

Towards Faster Conflict-Driven Pseudo-Boolean Solving

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Theory and Practice of Satisfiability Solving
Casa Matemática Oaxaca
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Some Acknowledgements and Caveats Right Away...

Survey heavily indebted to:

- Jan Elffers
- Stephan Gocht
- Daniel Le Berre
- João Marques-Silva
- and many, many others...

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By necessity, very selective—more material and references in:

- Dixon's thesis [Dix04]
- Pseudo-Boolean chapter [RM09] in *Handbook of Satisfiability*
- Full details of cited papers at end of slides (also online at www.csc.kth.se/~jakobn/research/TalkCM018.pdf)

The Satisfiability Problem (SAT)

$$(u \vee x \vee y) \wedge (x \vee \bar{y} \vee z) \wedge (\bar{x} \vee z) \wedge (\bar{y} \vee \bar{z}) \wedge (\bar{x} \vee \bar{z}) \wedge (\bar{u} \vee w) \wedge (\bar{u} \vee \bar{w})$$

- Variables should be set to true (= 1) or false (= 0)
- Constraint $(x \vee \bar{y} \vee z)$: means x or z should be true or y false
- \wedge means all constraints should hold simultaneously

Is there a truth value assignment satisfying all constraints?

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Can computers solve the SAT problem efficiently?

- Mentioned already in Gödel's famous letter in 1956 to von Neumann
- Intense research in TCS ever since early 1970s [Coo71, Lev73]
- Now one of Millennium Prize Problems in mathematics [Mil00]

SAT in Practice

- **Dramatic progress last 15–20 years** on SAT solvers using **conflict-driven clause learning (CDCL)** [MS96, BS97, MMZ⁺01]
- Today routinely used to solve large-scale real-world problems (100,000s or 1,000,000s of variables)
- But... There are also **small formulas** (just ~ 100 variables) that are **completely beyond reach** of even the very best SAT solvers
- Limitations of CDCL
 - 1 Clauses weak formalism for encoding constraints
 - 2 Method of reasoning used (resolution) also weak

Pseudo-Boolean Reasoning to the Rescue?

- Pseudo-Boolean (PB) linear constraints are stronger than clauses

Compare

$$x_1 + x_2 + x_3 + x_4 + x_5 + x_6 \geq 3$$

and

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

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- And pseudo-Boolean reasoning exponentially more powerful in theory
- But PB solvers less efficient than CDCL in practice(!?)

Outline

- 1 Conflict-Driven Clause Learning
 - CDCL by Example
 - Pseudocode and Analysis
- 2 Conflict-Driven Pseudo-Boolean Solving
 - Some Preliminaries
 - Pseudo-Boolean Solving Using Saturation
 - Pseudo-Boolean Solving Using Division
 - More About Pseudo-Boolean Reasoning
- 3 Open Problems and Future Directions

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Modern SAT Solving

DPLL method [DP60, DLL62]

- Assign values to variables (in some smart way)
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Variable Assignments

Two kinds of assignments — illustrate on our example formula:

$$(u \vee x \vee y) \wedge (x \vee \bar{y} \vee z) \wedge (\bar{x} \vee z) \wedge (\bar{y} \vee \bar{z}) \wedge (\bar{x} \vee \bar{z}) \wedge (\bar{u} \vee w) \wedge (\bar{u} \vee \bar{w})$$

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Free choice to assign value to variable

Notation $w \stackrel{d}{=} 0$

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$$y \stackrel{u \vee x \vee y}{=} 1$$

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Conflict-Driven Clause Learning

Time to analyse this conflict!

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Could backtrack by flipping last decision

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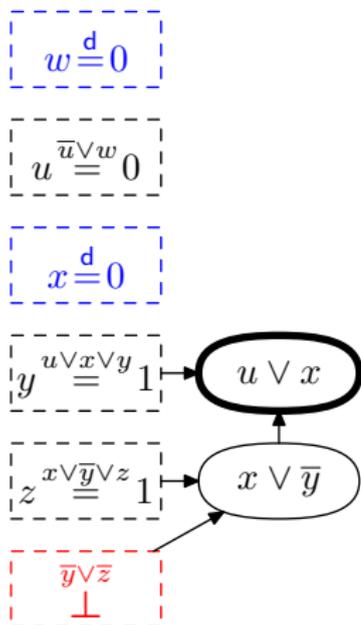
Case analysis over z for last two clauses:

- $x \vee \bar{y} \vee z$ wants $z = 1$
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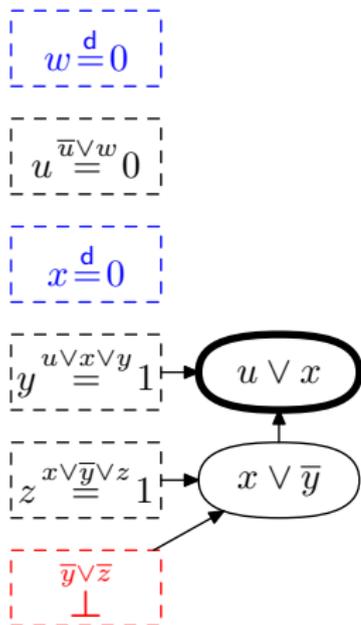
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Repeat until only 1 variable after last decision
— **learn** that clause (**1UIP**) and **backjump**

