# Logical Agent & Propositional Logic



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References:

- 1. S. Russell and P. Norvig. Artificial Intelligence: A Modern Approach. Chapter 7
- 2. S. Russell's teaching materials

#### Introduction

- The representation of knowledge and the processes of reasoning will be discussed
  - Important for the design of artificial agents
    - Reflex agents
      - Rule-based, table-lookup
    - Problem-solving agents
      - Problem-specific and inflexible
    - Knowledge-based agents
      - Flexible
      - Combine knowledge with current percepts to infer hidden aspects of the current state prior to selecting actions
  - Logic is the primary vehicle for knowledge representation
  - Reasoning copes with different infinite variety of problem states using a finite store of knowledge

## Introduction (cont.)

• Example: Natural Language Understanding

John saw the diamond through the window and coveted it

John threw the brick through the window and broke it

 Understanding natural language requires inferring the intention of the speaker

## Knowledge-Based Agents

- Knowledge base (background knowledge)
  - A set of sentences of formal (or knowledge representation) language
     is a declarative approach
    - Represent facts (assertions) about the world
  - Sentences have their syntax and semantics
- Declarative approach to building an agent
  - Tell: tell it what it needs to know
- (add new sentences to KB)

(query what is known)

Ask: ask itself what to do



- Inference
  - Derive new sentences from old ones

#### Knowledge-Based Agents (cont.)



- KB initially contains some background knowledge
- Each time the agent function is called

the internal state

- It Tells KB whit it perceives
- It Asks KB what action it should perform
- Once the action is chosen
  - The agent records its choice with Tell and executes the action

#### Knowledge-Based Agents (cont.)

- Agents can be viewed at knowledge level
  - What they know, what the goals are, ...
- Or agents can be viewed at the implementation level
  - The data structures in KB and algorithms that manipulate them
- In summary, the agents must be able to
  - Represent states, actions, etc.
  - Incorporate new percepts
  - Update internal representations of the world
  - Deduce hidden properties of the world
  - Deduce appropriate actions

## Wumpus World

- Wumpus world was an early computer game, based on an agent who explores a cave consisting of rooms connected by passageways
- Lurking somewhere in the cave is the wumpus, a beast that eats anyone who enters a room
- Some rooms contain bottomless pits that will trap anyone who wanders into these rooms (except the wumpus, who is too big to fall in)
- The only mitigating features of living in the environment is the probability of finding a heap of gold

#### Wumpus World PEAS Description

- Performance measure
  - gold +1000, death -1000,
    - -1 per step, -10 for using the arrow
- Environment
  - Squares adjacent to wumpus are smelly
  - Squares adjacent to pits are breezy
  - Glitter if gold is in the same square
  - Shooting kills wumpus if you are facing it
  - Shooting uses up the only one arrow
  - Grabbing picks up gold if in same square
  - Releasing drops the gold in same square
- Actuators
  - Forward, Turn Right, Turn Left, Grab, Release, Shoot
- Sensors
  - Breeze, Glitter, Smell, ...



#### Wumpus World Characterization

- Observable?? No --- only local perception
- Deterministic?? Yes --- outcomes exactly specified
- Episodic?? No --- sequential at the level of actions
- Static?? Yes --- Wumpus and pits can not move
- Discrete?? Yes
- Single-agent?? Yes --- Wumpus is essentially a nature feature

## Exploring a Wumpus World



• Initial percept [None, None, None, None, None]

stench breeze glitter bump

scream



After the first move, with percept
 [None, Breeze, None, None, None]





• After the third move, with percept [Stench, None, None, None, None]





• After the fourth move, with percept [None, None, None, None, None]

W	ОК		
S OK	АОК	ОК	
OK A	B OK	Ρ	



• After the fifth move, with percept [Stench, Breeze, Glitter, None, None]

## **Other Tight Spots**



Breeze in (1,2) and (2,1)  $\Rightarrow$  No safe actions

Smell in (1,1) ⇒ Cannot move Can use a strategy of coercion shot straight ahead wumpus there →dead →safe wumpus wasn't there → safe



