

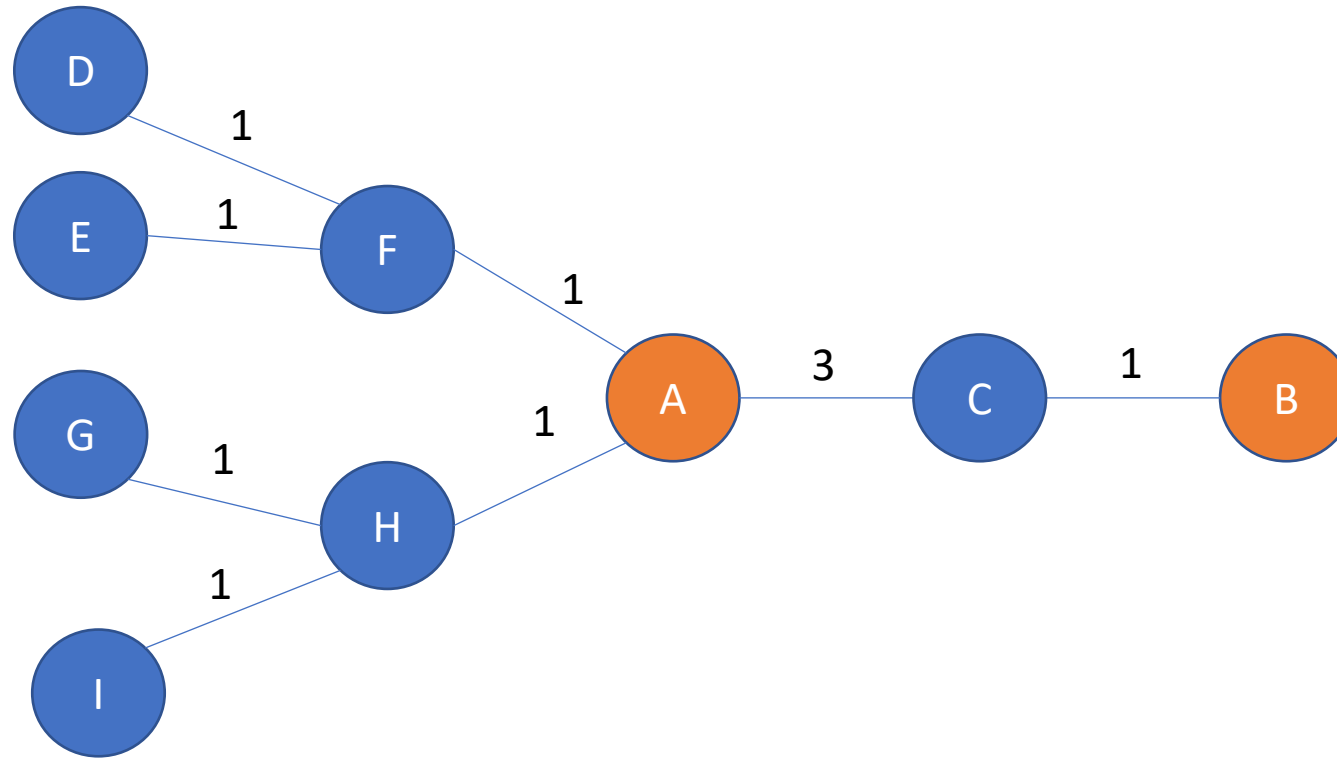
# Artificial Intelligence

## Lecture 04: Searching

# Lecture Outline

- Best-first Search
  - Greedy Search
  - A\* Search
  - Heuristics

# When UCS goes wrong...



**Goal:** Get from A... to... B

# Best-First Search

The idea:

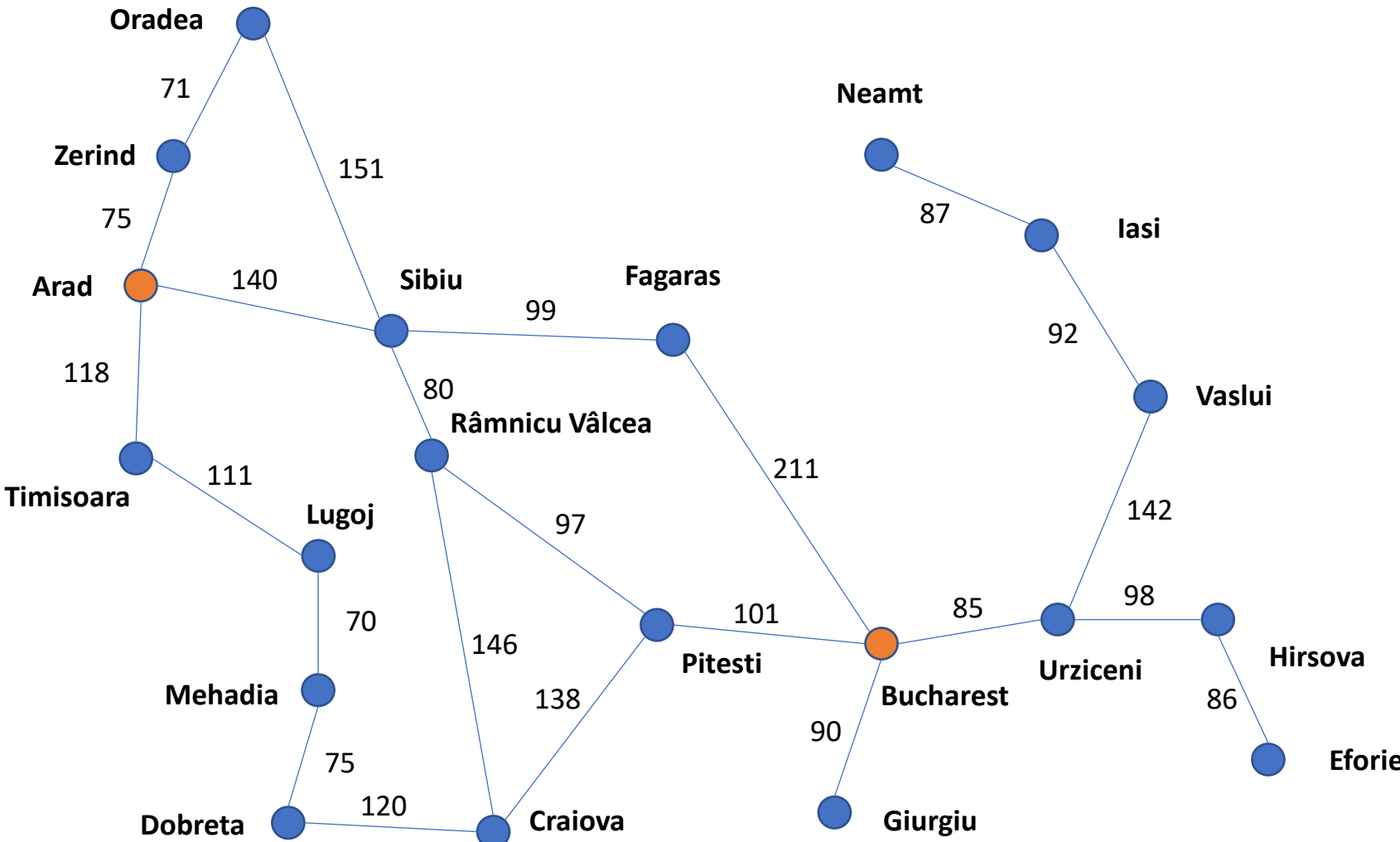
- What if we had an '**evaluation function**' which gave us an idea of the '**desirability**' of each node
- We then expand the most '**desirable**' node first

Implementation:

- The **fringe** is now sorted by **desirability**

# The Problem!

How to get from...  
**Arad to Bucharest!**  
But...



## Straightline to Bucharest

|                |     |
|----------------|-----|
| Arad           | 366 |
| Bucharest      | 0   |
| Craiova        | 160 |
| Dobreta        | 242 |
| Eforie         | 161 |
| Faragas        | 178 |
| Giurgiu        | 77  |
| Hirsova        | 151 |
| Iasi           | 226 |
| Lugoj          | 244 |
| Mehadia        | 241 |
| Neamt          | 234 |
| Oradea         | 380 |
| Pitesti        | 98  |
| Rimnicu Vilcea | 193 |
| Sibiu          | 253 |
| Timisoara      | 329 |
| Urziceni       | 80  |
| Vaslui         | 199 |
| Zerind         | 374 |

