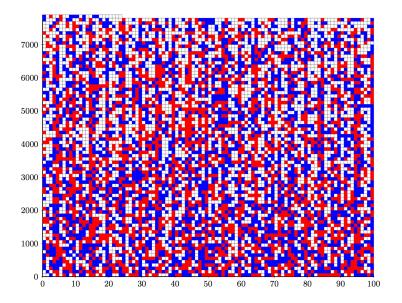
Can the set of natural numbers $\{1,2,3,\dots\}$ be partitioned into two parts such that no part contains a Pythagorean triple $(a,b,c\in\mathbb{N}$ with $a^2+b^2=c^2)$?

A partition into two parts is encoded using Boolean variables x_i with $i \in \{1, 2, 3, ...\}$ such that $x_i = 1$ (= 0) means that i occurs in the Part 1 (Part 2). For each Pythagorean triple (a, b, c) two clauses are added: $(x_a \lor x_b \lor x_c) \land (\bar{x}_a \lor \bar{x}_b \lor \bar{x}_c)$.

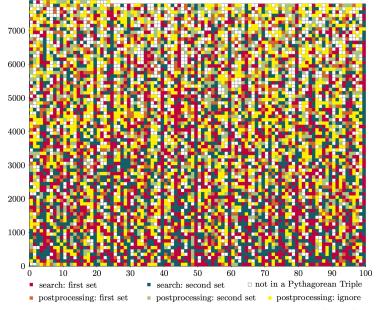
Theorem (Main result via parallel SAT solving)

[1,7824] can be partitioned into two parts, such that no part contains a Pythagorean triple. This is impossible for [1,7825].

An Extreme Solution (a valid partition of [1,7824]) I



An Extreme Solution (a valid partition of [1,7824]) II



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Main Contribution

We present a framework that combines, for the first time, all pieces to produce verifiable SAT results for very hard problems.

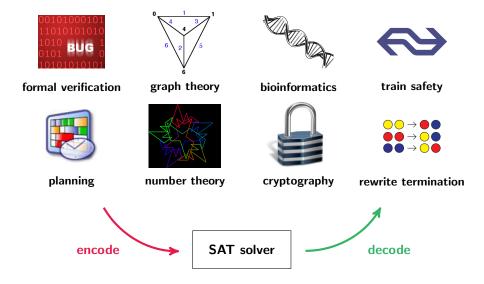
The status quo of using combinatorial solvers and years of computation is arguably intolerable for mathematicians:

- Nouril and Paul [2008] computed the sixth van der Waerden number (W(2,6) = 1132) using dedicated hardware without producing a proof.
- McKay's and Radziszowski's big result [1995] in Ramsey Theory (R(4,5) = 25) still cannot be reproduced.

We demonstrate our framework on the Pythagorean triples problem, potentially the hardest problem solved with SAT yet.

