# MIDTERM REVIEW

#### AI Class 13

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Material from David Matuszek @ Penn

# Bookkeeping

- HW3
- Midterm is next class period, in class
  - Covers **through** multi-agent systems
  - Study material should include all readings
- Project
  - Overview today, details after exam
  - Please fill out Google team form (posted on schedule)
  - If you aren't part of a 2-4 person team or would like more members, you must solve it by next Tuesday

# Today's Class

- Midterm review
- Project overview
- New Eleusis practice
  - Maybe.
  - The ideal # for New Eleusis is 6 players, so we will combine teams
  - You should play with your team, however

# Midterm Review

**10/19** In class

## What Will it Be Like?

- Broadly:
  - Turn a problem description into a solution
  - Work through a problem to reach a solution
  - Demonstrate a conceptual grasp of the material
- Concepts  $\leftarrow \rightarrow$  Algorithms and Implementations
- Basic idea: you need to understand the ideas behind the material we have covered, and be able to apply it to solving problems.

# What Kind of Questions?

- T/F, multiple choice, fill in the blank
- Write definitions of terms
- Work through an {algorithm | solution type | problem}
- Draw something search trees, states, Bayes nets, paths through a map, ...
- Write a **short** answer to English questions
  - E.g.: What approach would you use to solve this problem?
  - E.g.: "We know these are independent. Why?"

#### What Do I Need To Do?

- Homeworks and lectures should be good practice
- Coding questions (not minor syntax mistakes, etc.)
  - We're looking for "I understand this well enough to implement it," not "I know Python really well"
- Look at **homeworks**, **sample problems** in lectures, and **class exercises**
- Look at lectures' "Why?" questions.

# Scoring

- Follow directions.
- Start with a perfect score, mark down for mistakes
  If I ask for 2 examples, and you give 3, one of which is wrong, it's -1, not -<sup>1</sup>/<sub>3</sub>
- Read carefully.
  - You have time.
  - "I didn't see the part that said..."
- Ask for clarification on, e.g., unfamiliar words

# Topics: AI

- What is intelligence?
- What is AI?
- What is it used for? Good for?
- Historical events and figures

# Topics: Agents

- Agents
  - What kinds are there?
  - What do they do?
  - How do we characterize them (what traits do they have)?
    - Autonomy, rationality, ..?
  - How do they interact with an environment?
- Environments
  - What's an environment?
  - How is it characterized?

# Topics: Search

- What is it for?
- Elements of a search problem
  - State spaces, actions, costs, ...
  - How do state spaces pertain to search?
  - To problem-solving?
- Exploring search space (selecting, expanding, generating, evaluating)
- Specific algorithms: How do they work? What are they good for? What are their weaknesses?

## **Topics: Formalizing Search**

- What are the elements of a search problem?
  - "Express [X] as a search problem." What does that mean?
- **States:** every state a puzzle can be in
- Actions/Operations: how you get between states
- **Solutions:** you need a goal test (and sometimes a *heuristic*, or estimate of distance from goal)
  - Sometimes we care about path (planning), sometimes just goal (identification). Can you say which, for a given problem?
- **Costs:** not all solutions or actions are equal

# Topics: Uninformed Search

- Why do uninformed search?
- Come up with some examples of uninformed search problems
- Important algorithms: BFS, DFS, iterative deepening, uniform cost
- A (very) likely question: "What would be the best choice of search method for [*problem*], **and why**?"
- Characteristics of algorithms
  - Completeness, optimality, time and space complexity, ...

## **Topics: Informed Search**

- Some external or pre-existing information says what part of state space is **more likely** to have a solution
- Heuristics encode this information: h(n) ←
  - Pop quiz: What does h(n) = 0 mean?
- Admissibility & Optimality
  - Some algorithms can be optimal when using an admissible heuristic

A heuristic applies to a **node**, can give optimal solution with the right **algorithm** 

- Algorithms: best-first, greedy search, A\*, IDA\*, SMA\*
- What's a good heuristic for a problem? Why?



Apply the following to search this space. At each search step, show: current node being expanded, g(n) (path cost so far), h(n) (heuristic estimate), f(n) (evaluation function), and  $h^*(n)$  (true goal distance).

Depth-first search Uniform-cost search Breadth-first search A\* search Greedy search

## Topics: Local Search

- Idea: Keep a single "current" state, try to improve it
  - Don't keep path to goal
  - Don't keep entire search in memory
  - Go to "successor states"
- Concepts: hill climbing, local maxima/minima, random restarts
- Important algorithms: hill climbing, local beam search, simulated annealing



- If there are loops?
- What's a good algorithm? A bad one?

# Topics: CSPs

- Constraint Satisfaction: subset of search problems
  - 1. State is defined by random variables  $X_i$
  - 2. With values from a **domain** *D*
  - 3. Knowledge about problem can be expressed as **constraints** on what values  $X_i$  can take
- Special algorithms, esp. on constraint networks
- How would you express [X] as a CSP? As a search? How would you represent the constraints?
  - E.g.: "Must be alphabetical"

# Topics: Constraint Networks

- **Constraint propagation**: constraints can be propagated through a constraint network
  - Goal: maintain **consistency** (constraints aren't violated)
- Concepts: Variable ordering, value ordering, fail-first
- Important algorithms:
- 1. Backtracking: DFS, but at each point:
  - Only consider a single variable
  - Only allow legal assignments
- 2. Forward checking



