

# Informed (heuristic) search

- Assign a cost to each action/rule/arc
- Heuristic function:  
 $h(n)$  = estimated cost of the cheapest path from the state at node  $n$  to a goal state.

# Example: 8-puzzle

Initial node  $n_I$

5	4	
6	1	8
7	3	2

Goal node  $n_G$

1	2	3
8		4
7	6	5

$h(n)$  = number of tiles out of place in node  $n$

$$h(n_I) = 7$$

$$h(n_G) = 0$$

# Example: 8-puzzle

Initial node  $n_I$

5	4	
6	1	8
7	3	2

Goal node  $n_G$

1	2	3
8		4
7	6	5

$h(n)$  = sum of Manhattan distance of tiles out of place

$$h(n_I) = 2 + 3 + 3 + 2 + 4 + 2 + 0 + 2 = 18$$

$$h(n_G) = 0$$

# Greedy search algorithm

$L \leftarrow [\text{start nodes}]$

while  $L \neq \text{empty}$  do

remove node with lowest  $h(n)$  value from  $L$ , call it  $p$

if  $p$  is a goal node, return(success)

generate all successor states of  $p$ , and add them to  $L$

endwhile

return(fail)

# Greedy search algorithm

- Not guaranteed to find optimal path
- May not terminate

# A\* search: Minimizing total path cost

- Let

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

be the cost of an optimal path going through node  $n$ ,  
where

$g^*(n)$  is cost of an optimal path from initial node to  $n$ ,

$h^*(n)$  is cost of an optimal path from  $n$  to a goal node

- But  $f^*(n)$  is difficult to know, so we *estimate*

# A\* search: Minimizing total path cost

- Let

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

be the cost of a path going through node  $n$  to a goal node,  
where

$g(n)$  is cost of path from initial state to  $n$  that was found,

$h(n)$  is a heuristic estimate of cost from  $n$  to a goal

# Algorithm A\* for searching a tree

$L \leftarrow [\text{start nodes}]$

while  $L \neq \text{empty}$  do

remove node with lowest  $f(n)$  value from  $L$ , call it  $p$

if  $p$  is a goal node, return(success)

generate all successor states of  $p$ , and add them to  $L$

endwhile

return(fail)



# Admissible heuristics

**Definition:** A heuristic function  $h(n)$  is *admissible* if  $h(n) \leq h^*(n)$ , for all  $n$

**Theorem:** If  $h(n)$  is admissible, the  $A^*$  algorithm is guaranteed to find an optimal path to a goal node

# Dominating heuristics

**Definition:** Given admissible heuristics  $h_1(n)$  and  $h_2(n)$ ,  $h_2(n)$  *dominates*  $h_1(n)$  if, for all nodes  $n$ ,  
 $h_2(n) \geq h_1(n)$ ,  
and there exists a node  $n'$  such that  
 $h_2(n') > h_1(n')$

**Theorem:**  $A^*$  with  $h_1(n)$  expands at least as many nodes as  $A^*$  with  $h_2(n)$ , if  $h_2(n)$  dominates  $h_1(n)$

# Complexity results for A\*

- Good news:
  - A\* is complete, optimal, and optimally efficient
  - no algorithm with the same information can do better
- Bad news:
  - assume a single goal, search a tree, and each action is reversible
  - time complexity of A\* is exponential,  
$$O((b^\epsilon)^d) = O(b^{\epsilon d})$$
where  $\epsilon$  is the maximum relative error  $(h^*(n) - h(n)) / h^*(n)$   
 $b$  is the branching factor  
 $d$  is the depth of the goal node

# Iterative-deepening $A^*$

- Amount of memory needed is a problem for  $A^*$
- IDA\*
  - each iteration is a complete DFS with a cutoff (so, no priority queue)
  - branch is cutoff (node is not added to L) if  $f(n)$  exceeds some threshold
  - next iteration sets threshold to be minimum of values that exceeded old threshold
  - time complexity depends strongly on number of different values the heuristic can take on