

# CS 480 NoteSheet

## Chapter 2

### PEAS

- The *Performance* measure
- The *Environment* in which the agent will operate
- The *Actuators* that the agent will use to affect the environment
- The *Sensors* that the agent will use to perceive the environment

### Env prop

- Fully vs partially observable (can be unobservable too)
- Single agent vs multiagent
- multiagent: competitive vs. cooperative
- Deterministic vs. non deterministic (stochastic)
  - non deterministic: next state is NOT completely determined by the current state and agent action
- Episodic vs. sequential
  - sequential: current decision / action COULD affect all future decisions / actions
- Static vs. dynamic
  - Static: environment CANNOT change while the agent is taking its time to decide
- Discrete vs. continuous
  - continuous: time changes are continuous
- Known vs. unknown
  - known: agent knows all outcomes to its actions
  - unknown: learning and exploration can be necessary

### State representations

- Atomic
  - state representation has NO internal structure
- Factored
  - state representation includes fixed attributes (which can have values)
- Structured
  - state representation includes objects and their relationships

### Typical agent arch

- Simple reflex agent
  - uses condition-action rules
- Model-based reflex agent
  - keeps track of the unobserved parts of the environment by maintaining internal state:

- “how the world works”: state transition model
- how percepts and environment is related: sensor model
- Goal-based reflex agent
  - maintains the model of the world and goals to select decisions (that lead to goal)
- Utility-based reflex agent
  - maintains the model of the world and utility function to select PREFERRED decisions (that lead to the best expected utility:  $\text{avg}(\text{EU} * p)$ )

### Search problem: Dracula’s Roadtrip

State Space: a map of Romania

Initial State: Arad

Goal State: Bucharest

Actions:  $\text{ACTIONS}(\text{Arad}) = \{\text{ToSibiu}, \text{ToTimisoara}, \text{ToZerind}\}$

Transition Model:  $\text{RESULT}(\text{Arad}, \text{ToZerind}) = \text{Zerind}$

Action Cost Function [ $\text{ActionCost}(\text{Scurrent}, a, \text{Snext})$ ]:  $\text{ActionCost}(\text{Arad}, \text{ToSibiu}, \text{Sibiu}) = 140$

## Chapter 3

### Search perf

- Completeness: Is the algorithm guaranteed to find a solution when there is one, and to correctly report failure when there is not?
- Cost optimality: Does it find a solution with the lowest path cost of all solutions?

### Informed Search and Heuristics

Informed search relies on domain-specific knowledge / hints that help locate the goal state

An *admissible heuristics* is guaranteed to give you the optimal solution

Every *consistent heuristics*, heuristics that only makes the estimate better, is admissible heuristics, but not the other way around

### Greedy

Single heuristic (eg. distance to goal)

### A\*

heuristic and total path cost (eg. dist. to goal + path cost from initial node)

## Chapter 5

### Min-Max

I don't know what move my opponent will choose, but I am going to ASSUME that it is going to be the best / optimal option

- At every leaf node the MinMax value (utility at leaf node) is calculated,
- For every MAX Player node, the current LARGEST child MinMax value is saved in  $\alpha$
- For every MIN Player node, the current SMALLEST child MinMax value is saved in  $\beta$
- If at a MIN Player node m the current value  $\beta \leq \alpha$ , then the search at node m can end. Here  $\beta$  is the LARGEST value of a MAX Player node in the path from the root to node m,
- If at a MAX Player node n the current value  $\beta \geq \alpha$ , then the search at node n can end. Here  $\alpha$  is the SMALLEST value of a MIN Player node in the path from the root to node n.

## Chapter 6

### Constraint Satisfaction Problem (CSP)

- a set of variables  $X = X_1, \dots, X_n$
- a set of domains  $D = D_1, \dots, D_n$
- a set of constraints C that specify allowable combinations of value
- If NO constraints violated: consistent assignment
- If ALL variables have a value: complete assignment
- If SOME variables have NO value: partial assignment
- SOLUTION: consistent and complete assignment
- PARTIAL SOLUTION: consistent and partial assignment

### Variable Types

- Domains can be
  - finite, for example:  $\{1, 2, 3, 5, 8, 20\}$  (simpler)
  - infinite, for example: a set of all integers
- Variables can be:
  - discrete, for example:  $X = \{X_1, \dots, X_n\}$  (simpler)
  - continuous, for example:  $\mathbb{R}^+$
- Constraints can be:
  - unary (involve single variable), for example:  $X_1 = 5$
  - binary (involve two variables), for example:  $X_1 = X_2$
  - higher order (involve  $> 2$  variables), for example:  $X_1 = X_2 * X_3$

**local consistency** Remove inconsistent values from variable domains as we go as they would make certain assignments inconsistent later anyway

- Node consistency
  - a single variable is node-consistent (in a constraint graph) if all the values in its domain satisfy variable unary constraints
- Arc consistency
  - a single variable is arc-consistent (in a constraint graph) if all the values in its domains satisfy ALL its binary constraints
- Path consistency
  - two variable set  $\{X_i, X_j\}$  is path-consistent (in a constraint graph) with respect to a third variable  $X_m$  if for EVERY assignment  $\{X_i = a, X_j = b\}$  there is an assignment to  $X_m$  (between  $X_i$  and  $X_j$ ) that satisfies constraints on  $\{X_i, X_m\}$  and  $\{X_m, X_j\}$ .

## Chapter 7

$$(((p \implies q) \wedge (r \implies s)) \vee (\neg q \implies \neg s)) \\ ((p \implies q) \implies (qq)p))$$

### Logical Entailment

A set of sentences (called premises) logically entails a sentence (called a conclusion) if and only if every truth assignment that satisfies the premises also satisfies the conclusion

PREMISES  $\square$  CONCLUSION

### Conjunctive Normal Form CNF

A sentence is in CNF if and only if consists of conjunction:  $K_1 \wedge K_2 \wedge \dots \wedge K_m$  of clauses.

A clause  $K_i$  consists of a disjunction  $(l_{i1} \vee l_{i2} \vee \dots \vee l_{ini})$  of literals

eg.  $(a \vee b \vee \neg c) \wedge (a \vee b \vee \neg c) \wedge (\neg b \vee \neg c)$