

# Exploitation of dynamic symmetries for solving SAT problems

Doctorat de Sorbonne Université

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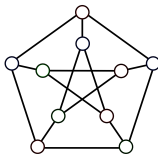
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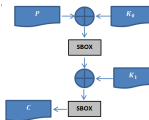


# Motivation

Graph coloring



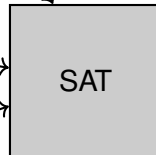
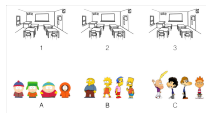
Cryptanalysis



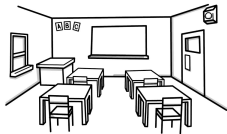
Hardware model  
checking



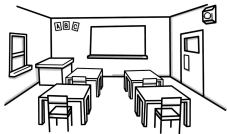
Planning



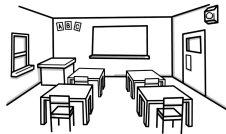
## SAT: an example (1/2)



1



2



3



A



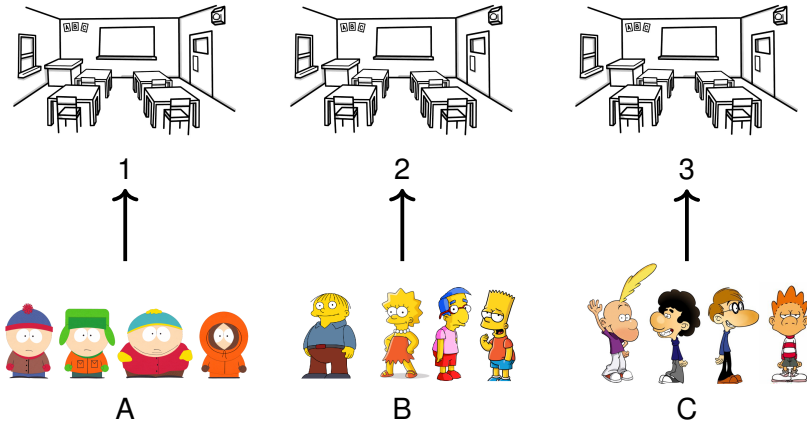
B



C

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

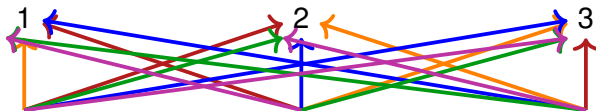
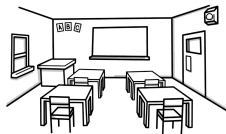
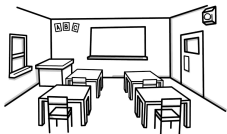
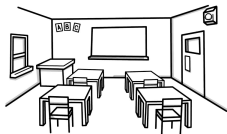
# SAT: an example (1/2)



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

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

# SAT: an example (1/2)



A



B



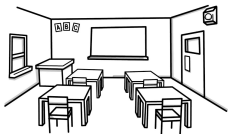
C

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

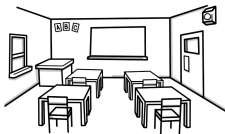
YES! SAT!  $\alpha = (A, 1), (B, 2), (C, 3)$

Many solutions  $\alpha = (A, 2), (B, 3), (C, 1); \dots$

## SAT: an example (2/2)



1



2



A



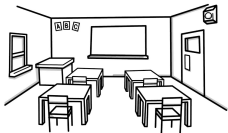
B



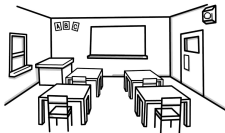
C

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

## SAT: an example (2/2)



1



2



A



B



C

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

No! UNSAT

# Encoding the problem

$$\begin{array}{ll} (A, 1)(A, 2)(A, 3) & (x_1 \vee x_2 \vee x_3) \wedge \\ (B, 1)(B, 2)(B, 3) & (x_4 \vee x_5 \vee x_6) \wedge \\ (C, 1)(C, 2)(C, 3) & (x_7 \vee x_8 \vee x_9) \wedge \end{array}$$

$$\begin{array}{ll} \neg(A, 1)\neg(B, 1) & (\neg x_1 \vee \neg x_4) \wedge \\ \neg(A, 1)\neg(C, 1) & (\neg x_1 \vee \neg x_7) \wedge \\ \neg(B, 1)\neg(C, 1) & (\neg x_4 \vee \neg x_7) \wedge \end{array}$$

$$\begin{array}{ll} \neg(A, 2)\neg(B, 2) & (\neg x_2 \vee \neg x_5) \wedge \\ \neg(A, 2)\neg(C, 2) & (\neg x_2 \vee \neg x_8) \wedge \\ \neg(B, 2)\neg(C, 2) & (\neg x_5 \vee \neg x_8) \wedge \end{array}$$

$$\begin{array}{ll} \neg(A, 3)\neg(B, 3) & (\neg x_3 \vee \neg x_6) \wedge \\ \neg(A, 3)\neg(C, 3) & (\neg x_3 \vee \neg x_9) \wedge \\ \neg(B, 3)\neg(C, 3) & (\neg x_6 \vee \neg x_9) \wedge \end{array}$$




# Encoding the problem

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

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

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

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

  
 $(x_1 \vee x_2 \vee x_3) \wedge$   
 $(x_4 \vee x_5 \vee x_6) \wedge$   
 $(x_7 \vee x_8 \vee x_9) \wedge$   
 $(\neg x_1 \vee \neg x_4) \wedge$   
 $(\neg x_1 \vee \neg x_7) \wedge$   
 $(\neg x_4 \vee \neg x_7) \wedge$   
 $(\neg x_2 \vee \neg x_5) \wedge$   
 $(\neg x_2 \vee \neg x_8) \wedge$   
 $(\neg x_5 \vee \neg x_8) \wedge$   
 $(\neg x_3 \vee \neg x_6) \wedge$   
 $(\neg x_3 \vee \neg x_9) \wedge$   
 $(\neg x_6 \vee \neg x_9) \wedge$

