

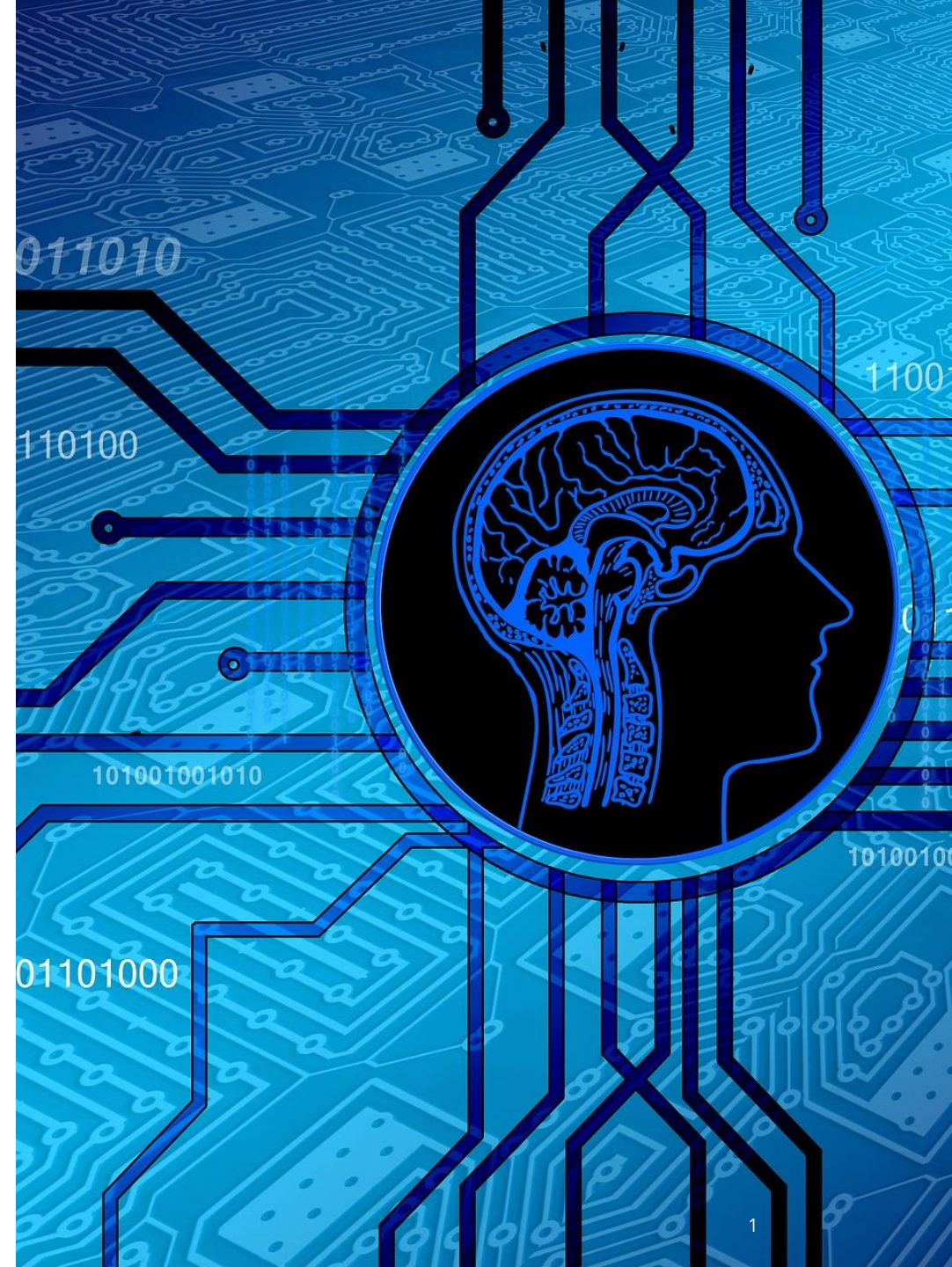
# Informed Search

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*Petros Papapanagiotou*

Informatics 2D: Reasoning and Agents

**Lecture 4**



# Review: Tree search

**function** TREE-SEARCH(*problem*) **returns** a solution, or failure

    initialize the frontier using the initial state of *problem*

**loop do**

**if** the frontier is empty **then return** failure

        choose a leaf node and remove it from the frontier

**if** the node contains a goal state **then return** the corresponding solution

        expand the chosen node, adding the resulting nodes to the frontier

A *search strategy* is defined by picking the order of node expansion from the **frontier**.

# Review: Tree search

```
function TREE-SEARCH(problem) returns a solution, or failure
  initialize the frontier using the initial state of problem
  loop do
    if the frontier is empty then return failure
    choose a leaf node and remove it from the frontier
    if the node contains a goal state then return the corresponding state
    expand the chosen node, adding the resulting nodes to the frontier
```

What if we **order** the nodes in the frontier by decreasing *desirability*?

A *search strategy* is defined by picking the order of node expansion from the **frontier**.

# Best-first search

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An instance of general TREE-SEARCH or GRAPH-SEARCH

- Use an evaluation function  $f(n)$  for each node  $n$ 
  - estimate of "desirability"
- Expand most desirable unexpanded node, usually the node with the lowest evaluation

# Heuristics

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*From the Greek «ευρίσκω» (/e 'vri.sko/), meaning to find, to discover.*

Any method that is believed or practically proven to be useful for the solution of a given problem.

- No guarantee that it will always work or lead to an optimal solution!

We use heuristics to guide tree search.

- This may not change the worst case complexity of the algorithm, but can help in the average case.

We introduce conditions (admissibility, consistency) in order to identify good heuristics, i.e. those which actually lead to an improvement over uninformed search.



# Greedy best-first search

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# Greedy best-first search

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Evaluation function  $f(n) = h(n)$  (heuristic)

- **estimated cost** of cheapest path from state at node  $n$  to a *goal* state
- e.g.,  $h_{SLD}(n)$  = straight-line distance from  $n$  to Bucharest

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



# Best-first search

Arad							
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Straight-line distance

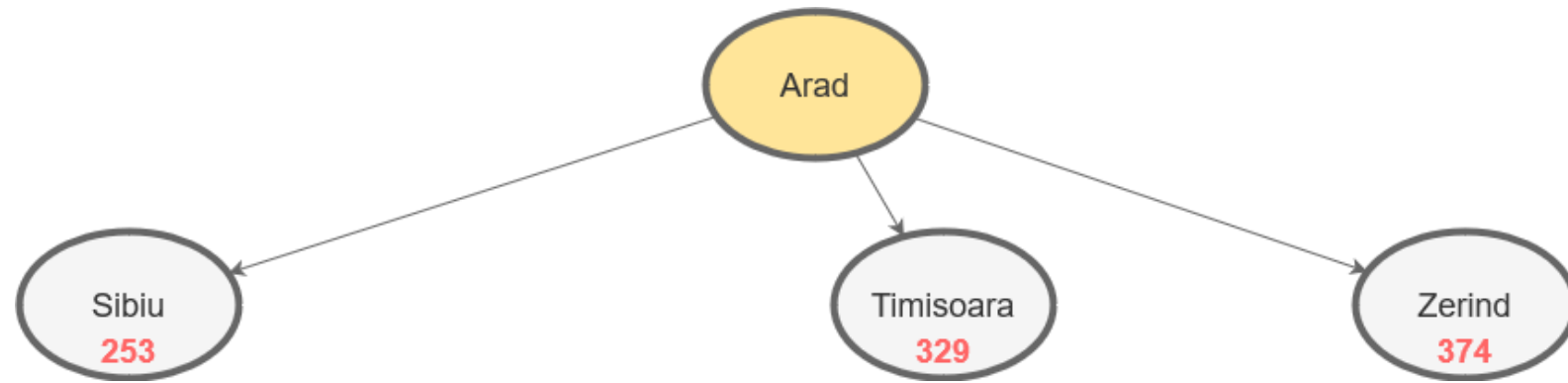
to Bucharest

Arad	366
Bucharest	0
Craiova	160
Dobreta	242
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# Best-first search

