

Computable Short Proofs

Marijn J.H. Heule



joint work with Benjamin Kiesl and Armin Biere

Oaxaca SAT workshop

August 27, 2018

"The Largest Math Proof Ever" [Nature 2016]

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143

Posted by [BeauHD](#) on Monday May 30, 2016 @08:10PM from the red-pill-and-blue-pill dept.

THE CONVERSATION

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SPIEGEL ONLINE

Introduction on Proofs

Interference-Based Proof Systems

Without New Variables

Shorter Clauses

Satisfaction-Driven Clause Learning (SDCL)

One More Thing...

Challenges and Conclusions

Introduction on Proofs

Certifying Satisfiability and Unsatisfiability

- Certifying **satisfiability** of a formula is easy:

$$(x \vee y) \wedge (x \vee \bar{y}) \wedge (\bar{y} \vee \bar{z})$$

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- Just consider a **satisfying assignment**: $x\bar{y}z$

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- Certifying **unsatisfiability** is not so easy:

- If a formula has n variables, there are 2^n possible assignments.
- ➔ Checking whether **every** assignment falsifies the formula is **costly**.
- More compact certificates of unsatisfiability are desirable.
 - ➔ Proofs

What Is a Proof in SAT?

- In general, a **proof** is a **string** that **certifies the unsatisfiability** of a formula.
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 - Proofs are **efficiently** (usually **polynomial-time**) **checkable**...
... but can be of exponential size with respect to a formula.

■ **Example:** Resolution (RES) proofs

- A **resolution proof** is a sequence C_1, \dots, C_m of clauses.
- Every clause is either contained in the formula or derived from two earlier clauses via the **resolution rule**:

$$\frac{C \vee x \quad \bar{x} \vee D}{C \vee D}$$

- C_m is the **empty clause** (containing no literals), denoted by \perp .
- There exists a resolution proof for every unsatisfiable formula.

Resolution Proofs

■ Example: $F = (\bar{x} \vee \bar{y} \vee z) \wedge (\bar{z}) \wedge (x \vee \bar{y}) \wedge (\bar{u} \vee y) \wedge (u)$

■ Resolution proof:

$(\bar{x} \vee \bar{y} \vee z), (\bar{z}), (\bar{x} \vee \bar{y}), (x \vee \bar{y}), (\bar{y}), (\bar{u} \vee y), (\bar{u}), (u), \perp$

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$$\frac{\frac{\frac{\bar{x} \vee \bar{y} \vee z \quad \bar{z}}{\bar{x} \vee \bar{y}} \quad x \vee \bar{y}}{\bar{y}} \quad \bar{u} \vee y}{\bar{u}} \quad u}{\perp}$$

Resolution Proofs

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■ **Drawbacks** of resolution:

- For **many** seemingly simple formulas, there are **only** resolution proofs of **exponential size**.
- **State-of-the-art solving techniques** are **not succinctly expressible**.

To cope with these drawbacks, we need advanced techniques...

Interference-Based Proof Systems

Traditional Proofs vs. Interference-Based Proofs

- In **traditional** proof systems, everything that is **inferred**, is **logically implied** by the premises.

$$\frac{C \vee x \quad \bar{x} \vee D}{C \vee D} \text{ (res)} \qquad \frac{A \quad A \rightarrow B}{B} \text{ (mp)}$$

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- ➔ Inference rules reason about the **presence** of facts.
 - If certain premises are present, infer the conclusion.
- **Different approach**: Allow **not only implied conclusions**.
 - **Require only** that the addition of facts preserves **satisfiability**.
 - Reason also about the **absence** of facts.
- ➔ This leads to **interference-based proof systems**.

Reasoning about Absence is as old as SAT Solving

The early SAT decision procedures used the **Pure Literal rule** [Davis and Putnam 1960; Davis, Logemann and Loveland 1962]:

$$\frac{\bar{x} \notin F}{(x)} \text{ (pure)}$$

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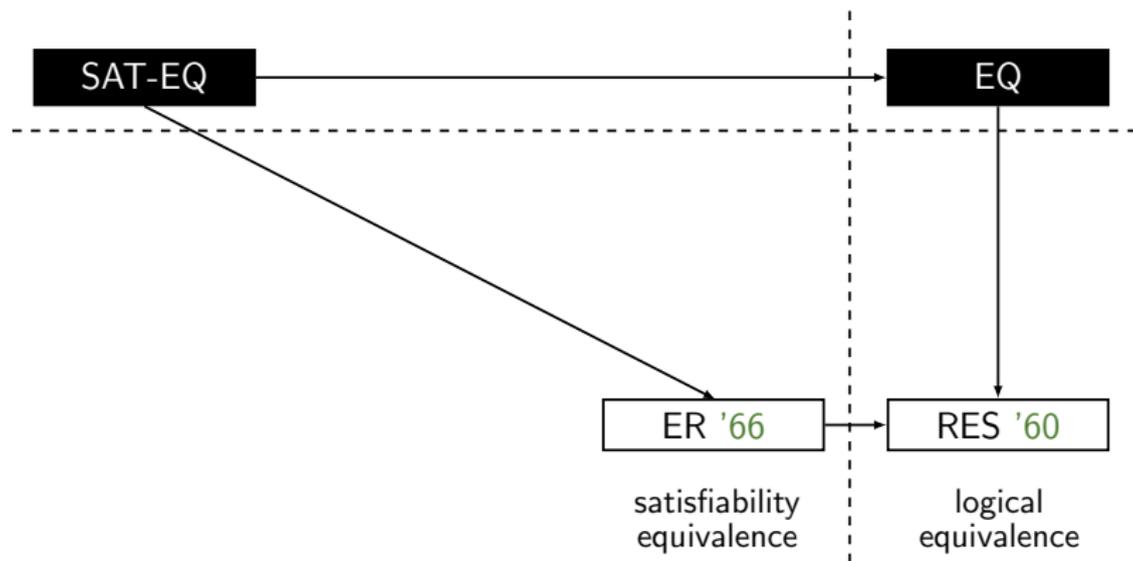
Extended Resolution (ER) [Tseitin 1966]

- Combines resolution with the **Extension rule**:

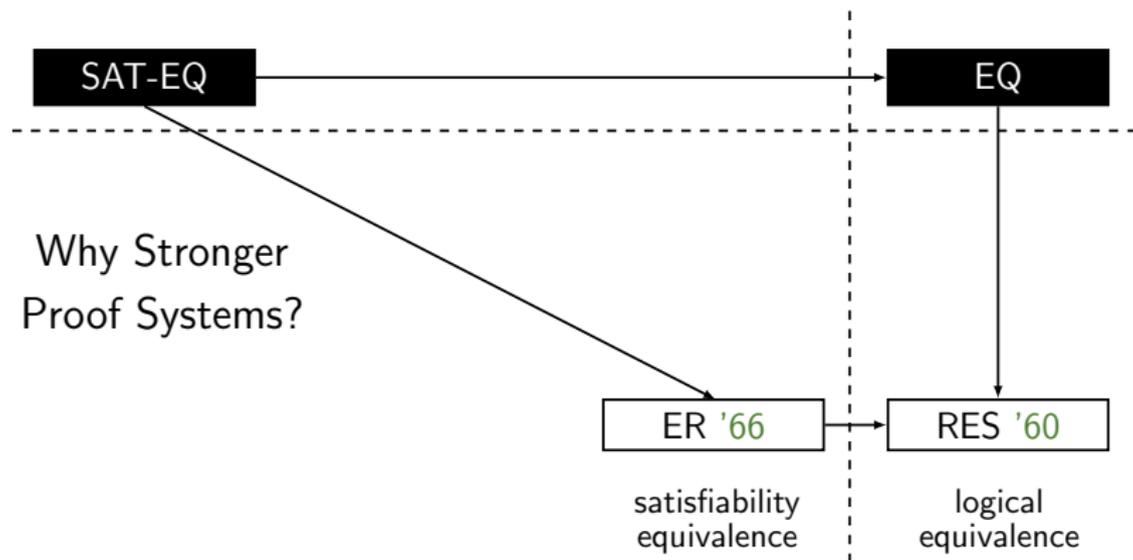
$$\frac{x \notin F \quad \bar{x} \notin F}{(x \vee \bar{a} \vee \bar{b}) \wedge (\bar{x} \vee a) \wedge (\bar{x} \vee b)} \text{ (er)}$$

- Equivalently, adds the definition $x := \text{AND}(a, b)$
- Can be considered the **first interference-based proof system**
- Is very powerful: **No known lower bounds**

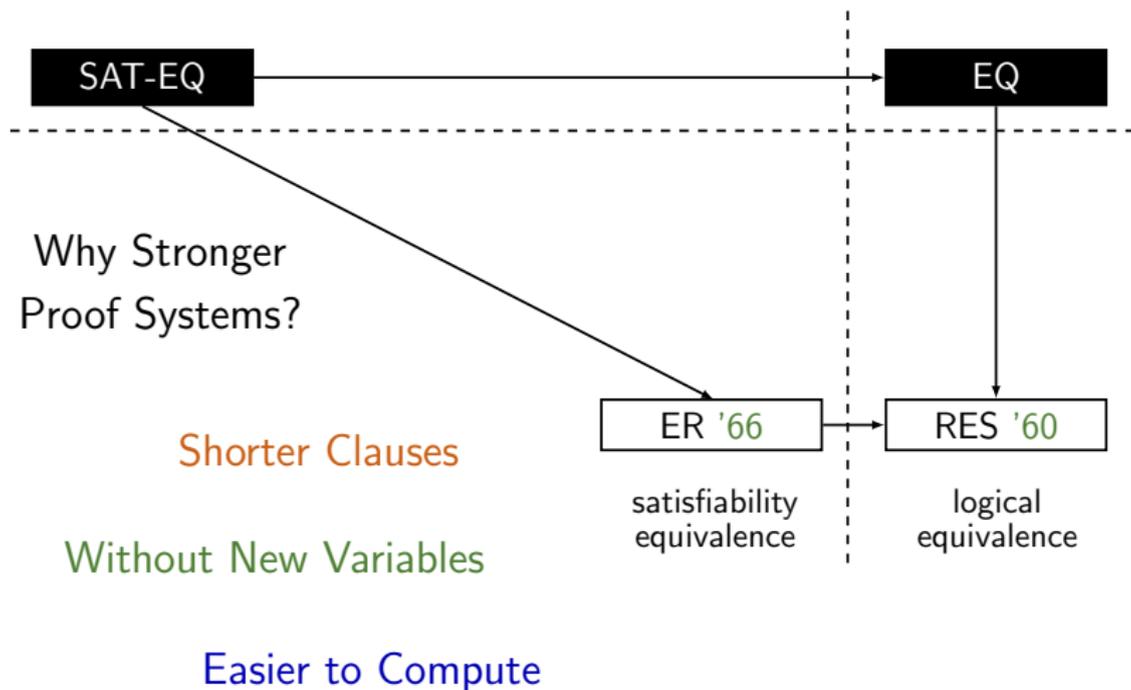
Classical Proof Systems for Propositional Logic



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Without New Variables

Short Proofs of Pigeon Hole Formulas [Cook 1967]

Can $n+1$ pigeons be placed in n holes (at-most-one pigeon per hole)?

$$PHP_n := \bigwedge_{1 \leq p \leq n+1} (x_{1,p} \vee \dots \vee x_{n,p}) \wedge \bigwedge_{1 \leq h \leq n} \bigwedge_{1 \leq p < q \leq n+1} (\bar{x}_{h,p} \vee \bar{x}_{h,q})$$

Resolution proofs of PHP_n formulas are **exponential** [Haken 1985]

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However, these proofs require introducing new variables:

- Hard to find such proofs automatically
- Existing ER approaches produce exponentially large proofs
- How to get rid of this hurdle? First approach: blocked clauses...

Blocked Clauses [Kullmann 1999]

Definition (Blocking literal)

A literal x blocks clause $(C \vee x)$ w.r.t. a CNF formula F if for every clause $(D \vee \bar{x}) \in F$, the resolvent $C \vee D$ is a tautology.

Definition (Blocked clause)

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Example

Consider the formula $(a \vee b) \wedge (a \vee \bar{b} \vee \bar{c}) \wedge (\bar{a} \vee c)$.

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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First clause is not blocked.

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

Third clause is blocked by c .

Theorem

Adding or removing a blocked clause preserves satisfiability.

Blocked Clause Addition and Blocked Clause Elimination

The Blocked Clause proof system (BC) combines the resolution rule with the addition of blocked clauses.

- BC generalizes ER [Kullmann 1999]

- Recall

$$\frac{x \notin F \quad \bar{x} \notin F}{(x \vee \bar{a} \vee \bar{b}) \wedge (\bar{x} \vee a) \wedge (\bar{x} \vee b)} \text{ (er)}$$

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Blocked clause elimination used in preprocessing and inprocessing

- Simulates many circuit optimization techniques [JAR 2012]
- Removes redundant Pythagorean Triples [SAT 2016]

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However, blocked clauses do not offer enough expressivity

- Increase expressivity with autarky-based reasoning...

Autarkies [Monien and Speckenmeyer 1985]

An **autarky** is an assignment that satisfies every clause it touches.

A **pure literal** and a **satisfying assignment** are autarkies.

Example

Consider the formula $F := (x \vee y) \wedge (x \vee \bar{y}) \wedge (\bar{y} \vee \bar{z})$.

Assignment $\alpha_1 = \bar{z}$ is an autarky: $(x \vee y) \wedge (x \vee \bar{y}) \wedge (\bar{y} \vee \bar{z})$.