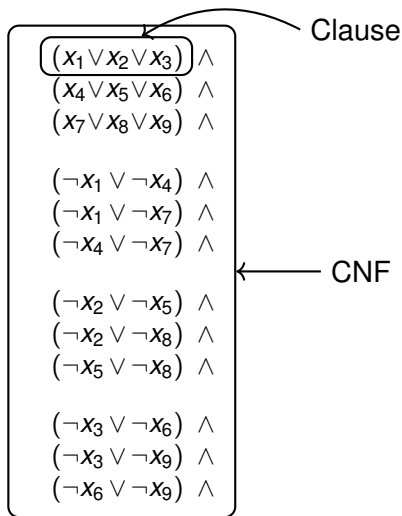
## Encoding the problem

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

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

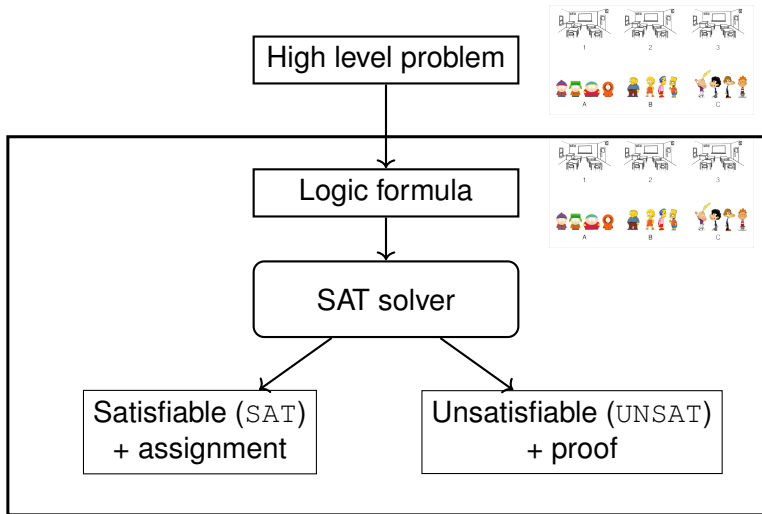
$\neg(A, 2)\neg(B, 2)$   
 $\neg(A, 2)\neg(C, 2)$   
 $\neg(B, 2)\neg(C, 2)$

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

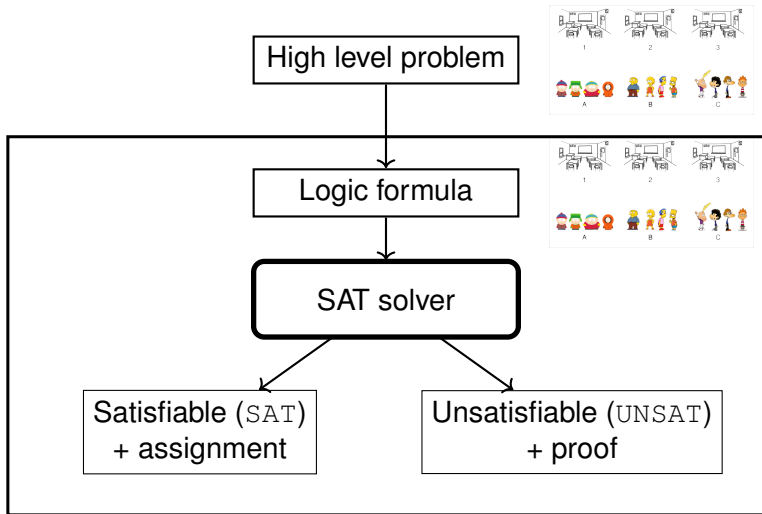


Conjunctive Normal Form (CNF)

# SAT



# SAT



# SAT Solving

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

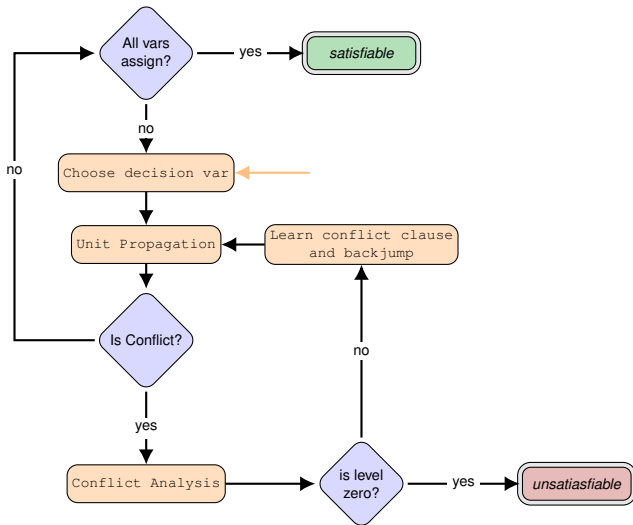
Good performance in practice:

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

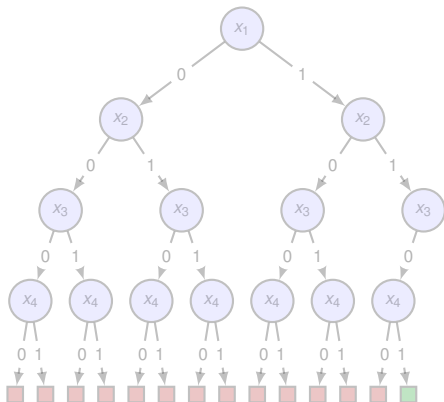
Enumerative algorithms:

- 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



# CDCL in action



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

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

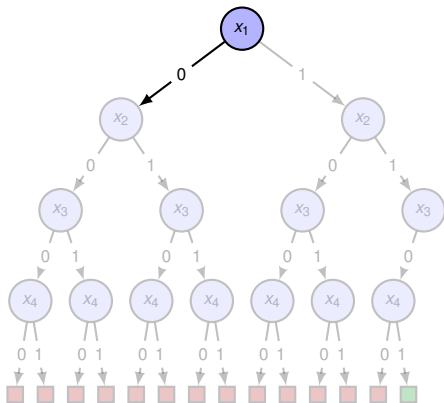
$$\omega_3 = \{x_1, x_4\}$$

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

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

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

# CDCL in action



$$\omega_1 = \{\mathbf{x}_1, x_2, x_3, x_4\}$$

$$\omega_2 = \{\mathbf{x}_1, \neg x_4\}$$

$$\omega_3 = \{\mathbf{x}_1, x_4\}$$

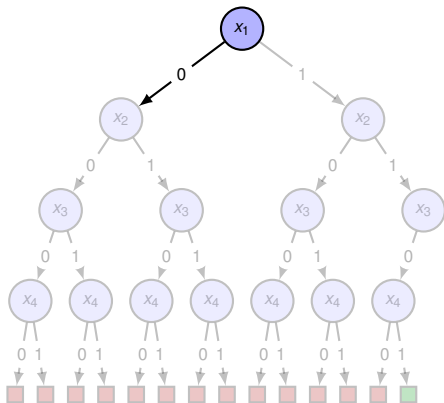
$$\omega_4 = \{x_2, \neg x_4\}$$

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

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



# CDCL in action



$$\omega_1 = \{\mathbf{x}_1, x_2, x_3, x_4\}$$

$$\omega_2 = \{\mathbf{x}_1, \neg \mathbf{x}_4\}$$

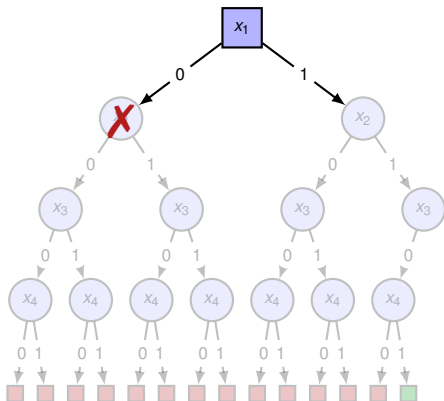
$$\omega_3 = \{\mathbf{x}_1, \mathbf{x}_4\}$$

