# Exploitation of dynamic symmetries for solving SAT problems

Thèse de doctorat de Sorbonne Université

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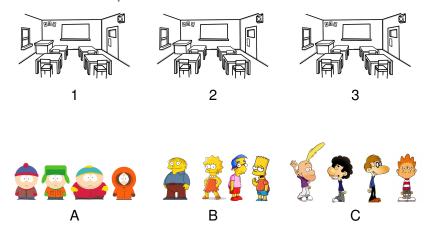


## Motivation

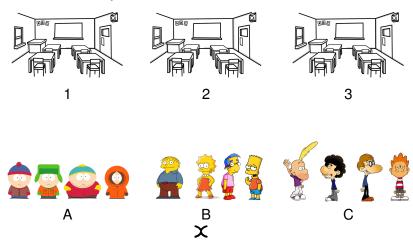
## SAT is widely used in different domains

- Artificial intelligence (planning, games, ...)
- Bioinformatics (haplotype inference, ...)
- Security (cryptanalysis, inversion attack on hash, ...)
- Computationally hard problems (graph coloring, ...)
- Formal methods (hardware model checking, ...)

# SAT an example



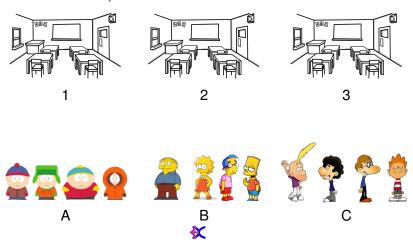
# SAT an example



Is it possible to attribute each group to a classroom?

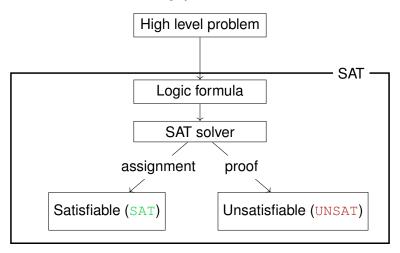
YES!

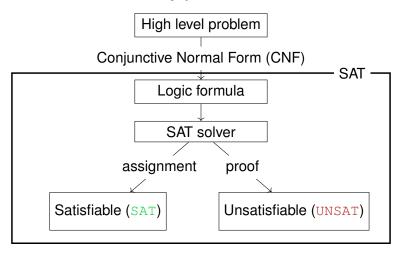
# SAT an example



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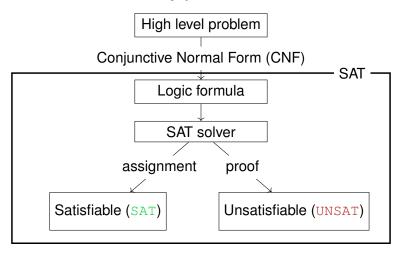
YES! Many solutions





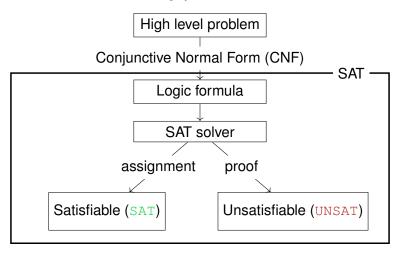
## CNF representation:

$$\underbrace{\left(x_1 \lor x_2 \lor \neg x_3\right)}_{\text{Clause with literals } x_1, x_2, \neg x_3}$$



## CNF representation:

Formula (CNF)
$$\underbrace{\left(x_1 \lor x_2 \lor \neg x_3\right)}_{Clause} \land \left(\neg x_1 \lor \neg x_2\right) \land \left(x_2 \lor \neg x_4\right)$$



Clause representation as a set:

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

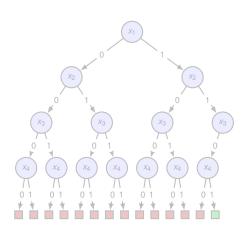
# **SAT Solving**

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

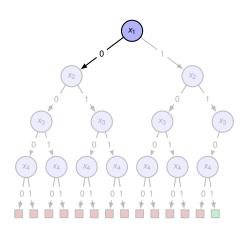
## Enumerative algorithm:

- Davis, Putnam, Logemann, and Loveland (DPLL) [?]
  - Boolean Constraint Propagation (BCP)

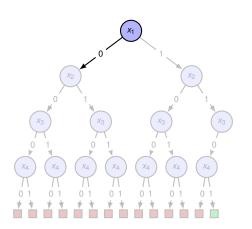
- Conflict Driven Clause Learning (CDCL) [?]
  - Derived from DPLL
  - Clause learning



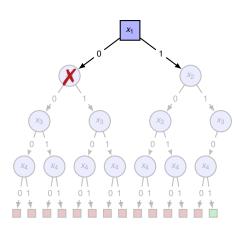
$$\omega_{1} = \{x_{1}, x_{2}, x_{3}, x_{4}\} 
\omega_{2} = \{x_{1}, \neg x_{4}\} 
\omega_{3} = \{x_{1}, x_{4}\} 
\omega_{4} = \{x_{2}, \neg x_{4}\} 
\omega_{5} = \{x_{2}, x_{4}\} 
\omega_{6} = \{x_{3}, x_{4}\}$$



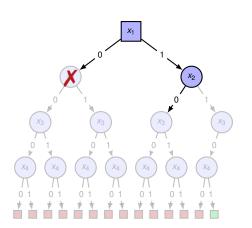
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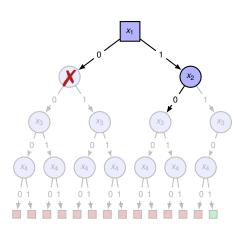
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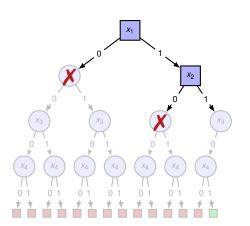
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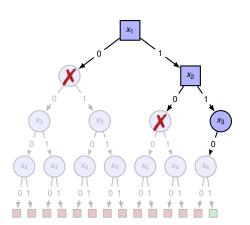
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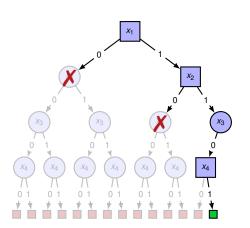
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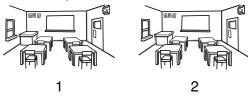
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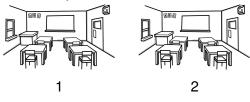
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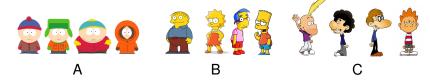


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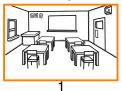


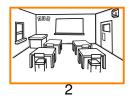


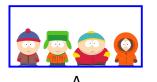


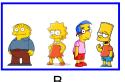
Is it possible to attribute each group to a classroom?

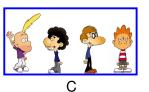
No!











Is it possible to attribute each group to a classroom?

No!

Presence of symmetries hinders the performance of the solver

## Outline

SAT overview

SAT basics SAT and symmetries

2 Existing approaches

Static symmetry breaking Dynamic symmetry breaking

3 Contribution and results

## Symmetry

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

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

$$\begin{array}{c} \omega_1 = \{x_1, x_2, x_3\} \\ \omega_2 = \{x_4, x_5, x_6\} \\ \omega_3 = \{x_7, x_8, x_9\} \\ \omega_4 = \{\neg x_1, \neg x_4\} \\ \omega_5 = \{\neg x_1, \neg x_7\} \\ \omega_6 = \{\neg x_4, \neg x_7\} \\ \omega_7 = \{\neg x_2, \neg x_5\} \\ \omega_7 = \{\neg x_2, \neg x_5\} \\ \omega_8 = \{\neg x_2, \neg x_8\} \\ \omega_9 = \{\neg x_5, \neg x_8\} \\ \omega_9 = \{\neg x_5, \neg x_8\} \\ \omega_9 = \{\neg x_5, \neg x_6\} \\ \omega_{11} = \{\neg x_3, \neg x_6\} \\ \omega_{11} = \{\neg x_3, \neg x_9\} \\ \omega_{12} = \{\neg x_6, \neg x_9\} \\ \end{array}$$

