### Conflict-Driven Clause Learning

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http://www.cs.cmu.edu/~mheule/15816-f21/ Automated Reasoning and Satisfiability September 22, 2021

# The Satisfiability (SAT) problem

$$(x_5 \lor x_8 \lor \overline{x}_2) \land (x_2 \lor \overline{x}_1 \lor \overline{x}_3) \land (\overline{x}_8 \lor \overline{x}_3 \lor \overline{x}_7) \land (\overline{x}_5 \lor x_3 \lor x_8) \land (\overline{x}_6 \lor \overline{x}_1 \lor \overline{x}_5) \land (x_8 \lor \overline{x}_9 \lor x_3) \land (x_2 \lor x_1 \lor x_3) \land (\overline{x}_1 \lor x_8 \lor x_4) \land (\overline{x}_9 \lor \overline{x}_6 \lor x_8) \land (x_8 \lor \overline{x}_9 \lor x_5) \land (x_9 \lor \overline{x}_3 \lor x_8) \land (x_6 \lor \overline{x}_9 \lor x_5) \land (x_2 \lor \overline{x}_3 \lor \overline{x}_8) \land (x_8 \lor \overline{x}_6 \lor \overline{x}_3) \land (x_8 \lor \overline{x}_3 \lor \overline{x}_1) \land (\overline{x}_8 \lor x_6 \lor \overline{x}_2) \land (x_7 \lor x_9 \lor \overline{x}_2) \land (x_8 \lor \overline{x}_9 \lor x_2) \land (\overline{x}_1 \lor \overline{x}_9 \lor x_4) \land (x_8 \lor x_1 \lor \overline{x}_2) \land (x_3 \lor \overline{x}_4 \lor \overline{x}_6) \land (\overline{x}_1 \lor \overline{x}_7 \lor x_5) \land (\overline{x}_7 \lor x_1 \lor x_6) \land (\overline{x}_5 \lor x_4 \lor \overline{x}_6) \land (\overline{x}_4 \lor x_9 \lor \overline{x}_8) \land (x_2 \lor x_9 \lor x_1) \land (x_5 \lor \overline{x}_7 \lor x_1) \land (\overline{x}_7 \lor \overline{x}_9 \lor \overline{x}_6) \land (x_2 \lor x_5 \lor x_4) \land (x_8 \lor \overline{x}_4 \lor x_5) \land (x_5 \lor x_9 \lor x_3) \land (\overline{x}_5 \lor \overline{x}_7 \lor x_9) \land (x_2 \lor \overline{x}_8 \lor x_1) \land (\overline{x}_7 \lor x_1 \lor x_5) \land (x_1 \lor x_4 \lor x_3) \land (x_1 \lor \overline{x}_9 \lor \overline{x}_4) \land (x_3 \lor x_5 \lor x_6) \land (\overline{x}_6 \lor x_3 \lor \overline{x}_9) \land (\overline{x}_7 \lor x_5 \lor x_9) \land (x_7 \lor \overline{x}_5 \lor \overline{x}_2) \land (x_4 \lor x_7 \lor x_3) \land (x_4 \lor \overline{x}_9 \lor \overline{x}_7) \land (x_5 \lor \overline{x}_1 \lor x_7) \land (x_6 \lor x_7 \lor \overline{x}_3) \land (\overline{x}_8 \lor \overline{x}_2 \lor x_5)$$

Does there exist an assignment satisfying all clauses?

