

PROPOSITIONAL LOGIC (2)

based on

Huth & Ruan

Logic in Computer Science:
Modelling and Reasoning about Systems
Cambridge University Press, 2004

Russell & Norvig

Artificial Intelligence:
A Modern Approach
Prentice Hall, 2010

Clauses

- Clauses are formulas consisting only of \vee and \neg

$$p \vee q \vee \neg r$$
$$\neg p \vee \neg q$$

(brackets within a clause are not allowed!)

they can also be written using \rightarrow , \vee (after \rightarrow) and \wedge (before \rightarrow)

$$r \rightarrow p \vee q$$
$$p \wedge q \rightarrow \perp$$
$$\top \rightarrow p \vee q$$
$$\top \rightarrow \perp$$

Clause without positive literal

Clause without negative literal

Empty clause is considered *false*

an atom or its negation is called a *literal*

Conjunctive & Disjunctive Normal Form

- A formula is in **conjunctive normal form** if it consists of a conjunction of clauses

$$(p \vee q \vee \neg r) \wedge (p \vee \neg q) \wedge (p \vee r)$$
$$(r \rightarrow p \vee q) \wedge (q \rightarrow p) \wedge (\top \rightarrow p \vee r)$$

- “conjunction of disjunctions”
- A formula is in **disjunctive normal form** if it consists of a disjunction of conjunctions

$$(p \wedge q \wedge \neg r) \vee (p \wedge \neg q) \vee (p \vee r)$$

Conjunctive & Disjunctive Normal Form

- The transformation from CNF to DNF is exponential

$$(p_1 \vee q_1) \wedge (p_2 \vee q_2) \wedge (p_3 \vee q_3) = \begin{aligned} & (p_1 \wedge p_2 \wedge p_3) \vee \\ & (p_1 \wedge p_2 \wedge q_3) \vee \\ & (p_1 \wedge q_2 \wedge p_3) \vee \\ & (p_1 \wedge q_2 \wedge q_3) \vee \\ & (q_1 \wedge p_2 \wedge p_3) \vee \\ & (q_1 \wedge p_2 \wedge q_3) \vee \\ & (q_1 \wedge q_2 \wedge p_3) \vee \\ & (q_1 \wedge q_2 \wedge q_3) \end{aligned}$$

Conjunctive Normal Form

- Any formula can be written in CNF

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

(consequently, any formula can also be written in DNF, but the DNF formula may be exponentially larger)

Checking Satisfiability of Formulas in DNF

- Checking DNF satisfiability is easy: process one conjunction at a time; if at least one conjunction is not a contradiction, the formula is satisfiable
 - DNF satisfiability can be decided in polynomial time

$$(p_1 \wedge p_3 \wedge \neg p_3) \vee$$

$$(p_1 \wedge \neg p_2 \wedge \neg p_3) \vee$$

$$(p_1 \wedge \neg p_2 \wedge p_3) \vee$$

$$(\neg p_1 \wedge p_3 \wedge \neg p_3) \vee$$

Conversion to DNF is not feasible in most cases
(exponential blowup)

Checking Satisfiability of Formulas in CNF

- No polynomial algorithm is known for checking the satisfiability of arbitrary CNF formulas

Example:

we could use such an algorithm to solve graph coloring with k colors

- for each node i , create a formula

$$\phi_i = p_{i1} \vee p_{i2} \vee \cdots \vee p_{ik}$$

indicating that each node i must have a color

- for each node i and different pair of colors c_1 and c_2 , create a formula

$$\phi_{ic_1c_2} = \neg(p_{ic_1} \wedge p_{ic_2}) = \neg p_{ic_1} \vee \neg p_{ic_2}$$

indicating a node may not have more than 1 color

- for each edge, create k formulas

$$\phi_{ijc} = \neg(p_{ic} \wedge p_{jc}) = \neg p_{ic} \vee \neg p_{jc}$$

indicating that a pair connected nodes i and j may not both have color c at the same time

“At-most-once” constraint

- Let us have variables x_1, \dots, x_n and require that at most one of these variables is one
- Constraints on the previous slide:

$$(\neg x_1 \vee \neg x_2) \wedge (\neg x_1 \vee \neg x_3) \wedge (\neg x_1 \vee \neg x_n) \wedge \dots \wedge (\neg x_{n-1} \vee \neg x_n)$$

→ $n(n - 1)/2$ clauses in total

- We can do better...

