## Exploitation of dynamic symmetries for solving SAT problems

Doctorat de Sorbonne Université

#### Hakan Metin

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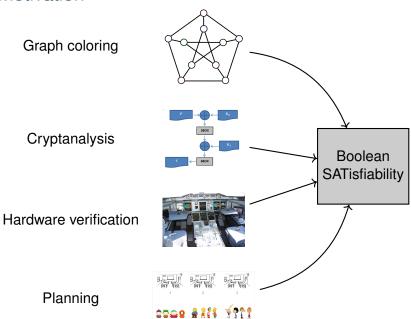


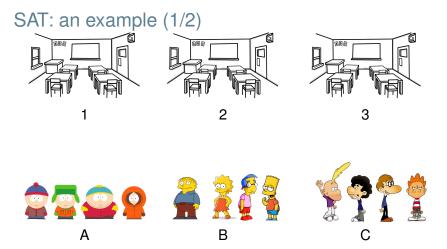


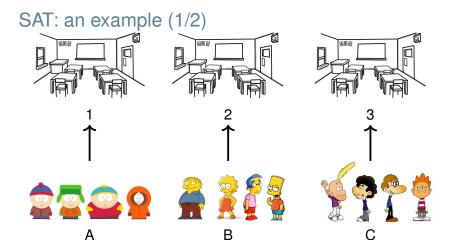




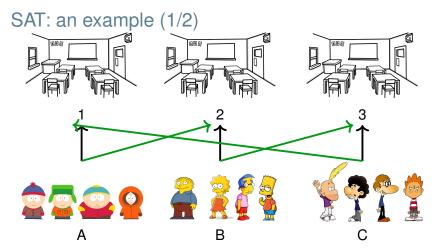
#### Motivation



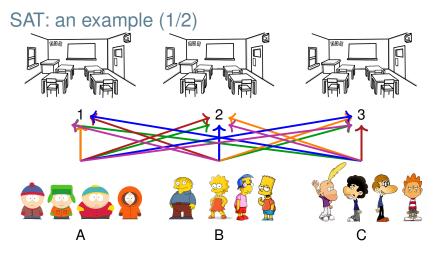




YES! SATisfiable 
$$\alpha = (A, 1), (B, 2), (C, 3)$$

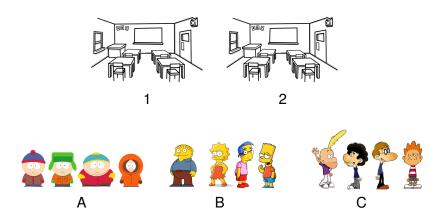


YES! SATisfiable 
$$\alpha = (A, 1), (B, 2), (C, 3)$$
  
Many solutions  $\alpha' = (A, 2), (B, 3), (C, 1)$ 

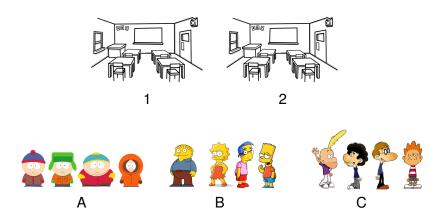


YES! SATisfiable 
$$\alpha = (A, 1), (B, 2), (C, 3)$$
  
Many solutions  $\alpha' = (A, 2), (B, 3), (C, 1)$   
 $\alpha'' = \cdots$ 

## SAT: an example (2/2)



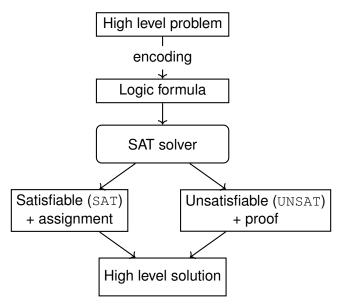
## SAT: an example (2/2)



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

No! UNSATisfiable

# From high level problem to the solution through SAT solving



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

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



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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

