# Chapter 3 Informed Searching

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#### Acknowledgement:

This lecture is based on (but not limited to) chapter 4 in "S. Russell and P. Norvig: *Artificial Intelligence: A Modern Approach".* 

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#### **Discussion and Motivation**



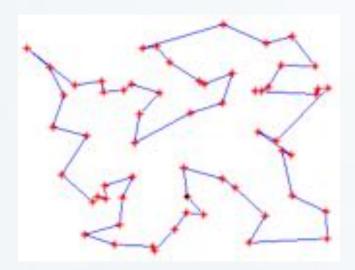
How to determine the minimum number of coins to give while making change?

→ The coin of the highest value first?

#### **Discussion and Motivation**

#### **Travel Salesperson Problem**

Given a list of cities and their pair wise distances, the task is to find a shortest possible tour that visits each city exactly once.



- Any idea how to improve this type of search?
- What type of information we may use to improve our search?
- Do you think this idea is useful: (At each stage visit the unvisited city nearest to the current city)?

#### **Best-first search**

Idea: use an evaluation function f(n) for each node

- family of search methods with various evaluation functions (estimate of "desirability")
- usually gives an estimate of the distance to the goal
- often referred to as heuristics in this context
- → Expand most desirable unexpanded node.
- → A heuristic function ranks alternatives at each branching step based on the available information (heuristically) in order to make a decision about which branch to follow during a search.

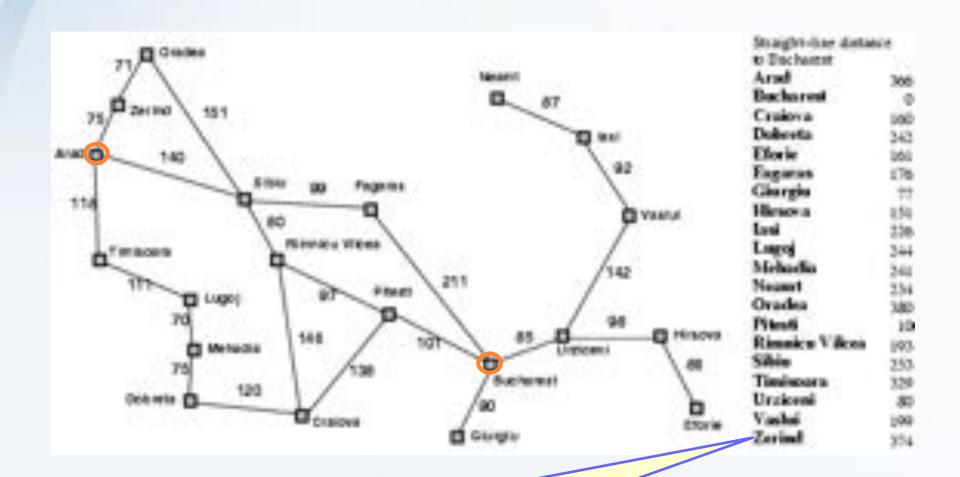
#### **Implementation**:

Order the nodes in fringe in decreasing order of desirability.

#### Special cases:

greedy best-first search
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### Romania with step costs in km



Suppose we can have this info (SLD)

STUDENTS UBLOWN can we use it to improve our search?

# **Greedy best-first search**

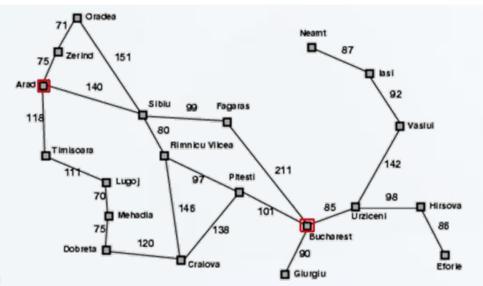
- Greedy best-first search expands the node that appears to be closest to goal.
- Estimate of cost from n to goal, e.g., h<sub>SLD</sub>(n) = straight-line distance from n to Bucharest.

Utilizes a heuristic function as evaluation function

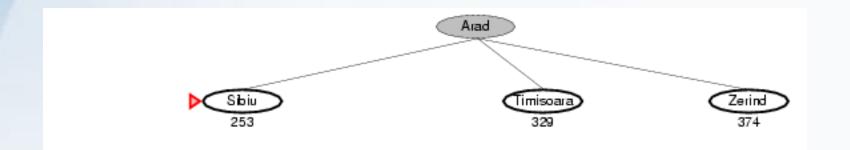
- -f(n) = h(n) = estimated cost from the current node to a goal.
- Heuristic functions are problem-specific.
- Often straight-line distance for route-finding and similar problems.
- Often better than depth-first, although worst-time complexities are equal or worse (space).

