Exploitation of dynamic symmetries for solving SAT problems

Doctorat de Sorbonne Université

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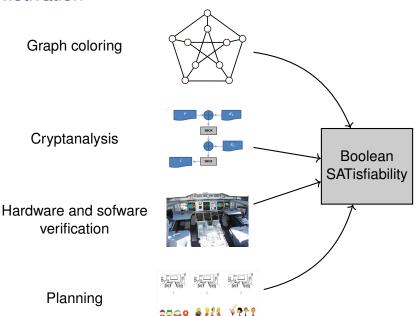


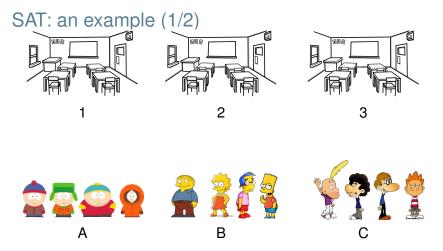


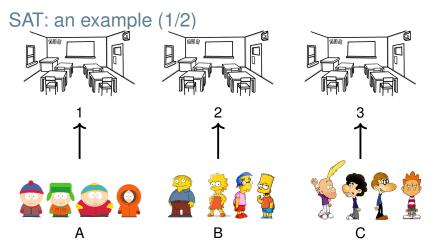




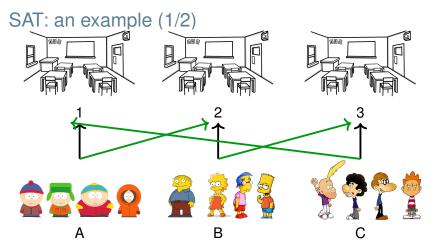
Motivation





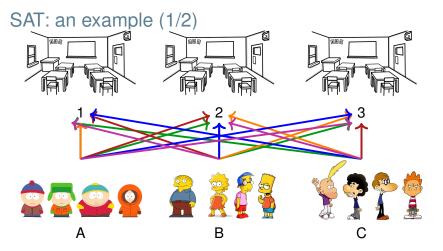


YES! SATisfiable
$$\alpha = \{(A, 1), (B, 2), (C, 3)\}$$



YES! SATISFIABLE
$$\alpha = \{(A, 1), (B, 2), (C, 3)\}$$

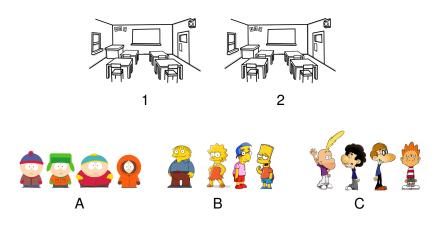
Many solutions $\alpha' = \{(A, 2), (B, 3), (C, 1)\}$



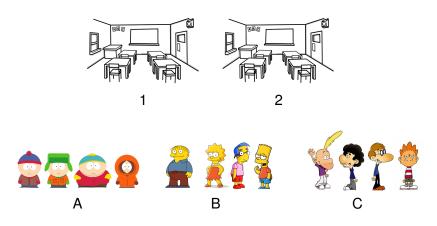
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:

SAT: an example (2/2)



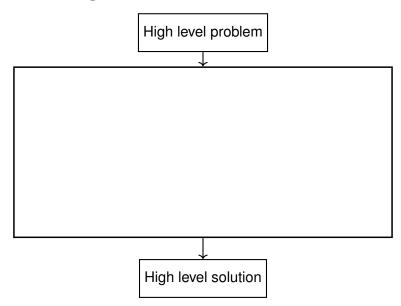
SAT: an example (2/2)



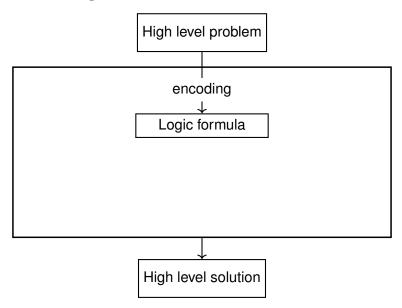
Is it possible to attribute each group to a unique classroom?

No! UNSATisfiable

From high level problem to the solution through SAT solving



From high level problem to the solution through SAT solving



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

 $(x_1 \lor x_2 \lor x_3) \land$



$$\overbrace{(A,1)(A,2)(A,3)}^{x_1}(A,1)(A,2)(A,3)$$
(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) \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)
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\neg (B, 3) \neg (C, 3)
\neg (B, 3) \neg (C, 3)
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(x_1 \lor x_2 \lor x_3)
(x_1 \lor x_2 \lor x_4)
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(x_2 \lor x_3)
(x_4 \lor x_5 \lor x_6)
(x_2 \lor x_3)
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(x_3 \lor x_4)
(x_4 \lor x_5 \lor x_8)
(x_5 \lor x_8)
(x_6 \lor x_9)
(x_6 \lor x_9)$$$$

