CS 5/7320 Artificial Intelligence

Logical AgentsAIMA Chapter 7

Slides by Michael Hahsler based on slides by Svetlana Lazepnik with figures from the AIMA textbook



What is logic?



Logic is a formal system for manipulating facts so that true conclusions may be drawn



Syntax: rules for constructing valid sentences

E.g., $x + 2 \ge y$ is a valid arithmetic sentence, $\ge x2y + is$ not

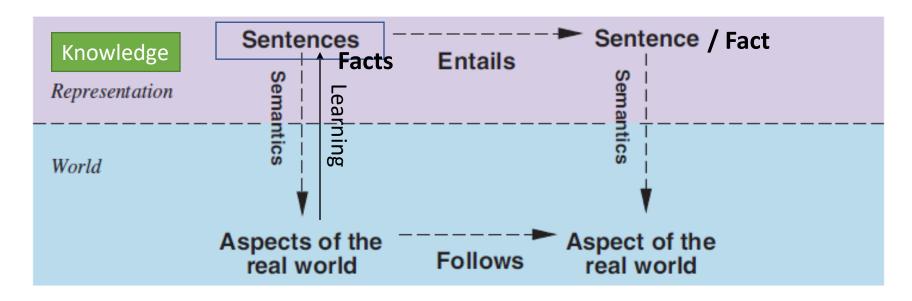


Semantics: "meaning" of sentences, or relationship between logical sentences and the real world

Specifically, semantics defines truth of sentences

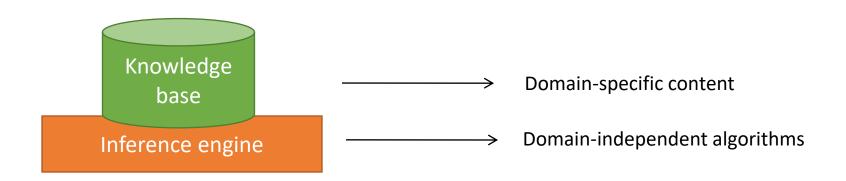
E.g., $x + 2 \ge y$ is true in a world where x = 5 and y = 7

Relation between Knowledge Representation and the World



- Facts: Sentences we know to be true.
- **Possible worlds**: all worlds/models which are consistent with the facts we know (compare with belief state).
- Learning new facts reduces the number of possible worlds.
- Entailment: A sentence logically follows from what we already know.

Knowledge-based agents



- Knowledge base (KB) = set of sentences in a formal language (knowledge representation) that are known to be true = set of facts
- Declarative approach to building an agent: Define what it needs to know
- Distinction between data (knowledge) and program (inference)
- Fullest realization of this philosophy was in the field of expert systems or knowledge-based systems in the 1970s and 1980s

Generic Knowledge-based Agent

function KB-AGENT(percept) returns an actionpersistent: KB, a knowledge base t, a counter, initially 0, indicating time

TELL(KB, MAKE-PERCEPT-SENTENCE(percept, t)) $action \leftarrow ASK(KB, MAKE-ACTION-QUERY(t))$ TELL(KB, MAKE-ACTION-SENTENCE(action, t)) $t \leftarrow t + 1$

