## Foundations of Artificial Intelligence E4. Propositional Logic: DPLL Algorithm

Malte Helmert

University of Basel

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# Foundations of Artificial Intelligence April 28, 2025 — E4. Propositional Logic: DPLL Algorithm

**E4.1 Motivation** 

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## Propositional Logic: Overview

#### Chapter overview: propositional logic

- ► E1. Syntax and Semantics
- ► E2. Equivalence and Normal Forms
- ► E3. Reasoning and Resolution
- ► E4. DPLL Algorithm
- ► E5. Local Search and Outlook

## **E4.1 Motivation**

## Propositional Logic: Motivation

- Propositional logic allows for the representation of knowledge and for deriving conclusions based on this knowledge.
- many practical applications can be directly encoded, e.g.
  - constraint satisfaction problems of all kinds
  - circuit design and verification
- many problems contain logic as ingredient, e.g.
  - automated planning
  - general game playing
  - description logic queries (semantic web)

### Propositional Logic: Algorithmic Problems

#### main problems:

- reasoning  $(\Phi \models \varphi?)$ : Does the formula  $\varphi$  logically follow from the formulas  $\Phi$ ?
- equivalence  $(\varphi \equiv \psi)$ : Are the formulas  $\varphi$  and  $\psi$  logically equivalent?
- ightharpoonup satisfiabile? If yes, find a model.

German: Schlussfolgern, Äquivalenz, Erfüllbarkeit

## The Satisfiability Problem

# The Satisfiability Problem (SAT) given:

propositional formula in conjunctive normal form (CNF) usually represented as pair  $\langle V, \Delta \rangle$ :

- V set of propositional variables (propositions)
- ▶  $\Delta$  set of clauses over V (clause = set of literals v or  $\neg v$  with  $v \in V$ )

#### find:

- satisfying interpretation (model)
- or proof that no model exists

SAT is a famous NP-complete problem (Cook 1971; Levin 1973).

#### Relevance of SAT

- ► The name "SAT" is often used for the satisfiability problem for general propositional formulas (instead of restriction to CNF).
- General SAT can be reduced to CNF case in linear time.
- All previously mentioned problems can be reduced to SAT or its complement UNSAT (is a given CNF formula unsatisfiable?) in linear time.
- → SAT algorithms important and intensively studied

this and next chapter: SAT algorithms

## E4.2 Systematic Search: DPLL

#### SAT vs. CSP

#### SAT can be considered a constraint satisfaction problem:

- ► CSP variables = propositions
- ightharpoonup domains =  $\{\mathbf{F}, \mathbf{T}\}$
- constraints = clauses

However, we often have constraints that affect > 2 variables.

Due to this relationship, all ideas for CSPs are applicable to SAT:

- search
- ▶ inference
- variable and value orders

## The DPLL Algorithm

The DPLL algorithm (Davis/Putnam/Logemann/Loveland) corresponds to backtracking with inference for CSPs.

- recursive call DPLL( $\Delta$ , I) for clause set  $\Delta$  and partial interpretation I
- result is a model of Δ that extends I; unsatisfiable if no such model exists
- ▶ first call  $DPLL(\Delta, \emptyset)$

#### inference in DPLL:

- simplify: after assigning value d to variable v,
   simplify all clauses that contain v
   forward checking (for constraints of arbitrary arity)
- unit propagation: variables that occur in clauses without other variables (unit clauses) are assigned immediately
  - → minimum remaining values variable order

## The DPLL Algorithm: Pseudo-Code

```
function DPLL(\Delta, I):
if \bot \in \Delta:
                                              [empty clause exists \rightsquigarrow unsatisfiable]
      return unsatisfiable
else if \Delta = \emptyset: [no clauses left \rightsquigarrow interpretation / satisfies formula]
      return /
else if there exists a unit clause \{v\} or \{\neg v\} in \Delta: [unit propagation]
      Let v be such a variable, d the truth value that satisfies the clause.
      \Delta' := simplify(\Delta, v, d)
      return DPLL(\Delta', I \cup \{v \mapsto d\})
else:
                                                                          splitting rule
      Select some variable v which occurs in \Delta.
      for each d \in \{F, T\} in some order:
            \Delta' := simplify(\Delta, v, d)
            I' := \mathsf{DPLL}(\Delta', I \cup \{v \mapsto d\})
            if I' \neq unsatisfiable
                  return I'
      return unsatisfiable
```