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Sibiu	Timiso- ara	Zerind					
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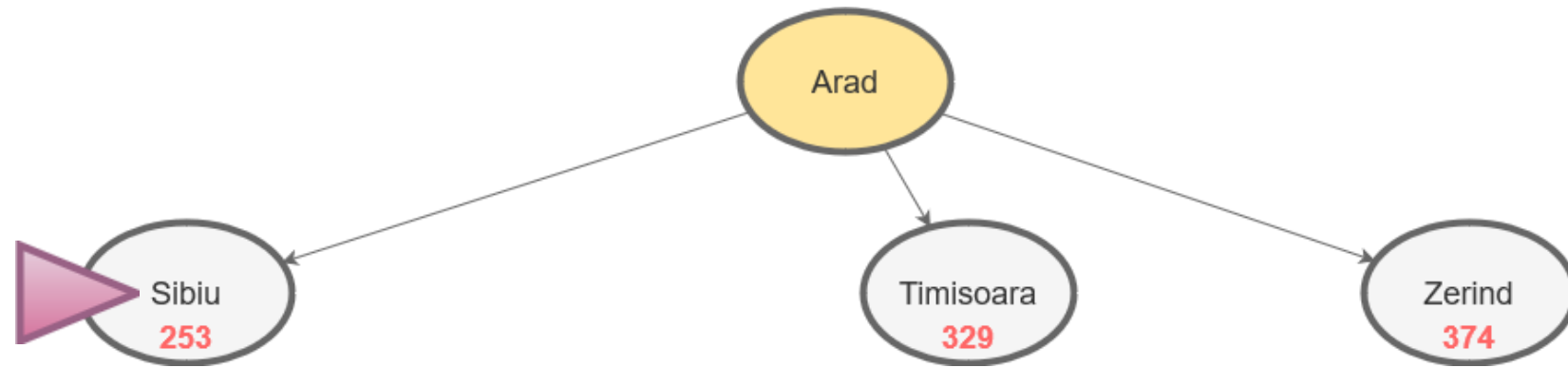


# Best-first search

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to Bucharest

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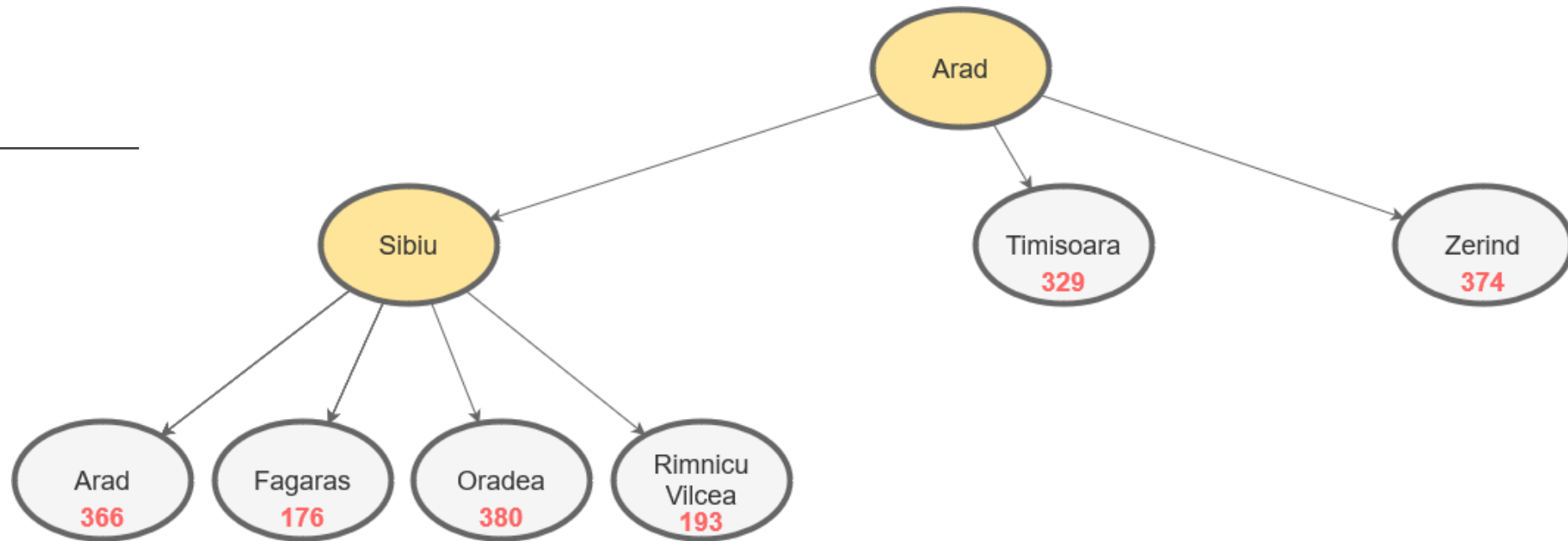
Sibiu	Timiso- ara	Zerind					
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Timiso- ara	Zerind	Arad	Fagaras	Oradea	RV		
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# Best-first search

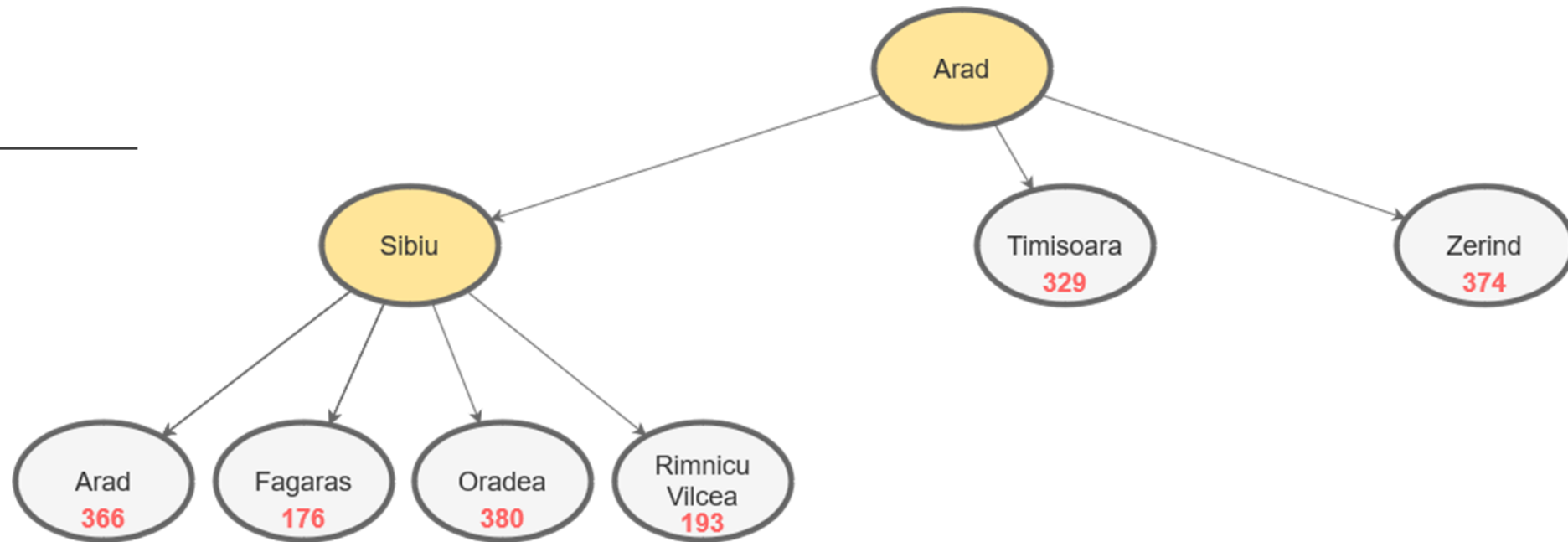
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Fagaras	RV	Timiso- ara	Arad	Zerind	Oradea		
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# Best-first search

