

# CS 1571 Introduction to AI

## Lecture 9

### Propositional logic.

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### Administration

- **PS-3 due today** Report
  - Programs through ftp
- **PS-4 is out**
  - on the course web page
  - due next week on Tuesday, September 31, 2003

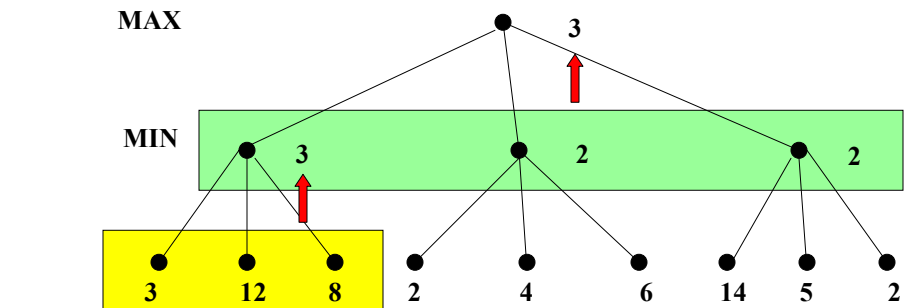
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## Minimax algorithm

How to deal with the contingency problem?

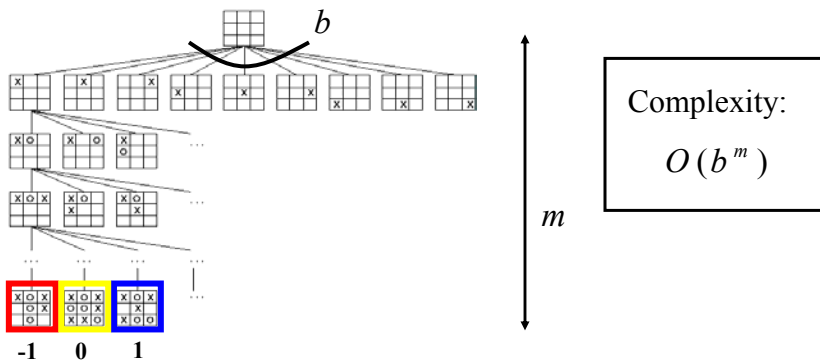
- Assuming that the opponent is rational and always optimizes its behavior (opposite to us) we consider **the best opponent's response**
- Then the **minimax algorithm** determines the best move



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## Complexity of the minimax algorithm

- We need to explore the complete game tree before making the decision



- Impossible for large games
  - Chess: 35 operators, game can have 50 or more moves

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## Solution to the complexity problem

Two solutions:

**1. Dynamic pruning of redundant branches** of the search tree

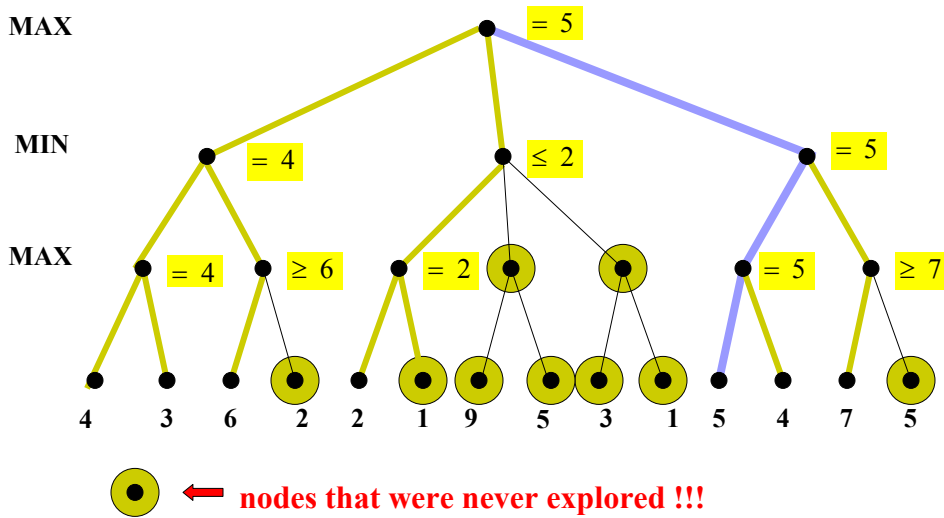
- identify provably suboptimal branch of the search tree even before it is fully explored
- Eliminate the suboptimal branch

**Procedure: Alpha-Beta pruning**

**2. Early cutoff of the search tree**

- uses imperfect minimax value estimate of non-terminal states (positions)