## Logic in General

- Logics are formal languages for representing information such that conclusions can be drawn
- Syntax defines the sentences in the language
- Semantics define the "meaning" of sentences; i.e., define truth or falsehood of a sentence in a world
- E.g., the language of arithmetic x+2>y is a sentence; x2+y> is not a sentence

 $x+2\ge y$  is true iff the number x+2 is no less than the number y

 $x+2 \ge y$  is true in a world where x=7, y=1

 $x+2 \ge y$  is false in a world where x=0, y=6

The term "model" will be used to replace the term "world"

 Sentences in an agent's KB are real physical configurations of it

## Entailment

• Entailment means that one thing follows from another:

**KB** |= α

- Knowledge base *KB* entails sentence  $\alpha$  if  $\alpha$  is true in all worlds where *KB* is true
  - E.g., the KB containing "the Giants won" and "the Reds won" entails "either the Giants or the Reds won"
  - E.g., *x*+*y*=4 entails 4=*x*+*y*
- The knowledge base can be considered as a statement
- Entailment is a relationship between sentences (i.e., syntax) that is based on semantics
  - E.g., α |= β
    - $\alpha$  entails  $\beta$
    - $\alpha \models \beta$  iff in every model in which  $\alpha$  is true,  $\beta$  is also true
    - Or, if  $\alpha$  is true,  $\beta$  must be true

#### Models

 Logicians typically think in terms of models, which are formally structured worlds with respect to which truth can be evaluated

*m* is a model of a sentence  $\alpha$  iff  $\alpha$  is true in *m* 

- IF M(α) is the set of all models of α
  Then KB |= α if and only if M(KB) ⊆ M(α)
  - I.e., every model in which *KB* is true,  $\alpha$  is also true
    - On the other hand, not every model in which  $\alpha$  is true, *KB* is also true



#### Entailment in the Wumpus World

- Situation after detecting nothing in [1,1], moving right, breeze in [2,1]
- Consider possible models for ?s
  assuming only pits

?	?		
[1,1]	в [2,1]	2	

• 3 Boolean choices  $\Rightarrow$  8 possible models

## Wumpus Models

• 8 possible models

















• *KB* = wumpus world-rules + observations



- *KB* = wumpus world-rules + observations
  - $\alpha_1$ = "[1,2] is safe" (no pit in [1, 2])

KB

- *KB* |=  $\alpha_1$ , proved by model checking

enumerate all possible models to check that  $\alpha_1$  is true in all models in which *KB* is true



• *KB* = wumpus world-rules + observations



- *KB* = wumpus world-rules + observations
  - $\alpha_2$ = "[2,2] is safe" (no pit in [2, 2])
  - *KB* |  $\neq \alpha_2$ , proved by model checking



## Inference

- **KB** |-<sub>i</sub> α
  - Sentence  $\alpha$  can be derived from *KB* by inference algorithm *i*
  - Think of

the set of all consequences of KB as a haystack

 $\alpha$  as a needle

entailment like the needle in the haystack

inference like finding it

- Soundness or truth-preserving inference
  - An algorithm *i* is sound if whenever *KB*  $|-_i \alpha$ , it is also true that *KB*  $|= \alpha$
  - That is the algorithm derives only entailed sentences
  - The algorithm won't announce " the discovery of nonexistent needles"

## Inference (cont.)

- Completeness
  - An algorithm *i* is complete if whenever *KB* |=  $\alpha$ , it is also true that *KB* |–*<sub>i</sub>*  $\alpha$
  - A sentence  $\alpha$  will be generated by an inference algorithm *i* if it is entailed by the *KB*
  - Or says, the algorithm will answer any question whose answer follows from what is known by the KB

## Inference (cont.)



- Sentences are physical configurations of the agent, and reasoning is a process of constructing new physical configurations from old ones
- Logical reasoning should ensure that the new configurations represent aspects of the world that actually follow from the aspects that the old configurations represent