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

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

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

# CDCL in action



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

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

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

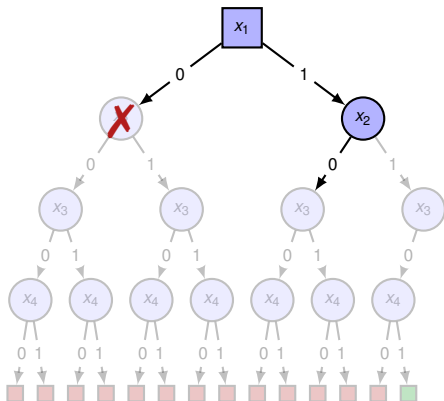
$$\omega_4 = \{x_2, \neg x_4\}$$

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

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

$$\omega_7 = \{x_1\}$$

# CDCL in action



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

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

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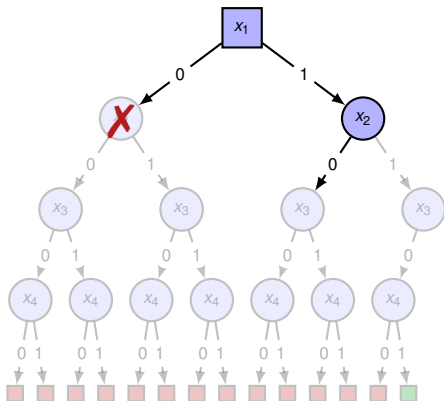
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$$\omega_7 = \{x_1\}$$

# CDCL in action



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

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

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

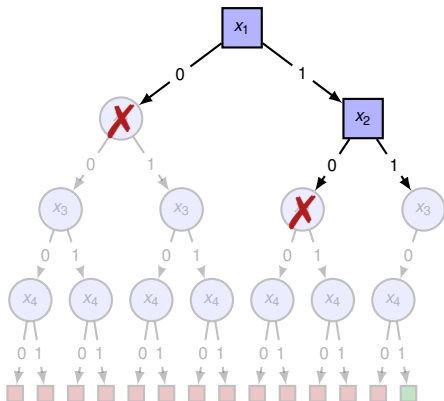
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# CDCL in action



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

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

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

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

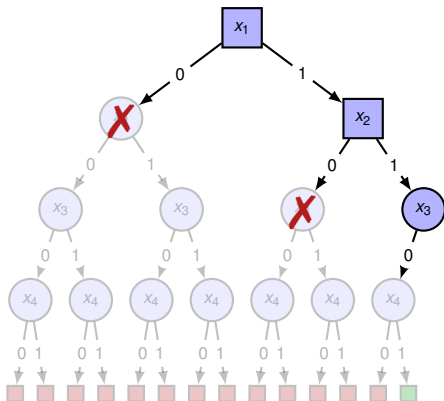
$$\omega_5 = \{x_2, x_4\}$$

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

$$\omega_7 = \{x_1\}$$

$$\omega_8 = \{x_2\}$$

# CDCL in action



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

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

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

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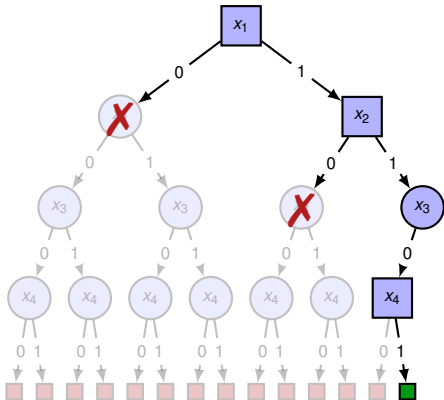
$$\omega_5 = \{x_2, x_4\}$$

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

$$\omega_7 = \{x_1\}$$

$$\omega_8 = \{x_2\}$$

## CDCL in action



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

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

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

$$\omega_4 = \{X_2, \neg X_4\}$$

$$\omega_5 = \{X_2, X_4\}$$

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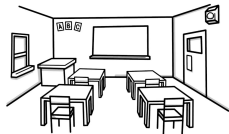
$$\omega_7 = \{x_1\}$$

$$\omega_8 = \{X_2\}$$

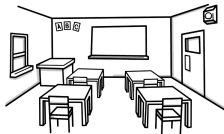
# SAT and symmetries



# Presence of symmetries



1



2



A

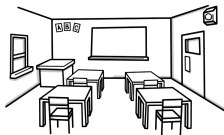
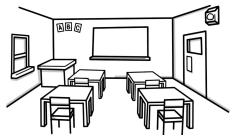