## Search for a satisfying assignment (or proof none exists)

$$\begin{array}{c} (\mathbf{x}_5 \vee \mathbf{x}_8 \vee \overline{\mathbf{x}}_2) \wedge (\mathbf{x}_2 \vee \overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_3) \wedge (\overline{\mathbf{x}}_8 \vee \overline{\mathbf{x}}_3 \vee \overline{\mathbf{x}}_7) \wedge (\overline{\mathbf{x}}_5 \vee \mathbf{x}_3 \vee \mathbf{x}_8) \wedge \\ (\overline{\mathbf{x}}_6 \vee \overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_5) \wedge (\mathbf{x}_8 \vee \overline{\mathbf{x}}_9 \vee \mathbf{x}_3) \wedge (\mathbf{x}_2 \vee \mathbf{x}_1 \vee \mathbf{x}_3) \wedge (\overline{\mathbf{x}}_1 \vee \mathbf{x}_8 \vee \mathbf{x}_4) \wedge \\ (\overline{\mathbf{x}}_9 \vee \overline{\mathbf{x}}_6 \vee \mathbf{x}_8) \wedge (\mathbf{x}_8 \vee \overline{\mathbf{x}}_9 \vee \mathbf{x}_3) \wedge (\mathbf{x}_9 \vee \overline{\mathbf{x}}_3 \vee \mathbf{x}_8) \wedge (\mathbf{x}_6 \vee \overline{\mathbf{x}}_9 \vee \mathbf{x}_5) \wedge \\ (\mathbf{x}_2 \vee \overline{\mathbf{x}}_3 \vee \overline{\mathbf{x}}_8) \wedge (\mathbf{x}_8 \vee \overline{\mathbf{x}}_6 \vee \overline{\mathbf{x}}_3) \wedge (\mathbf{x}_8 \vee \overline{\mathbf{x}}_3 \vee \overline{\mathbf{x}}_1) \wedge (\overline{\mathbf{x}}_8 \vee \mathbf{x}_6 \vee \overline{\mathbf{x}}_2) \wedge \\ (\mathbf{x}_7 \vee \mathbf{x}_9 \vee \overline{\mathbf{x}}_2) \wedge (\mathbf{x}_8 \vee \overline{\mathbf{x}}_9 \vee \mathbf{x}_2) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_9 \vee \mathbf{x}_4) \wedge (\mathbf{x}_8 \vee \mathbf{x}_1 \vee \overline{\mathbf{x}}_2) \wedge \\ (\mathbf{x}_3 \vee \overline{\mathbf{x}}_4 \vee \overline{\mathbf{x}}_6) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_7 \vee \mathbf{x}_5) \wedge (\overline{\mathbf{x}}_7 \vee \mathbf{x}_1 \vee \mathbf{x}_6) \wedge (\overline{\mathbf{x}}_5 \vee \mathbf{x}_4 \vee \overline{\mathbf{x}}_6) \wedge \\ (\overline{\mathbf{x}}_4 \vee \mathbf{x}_9 \vee \overline{\mathbf{x}}_8) \wedge (\mathbf{x}_2 \vee \mathbf{x}_9 \vee \mathbf{x}_1) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_7 \vee \mathbf{x}_1) \wedge (\overline{\mathbf{x}}_7 \vee \overline{\mathbf{x}}_9 \vee \overline{\mathbf{x}}_6) \wedge \\ (\overline{\mathbf{x}}_4 \vee \mathbf{x}_9 \vee \overline{\mathbf{x}}_8) \wedge (\mathbf{x}_2 \vee \mathbf{x}_9 \vee \mathbf{x}_1) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_7 \vee \mathbf{x}_1) \wedge (\overline{\mathbf{x}}_7 \vee \overline{\mathbf{x}}_9 \vee \overline{\mathbf{x}}_6) \wedge \\ (\mathbf{x}_2 \vee \mathbf{x}_5 \vee \mathbf{x}_4) \wedge (\mathbf{x}_8 \vee \overline{\mathbf{x}}_4 \vee \mathbf{x}_5) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_7 \vee \mathbf{x}_1) \wedge (\overline{\mathbf{x}}_7 \vee \overline{\mathbf{x}}_9 \vee \overline{\mathbf{x}}_6) \wedge \\ (\mathbf{x}_2 \vee \overline{\mathbf{x}}_8 \vee \mathbf{x}_1) \wedge (\overline{\mathbf{x}}_7 \vee \mathbf{x}_1 \vee \mathbf{x}_5) \wedge (\mathbf{x}_1 \vee \mathbf{x}_4 \vee \mathbf{x}_3) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_9 \vee \overline{\mathbf{x}}_4) \wedge \\ (\mathbf{x}_3 \vee \mathbf{x}_5 \vee \mathbf{x}_6) \wedge (\overline{\mathbf{x}}_6 \vee \mathbf{x}_3 \vee \overline{\mathbf{x}}_9) \wedge (\overline{\mathbf{x}}_7 \vee \mathbf{x}_5 \vee \mathbf{x}_9) \wedge (\mathbf{x}_7 \vee \overline{\mathbf{x}}_5 \vee \overline{\mathbf{x}}_2) \wedge \\ (\mathbf{x}_4 \vee \mathbf{x}_7 \vee \mathbf{x}_3) \wedge (\overline{\mathbf{x}}_6 \vee \overline{\mathbf{x}}_9 \vee \overline{\mathbf{x}}_7) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_1 \vee \mathbf{x}_7) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_1 \vee \mathbf{x}_7) \wedge (\mathbf{x}_6 \vee \mathbf{x}_7 \vee \overline{\mathbf{x}}_5) \wedge (\overline{\mathbf{x}}_8 \vee \mathbf{x}_2 \vee \mathbf{x}_5) \end{pmatrix}$$