## Alpha beta pruning. Example



## Alpha-Beta pruning

**function** MAX-VALUE(*state*, *game*,  $\alpha$ ,  $\beta$ ) **returns** the minimax value of *state*

**inputs:** *state*, current state in game

*game*, game description

$\alpha$ , the best score for MAX along the path to *state*

$\beta$ , the best score for MIN along the path to *state*

**if** GOAL-TEST(*state*) **then return** EVAL(*state*)

**for each** *s* **in** SUCCESSORS(*state*) **do**

$\alpha \leftarrow \text{MAX}(\alpha, \text{MIN-VALUE}(s, \text{game}, \alpha, \beta))$

**if**  $\alpha \geq \beta$  **then return**  $\beta$

**end**

**return**  $\alpha$

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**function** MIN-VALUE(*state*, *game*,  $\alpha$ ,  $\beta$ ) **returns** the minimax value of *state*

**if** GOAL-TEST(*state*) **then return** EVAL(*state*)

**for each** *s* **in** SUCCESSORS(*state*) **do**

$\beta \leftarrow \text{MIN}(\beta, \text{MAX-VALUE}(s, \text{game}, \alpha, \beta))$

**if**  $\beta \leq \alpha$  **then return**  $\alpha$

**end**

**return**  $\beta$

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## Using minimax value estimates

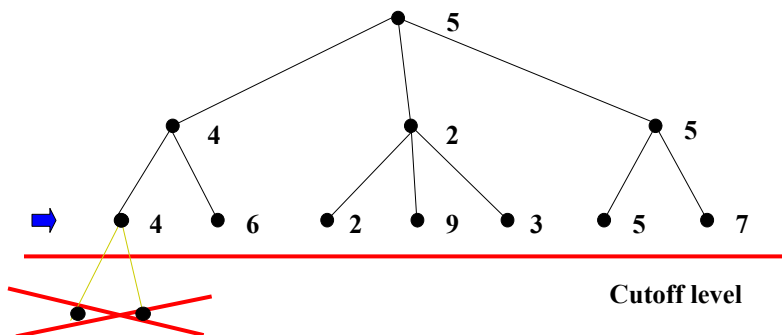
- Idea:**

- Cutoff the search tree before the terminal state is reached
- Use imperfect estimate of the minimax value at the leaves
  - Evaluation function

MAX

MIN

Heuristic  
evaluation  
function



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## Design of evaluation functions

- **Heuristic estimate** of the value for a sub-tree
- **Examples of a heuristic functions:**
  - **Material advantage in chess, checkers**
    - Gives a value to every piece on the board, its position and combines them
  - More general **feature-based evaluation function**
    - Typically a linear evaluation function:

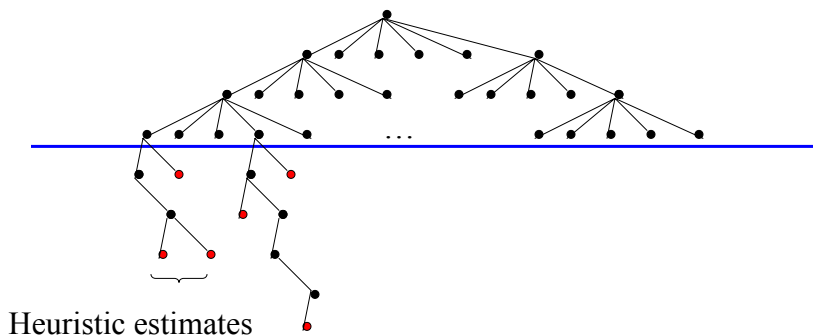
$$f(s) = f_1(s)w_1 + f_2(s)w_2 + \dots f_k(s)w_k$$

$f_i(s)$  - a feature of a state  $s$

$w_i$  - feature weight

## Further extensions to real games

- Restricted set of moves to be considered under **the cutoff level** to reduce branching and improve the evaluation function
  - E.g., consider only the capture moves in chess

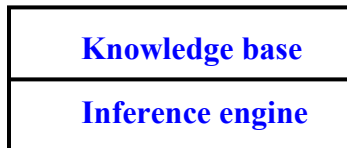


## Propositional logic

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## Knowledge-based agent



- **Knowledge base (KB):**
  - A set of sentences that describe facts about the world in some formal (representational) language
  - **Domain specific**
- **Inference engine:**
  - A set of procedures that work upon the representational language and can infer new facts or answer KB queries
  - **Domain independent**

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## Example: MYCIN

- MYCIN: an expert system for diagnosis of bacterial infections
- **Knowledge base** represents
  - Facts about a specific patient case
  - Rules describing relations between entities in the bacterial infection domain

<b>If</b>	1. The stain of the organism is gram-positive, and 2. The morphology of the organism is coccus, and 3. The growth conformation of the organism is chains
<b>Then</b>	the identity of the organism is streptococcus

- **Inference engine:**
  - manipulates the facts and known relations to answer diagnostic queries (consistent with findings and rules)

## Knowledge representation

- The **objective of knowledge representation** is to express the knowledge about the world in a computer-tractable form
- Key aspects of knowledge representation languages:
  - **Syntax:** describes how sentences are formed in the language
  - **Semantics:** describes the meaning of sentences, what is it the sentence refers to in the real world
  - **Computational aspect:** describes how sentences and objects are manipulated in concordance with semantic conventions

**Many KB systems rely on some variant of logic**

## Logic

A formal language for expressing knowledge and ways of reasoning.