B



C

# Presence of symmetries



A

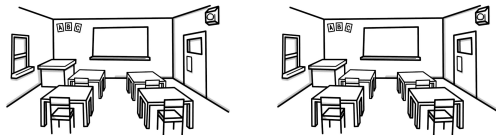


B



C

# Presence of symmetries



1 ← → 2



A

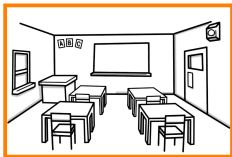


B

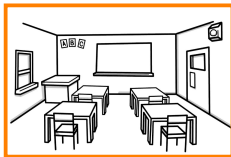


C

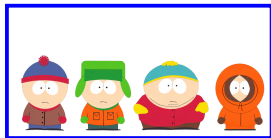
# Presence of symmetries



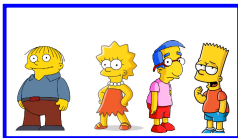
1



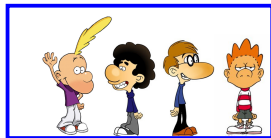
2



A

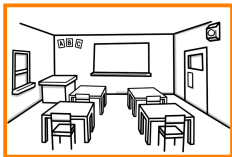


B

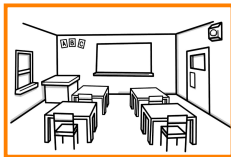


C

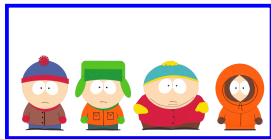
# Presence of symmetries



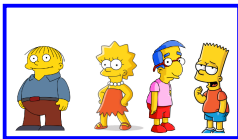
2



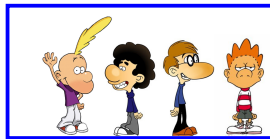
1



A



C

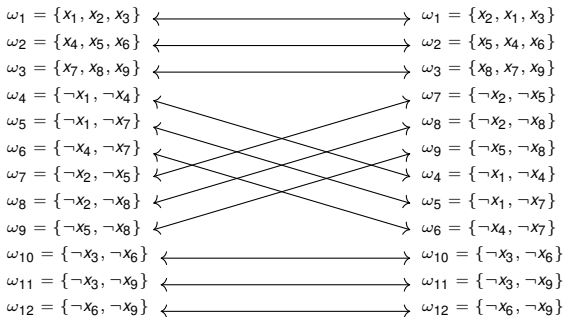


B

# Symmetry (Syntactic)

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

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



Equi-satisfiability:

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

# Computing symmetries of a SAT problem

*CNF formula*

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

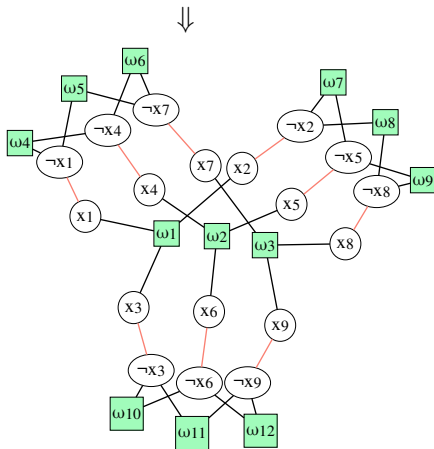
# Computing symmetries of a SAT problem

CNF formula

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colored graph





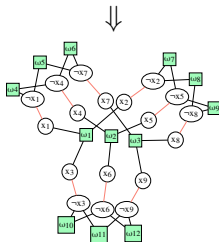
# Computing symmetries of a SAT problem

CNF formula

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colored graph



graph automorphism



(bliss, saucy, ...)

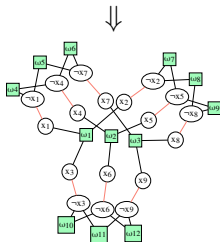
# Computing symmetries of a SAT problem

CNF formula

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



colored graph



⇓  
graph automorphism

⇓  
(bliss, saucy, ...)

⇓  
set of symmetries

⇓

$$\begin{aligned} g_1 &= (x_2 \ x_3)(x_5 \ x_6)(x_8 \ x_9) \\ g_2 &= (x_4 \ x_7)(x_5 \ x_8)(x_6 \ x_9) \\ g_3 &= (x_1 \ x_2)(x_4 \ x_5)(x_7 \ x_8) \\ g_4 &= (x_1 \ x_4)(x_2 \ x_5)(x_3 \ x_6) \end{aligned}$$

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

Exploitation of symmetries

Static symmetry breaking

# Orbit

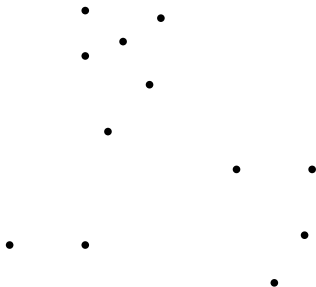
Orbit of an assignment  $\alpha = G.\alpha = \{g.\alpha \mid g \in G\}$

# Orbit

Orbit of an assignment  $\alpha = G.\alpha = \{g.\alpha \mid g \in G\}$

Example:

- full assignment

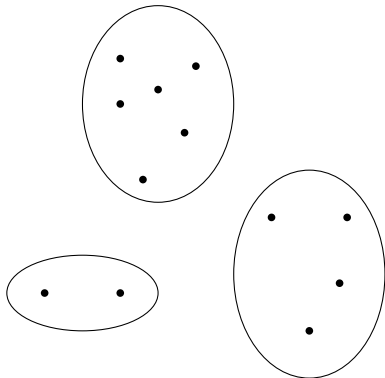


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Orbit of an assignment  $\alpha = G.\alpha = \{g.\alpha \mid g \in G\}$

Example:

- full assignment
- orbit

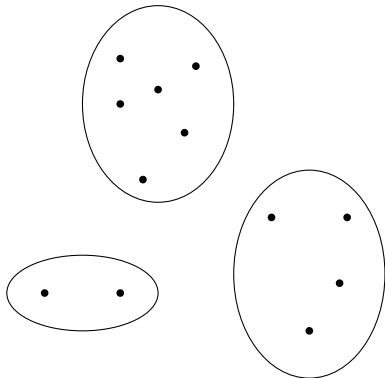


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Orbit of an assignment  $\alpha = G.\alpha = \{g.\alpha \mid g \in G\}$

Example:

- full assignment
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Equivalence relation with respect to SAT:

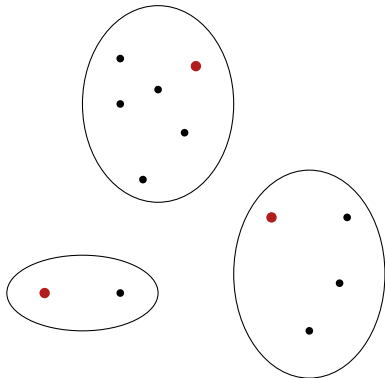
- Either  $G.\alpha$  contains no solution
- Or all elements of  $G.\alpha$  are solutions

