

# Symmetry in SAT (and ASP and CP): Breaking the right symmetries

Bart Bogaerts

Aalto University

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# Main Reference



Jo Devriendt, Bart Bogaerts, and Maurice Bruynooghe.

BreakIDGlucose: On the importance of row symmetry.

*In Proceedings of the Fourth International Workshop on the Cross-Fertilization Between CSP and SAT (CSPSAT), 2014.*

# Take Home Message

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Don't go implementing any of the methods I describe in this talk!  
(unless you really have to)

# Content

- 1 Motivation
- 2 Symmetries
  - Constraint Programming
  - SAT Solving
- 3 Row Interchangeability Detection: Take 1
- 4 Row Interchangeability Detection: Take 2
- 5 Row Interchangeability Detection: Take 3
- 6 Final thoughts

# Motivation

- 2012
- “Symmetry Propagation” for SAT [Devriendt et al., 2012]
- Add symmetric variants of learnt clauses lazily
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# CP perspective

- CSP:  $(V, D, C)$
- assignment  $\alpha : (V \rightarrow D)$
- solution satisfies constraints

For example:

- $V = \{a, b, c, d, e, f\}$
- $D = \{0, 1\}$
- $C = \{b \leq 0\}$
- $\alpha = \{a = 1, b = 0, c = 1, d = 1, e = 1, f = 1\}$
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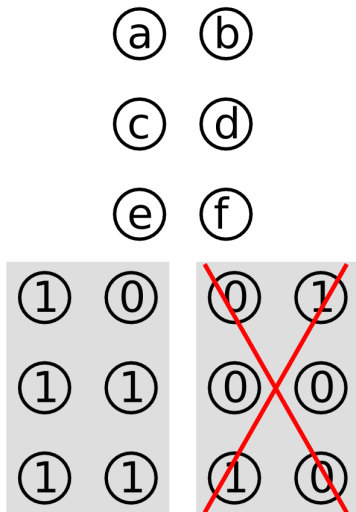
ⓐ	ⓑ		
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1	0	0	1
1	1	0	0
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# CP perspective on symmetry

$$S : (V \rightarrow D) \rightarrow (V \rightarrow D)$$

**symmetry**  $S$  is a permutation of the set of assignments preserving satisfaction to all constraints

Recall:

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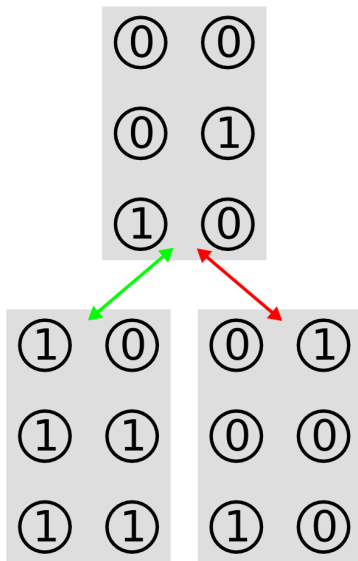
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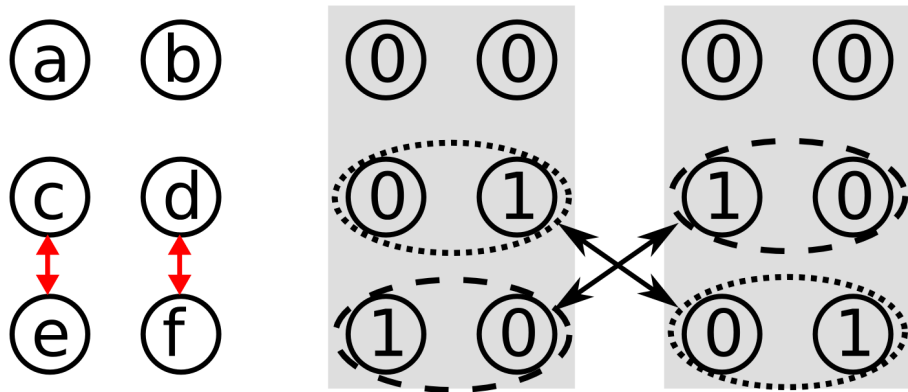
# CP perspective on symmetry

Common restriction:

**variable symmetry**  $S_P : \alpha \mapsto \alpha \circ P$

**induced** by variable permutation  $P : V \rightarrow V$

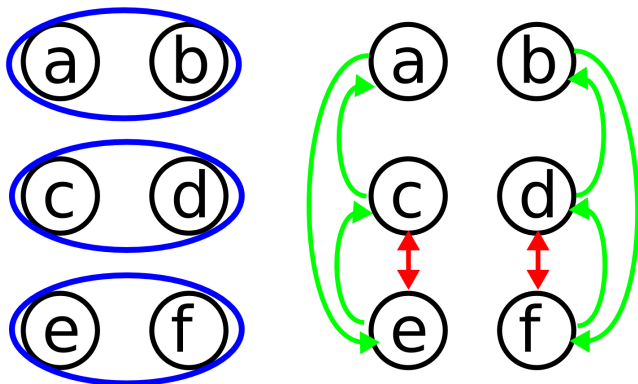
For example:  $P = (ce)(df)$



# CP perspective on row symmetry

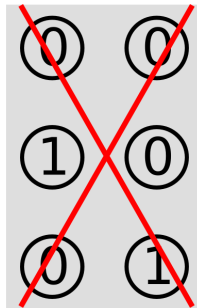
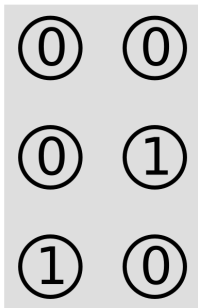
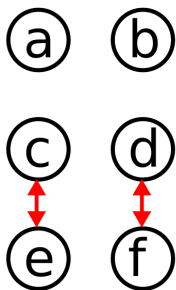
## Row symmetry:

- special case of variable symmetry
- assumption:  $V$  is ordered as a matrix  $M$
- $P$  permutes the **rows** of  $M$
- CSP is **row-interchangeable** for  $M$  iff all permutations on the rows of  $M$  induce a symmetry



## CP perspective on row symmetry

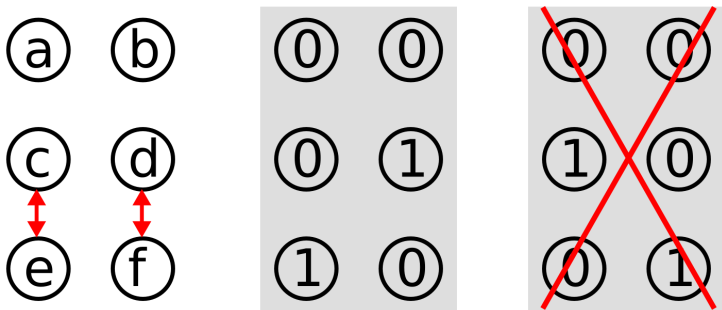
**Symmetry breaking:** speed up search by adding extra constraints to remove symmetrical assignments from the assignment space.





## CP perspective on row symmetry

**Symmetry breaking:** speed up search by adding extra constraints to remove symmetrical assignments from the assignment space.



**Complete** symmetry breaking: maximal number of symmetric assignments removed while retaining soundness.

CP result: row interchangeability can be broken completely by enforcing lexicographic order on rows. [Flener et al., 2002]

# SAT perspective

- SAT instance:  $(V, \{0, 1\}, C)$ ,  $C$  is a CNF
- assignment  $\alpha : (V \rightarrow D)$
- solution satisfies  $C$

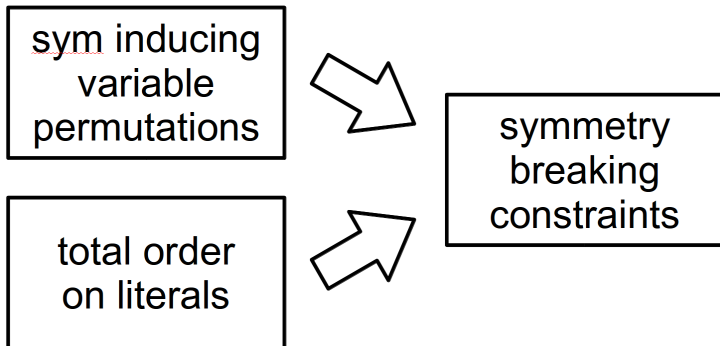
# SAT perspective on symmetry

- SAT instance:  $(V, \{\mathbf{t}, \mathbf{f}\}, C)$ ,  $C$  is a CNF
- assignment  $\alpha : (V \rightarrow D)$
- solution satisfies  $C$
- A *variable symmetry* is a permutation  $P$  of  $V$  such that  $\alpha \models C$  iff  $\alpha \circ P \models C$