⁻ Clause

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

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

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

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

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

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

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

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

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

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

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

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

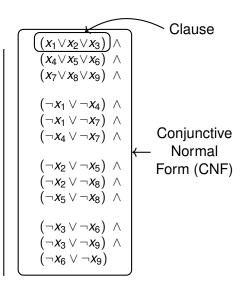
$$\neg (A, 2) \neg (C, 2)$$

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

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

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

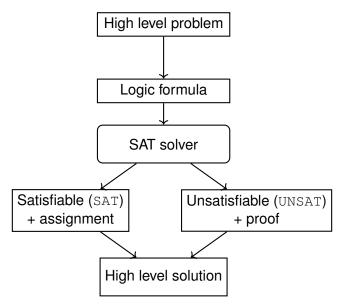
$$\neg (B, 3) \neg (C, 3)$$



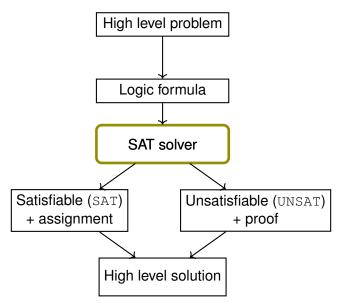
Clause represented as a set:

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

# From high level problem to the solution through SAT solving



# From high level problem to the solution through SAT solving



## **SAT Solving**

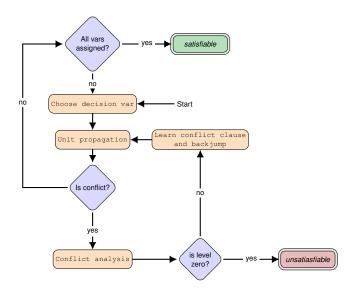
Solving SAT formula is known to be **NP-complete** [Coo71]

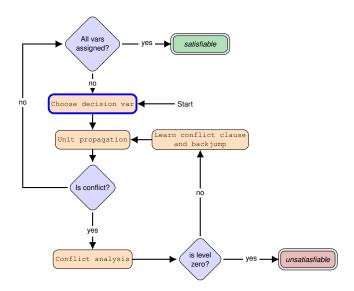
#### Good performance in practice:

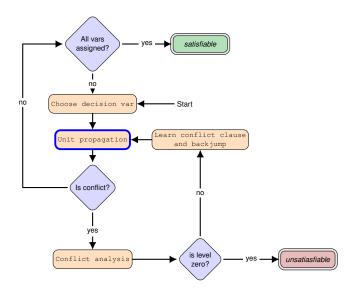
- Handle large problem (million variables and clauses)
- International SAT competition each year on academic and industrial problems

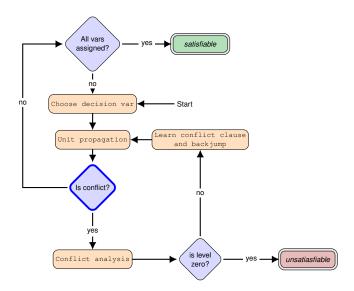
#### Enumerative algorithms:

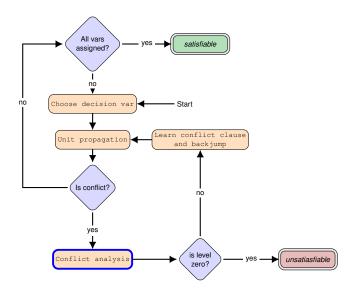
- Davis, Putnam, Logemann, and Loveland (DPLL) [DLL62]
  - Boolean Constraint Propagation (BCP)
- Conflict Driven Clause Learning (CDCL) [MSS99]
  - Derived from DPLL
  - Clause learning

