Exploitation of dynamic symmetries for solving SAT problems

Doctorat de Sorbonne Université

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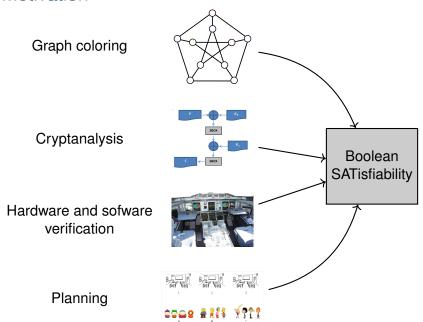


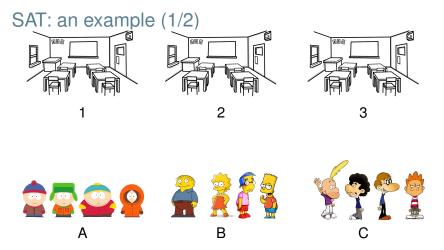


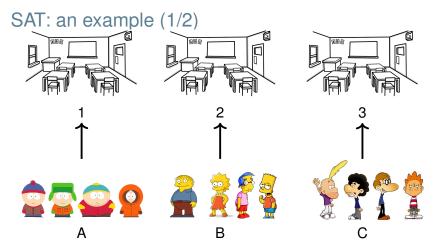




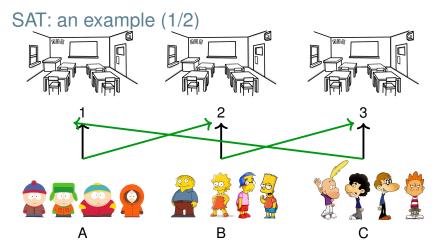
Motivation





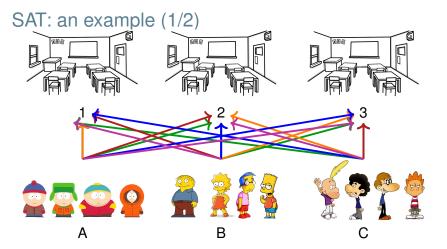


YES! SATISFIABLE
$$\alpha = \{(A, 1), (B, 2), (C, 3)\}$$



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

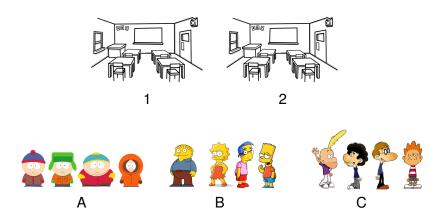


YES! SATisfiable
$$\alpha = \{(A, 1), (B, 2), (C, 3)\}$$

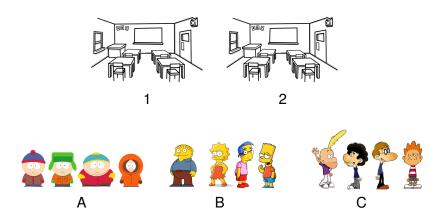
Many solutions $\alpha' = \{(A, 2), (B, 3), (C, 1)\}$
:

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SAT: an example (2/2)



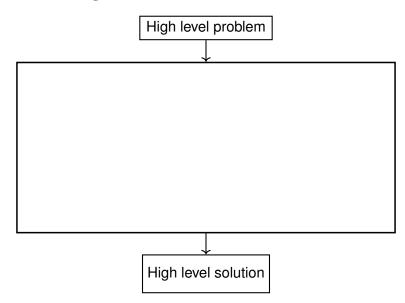
SAT: an example (2/2)



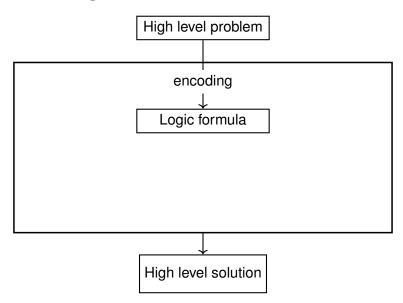
Is it possible to attribute each group to a unique classroom?

