

CS 540 Introduction to Artificial Intelligence Search II: Informed Search University of Wisconsin-Madison

Spring 2022

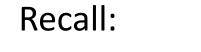
Outline

- Uninformed vs Informed Search
 - Review of uninformed strategies, adding heuristics
- A* Search
 - Heuristic properties, stopping rules, analysis
- Extensions: Beyond A*
 - Iterative deepening, beam search

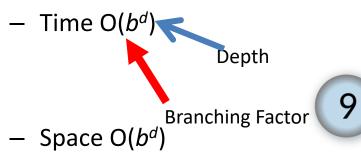
Breadth-First Search

5

10



- Data structure: queue
- Properties:
 - Complete
 - Optimal (if edge cost 1)



11

Q

12

3

2

6

Uniform Cost Search

Like BFS, but keeps track of cost

- Expand least cost node
- Data structure: priority queue
- Properties:
 - Complete
 - Optimal (if weight lower bounded by ϵ)
 - Time O($b^{C^*/\epsilon}$)
 - Space $O(b^{C^*/\varepsilon})$

Optimal goal path cost

Depth-First Search

3

5

2

6

Recall:

- Data structure: stack
- Properties:
 - Incomplete (stuck in infinite tree...)
 - Suboptimal
 - Time $O(b^m)$

Max Depth

- Space O(bm)

12

8

Q

10

Iterative Deepening DFS

Repeated depth-limited DFS

- Search like BFS, fringe like DFS
- Properties:
 - Complete
 - Optimal (if edge cost 1)
 - Time $O(b^d)$
 - Space O(bd)

A good option!

Uninformed vs Informed Search

hls

als

Uninformed search (all of what we saw). Knows:

- Path cost *g*(*s*) from start to node *s*
- Successors. start s



goa

Informed search. Knows:

• All uninformed search properties, plus

start

• Heuristic h(s) from s to goal (recall game heuristic)

Informed Search

Informed search. Know:

- All uninformed search properties, plus
- Heuristic *h(s)* from *s* to goal (recall game heuristic)

• Like in games, use information to **speed up search**.

Using the Heuristic

Back to uniform-cost search

- We had the priority queue
- Expand the node with the smallest g(s)
 - g(s) "first-half-cost"



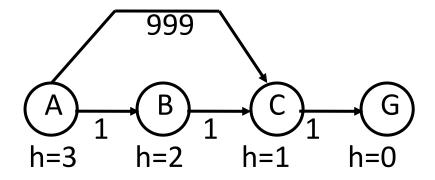
• Now let's use the heuristic ("second-half-cost")

- Several possible approaches: let's see what works

Attempt 1: Best-First Greedy

One approach: just use *h(s)* alone

- Specifically, expand node with smallest *h(s)*
- This isn't a good idea. Why?

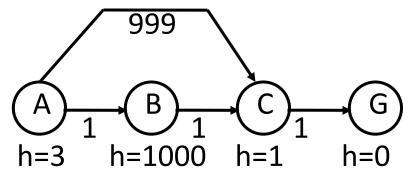


• Not optimal! **Get** $A \rightarrow C \rightarrow G$. **Want**: $A \rightarrow B \rightarrow C \rightarrow G$

Attempt 2: A Search

Next approach: use both *g*(*s*) + *h*(*s*) alone

- Specifically, expand node with smallest g(s) + h(s)
- Again, use a priority queue
- Called "A" search



• **Still not optimal!** (Does work for former example).

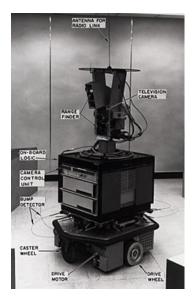
Attempt 3: A* Search

Same idea, use g(s) + h(s), with one requirement

- Demand that *h*(*s*) ≤ *h**(*s*), for all s
- If heuristic has this property, "admissible"
 - Optimistic! Never over-estimates
- Still need $h(s) \ge 0$
 - Negative heuristics can lead to strange behavior
- This is **A* search**

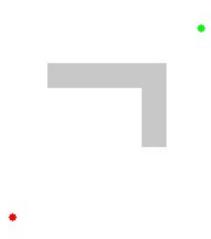
Attempt 3: A* Search

Origins: robots and planning



Shakey the Robot, 1960's

Credit: Wiki



Animation: finding a path around obstacle

Credit: Wiki

Admissible Heuristic Functions

Have to be careful to ensure admissibility (optimism!)

