

# Lecture 22

## Final Review

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# Review

- **Intelligent Agent**
- **Problem Solving Agent**
  - Classical search
  - Constraint satisfaction problem
  - Game tree search
- **Logical Agent**
  - Propositional Logic
  - First-Order Logic
- **Learning-based Agent**
  - Decision Tree
  - KNN algorithm
  - K-means algorithm

# Problem-Solving Agents

- **Problem-solving agent: goal-based agent**
- **Goal formulation**
  - Set of one or more desirable world states
- **Problem formulation**
  - What actions and states to consider given a goal and an initial state
    - States
    - Initial State
    - Actions
    - Transition model
    - Goal test
    - Path cost
- **Search for solution**
  - Given the problem, search for a solution --- a sequence of actions to achieve the goal starting from the initial state
- **Execution of the solution**

# Implementing a Problem-Solving Agent

- Search through the state space using a search tree
- Search tree is generated by the initial state and successor function
  - Expansion of nodes
    - generating a new set of child nodes and adding them to the frontier
  - Search strategy
    - determines the selection of the next node to be expanded
    - can be achieved by ordering the nodes in the frontier
- Search algorithms:
  - Tree-Search: frontier
  - Graph-Search: frontier and explored

# Search Strategies

- **Uninformed search (Blind Search)**
  - Can only generate successors and distinguish goals from non-goals
  - BFS, DFS, UCS
- **Informed search (Heuristic Search)**
  - Know whether one non-goal state is “more promising” than another
  - Greedy Search, A\* Search
- **Performance measure**
  - Complete?** Guaranteed to find a solution when there is one?
  - Optimal?** Find the cheapest solution?
  - Time complexity?** How long does it take to find a solution?
  - Space complexity?** How much memory is needed to perform the search?

# Constraint Satisfaction Problem

- A CSP is defined as a triple  $\langle V, D, C \rangle$ 
  - finite set of **variables**  $V = \{V_1, V_2, \dots, V_n\}$
  - non-empty **domains** of possible values for each variable
$$D = \{D_{V_1}, D_{V_2}, \dots, D_{V_n}\}$$
  - finite set of **constraints**  $C = \{C_1, C_2, \dots, C_m\}$  that specify allowable combinations of values
    - each constraint consists of a pair  $\langle \text{scope}, \text{relation} \rangle$
- A state is an assignment of values to some or all variables.
- A solution to a CSP is a complete and consistent assignment.

# Games

- Two players: MAX and MIN
- MAX moves first and they take turns until the game is over.
- A game can be defined as a search problem:
  - **initial state  $s_0$** : how the game is set up at the start
  - **Player( $s$ )**: which player has the move in state  $s$
  - **Action( $s$ )**: set of legal moves in state  $s$
  - **Result( $s, a$ )**: the state resulting from action  $a$  in state  $s$
  - **Terminal-Test( $s$ )**: true if game is over (terminal state) otherwise false
  - **Utility( $s, p$ )**: a numeric value of terminal state  $s$  for a player  $p$ ,  
e.g., win (+1), lose (-1) and draw (0) in chess.
- The  $s_0$ , **Action( $s$ )** and **Result( $s, a$ )** define the game tree, where the nodes are game states and the edges are moves.
- Players use game tree to determine next move.

# Minimax Algorithm

- **Assumption: both players play optimally.**
- **Given the game tree, the optimal strategy can be determined by using the minimax value of each node  $\text{MiniMax}(s)$ .**

**$\text{MiniMax}(s) =$**

**if  $\text{Terminal-Test}(s)$  then  $\text{Utility}(s)$**

**if  $\text{Player}(s) = \text{MAX}$  then**

**max of  $\text{MiniMax}(\text{Result}(s, a))$  for  $a$  in  $\text{Actions}(s)$**

**if  $\text{Player}(s) = \text{MIN}$  then**

**min of  $\text{MiniMax}(\text{Result}(s, a))$  for  $a$  in  $\text{Actions}(s)$**

**Minimax search problem: complete search is impractical for most games**



# Minimax with Alpha-Beta Pruning

## Key points:

1. Each node has to keep track of 3 values:  $\alpha$ ,  $\beta$ ,  $v$  (minimax value).
2. Pruning condition: if  $\alpha \geq \beta$  for node  $n$ , stop expanding the children of node  $n$  and return its current  $v$
3. MAX will update only  $\alpha$  values and MIN player will update only  $\beta$  values. Both of them will update  $v$  values.
4. Return  $v$  values to parent nodes of the game tree
5. Pass  $\alpha$  and  $\beta$  values to child nodes.

