

CS498: Algorithmic Engineering

Lecture 18: How SAT Solvers Actually Work

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Week 10 – 03/26/2026

Outline

- 1 Recap and Motivation
- 2 DPLL: Davis-Putnam-Logemann-Loveland
- 3 Where DPLL Fails: The Need for CDCL
- 4 CDCL: Conflict-Driven Clause Learning
- 5 Decision Heuristics and Engineering

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Today: we open the black box. How does a SAT solver actually work?

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The question is: can we do better?

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“**Simplify**”: remove satisfied clauses, drop false literals from remaining clauses.
Still 2^n worst-case, but we prune early \Rightarrow skip entire subtrees.

Unit Propagation: The Key Idea

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Key insight: UP *cascades*. Forcing x_3 might make another clause unit, which forces another variable, and so on.

Unit Propagation: Cascade Example

$$\underbrace{(x_1 \vee x_2)}_{C_1} \wedge \underbrace{(\neg x_1 \vee x_3)}_{C_2} \wedge \underbrace{(\neg x_2 \vee x_4)}_{C_3} \wedge \underbrace{(\neg x_3 \vee \neg x_4)}_{C_4} \wedge \underbrace{(x_2 \vee \neg x_3)}_{C_5}$$

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- $C_4 = (\neg x_3 \vee \neg \cancel{x_4})$ is unit. **Force** $x_3 = \text{False}$.

One decision triggered a cascade assigning *all* variables. **SAT**.

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- 1 **Unit Propagation:** while any unit clause exists, force and simplify.
- 2 **Pure Literal Elimination:** assign any pure literals.
- 3 If no clauses left: return **SAT**.
- 4 If any clause is empty: return **UNSAT**.
- 5 Pick an unassigned variable x .
- 6 If $\text{DPLL}(F \mid_{x=T}) = \text{SAT}$: return **SAT**.
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In words: propagate for free, then guess, then recurse. Dead end \Rightarrow backtrack.

DPLL Walkthrough: SAT Instance

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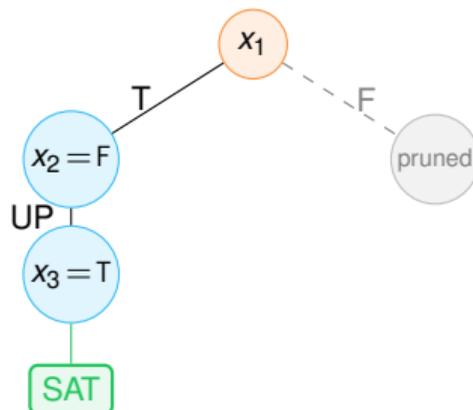
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DPLL Walkthrough: UNSAT Instance

An unsatisfiable formula (4 variables, 8 clauses):

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Both branches failed \Rightarrow **UNSAT.**

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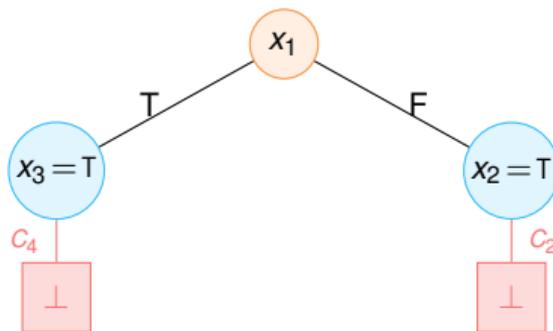
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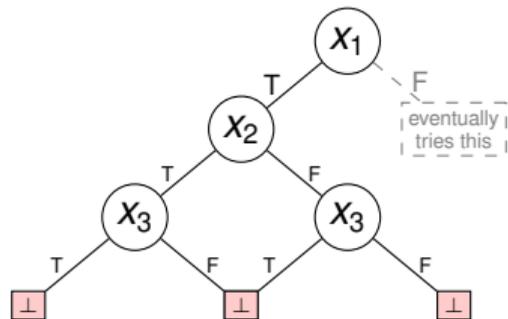
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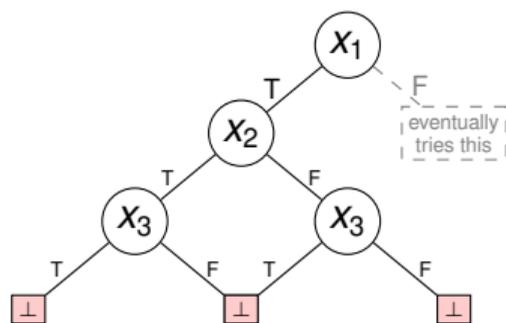
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- DPLL backtracks to level 9. But x_9 was irrelevant. Same conflict. Again at 8, 7, 6, ...
- Result: 2^8 useless combinations of x_2, \dots, x_9 before reaching the real culprit.