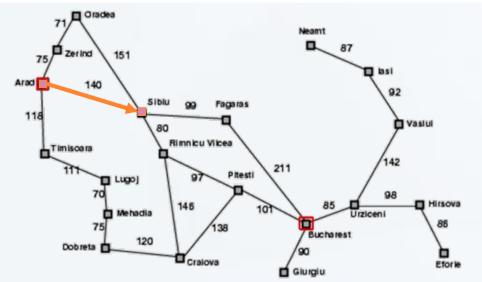
Example from [1]



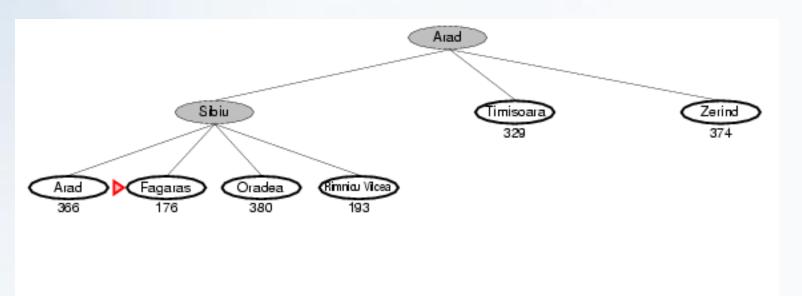


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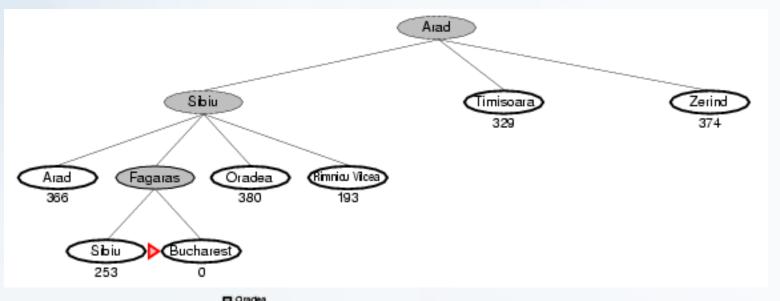


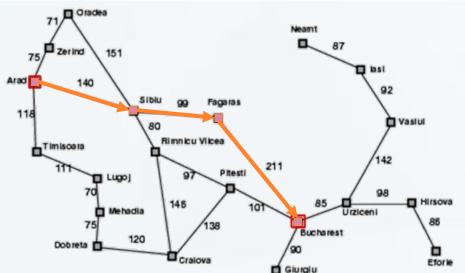
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# **Greedy best-first search**

```
function Greedy-Best-First-Search(initialState, goalTest)
     returns Success or Failure: /* Cost f(n) = h(n) */
     frontier = Heap.new(initialState)
     explored = Set.new()
     while not frontier.isEmpty():
          state = frontier.deleteMin()
          explored.add(state)
          if goalTest(state):
               return Success(state)
          for neighbor in state.neighbors():
               if neighbor not in frontier \cup explored:
                     frontier.insert(neighbor)
               else if neighbor in frontier:
                    frontier.decreaseKey(neighbor)
     return FAILURE
```

# Properties of greedy best-first search

```
Complete: No – can get stuck in loops (e.g., lasi → Neamt → lasi → Neamt → ....)
```

<u>Time</u>:  $O(b^m)$ , but a good heuristic can give significant improvement

Space:  $O(b^m)$  -- keeps all nodes in memory

Optimal: No

b branching factor

maximum depth of the search tree

#### **Discussion**

Do you think  $h_{SLD}(n)$  is admissible? Would you use  $h_{SLD}(n)$  in Palestine? How? Why?

Did you find the Greedy idea useful?

→ Ideas to improve it?

#### A\* search

Idea: avoid expanding paths that are already expensive.

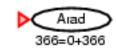
Evaluation function = path cost + estimated cost to the goal

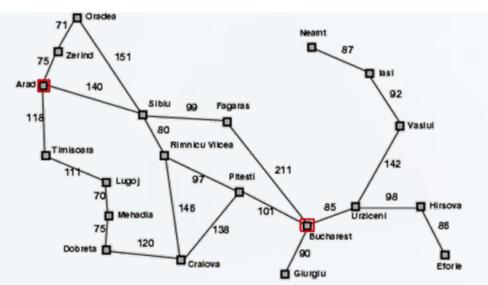
$$f(n) = g(n) + h(n)$$

- $-g(n) = \cos t$  so far to reach n
- -h(n) = estimated cost from n to goal
- -f(n) = estimated total cost of path through n to goal

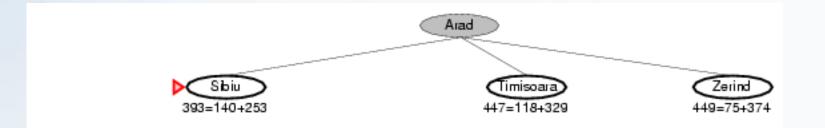
Combines greedy and uniform-cost search to find the (estimated) cheapest path through the current node