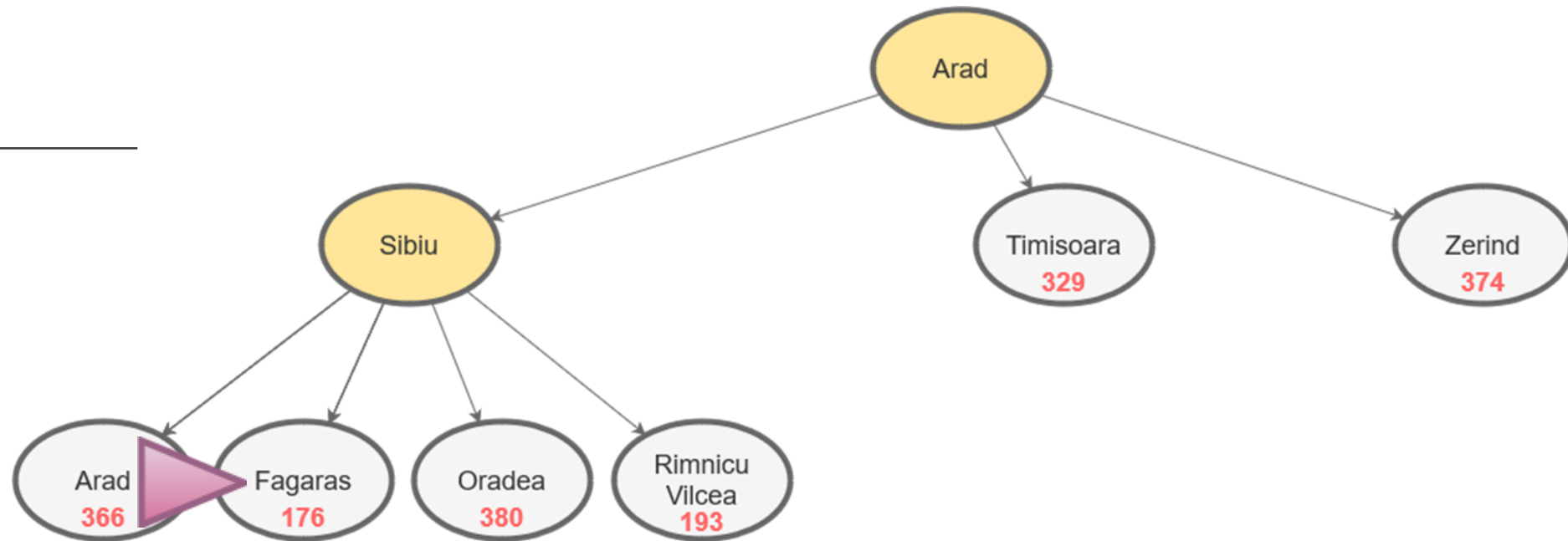
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Fagaras	RV	Timiso- ara	Arad	Zerind	Oradea		
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# Best-first search

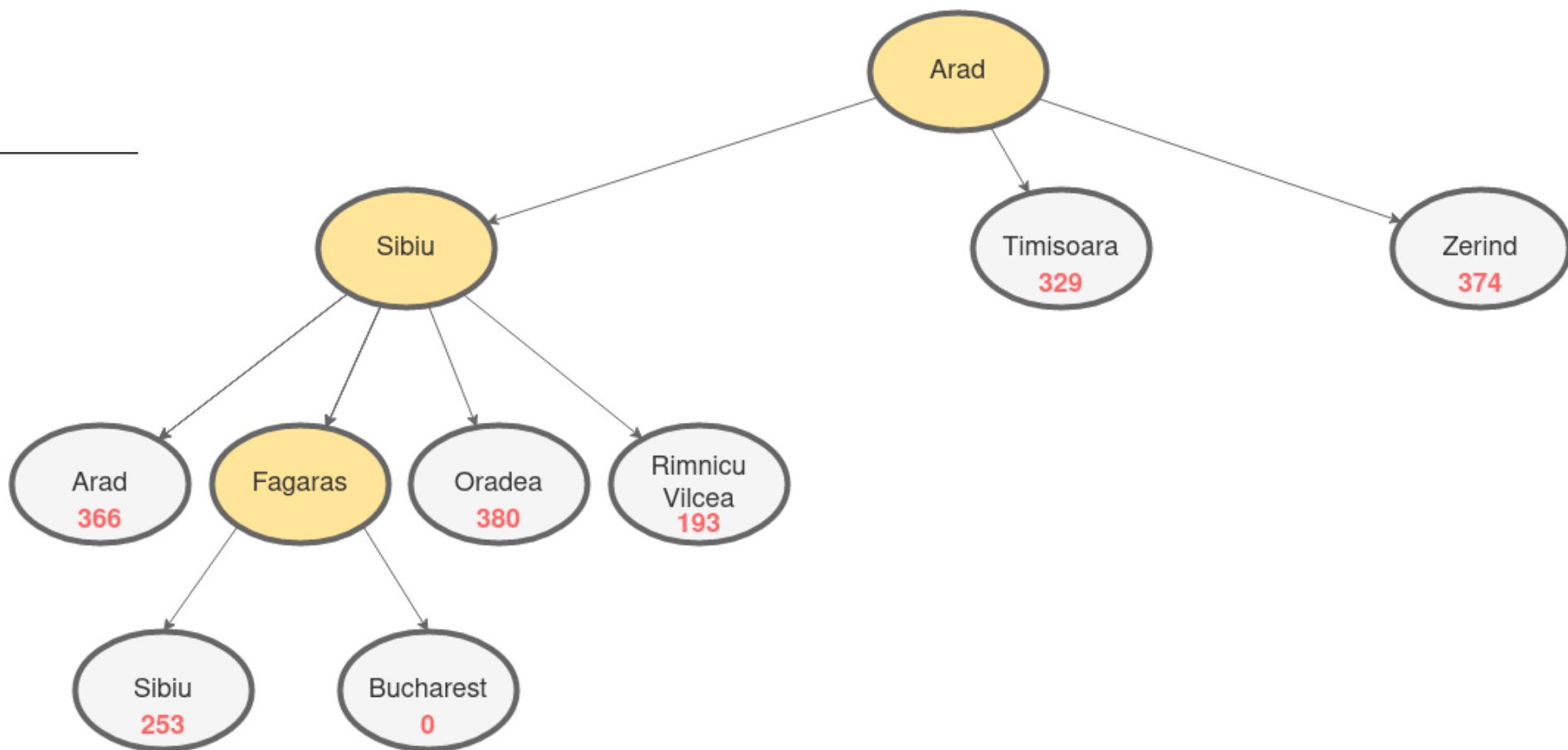
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Fagaras	RV	Timiso- ara	Arad	Zerind	Oradea	Sibiu	Bucha- rest
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# Properties of best-first search