#### Propositional Logic: Syntax

- Propositional logic is the simplest logic that illustrates basic ideas
- Syntax: defines the allowable sentences
  - Atomic sentences consist of a single propositional symbols
  - Propositional symbols: e.g., *P*, *Q* and *R* 
    - Each stands for a proposition (fact) that can be either true or false
  - Complex sentences are constructed from simpler one using logic connectives
    - $\wedge$  (and) conjunction
    - $\vee$  (or) disjunction
    - $\Rightarrow$  (implies) implication
    - $\Leftrightarrow$  (equivalent) equivalence, or biconditional
    - $\neg$  (not) negation

#### Propositional Logic: Syntax (cont.)

#### • BNF (Backus-Naur Form) grammar for propositional logic

Sentence  $\rightarrow$  Atomic Sentence | Complex Sentence Atomic Sentence  $\rightarrow$  True | False |Symbol Symbol  $\rightarrow$  P | Q | R ... Complex Sentence  $\rightarrow \neg$  Sentence | (Sentence  $\land$  Sentence) | (Sentence  $\lor$  Sentence) | (Sentence  $\Rightarrow$  Sentence) | (Sentence  $\Leftrightarrow$  Sentence)

Order of precedence: (from highest to lowest)
 ¬, ∧, ∨, ⇒, and ⇔

- E.g.,  $\neg \mathsf{P} \lor Q \land R \Rightarrow S$  means  $((\neg \mathsf{P}) \lor (Q \land R)) \Rightarrow S$  $A \Rightarrow B \Rightarrow C$  is not allowed !

#### **Propositional Logic: Semantics**

- Define the rules for determining the truth of a sentence with respect to a particular model
  - Each model fixes the truth value (true or false) for every propositional symbol

- · 3 symbols, 8 possible models, can be enumerated automatically
- A possible model  $m_1 \{P_{1,2} = false, P_{2,2} = false, P_{3,1} = true\}$

#### - Simple recursive process evaluates an arbitrary sentence, e.g.,

 $\neg P_{1,2} \land (P_{2,2} \lor P_{3,1}) = true \land (false \lor true) = true \land true = true$ 

Models for PL are just sets of truth values for the propositional symbols

#### **Truth Tables for Connectives**

P	Q	$\neg P$	$P \wedge Q$	$P \lor Q$	$P \Rightarrow Q$	$P \Leftrightarrow Q$
false	false	true	false	false	true	true
false	true	true	false	true	true	false
true	false	false	false	true	false	false
true	true	false	true	true	true	true

$\neg P$	is true iff	Р	is false			
$m{P}igwedge m{Q}$	is true iff	Ρ	is true	and	Q	is true
$P \lor Q$	is true iff	Р	is true	or	Q	is true
$P \!\! \Rightarrow \!\! Q$	is false iff	Р	is true	and	Q	is false
$P \Leftrightarrow Q$	is true iff	$P \Rightarrow 0$	Q is true	and	$Q {\Rightarrow} F$	'is true



## More about Implication

- For an implication:  $P \Rightarrow Q$ 
  - Which doesn't need any relation of causation or relevance between *P* and *Q*
    - "5 is odd implies Tokyo is the capital of Japan" is true
- We can think of " $P \Rightarrow Q$ " as saying
  - If *P* is true, then I am claiming that *Q* is true. Otherwise I am making no claim

#### Knowledge Base

- Knowledge base, consisting of a set of sentences, can be considered as a single sentence
  - A conjunction of these sentences
  - Knowledge base asserts that all the individual sentences are true
### Wumpus World Sentences

- 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]
- A square is breezy *if only if* there is an adjacent pit

 $\begin{array}{ll} R_{1} \colon \neg P_{1,1} & \text{no pit in [1,1]} \\ R_{2} \colon B_{1,1} \Leftrightarrow (P_{1,2} \lor P_{2,1}) & \text{pits cause breezes in adjacent squares} \\ R_{3} \colon B_{2,1} \Leftrightarrow (P_{1,1} \lor P_{2,2} \lor P_{3,1}) & \\ R_{4} \colon \neg B_{1,1} & \text{no breeze in [1,1]} \\ R_{5} \colon B_{2,1} & \text{breeze in [2,1]} \end{array}$ 