Complete Example of CDCL Execution

Backjump: roll back max #assignments so that last variable still flips

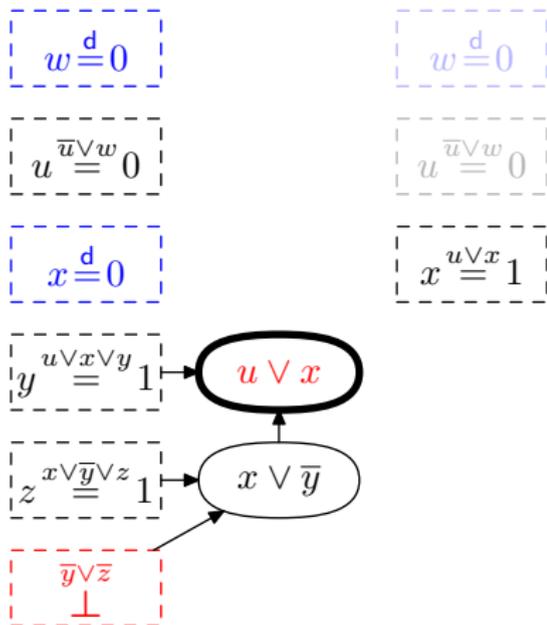
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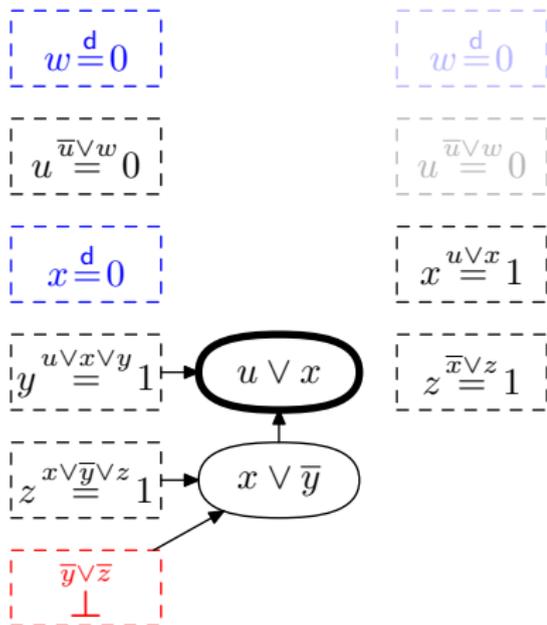
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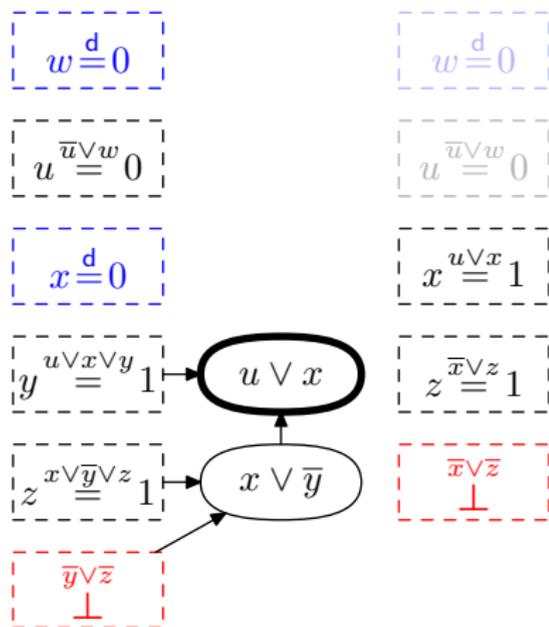
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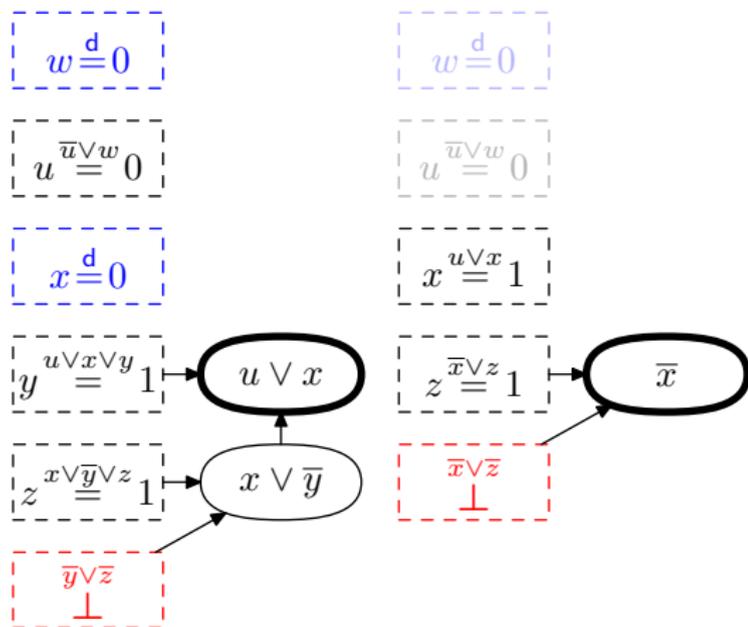
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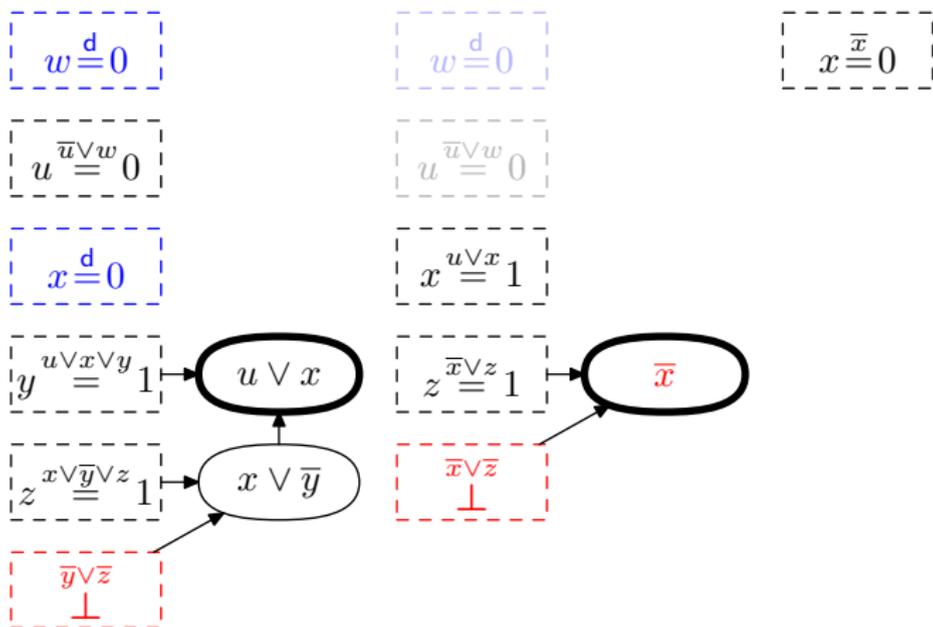
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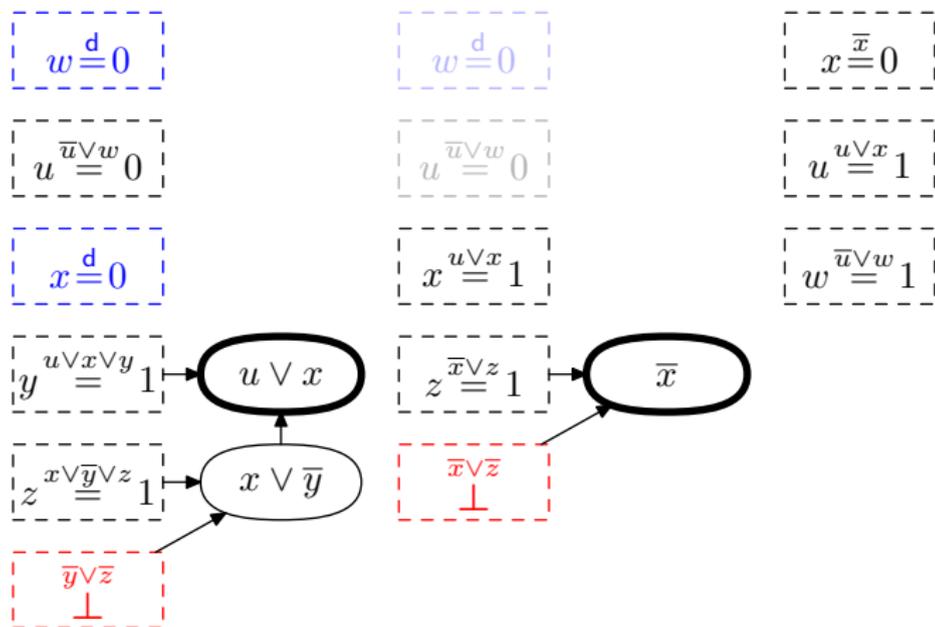
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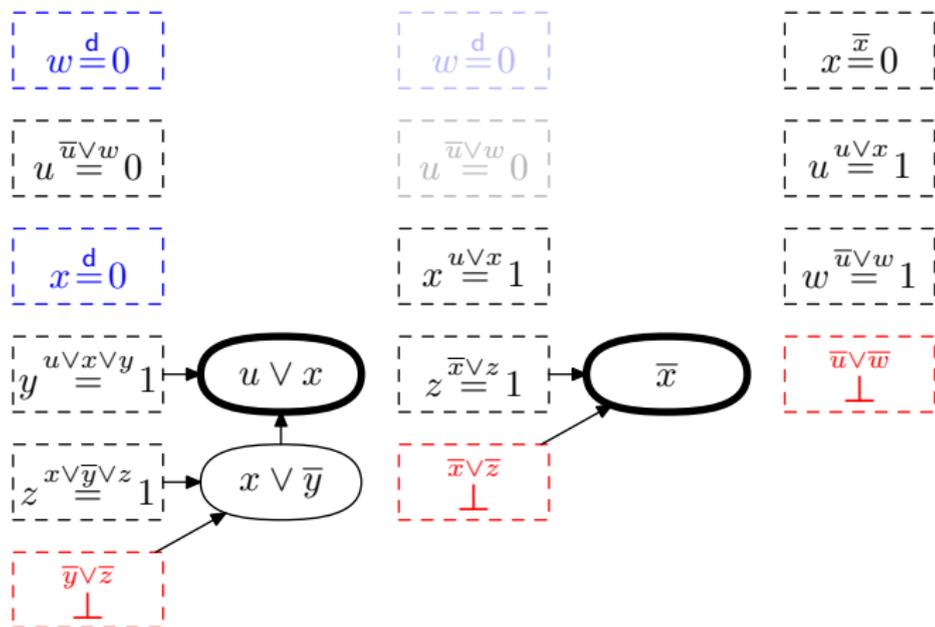
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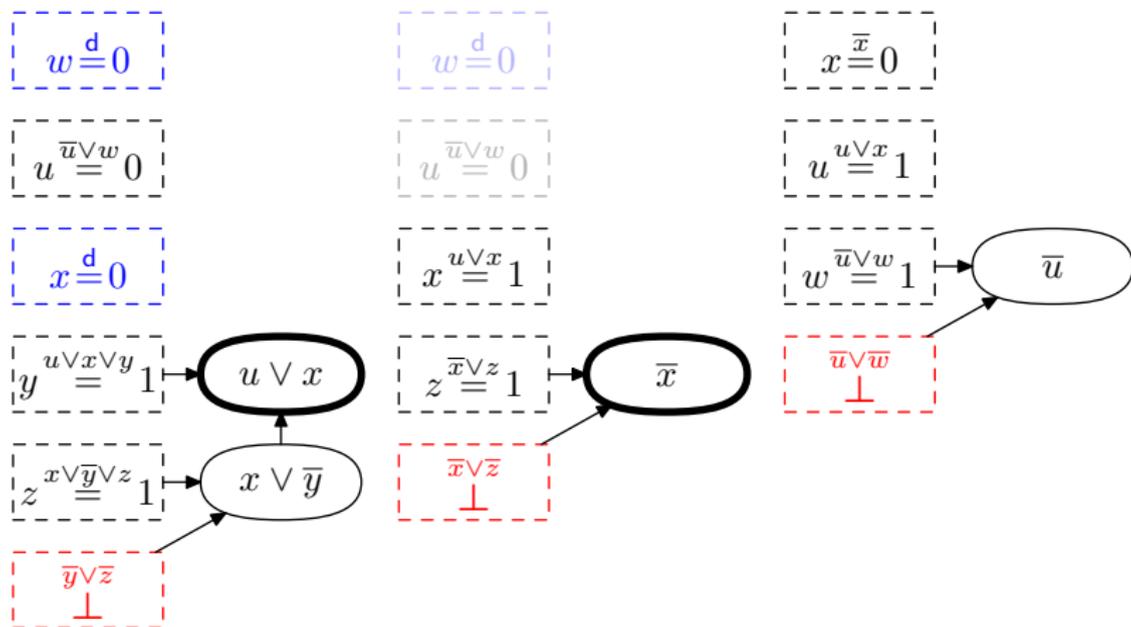
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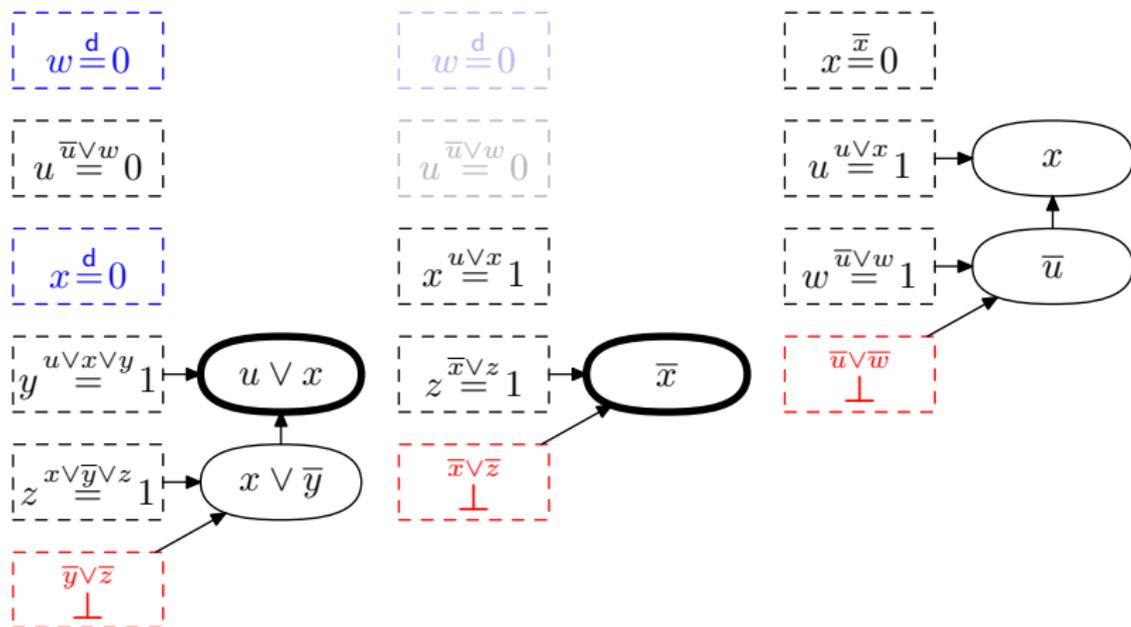