# Orbit

Orbit of an assignment  $\alpha = G.\alpha = \{g.\alpha \mid g \in G\}$

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



# Compare assignment

Define an ordering relation to compare assignment

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

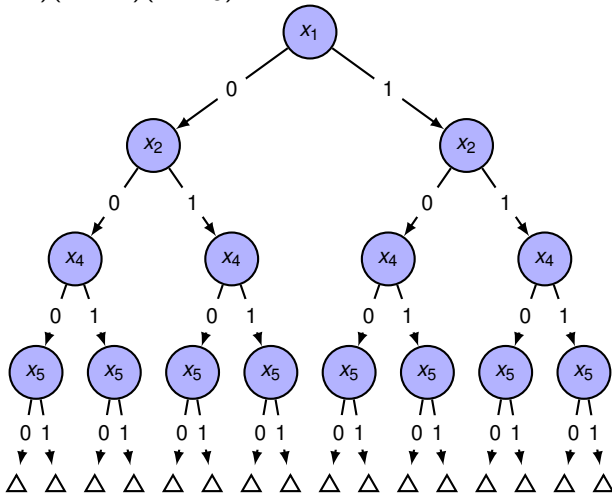
Allow only minimal (maximal) value

Forbids other assignment in each orbit

Adds Symmetry breaking predicates (SBP)

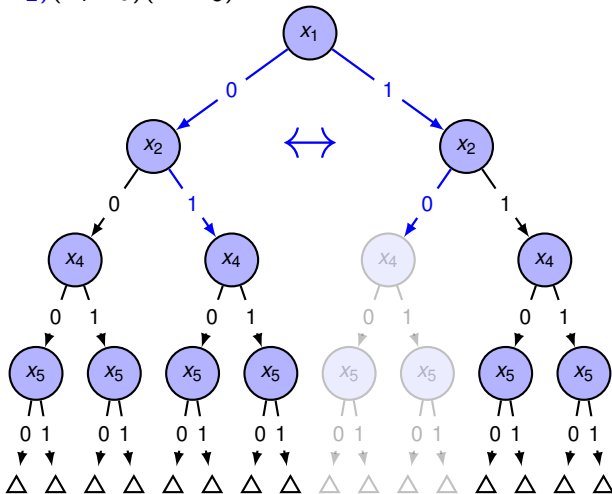
# Using symmetries to prune the search space

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

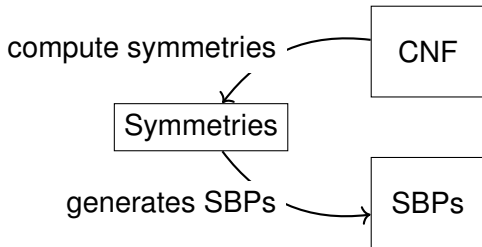


# Using symmetries to prune the search space

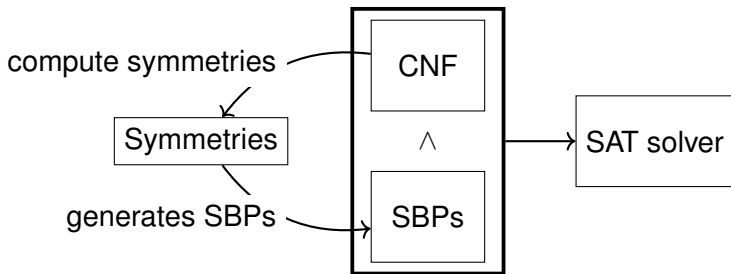
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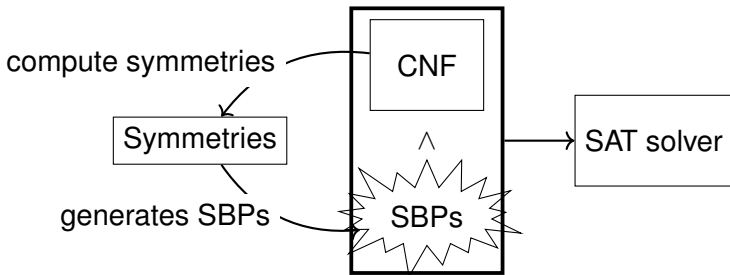
# Static symmetry breaking



# Static symmetry breaking



# Static symmetry breaking



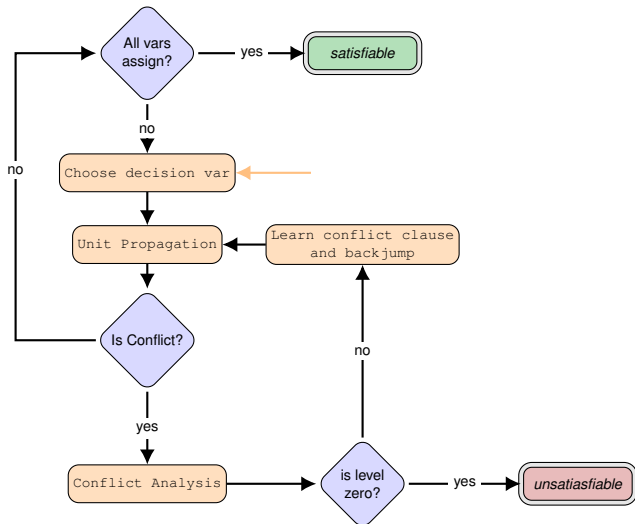
The solver can "explode" instead of being helped

Our contribution CDCL[sym]

TACAS'18 [MBCK18]

# Our contribution CDCL[Sym]

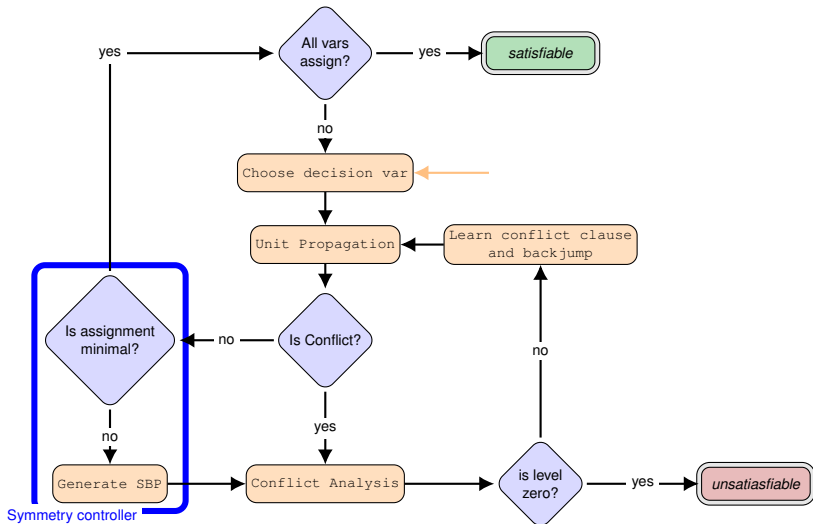
Compute and inject SBP **opportunistically**, during the solving





# Our contribution 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`  $\Rightarrow$  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

# CDCL[Sym] Implementation

- Packaged as a library **cosy**<sup>1</sup>
  - Lightweight
  - Fast update and low memory
  - Follows symmetry status
- 
- Works with any enumerative SAT solver
  - Can be integrated easily
- e.g. +3% LOC on MiniSAT.