- Note: there are 7 proposition symbols involved
  - $-B_{1,1}, B_{2,1}, P_{1,1}, P_{1,2}, P_{2,1}, P_{2,2}, P_{3,1}$
  - There are  $2^7 = 128$  models !
    - While only three of them satisfy the above 5 descriptions/sentences

### **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	$\alpha_1$	
false	true								
false	false	false	false	false	false	true	false	true	
:	:	:	:	:	:	:	:	:	
false	true	false	false	false	false	false	false	true	128
false	true	false	false	false	false	true	$\underline{true}$	true	models
false	true	false	false	false	true	false	$\underline{true}$	$\underline{true}$	
false	true	false	false	false	true	true	$\underline{true}$	$\underline{true}$	
false	true	false	false	true	false	false	false	true	
:	÷	:	:	:	:	:	:	:	
true	false	false							
							<b>_</b>	<b>†</b>	-

 $R_1 \wedge R_2 \wedge R_3 \wedge R_4 \wedge R_5$ 

• P<sub>2,2</sub> ?

Conjunction of sentences of KB

- P<sub>1,2</sub>

# Inference by Enumeration (Model Checking)

#### Test if KB is true $\alpha$ is also true



- A recursive depth-first enumeration of all models (assignments to variables)
  - Sound and complete
    - Time complexity:  $O(2^n)$  exponential in the size of the input
  - Space complexity: O(n)

# Logical Equivalences

• Two sentences are logically equivalent iff true in same set of models

 $\alpha \equiv \beta \quad \text{if and only if} \quad \alpha \models \beta \text{ and } \beta \models \alpha$ 

 $M(\alpha) \subseteq M(\beta) \text{ and}$  $M(\beta) \subseteq M(\alpha)$  $\therefore M(\beta) = M(\alpha)$ 

AI -Berlin Chen 40

# Logical Equivalences (cont.)

$$\begin{array}{l} (\alpha \wedge \beta) \equiv (\beta \wedge \alpha) \quad \text{commutativity of } \wedge \\ (\alpha \vee \beta) \equiv (\beta \vee \alpha) \quad \text{commutativity of } \vee \\ ((\alpha \wedge \beta) \wedge \gamma) \equiv (\alpha \wedge (\beta \wedge \gamma)) \quad \text{associativity of } \wedge \\ ((\alpha \vee \beta) \vee \gamma) \equiv (\alpha \vee (\beta \vee \gamma)) \quad \text{associativity of } \vee \\ \neg (\neg \alpha) \equiv \alpha \quad \text{double-negation elimination} \\ (\alpha \Rightarrow \beta) \equiv (\neg \beta \Rightarrow \neg \alpha) \quad \text{contraposition} \\ (\alpha \Rightarrow \beta) \equiv (\neg \alpha \vee \beta) \quad \text{implication elimination} \\ (\alpha \Leftrightarrow \beta) \equiv ((\alpha \Rightarrow \beta) \wedge (\beta \Rightarrow \alpha)) \quad \text{biconditional elimination} \\ \neg (\alpha \wedge \beta) \equiv (\neg \alpha \vee \neg \beta) \quad \text{De Morgan} \\ \neg (\alpha \vee \beta) \equiv ((\alpha \wedge \beta) \vee (\alpha \wedge \gamma)) \quad \text{distributivity of } \wedge \text{ over } \vee \\ (\alpha \vee (\beta \wedge \gamma)) \equiv ((\alpha \vee \beta) \wedge (\alpha \vee \gamma)) \quad \text{distributivity of } \vee \text{ over } \wedge \end{array}$$

### Validity and Satisfiability

A sentence is valid (or tautological) if it is true in all models

*True*,  $A \lor \neg A$ ,  $A \Rightarrow A$ ,  $(A \land (A \Rightarrow B)) \Rightarrow B$ 

- Validity is connected to inference via Deduction Theorem:  $KB \models \alpha$  if only if  $(KB \Rightarrow \alpha)$  is valid
- A sentence is satisfiable if it is true in some model
   A, B ∧ ¬C
- A sentence is unsatifiable if it is true in no models  $A \wedge \neg A$
- Satisfiablity is connected to inference via refutation (or proof by contradiction)

