# Introduction to Artificial Intelligence

#### Informed Search



Instructor: Tatjana Petrov

University of Trieste, Italy

[slides adapted from Dan Klein, Pieter Abbeel, Stuart Russell, et al for CS188 Intro to AI at UC Berkeley. All materials available at http://ai.berkeley.edu.]

# Today

- Informed Search
	- Heuristics
	- § Greedy Search
	- $A^*$  Search



### Recap: Search



#### Recap: Search

- Search problem:
	- States (configurations of the world)
	- Actions and costs
	- Successor function (world dynamics)
	- Start state and goal test

#### ■ Search tree:

- Nodes: represent plans for reaching states
- Plans have costs (sum of action costs)
- Search algorithm:
	- Systematically builds a search tree
	- Chooses an ordering of the fringe (unexplored nodes)
	- Optimal: finds least-cost plans



### The One Queue

#### All these search algorithms are the same except for fringe strategies

- Conceptually, all fringes are priority queues (i.e. collections of nodes with attached priorities)
- Practically, for DFS and BFS, you can avoid the log(n) overhead from an actual priority queue, by using stacks and queues
- Can even code one implementation that takes a variable queuing object



#### Search and Models

#### Search operates over models of the world

The agent doesn't actually try all the plans out in the real world!

Planning is all "in simulation"

Your search is only as good as your models…



#### Example: Pancake Problem



Cost: Number of pancakes flipped

#### Example: Pancake Problem

#### **BOUNDS FOR SORTING BY PREFIX REVERSAL**

William H. GATES

Microsoft, Albuquerque, New Mexico

Christos H. PAPADIMITRIOU\*†

Department of Electrical Engineering, University of California, Berkeley, CA 94720, U.S.A.

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For a permutation  $\sigma$  of the integers from 1 to n, let  $f(\sigma)$  be the smallest number of prefix reversals that will transform  $\sigma$  to the identity permutation, and let  $f(n)$  be the largest such  $f(\sigma)$ for all  $\sigma$  in (the symmetric group) S<sub>n</sub>. We show that  $f(n) \le (5n+5)/3$ , and that  $f(n) \ge 17n/16$  for n a multiple of 16. If, furthermore, each integer is required to participate in an even number of reversed prefixes, the corresponding function  $g(n)$  is shown to obey  $3n/2 - 1 \le g(n) \le 2n + 3$ .

#### Example: Pancake Problem

State space graph with costs as weights



#### General Tree Search



#### Uninformed Search



#### Uniform Cost Search

■ Strategy: expand lowest path cost

■ The good: UCS is complete and optimal!

- The bad:
	- Explores options in every "direction"
	- No information about goal location





#### Video of Demo Contours UCS Empty



#### Video of Demo Contours UCS Pacman Small Maze



#### Informed Search



#### Search Heuristics

- An heuristic function  $h(n)$ :
	- **Extimates how close a state**  $n$  **is to a goal**
	- **•** Designed for a particular search problem
	- Examples: Manhattan distance, Euclidean distance for pathing







#### Example: Heuristic Function



 $h(x)$ 

#### Example: Heuristic Function

Heuristic: the number of the largest pancake that is still out of place



### Greedy Search





## Greedy Search

- Strategy: expand a node that you think is closest to a goal state
	- Heuristic: estimate of distance to nearest goal for each state

- A common case:
	- Best-first takes you straight to the (wrong) goal

■ Worst-case: like a badly-guided DFS





### Video of Demo Contours Greedy (Empty)



#### Video of Demo Contours Greedy (Pacman Small Maze)



#### A\* Search



#### A\* Search





UCS

Greedy



### Combining UCS and Greedy

- § Uniform-cost orders by path cost, or *backward cost* g(n)
- § Greedy orders by goal proximity, or *forward cost* h(n)



A\* Search orders by the sum:  $f(n) = g(n) + h(n)$ 

Example: Teg Grenager

#### When should A\* terminate?

#### ■ Should we stop when we enqueue a goal?



■ No: only stop when we dequeue a goal

#### Is A\* Optimal?



- What went wrong?
- Actual bad goal cost < estimated good goal cost
- We need estimates to be less than actual costs!