Assignment $\alpha_2 = x \bar{y} z$ is an autarky: $(x \vee y) \wedge (x \vee \bar{y}) \wedge (\bar{y} \vee \bar{z})$.

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Given an assignment α , $F|_{\alpha}$ denotes a formula F without the clauses satisfied by α and without the literals falsified by α .

Theorem

Let α be an autarky for formula F .

Then, F and $F|_{\alpha}$ are satisfiability equivalent.

Conditional Autarkies [Heule, Kiesl, Seidl, Biere 2017]

An assignment $\alpha = \alpha_{\text{con}} \cup \alpha_{\text{aut}}$ is a **conditional autarky** for formula F if α_{aut} is an autarky for $F | \alpha_{\text{con}}$.

Example

Consider the formula $F := (x \vee y) \wedge (x \vee \bar{y}) \wedge (\bar{y} \vee \bar{z})$.

Let $\alpha_{\text{con}} = x$ and $\alpha_{\text{aut}} = \bar{y}$, then $\alpha = \alpha_{\text{con}} \cup \alpha_{\text{aut}} = x \bar{y}$ is a conditional autarky for F :

$$\alpha_{\text{aut}} = \bar{y} \text{ is an autarky for } F | \alpha_{\text{con}} = (\bar{y} \vee \bar{z}).$$

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Theorem

Let $\alpha = \alpha_{\text{con}} \cup \alpha_{\text{aut}}$ be a **conditional autarky** for formula F .
Then F and $F \wedge (\alpha_{\text{con}} \rightarrow \alpha_{\text{aut}})$ are satisfiability-equivalent.

In the above example, we could therefore learn $(\bar{x} \vee \bar{y})$.

Conditional Autarkies and Blocked Clauses

Blocked clauses and conditional autarkies are strongly related:

Theorem

A clause $(c_1 \vee \dots \vee c_i \vee x)$ is *blocked* on x w.r.t. formula F if and only if $\bar{c}_1 \wedge \dots \wedge \bar{c}_i \wedge x$ is a *conditional autarky* for F .

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The *blocking literal set* generalizes the blocking literal concept resulting in *set-blockedness* [Kiesl, Seidl, Tompits, Biere 2016]

Theorem ([Heule, Kiesl, Seidl, Biere 2017])

A clause $(c_1 \vee \dots \vee c_i \vee x_1 \vee \dots \vee x_k)$ is *set-blocked* (SBC) on literal set $\{x_1, \dots, x_k\}$ w.r.t. formula F if and only if $\bar{c}_1 \wedge \dots \wedge \bar{c}_i \wedge x_1 \wedge \dots \wedge x_k$ is a *conditional autarky* for F .

Conditional Autarkies and Pigeon Hole Formulas

$$\begin{aligned} & (x_{1,1} \vee x_{2,1} \vee \dots \vee x_{n,1}) \quad \wedge \quad (x_{1,3} \vee x_{2,3} \vee x_{3,3} \vee \dots \vee x_{n,3}) \wedge \\ & (x_{1,2} \vee x_{2,2} \vee \dots \vee x_{n,2}) \wedge \dots \wedge (x_{1,n+1} \vee x_{2,n+1} \vee x_{3,n+1} \vee \dots \vee x_{n,n+1}) \wedge \\ & (\bar{x}_{1,1} \vee \bar{x}_{1,2}) \wedge (\bar{x}_{2,1} \vee \bar{x}_{2,2}) \wedge (\bar{x}_{3,1} \vee \bar{x}_{3,2}) \wedge \dots \wedge (\bar{x}_{n,1} \vee \bar{x}_{n,2}) \wedge \\ & (\bar{x}_{1,1} \vee \bar{x}_{1,3}) \wedge (\bar{x}_{2,1} \vee \bar{x}_{2,3}) \wedge (\bar{x}_{3,1} \vee \bar{x}_{3,3}) \wedge \dots \wedge (\bar{x}_{n,1} \vee \bar{x}_{n,3}) \wedge \\ & (\bar{x}_{1,2} \vee \bar{x}_{1,3}) \wedge (\bar{x}_{2,2} \vee \bar{x}_{2,3}) \wedge (\bar{x}_{3,2} \vee \bar{x}_{3,3}) \wedge \dots \wedge (\bar{x}_{n,2} \vee \bar{x}_{n,3}) \wedge \\ & \quad \dots \quad \wedge \quad \dots \\ & (\bar{x}_{1,n} \vee \bar{x}_{1,n+1}) \wedge (\bar{x}_{2,n} \vee \bar{x}_{2,n+1}) \wedge (\bar{x}_{3,n} \vee \bar{x}_{3,n+1}) \wedge \dots \wedge (\bar{x}_{n,n} \vee \bar{x}_{n,n+1}) \end{aligned}$$

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- Consider $\alpha_{\text{con}} = \bar{x}_{1,3} \wedge \bar{x}_{1,4} \wedge \dots \wedge \bar{x}_{1,n+1} \wedge \bar{x}_{2,3} \wedge \bar{x}_{2,4} \wedge \dots \wedge \bar{x}_{2,n+1}$

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- Notice that $\alpha_{\text{aut}} = x_{1,1} \wedge \bar{x}_{1,2} \wedge \bar{x}_{2,1} \wedge x_{2,2}$ is an autarky of $F \mid \alpha_{\text{con}}$

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$$\begin{aligned} & (x_{1,1} \vee x_{2,1} \vee \dots \vee x_{n,1}) \quad \wedge \quad (x_{1,3} \vee x_{2,3} \vee x_{3,3} \vee \dots \vee x_{n,3}) \wedge \\ & (x_{1,2} \vee x_{2,2} \vee \dots \vee x_{n,2}) \wedge \dots \wedge (x_{1,n+1} \vee x_{2,n+1} \vee x_{3,n+1} \vee \dots \vee x_{n,n+1}) \wedge \\ & (\bar{x}_{1,1} \vee \bar{x}_{1,2}) \wedge (\bar{x}_{2,1} \vee \bar{x}_{2,2}) \wedge (\bar{x}_{3,1} \vee \bar{x}_{3,2}) \wedge \dots \wedge (\bar{x}_{n,1} \vee \bar{x}_{n,2}) \wedge \\ & (\bar{x}_{1,1} \vee \bar{x}_{1,3}) \wedge (\bar{x}_{2,1} \vee \bar{x}_{2,3}) \wedge (\bar{x}_{3,1} \vee \bar{x}_{3,3}) \wedge \dots \wedge (\bar{x}_{n,1} \vee \bar{x}_{n,3}) \wedge \\ & (\bar{x}_{1,2} \vee \bar{x}_{1,3}) \wedge (\bar{x}_{2,2} \vee \bar{x}_{2,3}) \wedge (\bar{x}_{3,2} \vee \bar{x}_{3,3}) \wedge \dots \wedge (\bar{x}_{n,2} \vee \bar{x}_{n,3}) \wedge \\ & \quad \dots \quad \wedge \quad \dots \\ & (\bar{x}_{1,n} \vee \bar{x}_{1,n+1}) \wedge (\bar{x}_{2,n} \vee \bar{x}_{2,n+1}) \wedge (\bar{x}_{3,n} \vee \bar{x}_{3,n+1}) \wedge \dots \wedge (\bar{x}_{n,n} \vee \bar{x}_{n,n+1}) \end{aligned}$$

