

Artificial Intelligence

Propositional logic:
inference algorithms

Outline

- Propositional (Boolean) logic
- Equivalence, validity, satisfiability
- Inference rules and theorem proving
 - forward chaining
 - backward chaining
 - resolution

Propositional logic: Syntax

- Propositional logic is the simplest logic – illustrates basic ideas
- The proposition symbols P_1, P_2 etc are sentences
 - If S is a sentence, $\neg S$ is a sentence (*negation*)
 - If S_1 and S_2 are sentences, $S_1 \wedge S_2$ is a sentence (*conjunction*)
 - If S_1 and S_2 are sentences, $S_1 \vee S_2$ is a sentence (*disjunction*)
 - If S_1 and S_2 are sentences, $S_1 \Rightarrow S_2$ is a sentence (*implication*)
 - If S_1 and S_2 are sentences, $S_1 \Leftrightarrow S_2$ is a sentence (*biconditional*)

3

Truth tables for connectives

P	Q	$\neg P$	$P \wedge Q$	$P \vee Q$	$P \Rightarrow Q$	$P \Leftrightarrow Q$
<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>true</i>
<i>false</i>	<i>true</i>	<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>	<i>false</i>
<i>true</i>	<i>false</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>
<i>true</i>	<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>	<i>true</i>	<i>true</i>

4

Logical equivalence

- Two sentences are **logically equivalent** iff true in same models: $\alpha \equiv \beta$ iff $\alpha \models \beta$ and $\beta \models \alpha$

$$(\alpha \wedge \beta) \equiv (\beta \wedge \alpha) \quad \text{commutativity of } \wedge$$

$$(\alpha \vee \beta) \equiv (\beta \vee \alpha) \quad \text{commutativity of } \vee$$

$$((\alpha \wedge \beta) \wedge \gamma) \equiv (\alpha \wedge (\beta \wedge \gamma)) \quad \text{associativity of } \wedge$$

$$((\alpha \vee \beta) \vee \gamma) \equiv (\alpha \vee (\beta \vee \gamma)) \quad \text{associativity of } \vee$$

$$\neg(\neg\alpha) \equiv \alpha \quad \text{double-negation elimination}$$

$$(\alpha \Rightarrow \beta) \equiv (\neg\beta \Rightarrow \neg\alpha) \quad \text{contraposition}$$

$$(\alpha \Rightarrow \beta) \equiv (\neg\alpha \vee \beta) \quad \text{implication elimination}$$

$$(\alpha \Leftrightarrow \beta) \equiv ((\alpha \Rightarrow \beta) \wedge (\beta \Rightarrow \alpha)) \quad \text{biconditional elimination}$$

$$\neg(\alpha \wedge \beta) \equiv (\neg\alpha \vee \neg\beta) \quad \text{de Morgan}$$

$$\neg(\alpha \vee \beta) \equiv (\neg\alpha \wedge \neg\beta) \quad \text{de Morgan}$$

$$(\alpha \wedge (\beta \vee \gamma)) \equiv ((\alpha \wedge \beta) \vee (\alpha \wedge \gamma)) \quad \text{distributivity of } \wedge \text{ over } \vee$$

$$(\alpha \vee (\beta \wedge \gamma)) \equiv ((\alpha \vee \beta) \wedge (\alpha \vee \gamma)) \quad \text{distributivity of } \vee \text{ over } \wedge$$

5

Validity and satisfiability

A sentence is **valid** if it is true in **all** models,

e.g., *True*, $A \vee \neg A$, $A \Rightarrow A$, $(A \wedge (A \Rightarrow B)) \Rightarrow B$

Validity is connected to inference via the **Deduction Theorem**:

$KB \models \alpha$ if and only if $(KB \Rightarrow \alpha)$ is valid

A sentence is **satisfiable** if it is true in **some** model

e.g., $A \vee B$, C

A sentence is **unsatisfiable** if it is true in **no** models

e.g., $A \wedge \neg A$

Satisfiability is connected to inference via the following:

$KB \models \alpha$ if and only if $(KB \wedge \neg\alpha)$ is unsatisfiable

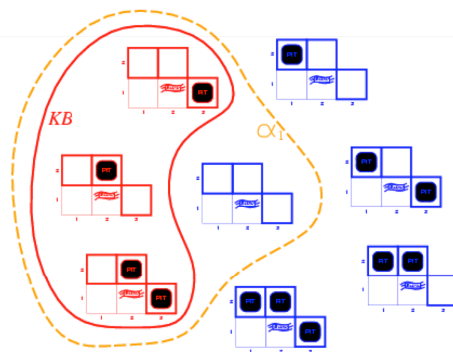
6

Example: Wumpus world KB

- Let $P_{i,j}$ be true if there is a pit in $[i, j]$.
- Let $B_{i,j}$ be true if there is a breeze in $[i, j]$.
- Pits cause breezes in adjacent squares
- The KB has 5 sentences:
 - $\neg P_{1,1}$
 - $\neg B_{1,1}$
 - $B_{2,1}$
 - $B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})$
 - $B_{2,1} \Leftrightarrow (P_{1,1} \vee P_{2,2} \vee P_{3,1})$

7

Wumpus models



- KB = wumpus-world rules + observations
- α_1 = "[1,2] is safe", $KB \models \alpha_1$, proved by **model checking**

8

Truth tables for inference

$B_{1,1}$	$B_{2,1}$	$P_{1,1}$	$P_{1,2}$	$P_{2,1}$	$P_{2,2}$	$P_{3,1}$	KB	α_1
false	false	false	false	false	false	false	false	true
false	false	false	false	false	false	true	false	true
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
false	true	false	false	false	false	false	false	true
false	true	false	false	false	false	true	<u>true</u>	<u>true</u>
false	true	false	false	false	true	false	<u>true</u>	<u>true</u>
false	true	false	false	false	true	true	<u>true</u>	<u>true</u>
false	true	false	false	true	false	false	false	true
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
true	true	true	true	true	true	true	false	false

9

Inference by enumeration

```

function TT-ENTAILS?( $KB, \alpha$ ) returns true or false
     $symbols \leftarrow$  a list of the proposition symbols in  $KB$  and  $\alpha$ 
    return TT-CHECK-ALL( $KB, \alpha, symbols, []$ )

function TT-CHECK-ALL( $KB, \alpha, symbols, model$ ) returns true or false
    if EMPTY?( $symbols$ ) then
        if PL-TRUE?( $KB, model$ ) then return PL-TRUE?( $\alpha, model$ )
        else return true
    else do
         $P \leftarrow$  FIRST( $symbols$ );  $rest \leftarrow$  REST( $symbols$ )
        return TT-CHECK-ALL( $KB, \alpha, rest, EXTEND(P, true, model)$ ) and
            TT-CHECK-ALL( $KB, \alpha, rest, EXTEND(P, false, model)$ )
    
```

- For n symbols, time complexity is $O(2^n)$, space complexity is $O(n)$

10

Proof methods

- Proof methods divide into (roughly) two kinds:
 - **Application of inference rules**
 - Legitimate (sound) generation of new sentences from old
 - **Proof** = a sequence of inference rule applications
 - Can use inference rules as operators in a standard search algorithm
 - Typically require transformation of sentences into a **normal form**
 - **Model checking**
 - truth table enumeration (always exponential in n)
 - improved backtracking, e.g., Davis--Putnam-Logemann-Loveland (DPLL) algorithm
 - heuristic search in model space (sound but incomplete)

11

Inference rule: Resolution

- **Conjunctive Normal Form (CNF)** : **conjunction of disjunctions of literals**

E.g., $(A \vee \neg B) \wedge (B \vee \neg C \vee \neg D)$

- **Resolution** inference rule (for CNF):

$$\begin{array}{c} \ell_i \vee \dots \vee \ell_k, \quad m_1 \vee \dots \vee m_n \\ \hline \ell_i \vee \dots \vee \ell_{i-1} \vee \ell_{i+1} \vee \dots \vee \ell_k \vee m_1 \vee \dots \vee m_{i-1} \vee m_{i+1} \vee \dots \vee m_n \\ \text{where } \ell_i \text{ and } m_i \text{ are complementary literals.} \end{array}$$