#### Admissible Heuristics



### Idea: Admissibility



Inadmissible (pessimistic) heuristics break optimality by trapping good plans on the fringe



Admissible (optimistic) heuristics slow down bad plans but never outweigh true costs

#### Admissible Heuristics

§ A heuristic *h* is *admissible* (optimistic) if:

 $0 \leq h(n) \leq h^*(n)$ 

where  $h^*(n)$  is the true cost to a nearest goal

■ Examples:





■ Coming up with admissible heuristics is most of what's involved in using A\* in practice.

#### Consistent Heuristics

A heuristic h is consistent if, for every node  $n$  and every successor  $n'$  of  $n$  generated by an action  $a$ , we have:

$$
h(n) \leq c(n, a, n') + h(n')
$$



### Optimality of A\* Tree Search



### Optimality of A\* Tree Search

#### Assume:

- A is an optimal goal node
- B is a suboptimal goal node
- § h is admissible

#### Claim:

■ A will exit the fringe before B



#### Proof:

- Imagine B is on the fringe
- Some ancestor *n* of A is on the fringe, too (maybe A!)
- Claim: *n* will be expanded before B 1.  $f(n) \leq f(A)$



…

 $\, B \,$ 

#### 1.  $f(n) \leq f(A)$

- Definition of f-cost says:  $f(n) = g(n) + h(n)$  = (path cost to n) + (est. cost of n to A)  $f(A) = g(A) + h(A)$  =(path cost to A) + (est. cost of A to A)
- The admissible heuristic must underestimate the true cost  $h(A) = (est. cost of A to A) = 0$
- So now, we have to compare:

 $f(n) = g(n) + h(n)$  $f(A) = g(A)$ 

•  $h(n)$  must be an underestimate of the true cost from n to A  $g(n) + h(n) \leq g(A)$  $f(n) \leq f(A)$ 

#### Proof:

- Imagine B is on the fringe
- Some ancestor *n* of A is on the fringe, too (maybe A!)
- Claim: *n* will be expanded before B
	- 1.  $f(n) \leq f(A)$ 2.  $f(A) \leq f(B)$



…

 $\overline{B}$ 

#### 2.  $f(A) \leq f(B)$

■ We know that:

 $f(A) = g(A) + h(A)$  = (path cost to A) + (est. cost of A to A)  $\boldsymbol{n}$  $f(B) = g(B) + h(B)$  = (path cost to B) + (est. cost of B to B)

- The heuristic must underestimate the true cost:  $h(A) = h(B) = 0$
- So now, we have to compare:  $f(A) = g(A)$  $f(B) = g(B)$
- We assumed that B is suboptimal! So

 $g(A) < g(B)$  $f(A) < f(B)$ 

#### Proof:

- Imagine B is on the fringe
- Some ancestor *n* of A is on the fringe, too (maybe A!)
- Claim: *n* will be expanded before B
	- 1.  $f(n) \leq f(A)$
	- 2.  $f(A) < f(B)$
	- 3. *n* expands before B  $\equiv$
- All ancestors of A expand before B
- § A expands before B
- § A\* search is optimal



 $f(n) \leq f(A) < f(B)$ 

# Properties of A\*

#### Properties of A\*



### UCS vs A\* Contours

■ Uniform-cost expands equally in all "directions"

 $\blacksquare$  A\* expands mainly toward the goal, but does hedge its bets to ensure optimality





#### Video of Demo Contours (Empty) -- UCS



#### Video of Demo Contours (Empty) -- Greedy



#### Video of Demo Contours (Empty) - A\*



#### Video of Demo Contours (Pacman Small Maze) - A\*



#### Comparison



Greedy

#### **Uniform Cost**

### A\* Applications



# A\* Applications

- Video games
- Pathing / routing problems
- Resource planning problems
- Robot motion planning
- Language analysis

§ …



## Video of Demo Pacman (Tiny Maze) - UCS / A\*



#### Video of Demo Empty Water Shallow/Deep – Guess Algorithm



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