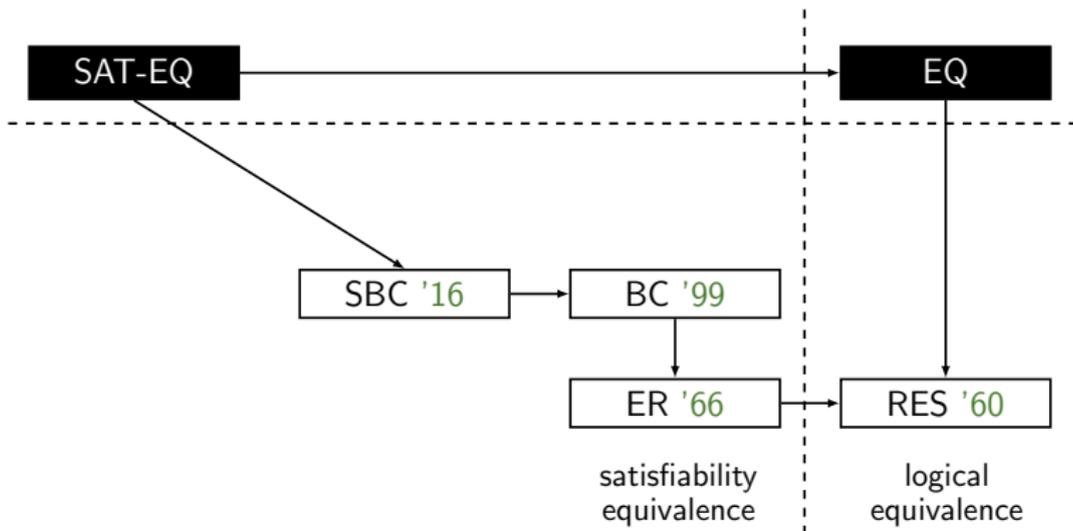
- Consider $\alpha_{\text{con}} = \bar{x}_{1,3} \wedge \bar{x}_{1,4} \wedge \dots \wedge \bar{x}_{1,n+1} \wedge \bar{x}_{2,3} \wedge \bar{x}_{2,4} \wedge \dots \wedge \bar{x}_{2,n+1}$
- Notice that $\alpha_{\text{aut}} = x_{1,1} \wedge \bar{x}_{1,2} \wedge \bar{x}_{2,1} \wedge x_{2,2}$ is an autarky of $F \mid \alpha_{\text{con}}$
- Resolution can reduce the constraint $\alpha_{\text{con}} \rightarrow \alpha_{\text{aut}}$ to $(\bar{x}_{1,2} \vee \bar{x}_{2,1})$

Conditional Autarkies and Pigeon Hole Formulas

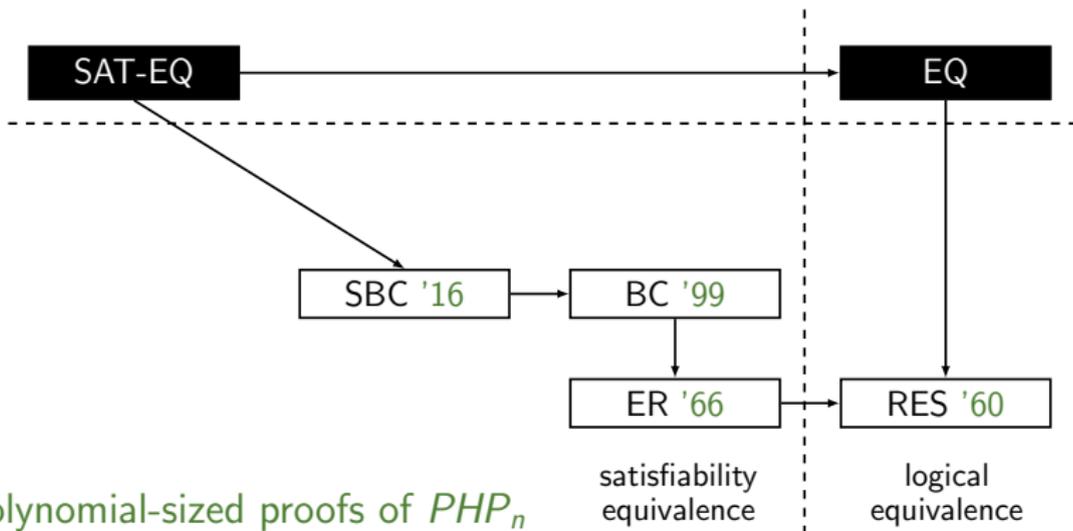
$$\begin{aligned} & (x_{1,1} \vee x_{2,1} \vee \dots \vee x_{n,1}) \quad \wedge \quad (x_{1,3} \vee x_{2,3} \vee x_{3,3} \vee \dots \vee x_{n,3}) \wedge \\ & (x_{1,2} \vee x_{2,2} \vee \dots \vee x_{n,2}) \wedge \dots \wedge (x_{1,n+1} \vee x_{2,n+1} \vee x_{3,n+1} \vee \dots \vee x_{n,n+1}) \wedge \\ & (\bar{x}_{1,1} \vee \bar{x}_{1,2}) \wedge (\bar{x}_{2,1} \vee \bar{x}_{2,2}) \wedge (\bar{x}_{3,1} \vee \bar{x}_{3,2}) \wedge \dots \wedge (\bar{x}_{n,1} \vee \bar{x}_{n,2}) \wedge \\ & (\bar{x}_{1,1} \vee \bar{x}_{1,3}) \wedge (\bar{x}_{2,1} \vee \bar{x}_{2,3}) \wedge (\bar{x}_{3,1} \vee \bar{x}_{3,3}) \wedge \dots \wedge (\bar{x}_{n,1} \vee \bar{x}_{n,3}) \wedge \\ & (\bar{x}_{1,2} \vee \bar{x}_{1,3}) \wedge (\bar{x}_{2,2} \vee \bar{x}_{2,3}) \wedge (\bar{x}_{3,2} \vee \bar{x}_{3,3}) \wedge \dots \wedge (\bar{x}_{n,2} \vee \bar{x}_{n,3}) \wedge \\ & \quad \dots \quad \wedge \quad \dots \\ & (\bar{x}_{1,n} \vee \bar{x}_{1,n+1}) \wedge (\bar{x}_{2,n} \vee \bar{x}_{2,n+1}) \wedge (\bar{x}_{3,n} \vee \bar{x}_{3,n+1}) \wedge \dots \wedge (\bar{x}_{n,n} \vee \bar{x}_{n,n+1}) \end{aligned}$$

