

CS 331: Artificial Intelligence

Informed Search

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Informed Search

- How can we make search smarter?
- Use problem-specific knowledge beyond the definition of the problem itself
- Specifically, incorporate knowledge of how good a non-goal state is

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Best-First Search

- Node selected for expansion based on an evaluation function $f(n)$. i.e. expand the node that *appears* to be the best
- Node with lowest evaluation is selected for expansion
- Uses a priority queue
- We'll talk about Greedy Best-First Search and A* Search

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Heuristic Function

- $h(n)$ = estimated cost of the cheapest path from node n to a goal node
- $h(\text{goal node}) = 0$
- Contains additional knowledge of the problem

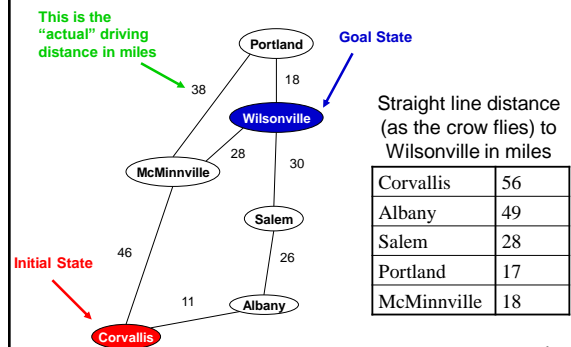
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Greedy Best-First Search

- Expands the node that is closest to the goal
- $f(n) = h(n)$

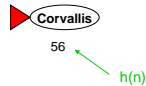
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Greedy Best-First Search Example



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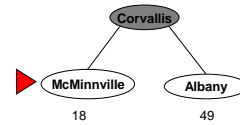
Greedy Best-First Search Example



Corvallis	56
Albany	49
Salem	28
Portland	17
McMinnville	18

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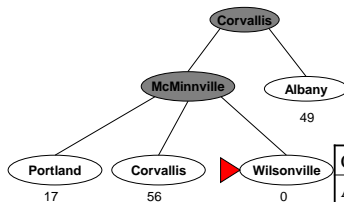
Greedy Best-First Search Example



Corvallis	56
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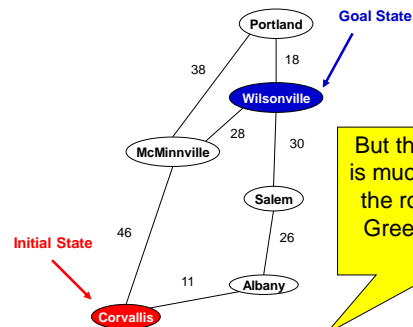
Greedy Best-First Search Example



Corvallis	56
Albany	49
Salem	28
Portland	17
McMinnville	18

Corvallis → McMinnville → Wilsonville = 74 miles

Greedy Best-First Search Example



Corvallis → Albany → Salem → Wilsonville = 67 miles

Evaluating Greedy Best-First Search

Complete?	No (could start down an infinite path)
Optimal?	No
Time Complexity	$O(b^m)$
Space Complexity	$O(b^m)$

Greedy Best-First search results in lots of dead ends which leads to unnecessary nodes being expanded

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A* Search

- A much better alternative to greedy best-first search
- Evaluation function for A* is:

$$f(n) = g(n) + h(n)$$
 where $g(n)$ = path cost from the start node to n
- If $h(n)$ satisfies certain conditions, A* search is optimal and complete!

