

Decision Procedures  
An Algorithmic Point of View  
Basic Concepts and Background

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Part I

Basic Concepts and Background

Outline

- 1 What is Logic?
- 2 Deductive Systems
- 3 Soundness and Completeness
- 4 Decidability
- 5 Expressiveness
- 6 Background on Propositional Logic

Logic in Computer Science

- Reasoning in AI
- Proofs in verification
- Queries in Databases
- ... many more

What is Logic?

Some useful definitions from the web:

- "*Science dealing with the principles of valid reasoning and argument*"
- "*A formal and powerful method of explaining why the program doesn't work*"
- "*The art of being wrong with confidence*"

So what is Logic?

- Defined by
  - **Syntax**  
(including the Signature  $\Sigma$  of the logic: variables and their domain, function and predicate symbols, quantifiers, etc.)
  - **Semantics: Axioms and Inference rules**
- A logic allows us to *infer theorems*

## Example: Propositional Logic

- Syntax

formula : Boolean-var |  $\neg$ formula | formula  $\vee$  formula | (formula) | T | F

(Syntactic sugar: formula  $\wedge$  formula | formula  $\rightarrow$  formula ...)

- Axioms:

1.  $\vdash (A \rightarrow (B \rightarrow A))$
2.  $\vdash ((A \rightarrow (B \rightarrow C)) \rightarrow ((A \rightarrow B) \rightarrow (A \rightarrow C)))$
3.  $\vdash (\neg B \rightarrow \neg A) \rightarrow (A \rightarrow B)$

- (Single) Inference Rule: Modus Ponens (MP)

$$\frac{\vdash A \quad \vdash A \rightarrow B}{\vdash B}$$

A specific (one of many possible) Deductive System for Propositional Logic. It is known as the Hilbert System  $\mathcal{H}$ .

## A proof by deduction: example

- Notation:  $\vdash_{\mathcal{H}} \varphi$  'there exists a proof of  $\varphi$  in  $\mathcal{H}$ '

- Theorem:  $\vdash_{\mathcal{H}} (A \rightarrow B) \rightarrow ((B \rightarrow C) \rightarrow (A \rightarrow C))$

- Proof:

1.  $\{A \rightarrow B, B \rightarrow C, A\} \vdash_{\mathcal{H}} A$  Premise
2.  $\{A \rightarrow B, B \rightarrow C, A\} \vdash_{\mathcal{H}} A \rightarrow B$  Premise
3.  $\{A \rightarrow B, B \rightarrow C, A\} \vdash_{\mathcal{H}} B$  M.P. 1,2
4.  $\{A \rightarrow B, B \rightarrow C, A\} \vdash_{\mathcal{H}} B \rightarrow C$  Premise
5.  $\{A \rightarrow B, B \rightarrow C, A\} \vdash_{\mathcal{H}} C$  M.P. 3,4
6.  $\{A \rightarrow B, B \rightarrow C\} \vdash_{\mathcal{H}} (A \rightarrow C)$  Deduction 5
7.  $\{A \rightarrow B\} \vdash_{\mathcal{H}} ((B \rightarrow C) \rightarrow (A \rightarrow C))$  Deduction 6
8.  $\vdash_{\mathcal{H}} (A \rightarrow B) \rightarrow ((B \rightarrow C) \rightarrow (A \rightarrow C))$  Deduction 7

## More on Semantics

Can be given via

- axioms and inference rules, or
- using another (background) logic, or
- via **truth tables**:

$x_1$	$x_2$	$x_1 \wedge x_2$	$x_1 \vee x_2$	...
T	T	T	T	
T	F	F	T	
F	T	F	T	
F	F	F	F	

## Satisfying Interpretations

- If an assignment  $\alpha$  satisfies (according to the truth tables) a formula  $\varphi$ , we write:  $\alpha \models \varphi$ .

