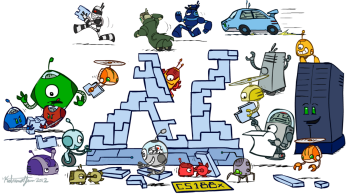


## CS 473: Artificial Intelligence Conclusion



Dieter Fox – University of Washington

[Many of these slides were created by Dan Klein and Pieter Abbeel for CS188 Intro to AI at UC Berkeley. All CS188 materials are available at <http://ai.berkeley.edu>.]

## Exam Topics

- **Search**
  - Problem spaces
  - BFS, DFS, UCS, A\* (tree and graph), local search
  - Completeness and Optimality
  - Heuristics: admissibility and consistency; pattern DBs
- **CSPs**
  - Constraint graphs, backtracking search
  - Forward checking, AC3 constraint propagation, ordering heuristics
- **Games**
  - Minimax, Alpha-beta pruning.
  - Expectimax
  - Evaluation Functions
- **MDPs**
  - Bellman equations
  - Value iteration, policy iteration
- **Reinforcement Learning**
  - Exploration vs Exploitation
  - Model-based vs. model-free
  - Q-learning
  - Linear value function approx.
  - Inverse optimal control
- **Hidden Markov Models**
  - DBNs
  - Forward algorithm
  - Particle Filters
- **Bayesian Networks**
  - Basic definition, independence (d-sep)
  - Variable elimination
- **Learning**
  - BN parameters with complete data
  - Search through space of BN structures
  - Expectation maximization

## What is intelligence?

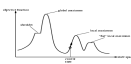
- **(bounded) Rationality**
  - Agent has a performance measure to optimize
  - Given its state of knowledge
  - Choose optimal action
  - With limited computational resources
- **Human-like intelligence/behavior**

## Search in Discrete State Spaces

- **Every discrete problem can be cast as a search problem.**
  - states, actions, transitions, cost, goal-test
- **Types**
  - **uninformed systematic:** often slow
    - DFS, BFS, uniform-cost, iterative deepening
  - **Heuristic-guided: better**
    - Greedy best first, A\*
    - relaxation leads to heuristics
  - **Local: fast, fewer guarantees; often local optimal**
    - Hill climbing and variations
    - Simulated Annealing: global optimal

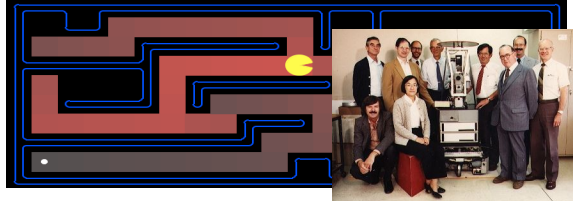
|   |   |   |
|---|---|---|
| 7 | 2 | 4 |
| 6 | 5 | 8 |
| 8 | 3 | 1 |

|   |   |
|---|---|
| 1 | 2 |
| 3 | 4 |
| 5 | 6 |
| 7 | 8 |



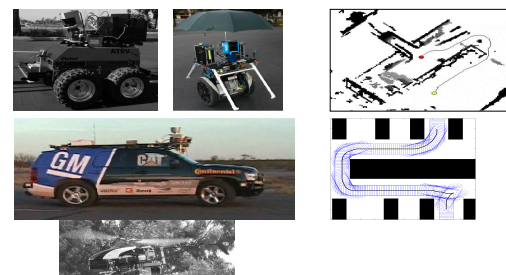
## Which Algorithm?

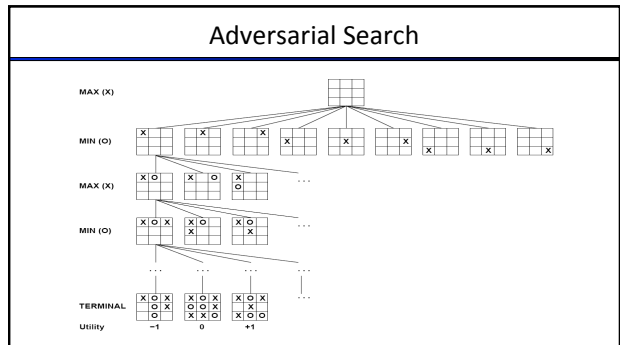
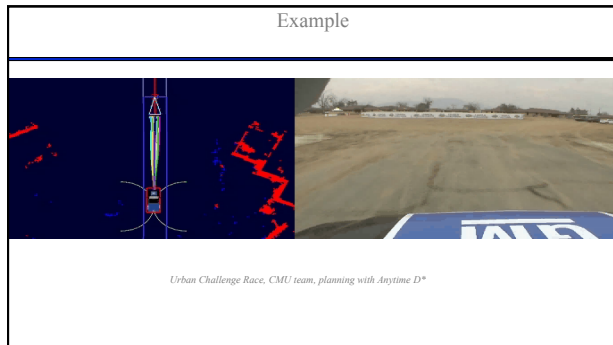
- **A\*, Manhattan Heuristic**