SAT Solving and Verification

Satisfiability (SAT) solving has many applications



A Small SAT Problem: Pythagorean Triples up to n = 55

$$\begin{array}{c} (x_3 \lor x_4 \lor x_5) \land (\bar{x}_3 \lor \bar{x}_4 \lor \bar{x}_5) \land (x_5 \lor x_{12} \lor x_{13}) \land (\bar{x}_5 \lor \bar{x}_{12} \lor \bar{x}_{13}) \land \\ (x_7 \lor x_{24} \lor x_{25}) \land (\bar{x}_7 \lor \bar{x}_{24} \lor \bar{x}_{25}) \land (x_9 \lor x_{40} \lor x_{41}) \land (\bar{x}_9 \lor \bar{x}_{40} \lor \bar{x}_{41}) \land \\ (x_6 \lor x_8 \lor x_{10}) \land (\bar{x}_6 \lor \bar{x}_8 \lor \bar{x}_{10}) \land (x_8 \lor x_{15} \lor x_{17}) \land (\bar{x}_8 \lor \bar{x}_{15} \lor \bar{x}_{17}) \land \\ (x_{10} \lor x_{24} \lor x_{26}) \land (\bar{x}_{10} \lor \bar{x}_{24} \lor \bar{x}_{26}) \land (x_{12} \lor x_{35} \lor x_{37}) \land (\bar{x}_{12} \lor \bar{x}_{35} \lor \bar{x}_{37}) \land \\ (x_{14} \lor x_{48} \lor x_{50}) \land (\bar{x}_{14} \lor \bar{x}_{48} \lor \bar{x}_{50}) \land (x_9 \lor x_{12} \lor x_{15}) \land (\bar{x}_9 \lor \bar{x}_{12} \lor \bar{x}_{15}) \land \\ (x_{15} \lor x_{36} \lor x_{39}) \land (\bar{x}_{15} \lor \bar{x}_{36} \lor \bar{x}_{39}) \land (x_{12} \lor x_{16} \lor x_{20}) \land (\bar{x}_{12} \lor \bar{x}_{16} \lor \bar{x}_{20}) \land \\ (x_{16} \lor x_{30} \lor x_{34}) \land (\bar{x}_{16} \lor \bar{x}_{30} \lor \bar{x}_{34}) \land (x_{20} \lor x_{48} \lor x_{52}) \land (\bar{x}_{20} \lor \bar{x}_{48} \lor \bar{x}_{52}) \land \\ (x_{15} \lor x_{20} \lor x_{25}) \land (\bar{x}_{15} \lor \bar{x}_{20} \lor \bar{x}_{25}) \land (x_{18} \lor x_{24} \lor x_{30}) \land (\bar{x}_{18} \lor \bar{x}_{24} \lor \bar{x}_{30}) \land \\ (x_{24} \lor x_{45} \lor x_{51}) \land (\bar{x}_{24} \lor \bar{x}_{45} \lor \bar{x}_{51}) \land (x_{21} \lor x_{28} \lor x_{35}) \land (\bar{x}_{21} \lor \bar{x}_{28} \lor \bar{x}_{35}) \land \\ (x_{20} \lor x_{21} \lor x_{29}) \land (\bar{x}_{20} \lor \bar{x}_{21} \lor \bar{x}_{29}) \land (x_{24} \lor x_{32} \lor x_{40}) \land (\bar{x}_{24} \lor \bar{x}_{32} \lor \bar{x}_{40}) \land \\ (x_{28} \lor x_{45} \lor x_{53}) \land (\bar{x}_{28} \lor \bar{x}_{45} \lor \bar{x}_{53}) \land (x_{27} \lor x_{36} \lor x_{45}) \land (\bar{x}_{27} \lor \bar{x}_{36} \lor \bar{x}_{45}) \land \\ (x_{30} \lor x_{40} \lor x_{50}) \land (\bar{x}_{30} \lor \bar{x}_{40} \lor \bar{x}_{50}) \land (x_{33} \lor x_{44} \lor x_{55}) \land (\bar{x}_{33} \lor \bar{x}_{44} \lor \bar{x}_{55}) \land (\bar{x}_{33} \lor$$

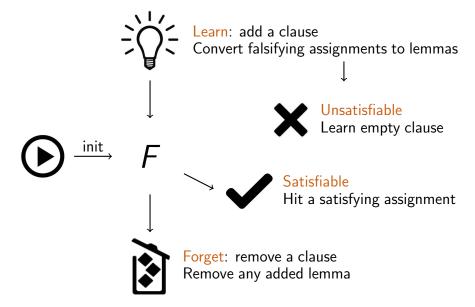
Does there exist an assignment satisfying all clauses of F_{55} ?

Search for a satisfying assignment (or proof none exists)