# Greedy Search

**Evaluation function**  $h(n)$

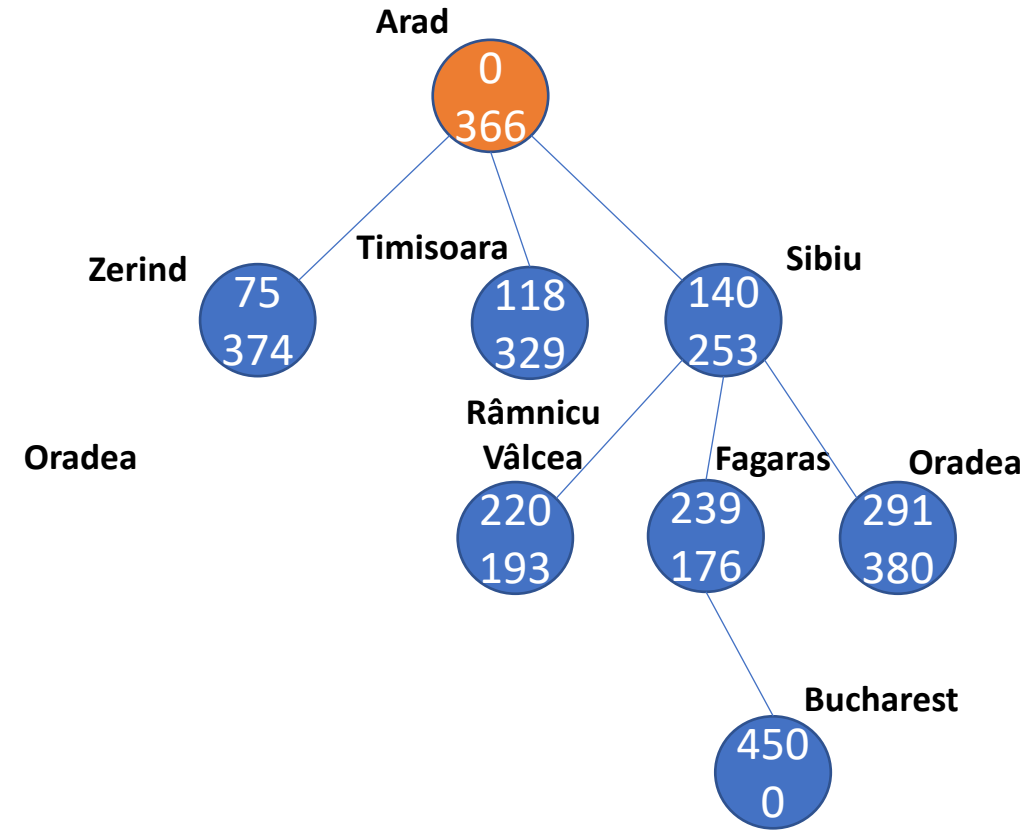
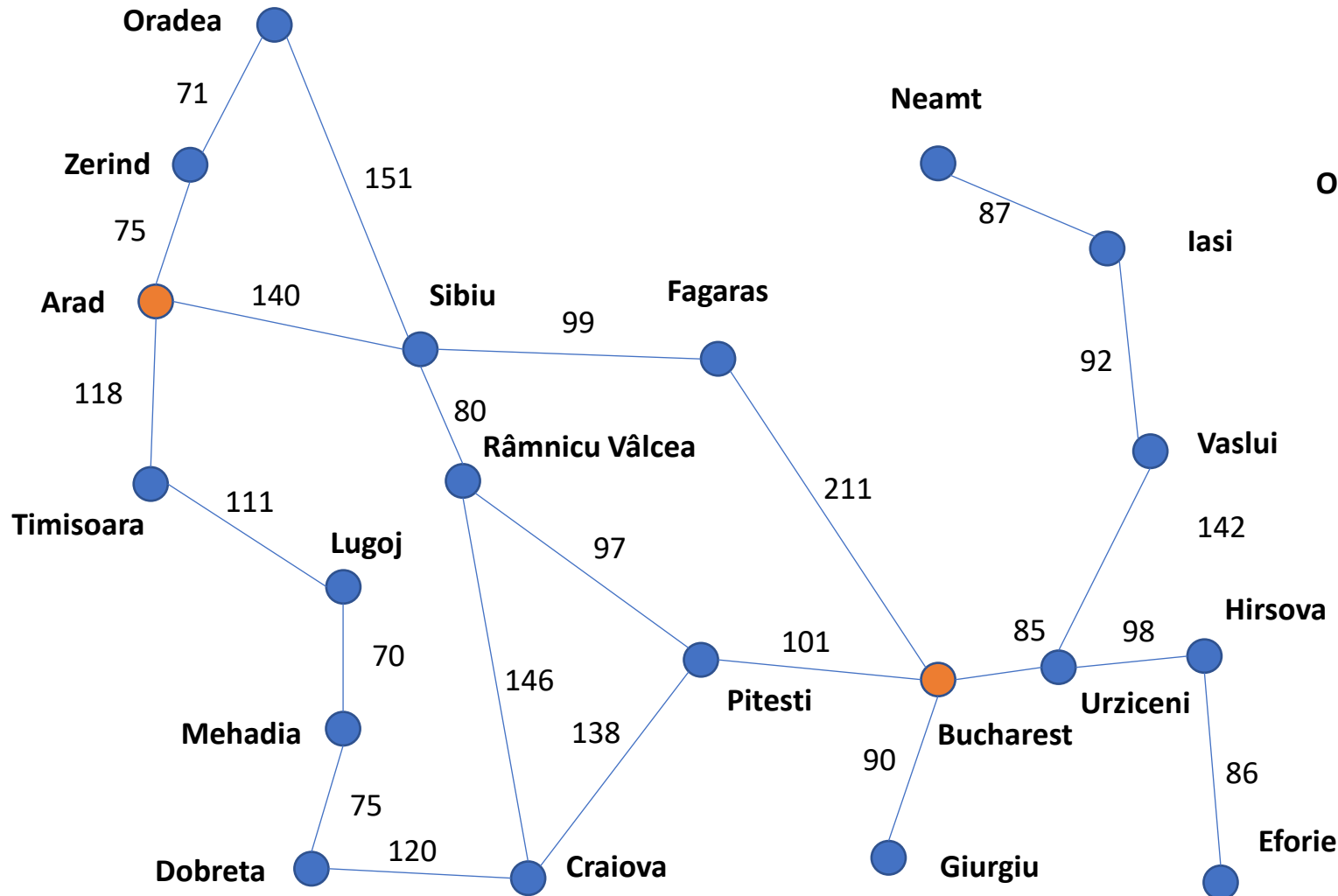
(h for heuristic)

In our case,  $h(n)$  is the straight-line distance to Bucharest

Greedy search expands the node that **appears** to be the closest to the goal

Side note: this use of the term **Greedy** isn't exactly the same as the usual use of a 'greedy algorithm'

# Greedy Search



# Greedy Search Analysis

Complete?

- No (Yes if finite and state checks)

Time Complexity?