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- Resolution can reduce the constraint $\alpha_{\text{con}} \rightarrow \alpha_{\text{aut}}$ to $(\bar{x}_{1,2} \vee \bar{x}_{2,1})$
- Allows constructing poly-sized proofs in SBC w/o new variables

Proof Systems based on Conditional Autarkies



Proof Systems based on Conditional Autarkies



Polynomial-sized proofs of PHP_n
in SBC without new variables

However, many clauses are long (making proof search hard)

Shorter Clauses

Reverse Unit Propagation [Goldberg and Novikov 2003]

- **Unit propagation** (UP) satisfies unit clauses by assigning their literal to true (until fixpoint or a conflict).
- Let F be a formula, C a clause, and α the smallest assignment that falsifies C . C is **implied by F via UP** (denoted by $F \vdash_{\perp} C$) if UP on $F|\alpha$ results in a conflict.
- $F \vdash_{\perp} C$ is also known as **Reverse Unit Propagation** (RUP).
- **Learned clauses** in CDCL solvers are RUP clauses.
- RUP typically summarizes **dozens of resolution steps**.

DRAT: An Interference-Based Proof System [SAT 2014]

- Popular **example** of an interference-based proof system: **DRAT**
- DRAT allows the addition of **RATs** (defined below) to a formula.
 - It can be **efficiently checked** if a clause is a RAT.
 - RATs are **not necessarily implied** by the formula.
 - But RATs are redundant: their **addition preserves satisfiability**.
- DRAT also allows clause **deletion**
 - Initially introduced to check proofs **more efficiently**
 - Clause deletion may **introduce clause addition** options (interference)

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A clause $(C \vee x)$ is a **resolution asymmetric tautology** (RAT) on x w.r.t. a CNF formula F if **for every clause** $(D \vee \bar{x}) \in F$, the resolvent $C \vee D$ is **implied by F via unit-propagation**, i.e., $F \vdash_1 C \vee D$.

Redundancy as an Implication [CADE 2017]

A formula G is **at least as satisfiable** as a formula F if $F \models G$.

Theorem ([Heule, Kiesl, Biere 2017])

*Let F be a formula, C a clause, and α the smallest assignment that falsifies C . Then, C is **redundant** w.r.t. F iff there exists an assignment ω such that 1) ω satisfies C ; and 2) $F|_{\alpha} \models F|_{\omega}$.*

This is the **strongest notion** of redundancy. However, it cannot be checked in polynomial time (assuming $P \neq NP$), unless **bounded**.

Propagation Redundancy [CADE 2017]

- Implied by F via UP is used in SAT solvers to determine redundancy of learned clauses and therefore \upharpoonright_1 is a **natural restriction** of \models .
- We bound $F|_\alpha \models F|_\omega$ by $F|_\alpha \upharpoonright_1 F|_\omega$.

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Definition (Propagation Redundant Clause)

Let F be a formula, C a clause, and α the smallest assignment that falsifies C . Then, C is **propagation redundant** (PR) w.r.t. F if there exists an assignment ω satisfying C with $F|_\alpha \vdash_1 F|_\omega$.

PR and Pigeon Hole Formulas [CADE 2017]

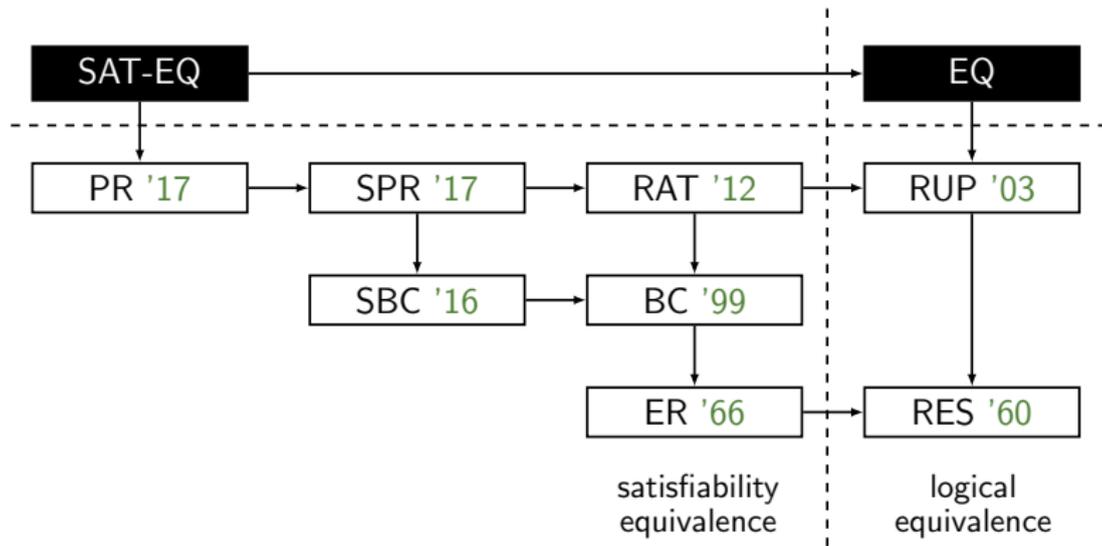
Can $n+1$ pigeons be placed in n holes (at-most-one pigeon per hole)?

$$PHP_n := \bigwedge_{1 \leq p \leq n+1} (x_{1,p} \vee \dots \vee x_{n,p}) \wedge \bigwedge_{1 \leq h \leq n, 1 \leq p < q \leq n+1} (\bar{x}_{h,p} \vee \bar{x}_{h,q})$$

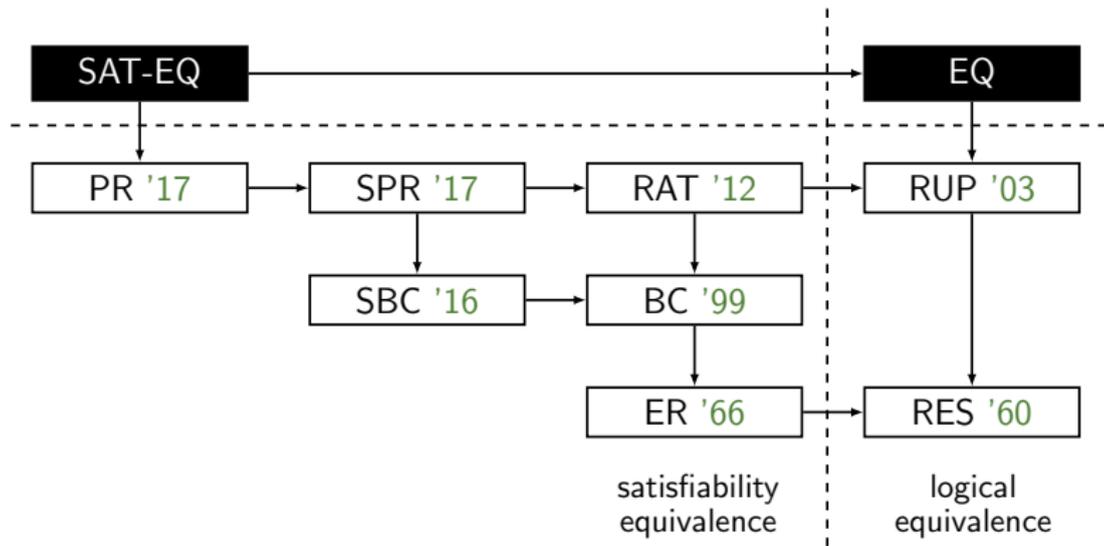
Any $(\bar{x}_{h,p} \vee \bar{x}_{k,q})$ with $h \neq k$ and $p \neq q$ is a **PR clause** w.r.t. PHP_n