return action

Memorize percept at time t

Ask for logical action

Record action taken at time t

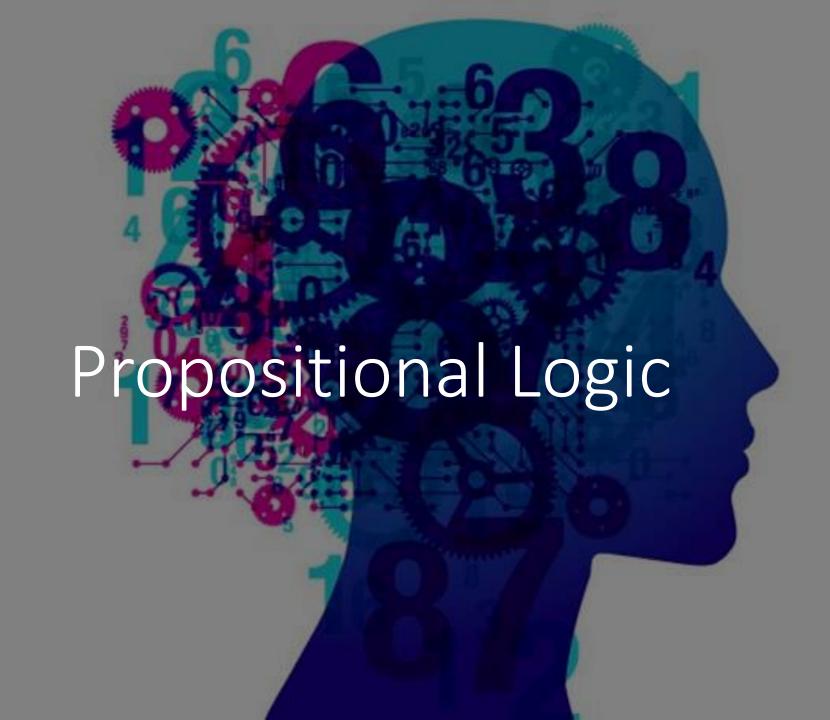
Overview





PROPOSITIONAL LOGIC

FIRST ORDER LOGIC



Propositional logic: Syntax in BN-Form

```
Sentence \rightarrow AtomicSentence \mid ComplexSentence
AtomicSentence \rightarrow True \mid False \mid P \mid Q \mid R \mid \dots = Symbols
ComplexSentence \rightarrow (Sentence) \mid \neg Sentence \mid Sentence \wedge Sentence \mid Conjunction \mid Sentence \vee Sentence \mid Disjunction \mid Sentence \Rightarrow Sentence \mid Implication \mid Sentence \Leftrightarrow Sentence \mid Biconditional
```

OPERATOR PRECEDENCE : $\neg, \land, \lor, \Rightarrow, \Leftrightarrow$

Validity and Satisfiability

A sentence is **valid** if it is true in **all** models (called a tautology)

e.g., *True*, $A \lor \neg A$, $A \Rightarrow A$, $(A \land (A \Rightarrow B)) \Rightarrow B$ useful to deduct new sentences.

A sentence is satisfiable if it is true in some model

e.g., AVB, C useful to find new facts that satisfy all possible worlds.

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

e.g., A∧¬A

Possible Worlds, Models and Truth Tables

A **model** specifies a "possible world" with the true/false status of each proposition symbol in the knowledge base

- E.g., **P** is true and **Q** is true
- With two symbols, there are $2^2 = 4$ possible worlds/models, and they can be enumerated exhaustively using:

A **truth table** specifies the truth value of a composite sentence for each possible assignments of truth values to its atoms. Each row is a model.

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

We have 3 possible worlds for $P \Rightarrow Q = true$

Propositional logic: Semantics

Rules for evaluating truth with respect to a model:

```
iff P
                                   is false
• \neg P is true
                     iff P
• P \wedge Q is true
                                  is true and Q
                                                            is true
                     iff P
• P \lor Q is true
                                  is true or Q
                                                            is true
                     iff
                                  is false or
• \mathbf{P} \Rightarrow \mathbf{Q} is true
                                                   Q
                                                           is true
                                           Model
    Sentence
```

Logical equivalence

Two sentences are logically equivalent iff (read if, and only if) they are true in same models

```
(\alpha \wedge \beta) \equiv (\beta \wedge \alpha) commutativity of \wedge
          (\alpha \vee \beta) \equiv (\beta \vee \alpha) commutativity of \vee
((\alpha \land \beta) \land \gamma) \equiv (\alpha \land (\beta \land \gamma)) associativity of \land
((\alpha \vee \beta) \vee \gamma) \equiv (\alpha \vee (\beta \vee \gamma)) associativity of \vee
            \neg(\neg \alpha) \equiv \alpha double-negation elimination
      (\alpha \Rightarrow \beta) \equiv (\neg \beta \Rightarrow \neg \alpha) contraposition
      (\alpha \Rightarrow \beta) \equiv (\neg \alpha \lor \beta) implication elimination
      (\alpha \Leftrightarrow \beta) \equiv ((\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)) biconditional elimination
       \neg(\alpha \land \beta) \equiv (\neg \alpha \lor \neg \beta) de Morgan
       \neg(\alpha \lor \beta) \equiv (\neg \alpha \land \neg \beta) de Morgan
(\alpha \land (\beta \lor \gamma)) \equiv ((\alpha \land \beta) \lor (\alpha \land \gamma)) distributivity of \land over \lor
(\alpha \vee (\beta \wedge \gamma)) \equiv ((\alpha \vee \beta) \wedge (\alpha \vee \gamma)) distributivity of \vee over \wedge
```

