PRuning Through Satisfaction

Marijn J.H. Heule, Benjamin Kiesl, Martina Seidl, and Armin Biere

UT Austin, Vienna University of Technology, and JKU Linz





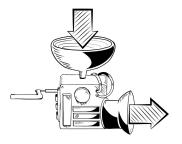


Haifa Verification Conference

November 15, 2017

SAT problem:

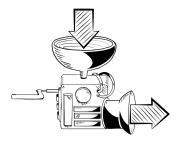
$$(x \lor y) \land (x \lor \bar{y}) \land (\bar{y} \lor \bar{z})$$



SAT problem:

Given a propositional formula, is it satisfiable?

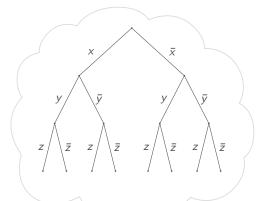
Input Formula in CNF $\downarrow \\ (x \vee y) \wedge (x \vee \bar{y}) \wedge (\bar{y} \vee \bar{z})$



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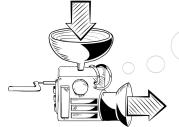
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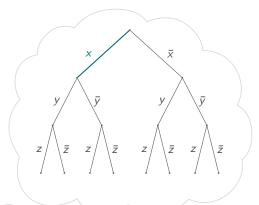




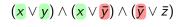
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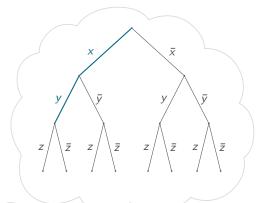




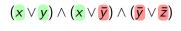
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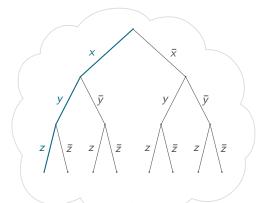




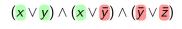
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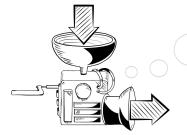


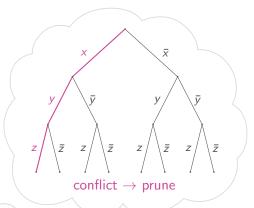




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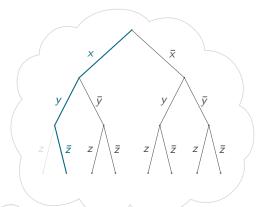




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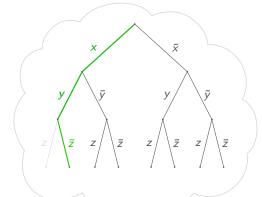


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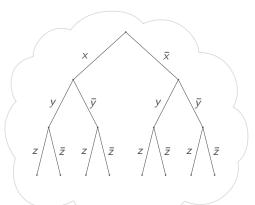




Satisfiable

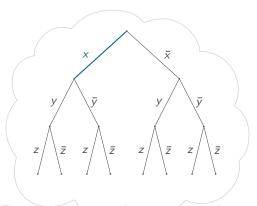
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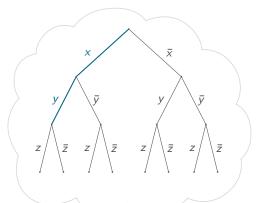
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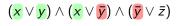


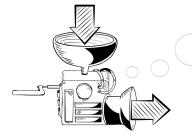


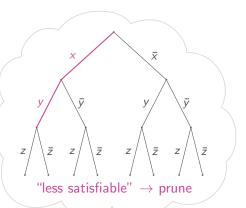




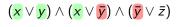




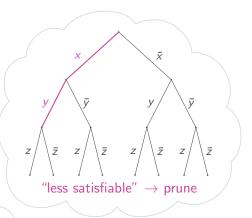




Can we prune earlier? Even satisfiable branches?

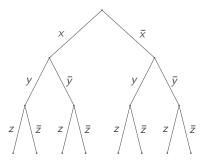




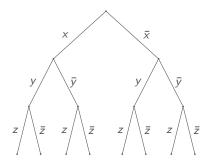


How to prune? Add redundant clauses!

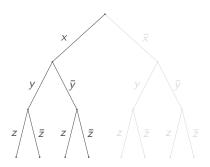
■ A clause prunes all branches that falsify the clause.



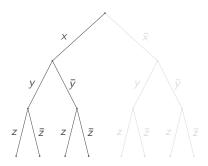
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- **Example**: The clause (x) prunes all branches where x is false.