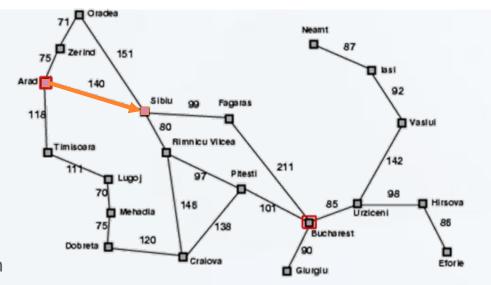
- Heuristics must be admissible
  - Never overestimate the cost to reach the goal
- Very good search method, but with complexity problems



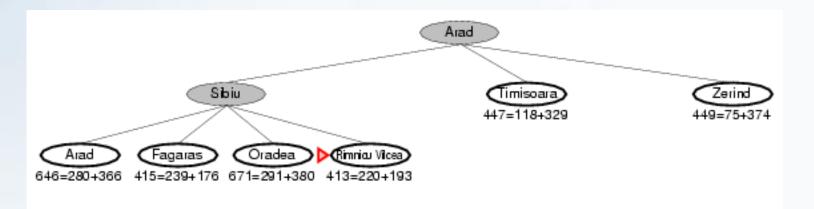


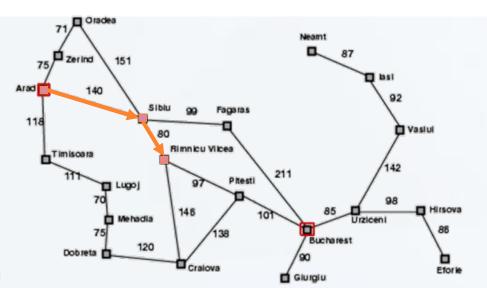
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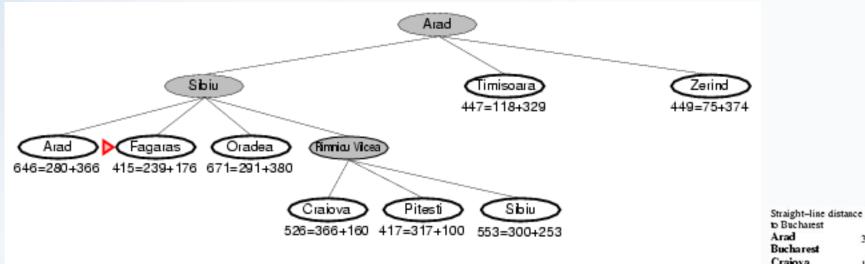


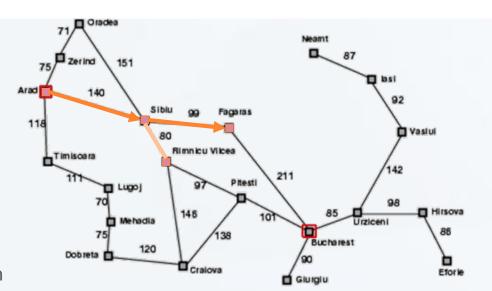
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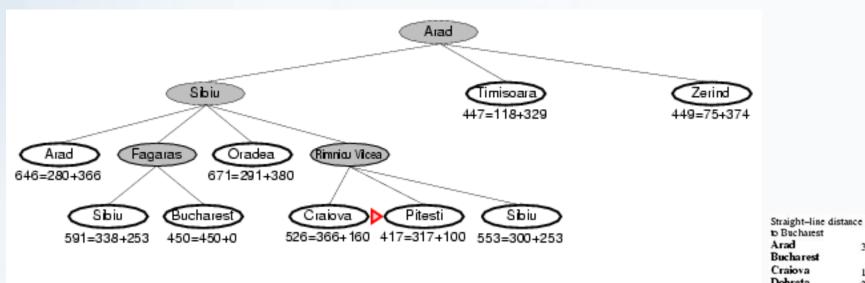


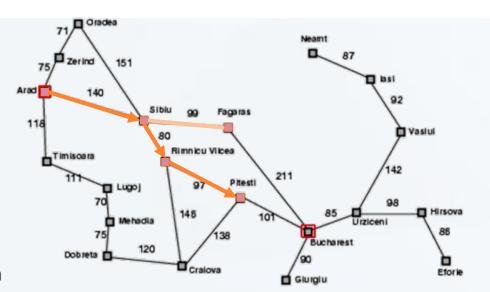
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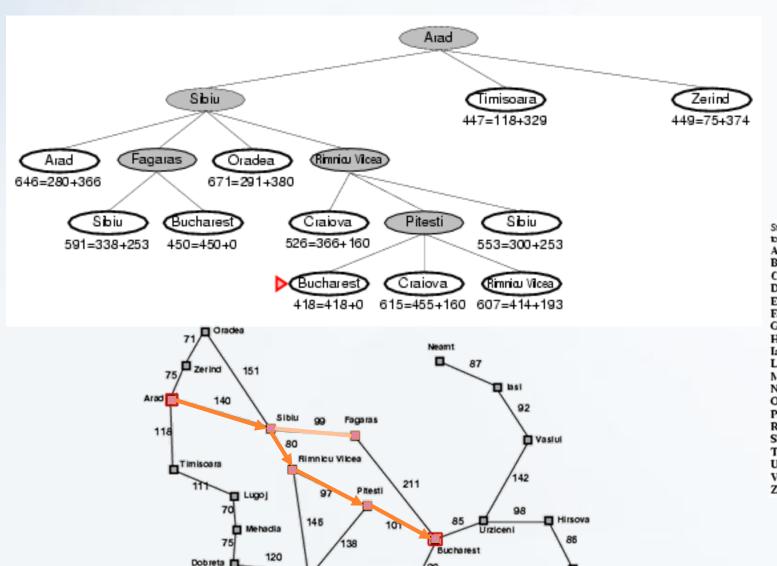


