Exploitation of dynamic symmetries for solving SAT problems

Doctorat de Sorbonne Université

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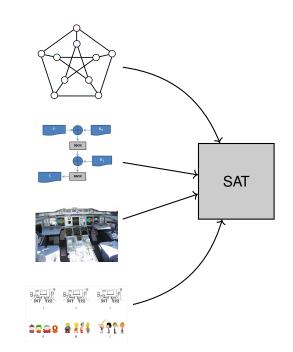
Motivation

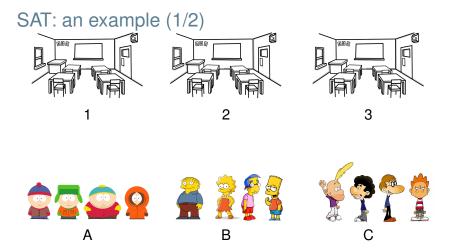
Graph coloring

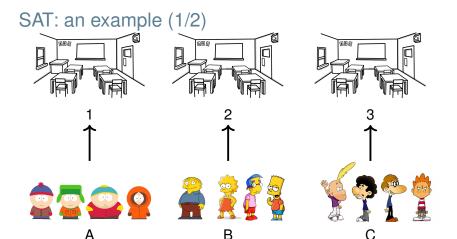
Cryptanalysis

Hardware model checking

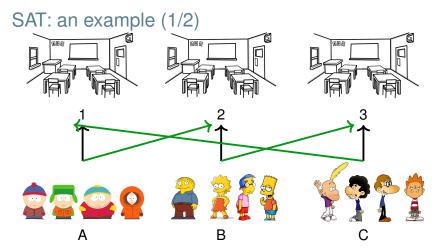
Planning





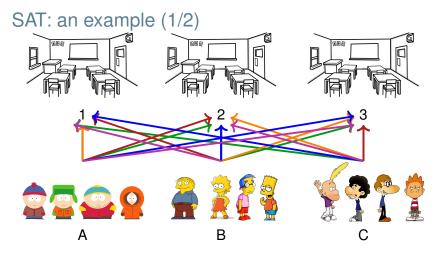


YES! SAT
$$\alpha = (A, 1), (B, 2), (C, 3)$$



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$$\alpha = (A, 1), (B, 2), (C, 3)$$

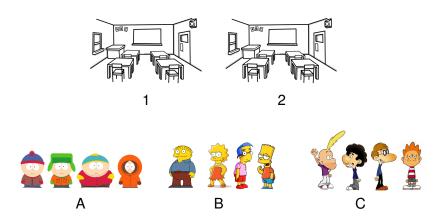
Many solutions $\alpha' = (A, 2), (B, 3), (C, 1)$



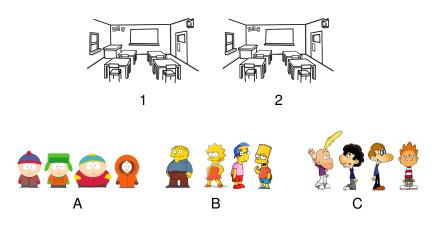
YES! SAT
$$\alpha = (A, 1), (B, 2), (C, 3)$$

Many solutions $\alpha' = (A, 2), (B, 3), (C, 1)$
 $\alpha'' = \cdots$

SAT: an example (2/2)



SAT: an example (2/2)



Is it possible to attribute each group to a unique classroom?

No! UNSAT

$$\overbrace{(A,1)(A,2)(A,3)}^{x_1}(A,1)(A,2)(A,3)$$
(B,1)(B,2)(B,3)
(C,1)(C,2)(C,3)

$$\begin{array}{c} (x_1 \lor x_2 \lor x_3) \land \\ (x_4 \lor x_5 \lor x_6) \land \\ (x_7 \lor x_8 \lor x_9) \land \end{array}$$



$$(A, 1) (A, 2) (A, 3)$$

$$(B, 1)(B, 2)(B, 3)$$

$$(C, 1)(C, 2)(C, 3)$$

$$\neg (A, 1) \neg (B, 1)$$

$$\neg (A, 1) \neg (C, 1)$$

$$\neg (B, 1) \neg (C, 1)$$

$$\begin{array}{c} (x_1 \lor x_2 \lor x_3) \land \\ (x_4 \lor x_5 \lor x_6) \land \\ (x_7 \lor x_8 \lor x_9) \land \\ \\ (\neg x_1 \lor \neg x_4) \land \\ (\neg x_1 \lor \neg x_7) \land \\ (\neg x_4 \lor \neg x_7) \land \end{array}$$