DPLL Makes the Same Mistake Over and Over

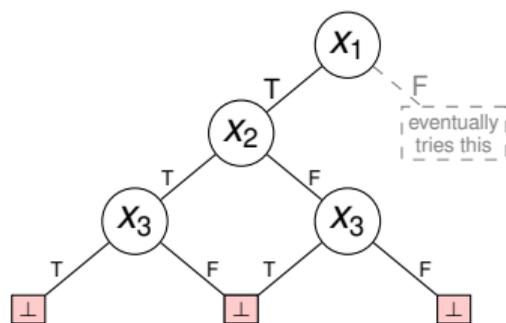


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The gap

DPLL does not learn *why* conflicts happen. It makes the same mistake over and over.

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This is exactly what **CDCL** (Conflict-Driven Clause Learning) does.

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- 3 **Non-chronological Backjumping.** Jump back to the *root cause*, not the most recent decision.

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- **Backjumping to level k** : undo all assignments from levels $> k$. Levels $0, \dots, k$ stay intact.

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- **Conflict node** (\perp , red): when propagation forces contradictory values.

CDCL Example: The Formula

$$C_1 = (x_1 \vee x_4)$$

$$C_3 = (\neg x_1 \vee x_2)$$

$$C_5 = (\neg x_2 \vee \neg x_4)$$

$$C_7 = (\neg x_3 \vee x_4)$$

$$C_2 = (x_1 \vee \neg x_3 \vee \neg x_5)$$

$$C_4 = (\neg x_1 \vee x_3 \vee x_5)$$

$$C_6 = (\neg x_2 \vee x_3)$$

$$C_8 = (x_2 \vee \neg x_4 \vee x_5)$$

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$$C_6 = (\neg x_2 \vee x_3)$$

$$C_8 = (x_2 \vee \neg x_4 \vee x_5)$$

5 variables, 8 clauses. Let's solve it with CDCL.

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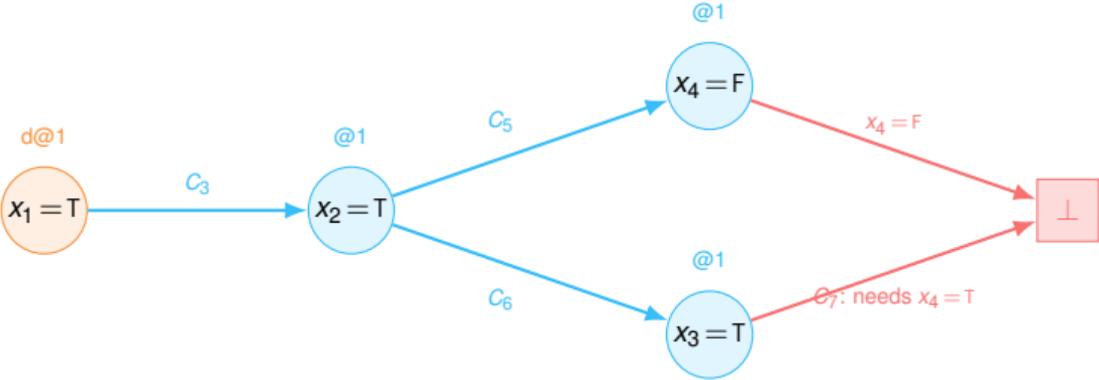
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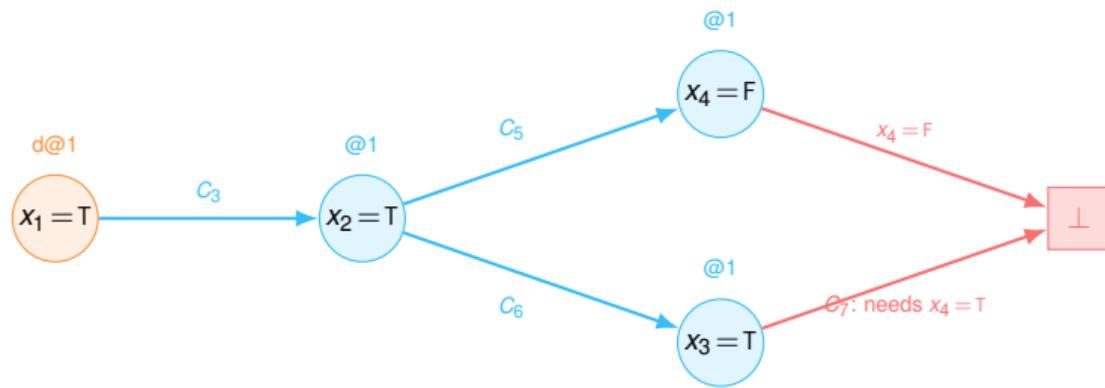
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Conflict: step 2 forced $x_4 = F$, but step 4 needs $x_4 = T$.

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The graph shows: x_2 is the **fork point**. It forced $x_4 = F$ (via C_5) and $x_3 = T$ (via C_6), and x_3 then demanded $x_4 = T$ (via C_7).