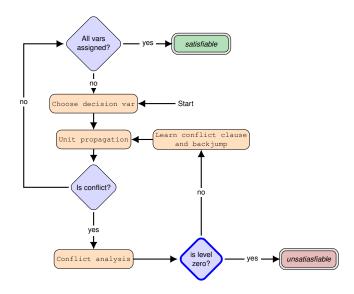


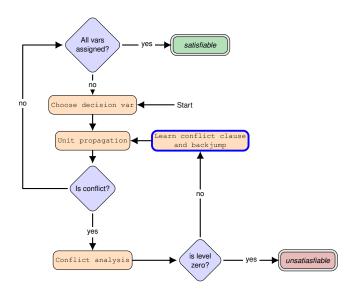


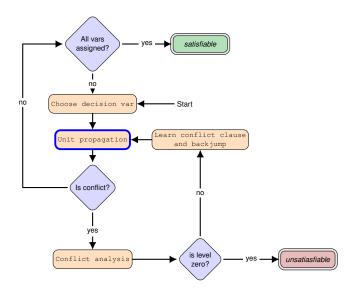


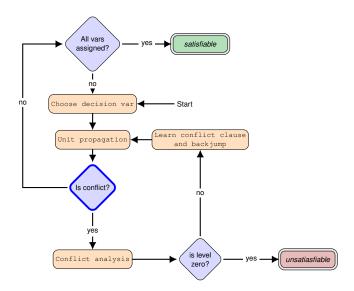


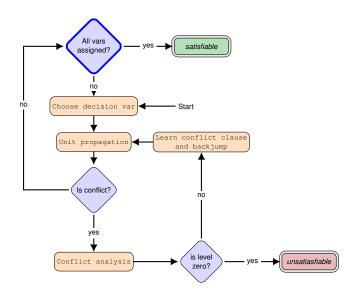


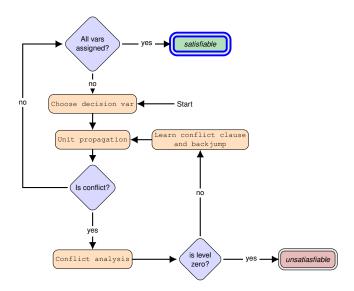


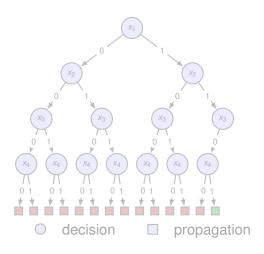












$$\omega_{1} = \{X_{1}, X_{2}, X_{3}, X_{4}\}$$

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

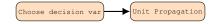
$$\omega_{3} = \{X_{1}, X_{4}\}$$

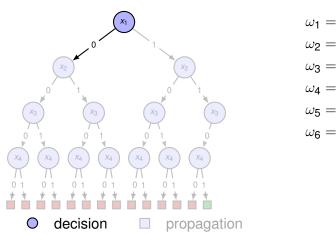
$$\omega_{4} = \{X_{2}, \neg X_{4}\}$$

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

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

$$\alpha = \{\}$$

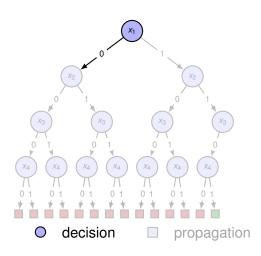




$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\} 
\omega_{2} = \{x_{1}, \neg x_{4}\} 
\omega_{3} = \{x_{1}, x_{4}\} 
\omega_{4} = \{x_{2}, \neg x_{4}\} 
\omega_{5} = \{x_{2}, x_{4}\} 
\omega_{6} = \{x_{3}, x_{4}\}$$

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





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

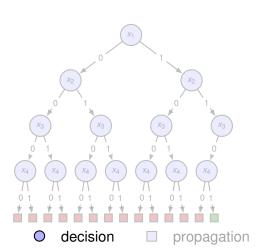
$$\omega_{2} = \{x_{1}, \neg x_{4}\}$$

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

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

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

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

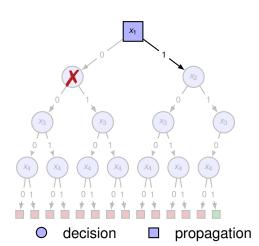


$$\alpha = \{\}$$

Learn conflict clause and backjump

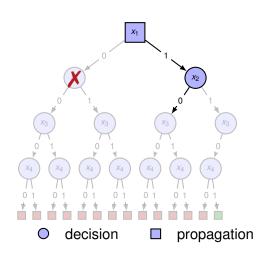
$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\} 
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\omega_{4} = \{x_{2}, \neg x_{4}\} 
\omega_{5} = \{x_{2}, x_{4}\} 
\omega_{6} = \{x_{3}, x_{4}\} 
\omega_{7} = \{x_{1}\}$$