## Motion/Path Planning

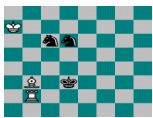
Examples (of what is usually referred to as path planning):





### Adversarial Search

- AND/OR search space (max, min)
- minimax objective function
- minimax algorithm (~dfs)
  - alpha-beta pruning
- Utility function for partial search
  - Learning utility functions by playing with itself
- Openings/Endgame databases

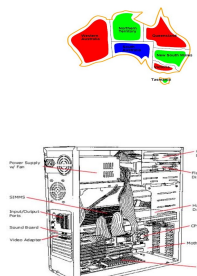


### Knowledge Representation and Reasoning

- **Representing: what agent knows**
  - Propositional logic
  - Constraint networks
  - HMMs
  - Bayesian networks
  - ...
- **Reasoning: what agent can infer**
  - Search
  - Dynamic programming
  - Preprocessing to simplify

### Search+KR&R Example: CSP

- **Representation**
  - Variables, Domains, Constraints
- **Reasoning:**
  - Arc Consistency (k-Consistency)
  - Solving
    - Backtracking search: partial var assignments
      - Heuristics: min remaining values, min conflicts
    - Local search: complete var assignments

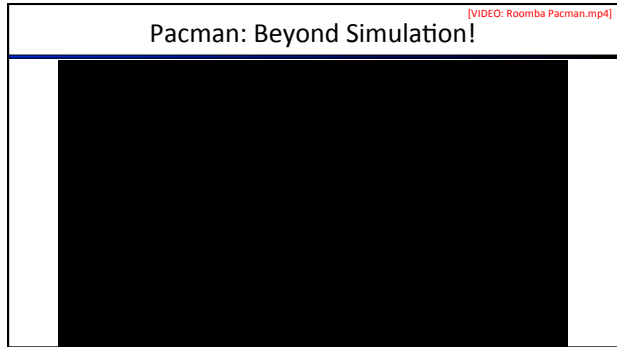
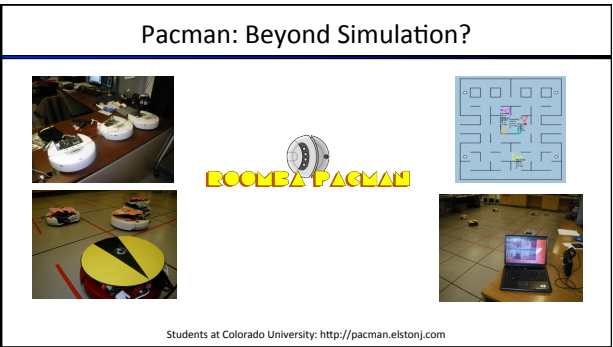
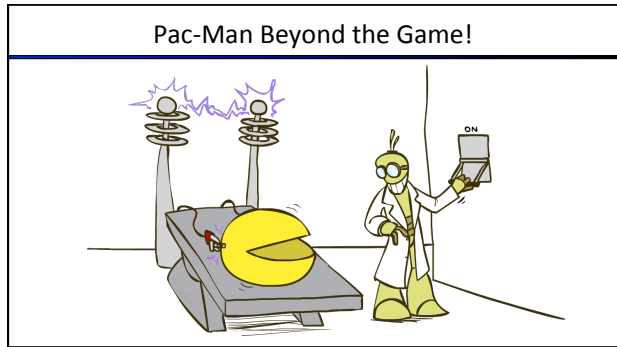
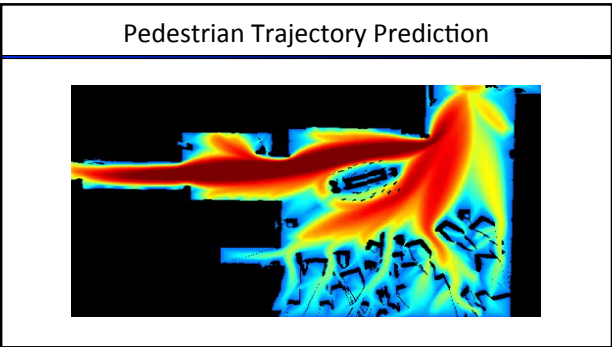
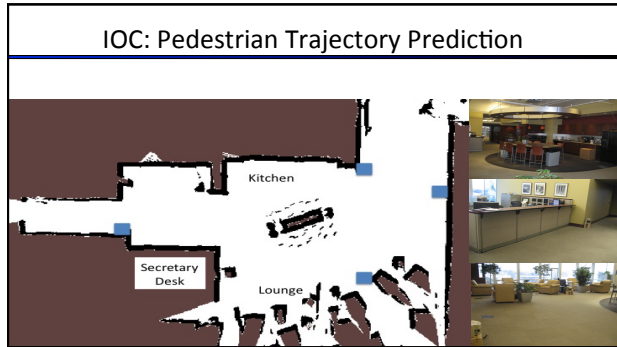


