# Exploitation of dynamic symmetries for solving SAT problems

Doctorat de Sorbonne Université

#### Hakan Metin

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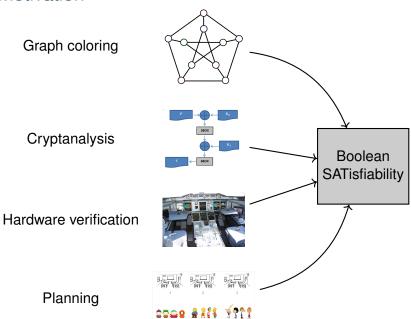


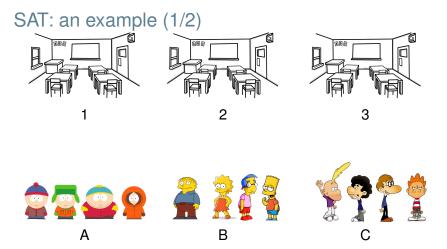


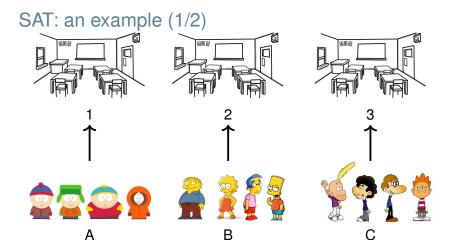




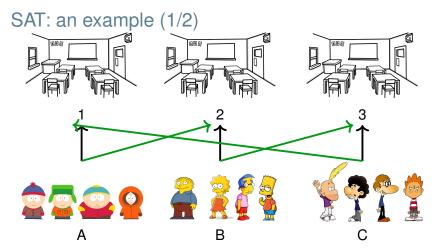
#### Motivation



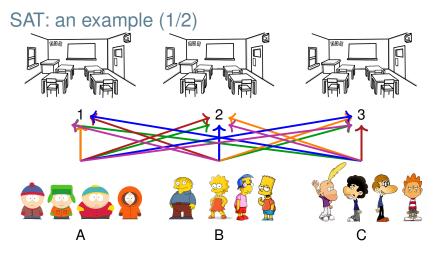




YES! SATisfiable 
$$\alpha = (A, 1), (B, 2), (C, 3)$$

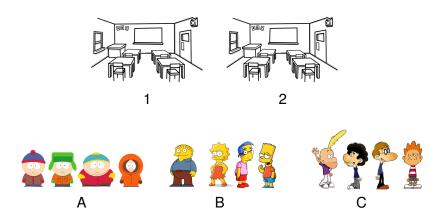


YES! SATisfiable 
$$\alpha = (A, 1), (B, 2), (C, 3)$$
  
Many solutions  $\alpha' = (A, 2), (B, 3), (C, 1)$ 

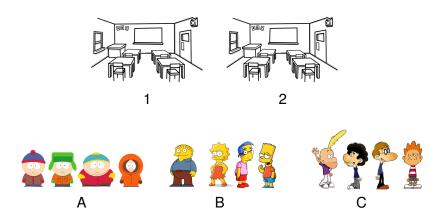


YES! SATisfiable 
$$\alpha = (A, 1), (B, 2), (C, 3)$$
  
Many solutions  $\alpha' = (A, 2), (B, 3), (C, 1)$   
 $\alpha'' = \cdots$ 

# SAT: an example (2/2)



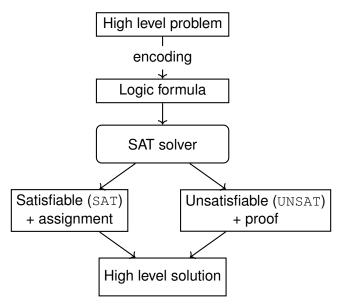
# SAT: an example (2/2)



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

No! UNSATisfiable

# From high level problem to the solution through SAT solving



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

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

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

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



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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