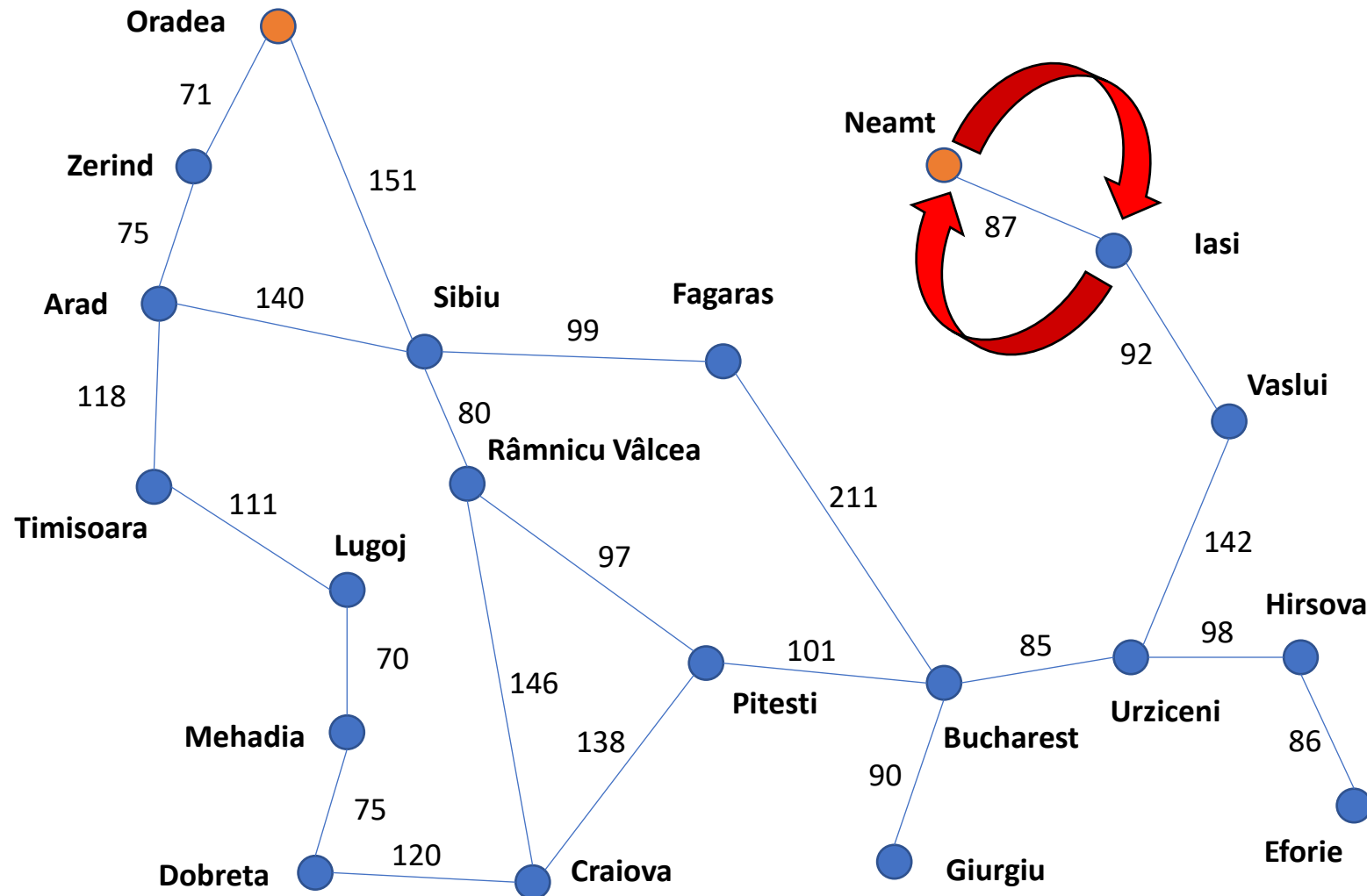
- $O(b^m)$
- Dependent on heuristic

Space Complexity?

- $O(b^m)$
- Keeps all nodes in memory

Optimal?

- No...





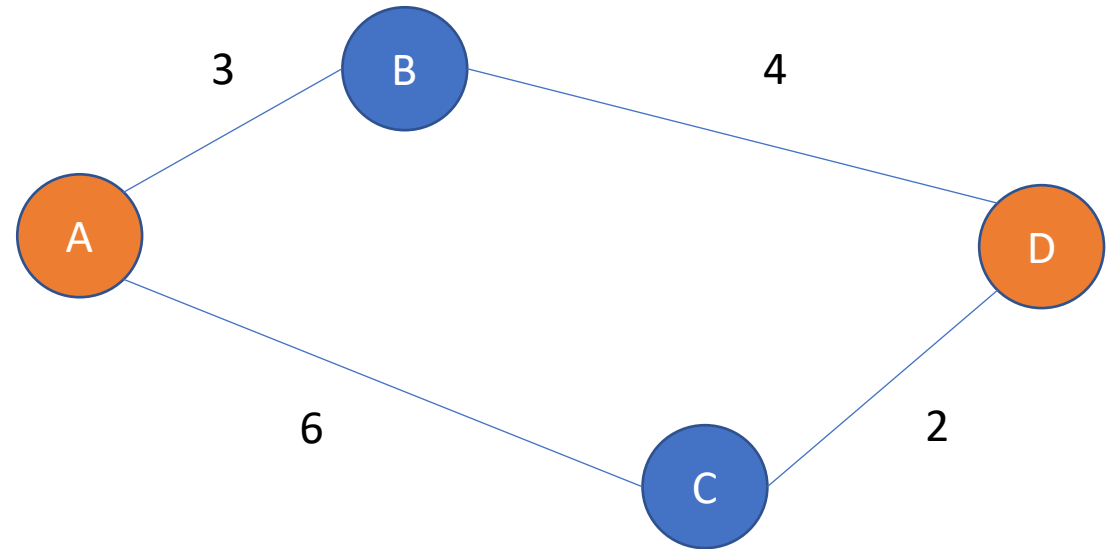
# A\* Search

Goal is  $A \Rightarrow D$

Greedy Search would expand C before B... because C is closer to D

But... , C has already cost me 6 plus an extra 2 is 8

Not as good as B which is 3 plus an extra 4 is 7



# A\* Search

A subtle variation on greedy search

Two main components:

#1: What is the current path cost already?