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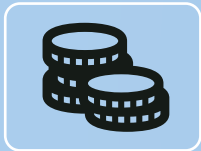
**Complete?**



**Time complexity?**



**Space complexity?**



**Optimal?**



# Properties of best-first search

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## **Complete?**

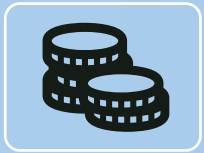
No! Can get stuck in loops.



## **Time complexity?**



## **Space complexity?**



## **Optimal?**

# Properties of best-first search

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## **Complete?**

No! Can get stuck in loops. • • •

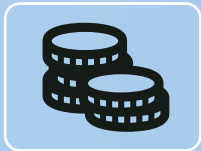


## **Time complexity?**

GRAPH-SEARCH  
is complete in  
finite spaces.



## **Space complexity?**



## **Optimal?**

# Properties of best-first search

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## **Complete?**

No! Can get stuck in loops.

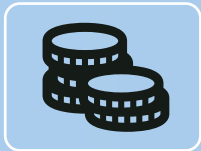


## **Time complexity?**

$O(b^m)$  for tree version



## **Space complexity?**



## **Optimal?**

# Properties of best-first search

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## **Complete?**

No! Can get stuck in loops.

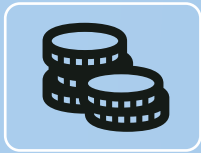


## **Time complexity?**

$O(b^m)$  for tree version



## **Space complexity?**



## **Optimal?**

A good heuristic  
can lead to  
dramatic  
improvement!

# Properties of best-first search

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## **Complete?**

No! Can get stuck in loops.



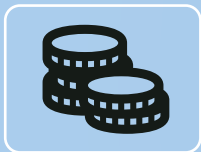
## **Time complexity?**

$O(b^m)$  for tree version



## **Space complexity?**

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



## **Optimal?**

# Properties of best-first search

---



## **Complete?**

No! Can get stuck in loops.



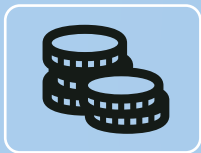
## **Time complexity?**

$O(b^m)$  for tree version



## **Space complexity?**

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



## **Optimal?**

No

# A\* Search

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# A\* search

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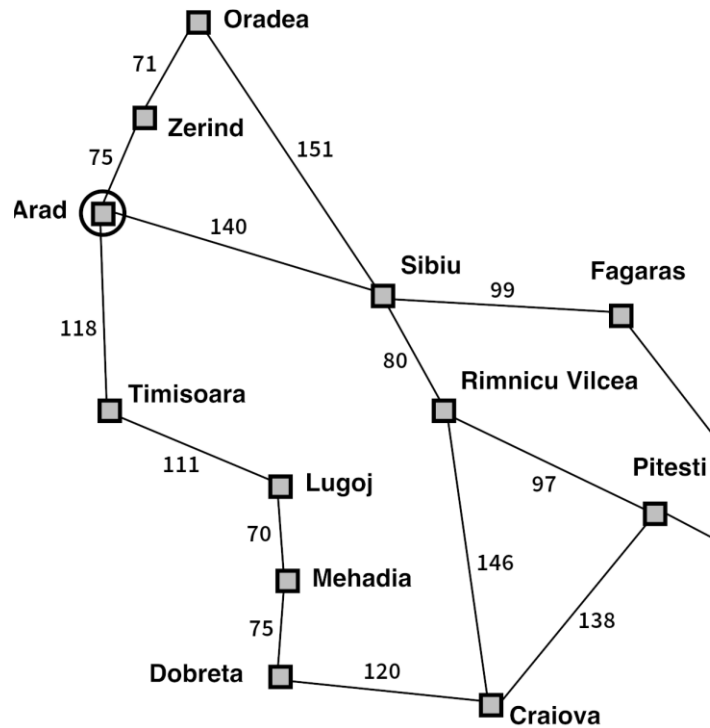
Evaluation function  $f(n) = g(n) + h(n)$

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

Avoid expanding paths that are already expensive

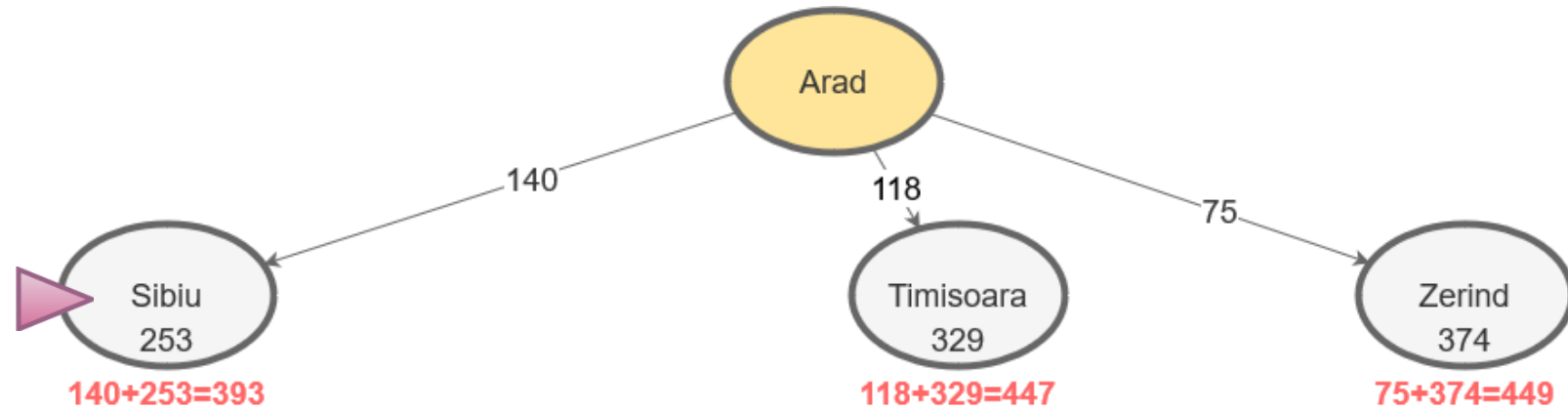
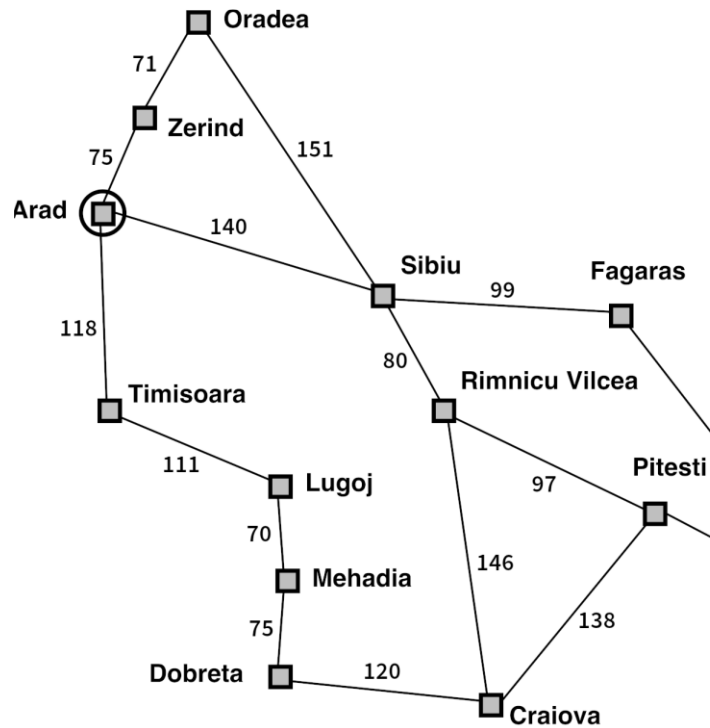