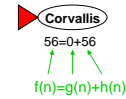
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Admissible Heuristics

- A* is optimal if $h(n)$ is an admissible heuristic
- An admissible heuristic is one that never overestimates the cost to reach the goal
- Admissible heuristic = optimistic
- Straight line distance was an admissible heuristic

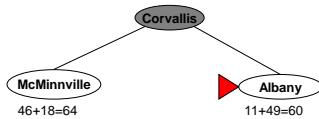
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A* Search Example



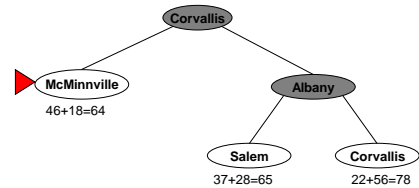
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A* Search Example



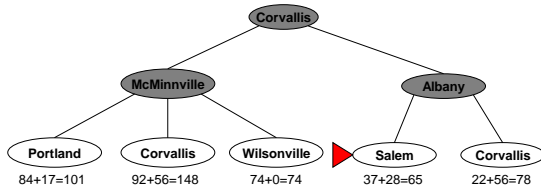
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A* Search Example



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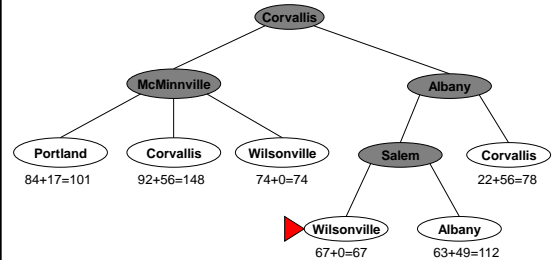
A* Search Example



Note: Don't stop when you put a goal state on the priority queue (otherwise you get a suboptimal solution)

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A* Search Example



Proper termination: Stop when you pop a goal state from the priority queue

Proof that A* using TREE-SEARCH is optimal if h(n) is admissible

- Suppose a suboptimal goal node G_2 appears on the fringe and let the cost of the optimal solution be C^*
- Because G_2 is suboptimal:
 $f(G_2) = g(G_2) + h(G_2) = g(G_2) > C^*$

$h(G_2) = 0$ because it is a goal node
- Now consider a fringe node n on an optimal solution path
- If $h(n)$ is admissible then:
 $f(n) = g(n) + h(n) \leq C^*$
- We have shown that $f(n) \leq C^* < f(G_2)$, so G_2 will not get expanded and A* must return an optimal solution

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What about search graphs (more than one path to a node)?

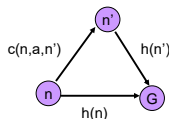
- What if we expand a state we've already seen?
- Suppose we use the GRAPH-SEARCH solution and not expand repeated nodes
- Could discard the optimal path if it's not the first one generated
- One simple solution: ensure optimal path to any repeated state is always the first one followed (like in Uniform-cost search)
- Requires an extra requirement on $h(n)$ called consistency (or monotonicity)

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Consistency

- A heuristic is **consistent** if, for every node n and every successor n' of n generated by any action a :

$$h(n) \leq \underbrace{c(n,a,n')}_{\text{Step cost of going from } n \text{ to } n' \text{ by doing action } a} + h(n')$$
- A form of the triangle inequality – each side of the triangle cannot be longer than the sum of the two sides



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Consistency

- Every consistent heuristic is also admissible
- A* using GRAPH-SEARCH is optimal if $h(n)$ is consistent
- Most admissible heuristics are also consistent

Consistency

- If $f(n)$ is consistent, then the values of $f(n)$ along any path are nondecreasing
- Proof:
 Suppose n' is a successor of n .
 Then $g(n') = g(n) + c(n,a,n')$ for some a , and we have

$$\begin{aligned} f(n') &= g(n') + h(n') \\ &= g(n) + c(n,a,n') + h(n') \\ &\geq g(n) + h(n) \quad \leftarrow \text{From defn of consistency: } c(n,a,n') + h(n') \geq h(n) \\ &= f(n) \end{aligned}$$
- Thus, the sequence of nodes expanded by A* is in nondecreasing order of $f(n)$
- First goal selected for expansion must be an optimal solution since all later nodes will be at least as expensive

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A* is Optimally Efficient

- Among optimal algorithms that expand search paths from the root, A* is **optimally efficient** for any given heuristic function
- Optimally efficient: no other optimal algorithm is guaranteed to expand fewer nodes than A*
 - Fine print: except A* might possibly expand more nodes with $f(n) = C^*$ where C^* is the cost of the optimal path – tie-breaking issues
- Any algorithm that does not expand all nodes with $f(n) < C^*$ runs the risk of missing the optimal solution

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Evaluating A* Search

With a consistent heuristic, A* is complete, optimal and optimally efficient. Could this be the answer to our searching problems?