- Example:

$$\varphi : \neg(x_1 \wedge \neg(x_2 \vee \neg x_3))$$

- Assignments for the example:

- $\alpha_1 : (x_1 = T, x_2 = F, x_3 = F)$
- $\alpha_2 : (x_1 = T, x_2 = F, x_3 = T)$
- $\alpha_1 \models \varphi$ , but  $\alpha_2 \not\models \varphi$



## Satisfiability, Validity, etc.

### Definition (Satisfiable)

A formula  $\varphi$  is *satisfiable* if  $\exists \alpha. \alpha \models \varphi$ .

### Definition (Valid)

A formula  $\varphi$  is *valid* if  $\forall \alpha. \alpha \models \varphi$ . If  $\varphi$  is valid, we write  $\models \varphi$ .

**Observation:**  $\varphi$  is valid if and only if  $\neg \varphi$  is unsatisfiable.

## A proof by enumeration: same example

A	B	C	$(A \rightarrow B) \rightarrow ((B \rightarrow C) \rightarrow (A \rightarrow C))$
T	T	T	T
T	T	F	T
T	F	T	T
T	F	F	T
F	T	T	T
F	T	F	T
F	F	T	T
F	F	F	T

$$\models (A \rightarrow B) \rightarrow ((B \rightarrow C) \rightarrow (A \rightarrow C))$$

## Soundness and completeness of a deductive system

Given a deductive system  $\mathcal{D}$ ,

- $\mathcal{D}$  is **sound** for a logic  $\mathcal{L}$ , if for every formula  $f$  in  $\mathcal{L}$ ,

$$\vdash_{\mathcal{D}} f \longrightarrow \models f$$

I.e., all formulas proven by the deductive system are valid.

- $\mathcal{D}$  is **complete** if for every formula  $f$  in  $\mathcal{L}$ ,

$$\models f \longrightarrow \vdash_{\mathcal{D}} f$$

I.e., the deductive system can prove all valid formulas.

## The decision problem

### Definition (Decision Problem)

The decision problem for a formula: given  $\varphi$ , is  $\varphi$  valid?

### Definition (Decision Procedure)

A *decision procedure* for a logic is an algorithm that solves the decision problem for any formula in this logic.

We are naturally interested in a sound and complete decision procedure.

## Soundness and Completeness

What does it mean that a decision procedure is sound and complete?

- **Soundness**: the answer returned by the decision procedure is always correct  
(Question: "correct" according to what?)
- **Completeness**: returns with a yes/no answer in finite time.  
(Question: How does this definition relate to the definition of completeness of a deduction system?)

## Soundness and Completeness

- **Soundness**: "when I say that it rains, it rains, and when I say it doesn't rain, it doesn't rain"
- **Completeness**: "When asked, I always reply (in a finite time) whether it rains"



### Definition

A logic is **decidable**  $\iff$  there is a sound and complete algorithm that decides if a well-formed expression in this logic is valid.

## Soundness and Completeness (cont'd)

Algorithm #1: for checking if it rains outside:  
**stand right outside the door and say 'it rains'**



- It is *not sound* because you might say it rains when it doesn't.
- But it is complete: you *always get an answer* in a finite time.

## Soundness and Completeness (cont'd)

Algorithm #2 for checking if it rains outside:  
**stand right outside the door and say 'it rains'**  
**if and only if you feel the rain.**



- It is *sound* because you say it rains only if it actually rains.
- It is incomplete because you do not say anything if it doesn't rain (we do not know whether it doesn't rain, or it takes the person too long to answer ...).

## Decidability

- Propositional logic is *decidable*  
⇒ there is a sound and complete algorithm (e.g., truth tables) to decide whether a propositional formula is valid.
- Arithmetic over integers is ...? ...*undecidable*  
(this is Gödel's incompleteness result)

## Inference engines

- We saw that in Propositional Logic we can infer using either a deductive system ('**deduction**') or truth tables ('**enumeration**').
- Which, in the general case, is the better method?
- All logics have a deductive definition.
- NOT all logics can be decided with an enumerative method.