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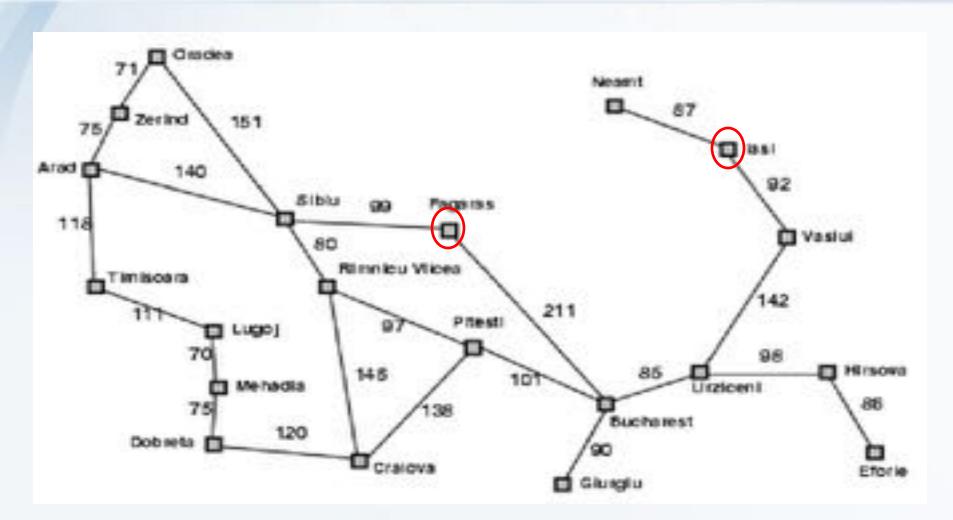
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## A\* Algorithm

```
function A-STAR-SEARCH(initialState, goalTest)
     returns Success or Failure: /* Cost f(n) = g(n) + h(n) */
     frontier = Heap.new(initialState)
     explored = Set.new()
     while not frontier.isEmpty():
          state = frontier.deleteMin()
          explored.add(state)
          if goalTest(state):
               return Success(state)
          for neighbor in state.neighbors():
               if neighbor not in frontier \cup explored:
                     frontier.insert(neighbor)
               else if neighbor in frontier:
                     frontier.decreaseKey(neighbor)
```

