

# Knowledge Representation

## Part 1

AI 109

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All men naturally desire knowledge.

Aristotle, *Metaphysics*

# What is Knowledge?

- What does it mean to *know* something? Someone?
- If you memorize a fact, do you truly know it?

# Aristotle's Approach

- Aristotle distinguished between “true belief” and *demonstrative knowledge* (ἐπιστήμη, episteme).
  - Reciting a truth didn't count as knowledge.
  - Having demonstrative knowledge of a thing is knowing its *causes* – what makes it to be what it is.
  - So, knowledge for Aristotle involved *explanation*.
- The kind of “explanation” that Aristotle accepted is called a *demonstration*.

# Example

- Argument:

“No human observed has a tail; therefore humans lack tails.”

- Is this a good argument?

- Is it convincing?

- Alternatively:

- All bipedal animals with upright posture lack tails, because a tail would not serve the functional balance appropriate to upright locomotion.
- All humans are bipedal rational animals with upright posture.
- **Therefore:** All humans lack tails.

- Is this better?

# Demonstration

- A **demonstration** is a *logical argument* (Aristotle used the term *sylogism*) whose **premises** are true, primary, immediate, prior to, and better known than the **conclusion**.
- The **conclusion** of an argument is the thing you want to prove.
- The **premises** are reasons given for accepting the conclusion.
- The process of moving from premises to a conclusion is called **inference**.
- Aristotle's goal was to organize all knowledge according to logical arguments that explain things in terms of basic definitions and principles.
  - Look! Another example of a graph.

# Knowledge Representation in AI

- ***Knowledge Representation*** (KR) studies
  - how to formally encode information about the world,
  - so that a machine can “reason” with it.
- Central question: What must be represented so that intelligent behavior becomes computationally possible?

# Components of a KR System

- Representation
  - We need some kind of machine-like language for representing knowledge.
- Inference
  - We need some kind of system to enable the machine to “reason” from premises to conclusions.
  - Is this really necessary? Why do we need it?

# Data vs Information vs Knowledge

- ***Data***

- Raw symbols or measurements without interpretation (e.g., “42”, “Alice”, “3.7”).

- ***Information***

- Data interpreted within a structure or ***schema*** (e.g., “Alice’s GPA is 3.7”).

- ***Knowledge***

- Structured information organized so that new conclusions can be derived (e.g., “Students with GPA > 3.5 graduate with honors; therefore Alice graduates with honors”).
- Knowledge implies the capacity for performing inference.

# Computers Need Formal Language

- ***Natural language*** is expressive but ambiguous and context-dependent.
- A ***formal language*** removes ambiguity by specifying exact ***formation rules***.
- Meaning is defined mathematically, not by intuition.
- This lets us ask precise questions: What follows from what?

# Propositional Logic

- **Atomic propositions** represent basic facts.
  - Example: Socrates is a man.
  - Can be either **true** or **false**.
    - This rules out subjective statements (in this system).
  - We represent atomic propositions using **symbols**: P, Q, R, ...
  - We can also represent them using more informative names: is\_hungry.
- We build logical **formulas** by combining atomic propositions using **connectives** and **formation rules**.
  - Connectives:  $\neg$  (not),  $\wedge$  (and),  $\vee$  (or),  $\rightarrow$  (implies).
  - Formation rules: If P and Q are formulas, so is  $(P \wedge Q)$

# Examples

- Example
  - P: “Alice is a student.”
  - Q: “Alice has GPA > 3.5.”
  - $P \wedge Q$ : “Alice is a student and has GPA > 3.5.”
- Example
  - P: “It is daytime.”
  - Q: “Eve is driving her car.”
  - $P \rightarrow Q$ : “If it is daytime, Eve is driving her car.”

# Syntax

- We can follow rules to build up complicated formulas without knowing what they mean.
  - This is what allows computers to “do logic.”
- The rules we follow to build up formulas from parts are called the ***syntax*** of our formal language.
- From  $P \rightarrow Q$  and  $Q \rightarrow R$ , we can form more complicated expressions like  $((((P \rightarrow Q) \wedge (Q \rightarrow R)) \wedge P) \rightarrow R)$ .

# Truth Values and Semantics

- Each atomic proposition is assigned a **truth value**: True or False.
- The truth of compound formulas is determined **compositionally**.
  - If you know the truth values of each part, you know the truth value of the whole.
- $P \wedge Q$  is true **if and only if** both P and Q are true.
- Semantics = mapping from formulas to truth values. What a formula “means.”
- Some statements are true no matter how you interpret them. These are called **tautologies**.
  - $((((P \rightarrow Q) \wedge (Q \rightarrow R)) \wedge P) \rightarrow R)$  is a tautology. Why?

# Syntax vs Semantics

- Syntax
  - formal structure of expressions (symbols, formation rules).
  - Like grammar.
- Semantics
  - what those expressions mean.
- Syntax and semantics are different, but work together
  - **Inference is correct when syntactic derivations preserve semantic truth.**
- KR requires an explicit semantic foundation.