⁻ Clause

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

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

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

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

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

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

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

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

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

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

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

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

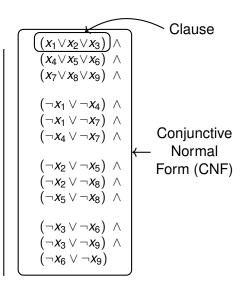
$$\neg (A, 2) \neg (C, 2)$$

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

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

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

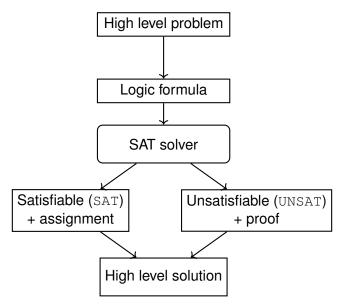
$$\neg (B, 3) \neg (C, 3)$$



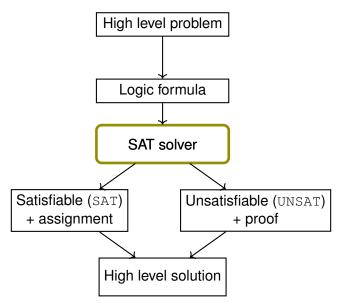
Clause represented as a set:

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

# 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]

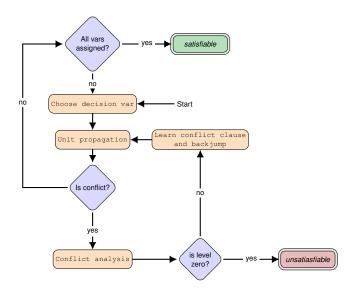
#### Good performance in practice:

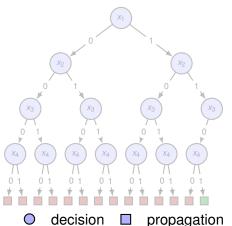
- Handle large problem (million variables and clauses)
- International SAT competition each year on academic and industrial problems

#### Enumerative algorithms:

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

## CDCL in detail





$$\omega_4 = \{x_2, \neg x_4\}$$

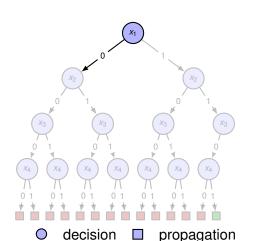
$$\omega_5 = \{x_2, x_4\}$$

$$\omega_6 = \{x_3, x_4\}$$

 $\omega_1 = \{x_1, x_2, x_3, x_4\}$ 

 $\omega_2 = \{x_1, \neg x_4\}$  $\omega_3 = \{x_1, x_4\}$ 

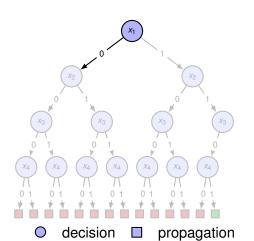
 $\alpha = \{\}$ 



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

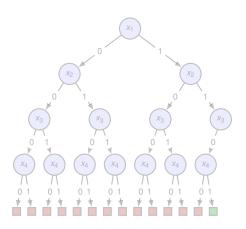
 $\alpha = \{\neg x_1\}$ 





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

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



O decision propagation

$$\alpha = \{\}$$

Learn conflict clause and backjump

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

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

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

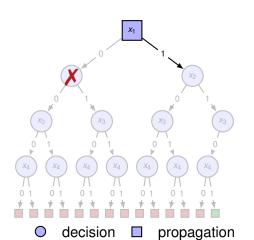
$$\omega_{4} = \{x_{2}, \neg x_{4}\}$$

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

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

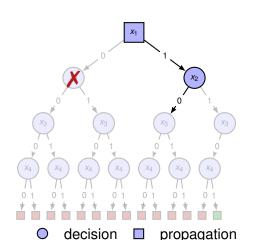
$$\omega_{7} = \{x_{1}\}$$





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

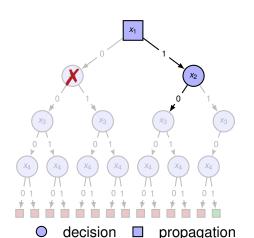
 $\alpha = \{x_1\}$ 



$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\} 
\omega_{2} = \{x_{1}, \neg x_{4}\} 
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\omega_{6} = \{x_{3}, x_{4}\} 
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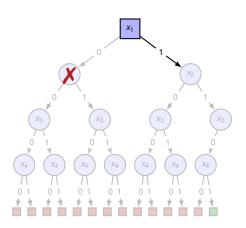
$$\alpha = \{x_1, \neg x_2\}$$





