

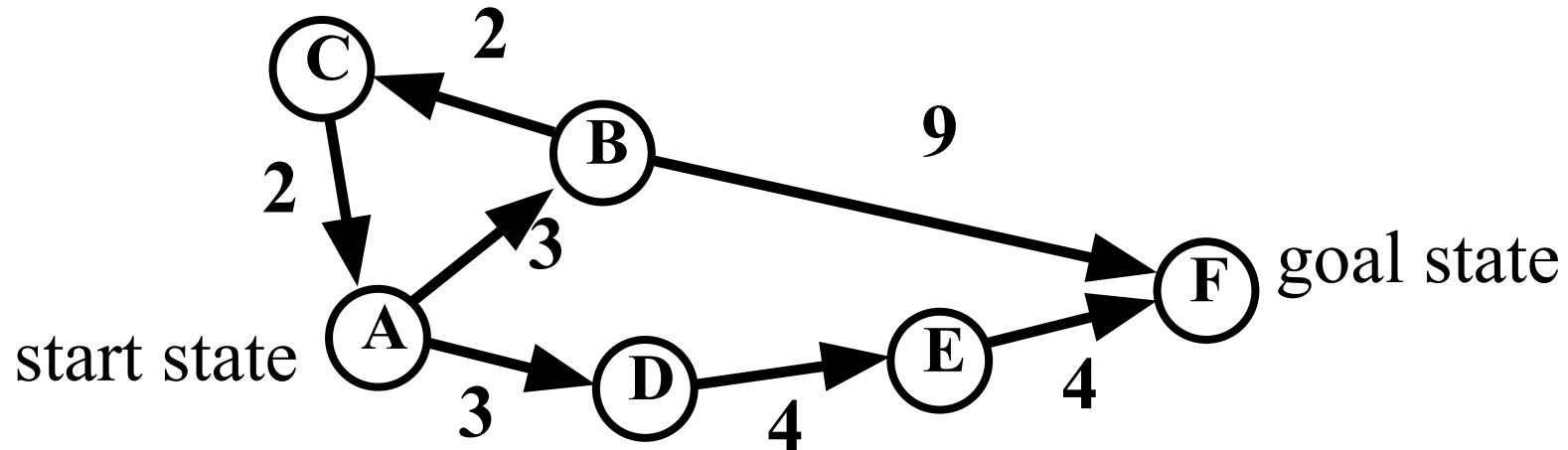
# **Informed /Heuristic Search**

# Informed search

- Uses problem-specific knowledge beyond the definition of the problem itself
- Can find solutions more efficiently than an uninformed strategy.
- Expand closer-seeming nodes first

# Heuristics

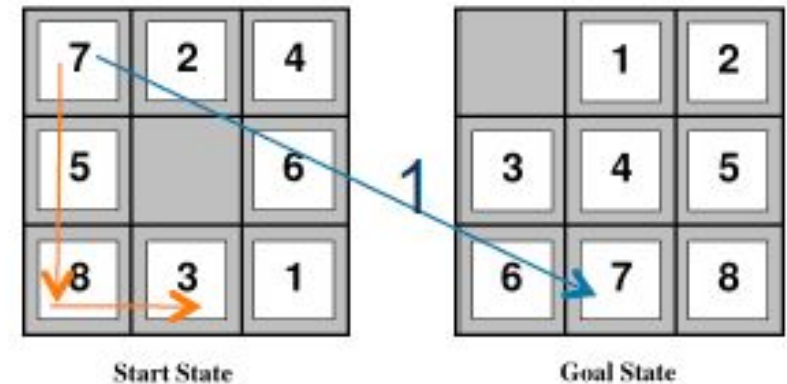
- Key notion: **heuristic function**  $h(n)$  gives an estimate of the distance from  $n$  to the goal
  - $h(n)=0$  for goal nodes
- E.g. **straight-line distance** for traveling problem



- Say:  $h(A) = 9$ ,  $h(B) = 8$ ,  $h(C) = 9$ ,  $h(D) = 6$ ,  $h(E) = 3$ ,  $h(F) = 0$
- **We're adding something new to the problem!**
- **Can use heuristic to decide which nodes to expand first**

# Examples - Heuristics

- ◆ Travel planning
  - Euclidean distance
- ◆ 8-puzzle
  - Manhattan distance
  - Number of misplaced tiles
- ◆ Traveling salesman problem
  - Minimum spanning tree