• Example: 8 Game

Example State	1		5	Goal State	1	2	3
	2	6	3	State	4	5	6
	7	4	8		7	8	

• One useful approach: relax constraints

- h(s) = number of tiles in wrong position

• allows tiles to fly to destination in a single step

Q 1.1: Consider finding the fastest driving route from one US city to another. Measure cost as the number of hours driven when driving at the speed limit. Let h(s) be the number of hours needed to ride a bike from city s to your destination. h(s) is

- A. An admissible heuristic
- B. Not an admissible heuristic

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- A. An admissible heuristic No: riding your bike take longer.
- B. Not an admissible heuristic

- **Q 1.2**: Which of the following are admissible heuristics?
- (i) **h(s)** = **h***(s)
- (ii) **h(s)** = max(2, **h*(s)**)
- (iii) **h(s)** = min(2, **h*(s)**)
- (iv) $h(s) = h^*(s)-2$
- (v) **h(s)** = sqrt(**h***(s))
- A. All of the above
- B. (i), (iii), (iv)
- C. (i), (iii)
- D. (i), (iii), (v)

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Q 1.2: Which of the following are admissible heuristics?

- (i) **h(s)** = **h*(s)**
- (ii) $h(s) = \max(2, h^*(s))$ No: h(s) might be too big
- (iii) **h(s)** = min(2, **h*(s)**)
- (iv) $h(s) = h^*(s)-2$ No: h(s) might be negative
- (v) $h(s) = \operatorname{sqrt}(h^*(s))$ No: if $h^*(s) < 1$ then h(s) is bigger
- A. All of the above
- B. (i), (iii), (iv)
- C. (i), (iii)
- D. (i), (iii), (v)

Heuristic Function Tradeoffs

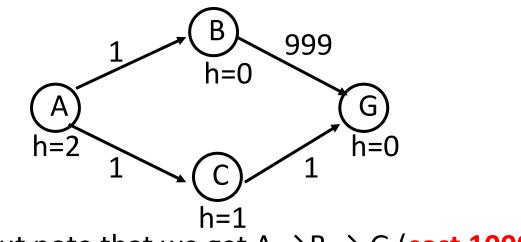
Dominance: h_2 dominates h_1 if for all states s, $h_1(s) \le h_2(s) \le h^*(s)$

- Idea: we want to be as close to *h** as possible
 - But not over!
- **Tradeoff**: being very close might require a very complex heuristic, expensive computation
 - Might be better off with cheaper heuristic & expand more nodes.

A* Termination

When should A* **stop**?

• One idea: as soon as we reach goal state?

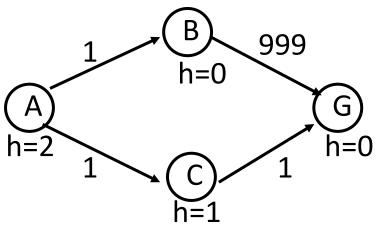


• *h* admissible, but note that we get $A \rightarrow B \rightarrow G$ (cost 1000)!

A* Termination

When should A* stop?

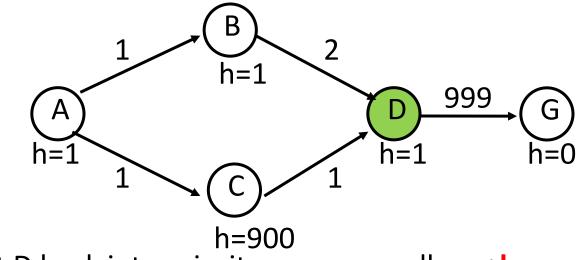
• Rule: terminate when a goal is popped from queue.



• Note: taking **h** =0 reduces to uniform cost search rule.

A* Revisiting Expanded States

Possible to revisit an expanded state, get a shorter path:



Put D back into priority queue, smaller g+h

A* Full Algorithm

- **1.** Put the start node **S** on the priority queue, called **OPEN**
- **2.** If OPEN is empty, exit with failure
- Remove from OPEN and place on CLOSED a node n for which f(n) is minimum (note that f(n)=g(n)+h(n))
- 4. If n is a goal node, exit (trace back pointers from n to S)
- 5. Expand n, generating all successors and attach to pointers back to n. For each successor n' of n
 - If n' is not already on OPEN or CLOSED estimate h(n'), g(n')=g(n)+ c(n,n'), f(n')=g(n')+h(n'), and place it on OPEN.
 - If n' is already on OPEN or CLOSED, then check if g(n') is lower for the new version of n'. If so, then:
 - 1. Redirect pointers backward from n' along path yielding lower g(n').
 - 2. Put n' on OPEN.
 - 3. If g(n') is not lower for the new version, do nothing.
- **6.** Goto 2.

A* Analysis

Some properties:

- Terminates!
- A* can use lots of memory: O(# states).
- Will run out on large problems.

• Next, we will consider some alternatives to deal with this.

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Q 2.1: Consider two heuristics for the 8 puzzle problem. h_1 is the number of tiles in wrong position. h_2 is the l_1 /Manhattan distance between the tiles and the goal location. How do h_1 and h_2 relate?