return FAILURE

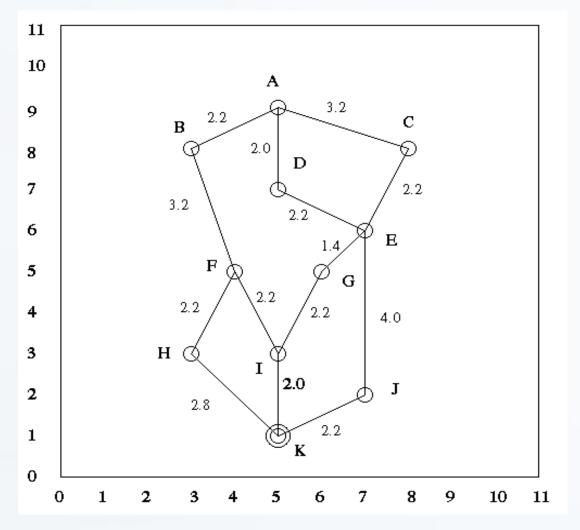
#### A\* Exercise



**How will A\* get from lasi to Fagaras?** 

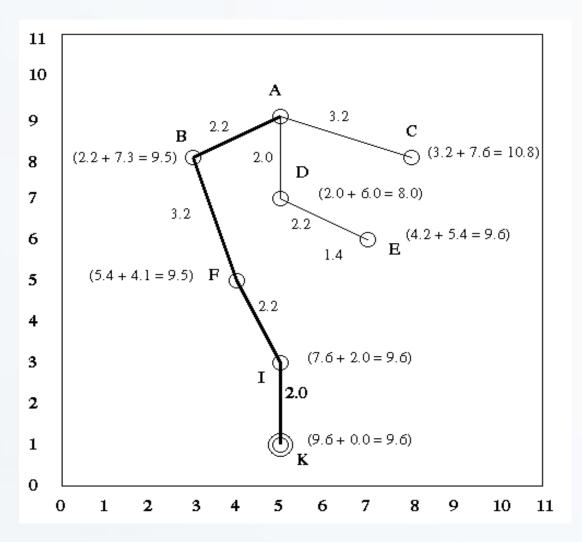
### **A\* Exercise**

Coordinates (5,9)	SL Distance 8.0
	8.0
(2.0)	
(3,8)	7.3
(8,8)	7.6
(5,7)	6.0
(7,6)	5.4
(4,5)	4.1
(6,5)	4.1
(3,3)	2.8
(5,3)	2.0
(7,2)	2.2
(5,1)	0.0
	(8,8) (5,7) (7,6) (4,5) (6,5) (3,3) (5,3) (7,2)



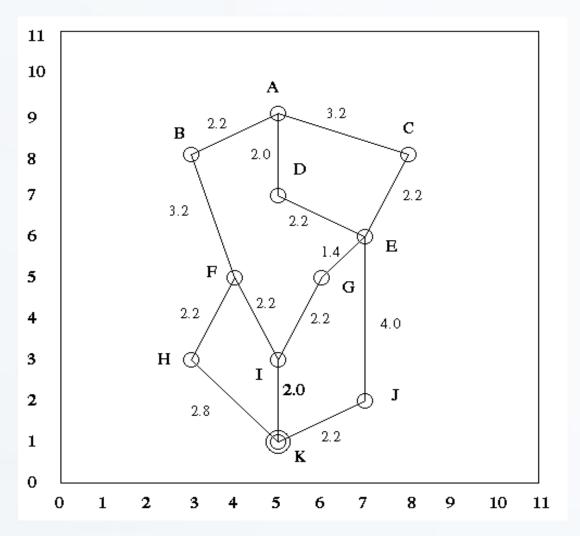
#### **Solution to A\* Exercise**

		a	
<u>Node</u>	<b>Coordinates</b>	<b>SL Distance</b>	
Α	(5,9)	8.0	
В	(3,8)	7.3	
C	(8,8)	7.6	
D	(5,7)	6.0	
E	(7,6)	5.4	
F	(4,5)	4.1	
G	(6,5)	4.1	
Н	(3,3)	2.8	
1	(5,3)	2.0	
J	(7,2)	2.2	
K	(5,1)	0.0	
	( , ,		



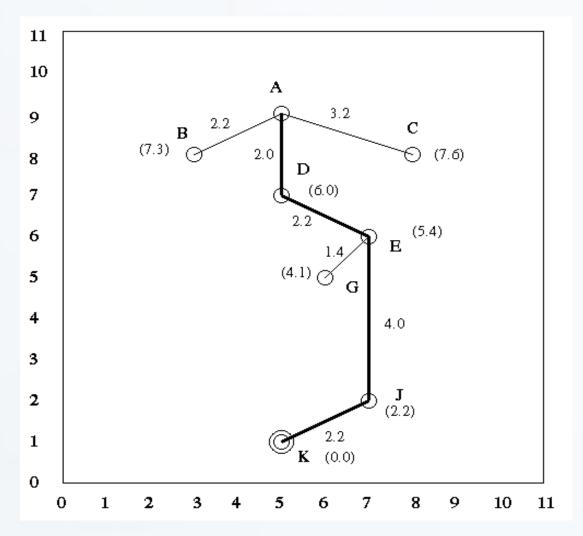
# **Greedy Best-First Exercise**

<u>Node</u>	<b>Coordinates</b>	nates <u>Distance</u>		
Α	(5,9)	8.0		
В	(3,8)	7.3		
C	(8,8)	7.6		
D	(5,7)	6.0		
E	(7,6)	5.4		
F	(4,5) 4.1	4.1		
G	(6,5)	4.1 2.8		
Н	(3,3)			
1	(5,3)	2.0		
J	(7,2)	2.2		
K	(5,1)	0.0		
	, ,			