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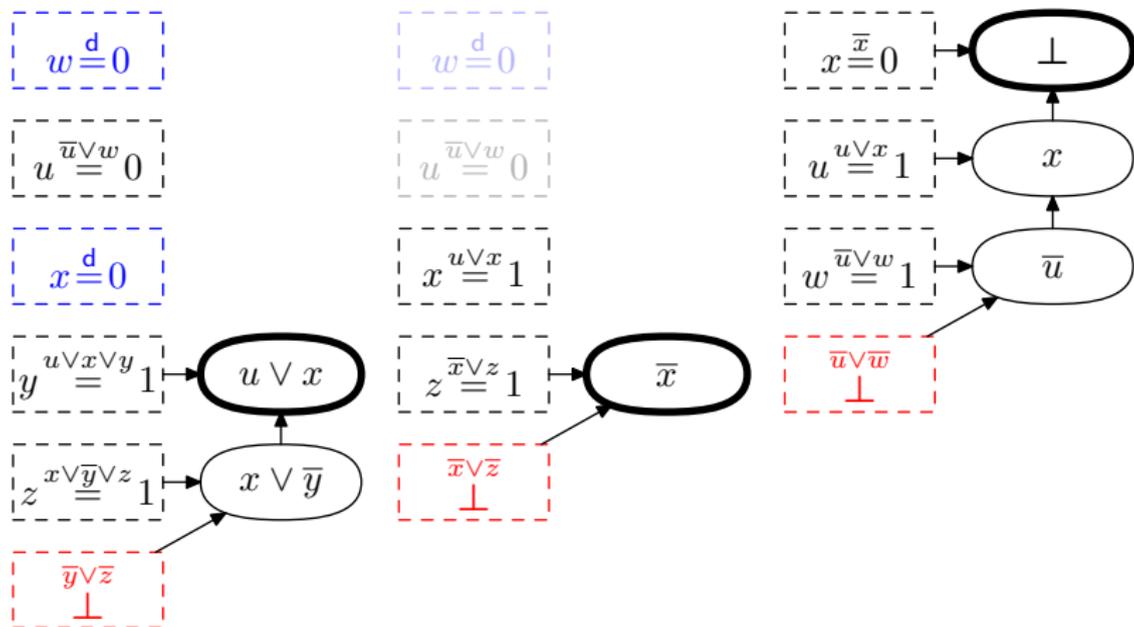
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CDCL Main Loop Pseudocode (High Level)

```

forever do
  if current assignment falsifies clause then
    apply learning scheme to derive new clause;
    if learned clause empty then output UNSATISFIABLE and exit;
    else
      | add learned clause and backjump
    end
  else if all variables assigned then output SATISFIABLE and exit;
  else if exists unit clause  $C$  propagating  $x$  to value  $b \in \{0, 1\}$  then
    | add propagated assignment  $x \stackrel{C}{=} b$ 
  else if time to restart then
    | remove all variable assignments
  else
    if time for clause database reduction then
      | erase (roughly) half of learned clauses in memory
    end
    use decision scheme to choose assignment  $x \stackrel{d}{=} b$ ;
  end
end