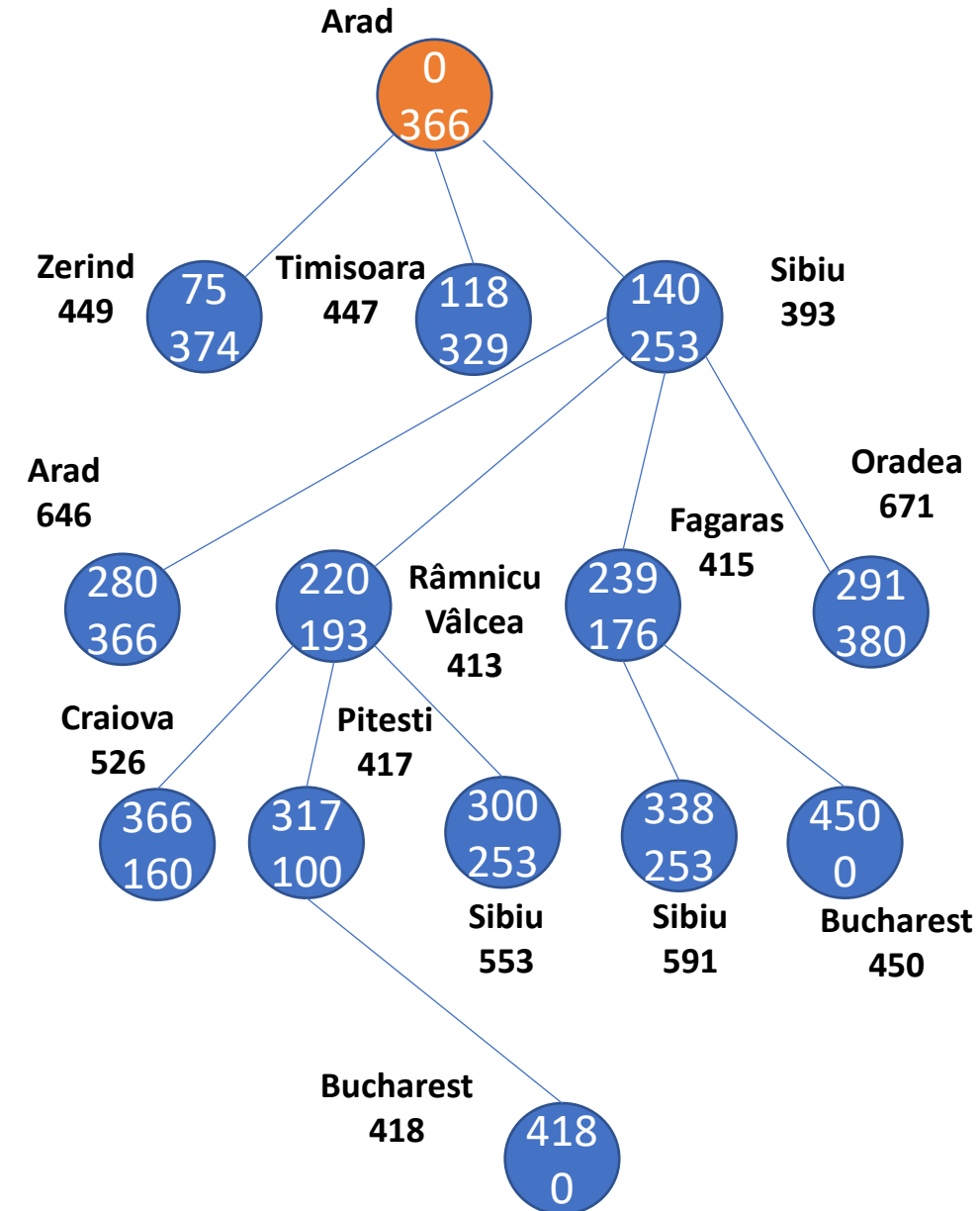
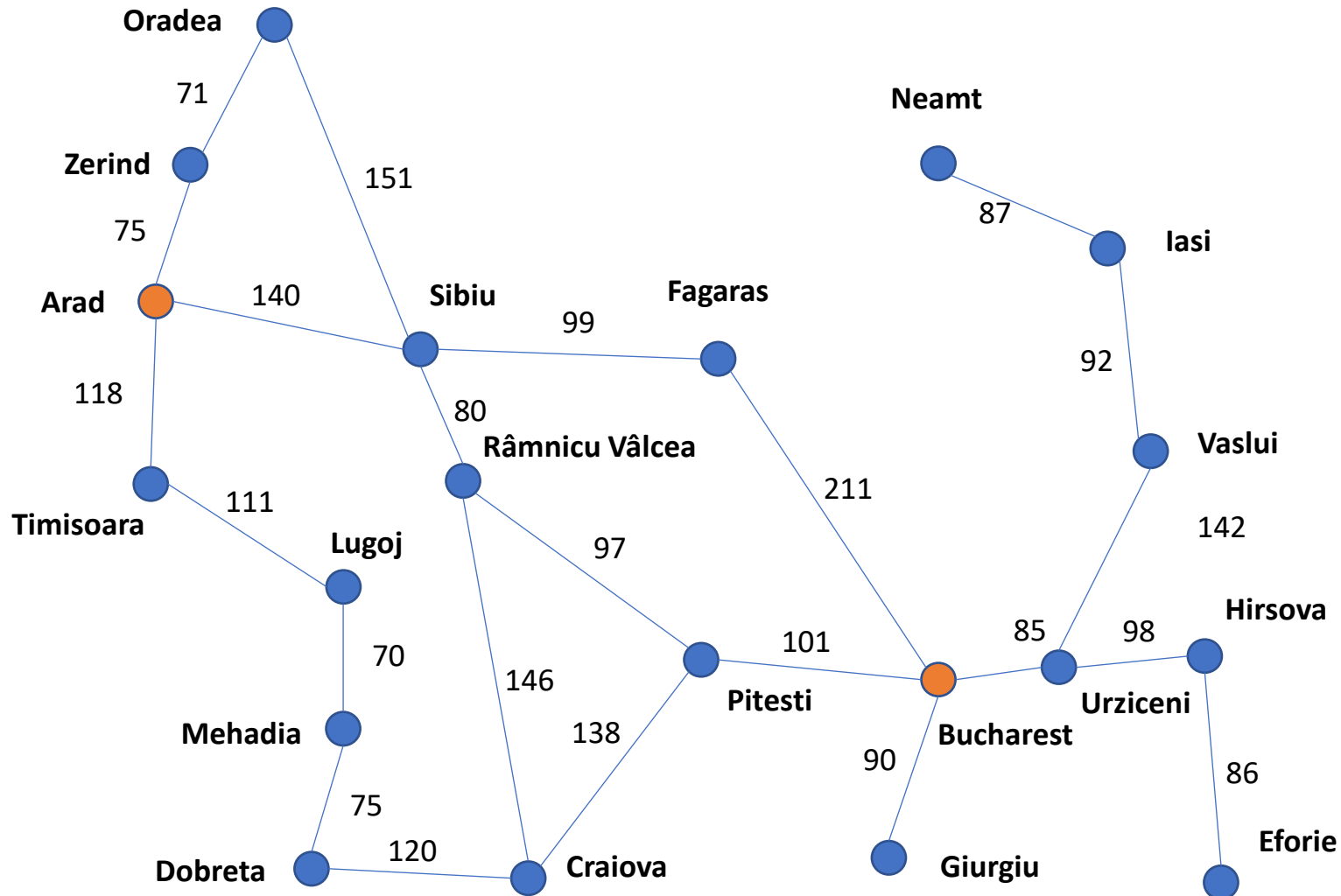
#2: What is the estimate of the remaining path?

