

# Decision Procedures

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DPLL(T)

Suppose we have a  $T_{\mathbb{Q}}$ -formulae that is not conjunctive:

$$(x \geq 0 \rightarrow y > z) \wedge (x + y \geq z \rightarrow y \leq z) \wedge (y \geq 0 \rightarrow x \geq 0) \wedge x + y \geq z$$

Our approach so far: Converting to DNF.

Yields in 8 conjuncts that have to be checked separately.

Is there a more efficient way to prove unsatisfiability?

Suppose we have the following  $T_{\mathbb{Q}}$ -formulae:

$$(x \geq 0 \rightarrow y > z) \wedge (x + y \geq z \rightarrow y \leq z) \wedge (y \geq 0 \rightarrow x \geq 0) \wedge x + y \geq z$$

Converting to CNF and restricting to  $\leq$ :

$$\begin{aligned} & (\neg(\underbrace{0 \leq x}_{P_1}) \vee \neg(\underbrace{y \leq z}_{P_2})) \wedge (\neg(\underbrace{z \leq x + y}_{P_3}) \vee \underbrace{y \leq z}_{P_2}) \\ & \wedge (\neg(\underbrace{0 \leq y}_{P_4}) \vee \underbrace{0 \leq x}_{P_1}) \wedge \underbrace{z \leq x + y}_{P_3} \end{aligned}$$

Now, introduce boolean variables for each atom:

$$P_1 : 0 \leq x$$

$$P_2 : y \leq z$$

$$P_3 : z \leq x + y$$

$$P_4 : 0 \leq y$$

Gives a propositional formula:

$$(\neg P_1 \vee \neg P_2) \wedge (\neg P_3 \vee P_2) \wedge (\neg P_4 \vee P_1) \wedge P_3$$

The core feature of the DPLL-algorithm is Unit Propagation.

$$\underbrace{(\neg P_1)}_{\downarrow \text{false}} \vee \underbrace{(\neg P_2)}_{\downarrow \text{false}} \wedge \underbrace{(\neg P_3)}_{\downarrow \text{false}} \vee \underbrace{P_2}_{\downarrow \text{true}} \wedge \underbrace{(\neg P_4)}_{\downarrow \text{false}} \vee \underbrace{P_1}_{\downarrow \text{false}} \wedge P_3$$

$\downarrow$  true

The core feature of the DPLL-algorithm is Unit Propagation.

$$(\neg P_1 \vee \neg P_2) \wedge (\neg P_3 \vee P_2) \wedge (\neg P_4 \vee P_1) \wedge P_3$$

The clause  $P_3$  is a unit clause; set  $P_3$  to  $\top$ .

Then  $\neg P_3 \vee P_2$  is a unit clause; set  $P_2$  to  $\top$ .

Then  $\neg P_1 \vee \neg P_2$  is a unit clause; set  $P_1$  to  $\perp$ .

Then  $\neg P_4 \vee P_1$  is a unit clause; set  $P_4$  to  $\perp$ .

Only solution is  $P_3 \wedge P_2 \wedge \neg P_1 \wedge \neg P_4$ .

Only solution is  $P_3 \wedge P_2 \wedge \neg P_1 \wedge \neg P_4$ .

$$P_1 : 0 \leq x$$

$$P_2 : y \leq z$$

$$P_3 : z \leq x + y$$

$$P_4 : 0 \leq y$$

This gives the **conjunctive**  $T_{\mathbb{Q}}$ -formula

$$(z \leq x + y) \wedge (y \leq z) \wedge (x < 0) \wedge (y < 0)$$

We describe DPLL(T) by a set of rules modifying a configuration.  
A configuration is a triple

$$\langle M, F, C \rangle,$$

where

- $M$  (model) is a sequence of literals (that are currently set to true) interspersed with backtracking points denoted by  $\square$ .
- $F$  (formula) is a formula in CNF, i. e., a set of clauses where each clause is a set of literals.
- $C$  (conflict) is either  $\top$  or a conflict clause (a set of literals).  
[ A conflict clause  $C$  is a clause with  $F \Rightarrow C$  and  $M \not\models C$ .  
Thus, a conflict clause shows  $M \not\models F$ . ]



We describe the algorithm by a set of rules, which each describe a set of transitions between configurations, e. g.,

Explain  $\frac{\langle M, F, C \cup \{l\} \rangle}{\langle M, F, C \cup \{l_1, \dots, l_k\} \rangle}$  where  $l \notin C, \{l_1, \dots, l_k, \bar{l}\} \in F,$   
 and  $\bar{l}_1, \dots, \bar{l}_k \prec \bar{l}$  in  $M$ .

*↑  
name*

Here,  $\bar{l}_1, \dots, \bar{l}_k \prec \bar{l}$  in  $M$  means the literals  $\bar{l}_1, \dots, \bar{l}_k$  occur in the sequence  $M$  before the literal  $\bar{l}$  (and all literals appear in  $M$ ).