---

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

# Experiments

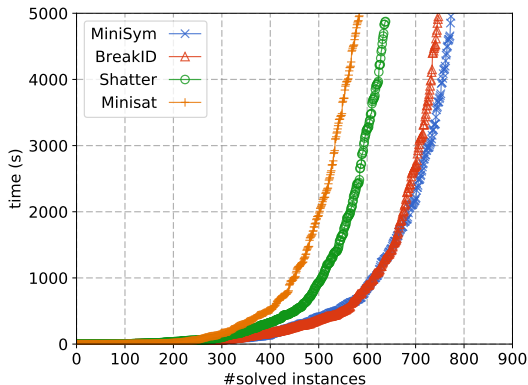
## Benchmark:

- from SAT contests 2012 – 2017
- filter: `bliss` finds symmetries in 1000s
- 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)
- 5000s timeout, 8GB memory
- includes time to compute symmetries (except for MiniSat)

# Experimental results

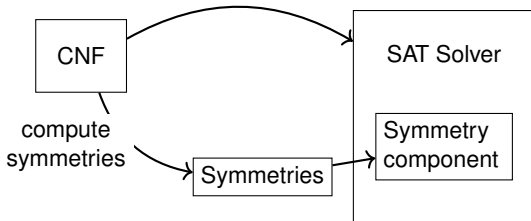


Solver	PAR-2	ALL	SAT	UNSAT
MiniSAT	2243h	586	325	261
Shatter	2088h	640	316	324
BreakID	1790h	749	334	415
MiniSym	<b>1735h</b>	<b>775</b>	<b>336</b>	<b>439</b>

Exploitation of symmetries

Dynamic symmetry breaking

# Dynamic Symmetry Breaking

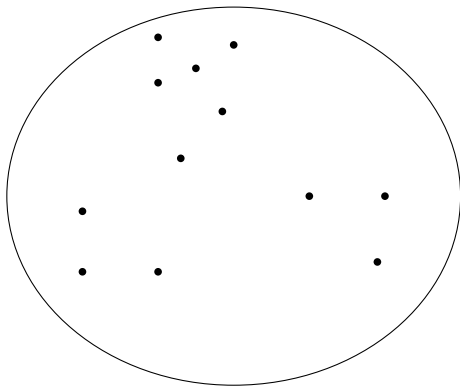


State-of-the-art:

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

# Learning symmetrical clauses

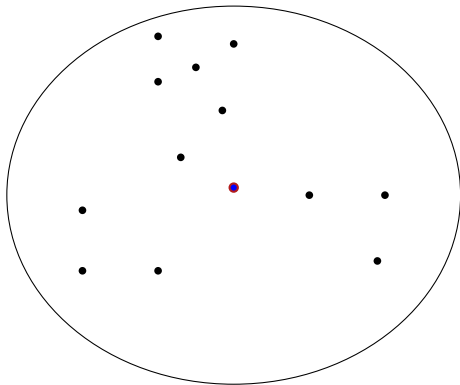
○ formula  
• clause





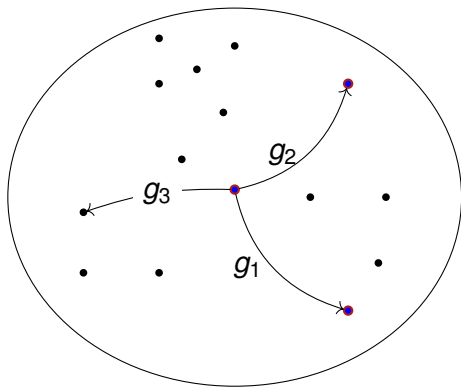
# Learning symmetrical clauses

- formula
- clause
- learnt clause



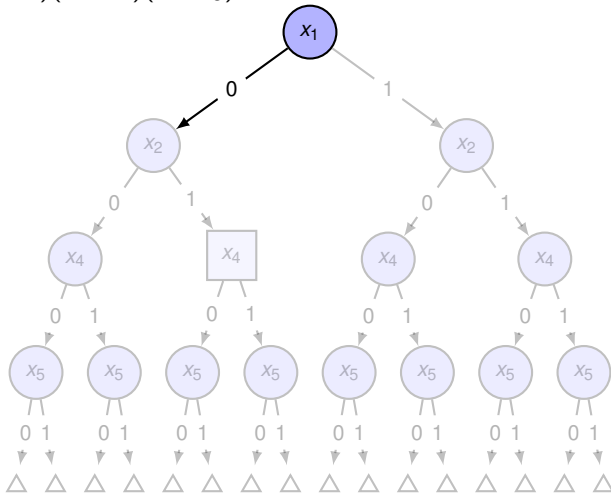
# Learning symmetrical clauses

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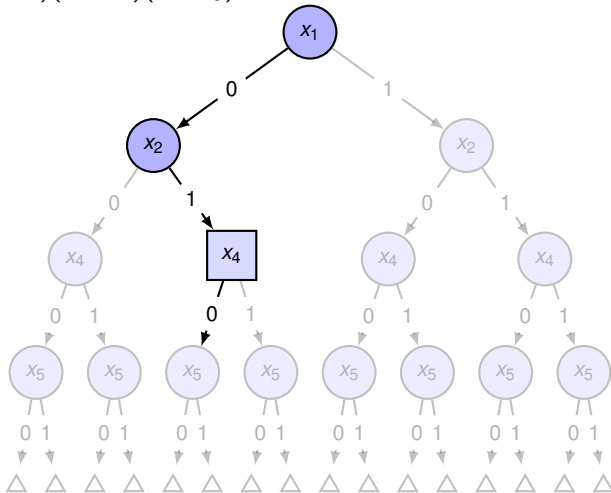
# Using symmetries to accelerate the tree traversal

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



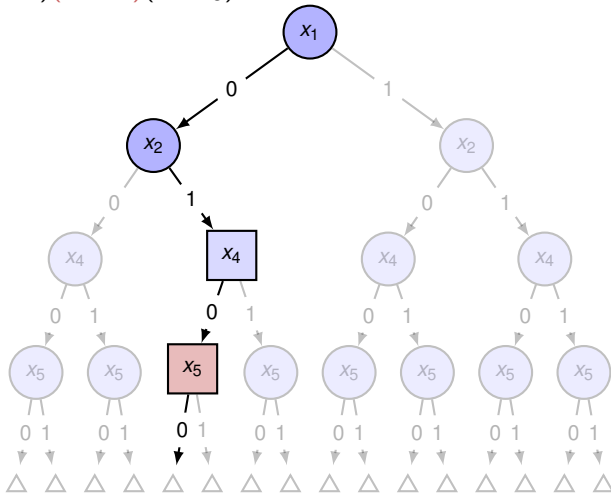
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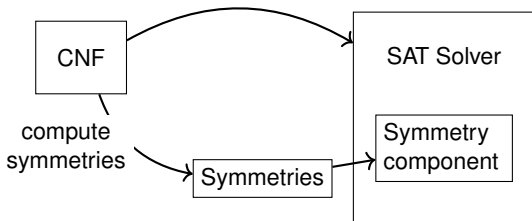
# Using symmetries to accelerate the tree traversal

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



Use symmetries to deduce symmetrical facts.