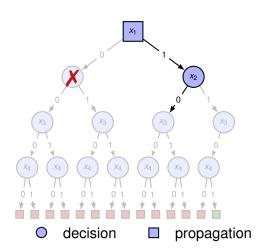
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\omega_{4} = \{x_{2}, \neg x_{4}\} 
\omega_{5} = \{x_{2}, x_{4}\} 
\omega_{6} = \{x_{3}, x_{4}\} 
\omega_{7} = \{x_{1}\}$$

 $\alpha = \{x_1\}$ 



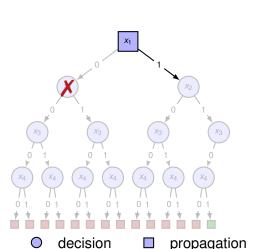
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\omega_{7} = \{x_{1}\}$$





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

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

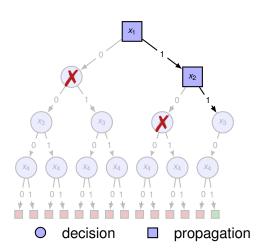


$$\alpha = \{x_1\}$$

Learn conflict clause and backjump

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





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

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

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

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

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

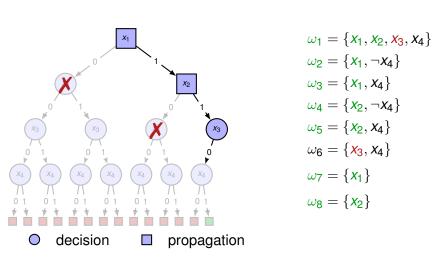
$$\omega_{6} = \{x_{3}, x_{4}\}$$

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

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

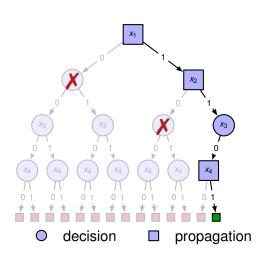
 $\alpha = \{x_1, x_2\}$ 





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

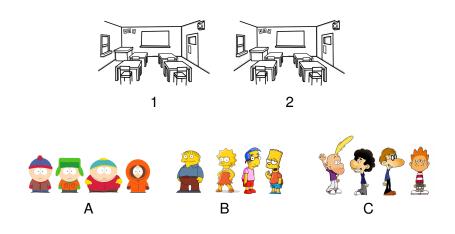


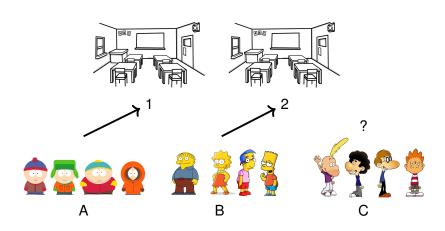


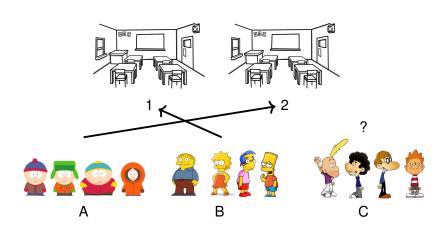
$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\} 
\omega_{2} = \{x_{1}, \neg x_{4}\} 
\omega_{3} = \{x_{1}, x_{4}\} 
\omega_{4} = \{x_{2}, \neg x_{4}\} 
\omega_{5} = \{x_{2}, x_{4}\} 
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\omega_{7} = \{x_{1}\} 
\omega_{8} = \{x_{2}\}$$