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

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



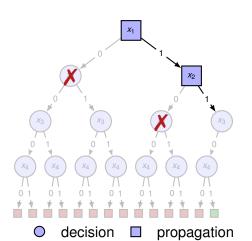
O decision propagation

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

Learn conflict clause and backjump

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

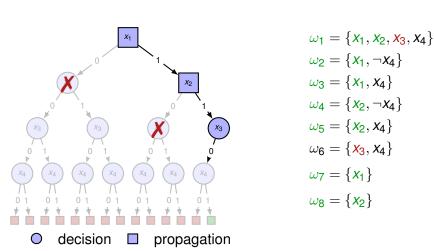




$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\} 
\omega_{2} = \{x_{1}, \neg x_{4}\} 
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\omega_{7} = \{x_{1}\} 
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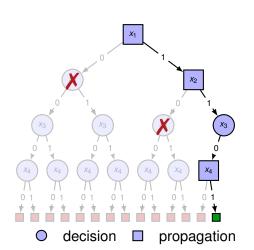
 $\alpha = \{x_1, x_2\}$ 





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

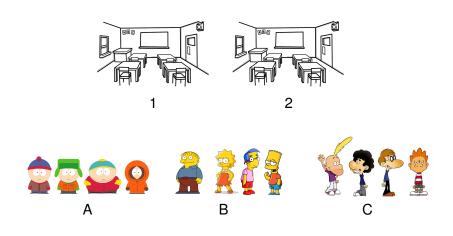


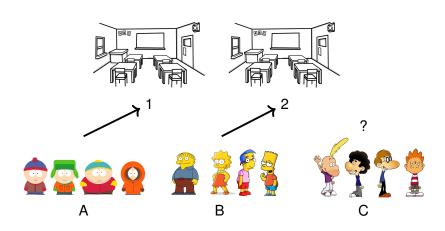


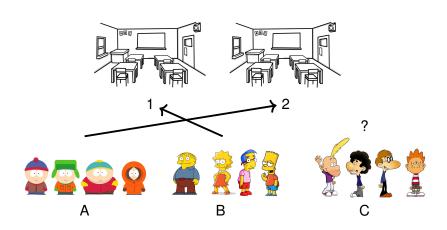
$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\} 
\omega_{2} = \{x_{1}, \neg x_{4}\} 
\omega_{3} = \{x_{1}, x_{4}\} 
\omega_{4} = \{x_{2}, \neg x_{4}\} 
\omega_{5} = \{x_{2}, x_{4}\} 
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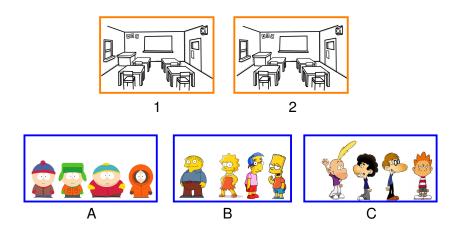
 $\alpha = \{x_1, x_2, \neg x_3, x_4\}$ 

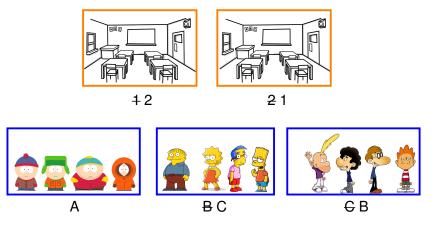
# SAT and symmetries





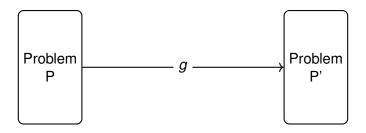






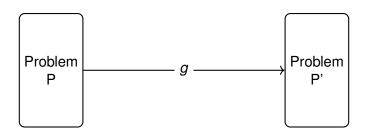
# Symmetry in high level

#### 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

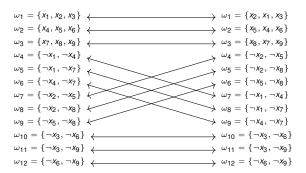
A symmetry (permuation) g is a bijective function (on variables) that leaves the formula  $\varphi$  invariant