**Logic** is defined by:

- **A set of sentences**
  - A sentence is constructed from a set of primitives according to syntax rules.
- **A set of interpretations**
  - An interpretation gives a semantic to primitives. It associates primitives with values.
- **The valuation (meaning) function  $V$** 
  - Assigns a value (typically the truth value) to a given sentence under some interpretation
$$V : \text{sentence} \times \text{interpretation} \rightarrow \{True, False\}$$

## Types of logic

- Different types of logic exist:
  - **Propositional logic**
  - First-order logic
  - Temporal logic
  - Numerical constraints logic
  - Map-coloring logic

In the following:

- **Propositional logic.**
  - Formal language for making symbolic inferences
  - Foundations of **propositional logic**: **George Boole** (1854)



## Propositional logic. Syntax

### Syntax:

- **Symbols (alphabet)** in  $\mathcal{P}$ :
  - **Constants**: True, False
  - **A set of propositional variables** (propositional symbols):  
Examples:  $P, Q, R, \dots$  or statements like:  
*Light in the room is on,*  
*It rains outside,* etc.
  - **A set of logical connectives**:  
 $\neg, \wedge, \vee, \Rightarrow, \Leftrightarrow$
- **Sentences**
  - **Build from symbols**

## Propositional logic. Syntax

### Sentences in the propositional logic:

- **Atomic sentences**:
  - **Constructed from constants and propositional symbols**
  - True, False are (atomic) sentences
  - $P, Q$  or *Light in the room is on,* *It rains outside* are (atomic) sentences
- **Composite sentences**:
  - **Constructed from valid sentences via logical connectives**
  - If  $A, B$  are sentences then  
 $\neg A$   $(A \wedge B)$   $(A \vee B)$   $(A \Rightarrow B)$   $(A \Leftrightarrow B)$   
or  $(A \vee B) \wedge (A \vee \neg B)$   
are sentences

## Propositional logic. Semantics.

The semantic gives the meaning to sentences.

In the propositional logic the semantics is defined by:

### 1. Interpretation of propositional symbols and constants

- Semantics of atomic sentences

### 2. Through the meaning of connectives

- Meaning (semantics) of composite sentences

## Semantic: interpretations

A **propositional symbol** (an atomic sentence) can stand for an arbitrary fact (statement) about the world

Examples: “*Light in the room is on*”,  
“*It rains outside*”, etc.

An **interpretation**:

- maps symbols to one of the two values: **True (T)**, or **False (F)**,
- the value depends on the world we want to describe

**World 1:**

**I:** *Light in the room is on* -> **True**, *It rains outside* -> **False**

**World 2:**

**I’:** *Light in the room is on* -> **False**, *It rains outside* -> **False**

## Semantics: symbols and constants

- The **meaning (truth)** of the propositional symbol for a **specific interpretation** is given by its interpretation

$$V(\textit{Light in the room is on}, \mathbf{I}) = \textit{True}$$

$$V(\textit{Light in the room is on}, \mathbf{I'}) = \textit{False}$$

- The meaning (truth) of constants:**
  - True** and **False** constants are always (under any interpretation) assigned the corresponding **True, False** value

$$V(\textit{True}, \mathbf{I}) = \textit{True}$$

$$V(\textit{False}, \mathbf{I}) = \textit{False}$$

For any interpretation **I**

## Semantics: composite sentences.

- The meaning (truth value) of complex propositional sentences.**
  - Determined using the “standard” rules for combining logical sentences:

<i>P</i>	<i>Q</i>	$\neg P$	$P \wedge Q$	$P \vee Q$	$P \Rightarrow Q$	$P \Leftrightarrow Q$
<i>True</i>	<i>True</i>	<i>False</i>	<i>True</i>	<i>True</i>	<i>True</i>	<i>True</i>
<i>True</i>	<i>False</i>	<i>False</i>	<i>False</i>	<i>True</i>	<i>False</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>False</i>	<i>False</i>	<i>True</i>	<i>False</i>	<i>False</i>	<i>True</i>	<i>True</i>