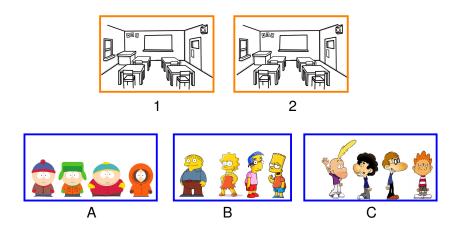
 $\alpha = \{x_1, x_2, \neg x_3, x_4\}$ 

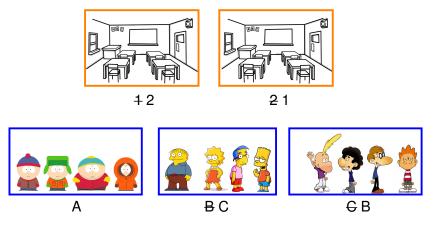
# SAT and symmetries





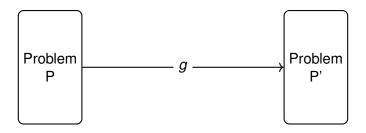






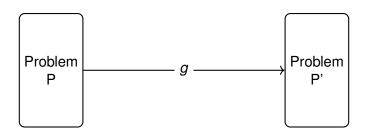
# Symmetry in high level

#### Let g a symmetry



## Symmetry in high level

Let g a symmetry



## Equi-satisfiability

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

# Syntactic symmetry

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

## Syntactic symmetry

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

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

$$\begin{matrix} \omega_1 = \{x_1, x_2, x_3\} \\ \omega_2 = \{x_4, x_5, x_6\} \\ \omega_3 = \{x_7, x_8, x_9\} \\ \omega_4 = \{-x_1, -x_4\} \\ \omega_5 = \{-x_1, -x_7\} \\ \omega_6 = \{-x_4, -x_7\} \\ \omega_6 = \{-x_2, -x_6\} \\ \omega_7 = \{-x_2, -x_6\} \\ \omega_9 = \{-x_5, -x_8\} \\ \omega_{10} = \{-x_3, -x_6\} \\ \omega_{11} = \{-x_3, -x_9\} \end{matrix}$$

$$\begin{matrix} \omega_1 = \{x_2, x_1, x_3\} \\ \omega_2 = \{x_5, x_4, x_6\} \\ \omega_3 = \{x_6, x_7, x_9\} \\ \omega_4 = \{-x_2, -x_8\} \\ \omega_6 = \{-x_2, -x_8\} \\ \omega_7 = \{-x_1, -x_7\} \\ \omega_9 = \{-x_4, -x_7\} \\ \omega_{10} = \{-x_3, -x_6\} \\ \omega_{11} = \{-x_3, -x_9\} \end{matrix}$$

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

## Computing symmetries of a SAT problem

CNF formula

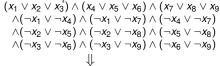
$$\begin{array}{l} (x_1 \vee x_2 \vee x_3^{}) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_7 \vee x_8 \vee x_9) \\ \wedge (\neg x_1 \vee \neg x_4) \wedge (\neg x_1 \vee \neg x_7) \wedge (\neg x_4 \vee \neg x_7) \\ \wedge (\neg x_2 \vee \neg x_5) \wedge (\neg x_2 \vee \neg x_8) \wedge (\neg x_5 \vee \neg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

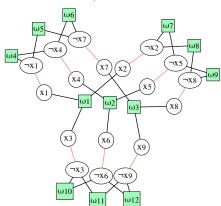
# Computing symmetries of a SAT problem $(x_1 \lor x_2 \lor x_3) \land (x_4 \lor x_5 \lor x_6) \land (x_7 \lor x_8 \lor x_9)$

CNF formula

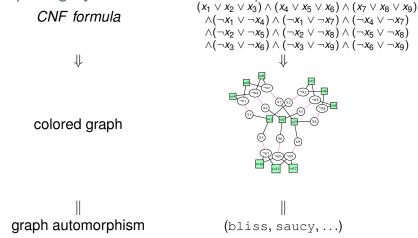


colored graph





## 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  $\alpha$  for a group G:

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

Orbit of an assignment  $\alpha$  for a group G:

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

## Example:

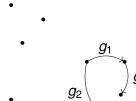
full assignment

Orbit of an assignment  $\alpha$  for a group G:

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

## Example:

full assignment

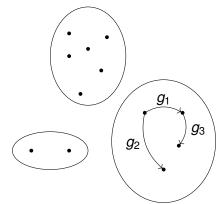


#### Orbit of an assignment $\alpha$ for a group G:

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

## Example:

- full assignment
- orbit



#### Equivalence relation with respect to SAT:

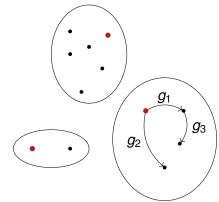
- Either  $G.\alpha$  contains no solution
- Or all elements of  $G.\alpha$  are solutions

#### Orbit of an assignment $\alpha$ 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 assignments  $(\prec)$ 

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

Allow only minimal value (lex-leader)

Forbid other assignments in each orbit

→ Add Symmetry breaking predicates (SBP)

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

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

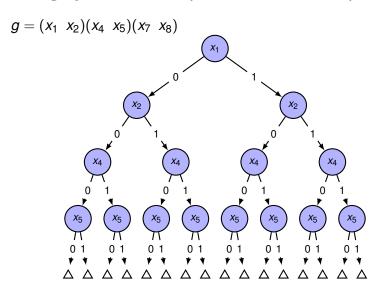
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$$g.\alpha \prec \alpha$$

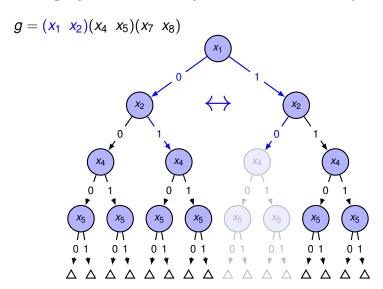
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$$g.\alpha \prec \alpha \Rightarrow \text{Generate SBP } \omega = \{\neg x_1, x_2\}$$

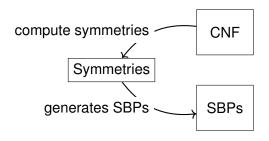
## Using symmetries to prune the search space



## Using symmetries to prune the search space



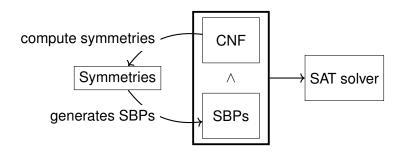
# Static symmetry breaking



State-of-the-art approaches:

Shatter [ASM06]BreakID [DBBD16]

## Static symmetry breaking

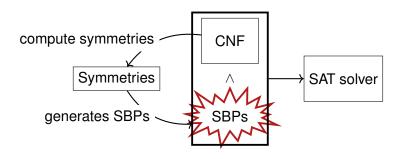


State-of-the-art approaches:

Shottor [ASM06]

Shatter [ASM06]BreakID [DBBD16]

# Static symmetry breaking



## State-of-the-art approaches:

- Shatter [ASM06]BreakID [DBBD16]

## The solver can "explode" instead of being helped

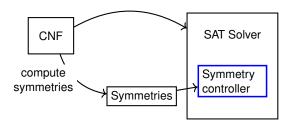
- generate not needed clause
- flooding the solver

#### First contribution

CDCL[sym] Introducing Effective Symmetry
Breaking in SAT Solving

TACAS'18 [MBCK18]

#### General idea

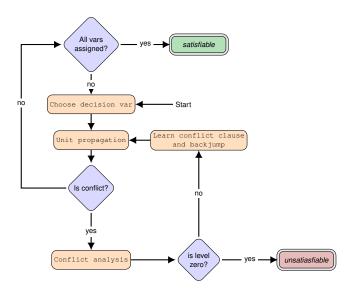


#### Symmetry controller:

- generates SBP on the fly
- only when needed
- intrusive on solver

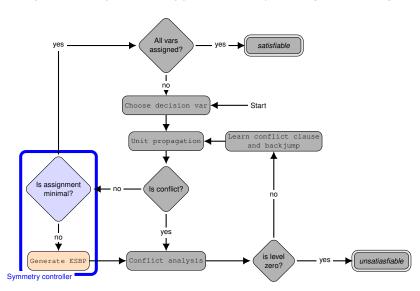
## CDCL[Sym]