$$\begin{array}{c} \left(x_{3} \lor x_{4} \lor x_{5} \right) \land \left(\overline{x}_{3} \lor \overline{x}_{4} \lor \overline{x}_{5} \right) \land \left(x_{5} \lor x_{12} \lor x_{13} \right) \land \left(\overline{x}_{5} \lor \overline{x}_{12} \lor \overline{x}_{13} \right) \land \\ \left(x_{7} \lor x_{24} \lor x_{25} \right) \land \left(\overline{x}_{7} \lor \overline{x}_{24} \lor \overline{x}_{25} \right) \land \left(x_{9} \lor x_{40} \lor x_{41} \right) \land \left(\overline{x}_{9} \lor \overline{x}_{40} \lor \overline{x}_{41} \right) \land \\ \left(x_{6} \lor x_{8} \lor x_{10} \right) \land \left(\overline{x}_{6} \lor \overline{x}_{8} \lor \overline{x}_{10} \right) \land \left(x_{8} \lor x_{15} \lor x_{17} \right) \land \left(\overline{x}_{8} \lor \overline{x}_{15} \lor \overline{x}_{17} \right) \land \\ \left(x_{10} \lor x_{24} \lor x_{26} \right) \land \left(\overline{x}_{10} \lor \overline{x}_{24} \lor \overline{x}_{26} \right) \land \left(x_{12} \lor x_{35} \lor x_{37} \right) \land \left(\overline{x}_{12} \lor \overline{x}_{35} \lor \overline{x}_{37} \right) \land \\ \left(x_{14} \lor x_{48} \lor x_{50} \right) \land \left(\overline{x}_{14} \lor \overline{x}_{48} \lor \overline{x}_{50} \right) \land \left(x_{9} \lor x_{12} \lor x_{15} \right) \land \left(\overline{x}_{9} \lor \overline{x}_{12} \lor \overline{x}_{15} \right) \land \\ \left(x_{15} \lor x_{36} \lor x_{39} \right) \land \left(\overline{x}_{15} \lor \overline{x}_{36} \lor \overline{x}_{39} \right) \land \left(x_{12} \lor x_{16} \lor x_{20} \right) \land \left(\overline{x}_{12} \lor \overline{x}_{16} \lor \overline{x}_{20} \right) \land \\ \left(x_{16} \lor x_{30} \lor x_{34} \right) \land \left(\overline{x}_{16} \lor \overline{x}_{30} \lor \overline{x}_{34} \right) \land \left(x_{20} \lor x_{48} \lor x_{52} \right) \land \left(\overline{x}_{20} \lor \overline{x}_{48} \lor \overline{x}_{52} \right) \land \\ \left(x_{15} \lor x_{20} \lor x_{25} \right) \land \left(\overline{x}_{15} \lor \overline{x}_{20} \lor \overline{x}_{25} \right) \land \left(x_{18} \lor x_{24} \lor x_{30} \right) \land \left(\overline{x}_{21} \lor \overline{x}_{28} \lor \overline{x}_{35} \right) \land \\ \left(x_{24} \lor x_{45} \lor x_{51} \right) \land \left(\overline{x}_{24} \lor \overline{x}_{45} \lor \overline{x}_{51} \right) \land \left(x_{21} \lor x_{28} \lor x_{35} \right) \land \left(\overline{x}_{21} \lor \overline{x}_{28} \lor \overline{x}_{35} \right) \land \\ \left(x_{20} \lor x_{21} \lor x_{29} \right) \land \left(\overline{x}_{20} \lor \overline{x}_{21} \lor \overline{x}_{29} \right) \land \left(x_{24} \lor x_{32} \lor x_{40} \right) \land \left(\overline{x}_{27} \lor \overline{x}_{36} \lor \overline{x}_{45} \right) \land \\ \left(x_{28} \lor x_{45} \lor x_{53} \right) \land \left(\overline{x}_{28} \lor \overline{x}_{45} \lor \overline{x}_{53} \right) \land \left(x_{27} \lor x_{36} \lor x_{45} \right) \land \left(\overline{x}_{33} \lor \overline{x}_{44} \lor \overline{x}_{55} \right) \land \\ \left(x_{30} \lor x_{40} \lor x_{50} \right) \land \left(\overline{x}_{30} \lor \overline{x}_{40} \lor \overline{x}_{50} \right) \land \left(x_{33} \lor x_{44} \lor x_{55} \right) \land \left(\overline{x}_{33} \lor \overline{x}_{44} \lor \overline{x}_{55} \right) \land \left(\overline{x}_{33} \lor \overline{x}_{44} \lor$$

Solving F_{55} is easy, but how to solve hard problems, e.g. F_{7825} ?