Entailment

 Entailment means that a sentence follows from the premises contained in the knowledge base:

$$KB \models \alpha$$

- The knowledge base KB entails sentence α iff α is true in all models where KB is true
 - E.g., KB with x = 0 entails sentence x * y = 0
- Tests for entailment
 - KB $\models \alpha$ iff (KB $\Rightarrow \alpha$) is valid
 - $KB = \alpha$ iff $(KB \land \neg \alpha)$ is unsatisfiable

Inference

 Logical inference: a procedure for generating sentences that follow from a knowledge base KB

• An inference procedure is **sound** if it derives a sentence α iff KB $\models \alpha$. I.e, it only derives **true sentences**.

• An inference procedure is **complete** if it can derive **all** α for which $KB \models \alpha$.

Inference

- How can we check whether a sentence α is entailed by KB?
- How about we **enumerate all possible models of the KB** (truth assignments of all its symbols), and check that α is true in every model in which KB is true?
 - Is this sound (if an answer is produced, it is correct)?
 - Is this complete (guaranteed to produce the correct answer)?
 - Problem: if KB contains n symbols, the truth table will be of size 2^n
- Better idea: use *inference rules*, or sound procedures to generate new sentences or *conclusions* given the *premises* in the KB

Complexity of inference

- Propositional inference is *co-NP-complete*
 - *Complement* of the SAT problem: $\alpha \models \beta$ if and only if the sentence $\alpha \land \neg \beta$ is unsatisfiable
 - Every known inference algorithm has worst-case exponential running time
- Efficient inference possible for restricted cases
 - e.g., Horn clauses are disjunctions of literals with at most one positive literal.

Example: Wumpus World

4 \$\frac{5\frac

1,4	2,4	3,4	4,4
1,3	2,3	3,3	4,3
1,2 OK	2,2	3,2	4,2
1,1 A OK	2,1 OK	3,1	4,1

_	
A	= Agent
В	= Breeze
\mathbf{G}	= Glitter, Gold
OK	= Safe square
P	= Pit
\mathbf{S}	= Stench

= Visited = Wumpus

1,4	2,4	3,4	4,4
1,3	2,3	3,3	4,3
1,2 OK	2,2 P?	3,2	4,2
1,1 V OK	2,1 A B OK	3,1 P?	4,1

(a) (b)

Example: Wumpus World

Initial KB needs to contain rules like these for each square:

$$Breeze(1,1) \Leftrightarrow Pit(1,2) \vee Pit(2,1)$$

 $Breeze(1,2) \Leftrightarrow Pit(1,1) \vee Pit(1,3) \vee Pit(2,2)$
 $Stench(1,1) \Leftrightarrow W(1,2) \vee W(2,1)$

. . .

Percepts at (1,1) are no breeze or stench. Add the following facts to the KB:

```
\neg Breeze(1,1)
\neg Stench(1,1)
```

Inference will tell us that the following facts are entailed:

$$\neg Pit(1,2), \neg Pit(2,1), \neg W(1,2), \neg W(2,1)$$

This means that (1,2) and (2,1) are safe.

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Percepts at (1,1) are no breeze or stench. Add the following facts to the KB:

 $\neg Breeze(1,1)$ $\neg Stench(1,1)$

Inference will tell us that the following facts are entailed:

 $\neg Pit(1,2), \neg Pit(2,1), \neg W(1,2), \neg W(2,1)$

This means that (1,2) and (2,1) are safe.