Compute and inject SBP opportunistically, during the solving



## CDCL[Sym]

Compute and inject SBP opportunistically, during the solving



## Is assignment minimal? Symmetry status

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

## Is assignment minimal? Symmetry status

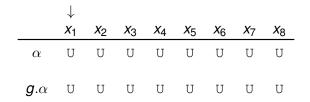
- reducer:  $g.\alpha \prec \alpha$ • inactive:  $\alpha \prec g.\alpha$
- active: not enough information

#### Efficient implementation of symmetry status

#### Keep track the smallest unassigned variable *x*:

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

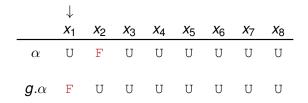
$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8; F < T$$
  $g = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$ 



 $g.\alpha$   $\alpha$ 

status of permutation g: active

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

status of permutation g: reducer

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

## CDCL[Sym] implementation

- C++ Implementation
- Packaged as a library cosy<sup>1</sup> (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 MiniSAT<sup>2</sup>.

<sup>1</sup>https://github.com/lip6/cosy

<sup>&</sup>lt;sup>2</sup>90 lines out of 3090

## Experiments

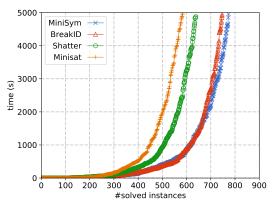
#### Benchmark:

- 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

**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
```

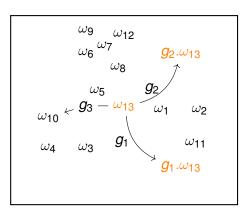
## Learn symmetrical clauses

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

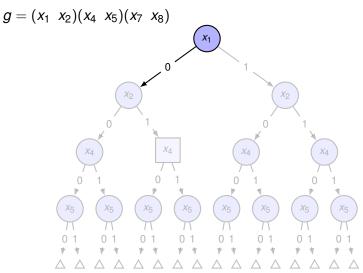
```
\omega12
                              \omega_8
                     \omega_5
                               \omega13
                                                \omega_1
\omega10
                                                             \omega11
 \omega_4
                \omegaз
```

## Learn symmetrical clauses

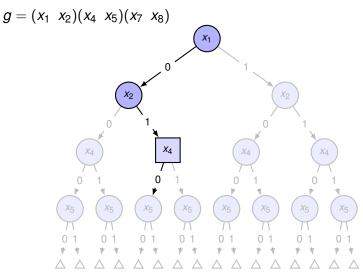
 $\begin{array}{ll} \mathbf{a} & \text{formula} \\ \omega & \text{clause} \\ \hline \omega & \text{learnt clause} \\ \end{array}$ 



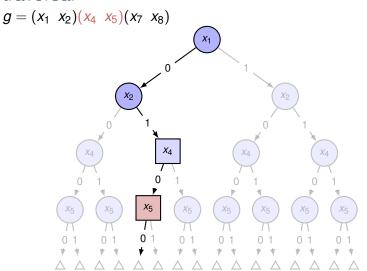
# Using symmetries to accelerate the tree traversal



# Using symmetries to accelerate the tree traversal

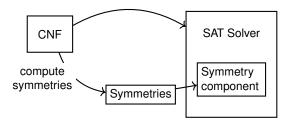


# Using symmetries to accelerate the tree traversal



Use symmetries to deduce symmetrical facts.

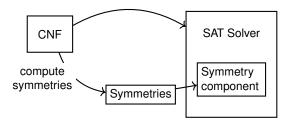
## Dynamic Symmetry Breaking



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

formula

 $\omega$  clause

 $\omega$  learnt clause

## Local symmetries:

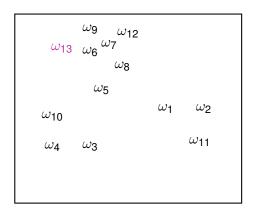
$$\omega \leftarrow \{g_1, g_2, g_3\}$$

formula

 $\omega$  clause

 $\omega$  learnt clause

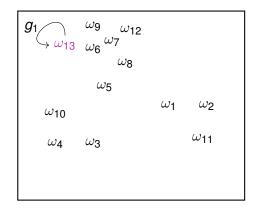
 $\omega$  ESBP



## Local symmetries:

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

- formula
- $\omega$  clause
- $\omega$  learnt clause
- $\omega$  ESBP

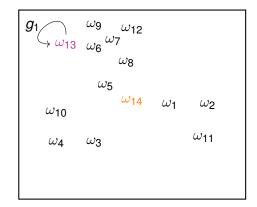


#### Local symmetries:

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

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

- formula
- $\omega$  clause
- $\omega$  learnt clause
- $\omega$  ESBP



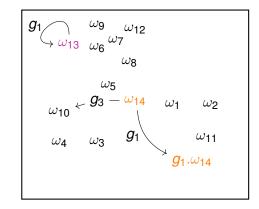
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#### Inductive construction

- formula
- $\omega$  clause
- $\omega$  learnt clause
- $\omega$  ESBP



#### Local symmetries:

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

$$\omega \leftarrow \{g_1\}$$

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

- Compute valid local symmetries
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- At minimal cost

Inductive construction

## Experimental results

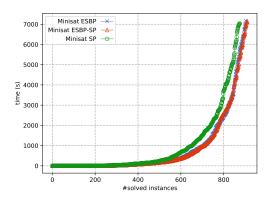
#### **Benchmark:**

- 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	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)
  - Take advantage of static and dynamic approach

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#### Perspectives

- Exploitation of partial symmetries
- Symmetries and parallel SAT solver

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#### Perspectives

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#### Thanks!



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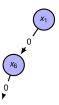
In AAAI, volume 5, pages 467-474, 2005.



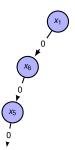
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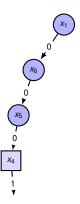
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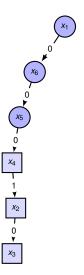
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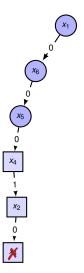
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$$\omega_7 = \{x_1, \neg x_4\}$$

# Weakly active symmetries

## Logical consequence

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

# Weakly active symmetries

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## Weakly active symmetries

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

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## Symmetry propagation

Let  $\sigma$  a weakly active symmetry, then

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

# Local symmetries

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### **Local Symmetries**

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

- $\varphi_o$  is the set of the original clauses
- $\varphi_e$  is the set of ESBPs
- φ<sub>d</sub> is the set of deduced clauses.

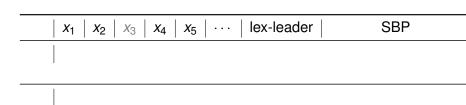
### Local symmetries

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

where  $\varphi_1$  is the set of clauses that derives  $\omega$ .

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

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



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

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

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O <sub>1</sub>	F	Т	-	-	_		✓ X	
	T	F	–	–	-	• • •	X	$\rightarrow \neg x_1 \lor x_2$
<i>O</i> <sub>2</sub>	F	F	-	F	Т		<b>✓</b>	

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

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

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

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

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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_9 $
$\neg (A, 1) \neg (B, 1)$ $\neg (A, 1) \neg (C, 1)$ $\neg (B, 1) \neg (C, 1)$	$   \begin{array}{c}     \neg X_1 \lor \neg X_4 \\     \neg X_1 \lor \neg X_7 \\     \neg X_4 \lor \neg X_7   \end{array} $
$\neg (A,2) \neg (B,2)$ $\neg (A,2) \neg (C,2)$ $\neg (B,2) \neg (C,2)$	
$\neg (A,3) \neg (B,3)$ $\neg (A,3) \neg (C,3)$ $\neg (B,3) \neg (C,3)$	$ \begin{array}{l} \neg x_3 \lor \neg x_6 \\ \neg x_3 \lor \neg x_9 \\ \neg x_6 \lor \neg x_9 \end{array} $