**Example:** for  $M = P_1 \bar{P}_3 \bar{P}_2 \bar{P}_4$ ,  $F = \{\{P_1\}, \{P_3, \bar{P}_4\}\}$ , and  $C = \{P_2\}$  the transition

$$\langle M, F, \underbrace{C \cup \{l\}}_{\bar{P}_4} \rangle \xrightarrow{\text{Explain}} \langle M, F, \{P_2, P_3\} \rangle$$

is possible.

# Rules for CDCL (Conflict Driven Clause Learning)

sequence  
concatenation

$R_1 < R_2$  iff  $R_1$  better than  $R_2$  (Heur.)

$$\text{Decide} \quad \frac{\langle M, F, T \rangle}{\langle M \cdot \square \cdot l, F, T \rangle}$$

✓

$$\text{Propagate} \quad \frac{\langle M, F, T \rangle}{\langle M \cdot l, F, T \rangle}$$

✓

$$\text{Conflict} \quad \frac{\langle M, F, T \rangle}{\langle M, F, \{l_1, \dots, l_k\} \rangle}$$

Explain

$$\frac{\langle M, F, C \cup \{l\} \rangle}{\langle M, F, C \cup \{l_1, \dots, l_k\} \rangle}$$

Heur.: go backwards  
in  $M$

Learn

$$\frac{\langle M, F, C \rangle}{\langle M, F \cup \{C\}, C \rangle}$$

Heur.: apply Learn  
right before  
Back

Back

$$\frac{\langle M, F, \{l_1, \dots, l_k, l\} \rangle}{\langle M' \cdot l, F, T \rangle}$$

backtrack exactly once

where  $l \in \text{lit}(F)$ ,  $l, \bar{l} \notin M$

where  $\{l_1, \dots, l_k, l\} \in F$   
and  $\bar{l}_1, \dots, \bar{l}_k \in M$ ,  $l, \bar{l} \notin M$ .

where  $\{l_1, \dots, l_k\} \in F$   
and  $\bar{l}_1, \dots, \bar{l}_k \in M$ .

where  $l \notin C$ ,  $\{l_1, \dots, l_k, \bar{l}\} \in F$ ,  
and  $\bar{l}_1, \dots, \bar{l}_k < \bar{l}$  in  $M$ .

where  $C \neq T$ ,  $C \notin F$ .

Heur.: choose  $M'$  small

where  $\{l_1, \dots, l_k, l\} \in F$ ,  
 $M = M' \cdot \square \dots \bar{l} \dots$ ,  
and  $\bar{l}_1, \dots, \bar{l}_k \in M'$ .  
conflict already learned

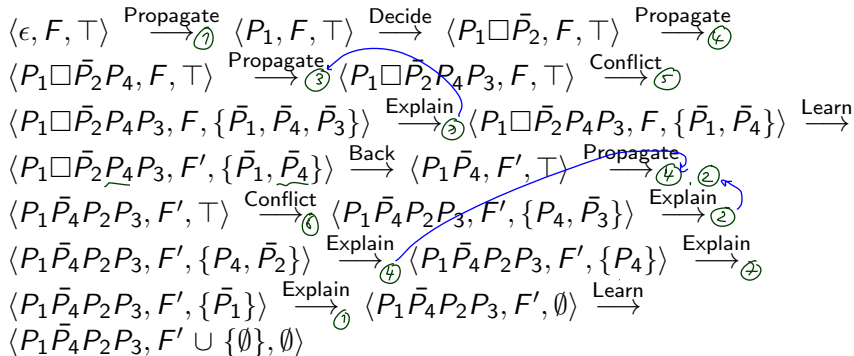
# Example: DPLL with Learning

$$P_1 \wedge (\neg P_2 \vee P_3) \wedge (\neg P_4 \vee P_3) \wedge (P_2 \vee P_4) \wedge (\neg P_1 \vee \neg P_4 \vee \neg P_3) \wedge (P_4 \vee \neg P_3)$$

①
②
③
④
⑤
⑥

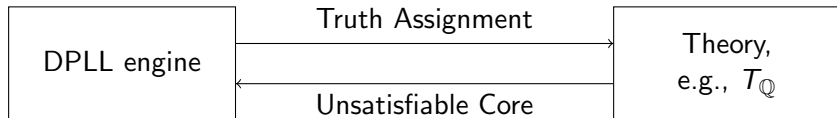
The algorithm starts with  $M = \epsilon$ ,  $C = \top$  and

$$F = \{\{P_1\}, \{\bar{P}_2, P_3\}, \{\bar{P}_4, P_3\}, \{P_2, P_4\}, \{\bar{P}_1, \bar{P}_4, \bar{P}_3\}, \{P_4, \bar{P}_3\}\}.$$



where  $F' = F \cup \{\{\bar{P}_1, \bar{P}_4\}\}$ .

The DPLL/CDCL algorithm is combined with a Decision Procedures for a Theory



DPLL takes the **propositional core** of a formula, assigns truth-values to **atoms**.