KB |=  $\alpha$  if only if (KB  $\wedge \neg \alpha$ ) is unsatifiable Determination of satisfiability of sentences in PL is NP-complete

Al –Berlin Chen 42

### Patterns of Inference: Inference Rules

- Applied to derive chains of conclusions that lead to the desired goal
- Modus Ponens (Implication Elimination, *if-then* reasoning)

$$\frac{\alpha \Rightarrow \beta, \qquad \alpha}{\beta}$$

And Elimination

$$\frac{\alpha \wedge \beta}{\alpha}$$

Biconditional Elimination

$$\frac{\alpha \Leftrightarrow \beta}{(\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)} \quad \text{and} \quad$$

$$\frac{(\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)}{\alpha \Leftrightarrow \beta}$$

# Patterns of Inference: Inference Rules (cont.)

- Example
  - With the KB as the following, show that  $\neg P_{1,2}$

 $R_1: \neg P_{1,1}$ no pit in [1,1] $R_2: B_{1,1} \Leftrightarrow (P_{1,2} \lor P_{2,1})$ pits cause breezes in adjacent squares $R_3: B_{2,1} \Leftrightarrow (P_{1,1} \lor P_{2,2} \lor P_{3,1})$ no breeze in [1,1] $R_4: \neg B_{1,1}$ no breeze in [1,1] $R_5: B_{2,1}$ breeze in [2,1]

- 1. Apply biconditional elimination to  $R_2$  $R_6$ :  $(B_{1,1} \Rightarrow (P_{1,2} \lor P_{2,1})) \land ((P_{1,2} \lor P_{2,1}) \Rightarrow B_{1,1})$
- 2. Apply And-Elimination to  $R_6$

$$R_7$$
: ( $P_{1,2} \lor P_{2,1}$ )  $\Rightarrow$   $B_{1,1}$ 

3. Logical equivalence for contrapositives

 $R_8$ :  $\neg B_{1,1}$   $\Rightarrow$   $\neg$ ( $P_{1,2} \lor P_{2,1}$ )

4. Apply Modus Ponens with  $R_8$  and the percept  $R_4$ 

5. Apply De Morgan's rule and give the conclusion

$$R_{10}$$
:  $\neg P_{1,2} \land \neg P_{2,2}$ 

6. Apply And-Elimination to  $R_{10}$  $R_{11}: \neg P_{1,2}$ 

## Patterns of Inference: Inference Rules (cont.)

• Unit Resolution

Resolution

$$\frac{\alpha \lor \beta, \quad \neg \beta}{\alpha}$$

$$\frac{l_1 \vee l_2 \vee \cdots \vee l_k, \quad m}{l_1 \vee \cdots \vee l_{i-1} \vee l_{i+1} \vee \cdots \vee l_k}$$

Resolution is used to either confirm or refute a sentence, but it can't be used to enumerate sentences

*I*, and *m* are complementary literals

$$\frac{\alpha \lor \beta, \quad \neg \beta \lor \gamma}{\alpha \lor \gamma} \quad \frac{l_1 \lor l_2 \lor \cdots \lor l_k, \quad m_1 \lor \cdots \lor m_n}{l_1 \lor \cdots \lor l_{i-1} \lor l_{i+1} \lor \cdots \lor l_k \lor m_1 \cdots \lor m_{j-1} \lor m_{j+1} \lor \cdots \lor m_n}$$

$$\frac{P_{1,1} \lor P_{3,1}, \quad \neg P_{1,1} \lor \neg P_{2,2}}{P_{3,1} \lor \neg P_{2,2}} \qquad I_i \text{ and } m_j \text{ are complementary literals}$$

$$\mathsf{E.g.}, \qquad \mathsf{E.g.}, \qquad \mathsf$$

 Multiple copies of literals in the resultant clause should be removed (such a process is called factoring)

AI – Berlin Chen 45

### Patterns of Inference: Inference Rules (cont.)

• Unit Resolution

$$\alpha \lor \beta$$
 is true  $(\alpha \lor \beta \equiv \neg \beta \Rightarrow \alpha)$   
and  $\neg \beta$  is true  
 $\therefore \alpha$  is true