Simplified Conflict-Driven Clause Learning (CDCL) Solving



Short history of CDCL improvements

Conflict-driven clause learning (CDCL) has been the dominant SAT solving paradigm and improved significantly in 20 years.

- ▶ Invented by Marques-Silva and Sakallah [1997];
- Dedicated data-structure and variable selection heuristics made CDCL really competitive [MoskewiczMZZM 2001];
- Efficient implementation [Een and Sörensson 2003];
- New value selection and rapid restarts make CDCL "local search for UNSAT" [Pipatsrisawat and Darwiche 2007];
- ▶ An alternative restart implementation makes ultra rapid restarts optimal [van der Tak, Ramos, and Heule 2011].

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CDCL is very powerful in solving large "easy" problems, but hard to parallelize effectively (no linear time speed-ups)

How to solve really hard problems efficiently?

Cube-and-Conquer [Heule, Kullmann, Wieringa, and Biere 2011]

There exists two main SAT solving paradigms:

- Conflict-driven clause-learning (CDCL) aims to find a short refutation using (cheap) local heuristics.
- ► Look-ahead aims to construct a small binary search-tree using (expensive) global heuristics.

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- ► Look-ahead aims to construct a small binary search-tree using (expensive) global heuristics.

The combination: Create a tautological DNF using look-ahead techniques and solve the problem multiple times using CDCL each time under the assumption that a given cube is true.

Example (Partitioning using a tautological DNF)

Given a formula F and a DNF $D=(a \wedge b) \vee (a \wedge \overline{b}) \vee (\overline{a})$. Instead of solving CDCL (F), we solve CDCL $(F \wedge (a \wedge b))$, CDCL $(F \wedge (a \wedge \overline{b}))$, and CDCL $(F \wedge (\overline{a}))$.

The approaches are equivalent if and only if D is a tautology.

Cube-and-Conquer [Heule, Kullmann, Wieringa, and Biere 2011]

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- ► Look-ahead aims to construct a small binary search-tree using (expensive) global heuristics.

Combining look-ahead and CDCL, called cube-and-conquer, does not work out of the box. Crucial details are:

- ▶ Partition a given formula into many (millions) of subproblems. When just a few subproblems are created, say only 32, the performance could actually decrease.
- ► Use heuristics to creates equally hard subproblems, i.e., not simply using the depth of the search-tree.

Cube-and-conquer solves many hard-combinatorial problems significantly faster than both pure CDCL and pure look-ahead.

Search for a satisfying assignment (or proof none exists)