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

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

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

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

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

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

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

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

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

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

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

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

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

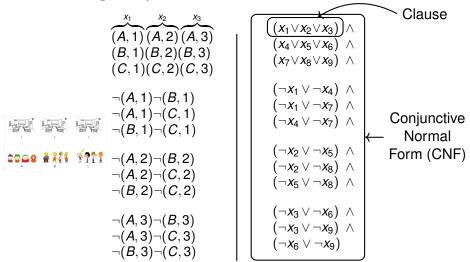
Clause

$$(x_1 \lor x_2 \lor x_3) \land (x_4 \lor x_5 \lor x_6) \land (x_7 \lor x_8 \lor x_9) \land$$

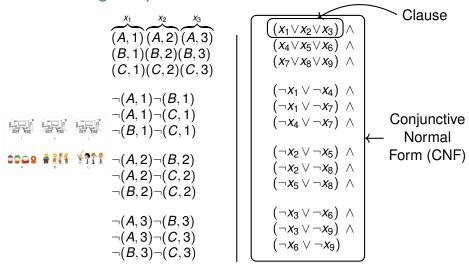
$$\begin{array}{c} (\neg x_1 \lor \neg x_4) \land \\ (\neg x_1 \lor \neg x_7) \land \\ (\neg x_4 \lor \neg x_7) \land \end{array}$$

$$\begin{array}{c} (\neg x_2 \lor \neg x_5) \land \\ (\neg x_2 \lor \neg x_8) \land \\ (\neg x_5 \lor \neg x_8) \land \end{array}$$

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



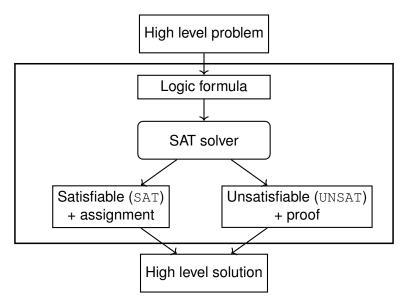
Any Boolean formula can be transformed into CNF in polynomial time



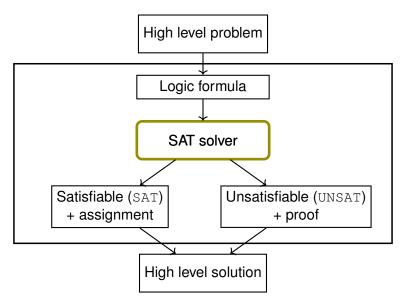
Clause represented as a set:

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

From high level problem to the solution through SAT solving



From high level problem to the solution through SAT solving



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

SAT Solving

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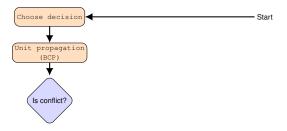
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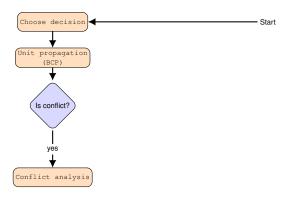
Enumerative algorithms:

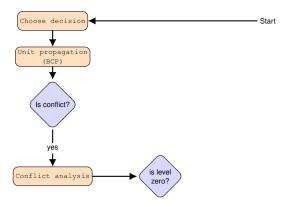
- Davis, Putnam, Logemann, and Loveland (DPLL) [DLL62]
 - Boolean Constraint Propagation (BCP)
- Conflict Driven Clause Learning (CDCL) [MSS99]
 - Derived from DPLL
 - Clause learning