No! UNSATisfiable

From high level problem to the solution through SAT solving



From high level problem to the solution through SAT solving



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

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



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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

⁻ Clause

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

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

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

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

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

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

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

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

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

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

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

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

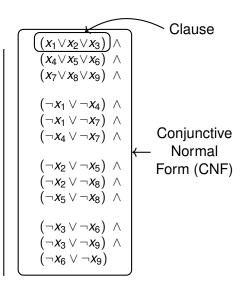
$$\neg (A, 2) \neg (C, 2)$$

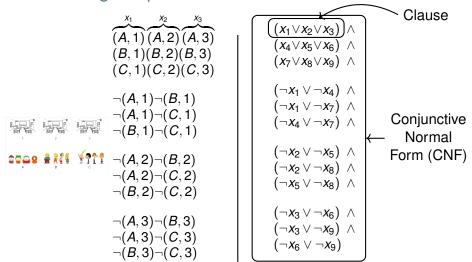
$$\neg (A, 3) \neg (C, 2)$$

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

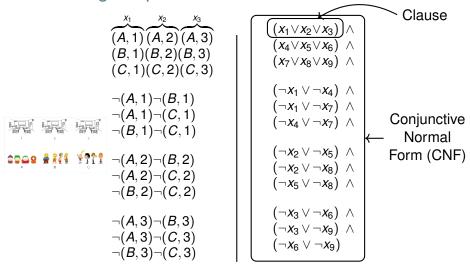
$$\neg (A, 3) \neg (C, 3)$$

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





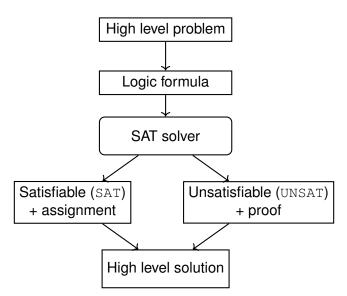
Any Boolean formula can be transformed into CNF in polynomial time



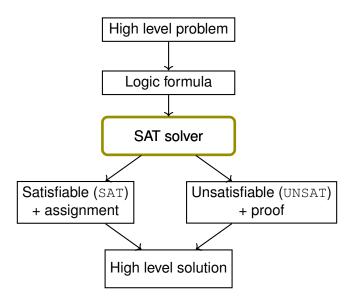
Clause represented as a set:

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

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]

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Good performance in practice:

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

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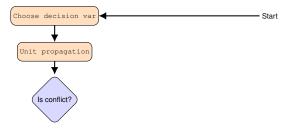
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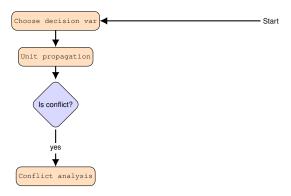
Enumerative algorithms:

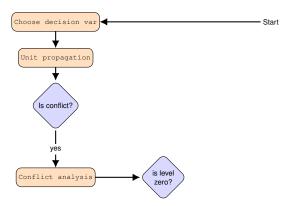
- Davis, Putnam, Logemann, and Loveland (DPLL) [DLL62]
 - Boolean Constraint Propagation (BCP)
- Conflict Driven Clause Learning (CDCL) [MSS99]
 - Derived from DPLL
 - Clause learning