# Best First Search

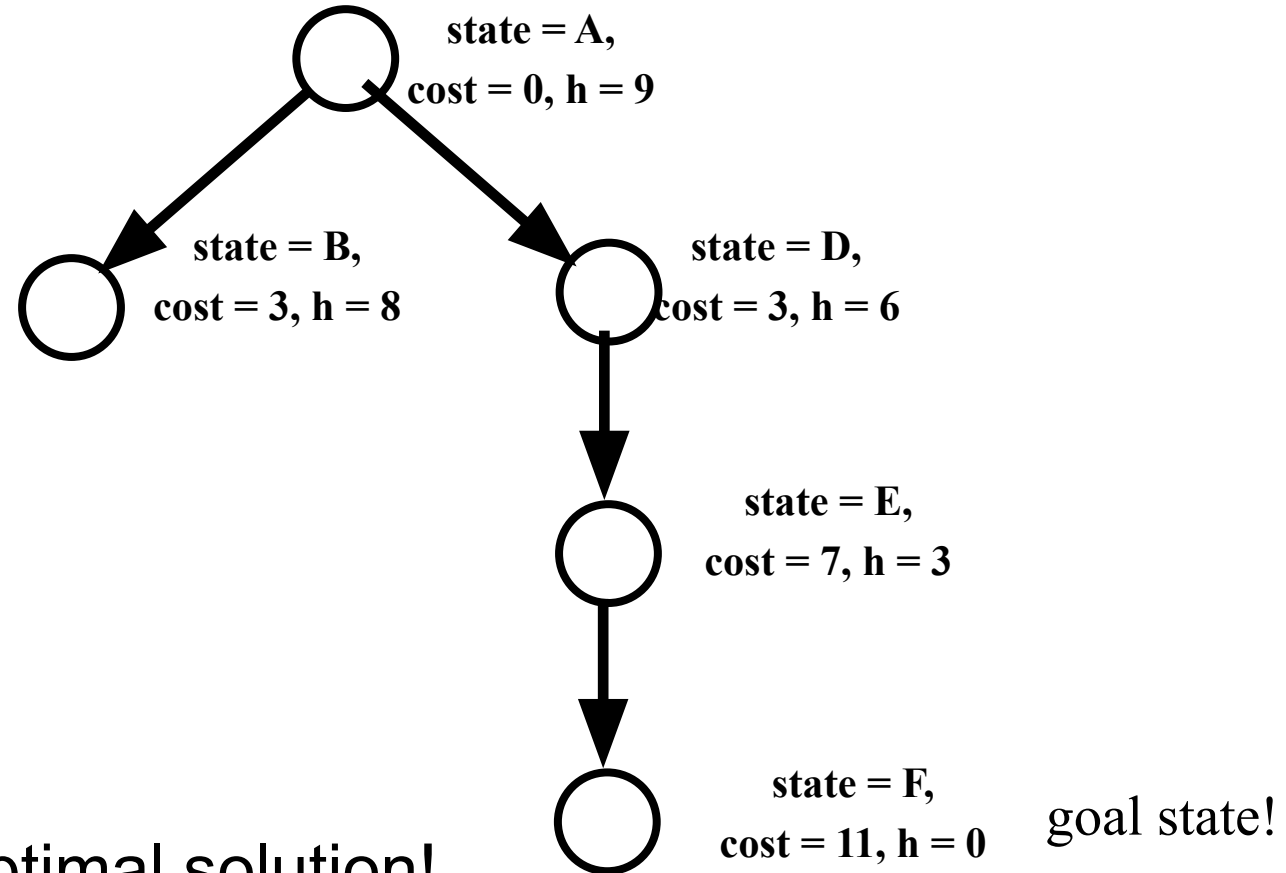
- Best-first search is an instance of the general TREE-SEARCH or GRAPH-SEARCH algorithm in which a node is selected for expansion based on an **evaluation function**,  $f(n)$
- Best-first search can be implemented within our general search framework via a priority queue, a data structure that will maintain the fringe in ascending order of  $f$ -values.
- If the evaluation function is exactly accurate, then this will indeed be the best node

# Best First Search

- There is a whole family of BEST-FIRST-SEARCH algorithms with different evaluation functions.
- A key component of these algorithms is a heuristic function  $h(n)$   
 $h(n)$  = estimated cost of the cheapest path from node  $n$  to a goal node
- Heuristic functions are the most common form in which additional knowledge of the problem is imparted to the search algorithm
- if  $n$  is a goal node, then  $h(n) = 0$ .

