

Informed Search

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[Based on slides from Jerry Zhu, Andrew Moore <http://www.cs.cmu.edu/~awm/tutorials>]

Main messages

- A*. Always be optimistic.



Uninformed vs. informed search

- **Uninformed search** (BFS, uniform-cost, DFS, ID etc.)
 - Knows the actual path cost $g(s)$ from start to a node s in the fringe, but that's it.



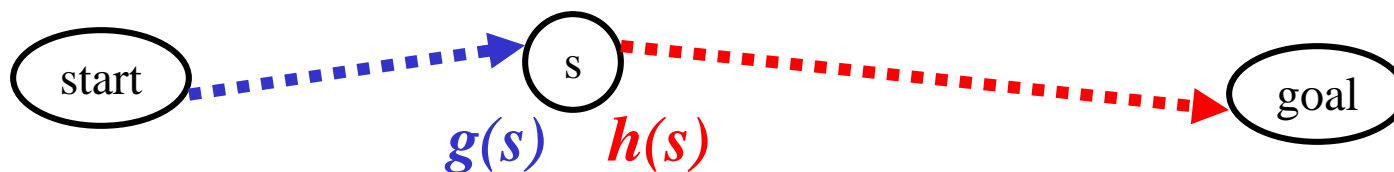
- **Informed search**



- also has a heuristic $h(s)$ of the cost from s to goal. ('h'= heuristic, non-negative)
- Can be **much faster** than uninformed search.

Recall: Uniform-cost search

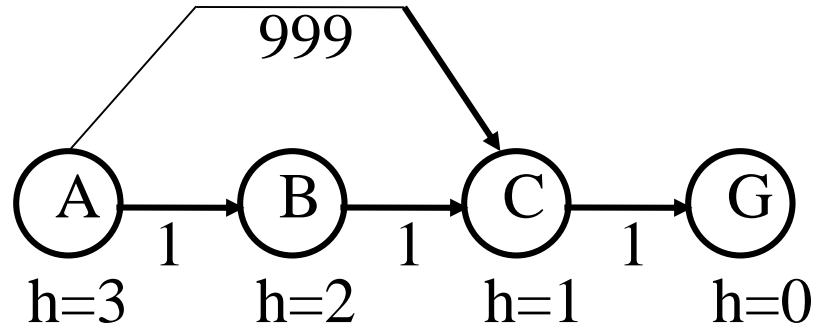
- Uniform-cost search: uninformed search when edge costs are not the same.
- Complete (will find a goal). Optimal (will find the least-cost goal).
- Always expand the node with the least $g(s)$
 - Use a **priority queue**:
 - Push in states with their first-half-cost $g(s)$
 - Pop out the state with the least $g(s)$ first.
- Now we have an estimate of the second-half-cost $h(s)$, how to use it?



First attempt: Best-first greedy search

- Idea 1: use $h(s)$ instead of $g(s)$
- Always expand the node with the least $h(s)$
 - Use a priority queue:
 - Push in states with their second-half-cost $h(s)$
 - Pop out the state with the least $h(s)$ first.
- Known as “best first greedy” search
- How’s this idea?

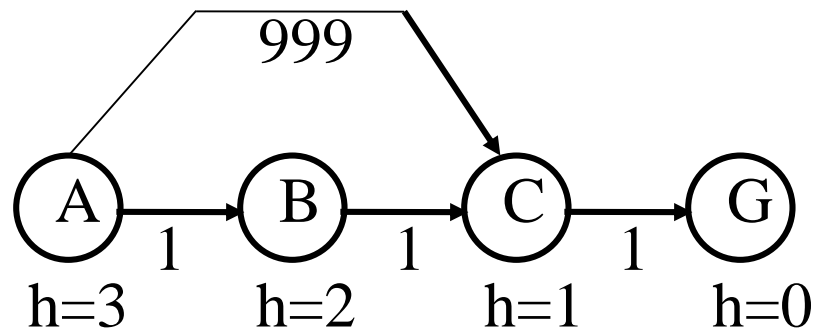
Best-first greedy search looking stupid



- It will follow the path $A \rightarrow C \rightarrow G$ (why?)
- Obviously not optimal

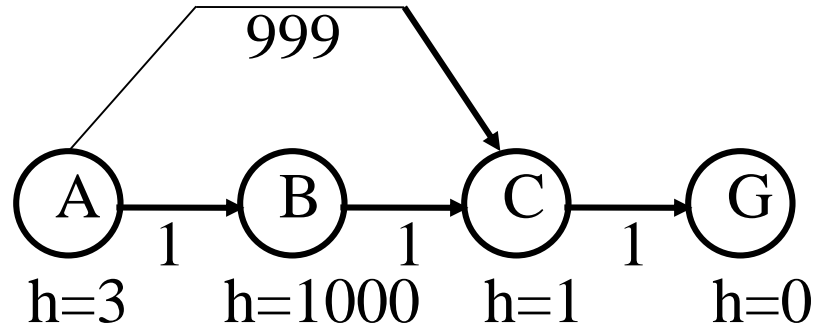
Second attempt: A search

- Idea 2: use $g(s)+h(s)$
- Always expand the node with the least $g(s)+h(s)$
 - Use a priority queue:
 - Push in states with their first-half-cost $g(s)+h(s)$
 - Pop out the state with the least $g(s)+h(s)$ first.
- Known as “A” search
- How’s this idea?



- Works for this example

A search still not quite right



- A search is not optimal.

Third attempt: A* search

- Same as A search, but the heuristic function $h()$ has to satisfy $h(s) \leq h^*(s)$, where $h^*(s)$ is the true cost from node s to the goal.
- Such heuristic function $h()$ is called **admissible**.
 - An admissible heuristic never over-estimates



It is always
optimistic

- A search with admissible $h()$ is called **A* search**.

Admissible heuristic functions h

- 8-puzzle example

Example State

1		5
2	6	3
7	4	8

Goal State

1	2	3
4	5	6
7	8	

- Which of the following are admissible heuristics?
 - $h(n)$ =number of tiles in wrong position
 - $h(n)=0$
 - $h(n)=1$
 - $h(n)$ =sum of Manhattan distance between each tile and its goal location

Admissible heuristic functions h

- In general, which of the following are admissible heuristics? $h^*(n)$ is the true optimal cost from n to goal.

- $h(n) = h^*(n)$

- $h(n) = \max(2, h^*(n))$

- $h(n) = \min(2, h^*(n))$

- $h(n) = h^*(n) - 2$

- $h(n) = \sqrt{h^*(n)}$

Heuristics for Admissible heuristics

- How to construct heuristic functions?

Example State

1		5
2	6	3
7	4	8

Goal State

1	2	3
4	5	6
7	8	

- Often by relaxing the constraints
 - $h(n)$ =number of tiles in wrong position
Allow tiles to fly to their destination in one step
 - $h(n)$ =sum of Manhattan distance between each tile and its goal location
Allow tiles to move on top of other tiles

“my heuristic is better than yours”

- A heuristic function h_2 **dominates** h_1 if for all s
 $h_1(s) \leq h_2(s) \leq h^*(s)$
- We prefer heuristic functions as close to h^* as possible, but not over h^* .

But

- Good heuristic function might need complex computation
- Time may be better spent, if we use a faster, simpler heuristic function and expand more nodes