# 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

#### **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

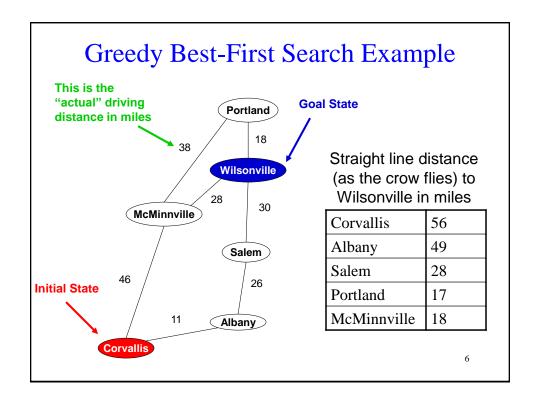
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#### **Heuristic Function**

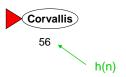
- h(n) = estimated cost of the cheapest path from node n to a goal node
- h(goal node) = 0
- Contains additional knowledge of the problem

### Greedy Best-First Search

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



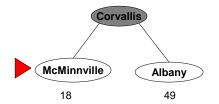
# Greedy Best-First Search Example



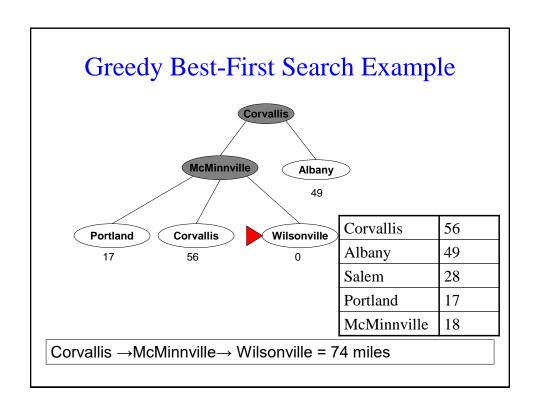
Corvallis	56
Albany	49
Salem	28
Portland	17
McMinnville	18

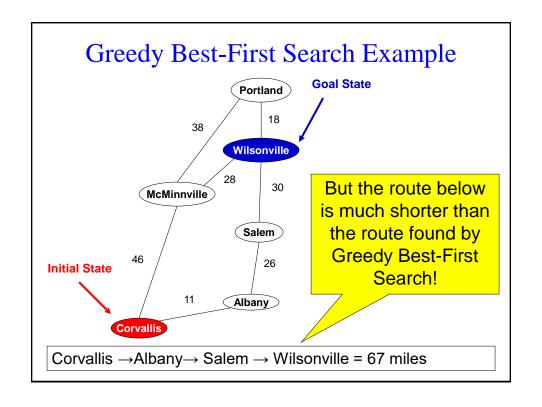
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# Greedy Best-First Search Example



Corvallis	56
Albany	49
Salem	28
Portland	17
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#### **Evaluating Greedy Best-First Search**

Complete?	No (could start down an infinite path)
Optimal?	No
Time Complexity	O(b <sup>m</sup> )
Space Complexity	O(b <sup>m</sup> )

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 bestfirst search
- Evaluation function for A\* is:

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

where g(n) = path cost from the start node to n

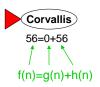
• If h(n) satisfies certain conditions, A\* search is optimal and complete!

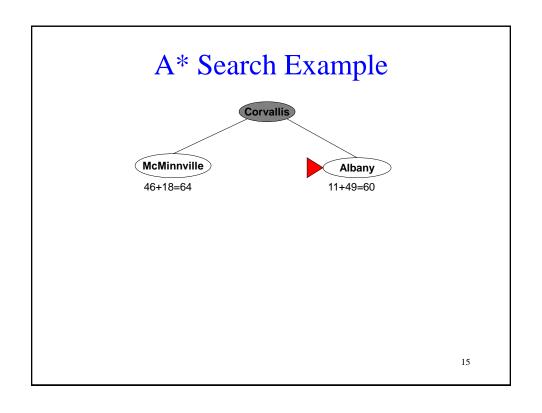
#### **Admissible Heuristics**

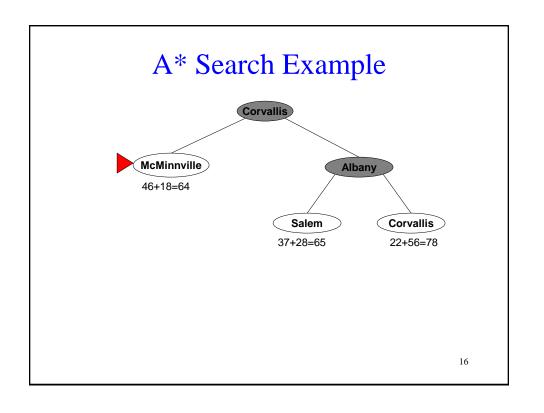
- A\* is optimal if h(n) is an admissible heuristic
- An admissible heuristic is one that never overestimates the cost to reach the goal
- Admissible heuristic = optimistic
- Straight line distance was an admissible heuristic

