Exploitation des symétries dynamiques pour la résolution des problèmes SAT

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Motivation

SAT is widely used in different domains:

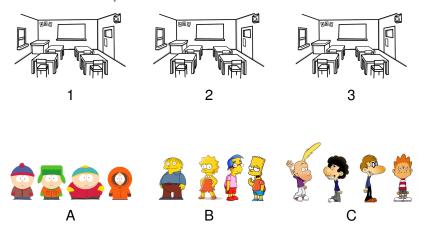
- Artificial intelligence (planning, games, ...)
- Bioinformatics (haplotype inference, ...)
- Security (cryptanalysis, inversion attack on hash function)
- Computationally hard problems (graph coloring, ...)
- Formal Methods (hardware model checking, ...)

Outline

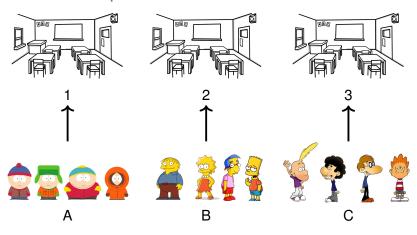
- SAT overview
 SAT basics
 SAT and symmetries
- 2 Existing approaches

3 Contribution and results

SAT an example



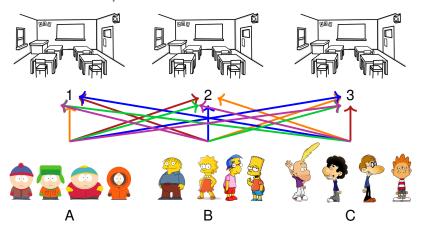
SAT an example



Is it possible to attribute each group to a classroom?

YES!

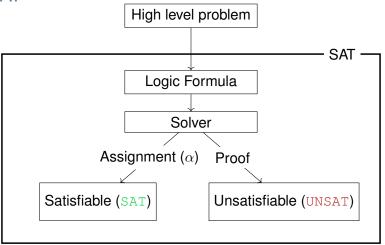
SAT an example



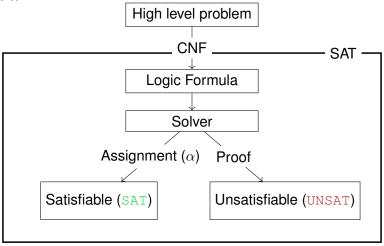
Is it possible to attribute each group to a classroom?

YES! Many solutions

SAT





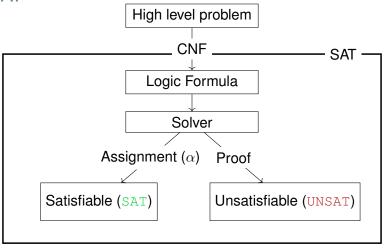


CNF Representation:

$$\underbrace{\left(X_1 \lor X_2 \lor \neg X_3\right)}_{\text{Clause with literals } X_1, X_2, \neg X_3}$$

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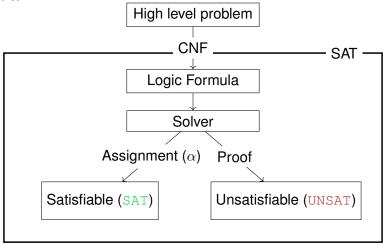
SAT



CNF Representation:

Formula (CNF)
$$\underbrace{\left(x_1 \lor x_2 \lor \neg x_3\right)}_{Clause} \land \left(\neg x_1 \lor \neg x_2\right) \land \left(x_2 \lor \neg x_4\right)$$





CNF Representation:

$$(x_1 \lor x_2 \lor \neg x_3) \to \{x_1, x_2, \neg x_3\}$$

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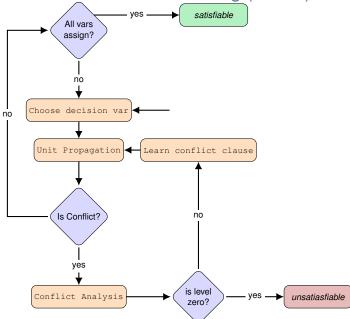
SAT Solving