```

CDCL Main Loop Pseudocode (High Level)

```
forever do
  if current assignment falsifies clause then
    apply learning scheme to derive new clause;
    if learned clause empty then output UNSATISFIABLE and exit;
    else
      | add learned clause and backjump
    end
  else if all variables assigned then output SATISFIABLE and exit;
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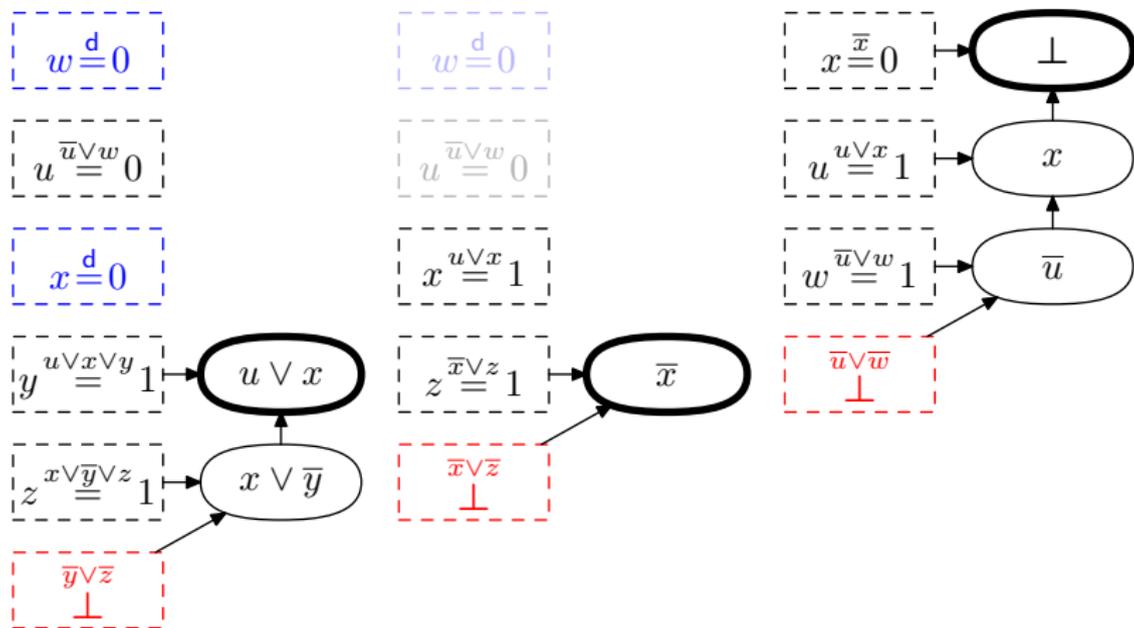
(*) Ignores preprocessing, but we don't have time to go into this

Resolution Proofs from CDCL Executions

Obtain resolution proof. . .

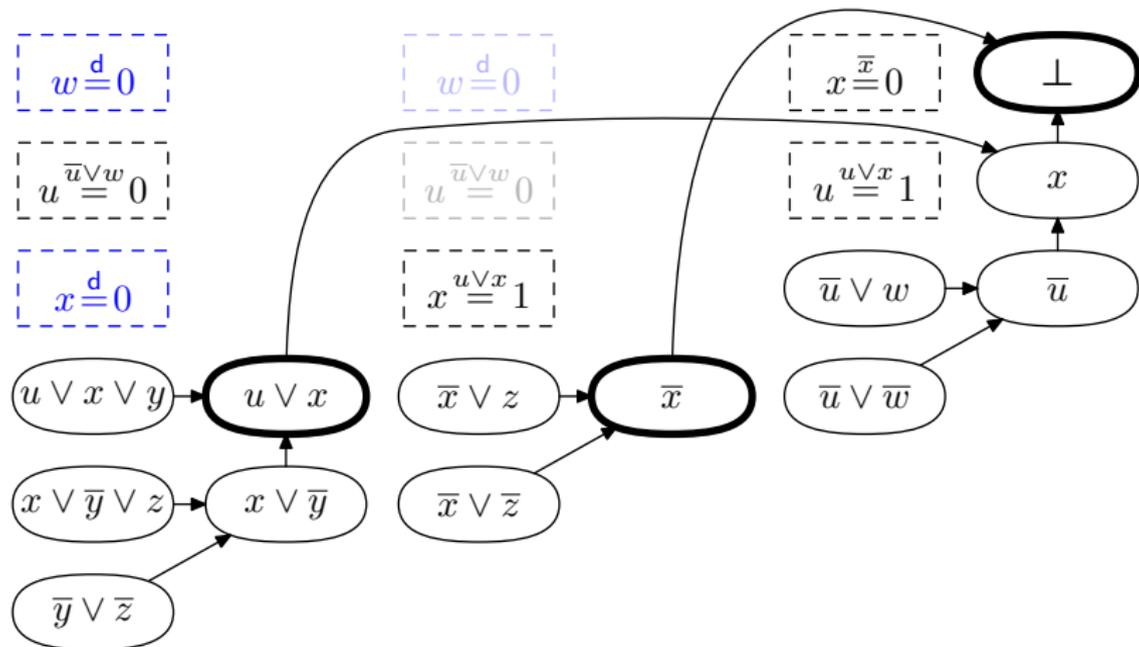
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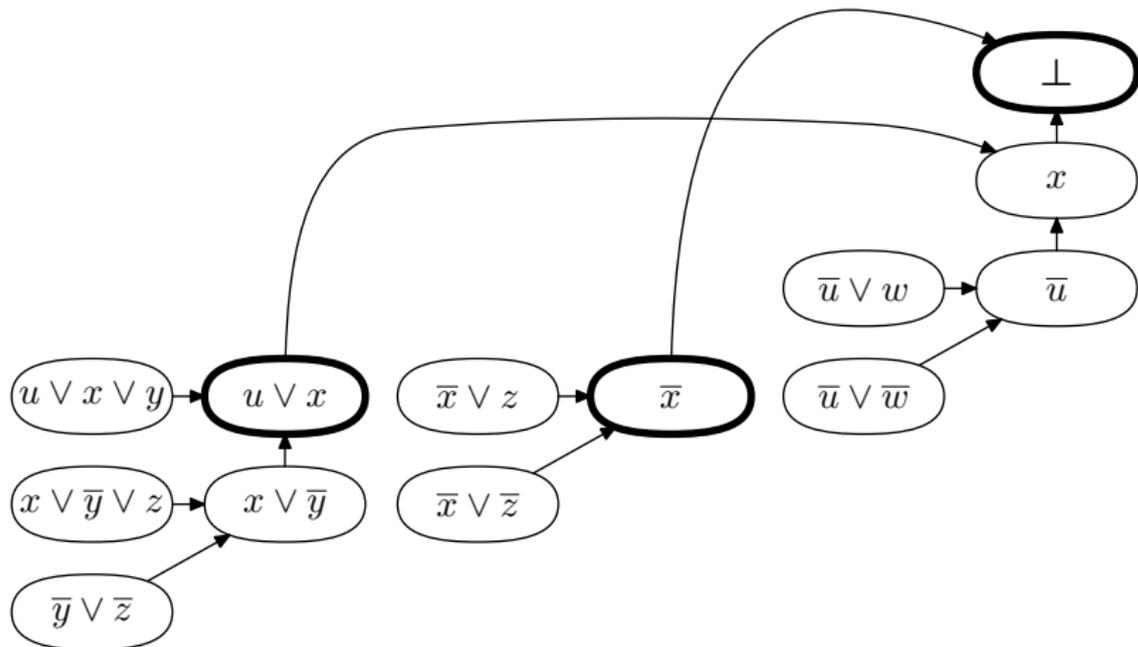
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Current state of affairs

- State-of-the-art CDCL solvers often perform amazingly well (“SAT is easy in practice”)
- Very poor theoretical understanding:
 - Why do heuristics work?
 - Why are applied instances easy?
- Paradox: resolution quite weak proof system; many strong lower bounds for “obvious” formulas, e.g., [Hak85, Urq87, BW01, MN14]
- Explore stronger reasoning methods (potential exponential speed-up)
- In particular, pseudo-Boolean solving (a.k.a. 0-1 integer programming) corresponding to cutting planes proof system
- Importantly, extends to pseudo-Boolean optimization (won’t talk about that—instead listen to Daniel Le Berre’s talk in the afternoon)

Pseudo-Boolean Constraints and Normalized Form

In this talk, “pseudo-Boolean” refers to 0-1 integer linear constraints

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- coefficients a_i : non-negative integers
- **degree (of falsity)** A : positive integer
- literals ℓ_i : x_i or \bar{x}_i (where $x_i + \bar{x}_i = 1$)

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(All constraints in what follows assumed to be implicitly normalized)

Some Types of Pseudo-Boolean Constraints

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Approaches to Pseudo-Boolean Solving

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Conflict-Driven Search in a Pseudo-Boolean Setting

Want to do “same thing” as CDCL but with linear constraints

- Variable assignments
 - 1 Always propagate forced assignment if possible
 - 2 Otherwise make assignment using decision heuristic
- At conflict
 - 1 Do conflict analysis to derive new constraint
 - 2 Add new constraint to instance
 - 3 Backjump by rolling back max #assignments so that variable flips

Propagation, Conflict, and Slack

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Note that constraint can be conflicting though not all variables assigned

Conflict Analysis Invariant

Look at our example CDCL conflict analysis again

$$(u \vee x \vee y) \wedge (x \vee \bar{y} \vee z) \wedge (\bar{x} \vee z) \wedge (\bar{y} \vee \bar{z}) \wedge (\bar{x} \vee \bar{z}) \wedge (\bar{u} \vee w) \wedge (\bar{u} \vee \bar{w})$$

$$w \stackrel{d}{=} 0$$

$$u \stackrel{\bar{u} \vee w}{=} 0$$

$$x \stackrel{d}{=} 0$$

$$y \stackrel{u \vee x \vee y}{=} 1$$

$$z \stackrel{x \vee \bar{y} \vee z}{=} 1$$

$$\bar{y} \vee \bar{z} \\ \perp$$

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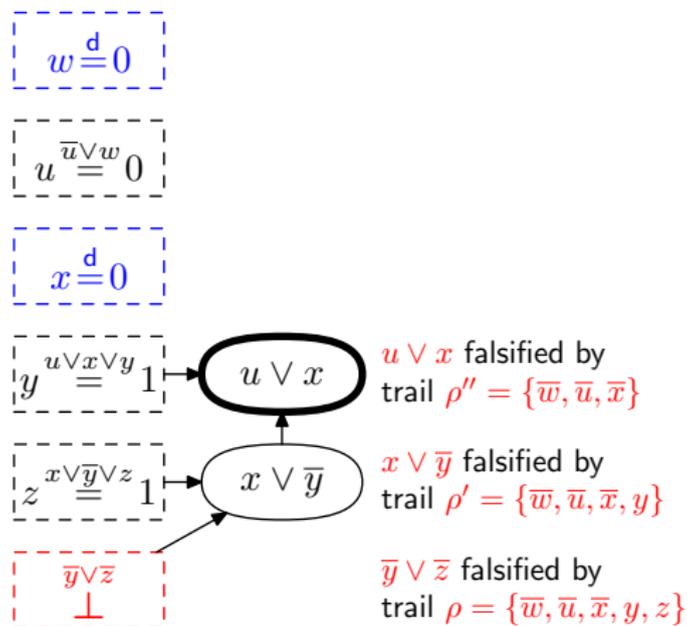
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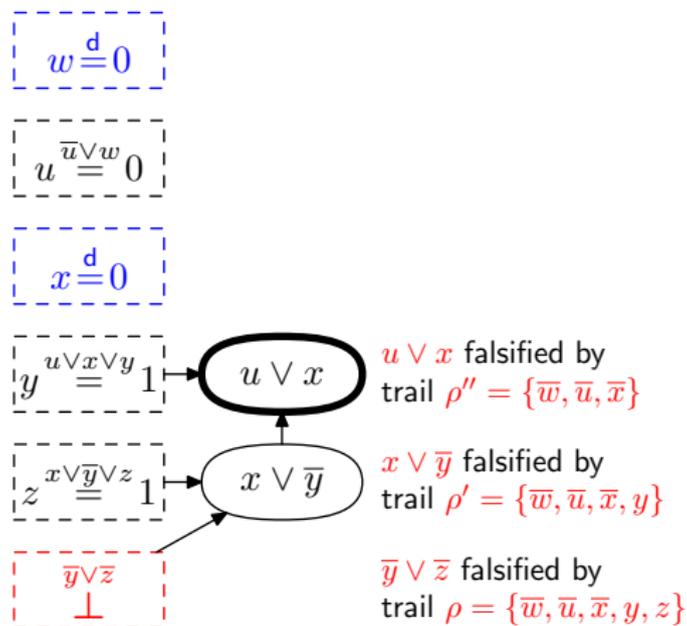
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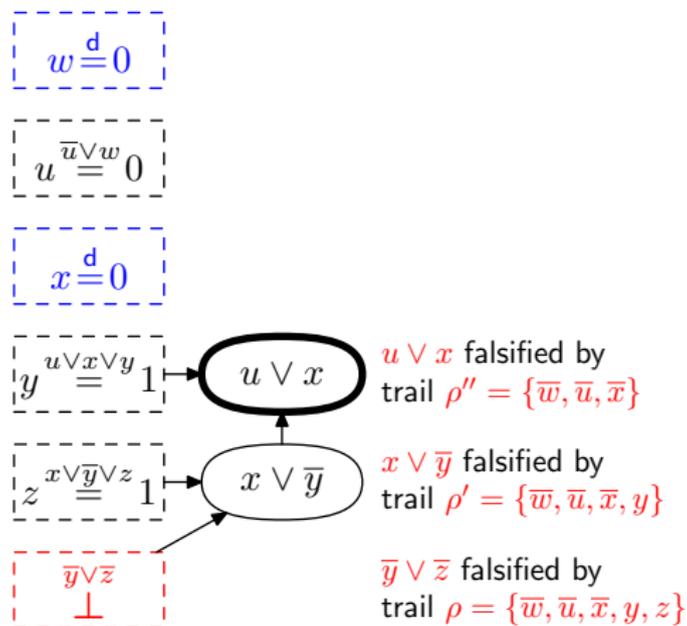
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Learn **asserting constraint**:
after backjump, some variable
guaranteed to flip

Generalized Resolution

Can mimic resolution step

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(Recall $z + \bar{z} = 1$)

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Generalized resolution rule [Hoo88, Hoo92]

Positive linear combination so that some variable cancels

$$\frac{a_1 x_1 + \sum_{i \geq 2} a_i l_i \geq A \quad b_1 \bar{x}_1 + \sum_{i \geq 2} b_i l_i \geq B}{\sum_{i \geq 2} \left(\frac{c}{a_1} a_i + \frac{c}{b_1} b_i \right) l_i \geq \frac{c}{a_1} A + \frac{c}{b_1} B - c} \quad [c = \text{lcm}(a_1, b_1)]$$

Saturation

Actually, don't get quite the right constraint in mimicking of resolution