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Evaluating A* Search

With a consistent heuristic, A* is complete, optimal and optimally efficient. Could this be the answer to our searching problems?

The Dark Side of A*...



Time complexity is exponential (although it can be reduced significantly with a good heuristic)

The really bad news: space complexity is exponential (usually need to store all generated states). Typically runs out of space on large-scale problems.

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Summary of A* Search

Complete?	Yes if $h(n)$ is consistent, b is finite, and all step costs exceed some finite ϵ ¹
Optimal?	Yes if $h(n)$ is consistent and admissible
Time Complexity	$O(b^d)$ (In the worst case but a good heuristic can reduce this significantly)
Space Complexity	$O(b^d)$ – Needs $O(\text{number of states})$, will run out of memory for large search spaces

¹ Since $f(n)$ is nondecreasing, we must eventually hit an $f(n) = \text{cost of the path to a goal state}$

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Iterative Deepening A*

- Use iterative deepening trick to reduce memory requirements for A*
- Cutoff is the $f\text{-cost}$ ($g+h$) rather than the depth
- At each iteration, the cutoff is the smallest $f\text{-cost}$ that exceeded the cutoff in the previous iteration

Complete, Optimal but more costly than A* and can take a while to run with real-valued costs

Examples of heuristic functions

The 8-puzzle

7	2	4
5		6
8	3	1

Start State

	1	2
3	4	5
6	7	8

End State

Heuristic #1: h_1 = number of misplaced tiles eg. start state has 8 misplaced tiles. This is an admissible heuristic

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Examples of heuristic functions

The 8-puzzle

7	2	4
5		6
8	3	1

Start State

	1	2
3	4	5
6	7	8

End State

Heuristic #2: h_2 = total Manhattan distance (sum of horizontal and vertical moves, no diagonal moves). Start state is $3+1+2+2+2+3+3+2=18$ moves away from the end state. This is also an admissible heuristic.

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Which heuristic is better?

- h_2 dominates h_1 . That is, for any node n , $h_2(n) \geq h_1(n)$.
- h_2 never expands more nodes than A^* using h_1 (except possibly for some nodes with $f(n) = C^*$)
- Better to use h_2 provided it doesn't overestimate and its computation time isn't too expensive.
(Remember that h_2 is also admissible)

Proof:

Every node with $f(n) < C^*$ will surely be expanded, meaning every node with $h(n) < C^* - g(n)$ will surely be expanded

Since h_2 is at least as big as h_1 for all nodes, every node expanded with A^* using h_2 will also be expanded with A^* using h_1 . But h_1 might expand other nodes as well.

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Which heuristic is better?

Depth	# nodes expanded		
	IDS	$A^*(h_1)$	$A^*(h_2)$
2	10	6	6
4	112	13	12
6	680	20	18
8	6384	39	25
10	47127	93	39
12	3644035	227	73
14		539	113
16		1301	211
18		3056	363
20		7276	676
22		18094	1219
24		39135	1641

From Russell and Norvig Figure 4.8 (Results averaged over 100 instances of the 8-puzzle for depths 2-24).

Inventing Admissible Heuristics

- Relaxed problem: a problem with fewer restrictions on the actions
- The cost of an optimal solution to a relaxed problem is an admissible heuristic for the original problem
- If we relax the rules so that a square can move anywhere, we get heuristic h_1
- If we relax the rules to allow a square to move to any adjacent square, we get heuristic h_2

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What you should know

- Be able to run A^* by hand on a simple example
- Why it is important for a heuristic to be admissible and consistent
- Pros and cons of A^*
- How do come up with heuristics
- What it means for a heuristic function to dominate another heuristic function

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