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Learned clause: $(\neg x_2)$. This clause is *permanently* added to the formula.

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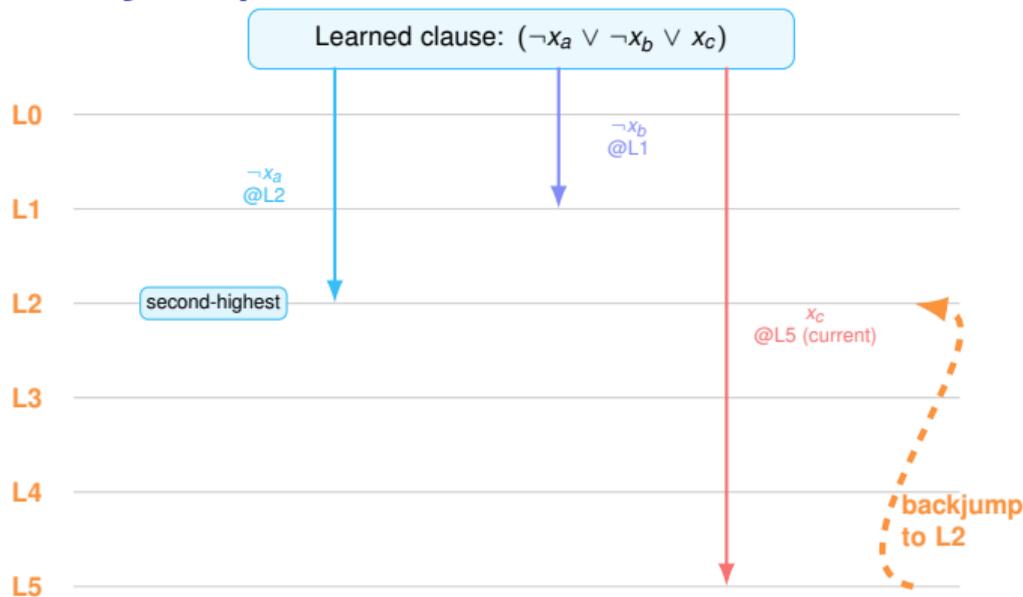
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Takeaway: resolve backwards until one literal remains at the current level. That gives you the learned clause.

Where to Backjump: The General Rule



Backjump Rule

Find the **second-highest decision level** among all literals in learned clause. Backjump there. The clause is **unit** at that level, forcing the remaining literal.

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CDCL does:

- **Backjump to level 0.**
- $(\neg x_2)$ becomes unit.
- Force $x_2 = F$.

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All 8 clauses satisfied. **SAT:** $x_1 = F$, $x_2 = F$, $x_3 = F$, $x_4 = T$, $x_5 = T$.

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Observation: one conflict, one learned clause, *zero* further decisions. UP did all the work.

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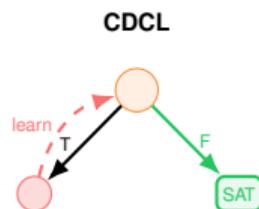
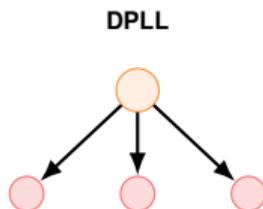
Key insight: CDCL might undo 9 decisions at once, because it *knows* the last 8 were irrelevant.

CDCL vs. DPLL: Summary

	DPLL	CDCL
Backtracking	Chronological	Non-chronological
Clause database	Static	Grows (learned clauses)
Conflict info	Discarded	Analyzed, recorded
Repeated mistakes	Yes	No

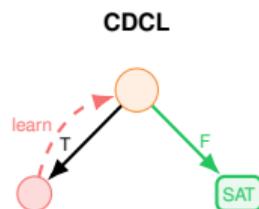
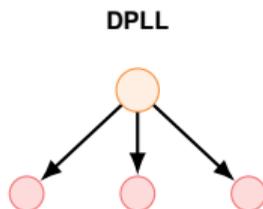
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Takeaway: CDCL = DPLL + memory. It remembers its mistakes.

- 1 Recap and Motivation
- 2 DPLL: Davis-Putnam-Logemann-Loveland
- 3 Where DPLL Fails: The Need for CDCL
- 4 CDCL: Conflict-Driven Clause Learning
- 5 Decision Heuristics and Engineering**

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- 2 **Restarts:** when to abandon and start over.

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Key insight: variables in recent conflicts are “hot.” VSIDS focuses the solver on the active part of the problem. Decay ensures old conflicts fade.

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Takeaway: restarts + clause learning = explore broadly without losing knowledge.

DPLL vs. CDCL: Complete Comparison

Feature	DPLL (1962)	CDCL (1996+)
Core idea	Backtracking + UP	Backtracking + UP
Conflict response	Backtrack one level	Analyze, learn, backjump
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The deal: CDCL adds complexity, but the payoff is **four orders of magnitude** in scalability.