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Saturation rule

$$\frac{\sum_i a_i \ell_i \geq A}{\sum_i \min\{a_i, A\} \cdot \ell_i \geq A}$$

Sound over integers, not over rationals (need such rules for SAT solving)

Analyze Conflict with Generalized Resolution + Saturation!

$$C_1 \doteq 2x_1 + 2x_2 + 2x_3 + x_4 \geq 4$$

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Trail $\rho = (x_1 \stackrel{d}{=} 0, x_2 \stackrel{C_1}{=} 1, x_3 \stackrel{C_1}{=} 1) \Rightarrow$ **Conflict with C_2**
(Note: same constraint can propagate several times!)

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- Non-negative slack w.r.t. $\rho' = (x_1 \stackrel{d}{=} 0, x_2 \stackrel{C_1}{=} 1)$ — **not conflicting!**

Analyze Conflict with Generalized Resolution + Saturation!

$$C_1 \doteq 2x_1 + 2x_2 + 2x_3 + x_4 \geq 4$$

$$C_2 \doteq 2\bar{x}_1 + 2\bar{x}_2 + 2\bar{x}_3 \geq 3$$

Trail $\rho = (x_1 \stackrel{d}{=} 0, x_2 \stackrel{C_1}{=} 1, x_3 \stackrel{C_1}{=} 1) \Rightarrow$ **Conflict with C_2**
 (Note: same constraint can propagate several times!)

- Resolve $\text{reason}(x_3, \rho) \doteq C_1$ with C_2 over x_3 to get $\text{resolve}(C_1, C_2, x_3)$

$$\frac{2x_1 + 2x_2 + 2x_3 + x_4 \geq 4 \quad 2\bar{x}_1 + 2\bar{x}_2 + 2\bar{x}_3 \geq 3}{x_4 \geq 1}$$

- Applying $\text{saturate}(x_4 \geq 1)$ does nothing
- Non-negative slack w.r.t. $\rho' = (x_1 \stackrel{d}{=} 0, x_2 \stackrel{C_1}{=} 1)$ — **not conflicting!**

Fix (non-obvious): Apply weakening to reason constraints

$$\text{weaken}(\sum_i a_i l_i \geq A, l_j) = \sum_{i \neq j} a_i l_i \geq A - a_j$$

Try to Reduce the Reason Constraint

$$C_1 \doteq 2x_1 + 2x_2 + 2x_3 + x_4 \geq 4$$

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Trail $\rho = (x_1 \stackrel{d}{=} 0, x_2 \stackrel{C_1}{=} 1, x_3 \stackrel{C_1}{=} 1) \Rightarrow$ Conflict with C_2

Let's try to

- 1 Weaken reason on non-falsified literal (but not last propagated)
- 2 Saturate weakened constraint
- 3 Resolve with conflicting constraint over propagated literal

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Let's try to

- ① Weaken reason on non-falsified literal (but not last propagated)
- ② Saturate weakened constraint
- ③ Resolve with conflicting constraint over propagated literal

$$\begin{array}{l}
 \text{weaken } x_2 \quad \frac{2x_1 + 2x_2 + 2x_3 + x_4 \geq 4}{2x_1 + 2x_3 + x_4 \geq 2} \\
 \text{saturate} \quad \frac{2x_1 + 2x_3 + x_4 \geq 2}{2x_1 + 2x_3 + x_4 \geq 2} \quad \frac{2\bar{x}_1 + 2\bar{x}_2 + 2\bar{x}_3 \geq 3}{2\bar{x}_2 + x_4 \geq 1} \\
 \text{resolve } x_3 \quad \frac{2x_1 + 2x_3 + x_4 \geq 2}{2\bar{x}_2 + x_4 \geq 1}
 \end{array}$$

Try to Reduce the Reason Constraint

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 \text{resolve } x_3 \quad \frac{2x_1 + 2x_3 + x_4 \geq 2}{2\bar{x}_2 + x_4 \geq 1}
 \end{array}$$

Bummer! Still non-negative slack — not conflicting

Try Again to Reduce the Reason Constraint. . .

$$C_1 \doteq 2x_1 + 2x_2 + 2x_3 + x_4 \geq 4$$

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Trail $\rho = (x_1 \stackrel{d}{=} 0, x_2 \stackrel{C_1}{=} 1, x_3 \stackrel{C_1}{=} 1) \Rightarrow$ Conflict with C_2

$$\begin{array}{l}
 \text{weaken } \{x_2, x_4\} \frac{2x_1 + 2x_2 + 2x_3 + x_4 \geq 4}{2x_1 + 2x_3 \geq 1} \\
 \text{saturnate} \\
 \text{resolve } x_3 \frac{x_1 + x_3 \geq 1}{2\bar{x}_2 \geq 1} \frac{2\bar{x}_1 + 2\bar{x}_2 + 2\bar{x}_3 \geq 3}{2\bar{x}_2 \geq 1}
 \end{array}$$

Try Again to Reduce the Reason Constraint. . .

$$C_1 \doteq 2x_1 + 2x_2 + 2x_3 + x_4 \geq 4$$

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Negative slack — conflicting! Saturate and resolve with reason for x_2

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Negative slack — conflicting! Saturate and resolve with reason for x_2

$$\text{resolve } x_2 \frac{2x_1 + 2x_2 + 2x_3 + x_4 \geq 4 \quad \frac{2\bar{x}_2 \geq 1}{\bar{x}_2 \geq 1} \text{ saturate}}{2x_1 + 2x_3 + x_4 \geq 4}$$

Try Again to Reduce the Reason Constraint. . .

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Asserting! Backjump propagates to conflict without decisions \Rightarrow **done**

Reason Reduction Using Saturation [CK05]

$\text{reduceSat}(C_{\text{confl}}, C_{\text{reason}}, \ell, \rho)$