# Inference Rules

- ***Inference rules*** tell us how to reach conclusions from premises.
- Example: ***Modus Ponens***
  - Given:  $P \rightarrow Q$  and  $P$
  - Conclude:  $Q$

# Limitations of Propositional Logic

- Propositional logic treats statements as indivisible atoms.
- No internal structure: cannot express “All students have GPA  $> 3.5$ .”
- Cannot quantify over individuals or describe relations.
- Useful for reasoning about fixed, finite sets of facts.
  
- The above lead to more sophisticated logical systems.

# KR as Modeling

- Representation is an abstraction: it omits detail to focus on relevant structure.
- Ontology: specification of what kinds of entities and relations exist in the model.
- Ontological commitments determine what can and cannot be expressed.
- Modeling choices constrain possible inferences.
- Good KR balances fidelity to the domain with computational feasibility.

# A Computer System for Logic

- Called ***Prolog***.
- Consists of ***facts*** and ***rules***.
  - Rules are written “backwards”
    - if X is human, then X is mortal.
  - commas mean “and.”
- Upper case symbols denote ***variables***.
- We can ***query*** the system to make Prolog perform inference.

```
% --- Facts ---
human(socrates).
human(plato).

% --- Rule ---
mortal(X) :-
    human(X).
```

# More Complicated Example

- Paths in Graphs

- $\text{path}(X,Y) \text{ :- edge}(X,Y)$ .
- $\text{path}(X,Y) \text{ :- edge}(X,Z), \text{path}(Z,Y)$ .

$\text{edge}(a,b)$ .  $\text{edge}(a,c)$ .  $\text{edge}(a,d)$ .

$\text{edge}(b,e)$ .  $\text{edge}(b,f)$ .

$\text{edge}(c,f)$ .  $\text{edge}(c,g)$ .

$\text{edge}(d,g)$ .  $\text{edge}(d,h)$ .

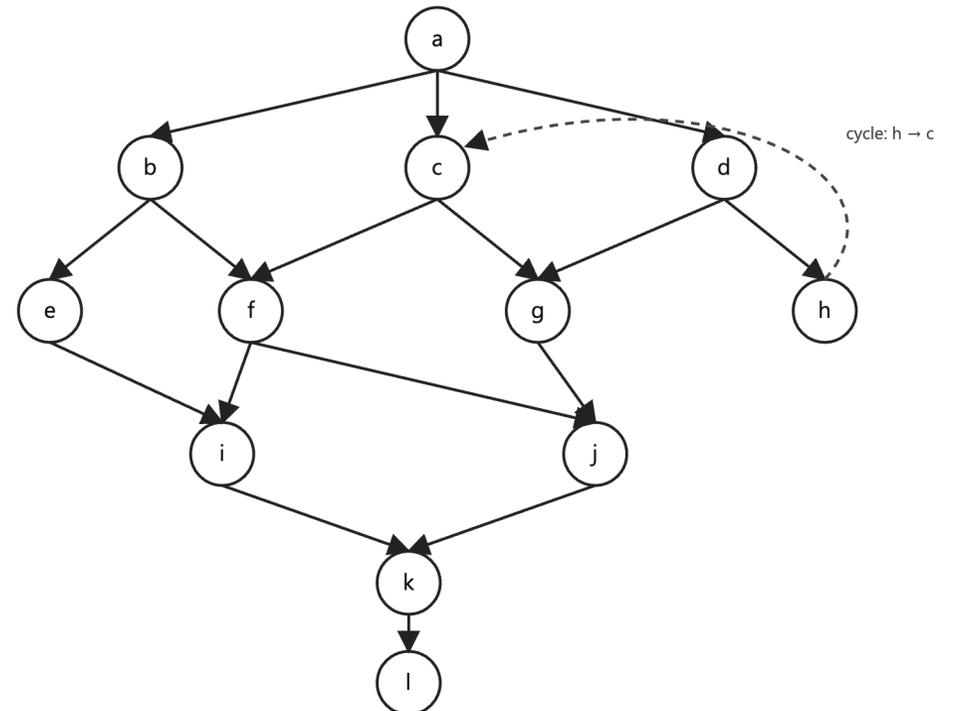
$\text{edge}(e,i)$ .

$\text{edge}(f,i)$ .  $\text{edge}(f,j)$ .

$\text{edge}(g,j)$ .  $\text{edge}(h,c)$ . % cycle

$\text{edge}(i,k)$ .  $\text{edge}(j,k)$ .

$\text{edge}(k,l)$ .



# Strengths? Weaknesses?

- Strengths
  - Fast
  - Systematic
- Weaknesses
  - Rules and facts are too rigid.
  - Doesn't handle synonymy, nuance.
  - Doesn't seem to match human reasoning.