■ Most successful architecture

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- State-of-the-art (sequential) CDCL solvers: CaDiCaL, Glucose, CryptoMiniSAT

Clause Learning

Data-structures

Heuristics

Clause Management

Conflict-Clause Minimization

Recent Advances and Conclusions

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Recent Advances and Conclusions

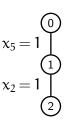
$$\begin{array}{c} (x_1 \vee x_4) \wedge \\ (x_3 \vee \overline{x}_4 \vee \overline{x}_5) \wedge \\ (\overline{x}_3 \vee \overline{x}_2 \vee \overline{x}_4) \wedge \\ \mathcal{F}_{\text{extra}} \end{array}$$



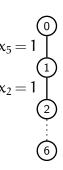
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$$x_5 = 1$$

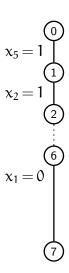
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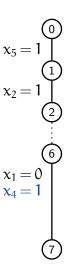
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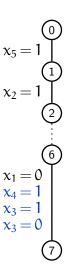
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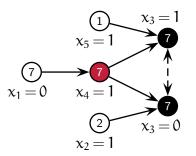
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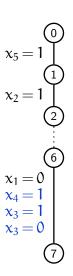


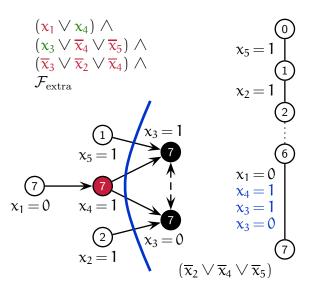
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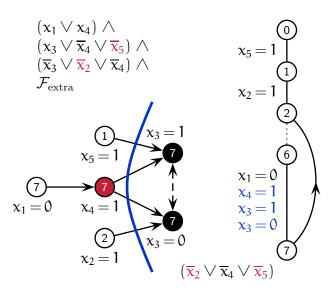


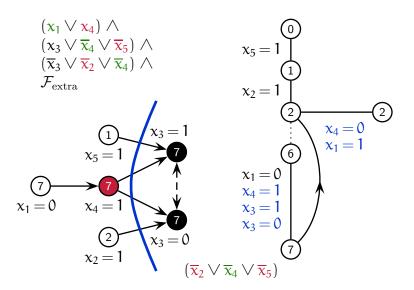


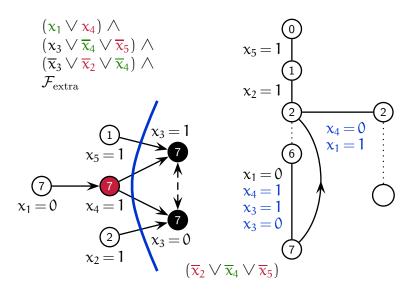












## Implication graph [Marques-SilvaSakallah '96]

#### CDCL in a nutshell:

- 1. Main loop combines efficient problem simplification with cheap, but effective decision heuristics; (> 90% of time)
- 2. Reasoning kicks in if the current state is conflicting;
- 3. The current state is analyzed and turned into a constraint;
- 4. The constraint is added to the problem, the heuristics are updated, and the algorithm (partially) restarts.

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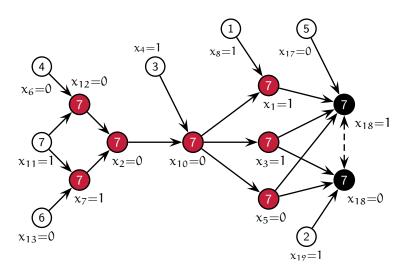
#### However, it has three weaknesses:

- CDCL is notoriously hard to parallelize;
- the representation impacts CDCL performance; and
- CDCL has exponential runtime on some "simple" problems.