- with witness $\omega = \bar{x}_{h,p} \wedge x_{k,p} \wedge x_{h,q} \wedge \bar{x}_{k,q}$
- learning n binary clauses $(\bar{x}_{1,1} \vee \bar{x}_{n,q})$ with $q \in \{2, \dots, n+1\}$ allows learning the unit clause $(\bar{x}_{1,1})$

New Proof Systems for Propositional Logic



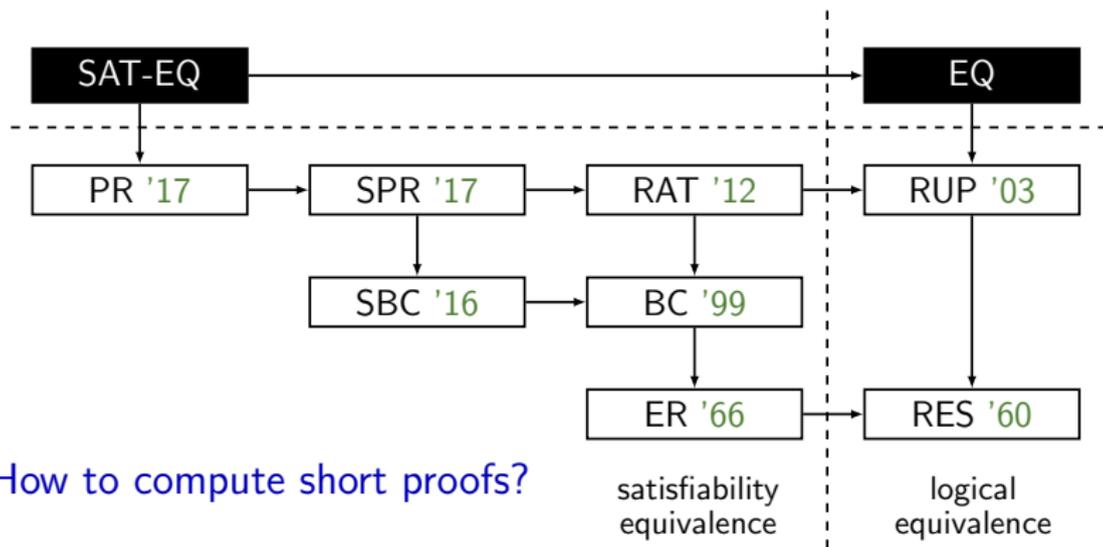
New Proof Systems for Propositional Logic



RAT simulates PR [Heule and Biere 2018]

ER simulates RAT [Kiesl, Rebola-Pardo, Heule 2018]

New Proof Systems for Propositional Logic



RAT simulates PR [Heule and Biere 2018]

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Satisfaction-Driven Clause Learning

Finding PR Clauses: The Positive Reduct [HVC 2017]

Determining whether a clause C is SBC or PR w.r.t. a formula F is an NP-complete problem.

How to find SBC and PR clauses? Encode it in SAT!

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How to find SBC and PR clauses? Encode it in **SAT**!

Given a formula F and a clause C . Let α denote the smallest assignment that falsifies C . The **positive reduct** of F and α is a formula which is satisfiable if and only if C is SBC w.r.t. F .

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Positive reducts are typically very easy to solve!

Key Idea: While solving a formula F , check whether the positive reduct of F and the current assignment α is **satisfiable**. In that case, **prune** the branch α .

The Positive Reduct: An Example [HVC 2017]

Given a formula F and a clause C . Let α denote the smallest assignment that falsifies C . The **positive reduct** of F and α , denoted by $p(F, \alpha)$, is the formula that contains C and all *assigned*(D, α) with $D \in F$ and D is satisfied by α .

Example

Consider the formula $F := (x \vee y) \wedge (x \vee \bar{y}) \wedge (\bar{y} \vee \bar{z})$.

Let $C_1 = (\bar{x})$, so $\alpha_1 = x$.

The positive reduct $p(F, \alpha_1) = (\bar{x}) \wedge (x) \wedge (x)$ is **unsatisfiable**.

Let $C_2 = (\bar{x} \vee \bar{y})$, so $\alpha_2 = x y$. The positive reduct $p(F, \alpha_2) = (\bar{x} \vee \bar{y}) \wedge (x \vee y) \wedge (x \vee \bar{y})$ is **satisfiable**.

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Theorem

Given a formula F and an assignment α . Every satisfying assignment ω of $p(F, \alpha)$ is a conditional autarky of F .

Pseudo-Code of CDCL (formula F)

```
1    $\alpha := \emptyset$ 
2   forever do
3      $\alpha := \text{Simplify}(F, \alpha)$ 
4     if  $F|_{\alpha}$  contains a falsified clause then
5        $C := \text{AnalyzeConflict}()$ 
6       if  $C$  is the empty clause then return unsatisfiable
7        $F := F \cup \{C\}$ 
8        $\alpha := \text{BackJump}(C, \alpha)$ 
13  else
14     $I := \text{Decide}()$ 
15    if  $I$  is undefined then return satisfiable
16     $\alpha := \alpha \cup \{I\}$ 
```

Pseudo-Code of SDCL (formula F) [HVC 2017]

```
1    $\alpha := \emptyset$ 
2   forever do
3      $\alpha := \text{Simplify}(F, \alpha)$ 
4     if  $F|\alpha$  contains a falsified clause then
5        $C := \text{AnalyzeConflict}()$ 
6       if  $C$  is the empty clause then return unsatisfiable
7        $F := F \cup \{C\}$ 
8        $\alpha := \text{BackJump}(C, \alpha)$ 
9     else if  $p(F, \alpha)$  is satisfiable then
10       $C := \text{AnalyzeWitness}()$ 
11       $F := F \cup \{C\}$ 
12       $\alpha := \text{BackJump}(C, \alpha)$ 
13    else
14       $I := \text{Decide}()$ 
15      if  $I$  is undefined then return satisfiable
16       $\alpha := \alpha \cup \{I\}$ 
```

Benchmark Suite: Pigeon Hole Formulas [HVC 2017]

Can $n+1$ pigeons be placed in n holes (at-most-one pigeon per hole)?

$$PHP_n := \bigwedge_{1 \leq p \leq n+1} (x_{1,p} \vee \dots \vee x_{n,p}) \wedge \bigwedge_{1 \leq h \leq n} \bigwedge_{1 \leq p < q \leq n+1} (\bar{x}_{h,p} \vee \bar{x}_{h,q})$$

The binary clauses encode the constraint $\leq_1 (x_{h,1}; \dots; x_{h,n+1})$.