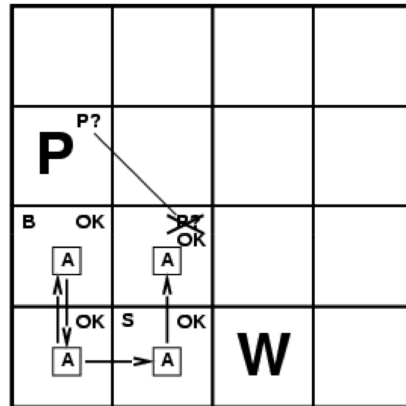
- Resolution is sound and complete for propositional logic

12

Resolution example

$$P_{1,3} \vee P_{2,2}, \neg P_{2,2}$$

$$P_{1,3}$$



13

Conversion to CNF

Convert $B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})$ into CNF:

1. Eliminate \Leftrightarrow , replacing $\alpha \Leftrightarrow \beta$ with $(\alpha \Rightarrow \beta) \wedge (\beta \Rightarrow \alpha)$.
 $(B_{1,1} \Rightarrow (P_{1,2} \vee P_{2,1})) \wedge ((P_{1,2} \vee P_{2,1}) \Rightarrow B_{1,1})$
2. Eliminate \Rightarrow , replacing $\alpha \Rightarrow \beta$ with $\neg\alpha \vee \beta$.
 $(\neg B_{1,1} \vee P_{1,2} \vee P_{2,1}) \wedge (\neg(P_{1,2} \vee P_{2,1}) \vee B_{1,1})$
3. Move \neg inwards using de Morgan's rules and double-negation:
 $(\neg B_{1,1} \vee P_{1,2} \vee P_{2,1}) \wedge ((\neg P_{1,2} \vee \neg P_{2,1}) \vee B_{1,1})$
4. Apply distributivity law (\wedge over \vee) and flatten:
 $(\neg B_{1,1} \vee P_{1,2} \vee P_{2,1}) \wedge (\neg P_{1,2} \vee B_{1,1}) \wedge (\neg P_{2,1} \vee B_{1,1})$

14

Resolution algorithm

- Proof by contradiction: To show $KB \models \alpha$, we show that $KB \wedge \neg\alpha$ is unsatisfiable
- First, $KB \wedge \neg\alpha$ is converted into CNF. Then, the resolution rule is applied to the resulting clauses. Each pair that contains complementary literals is resolved to produce a new clause, which is added to the set if it is not already present. The process continues until one of two things happens:
 - there are no new clauses that can be added, in which case KB does not entail α ; or,
 - Two clauses resolve to yield the empty clause, in which case KB entails α .

15

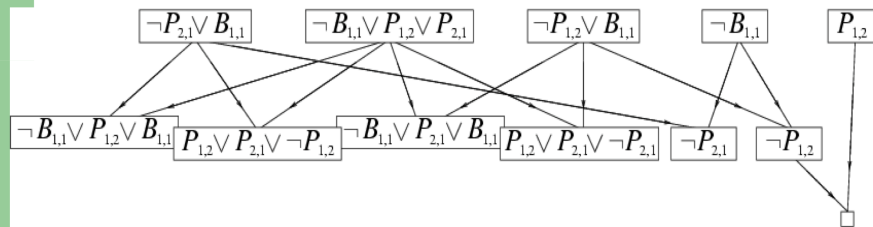
Resolution algorithm

```
function PL-RESOLUTION( $KB, \alpha$ ) returns true or false
   $clauses \leftarrow$  the set of clauses in the CNF representation of  $KB \wedge \neg\alpha$ 
   $new \leftarrow \{\}$ 
  loop do
    for each  $C_i, C_j$  in  $clauses$  do
       $resolvents \leftarrow$  PL-RESOLVE( $C_i, C_j$ )
      if  $resolvents$  contains the empty clause then return true
       $new \leftarrow new \cup resolvents$ 
    if  $new \subseteq clauses$  then return false
   $clauses \leftarrow clauses \cup new$ 
```

16

Resolution example

- $KB = (B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})) \wedge \neg B_{1,1}$
- $\alpha = \neg P_{1,2}$