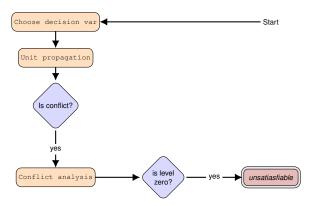


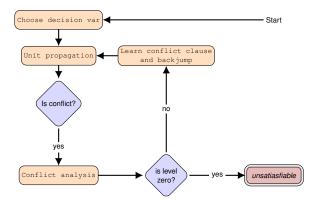


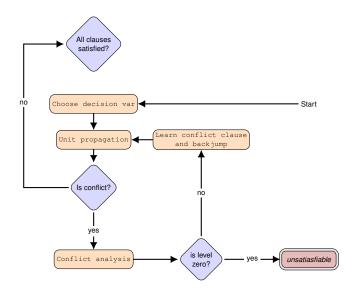


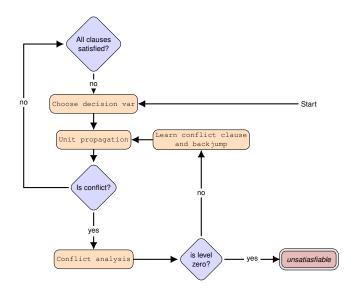


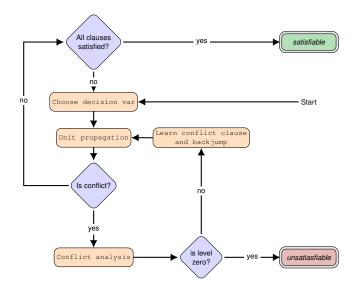


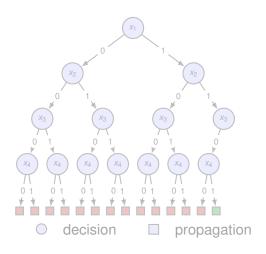












$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\}$$

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

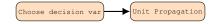
$$\omega_{3} = \{x_{1}, x_{4}\}$$

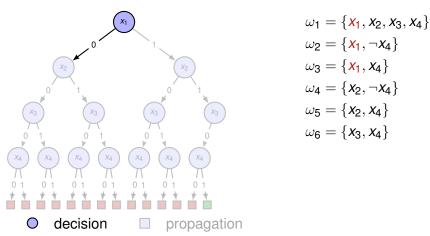
$$\omega_{4} = \{x_{2}, \neg x_{4}\}$$

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

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

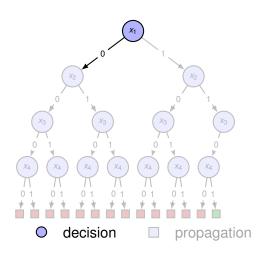
$$\alpha = \{\}$$



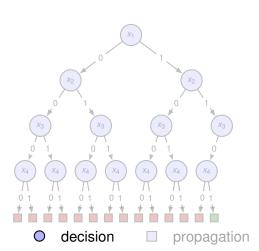


$$\omega_{2} = \{x_{1}, \neg x_{4}\}
\omega_{3} = \{x_{1}, x_{4}\}
\omega_{4} = \{x_{2}, \neg x_{4}\}
\omega_{5} = \{x_{2}, x_{4}\}
\omega_{6} = \{x_{3}, x_{4}\}$$





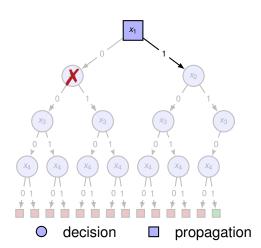
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Learn conflict clause and backjump

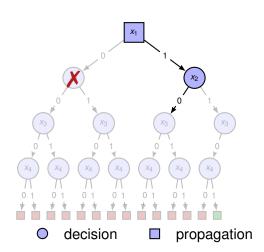
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\omega_{6} = \{x_{3}, x_{4}\}
\omega_{7} = \{x_{1}\}$$





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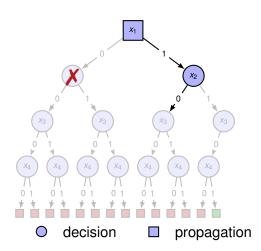
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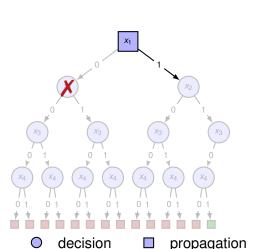
 $\alpha = \{x_1, \neg x_2\}$





$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\}
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\omega_{3} = \{x_{1}, x_{4}\}
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 $\alpha = \{x_1, \neg x_2\}$

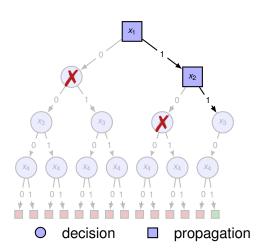


$$\alpha = \{x_1\}$$

Learn conflict clause and backjump

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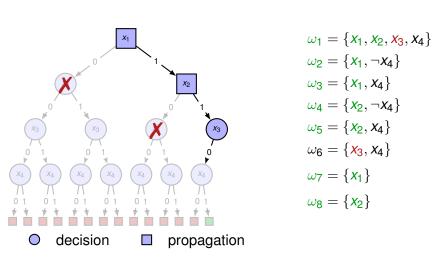
$$\omega_{6} = \{x_{3}, x_{4}\}$$