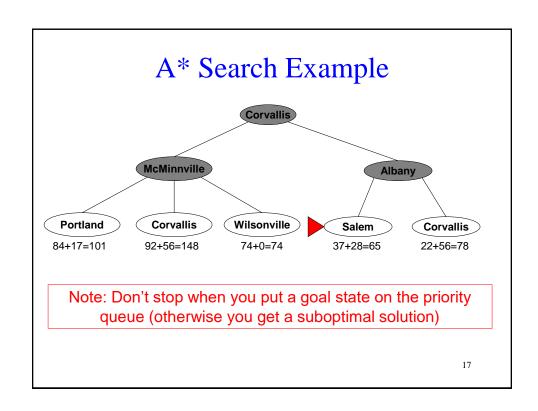
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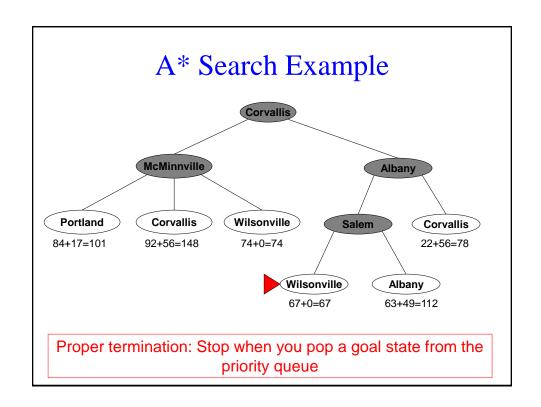
# A\* Search Example











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

- Suppose a suboptimal goal node G<sub>2</sub> appears on the fringe and let the cost of the optimal solution be C\*
- Because  $G_2$  is suboptimal:  $f(G_2) = g(G_2) + h(G_2) = g(G_2) > C^*$  is a goal node
- Now consider a fringe node n on an optimal solution path
- If h(n) is admissible then:

$$f(n) = g(n) + h(n) \le C^*$$

• We have shown that  $f(n) \le C^* < f(G_2)$ , 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)?

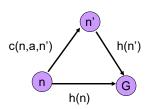
- What if we expand a state we've already seen?
- Suppose we use the GRAPH-SEARCH solution and not expand repeated nodes
- Could discard the optimal path if it's not the first one generated
- One simple solution: ensure optimal path to any repeated state is always the first one followed (like in Uniform-cost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

### Consistency

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

$$h(n) \leq \underbrace{c(n, a, n') + h(n')}_{\text{Step cost of going from n to n'}}$$
 by doing action a

• A form of the triangle inequality – each side of the triangle cannot be longer than the sum of the two sides



### 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.

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Then g(n') = g(n) + c(n,a,n') for some a, and we have f(n') = g(n') + h(n')= g(n) + c(n,a,n') + h(n')\geq g(n) + h(n) \qquad \qquad \qquad \text{From defn of consistency:}= f(n) \qquad \qquad c(n,a,n') + h(n') \geq h(n)
```

- 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

### 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.

# Summary of A\* Search

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

 $<sup>^{1}</sup>$  Since f(n) is nondecreasing, we must eventually hit an f(n) = cost of the path to a goal state

### Iterative Deepening A\*

- Use iterative deepening trick to reduce memory requirements for A\*
- Cutoff is the f-cost (g+h) rather than the depth
- At each iteration, the cutoff is the smallest f-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+3+3+2=18 moves away from the end state. This is also an admissible heuristic.

#### Which heuristic is better?

- $h_2$  dominates  $h_1$ . That is, for any node n,  $h_2(n) \ge h_1(n)$ .
- $h_2$  never expands more nodes than  $A^*$  using  $h_1$  (except possibly for some nodes with  $f(n) = C^*$ )
- Better to use h<sub>2</sub> provided it doesn't overestimate and its computation time isn't too expensive.

(Remember that h<sub>2</sub> 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?

	# nodes expanded		
Depth	IDS	A*(h <sub>1</sub> )	A*(h <sub>2</sub> )
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<sub>1</sub>
- If we relax the rules to allow a square to move to any adjacent square, we get heuristic h<sub>2</sub>

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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 if means for a heuristic function to dominate another heuristic function