# Topics: Games

- Why play games? What games, and how?
- Characteristics: zero-sum, deterministic, perfect information (or not)
- What's the search tree for a game? How big is it?
- How would you express game [X] as a search? What are the states, actions, etc.? How would you solve it?
- Algorithms: (expecti)minimax, alpha-beta pruning
  - Many examples on slides

# Topics: Basic Probability

- What is uncertainty?
- What are sources of uncertainty in a problem?
  - Non-deterministic, partially observable, noisy observations, noisy reasoning, uncertain cause/effect world model, continuous problem spaces...
- World of all possible states: a complete assignment of values to random variables
- Joint probability, conditional probability

#### Topics: Basic Probability

- Independence: A and B are independent
  - $P(A) \perp P(B)$  iff  $P(A \land B) = P(A) P(B)$
  - A and B do not affect each other's probability
- Conditional independence: A and B are independent *given C* 
  - $P(A \land B \mid C) = P(A \mid C) P(B \mid C)$
  - A and B don't affect each other if C is known

# **Topics: Basic Probability**

• 
$$P(a \mid b) = \frac{P(a \land b)}{P(b)}$$
 •  $P(a \land b) = P(a \mid b) P(b)$ 

P(smart ∧	smart		¬smart	
study ∧ prep)≈	study	¬study	study	¬study
prepared	.432	.16	.084	.008
¬prepared	.048	.16	.036	.072

- What is the prior probability of smart?
- What is the conditional probability of prepared, given study and smart?
- Is prepared independent of study?

# Topics: Probabilistic Reasoning

#### • Concepts:

- Posteriors and Priors; Bayesian Reasoning; Induction and Deduction; Probabilities of Events
- [In]dependence, conditionality, marginalization
- What is Bayes' Rule and what is it useful for?

$$\begin{split} P(H_i \mid E_j) &= \frac{P(E_j \mid H_i) P(H_i)}{P(E_j)} \\ P(cause \mid effect) &= \frac{P(effect \mid cause) P(cause)}{P(effect)} \end{split}$$

# **Topics: Joint Probability**

What is the **joint probability** of A and B?

- P(A,B)
- The probability of any set of legal assignments.
- Booleans: expressed as a matrix/table

	alarm	¬ alarm
burglary	0.09	0.01
¬ burglary	0.1	0.8



• Continuous domains  $\rightarrow$  probability functions

# Conditional Probability Tables

- For  $X_i$ , CPD  $P(X_i | Parents(X_i))$  quantifies effect of parents on  $X_i$
- **Parameters** are probabilities in conditional probability tables (CPTs):



Example from web.engr.oregonstate.edu / ~wong / slides / BayesianNetworksTutorial.ppt

# **BBN** Definition

- AKA Bayesian Network, Bayes Net
- A graphical model (as a DAG) of probabilistic relationships among a set of random variables
- Links represent direct influence of one variable on another



#### More Complex Bayesian Network



Slides from Dr. Oates, UMBC

#### More Complex Bayesian Network



# Age and Gender are independent.

P(A,G) = P(G) \* P(A)

P(A | G) = P(A)P(G | A) = P(G)

P(A,G) = P(G|A) P(A) = P(G)P(A)P(A,G) = P(A|G) P(G) = P(A)P(G)

Slides from Dr. Oates, UMBC

#### Examples of Worked BBNs

#### http://tiny.cc/bn-ex

- https://www.ics.uci.edu/~rickl/courses/cs-171/[etc]/ cs-171-17-BayesianNetworks.pdf
- http://tiny.cc/bn-ex2
  - http://chem-eng.utoronto.ca/~datamining/ Presentations/Bayesian\_Belief\_Network.pdf
- https://cw.fel.cvut.cz/wiki/\_media/courses/ ae4m33rzn/bn\_solved.pdf

## Topics: Reasoning Under Uncertainty

- How is the world represented over time?
  - Concepts: timesteps, world, observations
  - Transition model captures how the world changes
  - Sensor model capture what we see, given some world
  - Markov assumption (first-order) makes it all tractable
- What can we do with it?
  - Concepts: Filtering, predicting, smoothing, explaining

## Topics: Reasoning Under Uncertainty

- How would you represent this problem as a network and set of conditional probability tables?
  - The weather has a 30% chance of changing and a 70% chance of staying the same.
  - If it's raining, the probability of seeing someone carrying an umbrella is 90%; if it's not raining, it's 20%.
- I saw umbrellas Monday and Tuesday, but not today. What is the most likely weather pattern for those days?



# Topics: Utility

- How should rational agents make decisions?
- Concepts: rationality, utility functions, value functions, expected value, satisficing, preferences
- Utility is a function of world states
- Must have some preferences that pertain to perceived needs or wants

# **Topics: Decision Theory**

- What is the expected utility of an action?
  - Broadly: its probability times its value
  - The sum of that for all possible outcomes
- Maximum Expected Utility principle





# Topics: Nash Equilibrium

- Occurs when each player's strategy is optimal, given strategies of the other players
- No player benefits by **unilaterally** changing strategy while others stay fixed
- Example questions:
  - What strategy should you choose? Why?
  - What strateg(ies) are in a Nash equilibrium?

B	Confesses	Denies
Confesses	<b>(3, 3)</b>	(0, 5)
Denies	(5, 0)	(1, 1)



#### Various Reminders

- Everything in the readings is fair game.
- Look at homeworks, sample problems in lectures.
- Look at lectures' "Why?" questions.
- Slides are a good source of **conceptual** understanding.
- Book goes into **detail** and explains more deeply.