# Best-first search



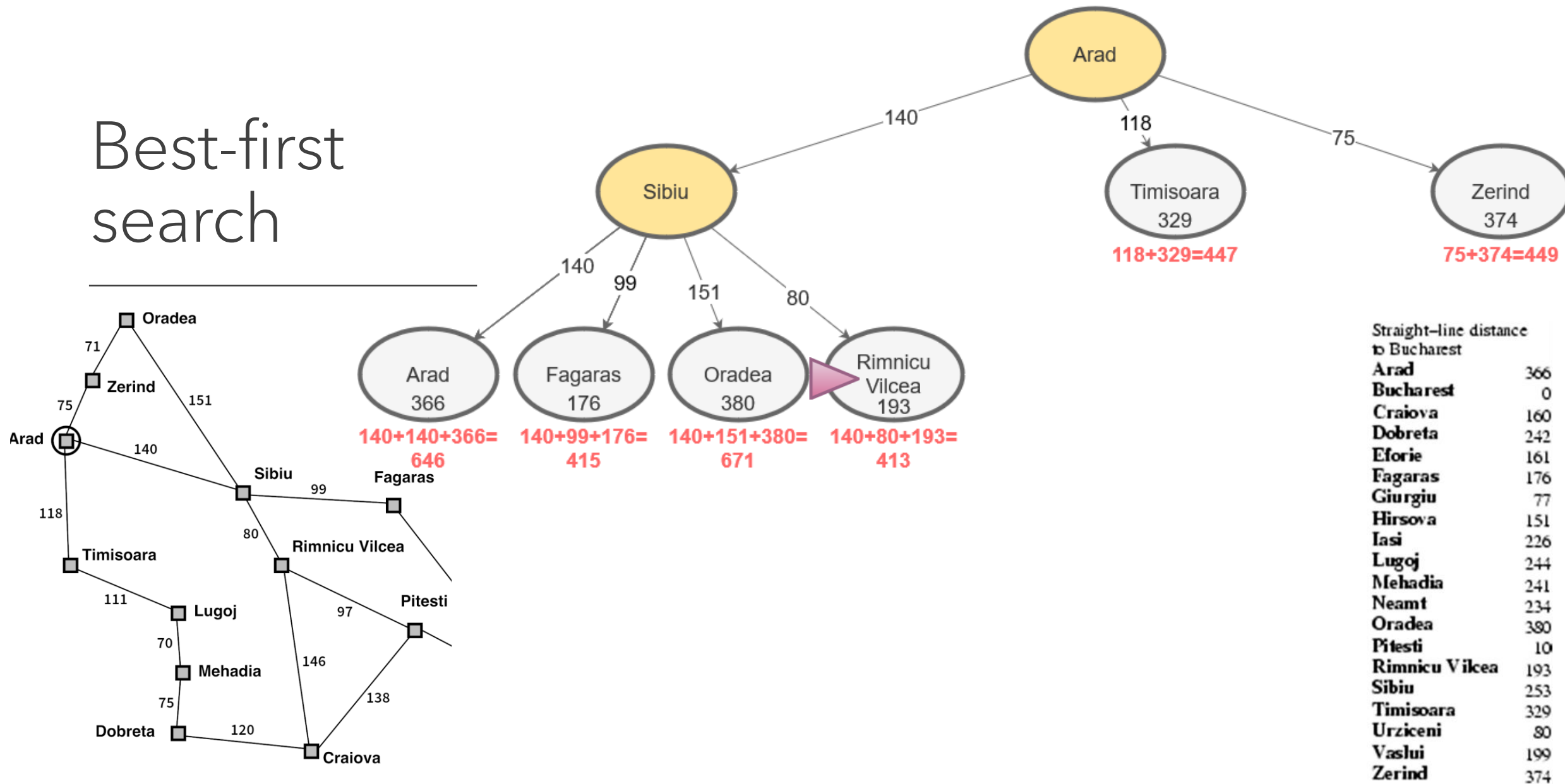
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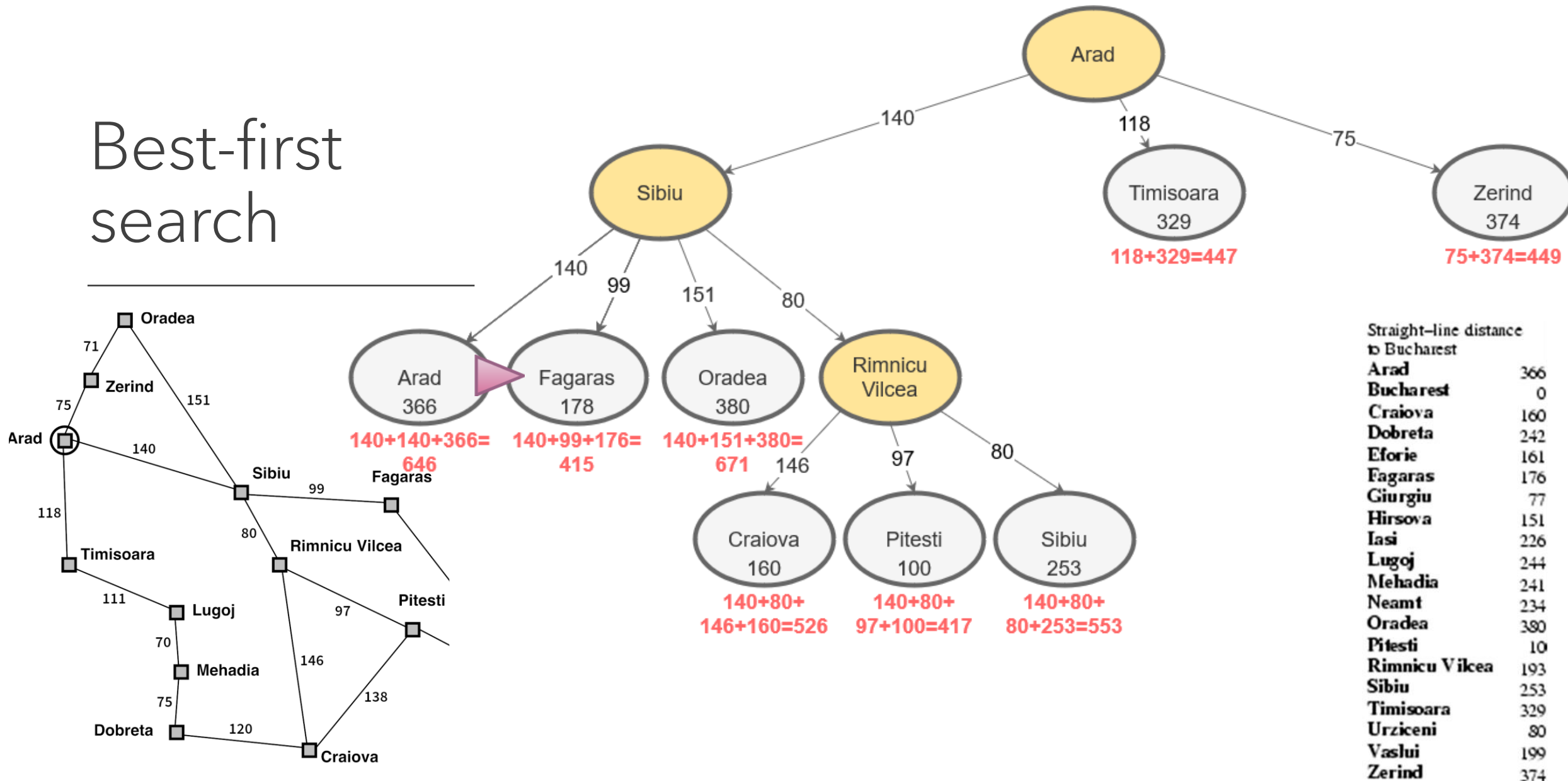