$$(A, 1) (A, 2) (A, 3)
(B, 1) (B, 2) (B, 3)
(C, 1) (C, 2) (C, 3)$$

$$-(A, 1) - (B, 1)
-(A, 1) - (C, 1)
-(B, 1) - (C, 1)$$

$$-(B, 1) - (C, 1)$$

$$-(A, 2) - (B, 2)
-(A, 2) - (C, 2)
-(B, 2) - (C, 2)$$

$$-(A, 3) - (B, 3)
-(A, 3) - (C, 3)
-(B, 3) - (C, 3)$$

$$-(B, 3) - (C, 3)$$

$$-(B, 3) - (C, 3)$$

$$(A, 1) (A, 2) (A, 3)$$

$$(B, 1)(B, 2)(B, 3)$$

$$(C, 1)(C, 2)(C, 3)$$

$$-(A, 1) - (B, 1)$$

$$-(A, 1) - (C, 1)$$

$$-(B, 1) - (C, 1)$$

$$-(B, 1) - (C, 1)$$

$$-(A, 2) - (C, 2)$$

$$-(A, 2) - (C, 2)$$

$$-(B, 2) - (C, 2)$$

$$-(A, 3) - (B, 3)$$

 $\neg (A,3) \neg (C,3)$

 $\neg (B,3) \neg (C,3)$

$$(x_2 \lor x_3) \land$$
 Clause

$$\begin{array}{c}
(x_1 \lor x_2 \lor x_3) \land \\
(x_4 \lor x_5 \lor x_6) \land \\
(x_7 \lor x_8 \lor x_9) \land
\end{array}$$

$$\begin{array}{c} (\neg x_1 \lor \neg x_4) \land \\ (\neg x_1 \lor \neg x_7) \land \\ (\neg x_4 \lor \neg x_7) \land \end{array}$$

$$\begin{array}{l} (\neg x_3 \lor \neg x_6) \land \\ (\neg x_3 \lor \neg x_9) \land \\ (\neg x_6 \lor \neg x_9) \end{array}$$

$$(A, 1) (A, 2) (A, 3)$$

$$(B, 1)(B, 2)(B, 3)$$

$$(C, 1)(C, 2)(C, 3)$$

$$\neg (A, 1) \neg (B, 1)$$

$$\neg (A, 1) \neg (C, 1)$$

$$\neg (B, 1) \neg (C, 1)$$

$$\neg (B, 1) \neg (C, 2)$$

$$\neg (A, 2) \neg (B, 2)$$

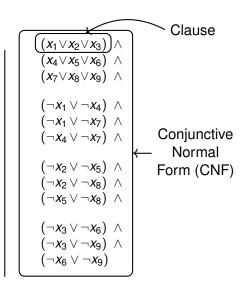
$$\neg (A, 2) \neg (C, 2)$$

$$\neg (A, 3) \neg (C, 2)$$

$$\neg (A, 3) \neg (C, 3)$$

$$\neg (A, 3) \neg (C, 3)$$

$$\neg (B, 3) \neg (C, 3)$$



Clause
$$(A, 1) (A, 2) (A, 3) \\
(B, 1)(B, 2)(B, 3) \\
(C, 1)(C, 2)(C, 3)$$

$$\neg (A, 1) \neg (B, 1) \\
\neg (A, 1) \neg (C, 1) \\
\neg (B, 1) \neg (C, 1)$$

$$\neg (B, 1) \neg (C, 1)$$

$$\neg (A, 2) \neg (B, 2) \\
\neg (A, 2) \neg (C, 2) \\
\neg (B, 2) \neg (C, 2)$$

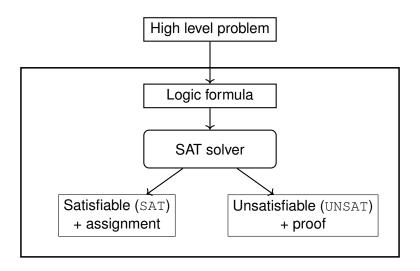
$$\neg (A, 3) \neg (B, 3) \\
\neg (A, 3) \neg (C, 3) \\
\neg (B, 3) \neg (C, 3)$$

$$\neg (B, 3) \neg (C, 3)$$

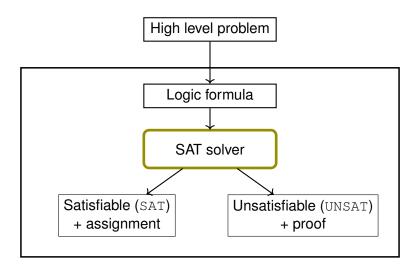
Clause represented as a set:

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

SAT design



SAT design



SAT Solving

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

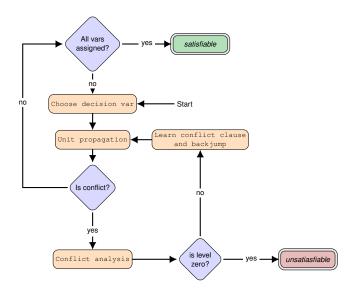
Good performance in practice:

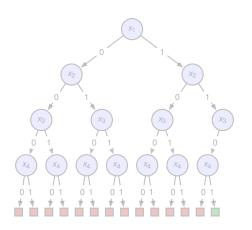
- Handle large problem (million variables and clauses)
- International SAT competition each year on academic and industrial problems

Enumerative algorithms:

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

CDCL in detail





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

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

$$\omega_{3} = \{x_{1}, x_{4}\}$$

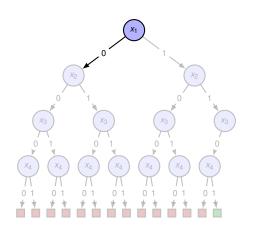
$$\omega_{4} = \{x_{2}, \neg x_{4}\}$$

$$\omega_{5} = \{x_{2}, x_{4}\}$$

$$\omega_{6} = \{x_{3}, x_{4}\}$$

$$\alpha = \{\}$$





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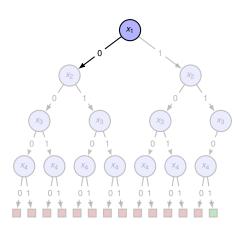
$$\omega_{3} = \{x_{1}, x_{4}\}$$

$$\omega_{4} = \{x_{2}, \neg x_{4}\}$$

$$\omega_{5} = \{x_{2}, x_{4}\}$$

$$\omega_{6} = \{x_{3}, x_{4}\}$$

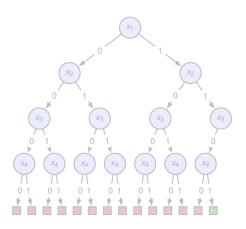
$$\alpha = \{\neg x_1\}$$



$$\alpha = \{\neg x_1\}$$

Conflict Analysis

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

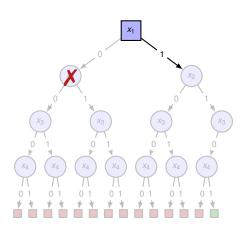


Learn conflict clause and backjump

$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\}
\omega_{2} = \{x_{1}, \neg x_{4}\}
\omega_{3} = \{x_{1}, x_{4}\}
\omega_{4} = \{x_{2}, \neg x_{4}\}
\omega_{5} = \{x_{2}, x_{4}\}
\omega_{6} = \{x_{3}, x_{4}\}
\omega_{7} = \{x_{1}\}$$

$$\alpha = \{\}$$





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

$$\omega_{3} = \{x_{1}, x_{4}\}$$

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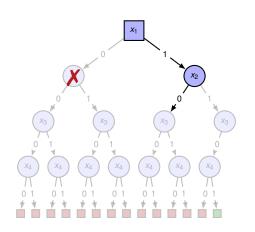
$$\omega_{5} = \{x_{2}, x_{4}\}$$

$$\omega_{6} = \{x_{3}, x_{4}\}$$

$$\omega_{7} = \{x_{1}\}$$

$$\alpha = \{x_1\}$$





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

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

$$\omega_{3} = \{x_{1}, x_{4}\}$$

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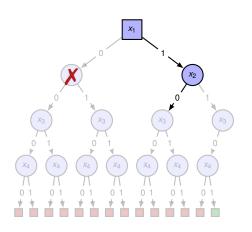
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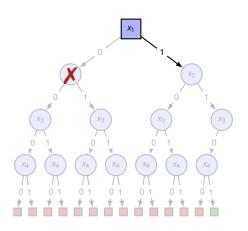
$$\alpha = \{x_1, \neg x_2\}$$





$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\}
\omega_{2} = \{x_{1}, \neg x_{4}\}
\omega_{3} = \{x_{1}, x_{4}\}
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\omega_{5} = \{x_{2}, x_{4}\}
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\omega_{7} = \{x_{1}\}$$

$$\alpha = \{x_1, \neg x_2\}$$

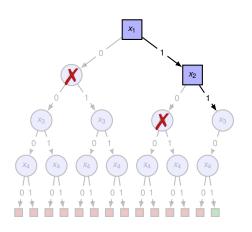


Learn conflict clause and backjump

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\omega_{5} = \{X_{2}, X_{4}\}
\omega_{6} = \{X_{3}, X_{4}\}
\omega_{7} = \{X_{1}\}
\omega_{8} = \{X_{2}\}$$

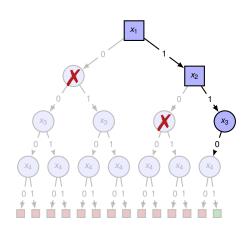
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\omega_{3} = \{x_{1}, x_{4}\}
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\omega_{5} = \{x_{2}, x_{4}\}
\omega_{6} = \{x_{3}, x_{4}\}
\omega_{7} = \{x_{1}\}
\omega_{8} = \{x_{2}\}$$

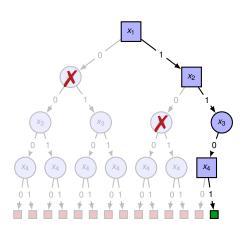
$$\alpha = \{x_1, x_2\}$$



$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\}
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\omega_{4} = \{x_{2}, \neg x_{4}\}
\omega_{5} = \{x_{2}, x_{4}\}
\omega_{6} = \{x_{3}, x_{4}\}
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\omega_{8} = \{x_{2}\}$$

$$\alpha = \{\mathbf{x_1}, \mathbf{x_2}, \neg \mathbf{x_3}\}$$

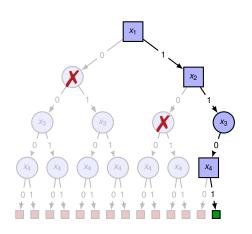




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$$\alpha = \{x_1, x_2, \neg x_3, x_4\}$$