Theory takes a **conjunctive** formula (conjunction of literals), returns a **(minimal) unsatisfiable core**.

Suppose we have a decision procedure for a conjunctive theory, e.g., Simplex Algorithm for  $T_{\mathbb{Q}}$ .

Given an unsatisfiable conjunction of literals  $l_1 \wedge \dots \wedge l_n$ .

Find a subset  $\text{UnsatCore} = \{l_{i_1}, \dots, l_{i_m}\}$ , such that  $1 \leq i_1, \dots, i_m \leq n$

- $l_{i_1} \wedge \dots \wedge l_{i_m}$  is unsatisfiable.
- For each subset of **UnsatCore** the conjunction is satisfiable. } *minimal*

Possible approach: check for each literal whether it can be omitted.  
→  $n$  calls to decision procedure.

Most decision procedures can give small unsatisfiable cores for free.

yields an min. unsat. core (not necessarily the smallest)

$$(x > 0) \wedge (x < 0)$$

$$l_1 \wedge \dots \wedge l_n \quad \text{unsat.}$$

Theory returns an unsatisfiable core:

- a conjunction of literals from current truth assignment
- that is unsatisfiable.

DPLL learns conflict clauses, a disjunction of literals

- that are implied by the formula
- and in conflict to current truth assignment.

Thus the negation of an unsatisfiable core is a conflict clause.

*i.e. valid*

$$\bar{l}_1 \vee \dots \vee \bar{l}_n$$

The DPLL part only needs one new rule:

TConflict  $\frac{\langle M, F, \top \rangle}{\langle M, F, C \rangle}$  where  $M$  is unsatisfiable in the theory  
and  $\neg C$  an unsatisfiable core of  $M$ .

$$F : \boxed{y \geq 1} \wedge (x \geq 0 \rightarrow y \leq 0) \wedge (x \leq 1 \rightarrow y \leq 0)$$

Atomic propositions:

$$P_1 : y \geq 1$$

$$P_2 : x \geq 0$$

$$P_3 : y \leq 0$$

$$P_4 : x \leq 1$$

Propositional core of  $F$  in CNF:

$$F_0 : (P_1) \wedge (\neg P_2 \vee P_3) \wedge (\neg P_4 \vee P_3)$$



## Running DPLL(T)

$$F_0 : \{\{P_1\}, \{\bar{P}_2, P_3\}, \{\bar{P}_4, P_3\}\}$$

$$P_1 : \underline{y \geq 1} \quad P_2 : \underline{x \geq 0} \quad P_3 : \underline{y \leq 0} \quad P_4 : \underline{x \leq 1}$$

$$\begin{aligned} &\langle \epsilon, F_0, T \rangle \xrightarrow{\text{Propagate } \textcircled{1}} \langle P_1, F_0, T \rangle \xrightarrow{\text{Decide}} \langle P_1 \square P_3, F_0, T \rangle \xrightarrow{\text{TConflict}} \\ &\langle P_1 \square P_3, F_0, \{\bar{P}_1, \bar{P}_3\} \rangle \xrightarrow{\text{Learn}} \langle P_1 \square P_3, F_1, \{\bar{P}_1, \bar{P}_3\} \rangle \xrightarrow{\text{Back}} \\ &\langle P_1 \bar{P}_3, F_1, T \rangle \xrightarrow{\text{Propagate } \textcircled{2}} \langle P_1 \bar{P}_3 \bar{P}_2, F_1, T \rangle \xrightarrow{\text{Propagate } \textcircled{3}} \\ &\langle P_1 \bar{P}_3 \bar{P}_2 \bar{P}_4, F_1, T \rangle \xrightarrow{\text{TConflict}} \langle P_1 \bar{P}_3 \bar{P}_2 \bar{P}_4, F_1, \{P_2, P_4\} \rangle \xrightarrow{\text{Explain } \textcircled{2}} \\ &\langle P_1 \bar{P}_3 \bar{P}_2 \bar{P}_4, F_1, \{P_2, P_3\} \rangle \xrightarrow{\text{Explain } \textcircled{2}} \langle P_1 \bar{P}_3 \bar{P}_2 \bar{P}_4, F_1, \{P_3\} \rangle \xrightarrow{\text{Explain } \textcircled{4}} \\ &\langle P_1 \bar{P}_3 \bar{P}_2 \bar{P}_4, F_1, \{\bar{P}_1\} \rangle \xrightarrow{\text{Explain } \textcircled{1}} \langle P_1 \bar{P}_3 \bar{P}_2 \bar{P}_4, F_1, \emptyset \rangle \xrightarrow{\text{Learn}} \\ &\langle P_1 \bar{P}_3 \bar{P}_2 \bar{P}_4, F_1 \cup \{\emptyset\}, \emptyset \rangle \end{aligned}$$

$$\text{where } F_1 := F_0 \cup \{\{\bar{P}_1, \bar{P}_3\}\}$$

No further step is possible; the formula  $F$  is unsatisfiable.