## **Solution to Greedy Best-First Exercise**

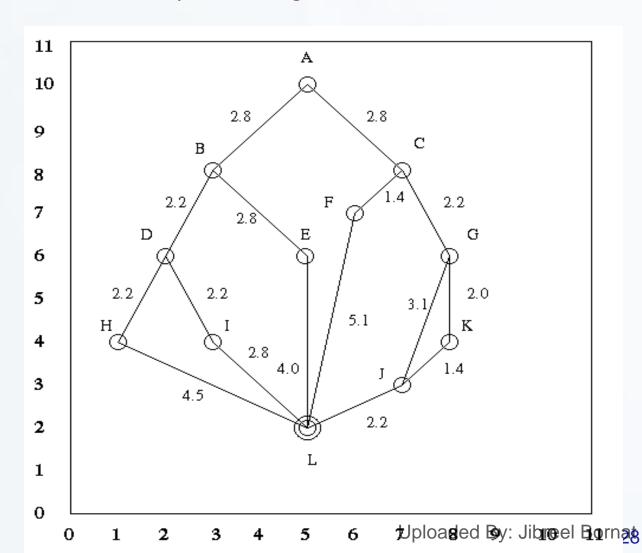
<u>Node</u>	<u>Coordinates</u>	<u>Distance</u>	
Α	(5,9)	8.0	
В	(3,8)	7.3	
C	(8,8)	7.6	
D	(5,7)	6.0	
E	(7,6)	5.4	
F	(4,5)	4.1	
G	(6,5)	4.1	
H	(3,3)	2.8	
1	(5,3)	2.0	
J	(7,2)	2.2	
K	(5,1)	0.0	



#### **Another Exercise**

#### Do 1) A\* Search and 2) Greedy Best-Fit Search

Nod	le C	<u>g(n)</u>	<u>h(n)</u>
Α	(5,10)	0.0	8.0
В	(3,8)	2.8	6.3
C	(7,8)	2.8	6.3
D	(2,6)	5.0	5.0
Ε	(5,6)	5.6	4.0
F	(6,7)	4.2	5.1
G	(8,6)	5.0	5.0
Н	(1,4)	7.2	4.5
1	(3,4)	7.2	2.8
J	(7,3)	8.1	2.2
K	(8,4)	7.0	3.6
L	(5,2)	9.6	0.0



#### **Admissible Heuristics**

A heuristic h(n) is admissible if for every node n,  $h(n) \le h^*(n)$ , where  $h^*(n)$  is the true cost to reach the goal state from n.

An admissible heuristic never overestimates the cost to reach the goal, i.e., it is optimistic.

The heuristic function  $h_{SLD}(n)$  is admissible because it never overestimates the actual road distance)

Theorem-1: If h(n) is admissible,  $A^*$  using TREE-SEARCH is optimal.

Recall that f(n) = g(n) + h(n)

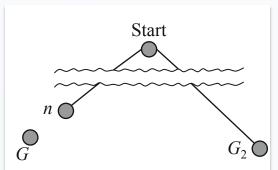
Now, suppose some suboptimal goal  $G_2$  has been generated and is in the fringe. Let n be an unexpanded node in the fringe such that n is on a shortest path to an optimal goal G.

#### We want to prove:

$$f(n) < f(G_2)$$
  
(then A\* will prefer  $n$  over  $G_2$ )

$$f(G_2) = g(G_2)$$
 since  $h(G_2) = 0$   
 $g(G_2) > g(G)$  since  $G_2$  is suboptimal  
 $f(G) = g(G)$  since  $h(G) = 0$   
Then  $f(G_2) > f(G)$  from above  
 $h(n) \le h^*(n)$  since  $h$  is admissible  
 $g(n) + h(n) \le g(n) + h^*(n)$   
Then  $f(n) \le f(G)$ 

Thus, A\* will never select G<sub>2</sub> for expansion



# **Optimality of A\* (proof)**

Recall that f(n) = g(n) + h(n)

Now, suppose some suboptimal goal  $G_2$  has been generated and is in the fringe. Let n be an unexpanded node in the fringe such that n is on a shortest path to an optimal goal G.

#### We want to prove:

$$f(n) < f(G_2)$$
  
(then A\* will prefer  $n$  over  $G_2$ )

#### In other words:

$$f(G_2) = g(G_2) + h(G_2) = g(G_2) > C^*$$
,  
since  $G_2$  is a goal on a non-optimal path (C\* is the optimal cost)

$$f(n) = g(n) + h(n) \le C^*$$
, since h is admissible

$$f(n) \le C^* < f(G_2)$$
, so  $G_2$  will never be expanded

→ A\* will not expand goals on sub-optimal paths

Start

# **Properties of A\***

Complete: Yes
 unless there are infinitely many nodes with f ≤ f(G)

Time: Exponential
 because all nodes such that f(n) ≤ C\* are expanded!

 Space: Keeps all nodes in memory fringe is exponentially large

Optimal: Yes

# **Memory Bounded Heuristic Search**

How can we solve the memory problem for A\* search?

Idea: Try something like iterative deeping search, but the cutoff is f-cost (g+h) at each iteration, rather than depth first.