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 in 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
- Algorithms use forward, backward chaining, are linear in time, and complete for special clauses (definite clauses).



Limitations of propositional logic

- Suppose you want to say "All humans are mortal"
 - In propositional logic, you would need ~6.7 billion statements of the form:
 MichaellsHuman and MichaellsMortal,
 SarahlsHuman and SarahlsMortal, ...
- Suppose you want to say "Some people can run a marathon"
 - You would need a disjunction of ~6.7 billion statements:

MichaelcanRunAMarathon or ... or SarahCanRunAMarathon

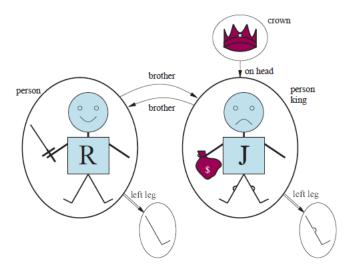
Other Languages to represent Knowledge

Language	Ontological Commitment (What exists in the world)	Epistemological Commitment (What an agent believes about facts)
Propositional logic First-order logic Temporal logic Probability theory Fuzzy logic	facts objects, relations facts, objects, relations, times facts facts with degree of truth $\in [0,1]$	true/false/unknown true/false/unknown true/false/unknown true/false/unknown degree of belief $\in [0,1]$ known interval value

• First-order Logic adds objects and relations.

Syntax of FOL

```
Sentence \rightarrow AtomicSentence \mid ComplexSentence
 AtomicSentence \rightarrow Predicate \mid Predicate(Term, ...) \mid Term = Term
ComplexSentence \rightarrow (Sentence)
                           \neg Sentence
                          Sentence \wedge Sentence
                          Sentence \lor Sentence
                          Sentence \Rightarrow Sentence
                          Sentence \Leftrightarrow Sentence
                           Quantifier Variable, . . . Sentence
             Term \rightarrow Function(Term,...)
```



Constant

Variable

 $Quantifier \rightarrow \forall \mid \exists$

 $Constant \rightarrow A \mid X_1 \mid John \mid \cdots$

 $Variable \rightarrow a \mid x \mid s \mid \cdots$

 $Predicate \rightarrow True \mid False \mid After \mid Loves \mid Raining \mid \cdots$

 $Function \rightarrow Mother \mid LeftLeg \mid \cdots \blacktriangleleft$

OPERATOR PRECEDENCE : $\neg, =, \land, \lor, \Rightarrow, \Leftrightarrow$

Objects

Relations. Predicate is/returns True or False

Function returns an object

Universal quantification

- ∀x P(x)
- Example: "Everyone at SMU is smart"
 ∀x At(x,SMU) ⇒ Smart(x)
 Why not ∀x At(x,SMU) ∧ Smart(x)?
- Roughly speaking, equivalent to the conjunction of all possible instantiations of the variable:

```
[At(John, SMU) \Rightarrow Smart(John)] \wedge ...
[At(Richard, SMU) \Rightarrow Smart(Richard)] \wedge ...
```

• $\forall x P(x)$ is true in a model m iff P(x) is true with x being each possible object in the model

Existential quantification

- ∃x P(x)
- Example: "Someone at SMU is smart"
 ∃x At(x,SMU) ∧ Smart(x)
 Why not ∃x At(x,SMU) ⇒ Smart(x)?
- Roughly speaking, equivalent to the disjunction of all possible instantiations:

```
[At(John,SMU) ∧ Smart(John)] ∨ [At(Richard,SMU) ∧ Smart(Richard)] ∨ ...
```

• $\exists x P(x)$ is true in a model m iff P(x) is true with x being some possible object in the model

Inference in FOL

- Inference is complicated.
- 1. Reduction to propositional logic and then use propositional logic inference.
- 2. Directly do inference on FOL (or a subset like definite clauses)
 - Unification: Combine two sentences into one.
 - Forward Chaining for FOL
 - Backward Chaining for FOL
 - Logical programming (e.g., Prolog)

Limitations of Logic

- What if we cannot collect enough knowledge to make a decision e.g., the environment if only partially observable
- What about facts about the future? grade(Michael, AI, this_semester)

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