```
while  $\text{slack}(\text{resolve}(C_{\text{confl}}, C_{\text{reason}}, \ell); \rho) \geq 0$  do  
  |  $\ell' \leftarrow$  literal in  $C_{\text{reason}} \setminus \{\ell\}$  not falsified by  $\rho$ ;  
  |  $C_{\text{reason}} \leftarrow \text{saturate}(\text{weaken}(C_{\text{reason}}, \ell'))$ ;  
end  
return  $C_{\text{reason}}$ ;
```

Reason Reduction Using Saturation [CK05]

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  |  $\ell'$   $\leftarrow$  literal in  $C_{\text{reason}} \setminus \{\ell\}$  not falsified by  $\rho$ ;
  |  $C_{\text{reason}} \leftarrow$  saturate(weaken( $C_{\text{reason}}$ ,  $\ell'$ ));
end
return  $C_{\text{reason}}$ ;
```

Why does this work?

- Slack is **subadditive**

$$\text{slack}(c \cdot C + d \cdot D; \rho) \leq c \cdot \text{slack}(C; \rho) + d \cdot \text{slack}(D; \rho)$$

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- **Weakening** leaves $\text{slack}(C_{\text{reason}}; \rho)$ unchanged

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  |  $C_{\text{reason}} \leftarrow$  saturate(weaken( $C_{\text{reason}}$ ,  $\ell'$ ));
end
return  $C_{\text{reason}}$ ;
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Why does this work?

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$$\text{slack}(c \cdot C + d \cdot D; \rho) \leq c \cdot \text{slack}(C; \rho) + d \cdot \text{slack}(D; \rho)$$

- By invariant have $\text{slack}(C_{\text{confl}}; \rho) < 0$
- **Weakening** leaves $\text{slack}(C_{\text{reason}}; \rho)$ **unchanged**
- **Saturation** **decreases slack** — reach 0 when max #literals weakened

Pseudo-Boolean Conflict Analysis

$\text{analyzePBconflict}(C_{\text{confl}}, \rho)$

while C_{confl} *not asserting* **do**

$\ell \leftarrow$ literal assigned last on trail ρ ;

if $\bar{\ell}$ *occurs in* C_{confl} **then**

$C_{\text{reason}} \leftarrow \text{reason}(\ell, \rho)$;

$C_{\text{reason}} \leftarrow \text{reduceSat}(C_{\text{reason}}, C_{\text{confl}}, \ell, \rho)$;

$C_{\text{confl}} \leftarrow \text{resolve}(C_{\text{confl}}, C_{\text{reason}}, \ell)$;

$C_{\text{confl}} \leftarrow \text{saturate}(C_{\text{confl}})$;

end

$\rho \leftarrow \text{removeLast}(\rho)$;

end

return C_{confl} ;

The need to reduce the reason is **new compared to CDCL**
Everything else is the same

Some Problems Compared to CDCL

- Compared to clauses **harder to detect propagation** for constraints like

$$\sum_{i=1}^n x_i \geq n - 1$$

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 - ⇒ **coefficient sizes can explode** (expensive arithmetic)

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- Compared to clauses **harder to detect propagation** for constraints like

$$\sum_{i=1}^n x_i \geq n - 1$$

- Generalized resolution for general pseudo-Boolean constraints
 - ⇒ lots of lcm computations
 - ⇒ **coefficient sizes can explode** (expensive arithmetic)
- For CNF inputs, **degenerates to resolution!**
 - ⇒ CDCL but with super-expensive data structures

The Cutting Planes Proof System

Cutting planes as defined in [CCT87] **doesn't use saturation** but instead **division** (a.k.a. **Chvátal-Gomory cut**)

$$\text{Literal axioms} \frac{}{l_i \geq 0}$$

$$\text{Linear combination} \frac{\sum_i a_i l_i \geq A \quad \sum_i b_i l_i \geq B}{\sum_i (c_A a_i + c_B b_i) l_i \geq c_A A + c_B B}$$

$$\text{Division} \frac{\sum_i a_i l_i \geq A}{\sum_i \lceil a_i / c \rceil l_i \geq \lceil A / c \rceil}$$

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- Cutting planes with division **implicationally complete**
- Cutting planes with **saturation** is **not** [VEG⁺18]
- Can division yield stronger conflict analysis?