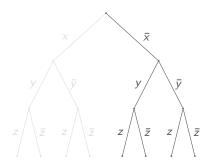
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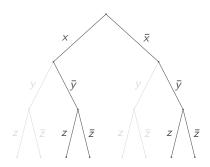
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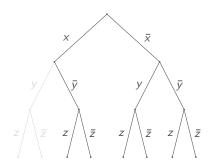
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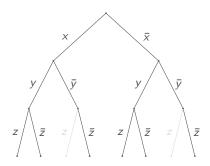
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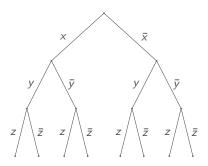
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Introduction

The Positive Reduct

Conditional Autarkies

The Algorithm

Evaluation

Conclusions and Future Work

The Positive Reduct

Traditional Proofs vs. Interference-Based Proofs

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

$$\frac{C \lor x \qquad \overline{x} \lor D}{C \lor D} \text{ (res)} \qquad \frac{A \qquad A \to B}{B} \text{ (mp)}$$

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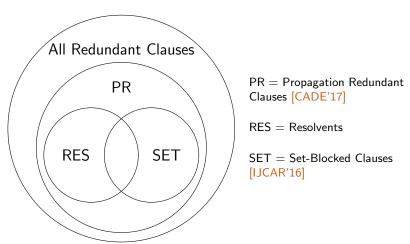
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- ➡ Inference rules reason about the presence of facts.
 - If certain premises are present, infer the conclusion.
 - Different approach: Allow not only implied conclusions.
 - Require only that the addition of facts preserves satisfiability.
 - Reason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

Redundant Clauses

A clause C is redundant w.r.t. a formula F if and only if F and $F \wedge C$ are either both satisfiable or both unsatisfiable.



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

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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

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 $assigned(D,\alpha)$ with $D \in F$ and D is satisfied by α .

Example

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

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

The positive reduct $p(F, \alpha_1) = (\bar{x}) \land (x) \land (x)$ 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.

Conditional Autarkies

Autarkies

A non-empty assignment α is an autarky for formula F if every clause $C \in F$ that is touched by α is also satisfied by α .

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 ([Monien and Speckenmeyer 1985])

Let α be an autarky for formula F. Then, F and $F|_{\alpha}$ are satisfiability equivalent.

Conditional Autarkies

An assignment $\alpha = \alpha_{\rm con} \cup \alpha_{\rm aut}$ is a conditional autarky for formula F if $\alpha_{\rm aut}$ is an autarky for $F \mid_{\alpha_{\rm con}}$.

Example

Consider the formula $F:=(x\vee y)\wedge(x\vee \bar{y})\wedge(\bar{y}\vee \bar{z})$. Let $\alpha_{\rm con}=x$ and $\alpha_{\rm aut}=\bar{y}$, then $\alpha=\alpha_{\rm con}\cup\alpha_{\rm aut}=x\,\bar{y}$ is a conditional autarky for F:

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

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

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

Learning PR clauses

Theorem

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

Recall: Given a formula F and a clause C. Let α denote the smallest assignment that falsifies C. C is SET w.r.t. F if and only if $p(F, \alpha)$ is satisfiable.

Let assignment ω satisfy $p(F, \alpha)$. Removing all but one of the literals in C that are satisfied by ω results in a PR clause w.r.t. F.

The Algorithm

Pseudo-Code of CDCL (formula F)

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

Pseudo-Code of SDCL (formula F)

```
\alpha := \emptyset
       forever do
2
          \alpha := Simplify (F, \alpha)
 3
          if F|_{\alpha} contains a falsified clause then
              C := AnalyzeConflict ()
             if C is the empty clause then return unsatisfiable
             F := F \cup \{C\}
             \alpha := \mathsf{BackJump} (C, \alpha)
          else if p(F, \alpha) is satisfiable then
              C := AnalyzeWitness ()
10
             F := F \cup \{C\}
11
             \alpha := \mathsf{BackJump} (C, \alpha)
12
          else
13
             I := Decide()
14
             if / is undefined then return satisfiable
15
             \alpha := \alpha \cup \{I\}
16
```

Evaluation

Benchmark Suite: Pigeon Hole Formulas

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

$$PHP_n := \bigwedge_{1 \le p \le n+1} (x_{1,p} \lor \cdots \lor x_{n,p}) \land \bigwedge_{1 \le h \le n, \ 1 \le p < q \le n+1} (\overline{x}_{h,p} \lor \overline{x}_{h,q})$$

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

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

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

OF = 32-bit overflow

Results on Large Pigeon Hole Formulas

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

OF = 32-bit overflow

TO = timeout of 9000 seconds

Conclusions and Future Work

Conclusions

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.

Our SDCL solver finds short proofs of pigeon hole formulas:

- Integrated in the state-of-the-art solver Lingeling;
- Linear sized proofs in $\mathcal{O}(n^3)$ can be found fully automatically;
- The implementation is efficient, robust, and open source.

Future Work

- SDCL likely requires different heuristics compared to CDCL
- Can more branches be pruned using stronger SAT calls?
- How to minimize clauses from pruning through satisfaction?
- Can SLS techniques be used to find conditional autarkies?

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