17

Horn clauses

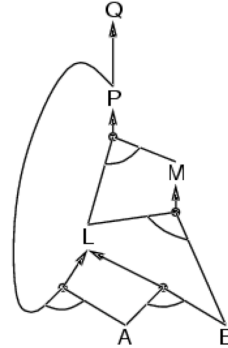
- Real-world knowledge bases often contain only clauses of a restricted kind called **Horn clauses**.
- Horn clause =
 - proposition symbol (**fact**); or
 - (conjunction of symbols) (called **body**) \Rightarrow symbol (called **head**)
 - E.g., $C \wedge D \Rightarrow B$
- Can be used with **forward chaining** or **backward chaining**.
- These algorithms are very natural and run in **linear** time

18

Forward chaining

- Idea: fire any rule whose premises are satisfied in the KB, add its conclusion to the KB, until query is found

$P \Rightarrow Q$
 $L \wedge M \Rightarrow P$
 $B \wedge L \Rightarrow M$
 $A \wedge P \Rightarrow L$
 $A \wedge B \Rightarrow L$
 A
 B



19

Forward chaining algorithm

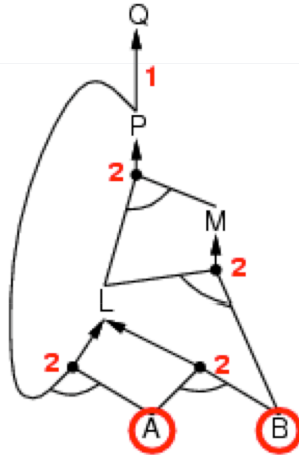
```
function PL-FC-ENTAILS?(KB, q) returns true or false
  local variables: count, a table, indexed by clause, initially the number of premises
                  inferred, a table, indexed by symbol, each entry initially false
                  agenda, a list of symbols, initially the symbols known to be true

  while agenda is not empty do
    p ← POP(agenda)
    unless inferred[p] do
      inferred[p] ← true
      for each Horn clause c in whose premise p appears do
        decrement count[c]
        if count[c] = 0 then do
          if HEAD[c] = q then return true
          PUSH(HEAD[c], agenda)
  return false
```