$$\begin{array}{c} (x_3 \lor x_4 \lor x_5) \land (\bar{x}_3 \lor \bar{x}_4 \lor \bar{x}_5) \land (x_5 \lor x_{12} \lor x_{13}) \land (\bar{x}_5 \lor \bar{x}_{12} \lor \bar{x}_{13}) \land \\ (x_7 \lor x_{24} \lor x_{25}) \land (\bar{x}_7 \lor \bar{x}_{24} \lor \bar{x}_{25}) \land (x_9 \lor x_{40} \lor x_{41}) \land (\bar{x}_9 \lor \bar{x}_{40} \lor \bar{x}_{41}) \land \\ (x_6 \lor x_8 \lor x_{10}) \land (\bar{x}_6 \lor \bar{x}_8 \lor \bar{x}_{10}) \land (x_8 \lor x_{15} \lor x_{17}) \land (\bar{x}_8 \lor \bar{x}_{15} \lor \bar{x}_{17}) \land \\ (x_{10} \lor x_{24} \lor x_{26}) \land (\bar{x}_{10} \lor \bar{x}_{24} \lor \bar{x}_{26}) \land (x_{12} \lor x_{35} \lor x_{37}) \land (\bar{x}_{12} \lor \bar{x}_{35} \lor \bar{x}_{37}) \land \\ (x_{14} \lor x_{48} \lor x_{50}) \land (\bar{x}_{14} \lor \bar{x}_{48} \lor \bar{x}_{50}) \land (x_9 \lor x_{12} \lor x_{15}) \land (\bar{x}_9 \lor \bar{x}_{12} \lor \bar{x}_{15}) \land \\ (x_{15} \lor x_{36} \lor x_{39}) \land (\bar{x}_{15} \lor \bar{x}_{36} \lor \bar{x}_{39}) \land (x_{12} \lor x_{16} \lor x_{20}) \land (\bar{x}_{12} \lor \bar{x}_{16} \lor \bar{x}_{20}) \land \\ (x_{16} \lor x_{30} \lor x_{34}) \land (\bar{x}_{16} \lor \bar{x}_{30} \lor \bar{x}_{34}) \land (x_{20} \lor x_{48} \lor x_{52}) \land (\bar{x}_{20} \lor \bar{x}_{48} \lor \bar{x}_{52}) \land \\ (x_{15} \lor x_{20} \lor x_{25}) \land (\bar{x}_{15} \lor \bar{x}_{20} \lor \bar{x}_{25}) \land (x_{18} \lor x_{24} \lor x_{30}) \land (\bar{x}_{18} \lor \bar{x}_{24} \lor \bar{x}_{30}) \land \\ (x_{24} \lor x_{45} \lor x_{51}) \land (\bar{x}_{24} \lor \bar{x}_{45} \lor \bar{x}_{51}) \land (x_{21} \lor x_{28} \lor x_{35}) \land (\bar{x}_{21} \lor \bar{x}_{28} \lor \bar{x}_{35}) \land \\ (x_{20} \lor x_{21} \lor x_{29}) \land (\bar{x}_{20} \lor \bar{x}_{21} \lor \bar{x}_{29}) \land (x_{24} \lor x_{32} \lor x_{40}) \land (\bar{x}_{24} \lor \bar{x}_{32} \lor \bar{x}_{40}) \land \\ (x_{28} \lor x_{45} \lor x_{53}) \land (\bar{x}_{28} \lor \bar{x}_{45} \lor \bar{x}_{53}) \land (x_{27} \lor x_{36} \lor x_{45}) \land (\bar{x}_{27} \lor \bar{x}_{36} \lor \bar{x}_{45}) \land \\ (x_{30} \lor x_{40} \lor x_{50}) \land (\bar{x}_{30} \lor \bar{x}_{40} \lor \bar{x}_{50}) \land (x_{33} \lor x_{44} \lor x_{55}) \land (\bar{x}_{33} \lor \bar{x}_{44} \lor \bar{x}_{55}) \land (\bar{x}_{33} \lor$$

Solutions are easy to verify, but what about unsatisfiability?

Motivation for validating unsatisfiability proofs

Satisfiability solvers are used in amazing ways...

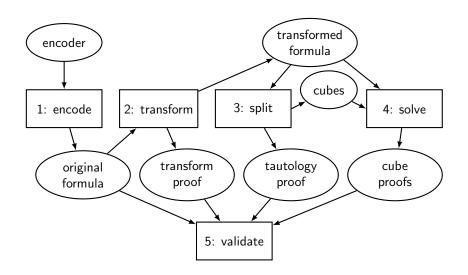
- Hardware and software verification (Intel and Microsoft)
- Hard-Combinatorial problems:
 - van der Waerden numbers
 [Dransfield, Marek, and Truszczynski, 2004; Kouril and Paul, 2008]
 - ► Gardens of Eden in Conway's Game of Life
 [Hartman, Heule, Kwekkeboom, and Noels, 2013]
 - ► Erdős Discrepancy Problem [Konev and Lisitsa, 2014]

..., but satisfiability solvers have errors and only return yes/no.

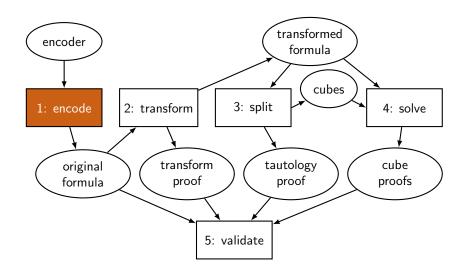
- ► Documented bugs in SAT, SMT, and QBF solvers
 [Brummayer and Biere, 2009; Brummayer et al., 2010]
- Implementation errors often imply conceptual errors
- ► Mathematical results require a stronger justification than a simple yes/no by a solver. UNSAT must be checkable.