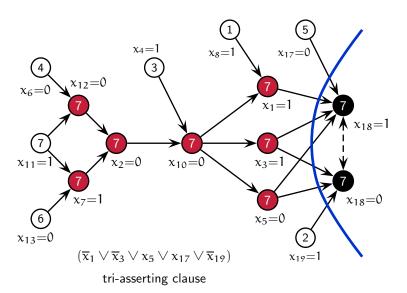
## Conflict-driven Clause Learning: Pseudo-code

```
1: while TRUE do
          l_{decision} := Decide ()
 2:
          If no l<sub>decision</sub> then return satisfiable
 3.
         \mathcal{F} := Simplify (\mathcal{F}(l_{decision} \leftarrow 1))
 4:
          while \mathcal{F} contains C_{\text{falsified}} do
 5:
               C_{conflict} := Analyze (C_{falsified})
 6:
               If C_{conflict} = \emptyset then return unsatisfiable
 7:
               BackTrack (C<sub>conflict</sub>)
 8.
               \mathcal{F} := Simplify (\mathcal{F} \cup \{C_{\text{conflict}}\})
 9:
          end while
10.
11: end while
```

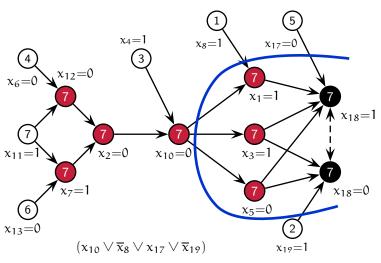
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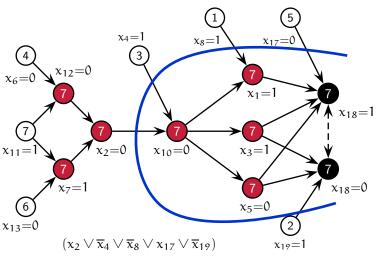


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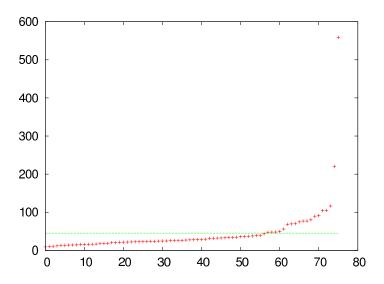
first unique implication point

# [Marques-SilvaSakallah'96]



second unique implication point

## Average Learned Clause Length



### Clause Learning

#### Data-structures

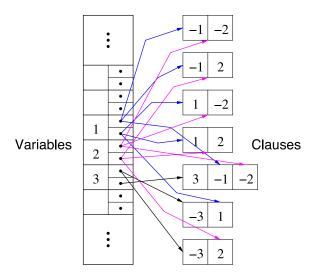
Heuristics

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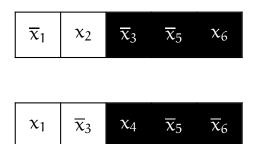
Conflict-Clause Minimization

Recent Advances and Conclusions

## Simple data structure for unit propagation



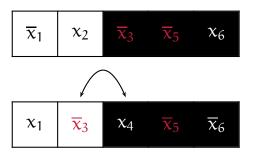
$$\phi = \{x_1 = *, x_2 = *, x_3 = *, x_4 = *, x_5 = *, x_6 = *\}$$



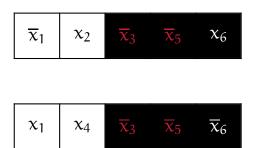
$$\varphi = \{x_1 = *, x_2 = *, x_3 = *, x_4 = *, x_5 = 1, x_6 = *\}$$



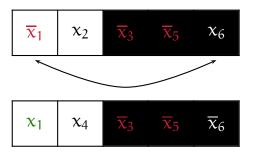
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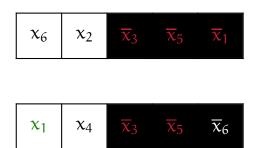
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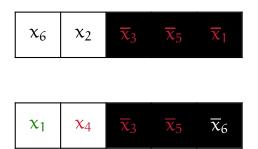
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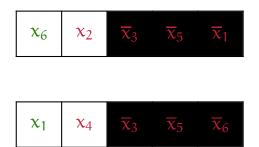
$$\phi = \{x_1 = 1, x_2 = *, x_3 = 1, x_4 = 0, x_5 = 1, x_6 = *\}$$