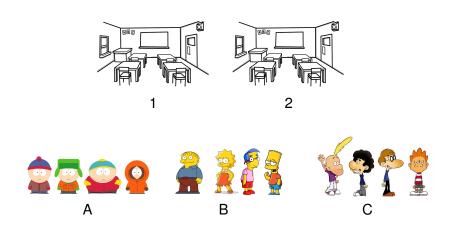


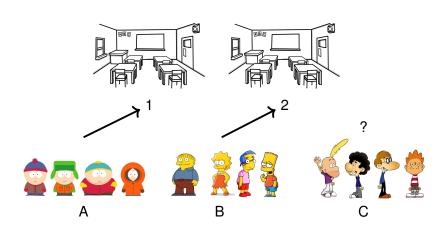


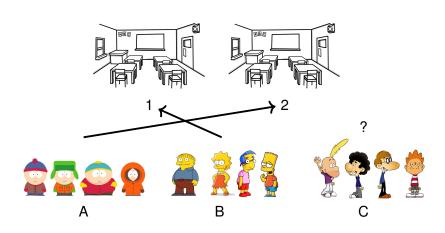
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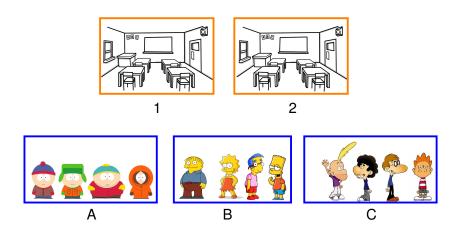
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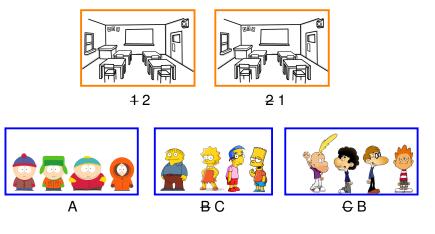
SAT and symmetries











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

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

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$$\begin{array}{c} \omega_1 = \{x_1, x_2, x_3\} \\ \omega_2 = \{x_4, x_5, x_6\} \\ \omega_3 = \{x_7, x_8, x_9\} \\ \omega_4 = \{-x_1, -x_4\} \\ \omega_5 = \{-x_1, -x_7\} \\ \omega_6 = \{-x_4, -x_7\} \\ \omega_8 = \{-x_2, -x_8\} \\ \omega_9 = \{-x_5, -x_8\} \\ \omega_9 = \{-x_5, -x_8\} \\ \omega_{11} = \{-x_3, -x_6\} \\ \omega_{11} = \{-x_3, -x_9\} \\ \omega_{12} = \{-x_6, -x_9\} \\ \end{array}$$

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Equi-satisfiability

$$\alpha \models \varphi \Leftrightarrow \mathbf{g}.\alpha \models \varphi$$

Computing symmetries of a SAT problem

CNF formula

$$\begin{array}{l} (x_1 \vee x_2 \vee x_3^*) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_7 \vee x_8 \vee x_9) \\ \wedge (\neg x_1 \vee \neg x_4) \wedge (\neg x_1 \vee \neg x_7) \wedge (\neg x_4 \vee \neg x_7) \\ \wedge (\neg x_2 \vee \neg x_5) \wedge (\neg x_2 \vee \neg x_8) \wedge (\neg x_5 \vee \neg x_8) \\ \wedge (\neg x_3 \vee \neg x_6) \wedge (\neg x_3 \vee \neg x_9) \wedge (\neg x_6 \vee \neg x_9) \end{array}$$

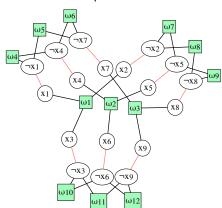
Computing symmetries of a SAT problem $(x_1 \lor x_2 \lor x_3) \land (x_4 \lor x_5 \lor x_6) \land (x_7 \lor x_8 \lor x_9)$

CNF formula

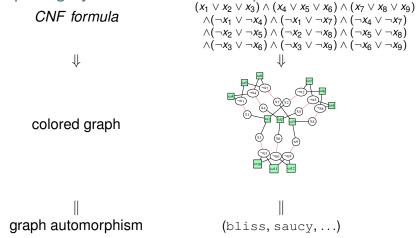


colored graph

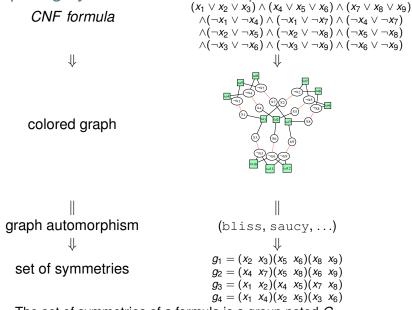




Computing symmetries of a SAT problem



Computing symmetries of a SAT problem



The set of symmetries of a formula is a group noted *G*

Exploitation of symmetries

Static symmetry breaking

Orbit of an assignment α for a group G:

$${\it G}.\alpha = \{{\it g}.\alpha \mid {\it g} \in {\it G}\}$$

Orbit of an assignment α for a group G:

$$G.\alpha = \{g.\alpha \mid g \in G\}$$

Example:

full assignment

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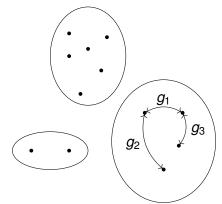