Equi-satisfiability:

$$\alpha \models \varphi \Leftrightarrow g.\alpha \models \varphi$$

# Computing symmetries of a SAT problem

CNF formula

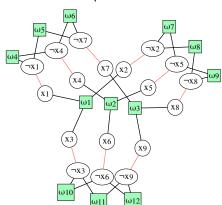
$$\begin{array}{l} (x_1 \vee x_2 \vee x_3) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_7 \vee x_8 \vee x_9) \\ \wedge (\neg x_1 \vee \neg x_4) \wedge (\neg x_1 \vee \neg x_7) \wedge (\neg x_4 \vee \neg x_7) \\ \wedge (\neg x_2 \vee \neg x_5) \wedge (\neg x_2 \vee \neg x_8) \wedge (\neg x_5 \vee \neg x_8)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

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

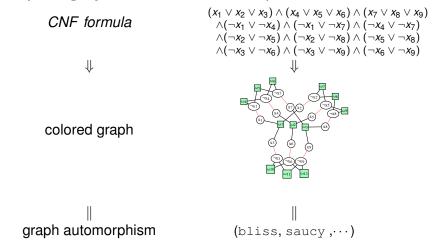
CNF formula

colored graph

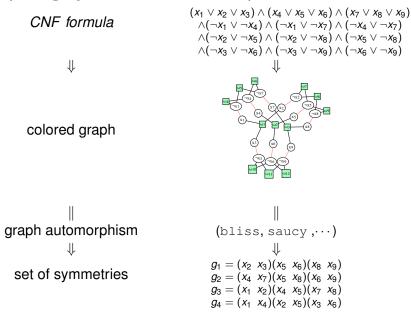




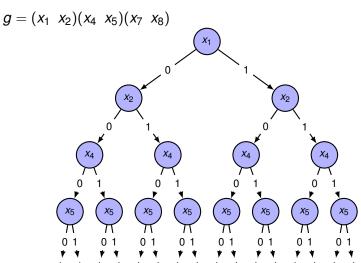
# Computing symmetries of a SAT problem



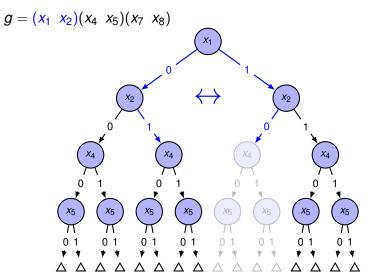
# Computing symmetries of a SAT problem



# Using symmetries to prune search space



# Using symmetries to prune search space



Adds symmetry breaking predicates to prune search space.

# Generates symmetry breaking predicates (SBP)

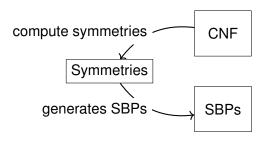
- Define lexicographic order
  - Define total order on variables
  - Define minimal value
- Forbid non minimal assignment with addition of SBP

## Example:

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

$x_1 \leq x_2$	$x_1 \vee \neg x_2$
$x_1=x_2\to x_4\le x_5$	$x_1 \vee x_2 \vee x_4 \vee \neg x_5$
	$\neg x_1 \lor \neg x_2 \lor x_4 \lor \neg x_5$
$x_1 = x_2 \wedge x_4 = x_5 \rightarrow x_8 \leq x_3$	$X_1 \vee X_2 \vee X_4 \vee X_5 \vee X_7 \vee \neg X_8$
	$\neg x_1 \vee \neg x_2 \vee x_4 \vee x_5 \vee x_7 \vee \neg x_8$
	• • •

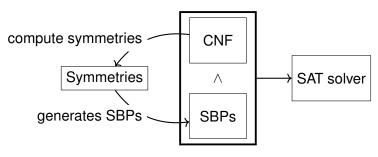
# Static symmetry breaking



## Different approaches:

- Shatter [?]
- BreakID [?]
- ..