## Deduction vs. enumeration

**Deductive methods**  
Axioms and inference rules



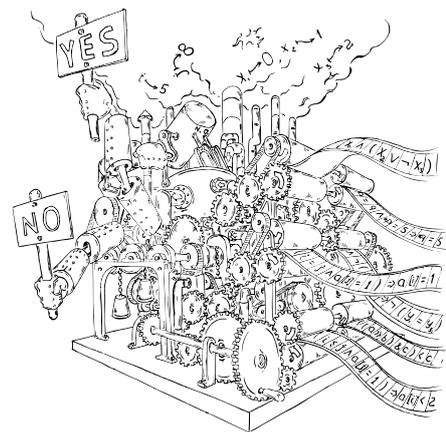
Requires thinking...

**Enumerative methods**  
Truth tables



Requires pressing 'Enter'...

Whenever we can: build an engine to think for us



## Expressiveness of a logic

- Each formula defines a **language**:  
the set of satisfying assignments ('models') are the words accepted by this language.
- Consider the logic '2-CNF'

*formula* : *literal* ∨ *literal* | *formula* ∧ *formula*  
*literal* : *Boolean-variable* | ¬*Boolean-variable*

- A '2-CNF' formula:  $(x_1 \vee \neg x_2) \wedge (\neg x_3 \vee x_2)$

## Expressiveness of a logic

- Now consider the Propositional Logic formula

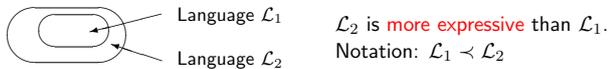
$$\varphi : x_1 \vee x_2 \vee x_3$$

- Q: Can we express this language with 2-CNF?
- A: No.

### Proof.

The language accepted by  $\varphi$  has 7 words: all assignments other than  $x_1 = x_2 = x_3 = \text{F}$ . The first 2-CNF clause removes  $\frac{1}{4}$  of the assignments, which leaves us with 6 accepted words. Additional clauses only remove more assignments. □

## Expressiveness of a logic



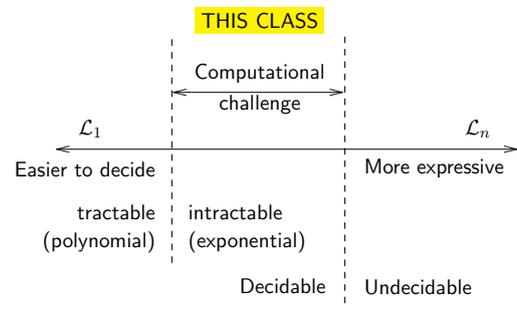
### Claim

$2\text{-CNF} < \text{Propositional Logic}$

Generally, there is only a **partial order** between logics.

## Tradeoff: expressiveness/computational hardness

Assume we are given logics  $\mathcal{L}_1 < \dots < \mathcal{L}_n$



## Expressiveness and complexity

- Q1: Let  $\mathcal{L}_1$  and  $\mathcal{L}_2$  be two logics whose satisfiability problem is **decidable** and in the **same complexity class**.  
Is the satisfiability problem of an  $\mathcal{L}_1$  formula reducible to a satisfiability problem of an  $\mathcal{L}_2$  formula?
- Q2: Let  $\mathcal{L}_1$  and  $\mathcal{L}_2$  be two logics whose satisfiability problems are reducible to one another.

Are  $\mathcal{L}_1$  and  $\mathcal{L}_2$  in the same complexity class?

## When is a specific logic useful?

- Expressible enough** to state something interesting.
- Decidable (or semi-decidable) and **more efficiently solvable** than richer logics.
- More expressive**, or more natural for expressing some models in comparison to 'leaner' logics.