- A. **h**₂ dominates **h**₁
- B. **h**₁ dominates **h**₂
- C. Neither dominates the other

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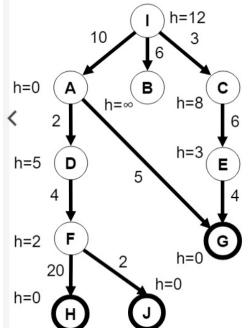
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- A. h₂ dominates h₁
- B. h₁ dominates h₂ (No: h₁ is a distance where each entry is at most 1, h₂ can be greater)
- C. Neither dominates the other

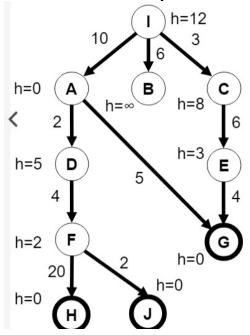
Q 2.2: Consider the state space graph below. Goal states have bold borders. h(s) is show next to each node. What node will be expanded by A* after the initial state I?

- A. A
- B. B
- C. C



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IDA*: Iterative Deepening A*

Similar idea to our earlier iterative deepening.

- Bound the memory in search.
- At each phase, don't expand any node with g(s) + h(s) > k,
 - Assuming integer costs, do this for k=0, then k=1, then k=2, and so on

- Complete + optimal, might be costly time-wise
 - Revisit many nodes
- Lower memory use than A*

IDA*: Properties

How many restarts do we expect?

• With integer costs, optimal solution C*, at most C*

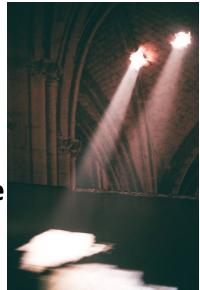
What about non-integer costs?

- Initial threshold k. Use the same rule for non-expansion
- Set new k to be the min g(s) + h(s) for non-expanded nodes
- Worst case: restarted for each state

Beam Search

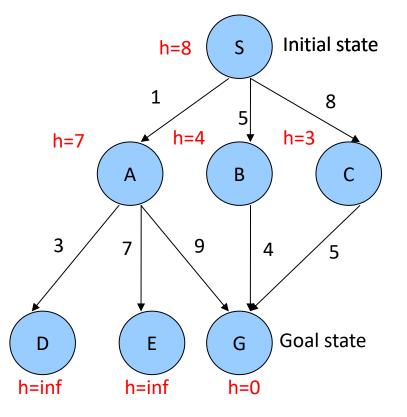
General approach (beyond A* too)

- Priority queue with fixed size k; beyond k nodes, discard!
- **Upside**: good memory efficiency
- **Downside**: not complete or optimal
- Variation:
- Priority queue with nodes that are at most ε worse than best node.