- Forward chaining is sound and complete for Horn KB

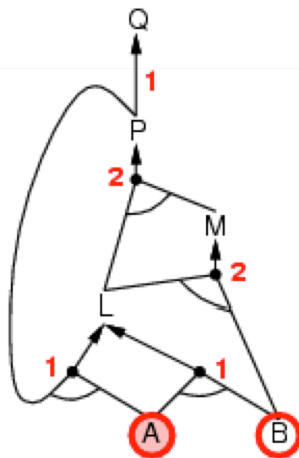
20

Forward chaining example



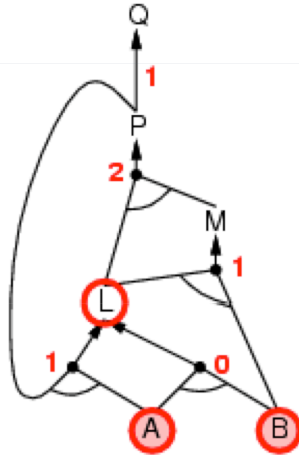
21

Forward chaining example



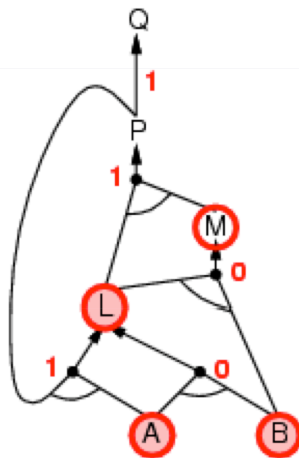
22

Forward chaining example



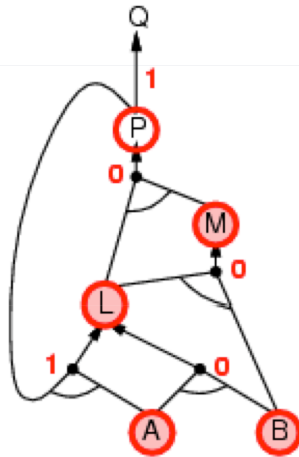
23

Forward chaining example



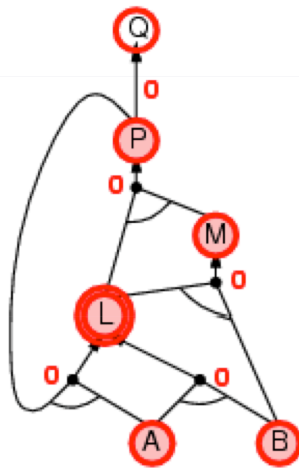
24

Forward chaining example



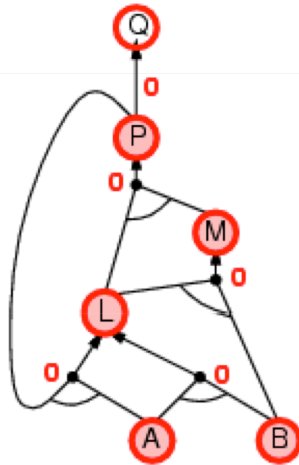
25

Forward chaining example



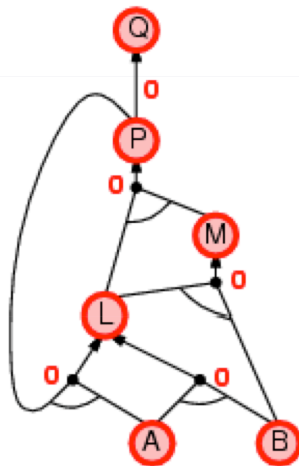
26

Forward chaining example



27

Forward chaining example



28

Proof of completeness

- FC derives every atomic sentence that is entailed by KB
 1. FC reaches a **fixed point** where no new atomic sentences are derived
 2. Consider the final state as a model m , assigning true/false to symbols
 3. Every clause in the original KB is true in m
$$a_1 \wedge \dots \wedge a_k \Rightarrow b$$
 4. Hence m is a model of KB
 5. If $KB \models q$, q is true in **every** model of KB , including m

29

Backward chaining

Idea: work backwards from the query q :

to prove q by BC,

check if q is known already, or

prove by BC all premises of some rule concluding q

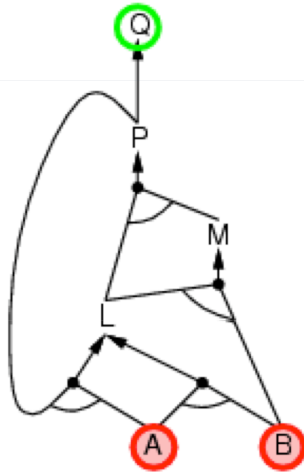
Avoid loops: check if new subgoal is already on the goal stack