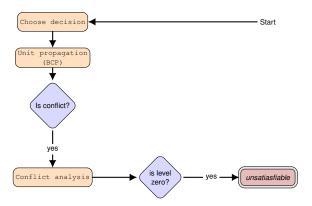


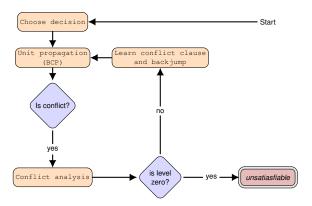


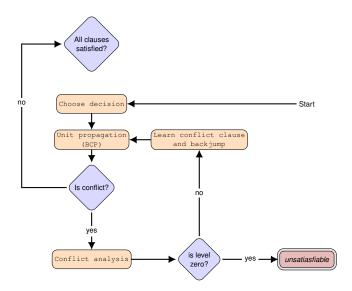


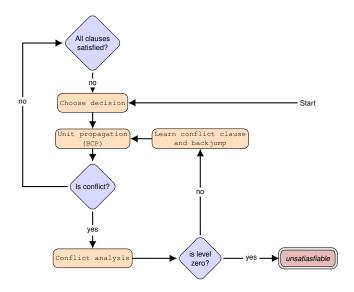


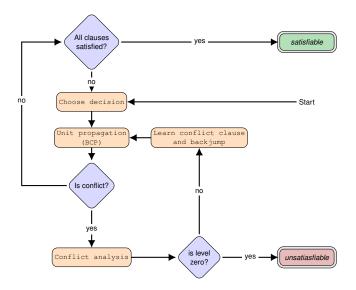


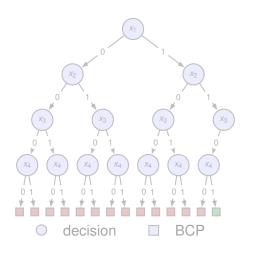












$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\}$$

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

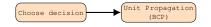
$$\omega_{3} = \{x_{1}, x_{4}\}$$

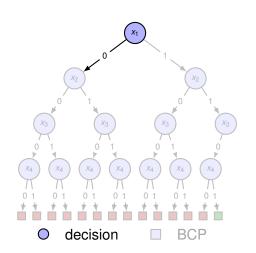
$$\omega_{4} = \{x_{2}, \neg x_{4}\}$$

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

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

$$\alpha = \{\}$$

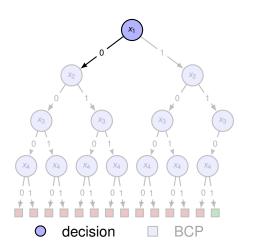




$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\}
\omega_{2} = \{x_{1}, \neg x_{4}\}
\omega_{3} = \{x_{1}, x_{4}\}
\omega_{4} = \{x_{2}, \neg x_{4}\}
\omega_{5} = \{x_{2}, x_{4}\}
\omega_{6} = \{x_{3}, x_{4}\}$$

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





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

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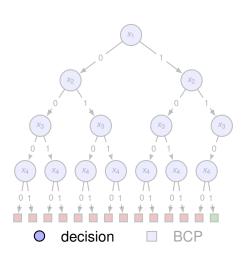
$$\omega_{3} = \{x_{1}, x_{4}\}$$

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

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

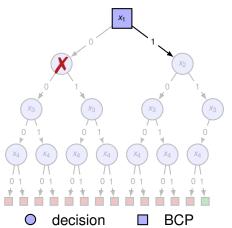


$$\alpha = \{\}$$

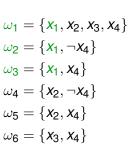
Learn conflict clause and backjump

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

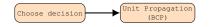


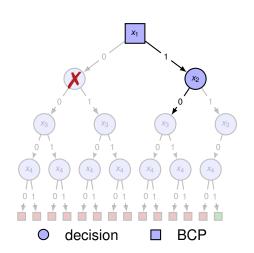


$$\alpha = \{x_1\}$$



 $\omega_7 = \{x_1\}$

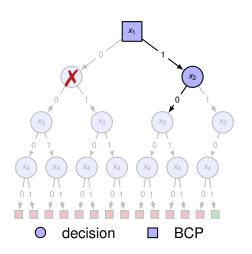




$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\}
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\omega_{3} = \{x_{1}, x_{4}\}
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\omega_{7} = \{x_{1}\}$$

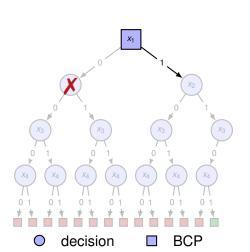
$$\alpha = \{x_1, \neg x_2\}$$





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

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

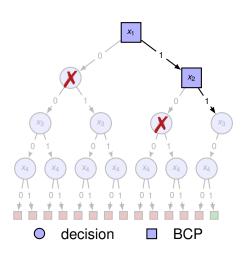


$$\alpha = \{x_1\}$$

Learn conflict clause and backjump

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

Unit Propagation (BCP)