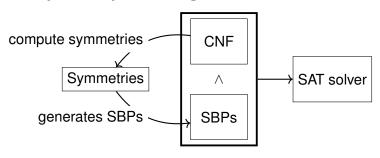
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# Static symmetry breaking



## Different approaches:

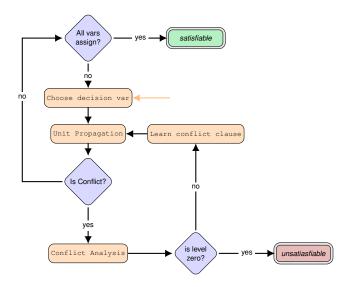
- Shatter [?]
- BreakID [?]
- ...

### Pros/Cons:

- Works well on many symmetric instances
- The solver can "explode" instead of being helped

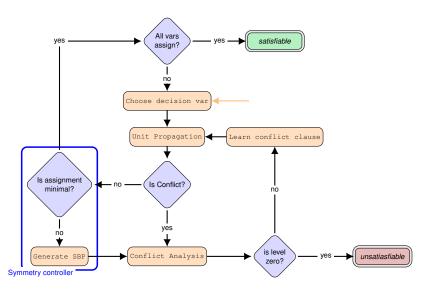
# Our contribution CDCL[Sym]

Compute and inject SBP opportunistically, during the solving



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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**  $\alpha(g.x)$  ≤  $\alpha(x)$ , 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

Update whenever variables are assigned / unassigned

- 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

$$x_1 \le x_2 \le x_3 \le x_4 \le x_5 \le x_6 \le x_7 \le x_8$$
; false < true   
 $g_1 = (x_2 \quad x_3) \quad (x_5 \quad x_6) \quad (x_8 \quad x_9) \quad x = x_2 \quad g.x = x_3$  active   
 $g_2 = (x_1 \quad x_2) \quad (x_4 \quad x_5) \quad (x_7 \quad x_8) \quad x = x_1 \quad g.x = x_2$  active   
...

 $\alpha = \{$ 

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

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

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

$$x_1 \leq x_2 \leq x_3 \leq x_4 \leq x_5 \leq x_6 \leq x_7 \leq x_8$$
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## CDCL[Sym] Implementation

 Packaged as a library cosy<sup>1</sup>, to be combined with your solver

$$ightarrow$$
 e.g. +3% LOC on MiniSAT.

- Follows symmetry status
- Should work with any enumerative SAT solver

https://github.com/lip6/cosy

## Experiments

#### Benchmark:

- from SAT contests 2012 2017
- retain only instances for which bliss finds significant symmetries in 1000s
- 1350 symmetric instances (out of 3700)

#### 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)
- 5000s timeout, 8GB memory
- includes time to compute symmetries (except for MiniSat)

# Experimental results

#### bliss gives more generators than saucy3

Figure: Cactus plot total number of instances

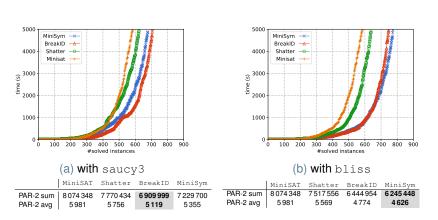


Figure: Time comparison

# Experimental results (UNSAT versus SAT)

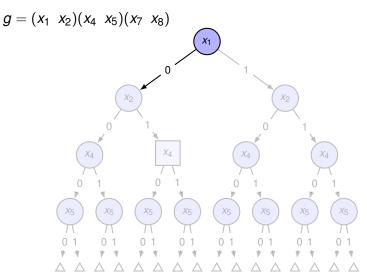
	MiniSAT	Shatter	BreakID	MiniSym		MiniSAT	Shatter	BreakID	MiniSym
TOTAL (no dup)	261	302	371	345	TOTAL (no dup)	261	324	415	439
(a) With saucy3					(b) With bliss				

Figure: Comparison on UNSAT instances

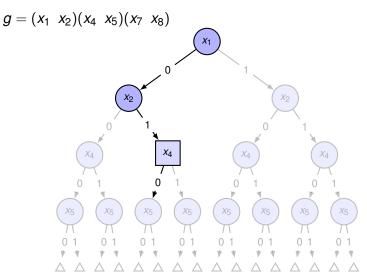


Figure: Comparison on SAT instances

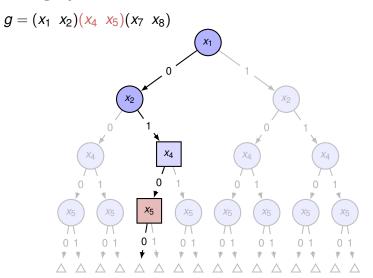
# Using symmetries to accelerate tree traversal