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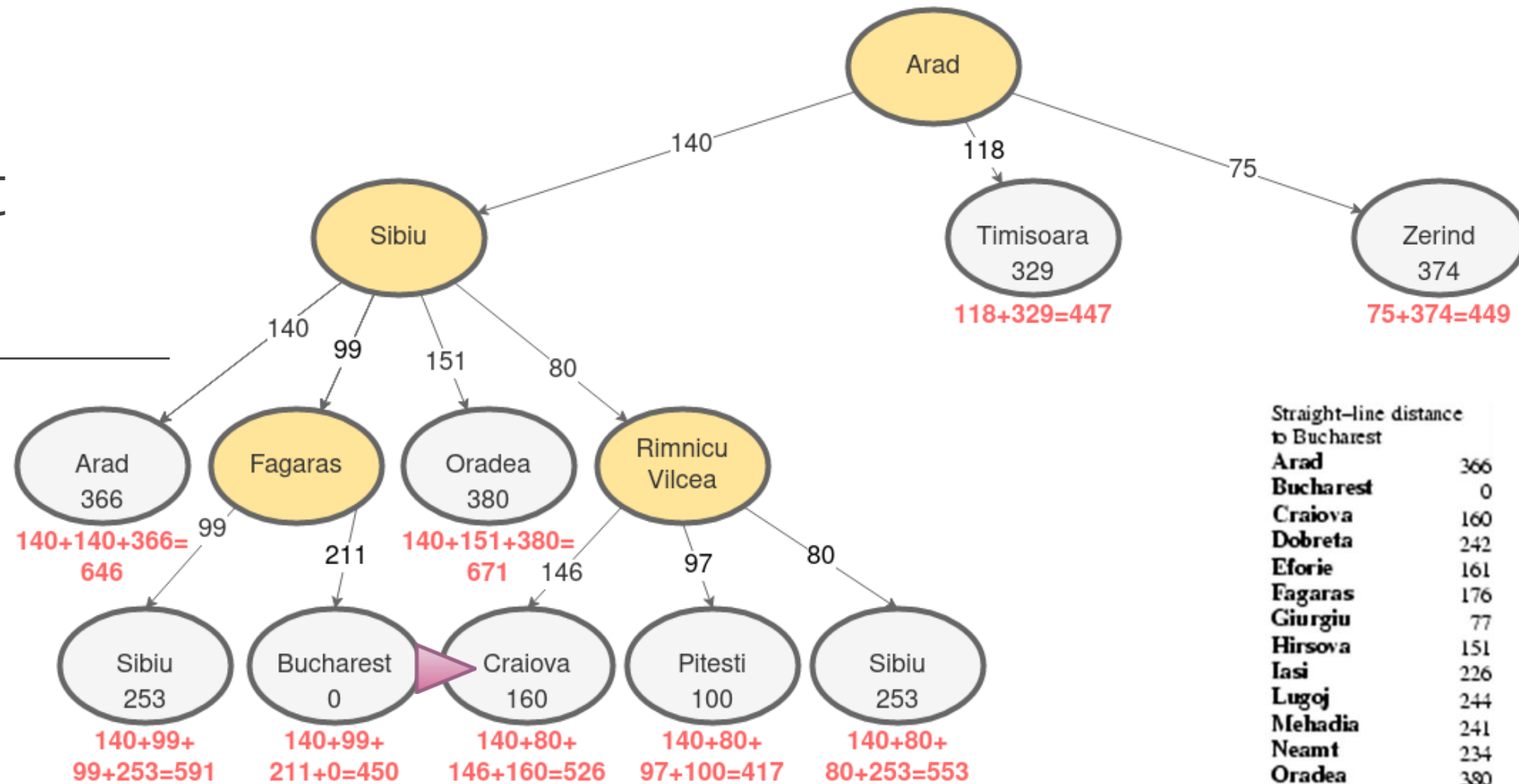
# Best-first search