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

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

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

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

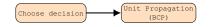
$$\omega_{5} = \{x_{2}, x_{4}\}$$

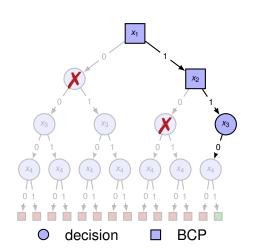
$$\omega_{6} = \{x_{3}, x_{4}\}$$

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

$$\omega_{8} = \{x_{2}\}$$

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

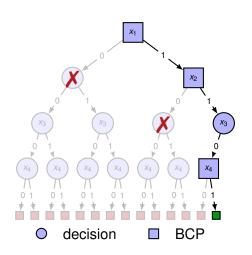




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\omega_{8} = \{x_{2}\}$$

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

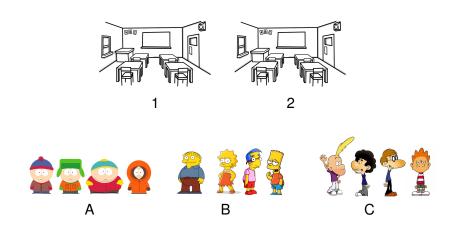


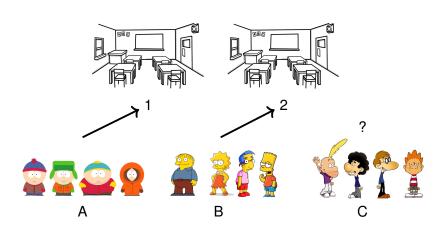


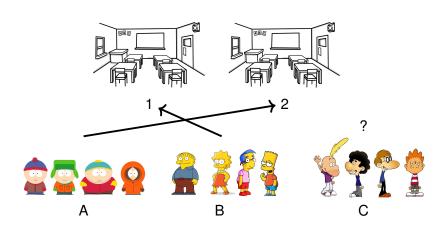
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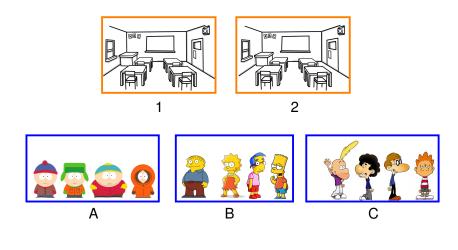
$$\alpha = \{x_1, x_2, \neg x_3, x_4\}$$

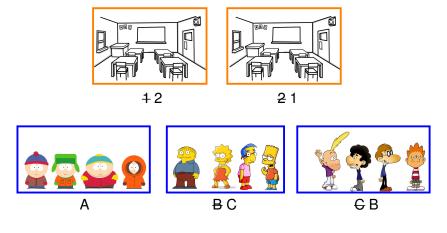
SAT and symmetries





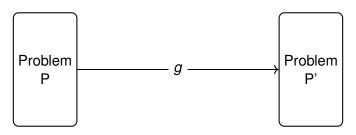






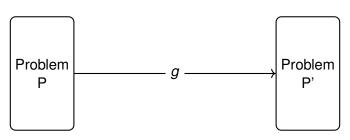
Symmetry in high level

g: a symmetry



Symmetry in high level

g: a symmetry

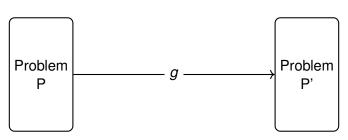


Equi-satisfiability

 $solution \models P \Leftrightarrow g.solution \models P'$

Symmetry in high level

g: a symmetry



Equi-satisfiability

$$solution \models P \Leftrightarrow g.solution \models P'$$

Semantic symmetries

Syntactic symmetries

Syntactic symmetry

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

Syntactic symmetry

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

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

Syntactic symmetry

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$$\begin{array}{c} \omega_1 = \{x_1, x_2, x_3\} \\ \omega_2 = \{x_4, x_5, x_6\} \\ \omega_3 = \{x_7, x_8, x_9\} \\ \omega_4 = \{-x_1, -x_4\} \\ \omega_5 = \{-x_1, -x_7\} \\ \omega_6 = \{-x_4, -x_7\} \\ \omega_6 = \{-x_4, -x_7\} \\ \omega_8 = \{-x_2, -x_8\} \\ \omega_9 = \{-x_5, -x_8\} \\ \omega_{10} = \{-x_3, -x_6\} \\ \omega_{11} = \{-x_3, -x_6\} \\ \omega_{11} = \{-x_3, -x_8\} \\ \omega_{11} =$$

g.P = P' = P

 $\omega_{12} = \{ \neg x_6, \neg x_9 \}$ $\omega_{12} = \{ \neg x_6, \neg x_9 \}$

Computing symmetries of a SAT problem

CNF formula

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

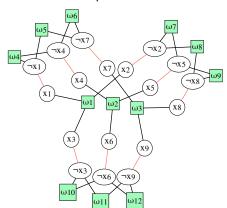
Computing symmetries of a SAT problem $(x_1 \lor x_2 \lor x_3) \land (x_4 \lor x_5 \lor x_6) \land (x_7 \lor x_8 \lor x_9)$

CNF formula

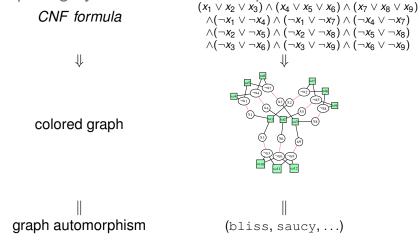


colored graph





Computing symmetries of a SAT problem



Computing symmetries of a SAT problem

The set of symmetries of a formula is a group noted < G, \circ >

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:

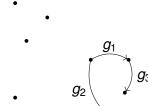
full assignment

Orbit of an assignment α for a group G:

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Example:

full assignment

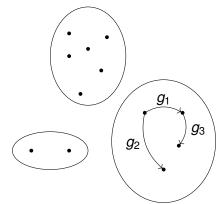


Orbit of an assignment α for a group G:

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Example:

- · full assignment
- orbit



Equivalence relation with respect to SAT:

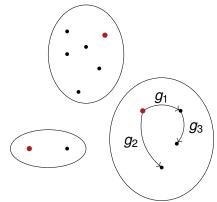
- Either $G.\alpha$ contains no solution
- Or all elements of $G.\alpha$ are solutions

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

Comparing assignments: Assessments

Define an ordering relation to compare assignments (\prec)

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

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Allow only minimal value (lex-leader)

Comparing assignments: Assessments

Define an ordering relation to compare assignments (\prec)

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

Allow only minimal value (lex-leader)

Forbid other assignments in each orbit

→ Add all symmetry breaking predicates (SBP) statically

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

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

Assignments:

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

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

Assignments:

	<i>X</i> ₁	<i>X</i> ₂	<i>X</i> 3	<i>X</i> ₄	<i>X</i> 5	<i>x</i> ₆	<i>X</i> 7	<i>X</i> ₈
α	Т	F	F	F	F	F	F	F
$oldsymbol{g}.lpha$	F	Т	F	F	F	F	F	F

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Assignments:

Comparing:

$$g.\alpha \prec \alpha$$

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

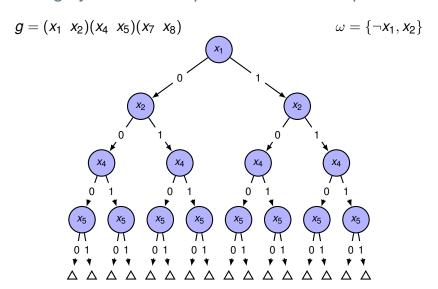
Symmetry: $g = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$

Assignments:

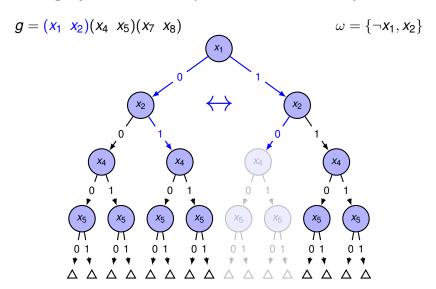
Comparing:

$$g.\alpha \prec \alpha \Rightarrow \mathsf{SBP}: \omega = \{\neg x_1, x_2\}$$

Using symmetries to prune the search space



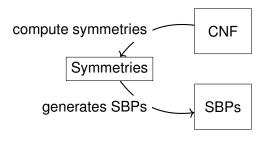
Using symmetries to prune the search space



State-of-the-art:

Shatter [ASM06]

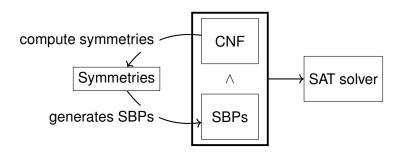
BreakID [DBBD16]



State-of-the-art:

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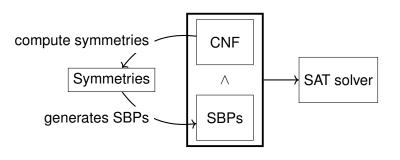
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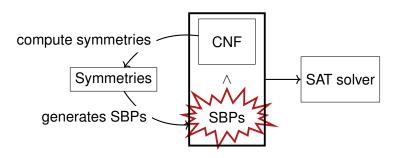


Works well on many symmetrical problems

State-of-the-art:

Shatter [ASM06]

BreakID [DBBD16]



Works well on many symmetrical problems

The solver can "explode" instead of being helped

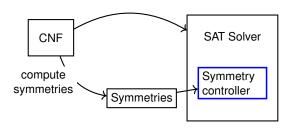
- generate not needed clause
- flooding the solver

First contribution:

CDCL[sym] Introducing Effective Symmetry Breaking in SAT Solving

TACAS'18 [MBCK18]

General idea

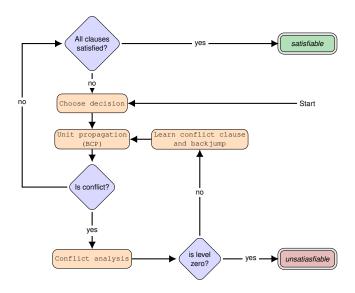


Symmetry controller:

- Generates SBP on-the-fly
- Only when needed
- Intrusive on solver

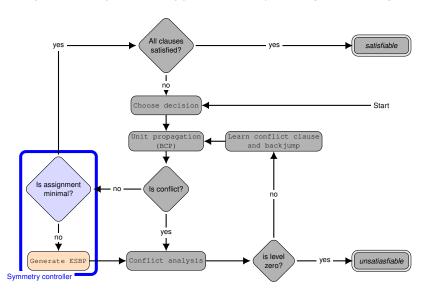
CDCL[Sym]

Compute and inject SBP opportunistically, during the solving



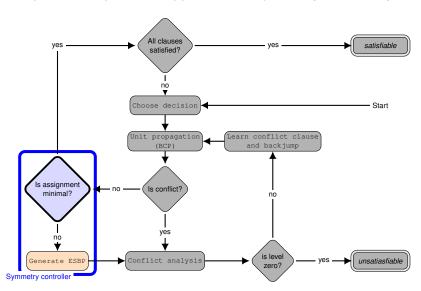
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CDCL[Sym]

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Is assignment minimal?