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- Cutting planes with division **implicationally complete**
- Cutting planes with **saturation** is **not** [VEG⁺18]
- Can division yield stronger conflict analysis?
(Used for general integer linear programming in *CutSat* [JdM13])

Using Division to Reduce the Reason

$$C_1 \doteq 2x_1 + 2x_2 + 2x_3 + x_4 \geq 4$$

$$C_2 \doteq 2\bar{x}_1 + 2\bar{x}_2 + 2\bar{x}_3 \geq 3$$

Trail $\rho = (x_1 \stackrel{d}{=} 0, x_2 \stackrel{C_1}{=} 1, x_3 \stackrel{C_1}{=} 1) \Rightarrow$ Conflict with C_2

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- ① Weaken reason on non-falsified literal(s) with coefficient not divisible by propagating literal coefficient
- ② Divide weakened constraint by propagating literal coefficient
- ③ Resolve with conflicting constraint over propagated literal

Reason Reduction Using Division [EN18]

$\text{reduceDiv}(C_{\text{confl}}, C_{\text{reason}}, \ell, \rho)$

$c \leftarrow \text{coeff}(C_{\text{reason}}, \ell);$

while $\text{slack}(\text{resolve}(C_{\text{confl}}, \text{divide}(C_{\text{reason}}, c), \ell); \rho) \geq 0$ **do**

$l_j \leftarrow$ literal in $C_{\text{reason}} \setminus \{\ell\}$ such that $\bar{l}_j \notin \rho$ and $c \nmid \text{coeff}(C, l_j);$
 $C_{\text{reason}} \leftarrow \text{weaken}(C_{\text{reason}}, l_j);$

end

return $\text{divide}(C_{\text{reason}}, c);$

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end

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So now why does **this** work?

- Sufficient to get **reason with slack 0** since
 - ① $\text{slack}(C_{\text{confl}}; \rho) < 0$
 - ② slack is subadditive

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reduceDiv( $C_{\text{confl}}$ ,  $C_{\text{reason}}$ ,  $\ell$ ,  $\rho$ )
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```
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```

```
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```

```
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So now why does **this** work?

- Sufficient to get **reason with slack 0** since
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- Weakening doesn't change slack \Rightarrow always $0 \leq \text{slack}(C_{\text{reason}}; \rho) < c$

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So now why does **this** work?

- Sufficient to get **reason with slack 0** since
 - 1 $\text{slack}(C_{\text{confl}}; \rho) < 0$
 - 2 slack is subadditive
- Weakening doesn't change slack \Rightarrow always $0 \leq \text{slack}(C_{\text{reason}}; \rho) < c$
- After max #weakenings have $0 \leq \text{slack}(\text{divide}(C_{\text{reason}}, c); \rho) < 1$

Round-to-1 Reduction used in *RoundingSat*

Reduction method used in *RoundingSat* does max weakening right away

```
roundToOne( $C, l, \rho$ )
```

```

 $c \leftarrow \text{coeff}(C, l);$ 
foreach literal  $l_j$  in  $C$  do
  | if  $\bar{l}_j \notin \rho$  and  $c \nmid \text{coeff}(C, l_j)$  then
  |   |  $C \leftarrow \text{weaken}(C, l_j);$ 
  |   end
end
return divide( $C, c$ );

```

And roundToOne used more aggressively in conflict analysis

RoundingSat Conflict Analysis

analyzePBconflict(C_{confl}, ρ)

while C_{confl} contains no or multiple falsified literals on last level **do**

if no current solver decisions **then**

 | output UNSATISFIABLE and terminate

end

$\ell \leftarrow$ literal assigned last on trail ρ ;

if $\bar{\ell}$ occurs in C_{confl} **then**

 | $C_{\text{confl}} \leftarrow$ roundToOne($C_{\text{confl}}, \bar{\ell}, \rho$);

 | $C_{\text{reason}} \leftarrow$ roundToOne(reason(ℓ, ρ), ℓ, ρ);

 | $C_{\text{confl}} \leftarrow$ resolve($C_{\text{confl}}, C_{\text{reason}}, \ell$);

end

$\rho \leftarrow$ removeLast(ρ);

end

$\ell \leftarrow$ literal in C_{confl} last falsified by ρ ;

return roundToOne($C_{\text{confl}}, \ell, \rho$);

Division vs. Saturation

- Higher conflict speed when PB reasoning doesn't help [EN18]
- Seems to perform better when PB reasoning crucial [EGNV18]
- Keeps coefficients small — can do fixed-precision integer arithmetic
- But still equally hard to detect propagation
- And still degenerates to resolution for CNF inputs

Other PB Rules I: Cardinality Constraint Reduction

Given PB constraint

$$3x_1 + 2x_2 + x_3 + x_4 \geq 4$$

can compute least #literals that have to be true

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Galena [CK05] only learns cardinality constraints — easier to deal with

Cardinality constraint reduction rule

$$\frac{\sum_i a_i l_i \geq A}{\sum_{i: a_i > 0} l_i \geq T} \quad T = \min\{|I| : I \subseteq [n], \sum_{i \in I} a_i \geq A\}$$

Other PB Rules II: Strengthening

Strengthening by example:

- Set $x = 0$ and propagate on constraints

$$x + y \geq 1 \quad x + z \geq 1 \quad y + z \geq 1$$

Other PB Rules II: Strengthening

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Strengthening rule (imported by [DG02] from operations research)

- Suppose $\ell = 0 \Rightarrow \sum_i a_i \ell_i \geq A$ oversatisfied by amount K
- Then can deduce $K\ell + \sum_i a_i \ell_i \geq A + K$

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In theory, can recover from bad encodings (e.g., CNF)

In practice, seems inefficient and hard to get to work...

Other PB Rules III: “Fusion Resolution”

Suppose have constraints

$$2x + 3y + 2z + w \geq 3 \quad 2\bar{x} + 3y + 2z + w \geq 3$$

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Then by eyeballing can conclude

$$3y + 2z + w \geq 3$$

Other PB Rules III: “Fusion Resolution”

Suppose have constraints

$$2x + 3y + 2z + w \geq 3 \quad 2\bar{x} + 3y + 2z + w \geq 3$$

Then by eyeballing can conclude

$$3y + 2z + w \geq 3$$

But only get from resolution

$$6y + 4z + 2w \geq 4$$

Other PB Rules III: “Fusion Resolution”

Suppose have constraints

$$2x + 3y + 2z + w \geq 3 \quad 2\bar{x} + 3y + 2z + w \geq 3$$

Then by eyeballing can conclude

$$3y + 2z + w \geq 3$$

But only get from resolution + saturation

$$4y + 4z + 2w \geq 4$$

Other PB Rules III: “Fusion Resolution”

Suppose have constraints

$$2x + 3y + 2z + w \geq 3 \quad 2\bar{x} + 3y + 2z + w \geq 3$$

Then by eyeballing can conclude

$$3y + 2z + w \geq 3$$

But only get from resolution + saturation + division

$$2y + 2z + w \geq 2$$

Other PB Rules III: “Fusion Resolution”

Suppose have constraints

$$2x + 3y + 2z + w \geq 3 \quad 2\bar{x} + 3y + 2z + w \geq 3$$

Then by eyeballing can conclude

$$3y + 2z + w \geq 3$$

But only get from resolution + saturation + division

$$2y + 2z + w \geq 2$$

“**Fusion resolution**” [Goc17]

$$\frac{a\bar{l} + \sum_i b_i l_i \geq B \quad a\bar{l} + \sum_i b_i l_i \geq B'}{\sum_i b_i l_i \geq \min\{B, B'\}}$$

No obvious way for cutting planes to immediately derive this

Shows up in some tricky benchmarks in [EGNV18]

Open Problems I: Some Implementation Challenges

- 1 Degrees of freedom in **PB conflict analysis**
 - Skip resolution steps when slack very negative?
 - How much to weaken?
 - Learn general PB constraints or more limited form?
- 2 **Efficient propagation detection** for PB constraints
- 3 Assessment of **quality of learned constraints**
- 4 **Distance to backjump?** (Constraint can be asserting at several levels)

Open Problems II: Some PB Reasoning Challenges

- 1 **Better conflict analysis** (also for CDCL)
Is trivial resolution optimal, or can it pay to be smarter?
- 2 Natural way to **recover from bad encodings** (e.g., CNF)
- 3 Efficient and concise **PB proof logging**
- 4 **Theoretical potential and limitations** poorly understood [VEG⁺18]
 - Separations of subsystems of cutting planes?
 - In particular, is division strictly stronger than saturation?