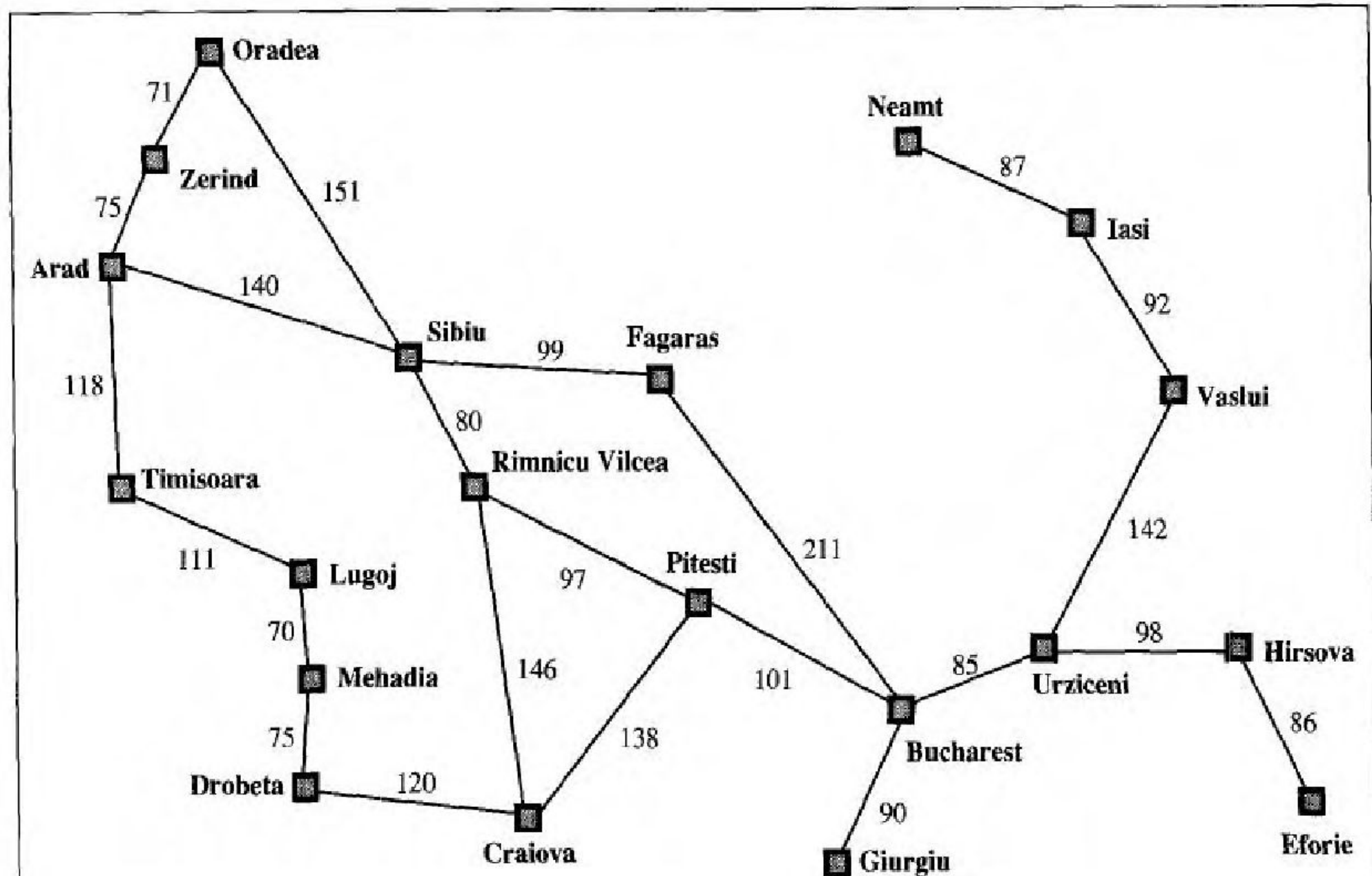
# Greedy best-first search

- Thus, it evaluates nodes by using just the heuristic function:  $f(n) = h(n)$ .
- Greedy best-first search: expand nodes with lowest  $h$  values first



- Rapidly finds the optimal solution!
- Does it always?

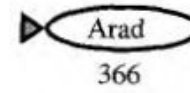
A simplified road map of part of Romania.



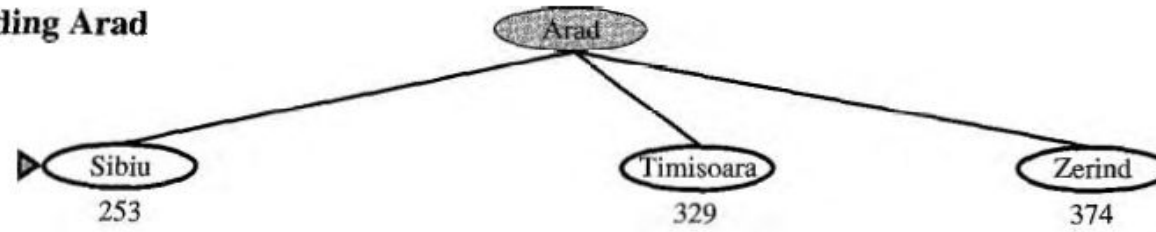


<b>Arad</b>	366	<b>Mehadia</b>	241
<b>Bucharest</b>	0	<b>Neamt</b>	234
<b>Craiova</b>	160	<b>Oradea</b>	380
<b>Drobeta</b>	242	<b>Pitesti</b>	100
<b>Eforie</b>	161	<b>Rimnicu Vilcea</b>	193
<b>Fagaras</b>	176	<b>Sibiu</b>	253
<b>Giurgiu</b>	77	<b>Timisoara</b>	329
<b>Hirsova</b>	151	<b>Urziceni</b>	80
<b>Iasi</b>	226	<b>Vaslui</b>	199
<b>Lugoj</b>	244	<b>Zerind</b>	374