### KR&R: Markov Decision Process

- **Representation**
  - states, actions, probabilistic outcomes, rewards
- **Reasoning:  $V^*(s)$** 
  - Expectimax
  - Value Iteration: dynamic programming
- **Reinforcement Learning:**
  - Exploration / exploitation
  - Learn model or learn Q-function?

$$V^*(s) = \max_a Q^*(s, a)$$

$$Q^*(s, a) = \sum_{s'} T(s, a, s') [R(s, a, s') + \gamma V^*(s')]$$

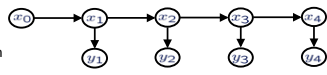




### KR&R: Probability

- **Representation: Bayesian Networks**
  - encode probability distributions compactly
    - by exploiting conditional independences
- **Reasoning**
  - Exact inference: var elimination
  - Approx inference: sampling based methods
    - rejection sampling, likelihood weighting, MCMC/Gibbs

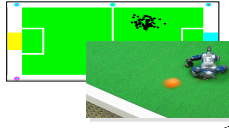

### KR&R: Hidden Markov Models

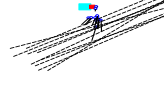
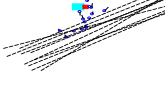
- **Representation**
  - Sequence model
  - One hidden state, one observation
- **Reasoning/Search**
  - most likely state sequence: Viterbi algorithm
  - marginal prob of one state: forward-backward

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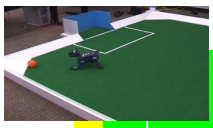
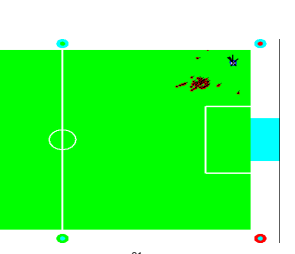
### Ball-Environment Interaction

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### Ball-Environment Interaction

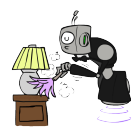
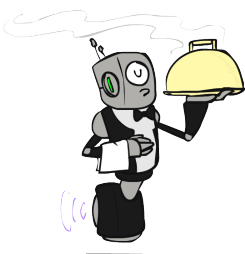



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### Learning Bayes Networks

- **Learning Structure of Bayesian Networks**
  - Search through space of BN structures
- **Learning Parameters for a Bayesian Network**
  - Fully observable variables
    - Maximum Likelihood (ML), MAP & Bayesian estimation
    - Example: Naive Bayes for text classification
  - Hidden variables
    - Expectation Maximization (EM)

### Personal Robotics

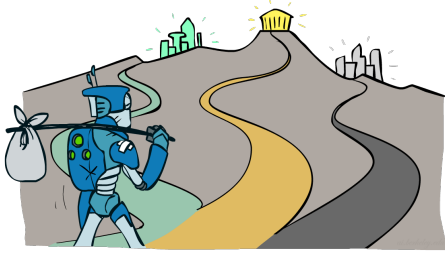



### PR2 (autonomous)



[VIDEO: 5pile\_200k.mp4]  
[Maitin-Shepard, Cusumano-Towner, Lei, Abbeel, 2010]

## Where to Go Next?



## That's It!

- Help out with some course evaluations
- Have a great summer, and always maximize your expected utilities!

