

Propositional Logic

CDCL: Conflict Driven Clause
Learning

CDCL: goal and idea

Goal: Combine DPLL and resolution into an algorithm oriented towards both satisfiability and unsatisfiability.

Idea: At every unsuccessful leaf of DPLL (called **conflict**), compute a **conflict clause**, and add it to the formula we are deciding about.

Conflict clauses “cache” previous search results, so we “learn from previous mistakes”.

Conflict clauses also determine backtracking.

We present a particular way of computing a conflict clause using resolution. There are other ways.

DPLL + CDCL algorithm

Given formula F and **partial assignment** $\mathcal{A}$:

$F|_{\mathcal{A}}$ denotes the result of deleting any clause containing a true literal, and deleting all false literals from each remaining clause.

Input: CNF formula F .

1. Initialise $\mathcal{A}$ to the empty assignment
2. While there is unit clause $\{L\}$ or pure literal L in $F|_{\mathcal{A}}$, update $\mathcal{A} \mapsto \mathcal{A}[\top/L]$
3. If $F|_{\mathcal{A}} = \emptyset$, stop and output $\mathcal{A}$.
4. If $F|_{\mathcal{A}} \ni \square$, add new clause C to F by **learning procedure**.
If $C = \square$, stop and output UNSAT; otherwise backtrack to highest level where C is unit clause.
Go to line 2.
5. Apply **decision strategy** to update $\mathcal{A}$.
Go to line 2.

Terminology

- ▶ **State** of algorithm is pair $(F, \mathcal{A})$, where F is CNF formula and $\mathcal{A}$ is partial assignment.
Successful state when $\mathcal{A} \models F$. **Conflict state** when $\mathcal{A} \not\models F$.
(Note: conflict state if $F|_{\mathcal{A}} \ni \square$, successful state if $F|_{\mathcal{A}} = \emptyset$)
- ▶ Each assignment $A_i \mapsto b_i$ classifies as **decision assignment** or **implied assignment**.
- ▶ $A_i \mapsto b_i$ denotes decision assignment with **decision variable** A_i .
- ▶ $A_i \xrightarrow{C} b_i$ denotes an implied assignment arising through **unit propagation** on clause C .

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- ▶ $A_i \xrightarrow{C} b_i$ denotes an implied assignment arising through **unit propagation** on clause C .
- ▶ **Decision level** of assignment $A_i \mapsto b_i$ in a given state $(F, \mathcal{A})$ is number of decision assignments in $\mathcal{A}$ that precede $A_i \mapsto b_i$.

Example: start with set of clauses $F = \{C_1, \dots, C_5\}$, where

$$C_1 = \{\neg A_1, \neg A_4, A_5\}$$

$$C_2 = \{\neg A_1, A_6, \neg A_5\}$$

$$C_3 = \{\neg A_1, \neg A_6, A_7\}$$

$$C_4 = \{\neg A_1, \neg A_7, \neg A_5\}$$

$$C_5 = \{A_1, A_4, A_6\}$$

Say current assignment is $(A_1 \mapsto 1, A_2 \mapsto 0, A_3 \mapsto 0, A_4 \mapsto 1)$.
Notice $F|_{\mathcal{A}}$ contains unit clause $\{A_5\}$.

Unit propagation further generates $(A_5 \xrightarrow{C_1} 1, A_6 \xrightarrow{C_2} 1, A_7 \xrightarrow{C_3} 1)$.
This leads to a conflict, with C_4 being made false.

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After unit propagation:

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1. $F \equiv F \cup \{C\}$
2. C is **conflict clause**: each literal of C is made false by $\mathcal{A}$
3. C mentions only decision variables in $\mathcal{A}$

Clause learning using resolution

Suppose $\mathcal{A} = (A_1 \mapsto b_1, \dots, A_k \mapsto b_k)$ leads to conflict.

Find associated clauses $D_1, \dots, D_{k+1}$ by backward induction:

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$C := A_1$, that is, the final clause A_1 is the **learned clause** .

Clause learning: example

Conflict of example above:

$C_1 = \{\neg A_1, \neg A_4, A_5\}$	$D_8 := \{\neg A_1, \neg A_7, \neg A_5\}$	(clause C_4)
$C_2 = \{\neg A_1, A_6, \neg A_5\}$	$D_7 := \{\neg A_1, \neg A_5, \neg A_6\}$	(resolve D_8, C_3)
$C_3 = \{\neg A_1, \neg A_6, A_7\}$	$D_6 := \{\neg A_1, \neg A_5\}$	(resolve D_7, C_2)
$C_4 = \{\neg A_1, \neg A_7, \neg A_5\}$	$D_5 := \{\neg A_1, \neg A_4\}$	(resolve D_6, C_1)
$C_5 = \{A_1, A_4, A_6\}$	$D_4 := \{\neg A_1, \neg A_4\}$	
$A_1 \mapsto 1, A_2 \mapsto 0,$	$D_3 := \{\neg A_1, \neg A_4\}$	
$A_3 \mapsto 0, A_4 \mapsto 1,$	$D_2 := \{\neg A_1, \neg A_4\}$	
$A_5 \xrightarrow{C_1} 1, A_6 \xrightarrow{C_2} 1,$	$D_1 := \{\neg A_1, \neg A_4\}$	
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Learned clause D_1 is conflict clause with only decision variables, including top-level one A_1 .