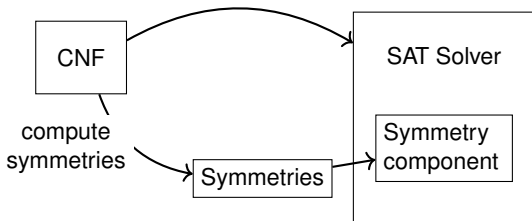
# Dynamic Symmetry Breaking



State-of-the-art:

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# 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]
- ...

Cannot handle some instances solved by static approach

Our contribution CDCL[sym]

NFM'19 [MBK19]



## ESBP + SP [MBK19]

Compose the symmetry propagation and the ESBP

*prune the decision tree while accelerating its traversal*

Problems:

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

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

# Local symmetries

Formula  $\leftarrow$  (Symmetries)

$\omega_1 \leftarrow$  (Local symmetries)

$\omega_2 \leftarrow$  (Local symmetries)

$\omega_3 \leftarrow$  (Local symmetries)

$\omega_4 \leftarrow$  (Local symmetries)

Macro level

$\rightarrow$

Micro level

# Local symmetries

Formula  $\leftarrow$  (Symmetries)

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$\omega_2 \leftarrow$  (Local symmetries)

$\omega_3 \leftarrow$  (Local symmetries)

$\omega_4 \leftarrow$  (Local symmetries)

$\omega_5$

Macro level

$\rightarrow$

Micro level

# Local symmetries

Formula  $\leftarrow$  (Symmetries)

$\omega_1 \leftarrow$  (Local symmetries)

$\omega_2 \leftarrow$  (Local symmetries)

$\omega_3 \leftarrow$  (Local symmetries)

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Macro level

$\rightarrow$

Micro level

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

# Local symmetries

Formula  $\leftarrow$  (Symmetries)

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$\omega_3 \leftarrow$  (Local symmetries)

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$\omega_5 \leftarrow$  (Local symmetries)

Macro level

$\rightarrow$

Micro level

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

- Inductive construction of the valid symmetries
- During the solving
- At a minimal cost

# Experimental results

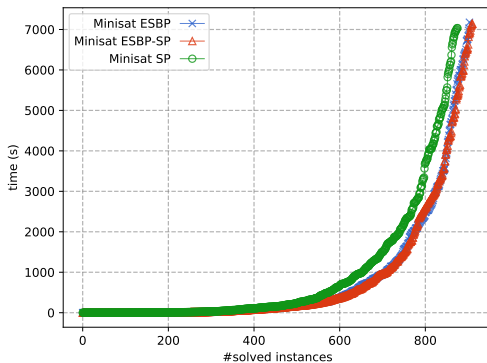
## Benchmark:

- 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

# Experimental results



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

# 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
- Combination with parallel SAT solver
- Exploitation of partial symmetries

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- Combination of CDCL[Sym] with other dynamic symmetry breaking approach
- Combination with parallel SAT solver
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**Thanks !**



Fadi A. Aloul, Karem A. Sakallah, and Igor L. Markov.  
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In *AAAI*, volume 5, pages 467–474, 2005.

# CDCL in action TODO



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

$$\omega_2 = \{x_4, x_5, x_6\}$$

$$\omega_3 = \{\neg x_1, \neg x_5\}$$

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

$$\omega_5 = \{\neg x_3, \neg x_4\}$$

$$\omega_6 = \{\neg x_3, \neg x_6\}$$

# CDCL in action TODO



$$\omega_1 = \{\mathbf{x}_1, x_2, x_3\}$$

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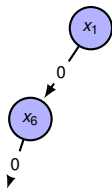
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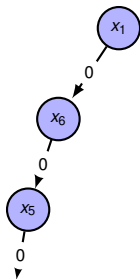
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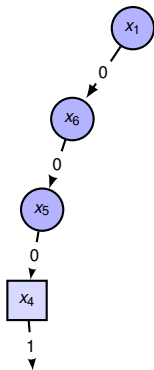
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# CDCL in action TODO



$$\omega_1 = \{\mathbf{x}_1, x_2, x_3\}$$

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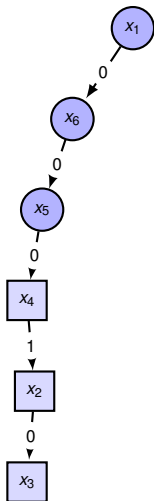
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# CDCL in action TODO



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

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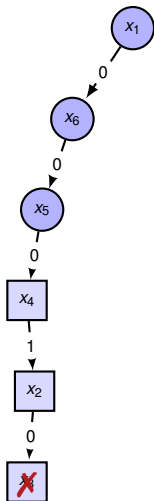
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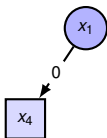
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$$\omega_7 = \{x_1, \neg x_4\}$$

# 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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Let a subset  $\delta \subseteq \alpha$ , a symmetry  $\sigma$  of  $\varphi$  such that  $\varphi \cup \delta \vdash \varphi \cup \alpha \wedge \sigma.\delta \subseteq \alpha$  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

## Logical consequence

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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
- $\varphi_d$  is the set of deduced clauses.