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

# Syntactic symmetry

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# Computing symmetries of a SAT problem

CNF formula

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

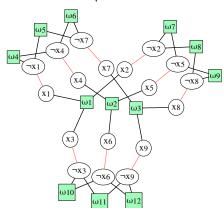
# Computing symmetries of a SAT problem $(x_1 \lor x_2 \lor x_3) \land (x_4 \lor x_5 \lor x_6) \land (x_7 \lor x_8 \lor x_9)$

CNF formula

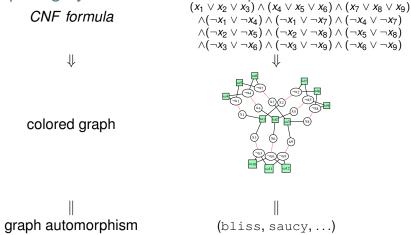


colored graph

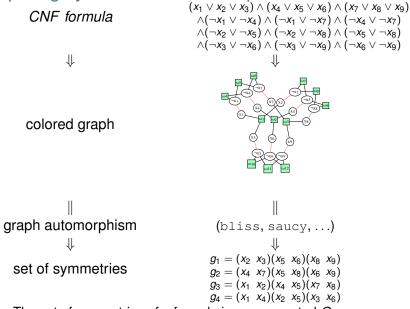




# Computing symmetries of a SAT problem



# Computing symmetries of a SAT problem



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

Exploitation of symmetries

Static symmetry breaking

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

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

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

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

### Example:

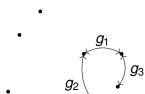
full assignment

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

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

### Example:

full assignment

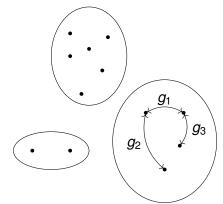


#### Orbit of an assignment $\alpha$ for a group G:

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

#### Example:

- full assignment
- orbit



#### Equivalence relation with respect to SAT:

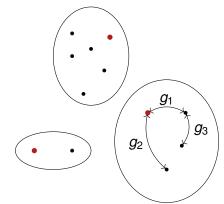
- Either  $G.\alpha$  contains no solution
- Or all elements of  $G.\alpha$  are solutions

#### 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

Allow only minimal value (lex-leader)

Forbid other assignments in each orbit

→ Add Symmetry breaking predicates (SBP)

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

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

	<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> 8
$\alpha$	Т	F	F	F	F	F	F	F
$oldsymbol{g}.lpha$	F	Т	F	F	F	F	F	F

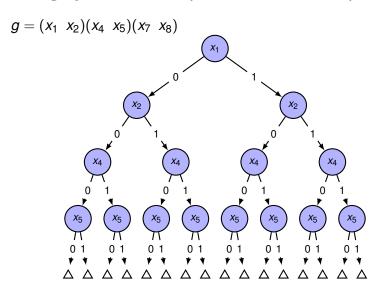
$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8; F < T$$
  $g = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$ 

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

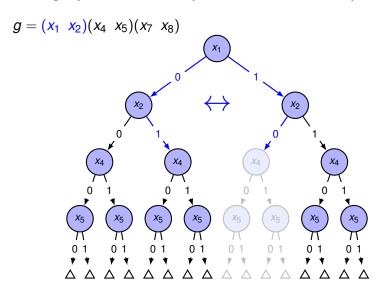
$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8; F < T$$
  $g = (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8)$ 

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

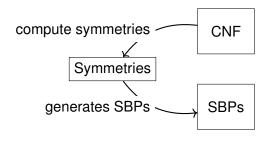
# Using symmetries to prune the search space



# Using symmetries to prune the search space



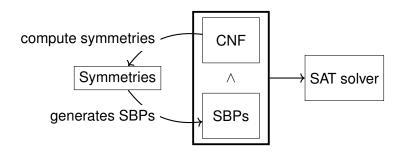
# Static symmetry breaking



State-of-the-art approaches:

Shatter [ASM06]BreakID [DBBD16]

# Static symmetry breaking

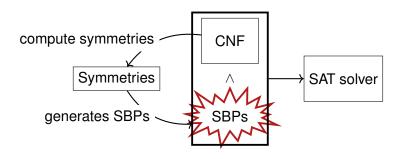


State-of-the-art approaches:

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

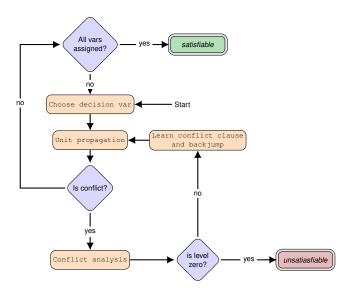
TACAS'18 [MBCK18]

### Idea

intrusive on the fly needed

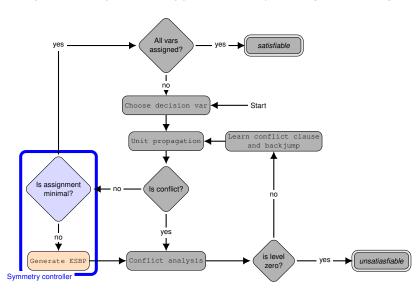
### CDCL[Sym]

Compute and inject SBP opportunistically, during the solving



### CDCL[Sym]

Compute and inject SBP opportunistically, during the solving



# Symmetry status

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

### Efficient implementation of symmetry status

Keep track the smallest unassigned variable x:

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

Update whenever variables are assigned / unassigned

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

status of permutation g: active

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status of permutation g: reducer

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

# CDCL[Sym] implementation

- C++ Implementation
- Packaged as a library 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

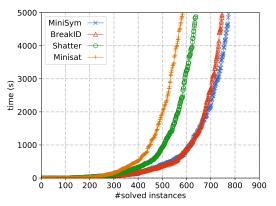
#### Benchmark:

- 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



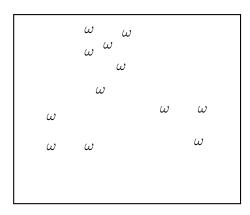
Solver	PAR-2	ALL	SAT	UNSAT
MiniSAT	2243h	586	325	261
Shatter	2088h	640	316	324
BreakID	1790h	749	334	415
MiniSym	1735h	775	336	439

**Exploitation of symmetries** 

Dynamic symmetry breaking

# Learn symmetrical clauses

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



# Learn symmetrical clauses

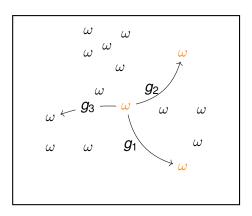
 $\begin{array}{ll} \mathbf{a} & \text{formula} \\ \omega & \text{clause} \\ \omega & \text{learnt clause} \end{array}$ 

```
\omega
             \omega
                                 \omega
                                                    \omega
```

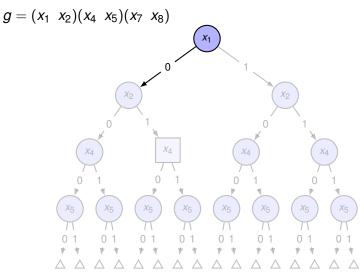
# Learn symmetrical clauses

 $_{\omega}$  formula  $_{\omega}$  clause

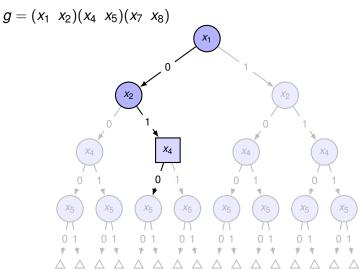
 $\omega$  learnt clause



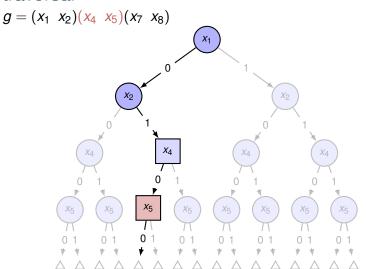
# Using symmetries to accelerate the tree traversal



# Using symmetries to accelerate the tree traversal

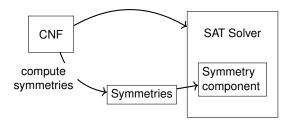


# Using symmetries to accelerate the tree traversal



Use symmetries to deduce symmetrical facts.