Orbit of an assignment α for a group G:

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Example:

- full assignment
- orbit



Equivalence relation with respect to SAT:

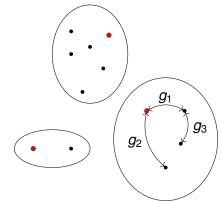
- Either $G.\alpha$ contains no solution
- Or all elements of $G.\alpha$ are solutions

Orbit of an assignment α for a group G:

$$G.\alpha = \{g.\alpha \mid g \in G\}$$

Example:

- full assignment
- orbit
 - representative



Equivalence relation with respect to SAT:

- Either $G.\alpha$ contains no solution
- Or all elements of $G.\alpha$ are solutions

Define an ordering relation to compare assignment (≺)

- Total ordering on variables
- Minimum value: F < T or T < F

Allow only minimal value (lex-leader)

Forbids other assignments in each orbit

→ Add Symmetry breaking predicates (SBP)

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Example:

$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8; \mathbb{F} < \mathbb{T}$$

 $g = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$

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Define an ordering relation to compare assignment (\prec)

- Total ordering on variables
- Minimum value: F < T or T < F

Allow only minimal value (lex-leader)

Forbids other assignments in each orbit

→ Add Symmetry breaking predicates (SBP)

Example:

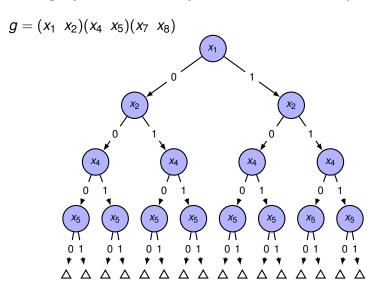
$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8; F < T$$

 $g = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$

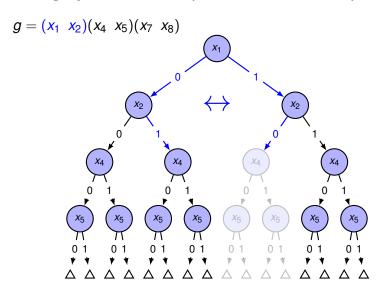
$$g.\alpha$$
 $r.0$
 $r.0$

Generate SBP
$$\omega = \{ \neg x_1, x_2 \}$$

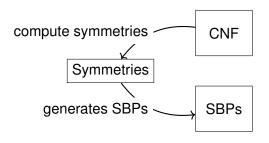
Using symmetries to prune the search space



Using symmetries to prune the search space



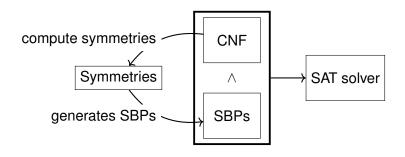
Static symmetry breaking



State-of-the-art approaches:

- Shatter [ASM06]
- BreakID [DBBD16]
- ...

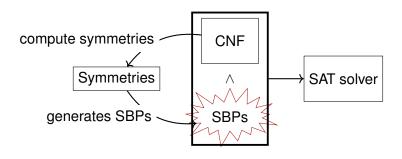
Static symmetry breaking



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Static symmetry breaking



State-of-the-art approaches:

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- ...

The solver can "explode" instead of being helped

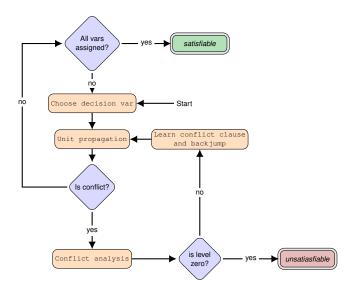
First contribution

CDCL[sym] Introducing Effective Symmetry Breaking in SAT Solving

TACAS'18 [MBCK18]

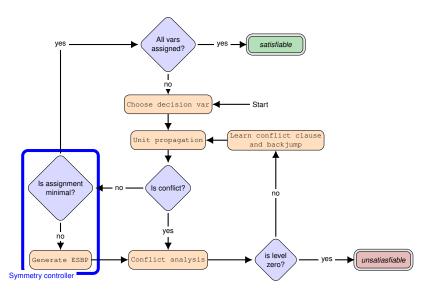
CDCL[Sym]

Compute and inject SBP opportunistically, during the solving



CDCL[Sym]

Compute and inject SBP opportunistically, during the solving



Symmetry status

- reducer: $g.\alpha \prec \alpha$
- inactive: $\alpha \prec g.\alpha$
- active: not enough information

Efficient implementation of symmetry status

Keep track the smallest unassigned variable x:

- **①** $\alpha(g.x) \leq \alpha(x)$, then *g* is reducer ⇒ Effective SBP (ESBP)
- 2 $\alpha(x) \leq \alpha(g.x)$, then g is inactive $\Rightarrow g$ cannot reduce α
- 3 $\alpha(g.x)$ or $\alpha(x)$ is unassigned then g is active

Update whenever variables are assigned / unassigned

$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8; F < T$$
 $g = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$

status of permutation g: active

$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8; F < T$$
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status of permutation g: reducer

Generate ESBP
$$\omega = \{\neg x_1, x_2\}$$

CDCL[Sym] implementation

- C++ Implementation
- Packaged as a library cosy¹ (Controller of Symmetry)
- Lightweight
- Fast update
- Low memory consumption
- Follows symmetry status
- Works with any enumerative SAT solver
- Can be integrated easily

 \rightarrow e.g. +3% LOC on MiniSAT².