#1:  $g(n)$ , #2:  $h(n)$

## **Evaluation Function**

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

# A\* Search



# How does A\* work?

A\* only works if the heuristic function  $h(n)$  underestimates the remaining distance

For our road-map version, the straight-line distance is always shorter than the actual road distance (can be equal).

i.e.,  $h(n) \leq h^*(n)$  ( $h^*(n)$  is the true cost)

For example...

$h(G) = 0$  (G is the goal node)

# Is it really optimal?

If we have generated a bad path...  $G_2$

Let us assume 'n' is on the real shortest path

$\Rightarrow f(G_2) = g(G_2)$  (as  $h(G_2) = 0$ )

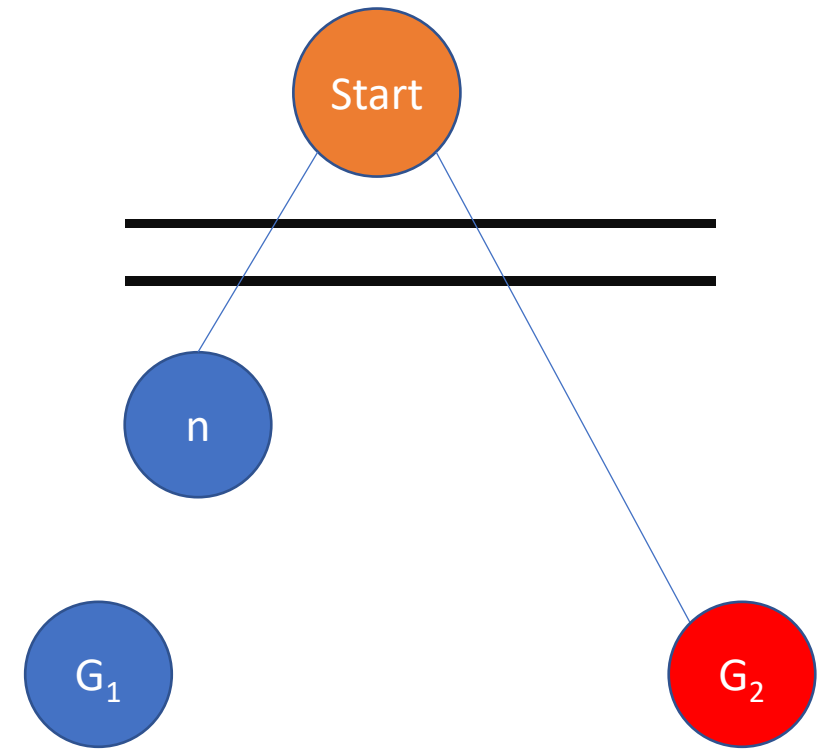
$\Rightarrow g(G_2) > g(G_1)$  (we know  $G_2$  is not optimal)