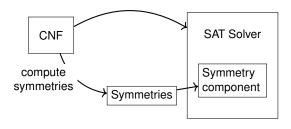
# Dynamic Symmetry Breaking



#### State-of-the-art:

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

# Dynamic Symmetry Breaking



#### State-of-the-art:

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- Symmetry Explanation Learning (SEL) [DBB17]

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

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

### 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

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

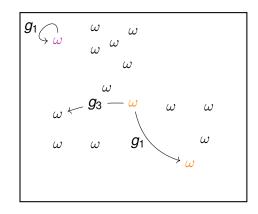
### Local symmetries:

$$\omega \leftarrow \{g_1, g_2, g_3\}$$
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- Compute valid local symmetries
- On the fly
- At minimal cost

#### 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
- On the fly
- At minimal cost

#### Inductive construction

### Experimental results

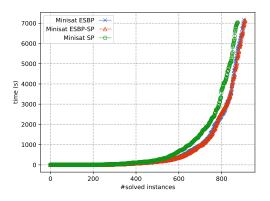
#### Benchmark:

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

### Setup:

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

# Experimental results



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

#### Conclusion

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

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### Perspectives

- Symmetries and parallel SAT solver
- Exploitation of partial symmetries

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### Perspectives

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#### Thanks!



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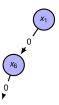
In AAAI, volume 5, pages 467-474, 2005.



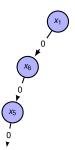
$$\omega_{1} = \{x_{1}, x_{2}, x_{3}\} 
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\omega_{3} = \{\neg x_{1}, \neg x_{5}\} 
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\omega_{6} = \{\neg x_{3}, \neg x_{6}\}$$



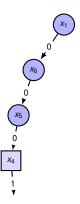
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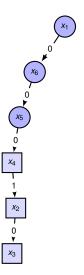
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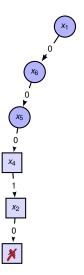
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\omega_{5} = \{\neg x_{3}, \neg x_{4}\} 
\omega_{6} = \{\neg x_{3}, \neg x_{6}\}$$

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

### 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.

## Weakly active symmetries

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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.

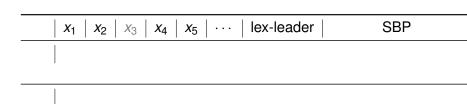
### Local symmetries

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

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

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



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	<i>x</i> <sub>1</sub>	<i>x</i> <sub>2</sub>	<i>X</i> <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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	<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
0.	F	Т	-	-	_		✓ ×	
O <sub>1</sub>	Т	F	-	–	-		X	$\rightarrow \neg x_1 \lor x_2$
	I							

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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
<i>O</i> <sub>1</sub>	F	T	_	_	_		✓ X	$\rightarrow \neg x_1 \lor x_2$

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

- Define lexicographic order
  - Define total order on variables
  - Define minimal value
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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

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

39/39

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

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

39/39

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

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

39/39

- 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   
 $g_1 = \begin{pmatrix} x_2 & x_3 \end{pmatrix} \begin{pmatrix} x_5 & x_6 \end{pmatrix} \begin{pmatrix} x_8 & x_9 \end{pmatrix} \begin{vmatrix} x = x_5 & g.x = x_6 \\ & & \text{active} \end{pmatrix}$ 
 $g_2 = \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} x_4 & x_5 \end{pmatrix} \begin{pmatrix} x_7 & x_8 \end{pmatrix} \begin{vmatrix} x = x_1 & g.x = x_2 \\ & & \text{reducer} \end{pmatrix}$ 
 $\cdots$ 
 $\alpha = \{ \neg x_2, \neg x_3, x_1 \}$ 

$$g_2$$
 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_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} $