## Some definitions

- A **model (in logic)**: An interpretation is a **model for a set of sentences** if it assigns true to each sentence in the set.
- A sentence is **satisfiable** if it has a model;
  - There is at least one interpretation under which the sentence can evaluate to **True**.
- A sentence is **valid** if it is **True** in all interpretations
  - i.e., if its negation is **not satisfiable** (leads to contradiction)

		Satisfiable sentence		Valid sentence
$P$	$Q$	$P \vee Q$	$(P \vee Q) \wedge \neg Q$	$((P \vee Q) \wedge \neg Q) \Rightarrow P$
<i>True</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>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>False</i>	<i>True</i>

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## Some definitions

- A **model (in logic)**:
  - An interpretation is a **model for a set of sentences S** if it assigns true to each sentence in the set.
- **Example:**
  - Assume two sentences:  
 $(P \vee Q) \wedge \neg Q$  and  $P$
  - Assume an interpretation:
 
$$\begin{aligned} \mathbf{I}: \quad P &\longrightarrow \text{False} \\ Q &\longrightarrow \text{False} \end{aligned}$$
  - Is I a model for  $(P \vee Q) \wedge \neg Q$  and  $P$  ?

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  - **Assume two sentences:**  
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  - **Assume an interpretation:**  
 $I: P \longrightarrow \text{False}$   
 $Q \longrightarrow \text{False}$
  - **Is I a model for  $(P \vee Q) \wedge \neg Q$  and  $P$  ?**  
**No !**

## Some definitions

- A **model (in logic)**:
  - An interpretation is **a model for a set of sentences S** if it assigns true to each sentence in the set.
- **Example**:
  - **Assume two sentences:**  
 $(P \vee Q) \wedge \neg Q$  and  $P$
  - **Assume an interpretation:**  
 $I': P \longrightarrow \text{True}$   
 $Q \longrightarrow \text{False}$
  - **I' is a model for  $(P \vee Q) \wedge \neg Q$  and  $P$  !!!**

## Some definitions

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**Example:**

$P$	$Q$	$P \vee Q$	$(P \vee Q) \wedge \neg Q$	$((P \vee Q) \wedge \neg Q) \Rightarrow P$
<i>True</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>True</i>
<i>False</i>	<i>True</i>	<i>True</i>	<i>False</i>	<i>True</i>
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<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>False</i>	<i>True</i>

**Satisfiable, but not valid**

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**Example:**

$P$	$Q$	$P \vee Q$	$(P \vee Q) \wedge \neg Q$	$((P \vee Q) \wedge \neg Q) \Rightarrow P$
<i>True</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>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>False</i>	<i>True</i>

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**Example:**

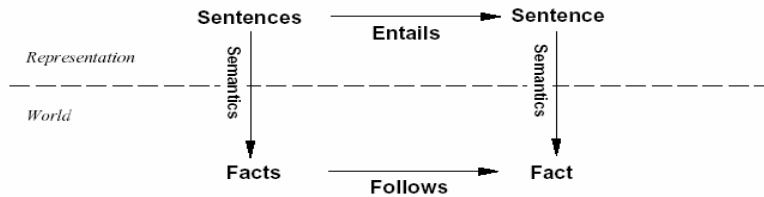
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<i>True</i>	<i>True</i>	<i>True</i>	<i>False</i>	<i>True</i>
<i>True</i>	<i>False</i>	<i>True</i>	<i>True</i>	<i>True</i>
<i>False</i>	<i>True</i>	<i>True</i>	<i>False</i>	<i>True</i>
<i>False</i>	<i>False</i>	<i>False</i>	<i>False</i>	<i>True</i>

**Satisfiable, and valid**

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## Entailment

- **Entailment** reflects the relation of one fact in the world following from the others



- Entailment  $KB \models \alpha$
- Knowledge base KB entails sentence  $\alpha$  **if and only if**  $\alpha$  is true in all worlds where KB is true

## Inference.

- **Inference** is a process by which conclusions are reached.
- **Our goal:**
  - We want to implement the inference process on a computer !!!
- Assume an **inference procedure  $i$**  that
  - derives a sentence  $\alpha$  from the KB :  $KB \vdash_i \alpha$
- **Important:**
  - We need to assure that our inference procedure derives correct conclusions



## Sound and complete inference.

Assume an **inference procedure  $i$**  that

- derives a sentence  $\alpha$  from the KB :  $KB \vdash_i \alpha$

### Properties of the inference procedure:

- **Soundness:** An inference procedure is **sound**

If  $KB \vdash_i \alpha$  then it is true that  $KB \models \alpha$

- **Completeness:** An inference procedure is **complete**

If  $KB \models \alpha$  then it is true that  $KB \vdash_i \alpha$