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

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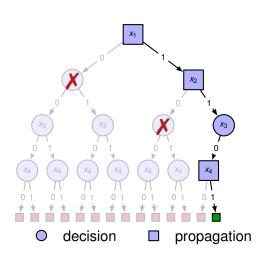
 $\alpha = \{x_1, x_2\}$





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

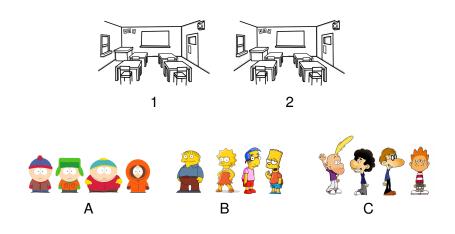


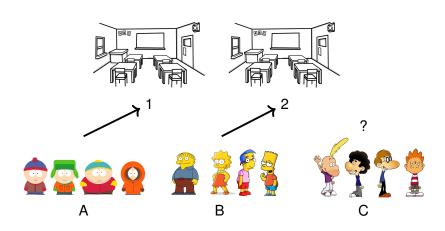


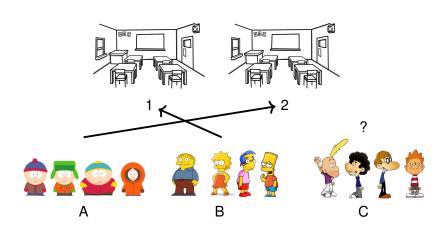
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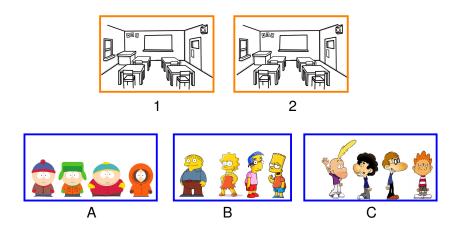
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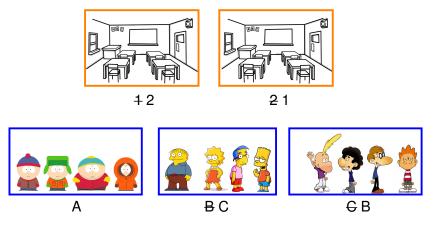
SAT and symmetries





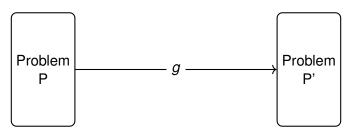






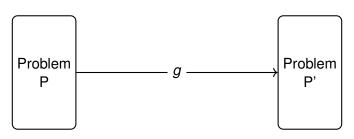
Symmetry in high level





Symmetry in high level

g a symmetry

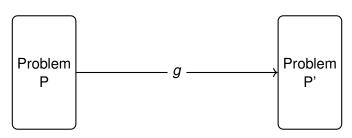


Equi-satisfiability

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

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

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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

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

 $\omega_{11} = \{\neg x_3, \neg x_9\} \qquad \qquad \omega_{12} = \{\neg x_6, \neg x_6\} \qquad \qquad \omega_{12} = \{\neg x_6, \neg x_6\} \qquad \qquad \omega_{12} = \{\neg x_6, \neg x_6\} \qquad \qquad \omega_{13} = \{\neg x_6, \neg x_6\} \qquad \qquad \omega_{14} = \{\neg x_6, \neg x_6\} \qquad \qquad \omega_{15} = \{\neg x_6, \neg x_6\} \qquad \qquad$

H

 $\omega_{10} = \{ \neg x_3, \neg x_6 \} \leftarrow$

$$g.P = P' = P$$

 $\rightarrow \omega_{10} = \{\neg x_3, \neg x_6\}$

Computing symmetries of a SAT problem

CNF formula

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

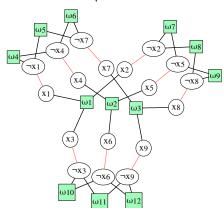
Computing symmetries of a SAT problem $(x_1 \lor x_2 \lor x_3) \land (x_4 \lor x_5 \lor x_6) \land (x_7 \lor x_8 \lor x_9)$