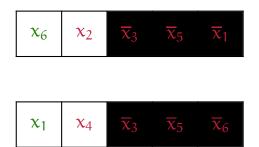
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Only examine (get in the cache) a clause when both

- a watch pointer gets falsified
- the other one is not satisfied

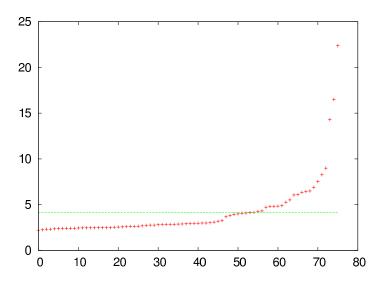
While backjumping, just unassign variables

Conflict clauses  $\rightarrow$  watch pointers

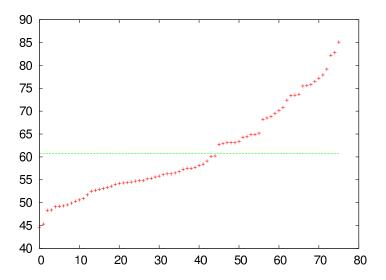
No detailed information available

Not used for binary clauses

# Average Number Clauses Visited Per Propagation



## Percentage visited clauses with other watched literal true



Clause Learning

Data-structures

### Heuristics

Clause Management

Conflict-Clause Minimization

Recent Advances and Conclusions

## Most important CDCL heuristics

#### Variable selection heuristics

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### Restart strategies

- aim: avoid heavy-tail behavior [GomesSelmanCrato'97]
- plus: focus search on recent conflicts when combined with dynamic heuristics

#### Variable selection heuristics

## Based on the occurrences in the (reduced) formula

- examples: Jeroslow-Wang, Maximal Occurrence in clauses of Minimal Size (MOMS), look-aheads
- not practical for CDCL solver due to watch pointers

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# Variable State Independent Decaying Sum (VSIDS)

- original idea (zChaff): for each conflict, increase the score of involved variables by 1, half all scores each 256 conflicts [MoskewiczMZZM'01]
- improvement (MiniSAT): for each conflict, increase the score of involved variables by  $\delta$  and increase  $\delta := 1.05\delta$ [EenSörensson'03]

### Visualization of VSIDS in PicoSAT

http://www.youtube.com/watch?v=MOjhFywLre8

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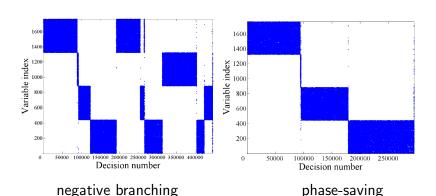
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# Based on the last implied value (phase-saving)

- introduced to CDCL [PipatsrisawatDarwiche'07]
- already used in local search [HirschKojevnikov'01]

Selecting the last implied value remembers solved components



#### Restarts

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- Luby sequence: e.g. 100, 100, 200, 100, 100, 200, 400, . . .

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# Rapid restarts by reusing trail: [vanderTakHeuleRamos'11]

- Partial restart same effect as full restart
- Optimal strategy Luby-1: 1, 1, 2, 1, 1, 2, 4, . . .

# Heuristics: SAT vs UNSAT [Oh'15]

The best heuristics choices depend on satisfiability: E.g.

- Restart frequently for UNSAT instances to get conflict early
- Restart sporadically for SAT instances to keep "progress"

Also, keeping learned clauses is less important on SAT instances and can actually slow down the search.

State-of-the-art CDCL solvers, such as CaDiCaL, have separate modes for SAT and UNSAT and they alternate between them.

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Conflict-Clause Minimization

Recent Advances and Conclusions

# Clause delection [EenSörensson'03, AudemardSimon'09]

Conflict clauses can significantly slow down CDCL solvers:

- Conflict clauses can quickly outnumber the original clauses
- Conflict clauses consists of important variables

Clause deletion is used to reduce the overhead:

- When the learned clause reach a limit, remove half
- Increase limit after every removal (completeness)

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#### Clause deletion heuristics:

- length of the clause
- relevance of the clause (when was it used in Analyze)
- the number of involved decision levels

Clause Learning

Data-structures

Heuristics

Clause Management

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### Self-Subsumption

Use self-subsumption to shorten conflict clauses

$$\frac{C \vee l \quad D \vee \overline{l}}{D} \quad C \subseteq D \qquad \frac{(a \vee b \vee l) \quad (a \vee b \vee c \vee \overline{l})}{(a \vee b \vee c)}$$

Conflict clause minimization is an important optimization.