¹https://github.com/lip6/cosy

²90 lines out of 3090

Experiments

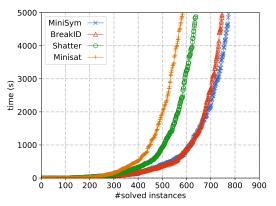
Benchmark:

- from SAT contests 2012 2017
- filter: bliss finds symmetries in 1000 seconds
- 36 % of instances, 1 350/3 700

Setup:

- four tools
 - MiniSat (no symmetry, baseline)
 - MiniSat + BreakID (SOTA SAT solver using symmetries)
 - MiniSat + Shatter (SOTA SAT solver using symmetries)
 - **MiniSym** = MiniSat + CDCL[Sym] (our approach)
- 5000 seconds timeout, 8GB memory
- includes time to compute symmetries (except for MiniSat)

Experimental results



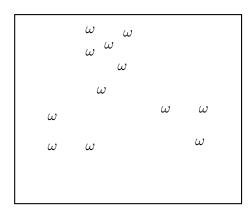
Solver	PAR-2	ALL	SAT	UNSAT
MiniSAT	2243h	586	325	261
Shatter	2088h	640	316	324
BreakID	1790h	749	334	415
MiniSym	1735h	775	336	439

Exploitation of symmetries

Dynamic symmetry breaking

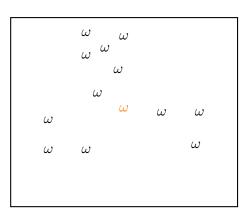
Learn symmetrical clauses

 $\begin{array}{ll} \square & \text{formula} \\ \omega & \text{clause} \end{array}$



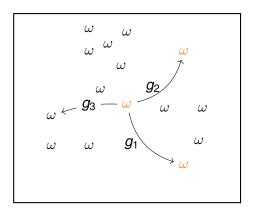
Learn symmetrical clauses

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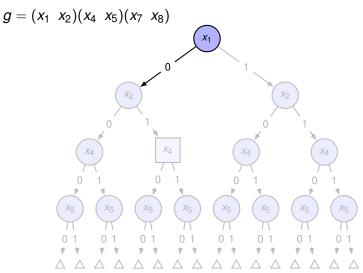


Learn symmetrical clauses

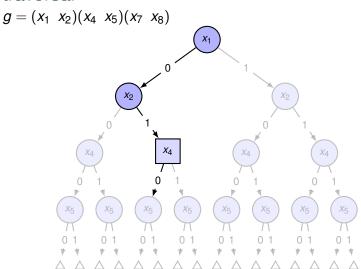
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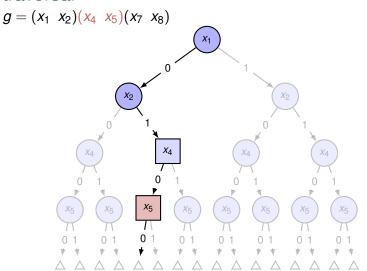
Using symmetries to accelerate the tree traversal



Using symmetries to accelerate the tree traversal

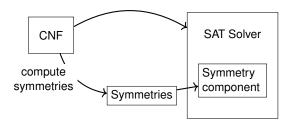


Using symmetries to accelerate the tree traversal



Use symmetries to deduce symmetrical facts.

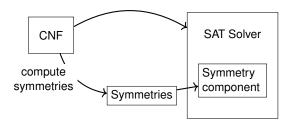
Dynamic Symmetry Breaking



State-of-the-art:

- Symmchaff [Sab05]
- Symmetry Propagation (SP) [DBdC⁺12]
- Symmetry Learning Scheme (SLS) [BNOS10]
- Symmetry Explanation Learning (SEL) [DBB17]

Dynamic Symmetry Breaking



State-of-the-art:

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Cannot handle some instances solved by static approach

Second contribution

Composing Symmetry Propagation and Effective Symmetry Breaking for SAT Solving

NFM'19 [MBK19]

ESBP + SP

Compose the symmetry propagation and the ESBP prune the decision tree while accelerating its traversal

Problems:

- ESBP breaks symmetries (incrementally)
- SP considers the manipulated symmetries valid all time

In a hybrid approach, SP must be able to identify valid symmetries

formula

 ω clause

 ω learnt clause

Local symmetries:

$$\omega \leftarrow \{g_1, g_2, g_3\}$$

formula

 ω clause

 ω learnt clause

 ω esbp

Local symmetries:

$$\omega \leftarrow \{g_1, g_2, g_3\}$$

- formula
- ω clause
- ω learnt clause
- ω esbp

g_1	$egin{array}{ccc} \omega & \omega & \omega & \omega & \omega \end{array}$			
	ω			
ω		ω	ω	
ω	ω		ω	