## The DPLL Algorithm: simplify

#### **function** simplify( $\Delta$ , v, d)

Let  $\ell$  be the literal for  $\nu$  that is satisfied by  $\nu \mapsto d$ .

$$\Delta' := \{C \mid C \in \Delta \text{ such that } \ell \notin C\}$$
  
$$\Delta'' := \{C \setminus \{\bar{\ell}\} \mid C \in \Delta'\}$$
  
**return**  $\Delta''$ 

- ► Remove clauses containing  $\ell$   $\rightsquigarrow$  clause is satisfied by  $v \mapsto d$
- lacktriangle Remove  $ar\ell$  from remaining clauses
  - ⇔ clause has to be satisfied with another variable

## Example (1)

$$\Delta = \{\{X,Y,\neg Z\}, \{\neg X, \neg Y\}, \{Z\}, \{X, \neg Y\}\}$$

- unit propagation:  $Z \mapsto T$  $\{\{X,Y\}, \{\neg X, \neg Y\}, \{X, \neg Y\}\}$
- splitting rule:
- 2a.  $X \mapsto \mathbf{F}$   $\{\{Y\}, \{\neg Y\}\}$
- 3a. unit propagation:  $Y \mapsto \mathbf{T}$   $\{\bot\}$

- 2b.  $X \mapsto \mathbf{T}$   $\{\{\neg Y\}\}$
- 3b. unit propagation:  $Y \mapsto F$  {}

## Example (2)

$$\Delta = \{\{W, \neg X, \neg Y, \neg Z\}, \{X, \neg Z\}, \{Y, \neg Z\}, \{Z\}\}$$

- unit propagation:  $Z \mapsto T$   $\{\{W, \neg X, \neg Y\}, \{X\}, \{Y\}\}\}$
- unit propagation:  $X \mapsto T$   $\{\{W, \neg Y\}, \{Y\}\}$
- unit propagation:  $Y \mapsto T$   $\{\{W\}\}$
- unit propagation:  $W \mapsto T$

## Properties of DPLL

- DPLL is sound and complete.
- DPLL computes a model if a model exists.
  - Some variables possibly remain unassigned in the solution *I*; their values can be chosen arbitrarily.
- time complexity in general exponential
- important in practice: good variable order and additional inference methods (in particular clause learning)
- ▶ Best known SAT algorithms are based on DPLL.

## E4.3 DPLL on Horn Formulas

#### Horn Formulas

#### important special case: Horn formulas

#### Definition (Horn formula)

A Horn clause is a clause with at most one positive literal, i.e., of the form

$$\neg x_1 \lor \cdots \lor \neg x_n \lor y \text{ or } \neg x_1 \lor \cdots \lor \neg x_n$$

$$(n = 0 \text{ is allowed.})$$

A Horn formula is a propositional formula in conjunctive normal form that only consists of Horn clauses.

#### German: Hornformel

- ► foundation of logic programming (e.g., PROLOG)
- critical in many kinds of practical reasoning problems

#### DPLL on Horn Formulas

#### Proposition (DPLL on Horn formulas)

If the input formula  $\varphi$  is a Horn formula, then the time complexity of DPLL is polynomial in the length of  $\varphi$ .

# Proof. properties:

- If  $\Delta$  is a Horn formula, then so is simplify $(\Delta, v, d)$ . (Why?)
  - → all formulas encountered during DPLL search are Horn formulas if input is Horn formula
- Every Horn formula without empty or unit clauses is satisfiable:
  - all such clauses consist of at least two literals
  - Horn property: at least one of them is negative
  - assigning F to all variables satisfies formula

. . .

## DPLL on Horn Formulas (Continued)

#### Proof (continued).

- From 2. we can conclude:
  - if splitting rule applied, then current formula satisfiable, and
  - if a wrong decision is taken, then this will be recognized without applying further splitting rules (i.e., only by applying unit propagation and by deriving the empty clause).
- Hence the generated search tree for n variables can only contain at most n nodes where the splitting rule is applied (i.e., where the tree branches).
- It follows that the search tree is of polynomial size, and hence the runtime is polynomial.



E4. Propositional Logic: DPLL Algorithm Summary

# E4.4 Summary

## Summary

- satisfiability basic problem in propositional logic to which other problems can be reduced
- here: satisfiability for CNF formulas
- Davis-Putnam-Logemann-Loveland procedure (DPLL): systematic backtracking search with unit propagation as inference method
- ▶ DPLL successful in practice, in particular when combined with other ideas such as clause learning
- polynomial on Horn formulas(= at most one positive literal per clause)