## Local symmetries

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

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

# Generates symmetry breaking predicates (SBP)

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

Example:

$$x_1 \leq x_2 \leq x_3 \leq x_4 \leq x_5 \leq x_6 \leq x_7 \leq x_8; \textcolor{red}{F} < \textcolor{green}{T}$$

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

$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$\dots$	lex-leader	SBP

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	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$\dots$	lex-leader	SBP
$O_1$	$\textcolor{red}{F}$	$\textcolor{green}{T}$	-	-	-	$\dots$	✓	

# Generates symmetry breaking predicates (SBP)

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

Example:

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

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

	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$\dots$	lex-leader	SBP
$O_1$	F	T	—	—	—	$\dots$	✓	$\rightarrow \neg x_1 \vee x_2$
	T	F	—	—	—	$\dots$	✗	

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$O_1$	F	T	—	—	—	$\dots$	✓	$\rightarrow \neg x_1 \vee x_2$
	T	F	—	—	—	$\dots$	✗	
$O_2$	F	F	—	F	T	$\dots$	✓	

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$O_1$	$\textcolor{red}{F}$	$\textcolor{green}{T}$	—	—	—	$\dots$	✓	$\rightarrow \neg x_1 \vee x_2$
	$\textcolor{green}{T}$	$\textcolor{red}{F}$	—	—	—	$\dots$	✗	
$O_2$	$\textcolor{red}{F}$	$\textcolor{red}{F}$	—	$\textcolor{red}{F}$	$\textcolor{green}{T}$	$\dots$	✓	$\rightarrow x_1 \vee x_2 \vee \neg x_4 \vee x_5$
	$\textcolor{red}{F}$	$\textcolor{red}{F}$	—	$\textcolor{green}{T}$	$\textcolor{red}{F}$	$\dots$	✗	

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$O_1$	F	T	—	—	—	$\cdots$	✓	$\rightarrow \neg x_1 \vee x_2$
	T	F	—	—	—	$\cdots$	✗	
$O_2$	F	F	—	F	T	$\cdots$	✓	$\rightarrow x_1 \vee x_2 \vee \neg x_4 \vee x_5$
	F	F	—	T	F	$\cdots$	✗	
$\cdots$								

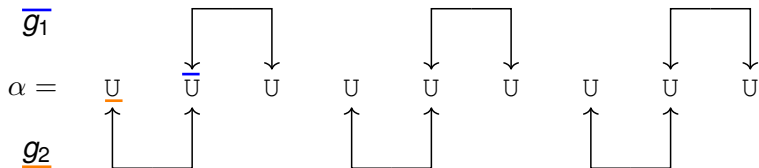


# Example

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

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

$$\textcolor{red}{F} < \textcolor{green}{T} \quad x_1 \leq x_2 \leq x_3 \leq x_4 \leq x_5 \leq x_6 \leq x_7 \leq x_8 \leq x_9$$

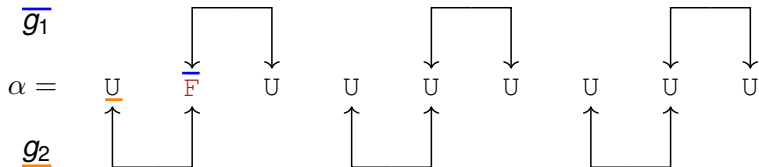


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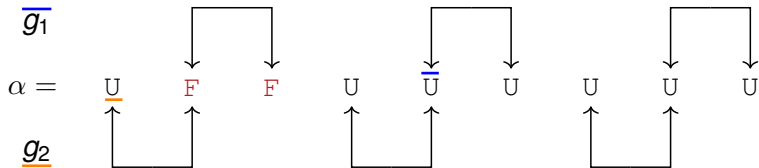


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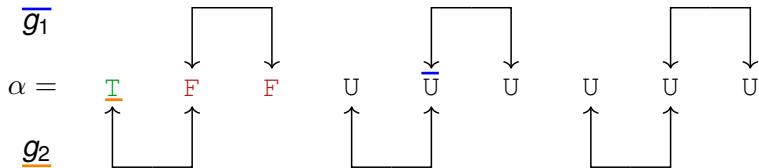


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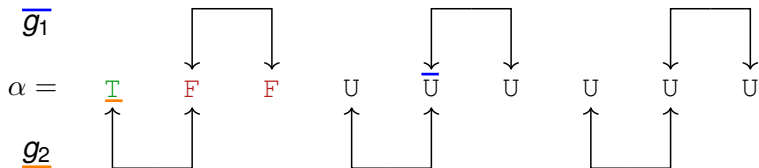


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$g_2$  generates ESBP  $\omega = \{\neg x_1, x_2\}$

# Example

- 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

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

$$g_1 = \begin{array}{ccc} (x_2 & x_3) & (x_5 & x_6) & (x_8 & x_9) \end{array} \left| \begin{array}{l} x = x_2 \\ g.x = x_3 \\ \text{active} \end{array} \right.$$

↑

$$g_2 = \begin{array}{ccc} (x_1 & x_2) & (x_4 & x_5) & (x_7 & x_8) \end{array} \left| \begin{array}{l} x = x_1 \\ g.x = x_2 \\ \text{active} \end{array} \right.$$

↑

...

$$\alpha = \{ \quad \quad \quad \}$$

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$$g_2 = \begin{array}{ccc} (x_1 & \textcolor{red}{x}_2) & (x_4 & x_5) & (x_7 & x_8) \end{array} \left| \begin{array}{l} x = x_1 \\ g.x = \textcolor{red}{x}_2 \\ \text{active} \end{array} \right.$$

...

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

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$$g_1 = \begin{array}{cc|cc} (x_2 & x_3) & (x_5 & x_6) & (x_8 & x_9) \\ \uparrow & & & & & \end{array} \quad \left| \quad x = x_5 \quad g.x = x_6 \right.$$

active

$$g_2 = \begin{array}{cc|cc} (x_1 & x_2) & (x_4 & x_5) & (x_7 & x_8) \\ \uparrow & & & & & \end{array} \quad \left| \quad x = x_1 \quad g.x = x_2 \right.$$

reducer

...

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



# Example

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$$g_2 = \begin{array}{cc|cc} (x_1 & x_2) & (x_4 & x_5) & (x_7 & x_8) \\ \uparrow & & & & & \end{array} \left| \begin{array}{l} x = x_1 \\ g.x = x_2 \\ \text{reducer} \end{array} \right.$$

...

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

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

# Encoding the problem

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

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

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

$x_1 \vee x_2 \vee x_3$

$x_4 \vee x_5 \vee x_6$

$x_7 \vee x_8 \vee x_9$

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

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

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

$\neg x_1 \vee \neg x_4$

$\neg x_1 \vee \neg x_7$

$\neg x_4 \vee \neg x_7$

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

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

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

$\neg x_2 \vee \neg x_5$

$\neg x_2 \vee \neg x_8$

$\neg x_5 \vee \neg x_8$

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

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

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

$\neg x_3 \vee \neg x_6$

$\neg x_3 \vee \neg x_9$

$\neg x_6 \vee \neg x_9$