

# 04a—Propositional Logic II

CS 3234: Logic and Formal Systems

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- 1 Recap: Syntax and Semantics of Propositional Logic
- 2 Questions
- 3 Conjunctive Normal Form
- 4 Algorithms for Satisfiability

- 1 Recap: Syntax and Semantics of Propositional Logic
  - Propositional Atoms
  - Syntax of Propositional Logic
  - Evaluation of Formulas
- 2 Questions
- 3 Conjunctive Normal Form
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# Atoms

## Convention

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## Atoms

More formally, we fix a set  $A$  of propositional atoms.

# Meaning of Atoms

Models assign truth values

A *model* assigns truth values ( $F$  or  $T$ ) to each atom.

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## More formally

A model (valuation) for a propositional logic for the set  $A$  of atoms is a mapping from  $A$  to  $\{T, F\}$ .

# Inductive Definition

## Definition

For a given set  $A$  of propositional atoms, the set of *well-formed formulas in propositional logic* is the least set  $F$  that fulfills the following rules:

- The constant symbols  $\perp$  and  $\top$  are in  $F$ .
- Every element of  $A$  is in  $F$ .
- If  $\phi$  is in  $F$ , then  $(\neg\phi)$  is also in  $F$ .
- If  $\phi$  and  $\psi$  are in  $F$ , then  $(\phi \wedge \psi)$  is also in  $F$ .
- If  $\phi$  and  $\psi$  are in  $F$ , then  $(\phi \vee \psi)$  is also in  $F$ .
- If  $\phi$  and  $\psi$  are in  $F$ , then  $(\phi \rightarrow \psi)$  is also in  $F$ .

# Parse trees

A formula

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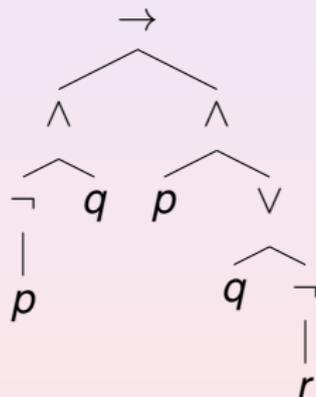
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# Evaluation of Formulas

## Definition

The result of *evaluating* a well-formed propositional formula  $\phi$  with respect to a valuation  $v$ , denoted  $v(\phi)$  is defined as follows:

- If  $\phi$  is the constant  $\perp$ , then  $v(\phi) = F$ .
- If  $\phi$  is the constant  $\top$ , then  $v(\phi) = T$ .
- If  $\phi$  is an propositional atom  $p$ , then  $v(\phi) = p^v$ .
- If  $\phi$  has the form  $(\neg\psi)$ , then  $v(\phi) = \neg v(\psi)$ .
- If  $\phi$  has the form  $(\psi \wedge \tau)$ , then  $v(\phi) = v(\psi) \& v(\tau)$ .
- If  $\phi$  has the form  $(\psi \vee \tau)$ , then  $v(\phi) = v(\psi) | v(\tau)$ .
- If  $\phi$  has the form  $(\psi \rightarrow \tau)$ , then  $v(\phi) = v(\psi) \Rightarrow v(\tau)$ .

# Valid and Satisfiable Formulas

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A formula is called *valid* if it evaluates to  $T$  with respect to every possible valuation.

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A formula is called *satisfiable* if it evaluates to  $T$  with respect to at least one valuation.

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# Questions about Propositional Formula

- Is a given formula valid?
- Is a given formula satisfiable?
- Is a given formula invalid?
- Is a given formula unsatisfiable?
- Are two formulas equivalent?

# Decision Problems

## Definition

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## Examples

The question whether a given propositional formula is satisfiable (unsatisfiable, valid, invalid) is a decision problem.

The question whether two given propositional formulas are equivalent is also a decision problem.

# How to Solve the Decision Problem?

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## The good news

We can construct a truth table for the formula and check if some/all rows have  $\top$  in the last column.

# Satisfiability is Decidable

## An algorithm for satisfiability

Using a truth table, we can implement an *algorithm* that returns “yes” if the formula is satisfiable, and that returns “no” if the formula is unsatisfiable.

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## Decidability

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## Decidability of satisfiability

The question, whether a given propositional formula is satisfiable, is decidable.

