

Heuristic Search

CPSC 322 Lecture 6

January 16, 2006

Textbook §2.5

Lecture Overview

Recap

Heuristic Search

Constructing Heuristics

Best-First Search

Depth-first Search

- ▶ **Depth-first search** treats the frontier as a stack
 - ▶ It always selects one of the last elements added to the frontier.

- ▶ **Complete** when the graph has no cycles and is finite
- ▶ **Time complexity** is $O(b^m)$
- ▶ **Space complexity** is $O(bm)$

Breadth-first Search

- ▶ **Breadth-first search** treats the frontier as a queue
 - ▶ it always selects one of the earliest elements added to the frontier.

- ▶ **Complete** when the graph has no cycles and is finite
- ▶ **Time complexity** is $O(b^m)$
- ▶ **Space complexity** is $O(b^m)$

Search with Costs

- ▶ Sometimes there are **costs** associated with arcs.
 - ▶ The cost of a path is the sum of the costs of its arcs.

$$\text{cost}(\langle n_0, \dots, n_k \rangle) = \sum_{i=1}^k |\langle n_{i-1}, n_i \rangle|$$

- ▶ An **optimal** search algorithm always finds the solution that **minimizes cost**

Lowest-Cost-First Search

- ▶ At each stage, lowest-cost-first search selects a path on the frontier with **lowest cost**.
 - ▶ The frontier is a priority queue ordered by path cost.

- ▶ **Complete** when the graph has strictly positive arc costs
- ▶ **Time complexity** is $O(b^m)$
- ▶ **Space complexity** is $O(b^m)$
- ▶ **Optimal** when the graph has non-negative arc costs

Past knowledge and search

- ▶ Some people believe that they are good at solving hard problems without search
 - ▶ However, consider e.g., public key encryption codes (or combination locks): the search problem is clear, but people can't solve it
 - ▶ When people do perform well on hard problems, it is usually because they have **useful knowledge** about the structure of the problem domain
- ▶ Computers can also improve their performance when given this sort of knowledge
 - ▶ in search, they can estimate the distance from a given node to the goal through a **search heuristic**
 - ▶ in this way, they can take the goal into account when selecting path

Heuristic Search

- ▶ $h(n)$ is an estimate of the cost of the shortest path from node n to a goal node.
 - ▶ h can be extended to paths: $h(\langle n_0, \dots, n_k \rangle) = h(n_k)$
- ▶ $h(n)$ uses only readily obtainable information (that is easy to compute) about a node.
- ▶ **Admissible heuristic:** $h(n)$ is an underestimate if there is no path from n to a goal that has path length less than $h(n)$.

Example Heuristic Functions

- ▶ If the nodes are points on a Euclidean plane and the cost is the distance, we can use the straight-line distance from n to the closest goal as the value of $h(n)$.
 - ▶ this makes sense if there are obstacles, or for other reasons not all adjacent nodes share an arc
- ▶ Likewise, if nodes are cells in a grid and the cost is the number of steps, we can use “Manhattan distance”
 - ▶ this is also known as the L_1 distance; Euclidean distance is L_2 distance
- ▶ In the 8-puzzle, we can use the number of moves between each tile's current position and its position in the solution

How to Construct a Heuristic

- ▶ Overall, a cost-minimizing search problem is a constrained optimization problem
 - ▶ e.g., find a path from A to B which minimizes distance traveled, subject to the constraint that the robot can't move through walls
- ▶ A **relaxed version of the problem** is a version of the problem where one or more constraints have been dropped
 - ▶ e.g., find a path from A to B which minimizes distance traveled, *allowing* the agent to move through walls
 - ▶ A relaxed version of a minimization problem will always return a value which is weakly smaller than the original value: thus, it's an admissible heuristic

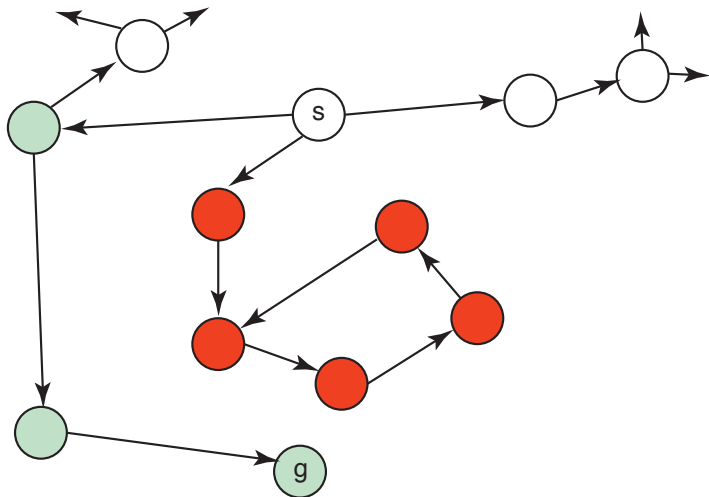
How to Construct a Heuristic

- ▶ It's usually possible to identify constraints which, when dropped, make the problem extremely easy to solve
 - ▶ this is important because heuristics are not useful if they're as hard to solve as the original problem!
- ▶ Another trick for constructing heuristics: if $h_1(n)$ is an admissible heuristic, and $h_2(n)$ is also an admissible heuristic, then $\min(h_1(n), h_2(n))$ is also admissible.

Best-First Search

- ▶ **Idea:** select the path whose end is closest to a goal according to the heuristic function.
- ▶ Best-First search selects a path on the frontier with minimal h -value.
- ▶ It treats the frontier as a priority queue ordered by h .
- ▶ This is a **greedy** approach: it always takes the path which appears locally best

Illustrative Graph — Best-First Search



Complexity of Best-First Search

- ▶ **Complete:** no: a heuristic of zero for an arc that returns to the same state can be followed forever.
- ▶ **Time complexity** is $O(b^m)$
- ▶ **Space complexity** is $O(b^m)$
- ▶ **Optimal:** no (why not?)