Conflict-Driven Clause Learning

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http://www.cs.cmu.edu/~mheule/15816-f20/ https://cmu.zoom.us/j/93095736668 Automated Reasoning and Satisfiability September 21, 2020

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}$$

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 - Addition conflict clauses
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- Complete local search (for a refutation)?
- 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

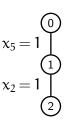
$$\begin{array}{c} (x_1 \vee x_4) \wedge \\ (x_3 \vee \overline{x}_4 \vee \overline{x}_5) \wedge \\ (\overline{x}_3 \vee \overline{x}_2 \vee \overline{x}_4) \wedge \\ \mathcal{F}_{\text{extra}} \end{array}$$



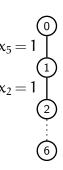
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$$x_5 = 1$$

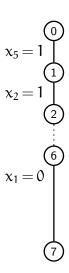
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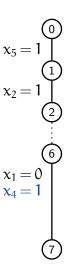
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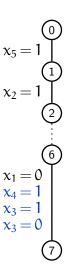
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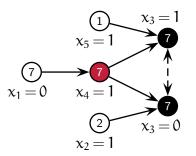
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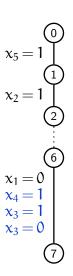


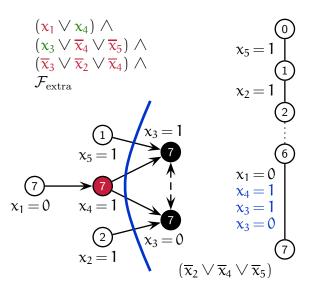
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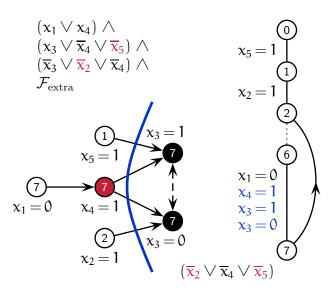


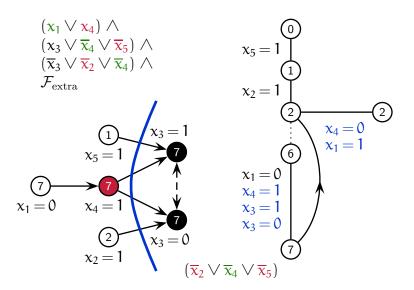


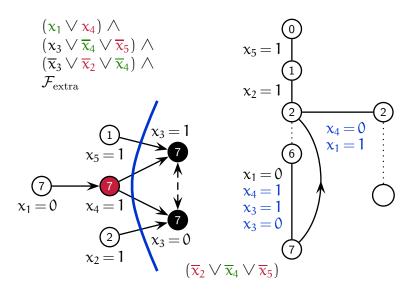












Implication graph [Marques-SilvaSakallah '96]

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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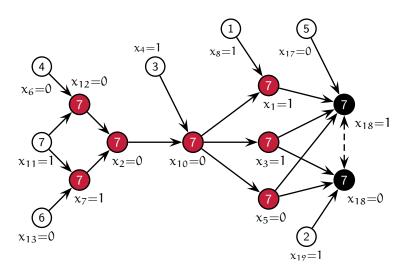
However, it has three weaknesses:

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

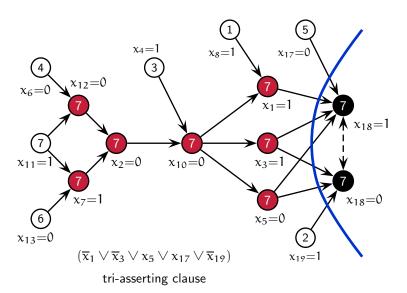
Conflict-driven Clause Learning: Pseudo-code

```
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
```

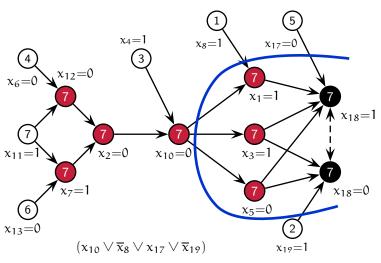
[Marques-SilvaSakallah'96]



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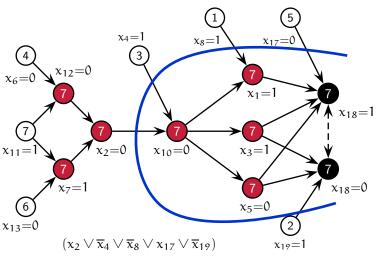


[Marques-SilvaSakallah'96]



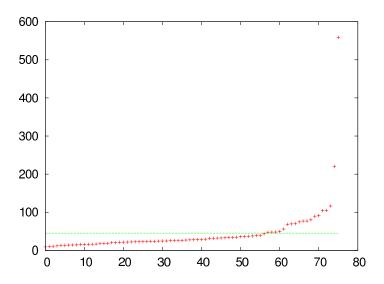
first unique implication point

[Marques-SilvaSakallah'96]



second unique implication point

Average Learned Clause Length



Clause Learning

Data-structures

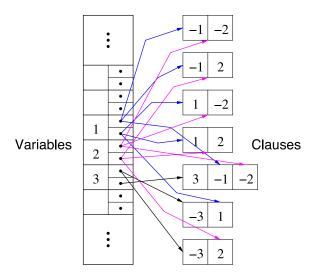
Heuristics

Clause Management

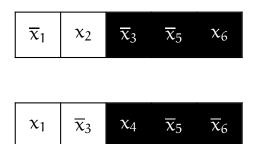
Conflict-Clause Minimization

Recent Advances and Conclusions

Simple data structure for unit propagation



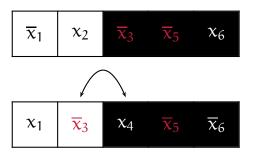
$$\phi = \{x_1 = *, x_2 = *, x_3 = *, x_4 = *, x_5 = *, x_6 = *\}$$