Two types of memory bounded heuristic searches:

- Recursive BFS
- ➤ SMA\*

# Recursive Best First Search (RBFS)

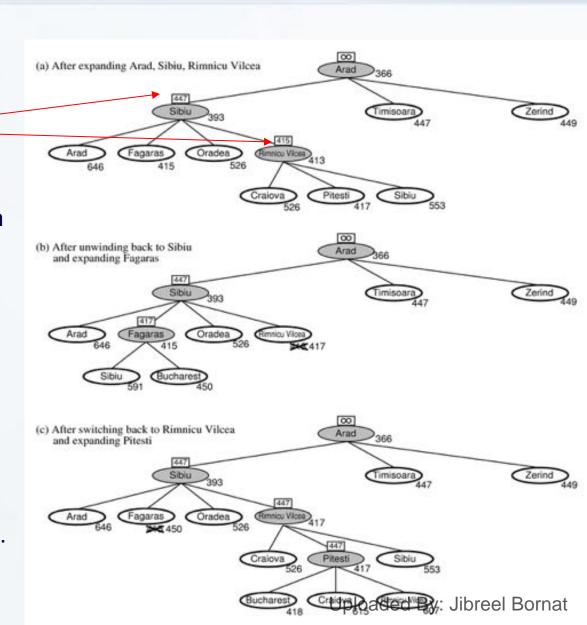
best alternative over fringe nodes, which are not children: do I want to back up?

RBFS changes its mind very often in practice.

This is because the f=g+h become more accurate (less optimistic) as we approach the goal. Hence, higher level nodes have smaller f-values and will be explored first.

**Problem?** If we have more memory we cannot make use of it.

Ay idea to improve this?



# Simple Memory Bounded A\* (SMA\*)

- This is like A\*, but when memory is full we delete the worst node (largest f-value).
- Like RBFS, we remember the best descendent in the branch we delete.
- If there is a tie (equal f-values) we first delete the oldest nodes first.
- SMA\* finds the optimal reachable solution given the memory constraint.
- But time can still be exponential.

# **SMA\*** pseudocode

```
function SMA*(problem) returns a solution sequence
  inputs: problem, a problem
  static: Queue, a queue of nodes ordered by f-cost
  Queue ← MAKE-QUEUE({MAKE-NODE(INITIAL-STATE[problem])})
  loop do
      if Queue is empty then return failure
      n \leftarrow deepest least-f-cost node in Queue
      if GOAL-TEST(n) then return success
      s \leftarrow \text{NEXT-SUCCESSOR}(n)
      if s is not a goal and is at maximum depth then
        f(s) \leftarrow \infty
      else
        f(s) \leftarrow MAX(f(n),g(s)+h(s))
      if all of n's successors have been generated then
        update n's f-cost and those of its ancestors if necessary
      if SUCCESSORS(n) all in memory then remove n from Queue
      if memory is full then
        delete shallowest, highest-f-cost node in Queue
        remove it from its parent's successor list
        insert its parent on Queue if necessary
      insert s in Queue
  end
```

# Simple Memory-bounded A\* (SMA\*)

SMA\* is a shortest path algorithm based on the A\* algorithm.

The advantage of SMA\* is that it uses a bounded memory, while the A\* algorithm might need exponential memory.

All other characteristics of SMA\* are inherited from A\*.

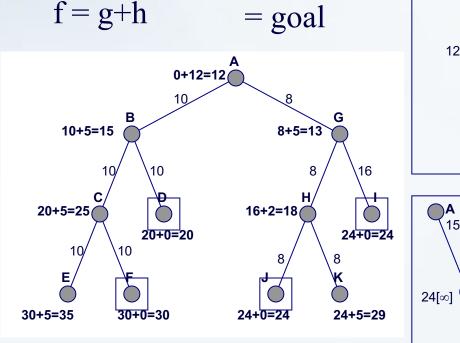
#### **How it works:**

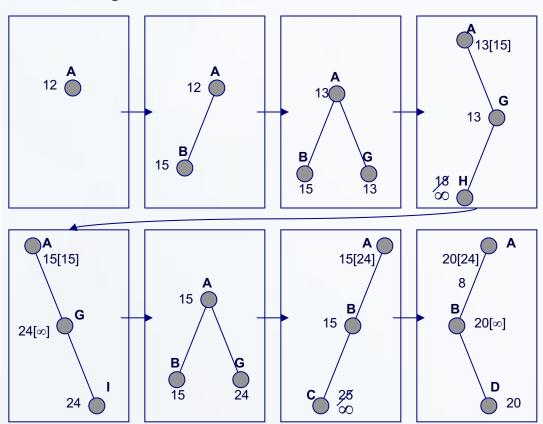
- Like A\*, it expands the best leaf until memory is full.
- Drops the worst leaf node- the one with the highest f-value.
- Like RBFS, SMA\* then backs up the value of the forgotten node to its parent.

# Simple Memory-bounded A\* (SMA\*) (Example with 3-node memory)

Search space

Progress of SMA\*. Each node is labeled with its current f-cost. Values in parentheses show the value of the best forgotten descendant.





 $\infty$  is given to nodes that the path up to it uses all available memory. Can tell when best solution found within memory constraint is optimal or not.