(this is exactly the same as in CP)

## SAT perspective on breaking symmetry

General symmetry breaking in SAT as per Saucy+Shatter  
[Aloul et al., 2006]:

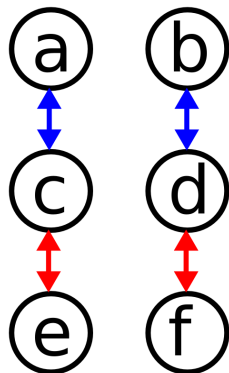


For each variable permutation  $P$  (**in some set of generators**) and for each variable  $v$ , add the lex-leader constraint  
 $(\forall v' \prec v : v' = P(v')) \Rightarrow v \leq P(v)$ .

# Can Shatter completely break row interchangeability?

- variable permutations:  $P_{12}$ ,  $P_{23}$   
that induce row interchangeability
- order on variables:  $a \prec b \prec c \prec d \prec e \prec f$
- symmetry breaking constraints:  
 $a \leq c$ ,  $a = c \Rightarrow b \leq d$ ,  
 $c \leq e$ ,  $c = e \Rightarrow d \leq f$

**These constraints actually state that the rows of each solution must be lexicographically ordered!**



Row interchangeability can be completely broken by standard SAT symmetry breaking methods, given the right variable permutations and variable ordering.

# What can go wrong?

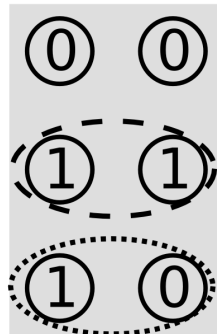
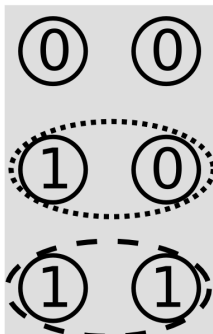
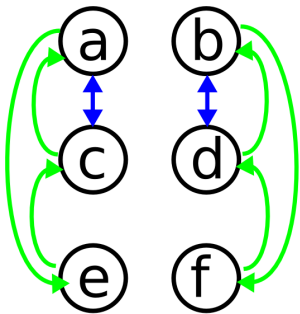
- variable permutations:  $P_{12}, P_{321}$   
that induce row interchangeability
- order on variables:  $a \prec b \prec c \prec d \prec e \prec f$
- symmetry breaking constraints:

$$a \leq c, a = c \Rightarrow b \leq d,$$

$$a \leq e, a = e \Rightarrow b \leq f,$$

$$a = e \wedge b = f \Rightarrow c \leq a, a = e \wedge b = f \wedge c = a \Rightarrow d \leq b$$

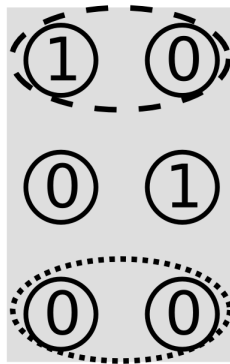
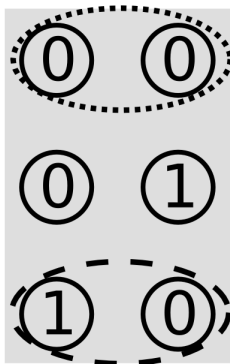
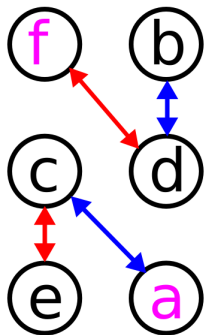
These do not represent lexicographic row ordering constraint



## What can go wrong?

- variable permutations:  $P_{12}$ ,  $P_{23}$   
that induce row interchangeability
- order on variables:  $f \prec b \prec c \prec d \prec e \prec a$
- symmetry breaking constraints:  
 $b \leq d$ ,  $b = d \Rightarrow c \leq a$ ,  
 $f \leq d$ ,  $f = d \Rightarrow c \leq e$

These do not represent lexicographic row ordering constraint



## SAT perspective: conclusion

To completely break row interchangeability:

- need for right generator symmetries
- need for right ordering on variables

Easy when you know the matrix structure of the variables. . .