“At-most-once” constraint

- Introduce additional variables a_1, \dots, a_n
 - Idea: let a_i be true if one of x_1, \dots, x_i is true
 - Formally:
 - $\neg a_i \vee \neg x_{i+1}$ (a_i and x_{i+1} may not be true at the same time)
 - $\neg a_i \vee a_{i+1}$ (if a_i is true, then a_{i+1} is true)
 - $\neg x_i \vee a_i$ (if x_i is true, then a_i is true)
- for all $1 \leq i \leq n - 1$
- $3(n-1)$ clauses in total!

Resolution Rule

Essential in most satisfiability solvers for CNF formulas is the **resolution rule** for clauses:

Given two clauses $l_1 \vee \dots \vee l_k$ and $m_1 \vee \dots \vee m_n$ where $l_1, \dots, l_k, m_1, \dots, m_n$ represent literals and it holds that $l_i = \neg m_j$, then it holds that

$$l_1 \vee \dots \vee l_k, m_1 \vee \dots \vee \dots \vee m_n \vdash_R \\ l_1 \vee \dots \vee l_{i-1} \vee l_{i+1} \vee \dots \vee l_k \vee m_1 \vee \dots \vee m_{j-1} \vee m_{j+1} \vee \dots \vee m_n$$

Example: $p \vee q \vee \neg r, r \vee s \vdash_R p \vee q \vee s$
 $r \rightarrow p \vee q, r \vee s \vdash_R p \vee q \vee s$

Proof for Resolution

on an example

1.	$p \vee q$	premise
2.	$q \rightarrow r$	premise
3.	p	assumption
4.	$p \vee r$	$\vee i$ 3
5.	q	assumption
6.	r	$\rightarrow e$ 2,5
7.	$p \vee r$	$\vee i$ 6
8.	$p \vee r$	$\vee e$ 1,3-4, 5-7

$\neg q \vee r$

Completeness of Resolution

- If it holds that $C_1, \dots, C_n \models \perp$ for clauses C_1, \dots, C_n (i.e. the clauses are a contradiction), then we can derive \perp from C_1, \dots, C_n by repeated application of the resolution rule

$$\begin{array}{rcl} p, p \rightarrow q \vee r, q \rightarrow \perp, r \rightarrow \perp & \vdash_R & q \vee r, q \rightarrow \perp, r \rightarrow \perp \\ & \vdash_R & r, r \rightarrow \perp \\ & \vdash_R & \perp \end{array}$$

How to find the resolution steps in general?
For some types of clauses it is easier...

Definite clauses & Horn clauses

- A **definite clause** is a clause with exactly one positive literal

$$p, q, p \wedge q \rightarrow t$$

- A **horn clause** is a clause with at most one positive literal

$$p, q, p \wedge q \rightarrow t, p \wedge q \rightarrow \perp$$

A clause with one positive literal is called a **fact**

Forward chaining for Definite clauses

- The forward chaining algorithm calculates facts that can be entailed from a set of definite clauses

C = initial set of definite clauses

repeat

if there is a clause $p_1, \dots, p_n \rightarrow q$ in C where p_1, \dots, p_n are
 facts in C **then**

 add fact q to C ←

end if

until no fact could be added

return all facts in C



Resolution

This algorithm is complete for facts: any fact that is entailed, will be derived.

Forward chaining for Horn clauses

- We now also allow to add \perp and other clauses without positive literals to \mathcal{C}
- We stop immediately \perp when is found, and return that the set of formulas is contradictory.

$$\mathbf{C}_1 = \{p, p \rightarrow q, p \wedge q \rightarrow r, r \rightarrow \perp\}$$

$$\mathbf{C}_2 = \{p, q, p \rightarrow q, p \wedge q \rightarrow r, r \rightarrow \perp\}$$

$$\mathbf{C}_3 = \{p, q, r, p \rightarrow q, p \wedge q \rightarrow r, r \rightarrow \perp\}$$

$$\mathbf{C}_4 = \{p, q, r, \perp, p \rightarrow q, p \wedge q \rightarrow r, r \rightarrow \perp\}$$

Note:

- 1) a set of definite clauses is always satisfiable.
- 2) we can decide in linear time whether a set of Horn clauses is satisfiable.

Deciding entailment for Horn clauses

- Suppose we would like to know whether

$$C_1, \dots, C_n \models p_1, \dots, p_n \rightarrow q$$

where C_1, \dots, C_n are Horn clauses; then it suffices to determine whether

$$C_1, \dots, C_n, p_1, \dots, p_n \vdash_R q$$

(we can show this by means of \rightarrow introduction)

- As entailment of facts can be decided in linear time, Horn clause entailment can be determined in linear time as well