Local symmetries:

$$\omega \leftarrow \{g_1, g_2, g_3\}$$
$$\omega \leftarrow \{g_1\}$$

- Compute valid local symmetries
- On the fly
- At minimal cost

- formula
- ω clause
- ω learnt clause
- $^\omega$ esbp

g_1	$egin{pmatrix} \omega & & \omega \ & \omega & & \omega \end{matrix}$			
ω	ω	ω	ω	
ω	ω		ω	

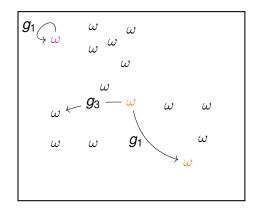
Local symmetries:

$$\omega \leftarrow \{g_1, g_2, g_3\}$$
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Inductive construction

- formula
- ω clause
- ω learnt clause
- ω esbp



Local symmetries:

$$\omega \leftarrow \{g_1, g_2, g_3\}$$
$$\omega \leftarrow \{g_1\}$$
$$\omega \leftarrow \{g_1, g_3\}$$

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- At minimal cost

Inductive construction

Experimental results

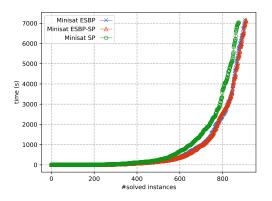
Benchmark:

- from SAT contests 2012 2018
- filter: bliss finds symmetries in 1000 seconds
- 1400 symmetric instances (out of 4000)

Setup:

- three tools
 - MiniSat SP (Minisat with Symmetry Propagation)
 - MiniSat ESBP (Minisat with CDCL[Sym])
 - Minisat ESBP-SP (our approach)
- 7200 seconds timeout

Experimental results



Solver	PAR-2	ALL	SAT	UNSAT
SP	1674h00	876	406	470
ESBP	1578h30	904	416	488
ESBP-SP	1570h15	911	420	491

Conclusion

- A new dynamic symmetry breaking approach
 - Generation of SBP on the fly
 - Package as a library cosy usable with any CDCL solver
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Perspectives

- Symmetries and parallel SAT solver
- Exploitation of partial symmetries

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Thanks!



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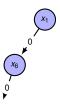
In AAAI, volume 5, pages 467-474, 2005.



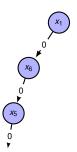
$$\omega_{1} = \{x_{1}, x_{2}, x_{3}\}
\omega_{2} = \{x_{4}, x_{5}, x_{6}\}
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\omega_{4} = \{\neg x_{2}, \neg x_{4}\}
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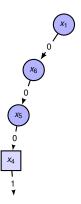
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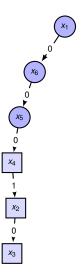
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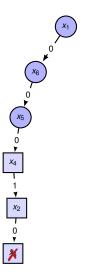
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$$\omega_7 = \{x_1, \neg x_4\}$$

Weakly active symmetries

Logical consequence

When ω is satisfied in all satisfying assignments of φ , we say that ω is a logical consequence of φ , and we denote this by $\varphi \vdash \omega$.

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Weakly active symmetries

Let a subset $\delta \subseteq \alpha$, a symmetry σ of φ such that $\varphi \cup \delta \vdash \varphi \cup \alpha \land \sigma.\delta \subseteq \alpha$ then σ is weakly active symmetry.

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Symmetry propagation

Let σ a weakly active symmetry, then

$$\varphi \cup \alpha \vdash \{I\} \Leftrightarrow \varphi \cup \alpha \vdash \sigma.\{I\}$$

Local symmetries

Logical consequence

When ω is satisfied in all satisfying assignments of φ , we say that ω is a logical consequence of φ , and we denote this by $\varphi \vdash \omega$.

Local Symmetries

Let φ be a formula. We define $L_{\omega,\varphi}$, the set of *local symmetries* for a clause ω , and with respect to a formula φ , as follows:

$$L_{\omega,\varphi} = \{ \sigma \in \mathfrak{S} \mid \varphi \vdash \sigma.\omega \}$$

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We can state that:

$$\bigcap_{\omega\in\varphi}L_{\omega,\varphi}\subseteq G.$$

Computing local symmetries

Formula can be decomposed as : $\varphi = \varphi_o \cup \varphi_e \cup \varphi_d$ where

- φ_o is the set of the original clauses
- φ_e is the set of ESBPs
- φ_d is the set of deduced clauses.

Local symmetries

- $\omega \in \varphi_o, L_{\omega,\varphi} \supseteq G$
- $\omega \in \varphi_e, L_{\omega,\varphi} \supseteq Stab(\omega) = \{ \sigma \in G \mid \omega = \sigma.\omega \}$
- $\omega \in \varphi_d, L_{\omega,\varphi} \supseteq (\bigcap_{\omega' \in \varphi_1} L_{\omega',\varphi}) \cup Stab(\omega)$

where φ_1 is the set of clauses that derives ω .