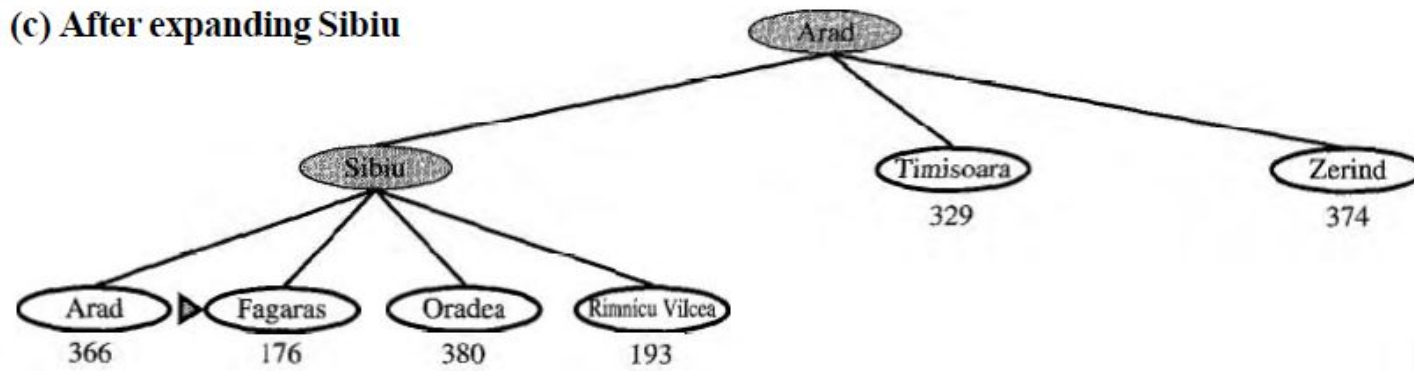
(a) The initial state



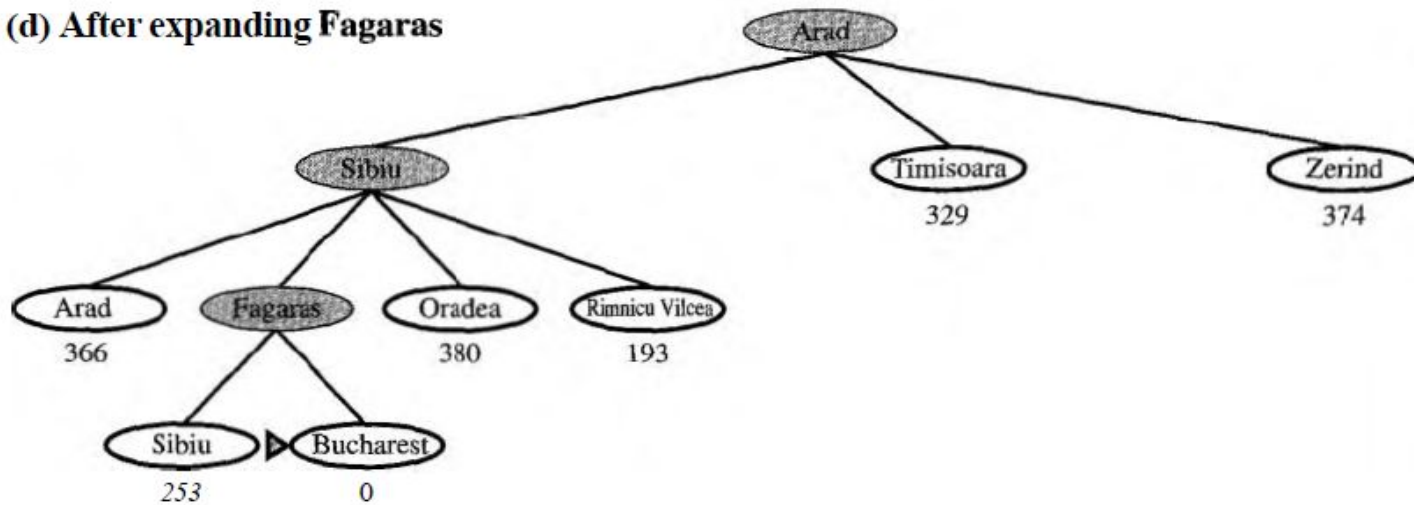
(b) After expanding Arad



(c) After expanding Sibiu



(d) After expanding Fagaras



# A\* search:

Minimizing the total estimated solution cost

- The most widely-known form of best-first search is called A\* search
- It evaluates nodes by combining  $g(n)$ , the cost to reach the node, and  $h(n)$ , the cost to get from the node to the goal

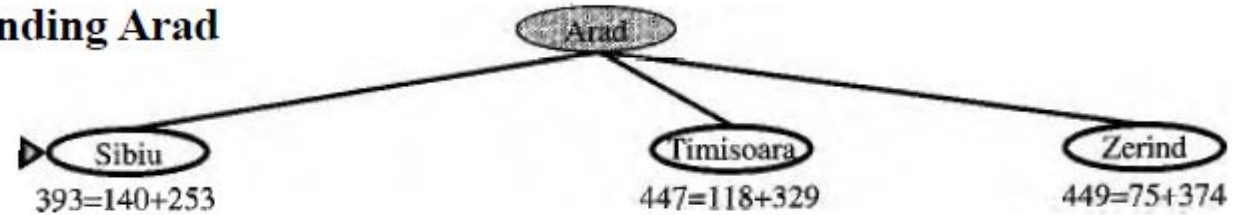
$$f(n) = g(n) + h(n)$$

$g(n)$  □ gives the path cost from the start node to node  $n$ , and  
 $h(n)$  □ the estimated cost of the cheapest path from  $n$  to the goal,