CNF formula

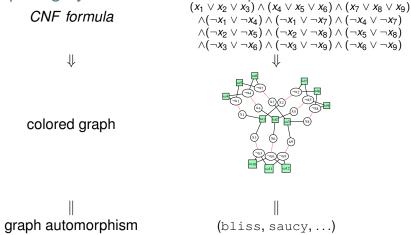


colored graph





Computing symmetries of a SAT problem



Computing symmetries of a SAT problem

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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:

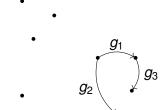
full assignment

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full assignment

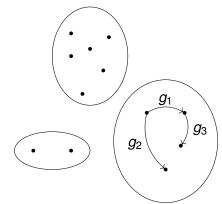


Orbit of an assignment α for a group G:

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

Example:

- full assignment
- orbit



Equivalence relation with respect to SAT:

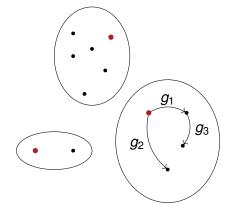
- Either $G.\alpha$ contains no solution
- Or all elements of $G.\alpha$ are solutions

Orbit of an assignment α for a group G:

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

Example:

- full assignment
- orbit
- representative



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- 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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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 Symmetry breaking predicates (SBP)

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

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

Assignments:

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

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Assignments:

Comparing: $g.\alpha \prec \alpha$

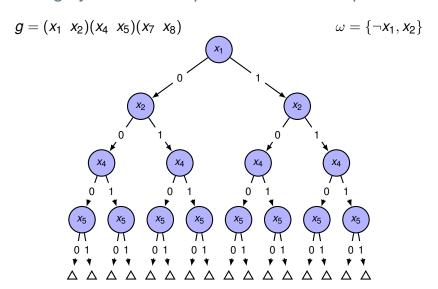
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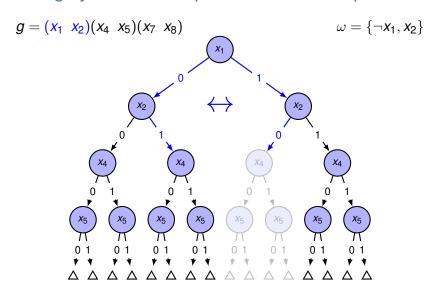
Assignments:

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

Using symmetries to prune the search space



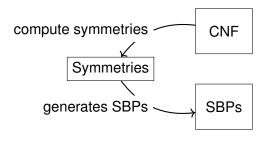
Using symmetries to prune the search space



State-of-the-art:

Shatter [ASM06]

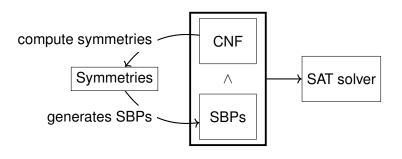
BreakID [DBBD16]



State-of-the-art:

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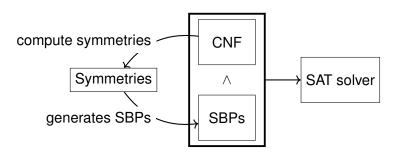
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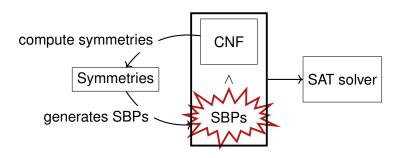


Works well on many symmetrical problems

State-of-the-art:

Shatter [ASM06]

BreakID [DBBD16]



Works well on many symmetrical problems

The solver can "explode" instead of being helped

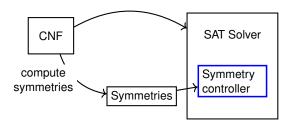
- generate not needed clause
- flooding the solver

First contribution:

CDCL[sym] Introducing Effective Symmetry Breaking in SAT Solving

TACAS'18 [MBCK18]

General idea

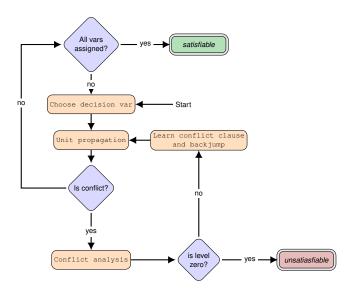


Symmetry controller:

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

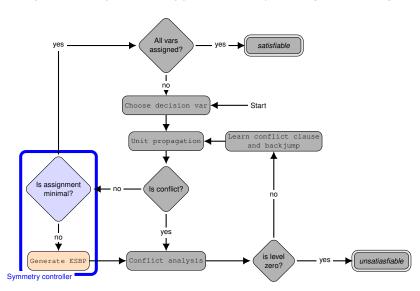
CDCL[Sym]

Compute and inject SBP opportunistically, during the solving



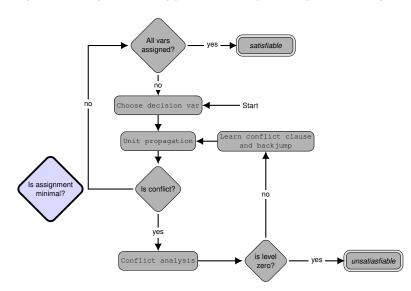
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CDCL[Sym]

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Is assignment minimal?

Our proposal: Symmetry status

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

Is assignment minimal?

Our proposal: Symmetry status

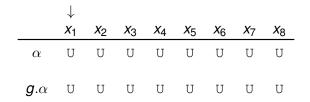
- reducer: $g.\alpha \prec \alpha$
- inactive: $\alpha \prec \mathbf{g}.\alpha$
- active: not enough information

Efficient implementation of symmetry status tracking

Keep track the smallest unassigned variable *x*:

- 1 $\alpha(g.x) \le \alpha(x)$, 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

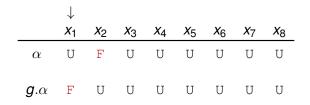
$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8; F < T$$
 $g = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$



 $g.\alpha$ α

status of permutation g: active

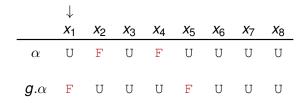
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$$g = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$$

$$\mathbf{g}.\alpha \prec \alpha$$

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 integrated easily

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

Experiments

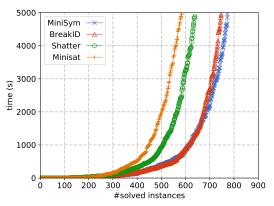
Benchmark:

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

Setup:

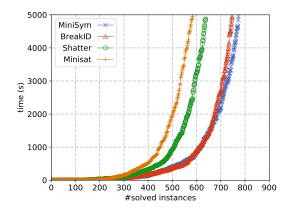
- four tools
 - MiniSat (no symmetry, baseline)
 - MiniSat + BreakID (SOTA SAT solver using symmetries)
 - MiniSat + Shatter (SOTA SAT solver using symmetries)
 - **MiniSym** = MiniSat + CDCL[Sym] (our approach)
- 5000 seconds timeout, 8GB memory
- includes time to compute symmetries (except for MiniSat)

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

recul

Exploitation of symmetries:

Dynamic symmetry breaking

Learn symmetrical clauses

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

```
\omega_8
                      \omega_5
                                                   \omega_1
                                                                   \omega_2
\omega10
                                                                 \omega11
 \omega_4
                  \omegaз
```

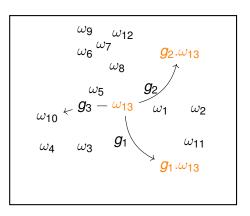
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 $\begin{array}{ll} \mathbf{u} & \text{formula} \\ \omega & \text{clause} \\ \hline \omega & \text{learnt clause} \end{array}$

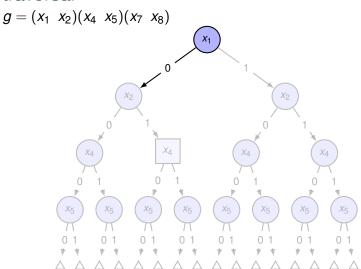
```
\omega12
                              \omega_8
                     \omega_5
                               \omega13
                                                \omega_1
\omega10
                                                             \omega11
 \omega_4
                \omegaз
```