## Self-Subsumption

### Use self-subsumption to shorten conflict clauses

$$\frac{C \vee l \quad D \vee \overline{l}}{D} \quad C \subseteq D \qquad \frac{(\alpha \vee b \vee l) \quad (\alpha \vee b \vee c \vee \overline{l})}{(\alpha \vee b \vee c)}$$

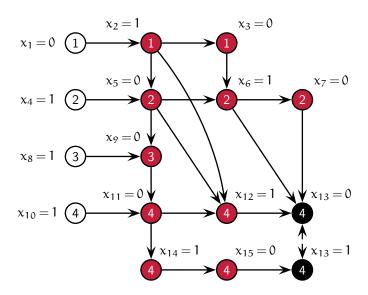
Conflict clause minimization is an important optimization.

# Use implication chains to further minimization:

$$\dots (\overline{a} \vee b)(\overline{b} \vee c)(\underline{a} \vee c \vee d) \dots \Rightarrow \\ \dots (\overline{a} \vee b)(\overline{b} \vee c)(c \vee d) \dots$$

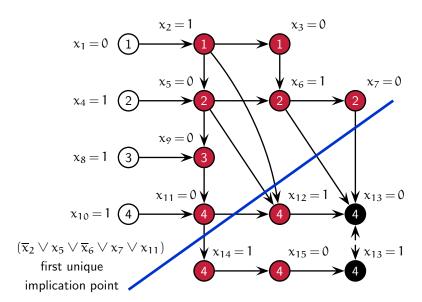
### Conflict-clause minimization

# [SörenssonBiere'09]



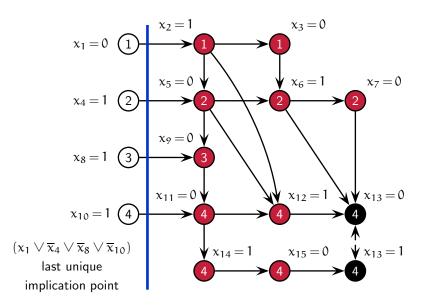
### Conflict-clause minimization

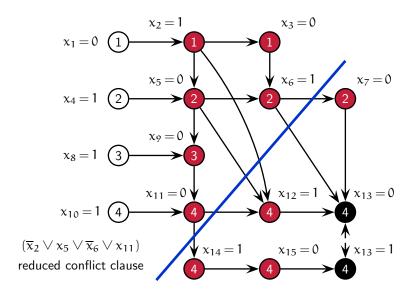
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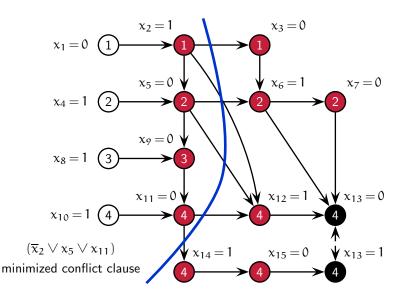


### Conflict-clause minimization

# [SörenssonBiere'09]







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#### Recent Advances

A new idea contributes to winning the competition.

Winner 2017: Clause vivification during search [LuoLiXiaoManyáLü'17]

Winner 2018: Chronological backtracking [NadelRyvchin'18]

Winner 2019: Multiple learnt clauses per conflict [KochemazovZaikinKondratievSemenov'19]

Winner 2020: Back to C and "target phases" [BiereFleury'20]

### Conclusions: state-of-the-art CDCL solver

### Key contributions to CDCL solvers:

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concept of conflict clauses (grasp) [Marques-SilvaSakallah'96]
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■ restart strategies [GomesSC'97,LubySZ'93]

■ 2-watch pointers and VSIDS (zChaff) [MoskewiczMZZM'01]

■ efficient implementation (Minisat) [EenSörensson'03]

■ phase-saving (Rsat) [PipatsrisawatDarwiche'07]

■ conflict-clause minimization [SörenssonBiere'09]

■ SAT vs UNSAT [Oh'15]

+ Pre- and in-processing techniques