# Using symmetries to accelerate tree traversal

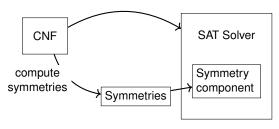


# Using symmetries to accelerate tree traversal



Use symmetries to deduce symmetrical facts.

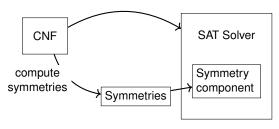
# Dynamic Symmetry Breaking



#### Different approaches:

- Symmchaff [?]
- Symmetry Propagation (SP) [?]
- Symmetry Learning Scheme (SLS) [?]
- Symmetry Explanation Leaning (SEL) [?]
- ...

# Dynamic Symmetry Breaking



#### Different approaches:

- Symmchaff [?]
- Symmetry Propagation (SP) [?]
- Symmetry Learning Scheme (SLS) [?]
- Symmetry Explanation Leaning (SEL) [?]
- ...

#### Pros/Cons:

- Works well on many symmetric instances
- Cannot handle some instances solved by static approach

#### 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 the time

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

 $\omega_1 \leftarrow \text{(Local symmetries)}$   $\omega_2 \leftarrow \text{(Local symmetries)}$   $\omega_3 \leftarrow \text{(Local symmetries)}$   $\omega_4 \leftarrow \text{(Local symmetries)}$ 

Macro level  $\rightarrow$  Micro level

 $\omega_1 \leftarrow \text{(Local symmetries)}$   $\omega_2 \leftarrow \text{(Local symmetries)}$   $\omega_3 \leftarrow \text{(Local symmetries)}$   $\omega_4 \leftarrow \text{(Local symmetries)}$   $\omega_5$   $\omega_5$ 

 $\omega_1 \leftarrow \text{(Local symmetries)}$   $\omega_2 \leftarrow \text{(Local symmetries)}$   $\omega_3 \leftarrow \text{(Local symmetries)}$   $\omega_4 \leftarrow \text{(Local symmetries)}$   $\omega_5 \leftarrow \text{(Local symmetries)}$   $\omega_6 \leftarrow \text{(Local symmetries)}$ 

Compute valid local symmetries on-the-fly at a minimal cost.

## Experimental results

#### Benchmark:

- from SAT contests 2012 2018
- retain only instances for which bliss finds significant symmetries in 1000s
- 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)
- 7200s timeout

#### 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
  - Overcomes drawbacks of the existing approaches

- A new hybrid approach (ESBP-SP)
  - Take advantage of static and dynamic approach
  - Introduce local symmetries

## Perspectives

 Combination of CDCL[Sym] with other dynamic symmetry breaking approach

Exploitation of partial symmetries

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 Combination of CDCL[Sym] with other dynamic symmetry breaking approach

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Thanks!



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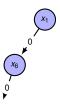
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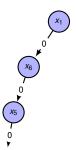
$$\omega_{1} = \{x_{1}, x_{2}, x_{3}\} 
\omega_{2} = \{x_{4}, x_{5}, x_{6}\} 
\omega_{3} = \{\neg x_{1}, \neg x_{5}\} 
\omega_{4} = \{\neg x_{2}, \neg x_{4}\} 
\omega_{5} = \{\neg x_{3}, \neg x_{4}\} 
\omega_{6} = \{\neg x_{3}, \neg x_{6}\}$$



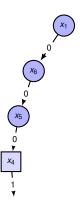
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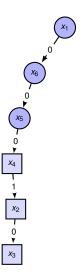
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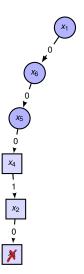
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\omega_{6} = \{\neg x_{3}, \neg x_{6}\}$$

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

## 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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### Symmetry propagation

Let  $\sigma$  a weakly active symmetry, then

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

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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  $\omega$ .