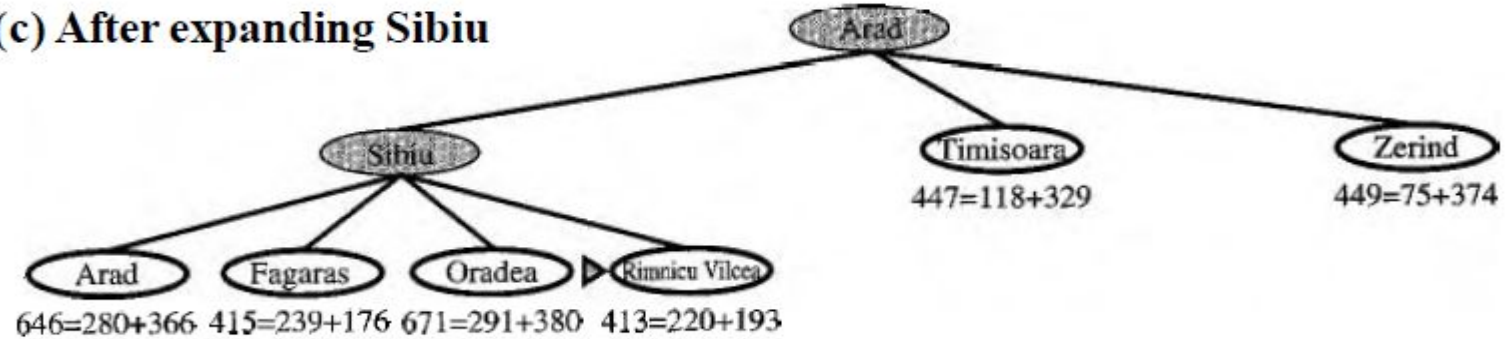
(a) The initial state

$$366=0+366$$

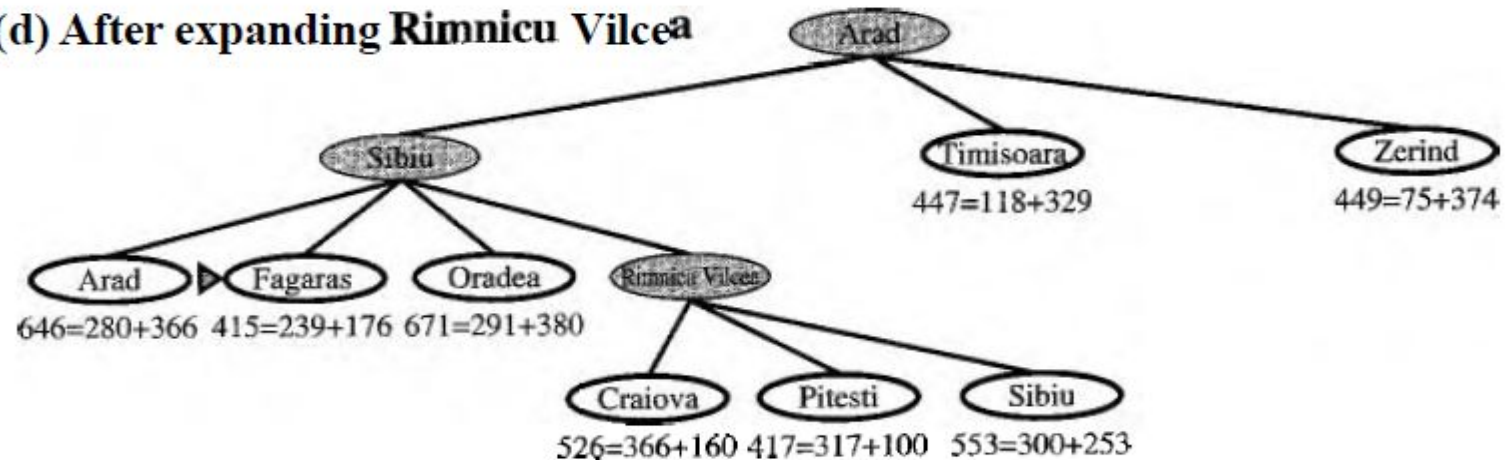
(b) After expanding Arad



(c) After expanding Sibiu

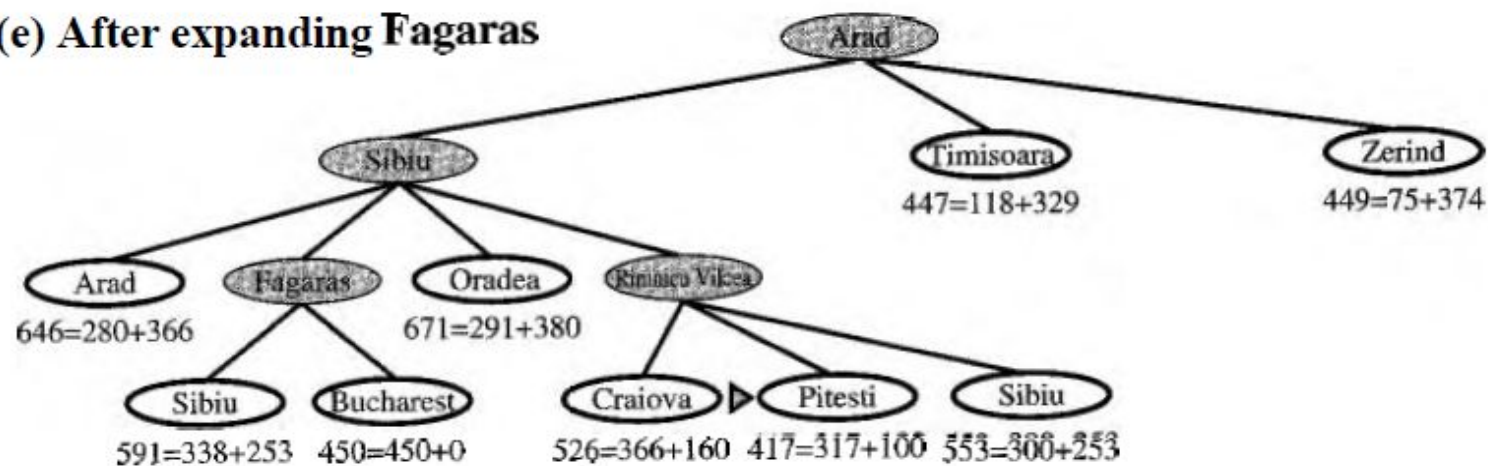


(d) After expanding Rimnicu Vilcea

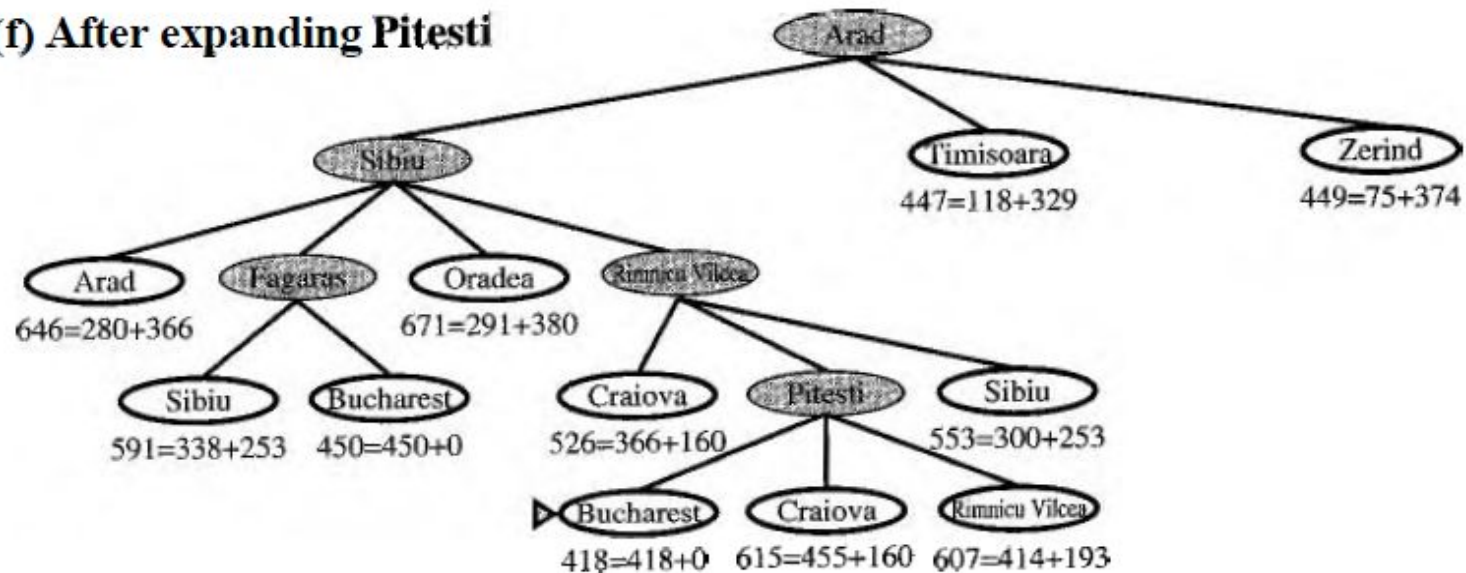




(e) After expanding Fagaras

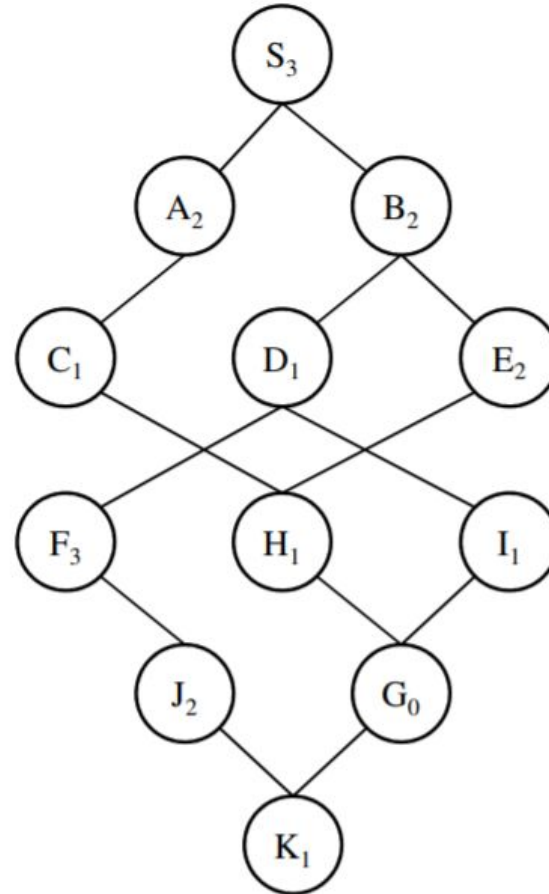


(f) After expanding Pitesti



# Example

$S_3$	$H_1$	$D_1$	$K_1$
$I_1$	$J_2$	$A_2$	$E_2$
$C_1$	$B_2$	$G_0$	$F_3$



Depth-first search.

Breadth-first search.

$S$  ( $0+3=3$ ),  $A$  ( $1+2=3$ ),  $B$  ( $1+2=3$ ),  
 $C$  ( $2+1=3$ ),  $D$  ( $2+1=3$ ),  $E$  ( $2+2=4$ ),  
 $H$  ( $3+1=4$ ),  $G$  ( $4+0=4$ )

$A^*$

The letter in each node is its name and the subscript digit is the heuristic value. All transitions have cost 1.



# Admissible heuristics

A heuristic  $h(n)$  is **admissible** if  
for every node  $n$ ,

$$h(n) \leq h^*(n)$$

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**

**Is the Straight Line Distance heuristic  $h_{SLD}(n)$   
*admissible?***