# Best-first search

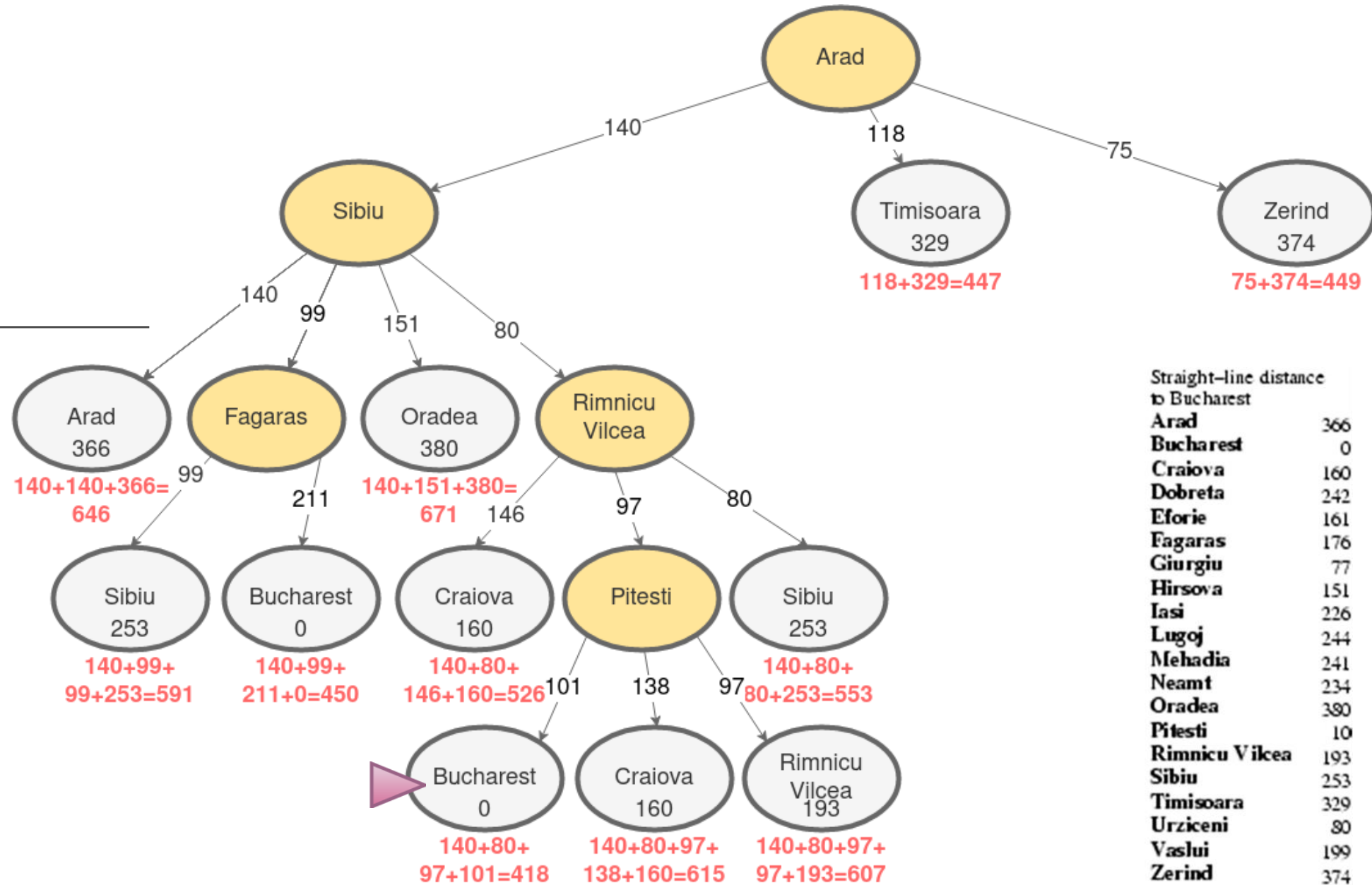


# Best-first search



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# Best-first search



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# Heuristics

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# Admissible heuristics

---

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

$$h(n) \leq \mathbf{h^*(n)}$$

where  $\mathbf{h^*(n)}$  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

Example:  $h_{SLD}(n)$  (never overestimates the actual road distance)



# Admissible heuristic = optimal A\*

---

$h(n)$  never overestimates the cost to reach the goal

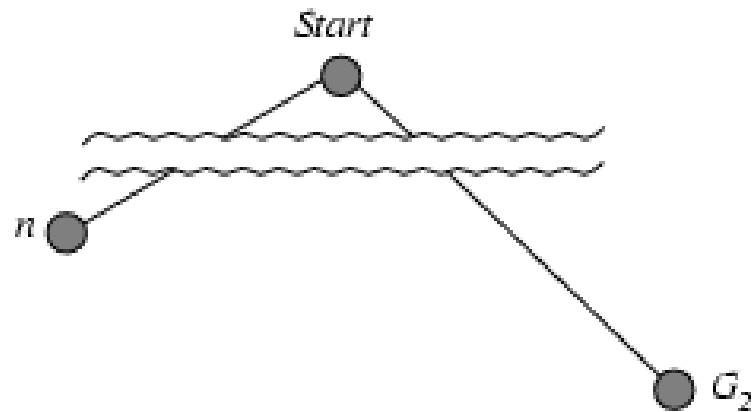
Thus,  $f(n) = g(n) + h(n)$  never overestimates the true cost of a solution

## THEOREM

If  $h(n)$  is admissible, A\* using **TREE-SEARCH** is optimal

# Proof: Optimality of $A^*$

---



Suppose some **suboptimal** goal  $G_2$  has been generated and is in the frontier.

Let  $n$  be an unexpanded node in the frontier such that  $n$  is on a **shortest path** to an **optimal** goal  $G$ .

$$f(G_2) = g(G_2)$$

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

$$g(G_2) > g(G)$$

$$\text{since } G_2 \text{ is suboptimal}$$

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

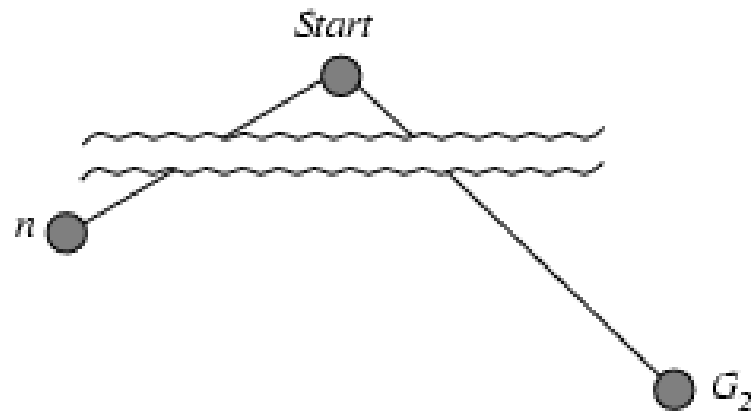
$$\text{since } h(G) = 0$$

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

$$\text{from above}$$

# Proof: Optimality of $A^*$

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Suppose some **suboptimal** goal  $G_2$  has been generated and is in the frontier.

Let  $n$  be an unexpanded node in the frontier such that  $n$  is on a **shortest path** to an **optimal** goal  $G$ .

$$f(G) < f(G_2) \quad \text{from above}$$

$$h(n) \leq h^*(n) \quad \text{since } h \text{ is admissible}$$

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

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

*Hence  $f(n) < f(G_2)$ , and  $A^*$  will never select  $G_2$  for expansion*

# Consistent heuristics

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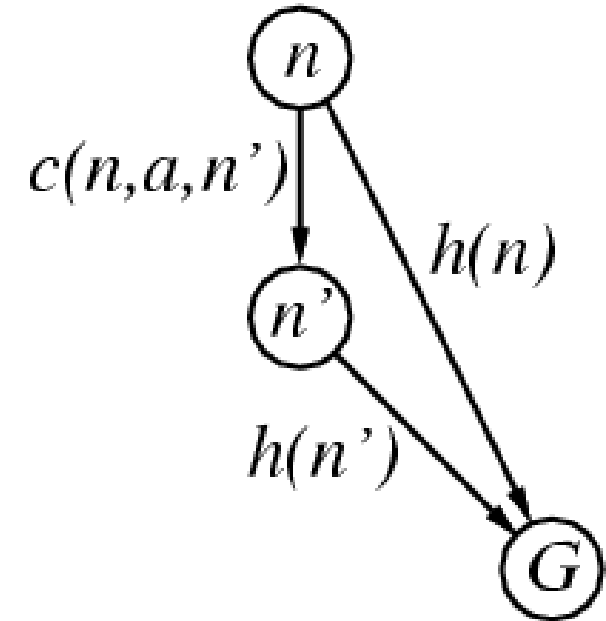
A heuristic  $h(n)$  is **consistent** if for every node  $n$ , every successor  $n'$  of  $n$  generated by any action  $a$ ,

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

If  $h$  is consistent, we have

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

i.e.,  $f(n)$  is non-decreasing along any path.