Solving SAT formula is known to be **NP-complete** [Coo71]

Enumerative Algorithm:

- Davis, Putnam, Logemann, and Loveland (DPLL) [DLL62]
 - Boolean Constraint Propagation (BCP)
- Conflict Driven Clause Learning (CDCL) [MSS99]
 - derived from DPLL
 - clause learning

Conflict Driven Clause Learning (CDCL)

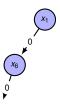




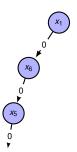
$$\omega_{1} = \{x_{1}, x_{2}, x_{3}\}
\omega_{2} = \{x_{4}, x_{5}, x_{6}\}
\omega_{3} = \{\neg x_{1}, \neg x_{5}\}
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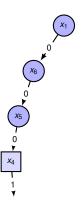
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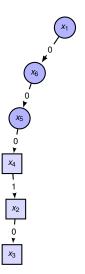
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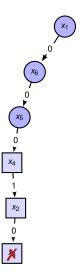
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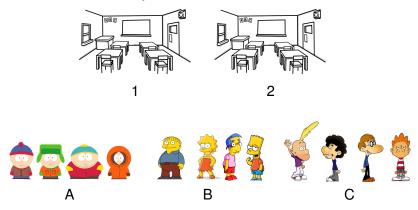


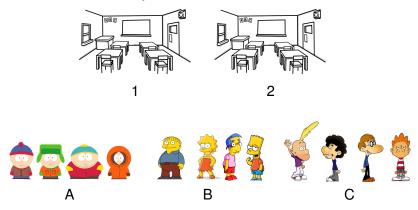
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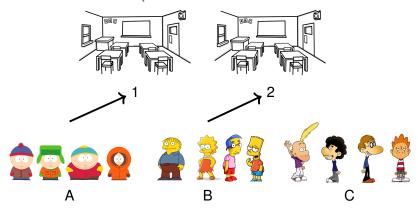


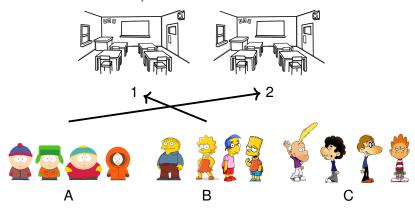
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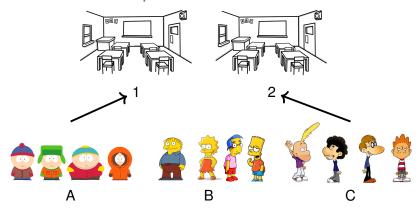
$$\omega_7 = \{x_1, \neg x_4\}$$

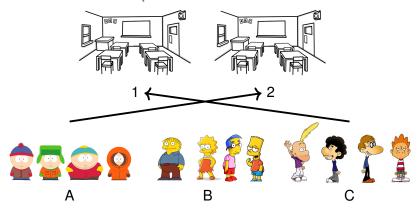


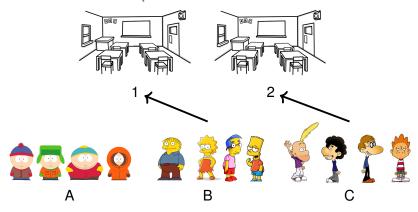


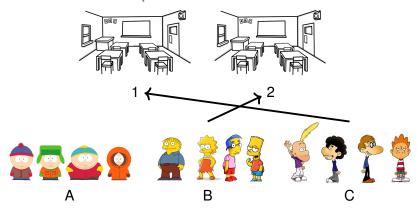






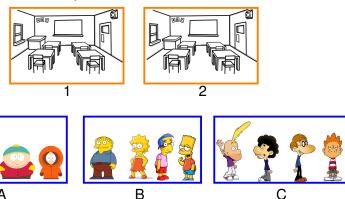






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No!



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No!