There exists **more compact encodings**, such as the sequential counter and minimal encoding, for at-most-one constraints.

We include these encodings to evaluate the robustness of the solver.

Tool Comparison

We used three tools in our evaluation:

- **EBDDRES**: A tool based on binary decision diagrams that can convert a refutation into an extended resolution proof.
- **GLUCOSER**: A SAT solver with **extended learning**, i.e., a technique that introduces new variables and could potentially solve pigeon hole formulas in polynomial time.
- **LINGELING (PR)**: Our SDCL solver.

Results on Small Pigeon Hole Formulas [HVC 2017]

<i>formula</i>	input		EBDDRES		GLUCOSER		LINGELING (PR)	
	#var	#cls	time	#node	time	#lemma	time	#lemma
<i>PHP</i> ₁₀ -std	110	561	1.00	3M	22.71	329,470	0.07	329
<i>PHP</i> ₁₁ -std	132	738	3.47	9M	146.61	1,514,845	0.11	439
<i>PHP</i> ₁₂ -std	156	949	10.64	27M	307.29	2,660,358	0.16	571
<i>PHP</i> ₁₃ -std	182	1,197	30.81	76M	982.84	6,969,736	0.22	727
<i>PHP</i> ₁₀ -seq	220	311	OF	—	1.62	25,712	0.07	327
<i>PHP</i> ₁₁ -seq	264	375	OF	—	6.94	77,747	0.10	437
<i>PHP</i> ₁₂ -seq	312	445	OF	—	19.40	174,084	0.14	569
<i>PHP</i> ₁₃ -seq	364	521	OF	—	172.76	1,061,318	0.18	725
<i>PHP</i> ₁₀ -min	180	281	28.60	81M	0.64	15,777	0.06	329
<i>PHP</i> ₁₁ -min	220	342	143.92	399M	1.82	34,561	0.10	439
<i>PHP</i> ₁₂ -min	264	409	OF	—	9.87	121,321	0.13	571
<i>PHP</i> ₁₃ -min	312	482	OF	—	57.66	483,789	0.18	727

OF = 32-bit overflow

Results on Large Pigeon Hole Formulas [HVC 2017]

<i>formula</i>	input		EBDDRES		GLUCOSER		LINGELING (PR)	
	#var	#cls	time	#node	time	#lemma	time	#lemma
<i>PHP</i> ₂₀ -std	420	4,221	OF	—	TO	—	1.61	2,659
<i>PHP</i> ₃₀ -std	930	13,981	OF	—	TO	—	13.45	8,989
<i>PHP</i> ₄₀ -std	1,640	32,841	OF	—	TO	—	67.41	21,319
<i>PHP</i> ₅₀ -std	2,550	63,801	OF	—	TO	—	241.14	41,649
<i>PHP</i> ₂₀ -seq	840	1,221	OF	—	TO	—	1.05	2,657
<i>PHP</i> ₃₀ -seq	1,860	2,731	OF	—	TO	—	6.55	8,987
<i>PHP</i> ₄₀ -seq	3,280	4,841	OF	—	TO	—	27.10	21,317
<i>PHP</i> ₅₀ -seq	5,100	7,551	OF	—	TO	—	86.30	41,647
<i>PHP</i> ₂₀ -min	760	1,161	OF	—	TO	—	1.03	2,659
<i>PHP</i> ₃₀ -min	1,740	2,641	OF	—	TO	—	6.30	8,989
<i>PHP</i> ₄₀ -min	3,120	4,721	OF	—	TO	—	26.65	21,319
<i>PHP</i> ₅₀ -min	4,900	7,401	OF	—	TO	—	85.00	41,649

OF = 32-bit overflow

TO = timeout of 9000 seconds

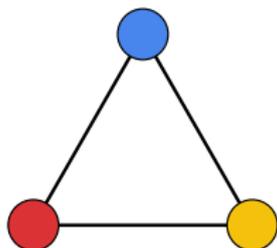
One More Thing...

Chromatic Number of the Plane

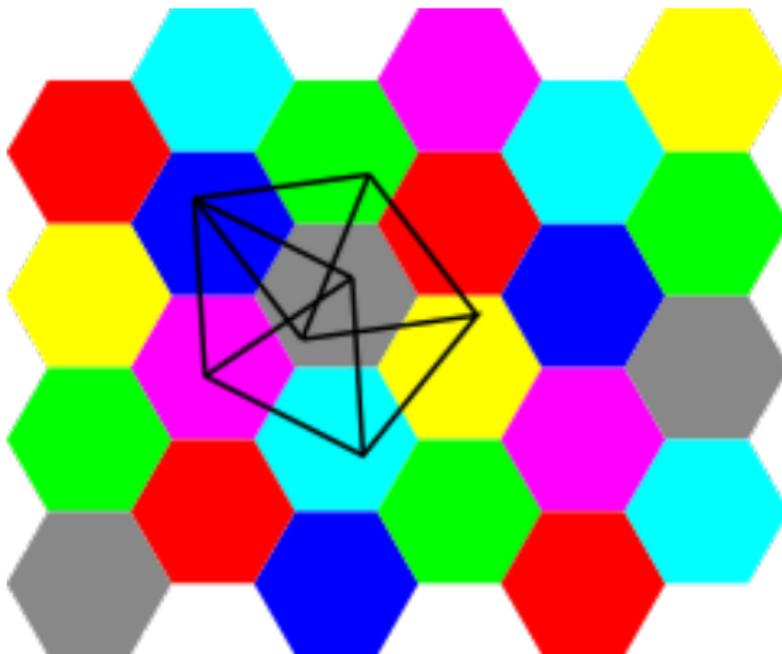
The Hadwiger-Nelson problem:

How many colors are required to color the plane such that each pair of points that are exactly 1 apart are colored differently?

The answer must be three or more because three points can be mutually 1 apart—and thus must be colored differently.



Bounds since the 1950s



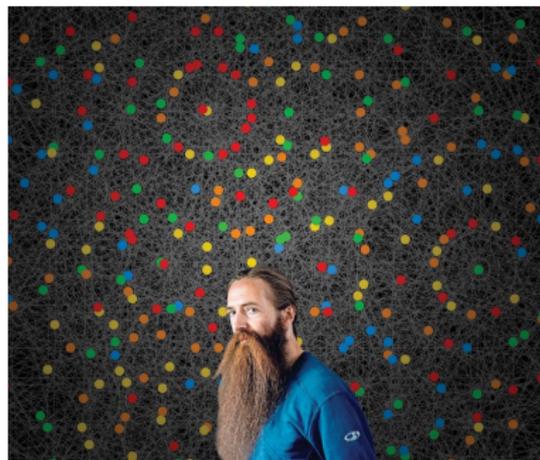
- The Moser Spindle graph shows the lower bound of 4
- A coloring of the plane showing the upper bound of 7

First Progress in Decades [De Grey 2018]

The first meaningful progress on this problem was by Aubrey de Grey, who found a unit-distance graph with **chromatic number 5**.