Our proposal: Symmetry status tracking

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

Is assignment minimal?

Our proposal: Symmetry status tracking

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

Efficient implementation of symmetry status tracking

Keep track the smallest unassigned variable *x*:

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

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

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

$$g.\alpha$$
 α

status of permutation g: active

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

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status of permutation g: reducer

On-the-fly generation of ESBP: $\omega = \{\neg x_1, x_2\}$

CDCL[Sym] implementation

- C++ Implementation: 1780 Loc
- Packaged as a library cosy (Controller of Symmetry)

```
https://github.com/lip6/cosy
```

Low memory consumption

- Virtually works with any enumerative CDCL SAT solver
- Can be easily integrated

```
ightarrow e.g. +3% LOC on MiniSAT 90 lines out of 3090
```

Experiments

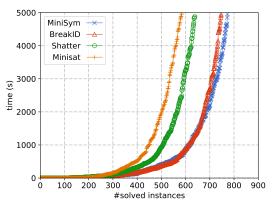
Benchmark:

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

Setup:

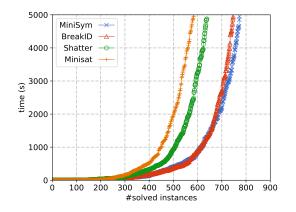
- four tools
 - MiniSat (no symmetry, baseline)
 - MiniSat + BreakID (SOTA SAT solver using symmetries)
 - MiniSat + Shatter (SOTA SAT solver using symmetries)
 - MiniSym = MiniSat + cosy (our approach)
- 5000 seconds timeout, 8GB memory
- includes time to compute symmetries (except for MiniSat)

Experimental results



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

Experimental results



Number of SBPs	BreakID	MiniSym
UNSAT (399)	2 576 349	913 339
SAT (320)	12 179 513	457 452

Discussion of the results

Change the ordering relation

- Choose another lex-leader
- Generate other SBP

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Change the ordering relation

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Composing permutations

- Observe more variables
- Earlier generation of ESBP

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Change the ordering relation

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Composing permutations

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Adapt the solver heuristics dynamically

- Restart
- Cleaning database

Exploitation of symmetries:

Dynamic symmetry breaking

Learn symmetrical clauses

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

```
\omega_8
                       \omega_5
                                                    \omega_1
                                                                     \omega_2
\omega10
                                                                 \omega11
 \omega_{4}
                  \omega_3
```

Learn symmetrical clauses

 $\begin{array}{cc} \mathbf{\Box} & \text{formula} \\ \omega & \text{clause} \\ \end{array}$

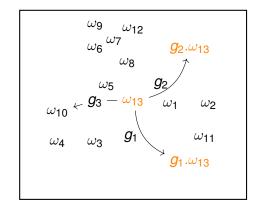
 ω learnt clause

```
\omega_5
                                \omega13
                                                  \omega_1
                                                                   \omega_2
\omega10
                                                                \omega11
 \omega_4
                 \omega_3
```

Learnt clauses are logical consequences of the formula

Learn symmetrical clauses

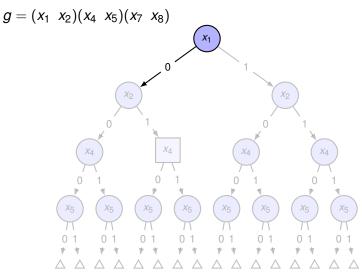
- $\begin{array}{cc} \mathbf{\square} & \text{formula} \\ \omega & \text{clause} \end{array}$
- ω learnt clause



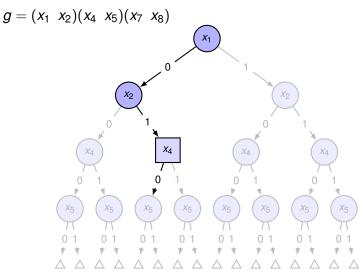
Learnt clauses are logical consequences of the formula

- $g \in G$ are symmetries of the formula
- → symmetrical learnt clauses are logical consequences too

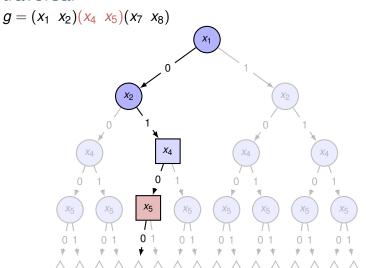
Using symmetries to accelerate the tree traversal



Using symmetries to accelerate the tree traversal



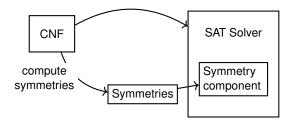
Using symmetries to accelerate the tree traversal



Use symmetries to deduce symmetrical facts.