Resolution

$$\alpha \lor \beta \text{ is true } (\alpha \lor \beta \equiv \neg \beta \Rightarrow \alpha)$$
  
and 
$$\neg \beta \lor \gamma \text{ is true } (\neg \beta \lor \gamma \equiv \neg \gamma \Rightarrow \neg \beta)$$
  
$$\therefore \neg \gamma \Rightarrow \alpha (\neg \gamma \Rightarrow \alpha \equiv \alpha \lor \gamma)$$

# Monotonicity

• The set of entailed sentences can only increase as information is added to the knowledge base

```
If KB |= \alpha then KB \wedge \beta |= \alpha
```

- The additional assertion  $\beta$  can't invalidate any conclusion  $\alpha$  already inferred
- E.g.,  $\alpha$ : there is not pit in [1,2]
  - $\boldsymbol{\beta}$  : there is eight pits in the world



# **Normal Forms**

- Conjunctive Normal Form (CNF)
  - A sentence expressed as a conjunction of disjunctions of literals
  - E.g.,  $(P \lor Q) \land (\neg P \lor R) \land (\neg S)$
- Also, Disjunction Normal Form (*DNF*)
  - A sentence expressed as a disjunction of conjunctions of literals
  - E.g.,  $(P \land Q) \lor (\neg P \land R) \lor (\neg S)$
- An arbitrary propositional sentence can be expressed in *CNF* (or *DNF*)

### Normal Forms (cont.)

- Example: convert  $B_{1,1} \Leftrightarrow (P_{1,2} \lor P_{2,1})$  into *CNF* 
  - 1. Eliminate  $\Leftrightarrow$ , replace  $\alpha \Leftrightarrow \beta$  with  $(\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)$  $(B_{1,1} \Rightarrow (P_{1,2} \lor P_{2,1})) \land ((P_{1,2} \lor P_{2,1}) \Rightarrow B_{1,1})$
  - 2. Eliminate  $\Rightarrow$ , replace  $\alpha \Rightarrow \beta$  with  $(\neg \alpha \lor \beta)$  $(\neg B_{11} \lor P_{12} \lor P_{21}) \land (\neg (P_{12} \lor P_{21}) \lor B_{11})$
  - 3. Move  $\neg$  inwards

 $(\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}) \land ((\neg P_{1,2} \land \neg P_{2,1}) \lor B_{1,1})$ 

4. Apply distributivity law

 $(\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}) \land (\neg P_{1,2} \lor B_{1,1}) \land (\neg P_{2,1} \lor B_{1,1})$ 

## **Resolution Algorithm**

```
 \begin{aligned} & \textbf{function PL-RESOLUTION}(KB, \alpha) \textbf{ returns } true \text{ or } false \\ & \textbf{inputs: } KB, \textbf{ the knowledge base, a sentence in propositional logic} \\ & \alpha, \textbf{ the query, a sentence in propositional logic} \\ & clauses \leftarrow \textbf{ the set of clauses in the CNF representation of } KB \land \neg \alpha \\ & new \leftarrow \{\} \\ & \textbf{loop do} \\ & \textbf{ for each } C_i, C_j \textbf{ in } clauses \textbf{ do} \\ & resolvents \leftarrow \textbf{PL-RESOLVE}(C_i, C_j) \\ & \textbf{ if } resolvents \text{ contains the empty clause then return } true \\ & new \leftarrow new \cup resolvents \\ & \textbf{ if } new \subseteq clauses \textbf{ then return } false // no new clauses resolved \\ & clauses \leftarrow clauses \cup new \end{aligned} } \end{aligned}
```

- To show that *KB* |=  $\alpha$  , we show that (*KB*  $\land \neg \alpha$ ) is unsatisfiable
- Each pair that contains complementary literals is resolved to produce new clause until one of the two things happens:
   (1) No new clauses can be added ⇒ KB does not entail α
   (2) Empty clause is derived ⇒ KB entails α

### **Resolution Example**

 $KB = (B_{1,1} \iff (P_{1,2} \lor P_{2,1})) \land \neg B_{1,1} \ \alpha = \neg P_{1,2}$ 



- Empty clause disjunction of no disjuncts
  - Equivalent to false
  - Represent a contradiction here

 $(B_{1,1} \Leftrightarrow (P_{1,2} \lor P_{2,1})) \land \neg B_{1,1}$  $(\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}) \land (\neg P_{1,2} \lor B_{1,1}) \land (\neg P_{2,1} \lor B_{1,1}) \land \neg B_{1,1}$ 

We have shown it before !