# Logical Agents

- Intelligent agents need **knowledge about the world** for making good decisions.
- The knowledge of an agent is stored in a knowledge base in the form of **sentences** in a **knowledge representation language**.
- A knowledge-based agent needs a **knowledge base** and an **inference mechanism**. It operates by storing sentences in its knowledge base, inferring new sentences with the inference mechanism, and using them to deduce which actions to take.
- A **representation language** is defined by its **syntax** and **semantics**, which specify structure of sentences and how they relate to world facts.
- The **interpretation** of a sentence is the fact to which it refers. If this fact is part of the actual world, then the sentence is true.

# Entailment and Inference

- **$KB \models \alpha$** 
  - If KB entails  $\alpha$ , then all models (assigning 'true' or 'false' values to symbols) that evaluate the KB to True also evaluate  $\alpha$  to True.
- **$KB \vdash \alpha$** 
  - Inference is a procedure for deriving a new sentence  $\alpha$  from KB following some inference approach.
- **Given:**
  - KB: a set of sentences
  - $\alpha$ : a sentence
- **The inference approaches:**
  - Truth-table approach
  - Inference rules
  - Resolution algorithm

# Truth-Table Approach

- **KB  $\models \alpha$  ?**
- **A two steps procedure:**
  - **Generate table for all possible models (n symbols  $\rightarrow 2^n$  entries)**
  - **Check whether the sentence  $\alpha$  is true whenever the sentences in KB are true**
- **To determine if one sentence is satisfiable and valid, we can also use truth-table to list out all the models for it.**
  - **If there exists at least one model that evaluates it to true, it is satisfiable.**
  - **If all the models evaluate it to true, it is valid.**

# Resolution Algorithm

- To show that  $KB \models \alpha$ , we show that  $(KB \wedge \neg\alpha)$  is unsatisfiable

**1.  $(KB \wedge \neg\alpha)$  is converted into CNF**

**2. Apply iteratively the resolution rule 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's not present

**3. Stop when:**

- Contradiction (empty clause  $\{\}$ ) is reached:
  - E.g.,  $P, \neg P \Rightarrow \{\}$
  - **Prove the entailment**
- No more new clauses can be derived
  - **Reject the entailment**

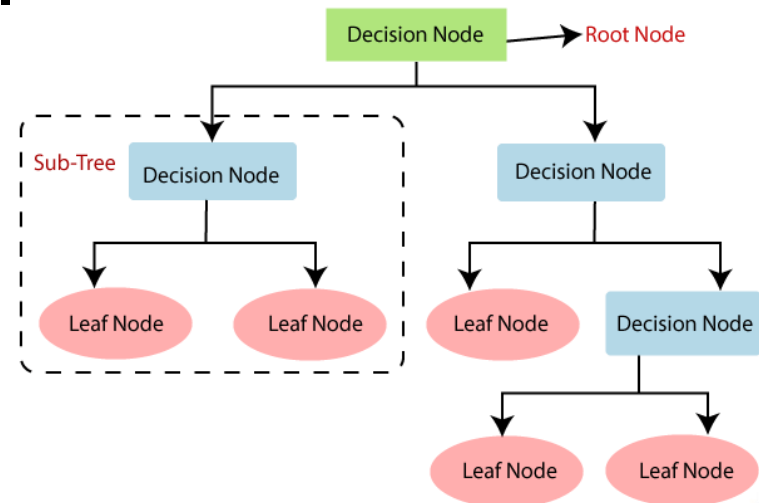
- $\{\}$  is a disjunction of no disjuncts is equivalent to False, thus the contradiction

# Basic Algorithm of Decision Tree

node = root of decision tree

Main loop:

1.  $A \leftarrow$  the “**best**” decision attribute for the next node.
2. Assign  $A$  as decision attribute for node.
3. For each value of  $A$ , create a new child (sub-tree) of the node.
4. Sort training examples to leaf nodes.
5. If training examples are perfectly classified, stop.  
Else, recurse over new leaf nodes.



# Final Exam

## Complete Student Opinion Survey for extra points

- 0am April 25th – 11:59pm April 26th under “Exams” on Canvas
- Time duration: 3 hours
- Open-book and open-note
- A final exam practice and its solutions under “Exams” on Canvas are available
- Total points: 100
- Five Problems
  - 1) Search problem: BFS, DFS, UCS, Greedy, A\*
  - 2) Games: minimax and alpha-beta pruning
  - 3) Inference in propositional logic
  - 4) Inference in propositional logic
  - 5) Decision tree