How to detect matrix structure of variables in a CNF?  
How to detect row interchangeability in a CNF?



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## Pigeon Hole

- SAT encoding of the CSP  $(V, D, C)$
- $V = \{1..n\}, D = \{1..n-1\}$
- $C = \{\forall d \in D : \#\{v \mid \alpha(v) = d\} \leq 1\}$ .
- (Boolean encoding of  $\alpha(i)$  is a “row”)

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- (Boolean encoding of  $\alpha(i)$  is a “row”)
- Any permutation of variables induces a symmetry
- Saucy: generators  $(12), (23), (34), (45), \dots$
- with order on variables  $1 \prec 2 \prec 3 \prec \dots$ : breaking constraints

$$\alpha(1) \leq \alpha(2), \alpha(2) \leq \alpha(3), \alpha(3) \leq \alpha(4) \dots$$

- Breaks the symmetry group completely (no two symmetric assignments satisfy this constraint)

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- Breaks the symmetry group completely (no two symmetric assignments satisfy this constraint)
- however... with order  $2 \prec 1 \prec 4 \prec 3 \prec \dots$ :

$$\alpha(2) \leq \alpha(1), \alpha(2) \leq \alpha(3), \alpha(4) \leq \alpha(3), \dots$$

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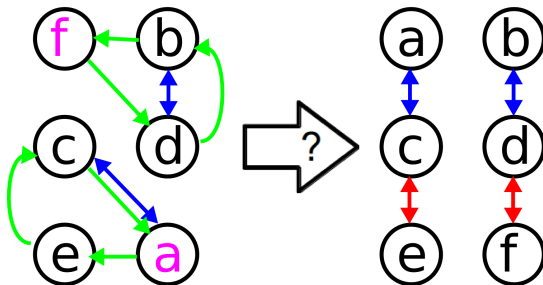
# Row Interchangeability Detection

Problem statement:

## Row Interchangeability Detection (RID)

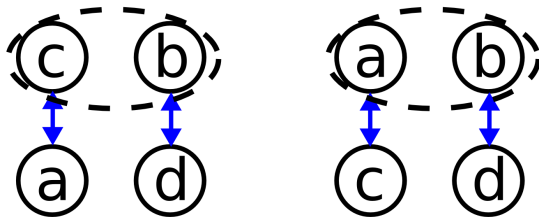
Given a CNF, find a maximal set of variables that form a variable matrix with interchangeable rows.

Chosen problem perspective: start from detected symmetry group.



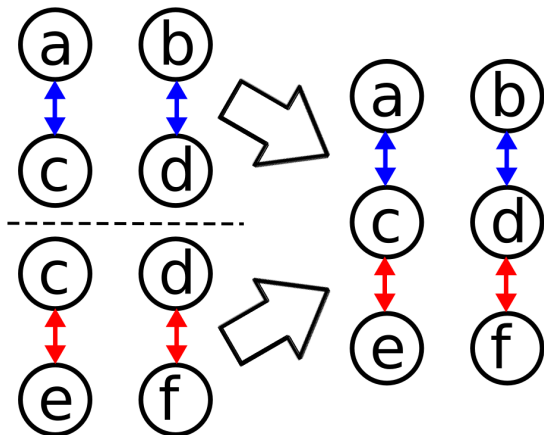
## Involution symmetries form rows

A permutation  $P$  for which  $P^2 = I$ , is an **involution**. Each variable involution forms multiple interchangeable row matrices, with only 2 rows:



## Involution symmetries form rows

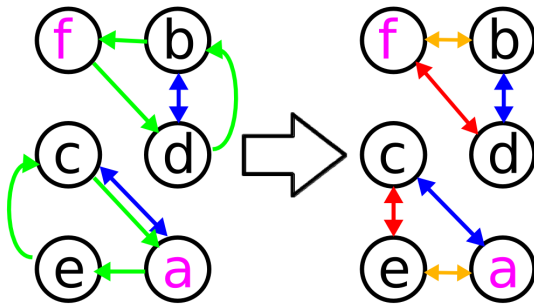
Compatible involutions can be combined to row interchangeable matrices with more than 2 rows:





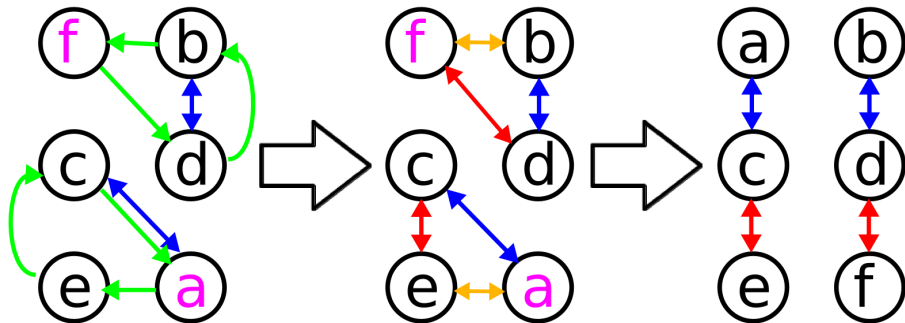
## Heuristically search for involutions

Start from generator symmetries returned by Saucy. Compose symmetries to form *small* involutions.



## Matrix structure detection by involutions

Combining involution generation & matrix extraction solves RID:



Can be extended to the **piecewise** case, where multiple disjoint row interchangeable variable matrices exist for one problem.

Piecewise Row Interchangeability Detection – the **PRID** problem.

# Working implementation: BreakIDGlucose

- extends Shatter with row interchangeability handling
  - ▶ use Saucy to detect symmetry inducing variable permutations
  - ▶ **new:** generate involutions to solve PRID
  - ▶ **new:** add row involutions to set of variable permutations
  - ▶ **new:** adjust order on variables as per detected rows
  - ▶ add Shatter's lex-leader constraints
- used Glucose 2.1 as SAT solving engine
- obtained gold medal at 2013's SAT competition hard combinatorial track
- could only solve 5 out of 14 two-pigeon-hole instances

Complete row interchangeability breaking is relevant in SAT  
Improvement possible with better detection of matrix structure – better answer to PRID.

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## Working implementation 2: BreakIDGlucose2

- extends Shatter with row interchangeability handling
  - ▶ use Saucy to detect a set  $G$  of symmetry inducing variable permutations
  - ▶ **new:** search for  $g_1, g_2 \in G$ : two “matching” involutions (rows  $r_1, r_2$  and  $r_3$ )
  - ▶ **new:** for each  $g \in G$ ,  $g(r_i)$  is a candidate row: check whether  $r_1 \leftrightarrow g(r_i)$  is a symmetry
  - ▶ **new:** use Saucy to find more permutations that do not permute rows  $r_2, r_3, \dots$
  - ▶ **new:** continue extending the row-interchangeability matrix
  - ▶ **new:** adjust order on variables as per detected rows
  - ▶ add Shatter’s lex-leader constraints
- used Glucose 4.0 as SAT solving engine
- obtained 10th place out of 28 in the 2015 SAT Race (best of all Glucose variants)

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# Work in progress: tackling the actual problem

Two main directions

- Modify Saucy to give “the right” generators
- Use algebraic tools (e.g., GAP) to modify the set of generators

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# Where does row interchangeability in SAT come from?

SAT encodings of

- variable interchangeable CSP's
- value interchangeable CSP's
- row interchangeable CSP's
- relational model generation problems where relation  $R : D_1 \times \dots \times D_n$  has to be found and some disjoint  $D_i$  contains interchangeable elements.

For example: the IDP system.

Note the triviality of solving the PRID problem in such a high level language!

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(unless you really have to)

“There are no CNF problems” (P.J. Stuckey, 2013)

# Thanks for your attention!

## Questions?



Aloul, F. A., Sakallah, K. A., and Markov, I. L. (2006).  
Efficient symmetry breaking for Boolean satisfiability.  
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