# The Bad News

## Concern

In practice, propositional formulas can be large. Example:

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## Techniques so far inadequate

Proving satisfiability/validity using truth tables or natural deduction is impractical for large formulas.

# Is there a *practical* way of deciding satisfiability?

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## More precisely...

Is there a *polynomial-time* algorithm that decides whether a given formula is satisfiable?

# Is there a *practical* way of deciding satisfiability?

## Question

Is there an *efficient* algorithm that decides whether a given formula is satisfiable?

## More precisely...

Is there a *polynomial-time* algorithm that decides whether a given formula is satisfiable?

## Answer

We do not know!

# What *do* we know about satisfiability?

## Truth assignment as witness

If the answer is “yes”, then a satisfying truth assignment can serve as a proof that the answer is indeed “yes”.

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## Witness for satisfiability

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## Checking the witness

We can quickly check whether indeed the witness assignment makes the formula true. This can be done in time proportional to the size of the formula.

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## Original definition

NP is the set of decision problems solvable in polynomial time by a non-deterministic Turing machine.

## Some History

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- In 1972, Richard Karp presented 21 mutually equivalent problems in NP, for which no polynomial time algorithms was known.
- Cook and Leonid Levin proved independently that propositional satisfiability is in this class (called NP-complete).

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- Many computer scientists assume  $P \neq NP$ , and therefore consider NP-complete problems as “intractable”.
- Many “proofs” for one or the other answer have been proposed, and subsequently rejected, most recently by Vinay Deolalikar (a researcher at HP), in August 2010.

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# Conjunctive Normal Form

## Definition

A literal  $L$  is either an atom  $p$  or the negation of an atom  $\neg p$ .  
A formula  $C$  is in *conjunctive normal form* (CNF) if it is a conjunction of clauses, where each clause is a disjunction of literals:

$$L ::= p \mid \neg p$$

$$D ::= L \mid L \vee D$$

$$C ::= D \mid D \wedge C$$

## Examples

$(\neg p \vee q \vee r) \wedge (\neg q \vee r) \wedge (\neg r)$  is in CNF.

$(\neg p \vee q \vee r) \wedge ((p \wedge \neg q) \vee r) \wedge (\neg r)$  is not in CNF.

$(\neg p \vee q \vee r) \wedge \neg(\neg q \vee r) \wedge (\neg r)$  is not in CNF.

## Usefulness of CNF

### Lemma

A disjunction of literals  $L_1 \vee L_2 \vee \dots \vee L_m$  is valid iff there are  $1 \leq i, j \leq m$  such that  $L_i$  is  $\neg L_j$ .

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## Proposition

Let  $\phi$  be a formula of propositional logic. Then  $\phi$  is satisfiable iff  $\neg\phi$  is not valid.

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## Satisfiability test

We can test satisfiability of  $\phi$  by transforming  $\neg\phi$  into CNF, and show that some clause is not valid.

# Transformation to CNF

## Theorem

Every formula in the propositional calculus can be transformed into an equivalent formula in CNF.

# Algorithm for CNF Transformation

- 1 Eliminate implication using:

$$A \rightarrow B \equiv \neg A \vee B$$

- 2 Push all negations inward using De Morgan's laws:

$$\neg(A \wedge B) \equiv (\neg A \vee \neg B)$$

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- 3 Eliminate double negations using the equivalence  $\neg\neg A \equiv A$

- 4 The formula now consists of disjunctions and conjunctions of literals. Use the distributive laws

$$A \vee (B \wedge C) \equiv (A \vee B) \wedge (A \vee C)$$

$$(A \wedge B) \vee C \equiv (A \vee C) \wedge (B \vee C)$$

to eliminate conjunctions within disjunctions.

## Example

$$\begin{aligned}(\neg p \rightarrow \neg q) \rightarrow (p \rightarrow q) &\equiv \neg(\neg\neg p \vee \neg q) \vee (\neg p \vee q) \\ &\equiv (\neg\neg\neg p \wedge q) \vee (\neg p \vee q) \\ &\equiv (\neg p \wedge q) \vee (\neg p \vee q) \\ &\equiv (\neg p \vee \neg p \vee q) \wedge (q \vee \neg p \vee q)\end{aligned}$$

## Algorithms for Proving Satisfiability of $\psi$

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propagation-based linear solver