Presence of symmetries hinders the performance of the solver

Outline

SAT overview
 SAT basics

SAT basics
SAT and symmetries

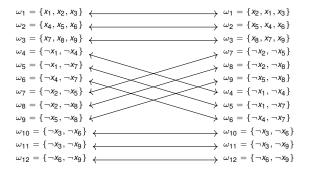
2 Existing approaches

Static symmetry breaking Dynamic symmetry breaking

3 Contribution and results

Symmetry

$$g = \begin{pmatrix} x_1 & x_2 & x_3 & x_4 & x_5 & x_6 & x_7 & x_8 & x_9 \\ x_2 & x_1 & x_3 & x_5 & x_4 & x_6 & x_8 & x_7 & x_9 \end{pmatrix} \rightarrow (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$$



A symmetry (permuation) g is a bijective function (on variables) that leaves φ invariant.

Computing symmetries of a SAT problem $(x_1 \lor x_2 \lor x_3) \land (x_4 \lor x_5 \lor x_6) \land (x_7 \lor x_8 \lor x_9)$

CNF formula

 $\begin{array}{c} (\overline{x}_1 \vee x_2 \vee x_3^{\vee}) \wedge (\overline{x}_4 \vee \overline{x}_5 \vee x_6) \wedge (x_7 \vee x_8 \vee x_6) \\ \wedge (\neg x_1 \vee \neg x_4) \wedge (\neg x_1 \vee \neg x_7) \wedge (\neg x_4 \vee \neg x_7) \\ \wedge (\neg x_2 \vee \neg x_5) \wedge (\neg x_2 \vee \neg x_8) \wedge (\neg x_5 \vee \neg x_8) \\ \wedge (\neg x_3 \vee \neg x_6) \wedge (\neg x_3 \vee \neg x_9) \wedge (\neg x_6 \vee \neg x_9) \end{array}$

¹http://www.tcs.hut.fi/Software/bliss/

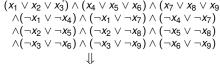
²http://vlsicad.eecs.umich.edu/BK/SAUCY/

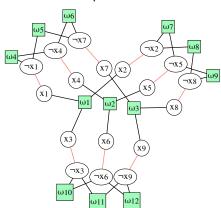
Computing symmetries of a SAT problem $(x_1 \lor x_2 \lor x_3) \land (x_4 \lor x_5 \lor x_6) \land (x_7 \lor x_8 \lor x_9)$

CNF formula



colored graph





Computing symmetries of a SAT problem $(x_1 \lor x_2 \lor x_3) \land (x_4 \lor x_5 \lor x_6) \land (x_7 \lor x_8 \lor x_9)$

CNF formula $\wedge(\neg x_1 \vee \neg x_4) \wedge (\neg x_1 \vee \neg x_7) \wedge (\neg x_4 \vee \neg x_7)$ $\wedge (\neg x_2 \vee \neg x_5) \wedge (\neg x_2 \vee \neg x_8) \wedge (\neg x_5 \vee \neg x_8)$ $\wedge(\neg x_3 \vee \neg x_6) \wedge (\neg x_3 \vee \neg x_9) \wedge (\neg x_6 \vee \neg x_9)$ colored graph (bliss 1 or saucy 2) graph automorphism

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Computing symmetries of a SAT problem

CNF formula

 \Downarrow

colored graph

graph automorphism ↓

set of symmetries

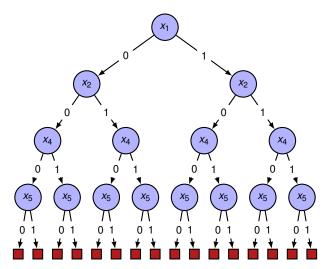
 $q_4 = (x_1 \ x_4)(x_2 \ x_5)(x_3 \ x_6)$