Hear more from Marc Vinyals this afternoon

Open Problems III: Beyond PB Reasoning

- Sometimes very poor performance even on LPs that are rationally infeasible! (And trivial for mixed integer linear programming solvers)
- But sometimes MIP solvers lost when learning from PB constraints crucial (and when conflict-driven PB solvers shine)
- Borrow techniques from (or merge with) MIP?
Stay tuned for Ambros Gleixner's talk. . .

Summing up

- Conflict-driven search hugely successful SAT solving paradigm
- This talk: Survey how to port from CDCL to PB constraints
- Potential exponential performance gains haven't materialized so far
- Instead highly nontrivial challenges regarding
 - Efficient implementation
 - Theoretical understanding
- But no obvious reason why efficient PB solvers should not be possible (remember CDCL took 50 years)
- And in any case lots of fun questions to work on! 😊

Summing up

- Conflict-driven search hugely successful SAT solving paradigm
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Thank you for your attention!

References I

- [BS97] Roberto J. Bayardo Jr. and Robert Schrag. Using CSP look-back techniques to solve real-world SAT instances. In *Proceedings of the 14th National Conference on Artificial Intelligence (AAAI '97)*, pages 203–208, July 1997.
- [BW01] Eli Ben-Sasson and Avi Wigderson. Short proofs are narrow—resolution made simple. *Journal of the ACM*, 48(2):149–169, March 2001. Preliminary version in *STOC '99*.
- [CCT87] William Cook, Collette Rene Coullard, and György Turán. On the complexity of cutting-plane proofs. *Discrete Applied Mathematics*, 18(1):25–38, November 1987.
- [CK05] Donald Chai and Andreas Kuehlmann. A fast pseudo-Boolean constraint solver. *IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems*, 24(3):305–317, March 2005. Preliminary version in *DAC '03*.
- [Coo71] Stephen A. Cook. The complexity of theorem-proving procedures. In *Proceedings of the 3rd Annual ACM Symposium on Theory of Computing (STOC '71)*, pages 151–158, 1971.
- [DG02] Heidi E. Dixon and Matthew L. Ginsberg. Inference methods for a pseudo-Boolean satisfiability solver. In *Proceedings of the 18th National Conference on Artificial Intelligence (AAAI '02)*, pages 635–640, July 2002.

References II

- [Dix04] Heidi E. Dixon. *Automating Psuedo-Boolean Inference within a DPLL Framework*. PhD thesis, University of Oregon, 2004. Available at <http://www.cirl.uoregon.edu/dixon/papers/dixonDissertation.pdf>.
- [DLL62] Martin Davis, George Logemann, and Donald Loveland. A machine program for theorem proving. *Communications of the ACM*, 5(7):394–397, July 1962.
- [DP60] Martin Davis and Hilary Putnam. A computing procedure for quantification theory. *Journal of the ACM*, 7(3):201–215, 1960.
- [EGNV18] Jan Elffers, Jesús Giráldez-Cru, Jakob Nordström, and Marc Vinyals. Using combinatorial benchmarks to probe the reasoning power of pseudo-Boolean solvers. In *Proceedings of the 21st International Conference on Theory and Applications of Satisfiability Testing (SAT '18)*, volume 10929 of *Lecture Notes in Computer Science*, pages 75–93. Springer, July 2018.
- [EN18] Jan Elffers and Jakob Nordström. Divide and conquer: Towards faster pseudo-Boolean solving. In *Proceedings of the 27th International Joint Conference on Artificial Intelligence (IJCAI '18)*, pages 1291–1299, July 2018.
- [ES06] Niklas Eén and Niklas Sörensson. Translating pseudo-Boolean constraints into SAT. *Journal on Satisfiability, Boolean Modeling and Computation*, 2(1-4):1–26, 2006.

References III

- [Goc17] Stephan Gocht. Personal communication, 2017.
- [Hak85] Armin Haken. The intractability of resolution. *Theoretical Computer Science*, 39(2-3):297–308, August 1985.
- [Hoo88] John N. Hooker. Generalized resolution and cutting planes. *Annals of Operations Research*, 12(1):217–239, 1988.
- [Hoo92] John N. Hooker. Generalized resolution for 0-1 linear inequalities. *Annals of Mathematics and Artificial Intelligence*, 6(1):271–286, 1992.
- [JdM13] Dejan Jovanovic and Leonardo de Moura. Cutting to the chase solving linear integer arithmetic. *Journal of Automated Reasoning*, 51(1):79–108, June 2013. Preliminary version in *CADE-23* 2011.
- [Lev73] Leonid A. Levin. Universal sequential search problems. *Problemy peredachi informatsii*, 9(3):115–116, 1973. In Russian. Available at <http://mi.mathnet.ru/ppi914>.
- [LP10] Daniel Le Berre and Anne Parrain. The Sat4j library, release 2.2. *Journal on Satisfiability, Boolean Modeling and Computation*, 7:59–64, 2010.

References IV

- [Mil00] The Millennium Problems of the Clay Mathematics Institute, May 2000. See <http://www.claymath.org/millennium-problems>.
- [MML14] Ruben Martins, Vasco M. Manquinho, and Inês Lynce. Open-WBO: A modular MaxSAT solver. In *Proceedings of the 17th International Conference on Theory and Applications of Satisfiability Testing (SAT '14)*, volume 8561 of *Lecture Notes in Computer Science*, pages 438–445. Springer, July 2014.
- [MMZ⁺01] Matthew W. Moskewicz, Conor F. Madigan, Ying Zhao, Lintao Zhang, and Sharad Malik. Chaff: Engineering an efficient SAT solver. In *Proceedings of the 38th Design Automation Conference (DAC '01)*, pages 530–535, June 2001.
- [MN14] Mladen Mikša and Jakob Nordström. Long proofs of (seemingly) simple formulas. In *Proceedings of the 17th International Conference on Theory and Applications of Satisfiability Testing (SAT '14)*, volume 8561 of *Lecture Notes in Computer Science*, pages 121–137. Springer, July 2014.
- [MS96] João P. Marques-Silva and Karem A. Sakallah. GRASP—a new search algorithm for satisfiability. In *Proceedings of the IEEE/ACM International Conference on Computer-Aided Design (ICCAD '96)*, pages 220–227, November 1996.

References V

- [RM09] Olivier Roussel and Vasco M. Manquinho. Pseudo-Boolean and cardinality constraints. In Armin Biere, Marijn J. H. Heule, Hans van Maaren, and Toby Walsh, editors, *Handbook of Satisfiability*, volume 185 of *Frontiers in Artificial Intelligence and Applications*, chapter 22, pages 695–733. IOS Press, February 2009.
- [SN15] Masahiko Sakai and Hidetomo Nabeshima. Construction of an ROBDD for a PB-constraint in band form and related techniques for PB-solvers. *IEICE Transactions on Information and Systems*, 98-D(6):1121–1127, June 2015.
- [SS06] Hossein M. Sheini and Karem A. Sakallah. Pueblo: A hybrid pseudo-Boolean SAT solver. *Journal on Satisfiability, Boolean Modeling and Computation*, 2(1-4):165–189, March 2006. Preliminary version in *DATE '05*.
- [Urq87] Alasdair Urquhart. Hard examples for resolution. *Journal of the ACM*, 34(1):209–219, January 1987.
- [VEG⁺18] Marc Vinyals, Jan Elffers, Jesús Giráldez-Cru, Stephan Gocht, and Jakob Nordström. In between resolution and cutting planes: A study of proof systems for pseudo-Boolean SAT solving. In *Proceedings of the 21st International Conference on Theory and Applications of Satisfiability Testing (SAT '18)*, volume 10929 of *Lecture Notes in Computer Science*, pages 292–310. Springer, July 2018.