**Yes, it never overestimates the actual road distance**

# Admissibility

- A heuristic is **admissible** if it never overestimates the distance to the goal. i.e, If  $n$  is the optimal solution reachable from  $n'$ ,  
then  $g(n) \geq g(n') + h(n')$
- $A^*$  is admissible if it uses an admissible heuristic, and  $h(\text{goal}) = 0$ .



# Consistent (monotone) Heuristics

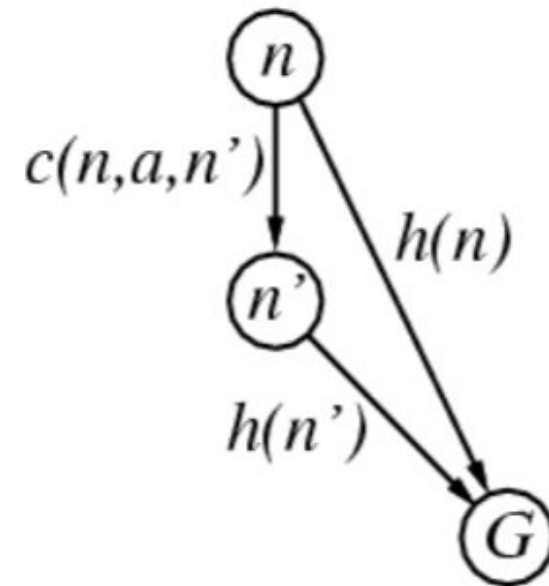
- A heuristic is **consistent** if for every node  $n$ , every successor  $n'$  of  $n$  generated by any action  $a$ ,

$$h(n) \leq c(n,a,n') + h(n')$$

- If  $h$  is consistent, we have

$$\begin{aligned} f(n') &= g(n') + h(n') \\ &= g(n) + c(n,a,n') + h(n') \\ &\geq g(n) + h(n) \\ &= f(n) \end{aligned}$$

- i.e.,  $f(n)$  is non-decreasing along any path.
- Theorem:** If  $h(n)$  is consistent,  $f$  along any path is non-decreasing.
- Corollary:** the  $f$  values seen by A\* are non-decreasing.



# Sample Heuristic functions

$h(n)$  = Number of tiles out of position.