## Example: First Order Peano Arithmetic

- Constants: 0, 1
- Function symbols: '+', '\*', predicate symbol: '='
- Domain: natural numbers
- Semantics given by axioms:
  - $\forall x : (0 \neq x + 1)$
  - $\forall x, y : (x \neq y) \rightarrow (x + 1 \neq y + 1)$
  - Induction
  - $\forall x : x + 0 = x$
  - $\forall x, y : (x + y) + 1 = x + (y + 1)$
  - $\forall x : x * 0 = 0$
  - $\forall x, y : x * (y + 1) = x * y + x$

**UNDECIDABLE!**

## Example: Presburger Arithmetic

- Constants: 0, 1
- Function symbols: '+', predicate symbol: '='
- Domain: natural numbers
- Semantics given by axioms:
  - $\forall x : (0 \neq x + 1)$
  - $\forall x, y : (x \neq y) \rightarrow (x + 1 \neq y + 1)$
  - Induction
  - $\forall x : x + 0 = x$
  - $\forall x, y : (x + y) + 1 = x + (y + 1)$
  - ~~$\forall x : x * 0 = 0$~~
  - ~~$\forall x, y : x * (y + 1) = x * y + x$~~

**DECIDABLE!**

## Some notes on Propositional Logic

- The simplest of them all
- NP-complete
- Exceptionally efficient solvers (SAT engines, BDDs)
- Formulas with  $10^5$  variables are being solved regularly
- All the logics that we will consider can be reduced directly to this logic

## Some notes on Propositional Logic

$v$                        $\neg v$   
positive literal    negative literal

- Also known as 'the phase', or 'the polarity' of the literal.
- The 'logical phase' of a literal can be computed by counting the number of negations that nest it.
  - $v$  is logically negative in:  
 $\neg v, \neg(\neg(\neg v)), v \rightarrow u, \neg(u \rightarrow v)$
  - $v$  is logically positive in:  
 $v, \neg(v \rightarrow u)$

## Normal forms

- **Conjunctive Normal Form** (CNF)

$$\bigwedge \bigvee l_i$$

- **Disjunctive Normal Form** (DNF)

$$\bigvee \bigwedge l_i$$

Satisfiability is in  $P$ !

- **Negation Normal Form** (NNF)  
(all negations are over literals, not sub-formulas)

Note that CNF and DNF are special cases of NNF.

## Normal forms

Conversion into normal forms:

- Convert  $\varphi$  to a CNF: with additional variables, in polynomial time
- Convert  $\varphi$  to DNF: exponential time and space
- Convert  $\varphi$  to NNF: polynomial time

## The 'Pure literal rule'

- Consider  $\varphi : (x \vee y) \wedge (\neg x \vee z) \wedge (x \vee y \vee \neg z)$
- $y$  is 'pure': it only appears in one phase
- Idea: when trying to satisfy  $\varphi$ , first assign  $y = \text{true}$
  
- Why? If there is a satisfying assignment to  $\varphi$ , there is a satisfying assignment in which  $y = \text{true}$ .
- Generalization: assign all pure literals according to their phase

## Pure literals in NNF

- CNF is a special case of NNF
- A pure literal in NNF is defined in the same way:  
a literal that only appears in one phase.
- We can always start satisfiability checking by assigning these pure literals true or false according to their phase.
- We will rely on a similar principle also when considering other logics.

Theorem

*NNF formulas are monotonically satisfied.*

In case of CNF, this is simply the pure literal rule.

Example

$$\begin{aligned} \varphi &: (x_1 \wedge \neg x_2) \vee (x_2 \vee (x_3 \wedge x_1)) \\ \alpha &: 0 \quad 1 \quad 1 \quad 1 \quad 0 \\ \alpha' &: 1 \quad \quad \quad \quad 1 \end{aligned}$$

Why is monotonicity relevant to decision procedures?

- We will use the fact that if we make unsatisfied predicates satisfied, we do not make the formula unsatisfied.
- We will rely heavily on this fact later: it simplifies decision procedures.