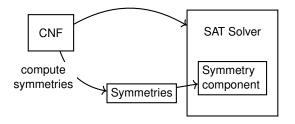
State-of-the-art:

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



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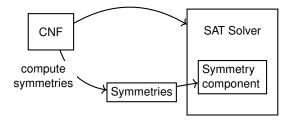


Observations:

Solve some instances very quickly

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Observations:

Solve some instances very quickly Cannot handle some instances solved by static approach

Second contribution

Composing Symmetry Propagation and Effective Symmetry Breaking for SAT Solving

NFM'19 [MBK19]

Composing ESBP and SP

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

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Problems:

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

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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

Is valid symmetry?

Our proposal: 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:

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

We can state that:

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

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We can state that:

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Guarantee that symmetrical clauses are logical consequences of the formula

formula

 ω clause

 ω learnt clause

$$\begin{array}{cccc} & \omega_9 & \omega_{12} \\ & \omega_6 & \omega_7 \\ & & \omega_8 \\ & & \omega_5 \\ & & \omega_{10} & & \omega_2 \\ & & & \omega_4 & \omega_3 & & \omega_{11} \end{array}$$

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

formula

 $^{\omega}$ clause

 ω learnt clause

 ω ESBP

```
\omega9
                                \omega_{12}
                               \omega_8
            \omega13
                     \omega_5
                                                  \omega_1
                                                                   \omega_2
\omega10
                                                                \omega11
 \omega_4
                 \omega_3
```

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

- formula
- ω clause
- ω learnt clause
- ω ESBP

$g_1, g_2 \xrightarrow{\omega_6} \begin{matrix} \omega_9 & \omega_{12} \\ \omega_7 & \omega_7 \\ \omega_{13} & \omega_5 \end{matrix}$			
ω_{10}	ω_1	ω_2	
ω_4 ω_3		ω_{11}	

$$\omega \leftarrow \{g_1, g_2, g_3\}$$
$$\omega_{13} \leftarrow \{g_1, g_2\}$$

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

- formula
- ω clause
- ω learnt clause
- ω ESBP

$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	g_1, g_3 ω_{14}
ω_{10}	ω_1 ω_2
ω_4 ω_3	ω 11

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

 $\omega_{13} \leftarrow \{g_1, g_2\}$
 $\omega_{14} \leftarrow \{g_1, g_3\}$

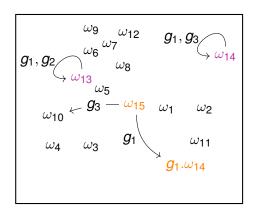
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formula

 ω clause

 ω learnt clause

 ω ESBP



Local symmetries:

$$\omega \leftarrow \{g_1, g_2, g_3\}$$
 $\omega_{13} \leftarrow \{g_1, g_2\}$
 $\omega_{14} \leftarrow \{g_1, g_3\}$
 $\omega_{15} \leftarrow \{g_1, g_3\}$

- Compute valid local symmetries
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Inductive construction

Experimental results

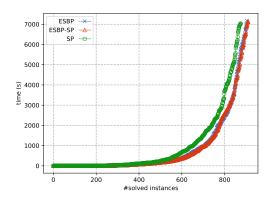
Benchmark:

- 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 cosy)
 - Minisat ESBP-SP (our approach)
- 7200 seconds timeout

Experimental results



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

Discussion of the results

SP and ESBP have separated symmetry managers \rightarrow costly

Discussion of the results

SP and ESBP have separated symmetry managers → costly

Combine ESBP with Symmetry Explanation Learning (SEL)

- SEL have less requirements than SP
- We believe that this will improves the performance

Conclusion

- A new dynamic symmetry breaking approach
 - Generation of SBP on the fly
 - Package as a library cosy usable with any CDCL solver
- A new hybrid approach (ESBP-SP)
 - Take advantage of static and dynamic approaches

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Perspectives

- Exploitation of partial symmetries
- Symmetries and parallel SAT solver

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Thanks!



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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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Symmetry propagation

Let σ a weakly active symmetry, then

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

Local symmetries

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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

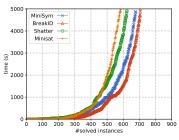
- φ_o is the set of the original clauses
- φ_e is the set of ESBPs
- φ_d is the set of deduced clauses.

Local symmetries

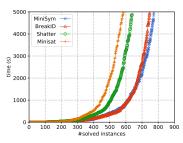
- $\omega \in \varphi_o, L_{\omega,\varphi} \supseteq G$
- $\omega \in \varphi_e, L_{\omega,\varphi} \supseteq Stab(\omega) = \{ \sigma \in G \mid \omega = \sigma.\omega \}$
- $\omega \in \varphi_d, L_{\omega,\varphi} \supseteq (\bigcap_{\omega' \in \varphi_1} L_{\omega',\varphi}) \cup Stab(\omega)$

where φ_1 is the set of clauses that derives ω .