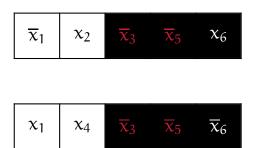
$$\varphi = \{x_1 = *, x_2 = *, x_3 = *, x_4 = *, x_5 = 1, x_6 = *\}$$



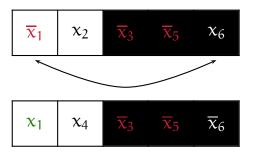
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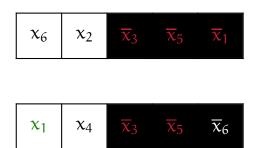
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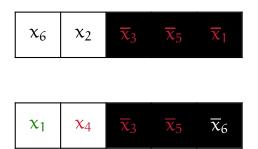
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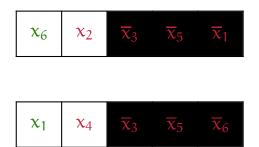
$$\phi = \{x_1 = 1, x_2 = *, x_3 = 1, x_4 = 0, x_5 = 1, x_6 = *\}$$



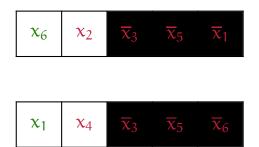
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Only examine (get in the cache) a clause when both

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

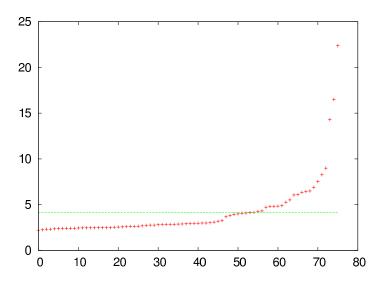
While backjumping, just unassign variables

Conflict clauses \rightarrow watch pointers

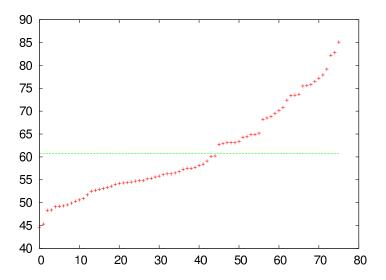
No detailed information available

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

- aim: minimize the search space
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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 δ and increase $\delta := 1.05\delta$ [EenSörensson'03]

Visualization of VSIDS in PicoSAT

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

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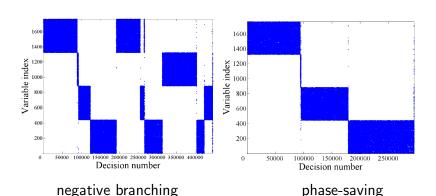
Based on the encoding / consequently

■ negative branching (early MiniSAT) [EenSörensson'03]

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

Restarts in CDCL solvers:

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- Geometrical restart: e.g. 100, 150, 225, 333, 500, 750, . . .
- 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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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)

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
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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

Conflict-Clause Minimization

Recent Advances and Conclusions

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

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

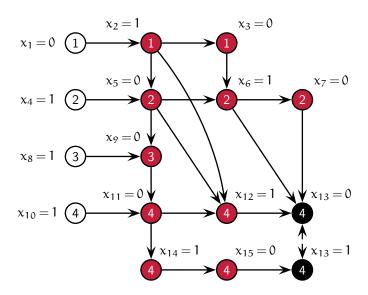
Conflict clause minimization is an important optimization.

Use implication chains to further minimization:

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

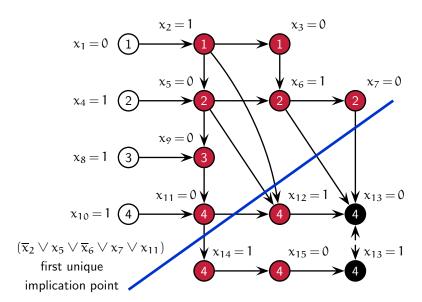
Conflict-clause minimization

[SörenssonBiere'09]



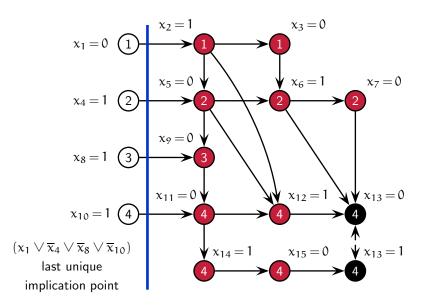
Conflict-clause minimization

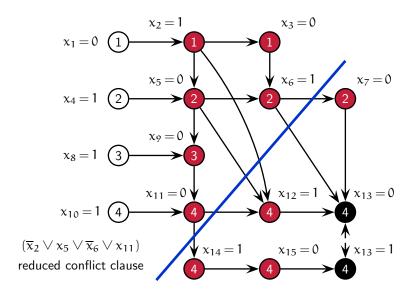
[SörenssonBiere'09]

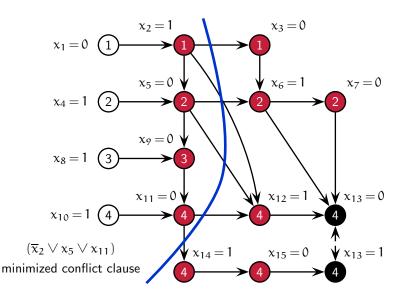


Conflict-clause minimization

[SörenssonBiere'09]







Clause Learning

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Recent Advances and Conclusions

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