THEOREM

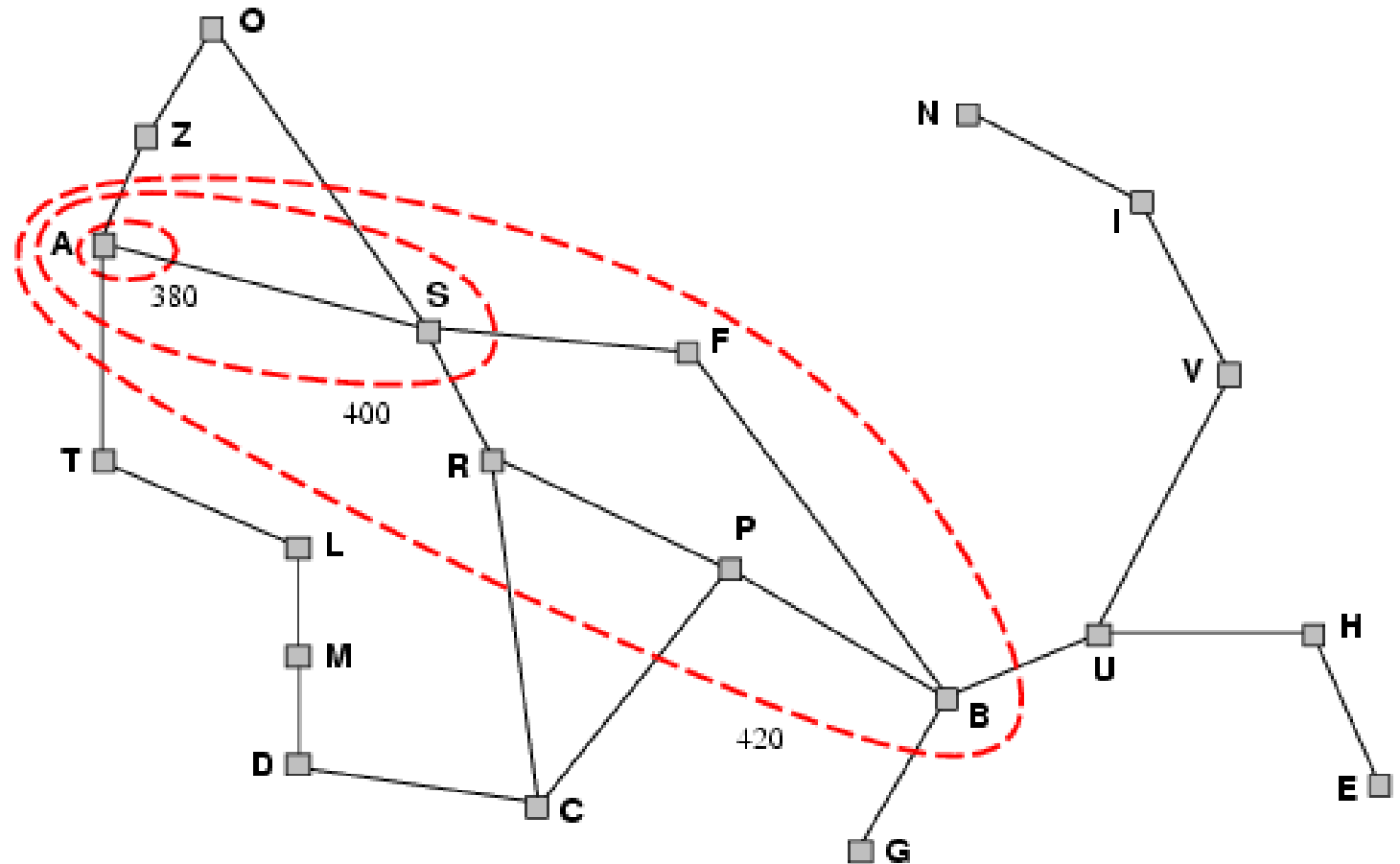
If  $h(n)$  is consistent,  $A^*$  using **GRAPH-SEARCH** is **optimal**

# Optimality of $A^*$

$A^*$  expands nodes in order of increasing  $f$  value

Gradually adds " $f$ -contours" of nodes

Contour  $i$  has all nodes with  $f=f_i$ , where  $f_i < f_{i+1}$



# Properties of $A^*$

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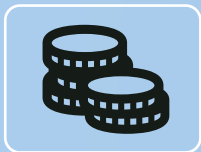
**Complete?**



**Time complexity?**



**Space complexity?**



**Optimal?**

# Properties of $A^*$

---



## **Complete?**

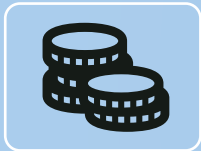
Yes (unless there are infinitely many nodes with  $f \leq f(G)$ )



## **Time complexity?**



## **Space complexity?**



## **Optimal?**

# Properties of A\*

---



## **Complete?**

Yes (unless there are infinitely many nodes with  $f \leq f(G)$ )



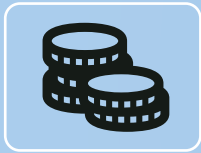
## **Time complexity?**

Exponential



## **Space complexity?**

Keeps all nodes in memory



## **Optimal?**



# Properties of A\*

---



## **Complete?**

Yes (unless there are infinitely many nodes with  $f \leq f(G)$ )



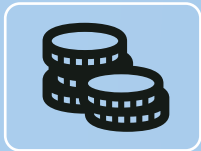
## **Time complexity?**

Exponential



## **Space complexity?**

Keeps all nodes in memory



## **Optimal?**

Yes

# Admissible heuristics

---

Example: 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)

$$h_1(S) = ?$$

$$h_2(S) = ?$$

7	2	4
5		6
8	3	1

Start State

	1	2
3	4	5
6	7	8

Goal State

# Dominance

---

If  $h_2(n) \geq h_1(n)$  for all  $n$  (both admissible) then

- $h_2$  dominates  $h_1$
- $h_2$  is better for search

Typical search costs (*average number of nodes expanded*):

- $d=12$ 
  - IDS = 3,644,035 nodes
  - $A^*(h_1) = 227$  nodes
  - $A^*(h_2) = 73$  nodes
- $d=24$ 
  - IDS  $\approx$  54,000,000,000 nodes
  - $A^*(h_1) = 39,135$  nodes
  - $A^*(h_2) = 1,641$  nodes

# 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 adjacent square,

- then  $h_2(n)$  gives the shortest solution

Use relaxation to automatically generate admissible heuristics!



# Summary

---

Smart search based on **heuristic scores**.

- Best-first search
- Greedy best-first search
- A\* search
- Admissible heuristics and optimality.

# Why?

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Informed search allows us to use domain knowledge to our advantage.

Optimality over some utility can often be the top priority.

A\* is **very** popular!

- e.g. pathfinding

A\* is simple, yet very efficient.

A\* is *too* good sometimes (e.g. in games).