&  $g(G_1) \geq f(n)$

As  $h$  is admissible we know:

$g(G_2) > f(n)$

So we'll always expand 'n' before  $G_2$



## Note:

**Admissible** means that the heuristic **never overestimates** the remaining distance to the target

# Analysing A\*

Complete?

- Yes (unless there are infinite 'closer nodes')

Time Complexity

- Exponential in [relative error in  $h$  \* length of solution]

Space Complexity

- Keeps all nodes in memory

Optimal?

- Yes

# Some further observations about $H^*$

What happens when  $h(n)$  is always set to zero.

I.e., the estimate is always zero.

Is this admissible?

What does the algorithm look like?

What happens when  $h(n) = h^*(n)$

I.e., the heuristic is magically correct?

Is this admissible?

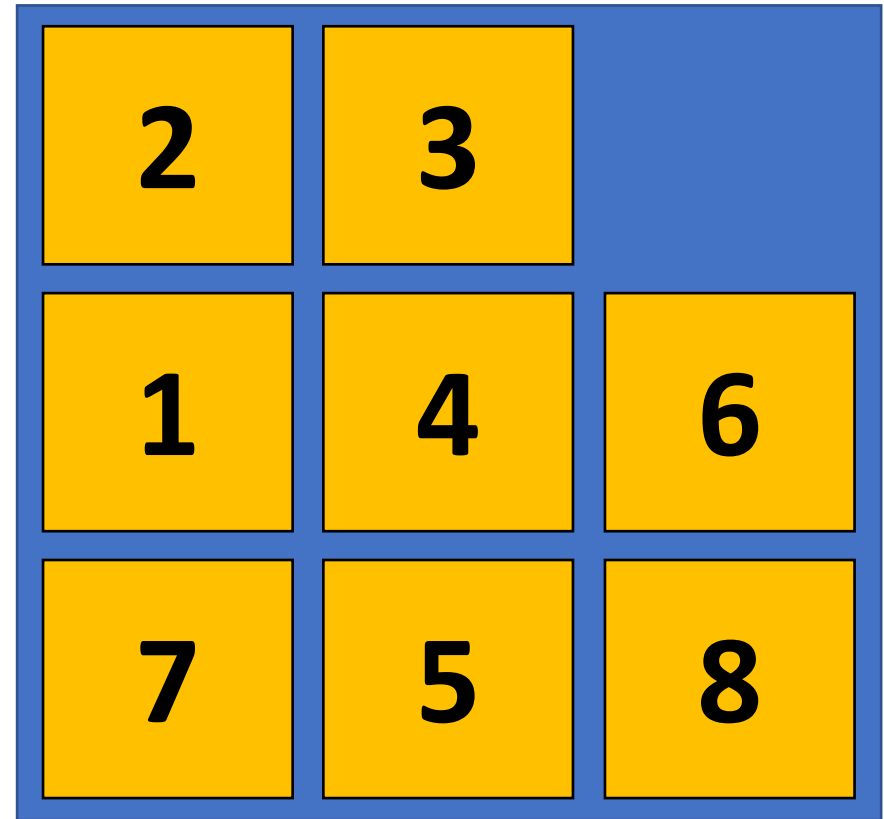
What does this algorithm look like?