Avoid repeated work: check if new subgoal

1. has already been proved true, or
2. has already failed

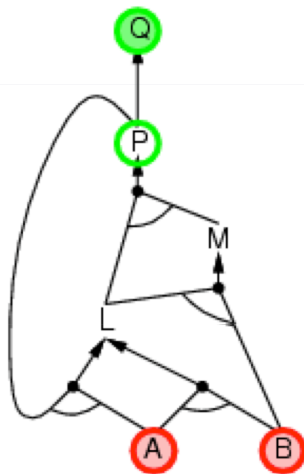
30

Backward chaining example



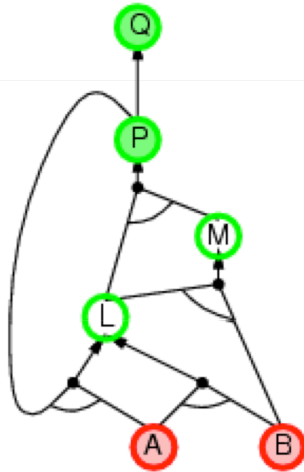
31

Backward chaining example



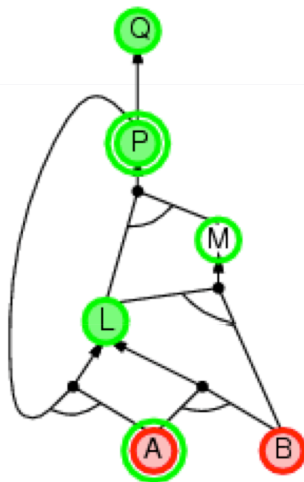
32

Backward chaining example



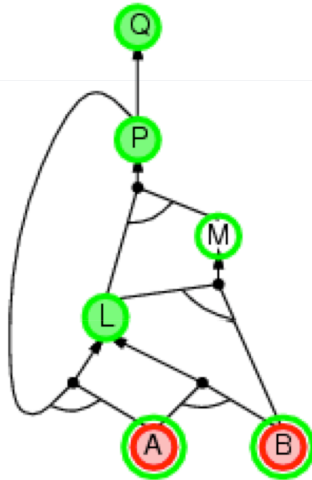
33

Backward chaining example



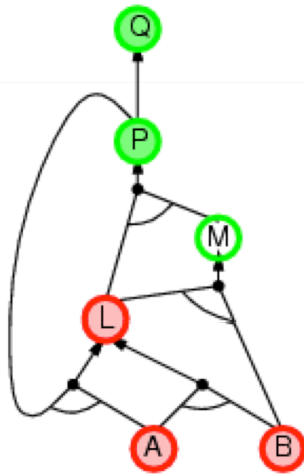
34

Backward chaining example



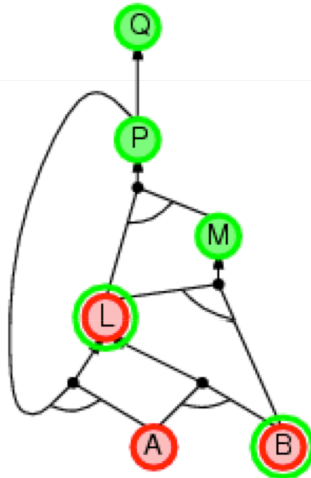
35

Backward chaining example



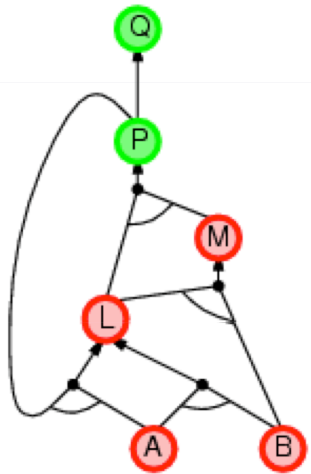
36

Backward chaining example



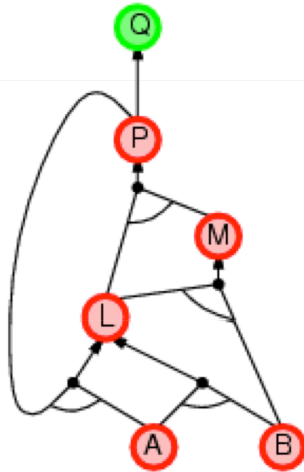
37

Backward chaining example



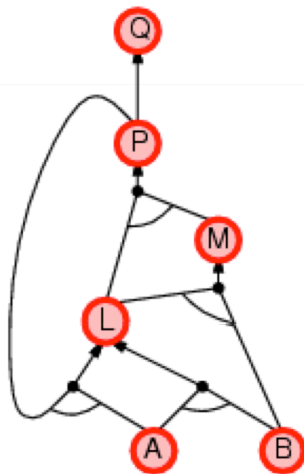
38

Backward chaining example



39

Backward chaining example



40

Forward vs. backward chaining

- FC is **data-driven reasoning**. It can be used within an agent to derive conclusions from incoming percepts, often without a specific query in mind.
- FC may do lots of work that is irrelevant to the goal
- BC is **goal-driven**, appropriate for problem-solving,
 - e.g., Where are my keys? What shall I do now?
- Complexity of BC can be **much less** than linear in size of KB

41

Expressiveness limitation of propositional logic

- In Wumpus world, KB must contain sentences for every single square, not “general rules”
- Propositional logic does not scale to environments of unbounded size because it lacks the expressive power to deal concisely with time, space, and universal patterns of relationships among objects.

42

Summary

- Logical agents apply **inference** to a **knowledge base** to derive new information and make decisions
- Basic concepts of logic:
 - **syntax**: formal structure of **sentences**
 - **semantics**: **truth** of sentences wrt **models**
 - **entailment**: necessary truth of one sentence given another
 - **inference**: deriving sentences from other sentences
 - **soundness**: derivations produce only entailed sentences
 - **completeness**: derivations can produce all entailed sentences
- Resolution is complete for propositional logic
- Forward, backward chaining are linear-time, complete for Horn clauses
- Propositional logic lacks expressive power