

# AAA528: Computational Logic

## Lecture 5 — DPLL(T)

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# Deciding $T$ -Satisfiability

- Equality

$$(x_1 = x_2 \vee x_1 = x_3) \wedge (x_1 = x_2 \vee x_1 = x_4) \wedge \\ x_1 \neq x_2 \wedge x_1 \neq x_3 \wedge x_1 \neq x_4$$

- Linear arithmetic

$$(x_1 + 2x_3 < 5) \vee \neg(x_3 \leq 1) \wedge (x_1 \geq 3)$$

- Arrays

$$(i = j \wedge a[j] = 1) \wedge \neg(a[i] = 1)$$

# DPLL(T)

- Generalization of CDCL to decidable quantifier-free first-order theories
- Based on an interplay between a CDCL SAT solver and a decision procedure  $DP_T$  for the conjunctive fragment of  $T$
- Implemented in most Satisfiability Module Theory (SMT) solvers

## Theory Solver ( $DP_T$ )

- A theory solver  $DP_T$  accepts conjunctive quantifier-free  $\Sigma$ -formulas, where conjunctive  $\Sigma$ -formulas are conjunctions of  $\Sigma$ -literals.
- This does not restrict the scope of the decision procedures. For arbitrary quantifier-free  $\Sigma$ -formula  $F$ , we can convert it to DNF:

$$F_1 \vee F_2 \vee \dots \vee F_k.$$

$F$  is  $T$ -satisfiable iff at least one  $F_i$  is  $T$ -satisfiable. E.g.,

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

- This method misses any opportunity for learning, as each clause is solved independently, e.g.,  $x_1 = x_2 \wedge x_1 \neq x_2$ .
- A better approach is to leverage the learning capabilities of SAT and combine it with  $DP_T$ .

## $DP_T$ for Equality Theory

- A literal is either equality or inequality.
- Given a conjunction of  $T$ -literals  $\phi$ , build an undirected graph  $G(N, E=, E\neq)$ .
  - ▶  $V$  is the set of variables in  $\phi$ .
  - ▶  $(x_1, x_2) \in E=$  iff  $(x_1 = x_2) \in \phi$
  - ▶  $(x_1, x_2) \in E\neq$  iff  $(x_1 \neq x_2) \in \phi$
- $\phi$  is unsatisfiable iff there exists an edge  $(x_1, x_2) \in E\neq$  such that  $x_2$  is reachable from  $x_1$  through a sequence of  $E=$  edges.
- Example:

$$x_1 \neq x_2 \wedge x_2 = x_3 \wedge x_1 = x_3$$

## Overview of DPLL(T)

- Let  $at(\phi)$  be the set of  $\Sigma$ -atoms in a given NNF formula  $\phi$ .
- Assuming some order, let  $at_i(\phi)$  be the  $i$ -th distinct atom in  $\phi$ .
- Given an atom  $a$ , let  $e(a)$  be the unique boolean variable associated with  $a$  (called boolean encoder of  $a$ ). Given a  $\Sigma$ -formula  $t$ , let  $e(t)$  be the boolean formula where each atom is replaced by  $e(a)$ .
- For example, if  $x = y$  is a  $\Sigma$ -atom, then  $e(x = y)$  denotes its encoder (boolean variable) and when  $\phi := x = y \vee x = z$ ,  $e(\phi) := e(x = y) \vee e(x = z)$ .
- $e(\phi)$  is called the propositional skeleton of  $\phi$ .

# Overview of DPLL(T)

- Example:  $\phi := x = y \wedge ((y = z \wedge \neg(x = z)) \vee x = z)$ .
- The propositional skeleton:

$$e(\phi) := e(x = y) \wedge ((e(y = z) \wedge \neg e(x = z)) \vee e(x = z))$$

- Let  $B$  be a boolean formula initially set to  $e(\phi)$ .
- Next, we check the satisfiability of  $B$  using a SAT solver. A satisfying assignment

$$\alpha := \{e(x = y) \mapsto \mathbf{true}, e(y = z) \mapsto \mathbf{true}, e(x = z) \mapsto \mathbf{false}\}$$

- This does not mean that  $\phi$  is satisfiable. We need to check the conjunction:

$$\hat{T}h(\alpha) := x = y \wedge y = z \wedge \neg(x = z)$$

- $\hat{T}h(\alpha)$  is not satisfiable. We conjoin  $B$  with

$$e(\neg \hat{T}h(\alpha)) := (\neg e(x = y) \vee \neg e(y = z) \vee e(x = z))$$

which is called a blocking clause or lemma.

- The SAT solver is invoked again and suggests another assignment:

$$\alpha' := \{e(x = y) \mapsto \mathbf{true}, e(y = z) \mapsto \mathbf{true}, e(x = z) \mapsto \mathbf{true}\}$$

- The corresponding  $\hat{T}h(\alpha') := x = y \wedge y = z \wedge x = z$  is satisfiable, so  $\phi$  is.

# Theory Propagation

- An improvement to the procedure.
- Invoke  $DP_T$  after some or all partial assignments, rather than waiting for a full assignment.
- When the partial assignment is not contradictory, propagate implications due to the theory  $T$  back to the SAT solver.
- Consider the partial assignment:

$$\alpha := \{e(x = y) \mapsto \mathbf{true}, e(y = z) \mapsto \mathbf{true}\}$$

and the corresponding formula fed to  $DP_T$ :

$$\hat{T}h(\alpha) := x = y \wedge y = z.$$

- $DP_T$  infers that  $x = z$  and informs the SAT solver that  $e(x = z) \mapsto \mathbf{true}$  is implied by the partial assignment.
- In addition to BCP, we now have theory propagation (TP). TP may lead to further BCP, which means that (BCP,TP) may iterate several times.



# DPLL(T)

## Algorithm 11.2.3: DPLL( $T$ )

**Input:** A formula  $\varphi$

**Output:** “Satisfiable” if the formula is satisfiable and “Unsatisfiable” otherwise

```
1. function DPLL( $T$ )
2.   ADDCLAUSES( $cnf(e(\varphi))$ );
3.   if BCP() = “conflict” then return “Unsatisfiable”;
4.   while (TRUE) do
5.     if  $\neg$ DECIDE() then return “Satisfiable”;  $\triangleright$  Full assignment
6.     repeat
7.       while (BCP() = “conflict”) do
8.          $backtrack\text{-}level :=$  ANALYZE-CONFLICT();
9.         if  $backtrack\text{-}level < 0$  then return “Unsatisfiable”;
10.        else BackTrack( $backtrack\text{-}level$ );
11.         $\langle t, res \rangle :=$  DEDUCTION( $\hat{T}h(\alpha)$ );
12.        ADDCLAUSES( $e(t)$ );
13.    until  $t \equiv$  TRUE
```