He published a graph with 1581 vertices on April 8, 2018.

Aubrey de Grey is known for his research to extend life.



The New Result Started a Media Hype



Decades-Old Graph Problem Yields to Amateur Mathematician

by KEVIN LAND



Одна из самых красивых и до сих пор не решенных задач математик формулируется следующим образом. Попытаемся раскрасить плоскость так, чтобы никакие две точки, находящиеся на расстоянии одного сантиметра друг от друга, не оказались покрашены в один цвет. Какое минимальное число цветов для этого потребуется? Несмотря на кажущуюся простоту, за почти 70 лет существования этой задачи точного ответа до сих пор нет, притом что над ней

業餘數學家為一道填色難題帶來突破！

2020年10月14日 10:00 AM



本圖來自 [圖網評述](#)，INSIDE 控疫情報網。

1950 年秋天，美國伊利諾大學厄巴納的數學家尼爾遜 (Edward Nelson) 對「四色問題」感到興趣。這個問題是：

nrc.nl [abonnement](#)

Man van 'we leven ooit 1.000 jaar' brengt oplossing oud pixelprobleem dichterbij

Wiskunde

De bekende Britse verouderingswetenschapper Aubrey de Grey heeft zich met succes op een meetkundevraagstuk uit de jaren vijftig van de vorige eeuw gestort. Zijn vondst werd zeer veel besproken op blogs van wiskundigen.

Alta van den Brundhof 24 april 2020

Een gedachtenexperiment: stel je een foto voor waarvan de pixels onszindig klein zijn. Dus kovert je ook inzoomen, de individuele pixels zijn nooit te herkennen, omdat ze oppervlakte nul hebben. Elke pixel - dat zijn er uiteraard onszindig veel - heeft een kleur. De vraag is: hoeveel verschillende kleuren zijn er minimaal nodig om ervoor te zorgen dat twee pixels die op een vaste afstand, bijvoorbeeld 1 centimeter, van elkaar vandaan liggen, nooit dezelfde kleur hebben?



Propositional Proofs for Graph Validation and Shrinking

Checking that a unit-distance graph has chromatic number 5:

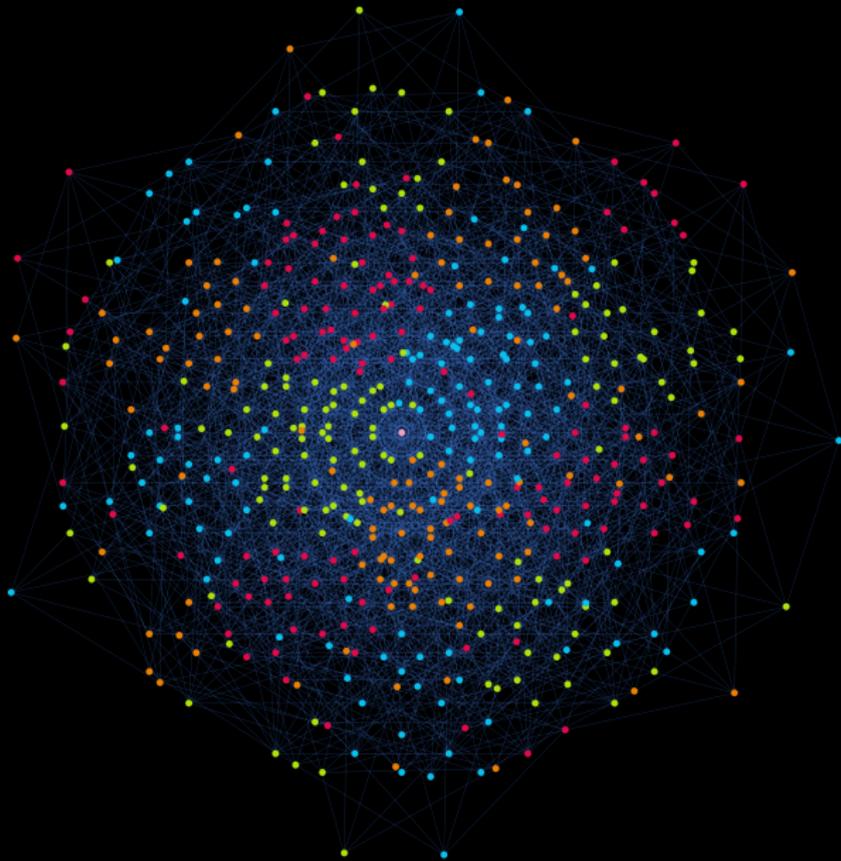
- Show that there exists a 5-coloring
- While there is no 4-coloring (formula is UNSAT)
- Unsatisfiable core represents a subgraph

SAT solvers find **short proofs** of unsatisfiability for these formulas

Proof minimization techniques allow further reduction

Combining the techniques allows finding **much smaller graphs**

Record by Proof Minimization: 553 Vertices [Heule 2018]



Challenges and Conclusions

Theoretical Challenges

Lower bounds for interference-based proof systems with new variables will be hard, but what about **without new variables**?

- Lower bound for BC w/o new variables? Pigeon-hole formulas?
- Lower bound for SBC w/o new variables? Tseitin formulas?
- Lower bound for PR w/o new variables?!

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Can the new proof systems without new variables **simulate** old ones, in particular **Frege systems** (or the other way around)?

What about **cutting planes**?

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Can the new proof systems without new variables **simulate** old ones, in particular **Frege systems** (or the other way around)?

What about **cutting planes**?

Can we design stronger proof systems that make it even **easier to compute** short proofs?

Practical Challenges

The current version of SDCL is just the beginning:

- Which heuristics allow learning short PR clauses?
- How to construct an AnalyzeWitness procedure?
- Can the positive reduct be improved?

Can **local search** be used to find short proofs of unsatisfiability?

Constructing positive reducts (or similar formulas) efficiently:

- Generating a positive reduct is **more costly** than solving them
- Can we design **data-structures** to cheaply compute them?

Conclusions

We introduced new redundancy notions for SAT.

Proof systems based on these redundancy notions are strong.

- They allow for **short proofs without new variables**; and
- They are more suitable for **mechanized** proof search.

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SDCL generalizes the well-known CDCL paradigm by allowing to prune branches that are potentially satisfiable:

- Such branches can be found using the **positive reduct**;
- Pruning can be expressed in the **PR proof system**;
- Runtime and proofs can be **exponentially** smaller.