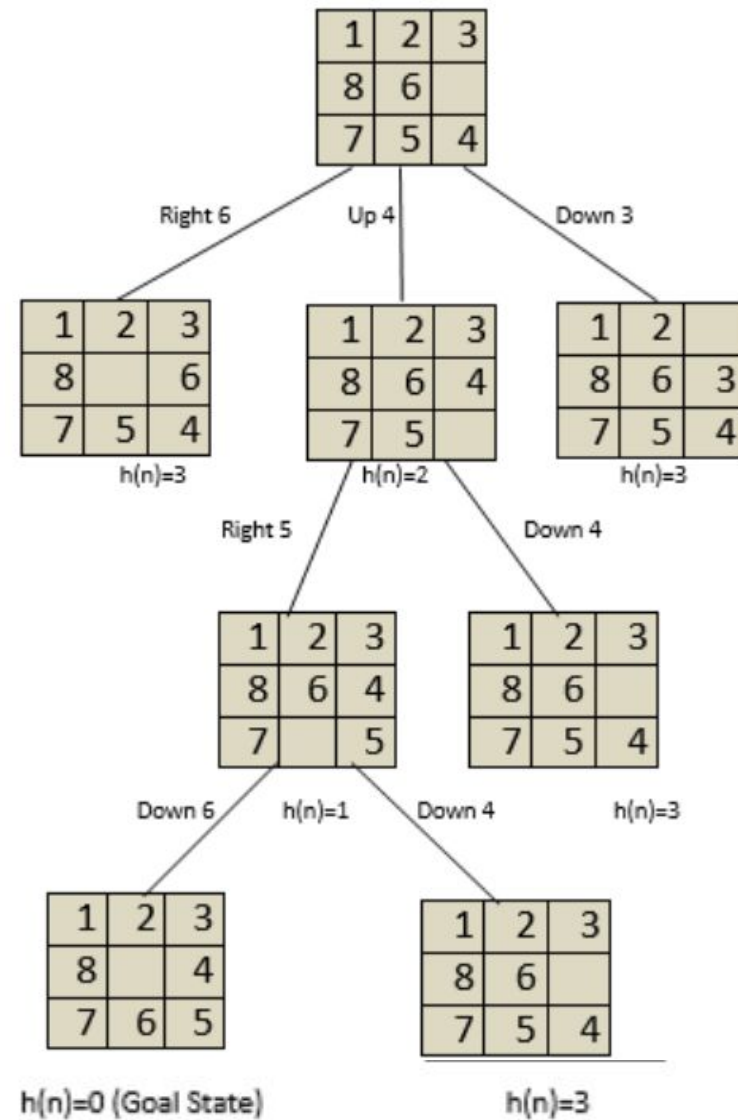
$$h(n)=3$$

1	2	3
8	6	
7	5	4

Start State

1	2	3
8		4
7	6	5

Goal State



1	2	3
	<b>4</b>	6
7	<b>5</b>	<b>8</b>

Initial state

1	2	3
4	5	6
7	8	

final state

Heuristic value= $1+1+1=3$

Set of action={up, down, left, right}

1	2	3
4	5	6
7	8	

final state

1	2	3
	4	6
7	5	8

H=3

up

right

down

	2	3
<b>1</b>	<b>4</b>	6
7	<b>5</b>	<b>8</b>

H=4

1	2	3
4		6
7	<b>5</b>	<b>8</b>

H=2

1	2	3
<b>7</b>	<b>4</b>	6
	<b>5</b>	<b>8</b>

H=4

right

down

left

up

1	2	3
	<b>4</b>	6
7	<b>5</b>	<b>8</b>

H=3

1	2	3
4	5	6
7		<b>8</b>

H=1

1	2	3
4	<b>6</b>	
7	<b>5</b>	<b>8</b>

H=3

1		3
4	<b>2</b>	6
7	<b>5</b>	<b>8</b>

H=3

7	2	4
5	*	6
8	3	1

*Initial State*

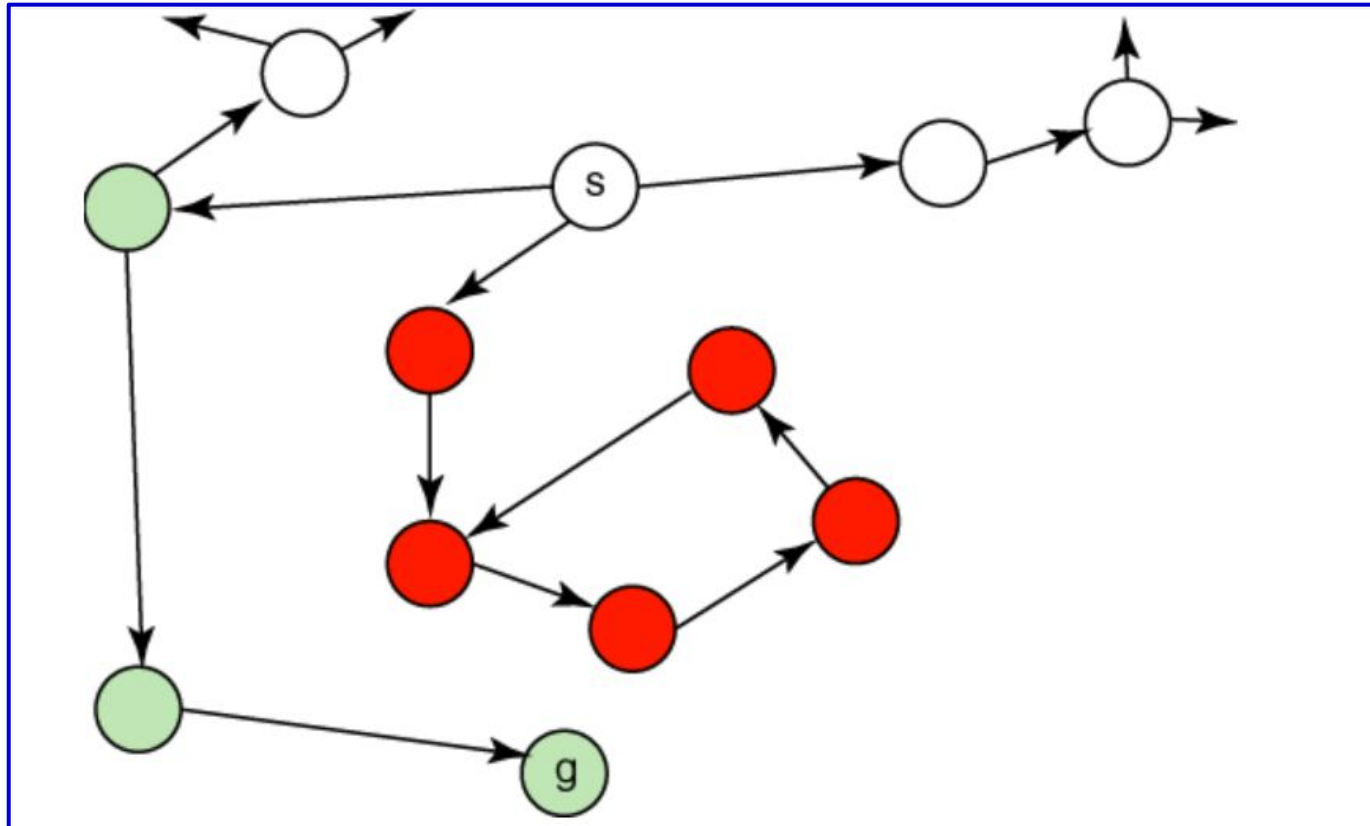
*	1	2
3	4	5
6	7	8

*Final State*

H1 : Number of misplaced tiles

H2: Sum of Euclidean distances of the tiles from their goal positions

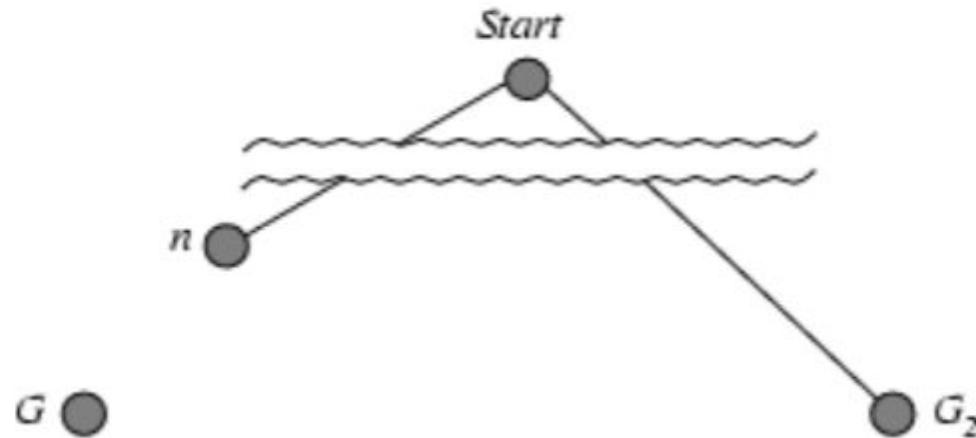
H3 : Sum of Manhattan distances of the tiles from their goal positions