Learn symmetrical clauses

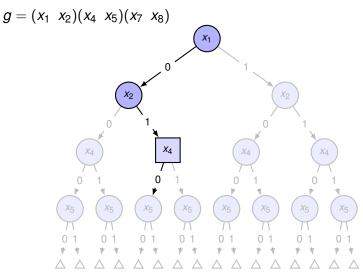
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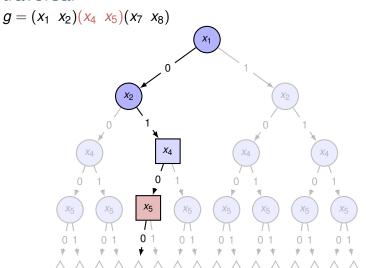
Using symmetries to accelerate the tree traversal



Using symmetries to accelerate the tree traversal

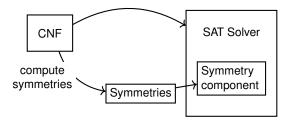


Using symmetries to accelerate the tree traversal



Use symmetries to deduce symmetrical facts.

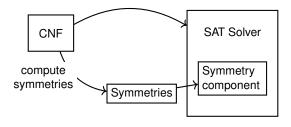
Dynamic Symmetry Breaking



State-of-the-art:

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

Dynamic Symmetry Breaking



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- Symmetry Explanation Learning (SEL) [DBB17]

Observation:

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

Local symmetry

formula

 ω clause

 ω learnt clause

Local symmetries:

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

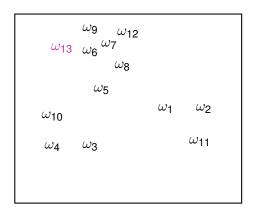
Local symmetry

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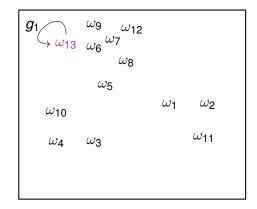


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Local symmetry

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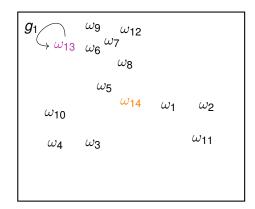
Local symmetries:

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

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

Local symmetry

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



Local symmetries:

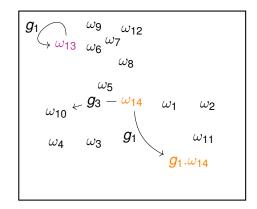
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Inductive construction

Local symmetry

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



Local symmetries:

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

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Inductive construction

Experimental results

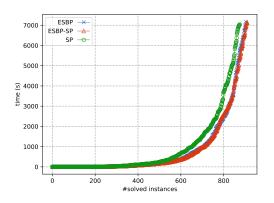
Benchmark:

- 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 CDCL[Sym])
 - 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

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)
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- Exploitation of partial symmetries
- Symmetries and parallel SAT solver

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Thanks!



Fadi A. Aloul, Karem A. Sakallah, and Igor L. Markov. Efficient symmetry breaking for boolean satisfiability. *IEEE Trans. Computers*, 55(5):549–558, 2006.



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Ashish Sabharwal

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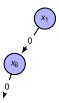
In AAAI, volume 5, pages 467-474, 2005.



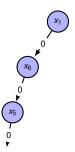
$$\omega_{1} = \{x_{1}, x_{2}, x_{3}\}
\omega_{2} = \{x_{4}, x_{5}, x_{6}\}
\omega_{3} = \{\neg x_{1}, \neg x_{5}\}
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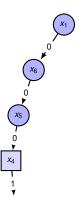
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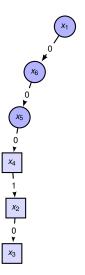
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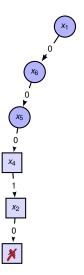
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$$\omega_7 = \{x_1, \neg x_4\}$$

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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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

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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 \cdots lex-leader $	SBP

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

	X ₁ X ₂	X ₃ X ₄	X ₅ · · ·	lex-leader	SBP
<i>O</i> ₁	FT	- -	-	· ✓	

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

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

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

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

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

	<i>x</i> ₁	<i>x</i> ₂	<i>X</i> ₃	<i>x</i> ₄	<i>x</i> ₅		lex-leader	SBP
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
 - Define total order on variables
 - Define minimal value
- Forbid non minimal assignment for each orbit

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

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

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

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

- 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

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

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

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