Solving Framework for Hard-Combinatorial Problems

Overview of Solving Framework



Overview of Solving Framework: Phase 1



Phase 1: Encode

Input: encoder program

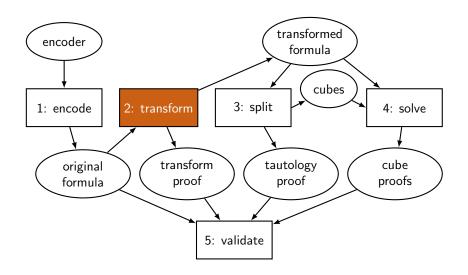
Output: the "original" CNF formula

Goal: make the translation to SAT as simple as possible

```
for (int a = 1; a <= n; a++)
  for (int b = a; b <= n; b++) {
    int c = sqrt (a*a + b*b);
    if ((c <= n) && ((a*a + b*b) == (c*c))) {
      addClause (a, b, c);
      addClause (-a, -b, -c); }
}</pre>
```

 F_{7824} has 6492 (occurring) variables and 18930 clauses, and F_{7825} has 6494 (occurring) variables and 18944 clauses.

Overview of Solving Framework: Phase 2



Phase 2: Transform

Input: original CNF formula

Output: transformed CNF formula and a transformation proof

Goal: optimize the formula regarding the later (solving) phases

We applied two transformations:

- Pythagorean Triple Elimination removes Pythagorean Triples that contain an element that does not occur in any other Pythagorean Triple, e.g. $3^2 + 4^2 = 5^2$. (till fixpoint)
- ▶ Symmetry breaking places the number most frequently occurring in Pythagorean triples (2520) in Part 1 (encode).

All transformation (pre-processing) techniques can be expressed using RAT steps [Järvisalo, Heule, and Biere 2012].

Phase 2: Blocked Clauses [Kullmann'99]

Definition (Blocking literal)

A literal I in a clause C of a CNF F blocks C w.r.t. F if for every clause $D \in F_{\overline{I}}$, the resolvent $(C \setminus \{I\}) \cup (D \setminus \{\overline{I}\})$ obtained from resolving C and D on I is a tautology.

With respect to a fixed CNF and its clauses we have:

Definition (Blocked clause)

A clause is blocked if it contains a literal that blocks it.

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Example (Blocking literals and blocked clauses)

Consider the formula $(a \lor b) \land (a \lor \bar{b} \lor \bar{c}) \land (\bar{a} \lor c)$.

First clause is not blocked.

Second clause is blocked by both a and \bar{c} .

Third clause is blocked by c.

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Third clause is blocked by c.

Proposition

Removal of an arbitrary blocked clause preserves satisfiability.



Phase 2: Blocked Clause Elimination (BCE)

Definition (BCE)

While a clause C in a formula F is blocked, remove C from F.

Example (BCE)

Consider $(a \lor b) \land (a \lor \overline{b} \lor \overline{c}) \land (\overline{a} \lor c)$. After removing either $(a \lor \overline{b} \lor \overline{c})$ or $(\overline{a} \lor c)$, the clause $(a \lor b)$ becomes blocked (no clause with either \overline{b} or \overline{a}).

An extreme case in which BCE removes all clauses!

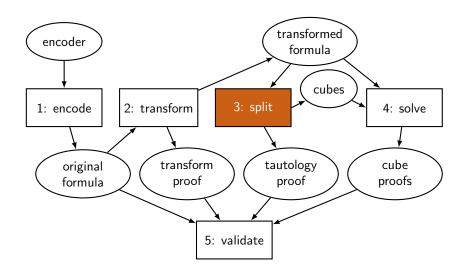
Example (Pythagorean Triples)

The clauses $(x_3 \lor x_4 \lor x_5)$ and $(\bar{x}_3 \lor \bar{x}_4 \lor \bar{x}_5)$ are blocked in F_{7824} and F_{7825} (actually in any F_n). BCE (F_{55}) is empty. BCE (F_{7824}) has 3740 variables and 14652 clauses, and BCE (F_{7825}) has 3745 variables and 14672 clauses.