 $⁽x_1 \lor x_2 \lor x_3) \land (x_4 \lor x_5 \lor x_6) \land (x_7 \lor x_8 \lor x_9)$ $\wedge(\neg x_1 \vee \neg x_4) \wedge (\neg x_1 \vee \neg x_7) \wedge (\neg x_4 \vee \neg x_7)$ $\wedge(\neg x_2 \vee \neg x_5) \wedge (\neg x_2 \vee \neg x_8) \wedge (\neg x_5 \vee \neg x_8)$ $\wedge(\neg x_3 \vee \neg x_6) \wedge (\neg x_3 \vee \neg x_9) \wedge (\neg x_6 \vee \neg x_9)$ (bliss 1 or saucy 2) $g_1 = (x_2 \ x_3)(x_5 \ x_6)(x_8 \ x_9)$ $g_2 = (x_4 \ x_7)(x_5 \ x_8)(x_6 \ x_9)$ $g_3 = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$

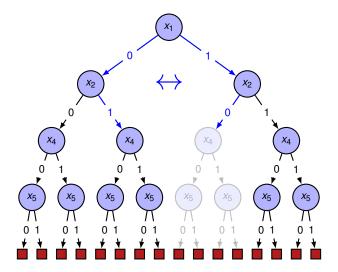
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Using symmetries to prune search space



Using symmetries to prune search space



Adds additional constraints to prune search space.

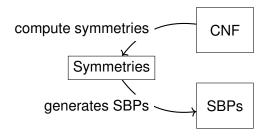
Generates symmetry breaking predicates (SBP)

- Define lexicographic order
 - Define total order on variables
 - Define minimal value
- Forbid non minimal assignment with addition of SBP

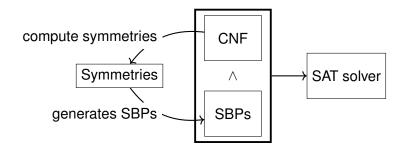
Example:

$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8;$$
 true < false $(x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$

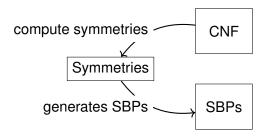
$x_1 \leq x_2$	$x_1 \vee \neg x_2$
$x_1=x_2\to x_4\le x_5$	$x_1 \vee x_2 \vee x_4 \vee \neg x_5$
	$\neg x_1 \lor \neg x_2 \lor x_4 \lor \neg x_5$
$x_1 = x_2 \wedge x_4 = x_5 \rightarrow x_8 \leq x_3$	$X_1 \vee X_2 \vee X_4 \vee X_5 \vee X_7 \vee \neg X_8$
	$\neg x_1 \vee \neg x_2 \vee x_4 \vee x_5 \vee x_7 \vee \neg x_8$
	• • •



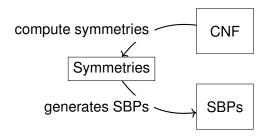
- Works well on many symmetric instances
- The solver can "explode" instead of being helped



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Dynamic Symmetry Breaking

TODO Accelerate tree traversal modify solver Different tools SP, SLS, SEL, ...

Symmetry Propagation

TODO Present SP

Example

TODO Build an example

Outline

SAT overview

SAT basics SAT and symmetries

2 Existing approaches

Static symmetry breaking Dynamic symmetry breaking

3 Contribution and results

CDCL [Sym]
Combination of different approaches

CDCL[Sym] idea

TODO

Tackling the explosion problem in the static symmetry breaking approaches.

Compute and inject ESBP opportunistically during the solving Symmetry Controller in CDCL

Symmetry status

TODO Reducer, Inactive, active

Example

Experimental results

ESBP + SP

TODO Symmetry propagation on top of ESBP Compose both approaches Is it possible?

Notion of local symmetries

Computation of local symmetries

Experimental results

Conclusion and Perspective

TODO

Conclusion:

Perspectives:

Thanks!



Stephen A Cook.

The complexity of theorem-proving procedures.

In Proceedings of the third annual ACM symposium on Theory of computing, pages 151–158. ACM, 1971.



Martin Davis, George Logemann, and Donald Loveland. A machine program for theorem-proving. *Commun. ACM*, 5(7):394–397, July 1962.



Joao P Marques-Silva and Karem A Sakallah.

Grasp: A search algorithm for propositional satisfiability. *IEEE Transactions on Computers*, 48(5):506–521, 1999.