- Define lexicographic order
 - Define total order on variables
 - Define minimal value
- Forbid non minimal assignment for each orbit

$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8; \mathbb{F} < \mathbb{T}$$

 $g = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$

x ₁ x ₂ x ₃ x ₄ x ₅ ··· lex-leader	SBP

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<i>O</i> ₁	F	Т	-	-	-		✓	

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0	F	Т	_	-	-		✓ X	
<i>U</i> ₁	Т	F	–	–	-		X	$\rightarrow \neg x_1 \lor x_2$

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Ω_{t}	F	Т	-	_	-		✓ X	
——	Т	F	_	_	_		X	$\rightarrow \neg x_1 \lor x_2$
	F	F	-	F	Т		/	$\bigg \to x_1 \vee x_2 \vee \neg x_4 \vee x_5$
O_2	F	F	–	Т	F		×	$ \rightarrow x_1 \lor x_2 \lor \neg x_4 \lor x_5 $

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Example:

$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8; \mathbb{F} < \mathbb{T}$$

 $g = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$

	<i>x</i> ₁	<i>X</i> ₂	<i>X</i> ₃	X ₄	<i>x</i> ₅		lex-leader	SBP
	F	Т	-	-	-		✓ ×	
O ₁	Т	F	-	_	-		X	$\rightarrow \neg x_1 \lor x_2$
_	F	F	-	F	Т		/	
O_2	F	F	-	Т	F		×	$\rightarrow x_1 \vee x_2 \vee \neg x_4 \vee x_5$

. .

$$g_1 = (x_2 \ x_3)(x_5 \ x_6)(x_8 \ x_9)$$
 $g_2 = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$

$$F < T \quad x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8 \le x_9$$

$$\overline{g_1}$$

$$\alpha = \quad \underline{U} \quad \overline{U} \quad U \quad U \quad U \quad U \quad U \quad U$$

$$\underline{g_2} \quad \Box$$

$$g_1 = (x_2 \ x_3)(x_5 \ x_6)(x_8 \ x_9)$$
 $g_2 = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$

$$F < T \quad x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8 \le x_9$$

$$\overline{g_1}$$

$$\alpha = \underline{T} \quad F \quad F \quad U \quad \overline{U} \quad U \quad U \quad U \quad U$$

$$\underline{g_2} \quad \Box$$

$$g_2$$
 generates ESBP $\omega = \{\neg x_1, x_2\}$

- 1 reducer: $\alpha(g.x) \leq \alpha(x)$
- 2 inactive: $\alpha(x) \leq \alpha(g.x)$
- 3 active: $\alpha(g.x)$ or $\alpha(x)$ is unassigned

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$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8$$
 ; F < T
 $g_1 = \begin{pmatrix} x_2 & x_3 \end{pmatrix} \begin{pmatrix} x_5 & x_6 \end{pmatrix} \begin{pmatrix} x_8 & x_9 \end{pmatrix} \begin{vmatrix} x = x_2 & g.x = x_3 \\ & & \text{active} \end{pmatrix}$ $g_2 = \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} x_4 & x_5 \end{pmatrix} \begin{pmatrix} x_7 & x_8 \end{pmatrix} \begin{vmatrix} x = x_1 & g.x = x_2 \\ & & \text{active} \end{pmatrix}$ \cdots $\alpha = \{ \neg x_2 \}$

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 $g_2 = \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} x_4 & x_5 \end{pmatrix} \begin{pmatrix} x_7 & x_8 \end{pmatrix} \begin{vmatrix} x = x_1 & g.x = x_2 \\ & \text{reducer} \end{pmatrix}$

 $\alpha = \{ \neg x_2, \neg x_3, x_1 \}$

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 \cdots
 $\alpha = \{ \neg x_2, \neg x_3, x_1 \}$

$$g_2$$
 generates $\omega = \{\neg x_1, x_2\}$

Encoding the problem

(A, 1)(A, 2)(A, 3) (B, 1)(B, 2)(B, 3) (C, 1)(C, 2)(C, 3)	$X_1 \lor X_2 \lor X_3 $ $X_4 \lor X_5 \lor X_6 $ $X_7 \lor X_8 \lor X_6 $
$\neg (A, 1) \neg (B, 1)$ $\neg (A, 1) \neg (C, 1)$ $\neg (B, 1) \neg (C, 1)$	$ \neg X_1 \lor \neg X_4 \neg X_1 \lor \neg X_7 \neg X_4 \lor \neg X_7 $
$\neg (A,2) \neg (B,2)$ $\neg (A,2) \neg (C,2)$ $\neg (B,2) \neg (C,2)$	$\neg x_2 \lor \neg x_5 \neg x_2 \lor \neg x_8 \neg x_5 \lor \neg x_8$
$\neg (A,3) \neg (B,3)$ $\neg (A,3) \neg (C,3)$ $\neg (B,3) \neg (C,3)$	$ \begin{array}{c} \neg X_3 \lor \neg X_6 \\ \neg X_3 \lor \neg X_9 \\ \neg X_6 \lor \neg X_9 \end{array} $