BCE can simulate many high-level reasoning techniques.

[Järvisalo, Biere, and Heule 2010]

Overview of Solving Framework: Phase 3



Phase 3: Split

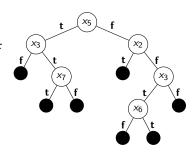
Input: transformed formula

Output: cubes and tautology proof

Goal: partition input in as many subproblems such that total wallclock time is minimal

Two layers of splitting F_{7824} :

- The top level split partitions the transformed formula into exactly a million subproblems;
- Each subproblem is partitioned into tens of thousands of subsubproblems.
 Total time: 35,000 CPU hours



$$D = (x_5 \wedge \bar{x}_3) \vee (x_5 \wedge x_3 \wedge x_7) \vee (x_5 \wedge x_3 \wedge \bar{x}_7) \vee (\bar{x}_5 \wedge x_2) \vee (\bar{x}_5 \wedge \bar{x}_2 \wedge x_3 \wedge \bar{x}_6) \vee (\bar{x}_5 \wedge \bar{x}_2 \wedge x_3 \wedge x_6) \vee (\bar{x}_5 \wedge \bar{x}_2 \wedge \bar{x}_3)$$

Details regarding the splitting heuristics

Splitting based on look-aheads:

- Count the number of assigned variables (large formulas)
- ► Count the number of clause reductions (medium formulas)
- Sum of the weighted new binary clauses (3-SAT)

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Equations (init, average, and update) for 3-SAT heuristics h_4 :

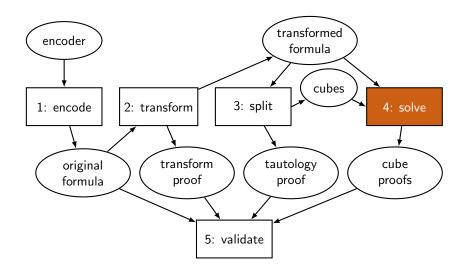
$$h_0(x) = h_0(\bar{x}) = 1$$

$$\mu_i = \frac{1}{2n} \sum_{x \in \mathsf{var}(F)} \left(h_i(x) + h_i(\bar{x}) \right)$$

$$h_{i+1}(x) = \max(\alpha, \min(\beta, \sum_{(x \vee y \vee z) \in F} \left(\frac{h_i(\bar{y})}{\mu_i} \cdot \frac{h_i(\bar{z})}{\mu_i}\right) + \gamma \sum_{(x \vee y) \in F} \frac{h_i(\bar{y})}{\mu_i})).$$

Rnd: $\alpha = 0.01, \beta = 25, \gamma = 3.3$ Ptn: $\alpha = 8, \beta = 550, \gamma = 25$

Overview of Solving Framework: Phase 4



Phase 4: Solve

Input: transformed formula and cubes

Output: cube proofs (or satisfying assignment)

Goal: solve (with proof logging) all cubes as fast as possible

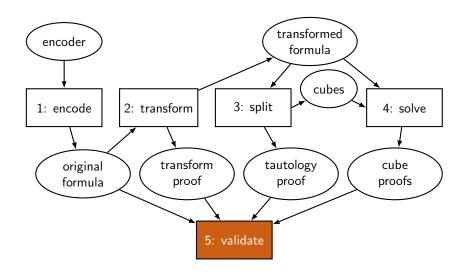
Let φ_i be the i^{th} cube with $i \in [1, 1000000]$.

We first solved all $F_{7824} \wedge \varphi_i$, total runtime was 13,000 CPU hours (less than a day on the cluster). One cube is satisfiable.