# Horn Clauses

- A Horn clause is a disjunction of literals of which at most one is positive
  - $E.g., \neg P_1 \lor \neg P_2 \lor \ldots \lor \neg P_n \lor Q$
- Every Horn clause can be written as an implication
  - The premise is a conjunction of positive literals
  - The conclusion is a single positive literal
  - E.g.,  $\neg P_1 \lor \neg P_2 \lor \ldots \lor \neg P_n \lor Q$  can be converted to  $(P_1 \land P_2 \land \dots \land P_n) \Rightarrow Q$
- Inference with Horn clauses can be done naturally through the forward chaining and backward chaining, which be will be discussed later on  $\frac{\alpha \Rightarrow \beta, \qquad \alpha}{\beta}$ 
  - The application of Modus Ponens
- Not every PL sentence can be represented as a conjunction of Horn clauses

# Forward Chaining

- As known, if all the premises of an implication are known, then its conclusion can be added to the set of known facts
- Forward Chaining fires any rule whose premises are satisfied in the KB, add its conclusion to the KB, until query is found or until no further inferences can be made

Applications of Modus Ponens



#### Forward Chaining: Example

















# Forward Chaining: Algorithm (cont.)

```
function PL-FC-ENTAILS?(KB, q) returns true or false
  inputs: KB, the knowledge base, a set of propositional Horn clauses
           q, the query, a proposition symbol
  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 in KB
                                                                                facts
  while agenda is not empty do
      p \leftarrow POP(agenda)
      unless inferred[p] do //each logical symbol checked at most once (avoiding repeated firings)
          inferred[p] \leftarrow 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: Properties

- Sound
  - Because every inference is an application of Modus Ponens
- Complete
  - Every entailed atomic sentence (i.e., propositional symbol) will be derived
  - But may do lots of work that is irrelevant to the goal
- A form of data-driven reasoning
  - Start with known data and derive conclusions from incoming percepts

# **Backward Chaining**

- Work backwards from the query *q* to prove *q* by backward chaining (*BC*)
- Check if q is known already, or prove by BC all premises of some rule concluding q
- A form of goal-directed reasoning

### Backward Chaining: Example





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

 $P \Rightarrow Q$ 







 $P \Rightarrow Q$  $L \land M \Rightarrow P$  $B \land L \Rightarrow M$  $A \land P \Rightarrow L$  $A \land B \Rightarrow L$ AB



 $A \land B \Rightarrow L$ 

$$P \Rightarrow Q$$

$$L \land M \Rightarrow P$$

$$B \land L \Rightarrow M$$

$$A \land P \Rightarrow L$$

$$A \land B \Rightarrow L$$

$$A$$

$$B$$





 $B \land L \Rightarrow M$ 

$$P \Rightarrow Q$$

$$L \land M \Rightarrow P$$

$$B \land L \Rightarrow M$$

$$A \land P \Rightarrow L$$

$$A \land B \Rightarrow L$$

$$A$$

$$B$$
### Backward Chaining: Example (cont.)



 $L \land M \Rightarrow P$ 

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

# Backward Chaining: Example (cont.)



#### Backward Chaining: Example (cont.)



# Forward vs. Backward Chaining

- FC (data-driven)
  - May do lots of work that is irrelevant to the goal
- BC (goal-driven)
  - Complexity of BC can be much less than linear in size of KB

# Propositional Logic: Drawbacks

- Propositional Logic is declarative and compositional
- The lack of expressive power to describe an environment with many objects concisely
  - E.g., we have to write a separate rule about breezes and pits for each square

 $B_{1,1} \Leftrightarrow (P_{1,2} \lor P_{2,1})$