# The Algorithm proceeds as follows [3]

- At each stage, one successor is added to the deepest lowest-f-cost node that has some successors not currently in the tree. The left child B is added to the root A.
- 2. Now f(A) is still 12, so we add the right child G(f = 13). Now that we have seen all the children of A, we can update its f-cost to the minimum of its children, that is, 13. The memory is now full.
- 3. G is now designated for expansion, but we must first drop a node to make room. We drop the shallowest highest-f-cost leaf, that is, B. When we have done this, we note that A's best forgotten descendant has f = 15, as shown in parentheses. We then add H, with f(H) = 18. Unfortunately, H is not a goal node, but the path to H uses up all the available memory. Hence, there is no way to find a solution through H, so we set  $f(H) = \infty$ .
- 4. G is expanded again. We drop H, and add I, with f(I) = 24. Now we have seen both successors of G, with values of  $\infty$  and 24, so f(G) becomes 24. f(A) becomes 15, the minimum of 15 (forgotten successor value) and 24. Notice that I is a goal node, but it might not be the best solution because A's f-cost is only 15.
- 5. A is once again the most promising node, so B is generated for the second time. We have found that the path through G was not so great after all.
- 6. C, the first successor of B, is a nongoal node at the maximum depth, so  $f(C) = \infty$ .
- 7. To look at the second successor, D, we first drop C. Then f(D) = 20, and this value is inherited by B and A.
- Now the deepest, lowest-f-cost node is D. D is therefore selected, and because it is a goal node, the search terminates.

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## **SMA\* Properties** [2]

- It works with a heuristic, just as A\*
- It is complete if the allowed memory is high enough to store the shallowest solution.
- It is optimal if the allowed memory is high enough to store the shallowest optimal solution, otherwise it will return the best solution that fits in the allowed memory.
- It avoids repeated states as long as the memory bound allows it
- It will use all memory available.
- Enlarging the memory bound of the algorithm will only speed up the calculation.
- When enough memory is available to contain the entire search tree, then calculation has an optimal speed

#### **Admissible Heuristics**

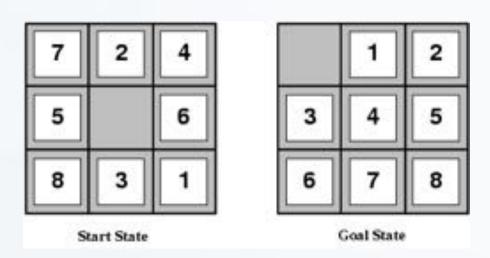
How can you invent a good admissible heuristic function? E.g., for the 8-puzzle

#### **Admissible heuristics**

E.g., for the 8-puzzle:

 $h_1(n)$  = number of misplaced tiles

 $h_2(n)$  = total Manhattan distance (i.e., no. of squares from desired location of each tile)



$$h_1(S) = ?$$

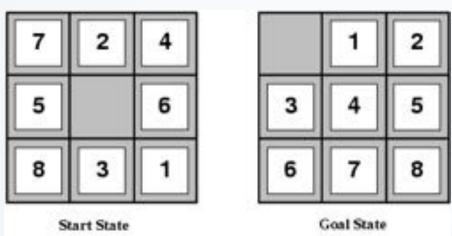
$$h_2(S) = ?$$

#### **Admissible heuristics**

E.g., for the 8-puzzle:

 $h_1(n)$  = number of misplaced tiles

 $h_2(n)$  = total Manhattan distance (i.e., no. of squares from desired location of each tile)



$$h_1(S) = 8$$
  
 $h_2(S) = 3+1+2+2+3+3+2 = 18$ 

#### **Dominance**

If  $h_2(n) \ge h_1(n)$  for all n, and both are admissible.

then  $h_2$  dominates  $h_1$ 

 $h_2$  is better for search: it is guaranteed to expand less nodes.

Typical search costs (average number of nodes expanded):

IDS = 3,644,035 nodes  

$$A^*(h_1) = 227$$
 nodes  
 $A^*(h_2) = 73$  nodes  
 $d=24$  IDS = too many nodes  
 $A^*(h_1) = 39,135$  nodes  
 $A^*(h_2) = 1,641$  nodes

What to do If we have  $h_1...h_m$ , but none dominates the other?

$$\rightarrow h(n) = \max\{h_1(n), \dots h_m(n)\}$$

#### **Relaxed Problems**

A problem with fewer restrictions on the actions is called a relaxed problem.

The cost of an optimal solution to a relaxed problem is an admissible heuristic for the original problem.

If the rules of the 8-puzzle are relaxed so that a tile can move anywhere, then  $h_1(n)$  gives the shortest solution.

If the rules are relaxed so that a tile can move to any near square, then  $h_2(n)$  gives the shortest solution.

#### **Admissible Heuristics**

#### How can you invent a good admissible heuristic function?

- Try to relax the problem, from which an optimal solution can be found easily.
- >Learn from experience.
- → Can machines invite an admissible heuristic automatically?

#### References

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