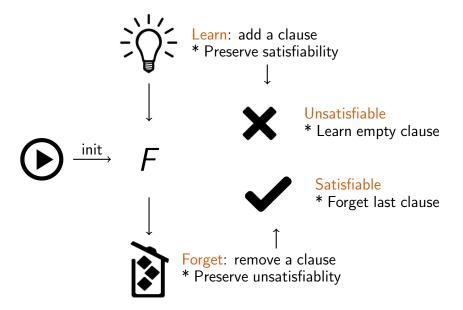
The backbone of a formula is the set of literals that are assigned to true in all solutions. The backbone of F_{7824} after symmetry breaking consists of 2304 literals, including

- x_{5180} and x_{5865} , while $5180^2 + 5865^2 = 7825^2$
- \bar{x}_{625} and \bar{x}_{7800} , while $625^2 + 7800^2 = 7825^2$

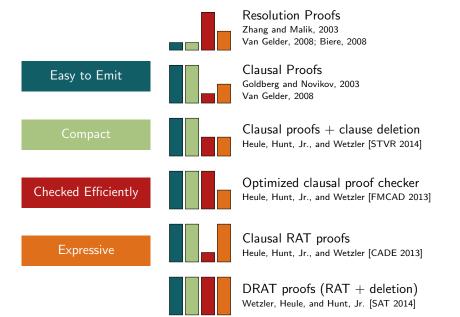
Overview of Solving Framework: Phase 5



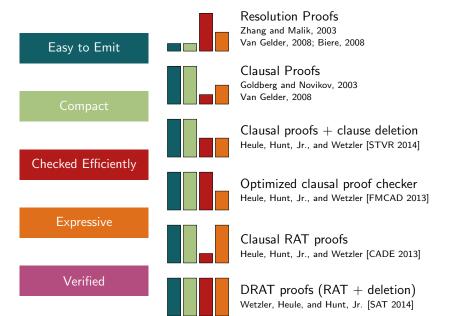
Clausal Proof System [Järvisalo, Heule, and Biere 2012]



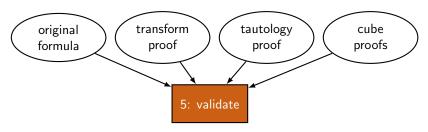
Ideal Properties of a Proof System for SAT Solvers



Ideal Properties of a Proof System for SAT Solvers



Phase 5: Validate Pythagorean Triples Proofs.



We check the proofs with the DRAT-trim checker, which has been used to validate the UNSAT results of the international SAT Competitions since 2013.

Recently it was shown how to validate DRAT proofs in parallel [Heule and Biere 2015].

The size of the merged proof is almost 200 terabyte and has been validated in 16,000 CPU hours.

Conclusions and Future Work

Conclusions

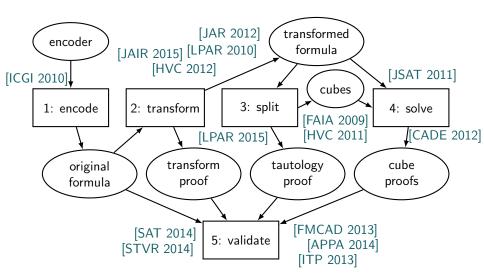
Theorem (Main result)

[1,7824] can be partitioned into two parts, such that no part contains a Pythagorean triple. This is impossible for [1,7825].

We solved and verified the theorem via SAT solving:

- Cube-and-conquer facilitated massive parallel solving.
- ▶ A new look-ahead heuristic was developed to substantially reduce the search space.
- ▶ The proof is huge (200 terabyte), but can be compressed to 68 gigabyte (13,000 CPU hours to decompress) and be validated in 16,000 CPU hours.

Heule's Contributions to Solving Framework



Future Directions

Apply our solving framework to other challenges in Ramsey Theory and elsewhere:

- Existing results for which no proof was produced, for example W(2,6) = 1132 [Kouril and Paul 2008].
- ▶ Century-old open problems appear solvable now, e.g. S(5).

Look-ahead heuristics are crucial and we had to develop dedicated heuristics to solve the Pythagorean triples problem.

- Develop powerful heuristics that work out of the box.
- Alternatively, add heuristic-tuning techniques to the tool chain [Hoos 2012].

Develop a mechanically-verified, fast clausal proof checker.

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Thanks!