# Making an Admissible heuristic

$h_1(n)$  = number of misplaced tiles

$h_2(n)$  = total **Manhattan** distance

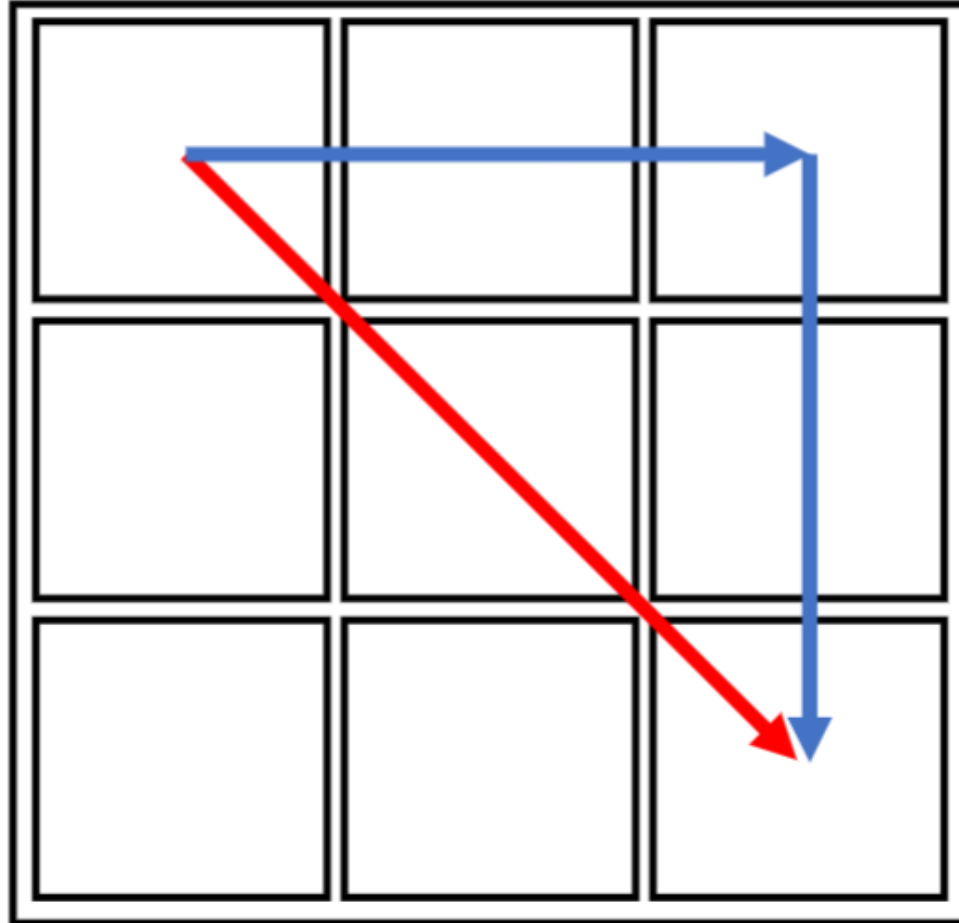
- (total number of squares from desired location no diagonals)





# Manhattan Distance

**Euclidean Distance**  
(straight line distance)  
 $= \text{sqrt}(2) * 2$



**Manhattan Distance** (horizontal + vertical distance)  
 $= 2 + 2$

# Making an Admissible heuristic

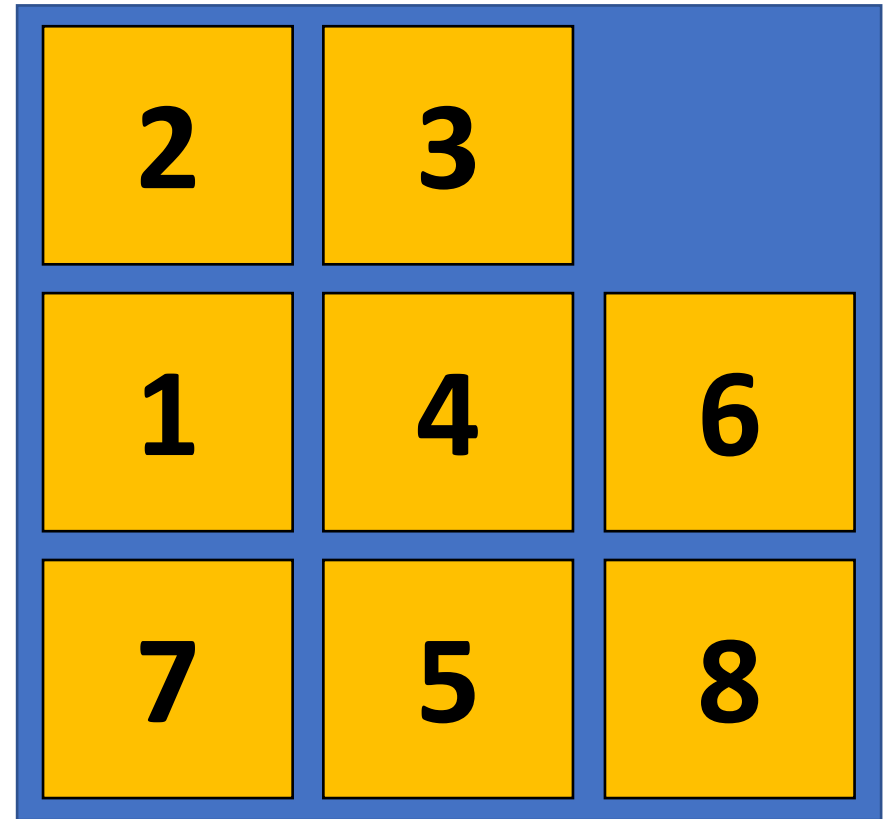
$h_1(n)$  = number of misplaced tiles

$h_2(n)$  = total **Manhattan** distance

- (total number of squares from desired location no diagonals)

$$h_1(n) = 6$$

$$h_2(n) = 1 + 1 + 1 + 1 + 0 + 0 + 1 + 1 \\ = 6$$



# Are these heuristics admissible?

$h_1(n)$  = number of misplaced tiles       $h_1$  ?

$h_2(n)$  = total **Manhattan** distance

- (total number of squares from  
desired location no diagonals)       $h_2$  ?

$$h_1(n) = 6$$

$$h_2(n) = 1 + 1 + 1 + 1 + 0 + 0 + 1 + 1 \\ = 6$$

Better to ask, how many moves  
would be necessary as a  
minimum?

Which is better?

# Intuitions about heuristics

What happens when your heuristic... is always zero?

What happens when your heuristic... is always 'correct' (i.e., magic)?

What happens when you don't take existing path into your calculation?