A graph that is bad for best-first search

# Properties of A\*

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  $G$ .



$$\begin{aligned} f(G_2) &= g(G_2) \\ &> g(G) \end{aligned}$$

$$f(G) = g(G)$$

$$f(G_2) > f(G)$$

$$\text{since } h(G_2) = 0$$

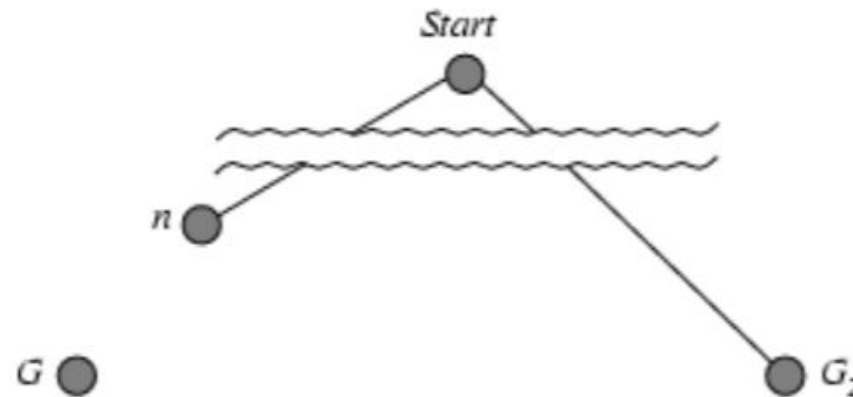
since  $G_2$  is suboptimal

$$\text{since } h(G) = 0$$

from above



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  $G$ .



$f(G_2) > f(G)$  from prev slide

$h(n) \leq h^*(n)$  since  $h$  is admissible

$g(n) + h(n) \leq g(n) + h^*(n)$

$f(n) \leq f(G) < f(G_2)$

Hence  $f(n) < f(G_2) \Rightarrow A^*$  will never select  $G_2$  for expansion.



# Properties of A\*

- Complete?
- Yes, (unless there are infinitely many nodes with  $f \leq f(G)$ )
- Time? Exponential (Worst case all the nodes are added)
- Space ? Keeps all nodes in memory
- Optimal?

**In general, A\* not practical for large scale problems due to memory requirements (all generated nodes in memory)**

**Idea: Use iterative deepening**