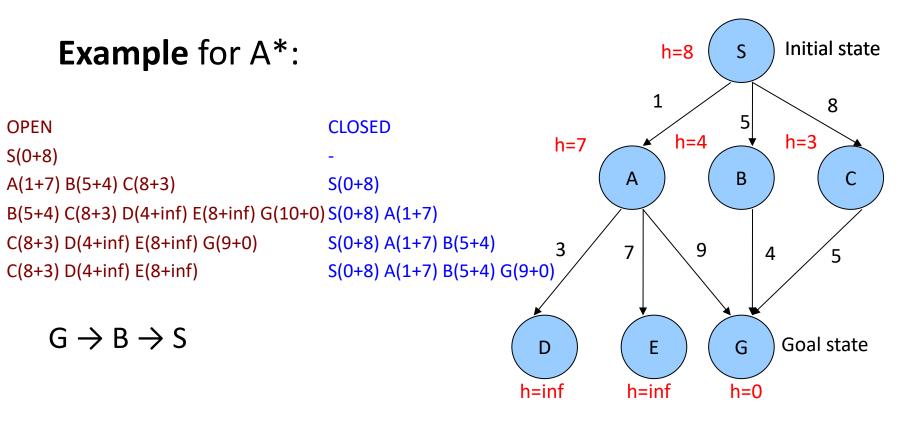


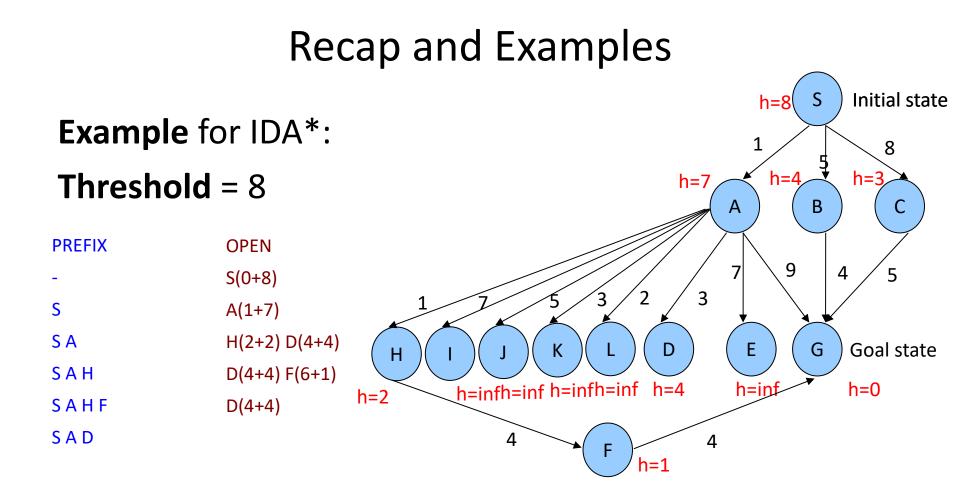
Recap and Examples

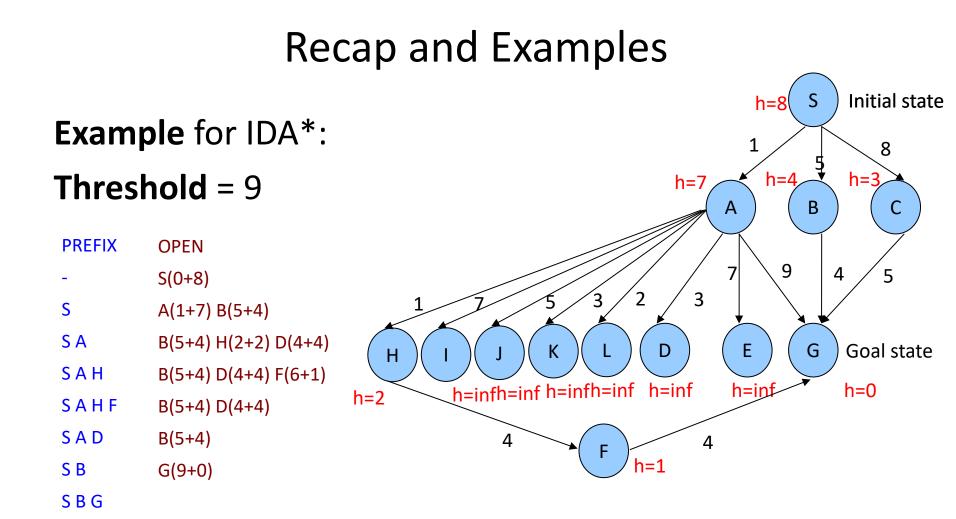
Example for A*:

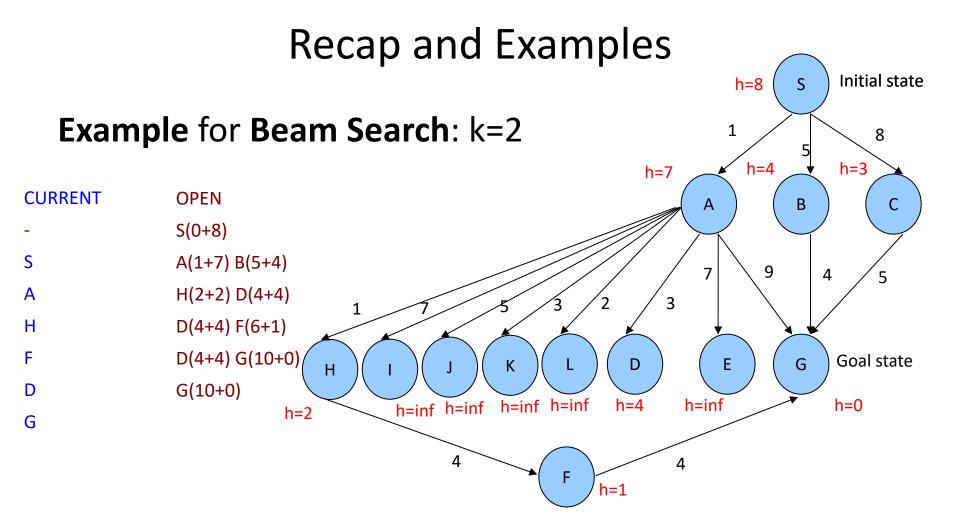


Recap and Examples









Summary

- Informed search: introduce heuristics
 - Not all approaches work: best-first greedy is bad
- A* algorithm
 - Properties of A*, idea of admissible heuristics
- Beyond A*
 - IDA*, beam search. Ways to deal with space requirements.