Experimental results



(a) with saucy3

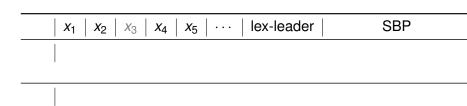


(b) with bliss

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

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

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



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

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

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

	<i>x</i> ₁	<i>X</i> ₂	<i>X</i> ₃	<i>x</i> ₄	<i>X</i> ₅		lex-leader	SBP
<i>O</i> ₁	F	Т	–	-	-		✓	

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

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	F	Т	-	_	-		✓ ×	
O_1	Т	F	–	_	-		x	$\rightarrow \neg x_1 \lor x_2$
	I							

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	<i>x</i> ₁	<i>X</i> ₂	<i>X</i> ₃	<i>x</i> ₄	<i>x</i> ₅		lex-leader	SBP
O ₁	F	T	_	_	_		✓ x	$\rightarrow \neg x_1 \lor x_2$

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	<i>x</i> ₁	<i>x</i> ₂	<i>X</i> ₃	<i>x</i> ₄	<i>x</i> ₅		lex-leader	SBP
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		×	$\rightarrow x_1 \vee x_2 \vee \neg x_4 \vee x_5$

- Define lexicographic order
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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		×	$\rightarrow x_1 \vee x_2 \vee \neg x_4 \vee x_5$

. .

$$g_1 = (x_2 \ x_3)(x_5 \ x_6)(x_8 \ x_9)$$
 $g_2 = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$
 $F < T \ x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8 \le x_9$
 g_1
 $\alpha = \underbrace{U} \ \overline{U} \ U \ U \ U \ U \ U \ U \ U$

$$g_1 = (x_2 \ x_3)(x_5 \ x_6)(x_8 \ x_9)$$
 $g_2 = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$
 $F < T \ x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8 \le x_9$
 g_1
 $\alpha = \underbrace{U}_F \quad U \quad U \quad U \quad U \quad U \quad U \quad U$
 g_2

$$g_1 = (x_2 \ x_3)(x_5 \ x_6)(x_8 \ x_9)$$

$$g_2 = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$$

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

$$\overline{g_1}$$

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

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

$$g_2$$
 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

$$x_1 \leq x_2 \leq x_3 \leq x_4 \leq x_5 \leq x_6 \leq x_7 \leq x_8$$
 ; F < T $g_1 = (x_2 \quad x_3) \quad (x_5 \quad x_6) \quad (x_8 \quad x_9) \mid x = x_2 \quad g.x = x_3$ active $g_2 = (x_1 \quad x_2) \quad (x_4 \quad x_5) \quad (x_7 \quad x_8) \mid x = x_1 \quad g.x = x_2$ active $\alpha = \{$

- 1 reducer: $\alpha(g.x) \leq \alpha(x)$
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$$x_1 \leq x_2 \leq x_3 \leq x_4 \leq x_5 \leq x_6 \leq x_7 \leq x_8$$
 ; 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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- 1 reducer: $\alpha(g.x) \leq \alpha(x)$
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 ; F < T
 $g_1 = \begin{pmatrix} x_2 & x_3 \end{pmatrix} \begin{pmatrix} x_5 & x_6 \end{pmatrix} \begin{pmatrix} x_8 & x_9 \end{pmatrix} \begin{vmatrix} x = x_5 & g.x = x_6 \\ & \text{active} \end{pmatrix}$ $g_2 = \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} x_4 & x_5 \end{pmatrix} \begin{pmatrix} x_7 & x_8 \end{pmatrix} \begin{vmatrix} x = x_1 & g.x = x_2 \\ & & \text{reducer} \end{pmatrix}$... $\alpha = \{ \neg x_2, \neg x_3, x_1 \}$

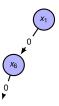
 g_2 generates $\omega = \{ \neg x_1, x_2 \}$



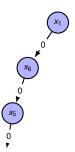
$$\omega_{1} = \{x_{1}, x_{2}, x_{3}\}
\omega_{2} = \{x_{4}, x_{5}, x_{6}\}
\omega_{3} = \{\neg x_{1}, \neg x_{5}\}
\omega_{4} = \{\neg x_{2}, \neg x_{4}\}
\omega_{5} = \{\neg x_{3}, \neg x_{4}\}
\omega_{6} = \{\neg x_{3}, \neg x_{6}\}$$



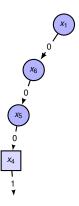
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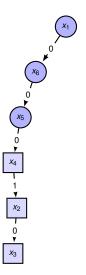
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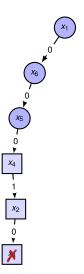
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\omega_{5} = \{\neg x_{3}, \neg x_{4}\}
\omega_{6} = \{\neg x_{3}, \neg x_{6}\}$$

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