Clause learning: example

Intuitively:

- ▶ D_1 records that conflict due to decision to make A_1, A_4 true.
- ▶ Adding D_1 ensures search does not explore assignments with $A_1 \mapsto 1, A_4 \mapsto 1$.
- ▶ DPLL backtracks to highest level where D_1 is unit clause (after $A_1 \mapsto 1$), unit propagation leads to $A_4 \mapsto 0$.

Clause learning

Proposition: The clause learning procedure satisfies the three desiderata.

Proof sketch: **Observation:** If $A_i \xrightarrow{C_i} b_i$, then the only literal of C_i true under $\mathcal{A}$ is the literal for A_i (that is, C_i contains either A_i or $\neg A_i$, and b_i is chosen to make the literal true).

1. $F \equiv F \cup \{C\}$

Because C is obtained from clauses of F through resolution steps.

2. C is **conflict clause**: each literal is made false by $\mathcal{A}$.

We show by induction that $D_{k+1}, D_k, \dots, D_1 = C$ are conflict clauses.

D_{k+1} is conflict clause by definition.

If D_{i+1} is conflict clause and $D_i = D_{i+1}$, then so is D_i .

If D_{i+1} is conflict clause and $D_i \neq D_{i+1}$, then D_i is the result of resolving D_{i+1} and C_i . By the **observation**, all literals of D_i are made false by $\mathcal{A}$.

3. C mentions only decision variables in $\mathcal{A}$.

Because every other variable, say A_i , disappears after resolving with D_{i+1} w.r.t. A_i .

Indeed, since $\mathcal{A}$ makes D_{i+1} false, by the **observation** A_i has opposite signs in D_{i+1} and C_i .

Example (without PLR)

$$\{\neg A_1\} \{A_1, A_3, A_4\} \{\neg A_2, \neg A_5\} \{A_3, \neg A_4, A_5, \neg A_6\} \{A_1, \neg A_2, \neg A_4, A_6\}$$

$$\text{OLR: } A_1 \mapsto 0 \quad \{A_3, A_4\} \quad \{\neg A_2, \neg A_5\} \quad \{A_3, \neg A_4, A_5, \neg A_6\} \quad \{\neg A_2, \neg A_4, A_6\}$$

$$\text{DE: } A_2 \mapsto 1 \quad \{A_3, A_4\} \quad \{\neg A_5\} \quad \{A_3, \neg A_4, A_5, \neg A_6\} \quad \{\neg A_4, A_6\}$$

$$\text{OLR: } A_5 \mapsto 0 \quad \{A_3, A_4\} \quad \{A_3, \neg A_4, \neg A_6\} \quad \{\neg A_4, A_6\}$$

$$\text{DE: } A_3 \mapsto 0 \quad \{A_4\} \quad \{\neg A_4, \neg A_6\} \quad \{\neg A_4, A_6\}$$

$$\text{OLR: } A_4 \mapsto 1 \quad \{\neg A_6\} \quad \{A_6\}$$

$$\text{OLR: } A_6 \mapsto 1 \quad \{\}$$

$$D_7 := \{A_3, \neg A_4, A_5, \neg A_6\} \quad (\text{conflict clause})$$

$$D_6 := \{A_1, \neg A_2, A_3, \neg A_4, A_5\} \quad (\text{resolve } D_7, \{A_1, \neg A_2, \neg A_4, A_6\})$$

$$D_5 := \{A_1, \neg A_2, A_3, A_5\} \quad (\text{resolve } D_6, \{A_1, A_3, A_4\})$$

$$D_4 := \{A_1, \neg A_2, A_3, A_5\}$$

$$D_3 := \{A_1, \neg A_2, A_3\} \quad (\text{resolve } D_4, \{\neg A_2, \neg A_5\})$$

$$D_2 := \{A_1, \neg A_2, A_3\}$$

$$D_1 := \{\neg A_2, A_3\} \quad (\text{resolve } D_2, \{\neg A_1\})$$

Backtracking to $\{A_1 \mapsto 0, A_2 \mapsto 1\}$. Unit propagation: $A_3 \mapsto 1$.