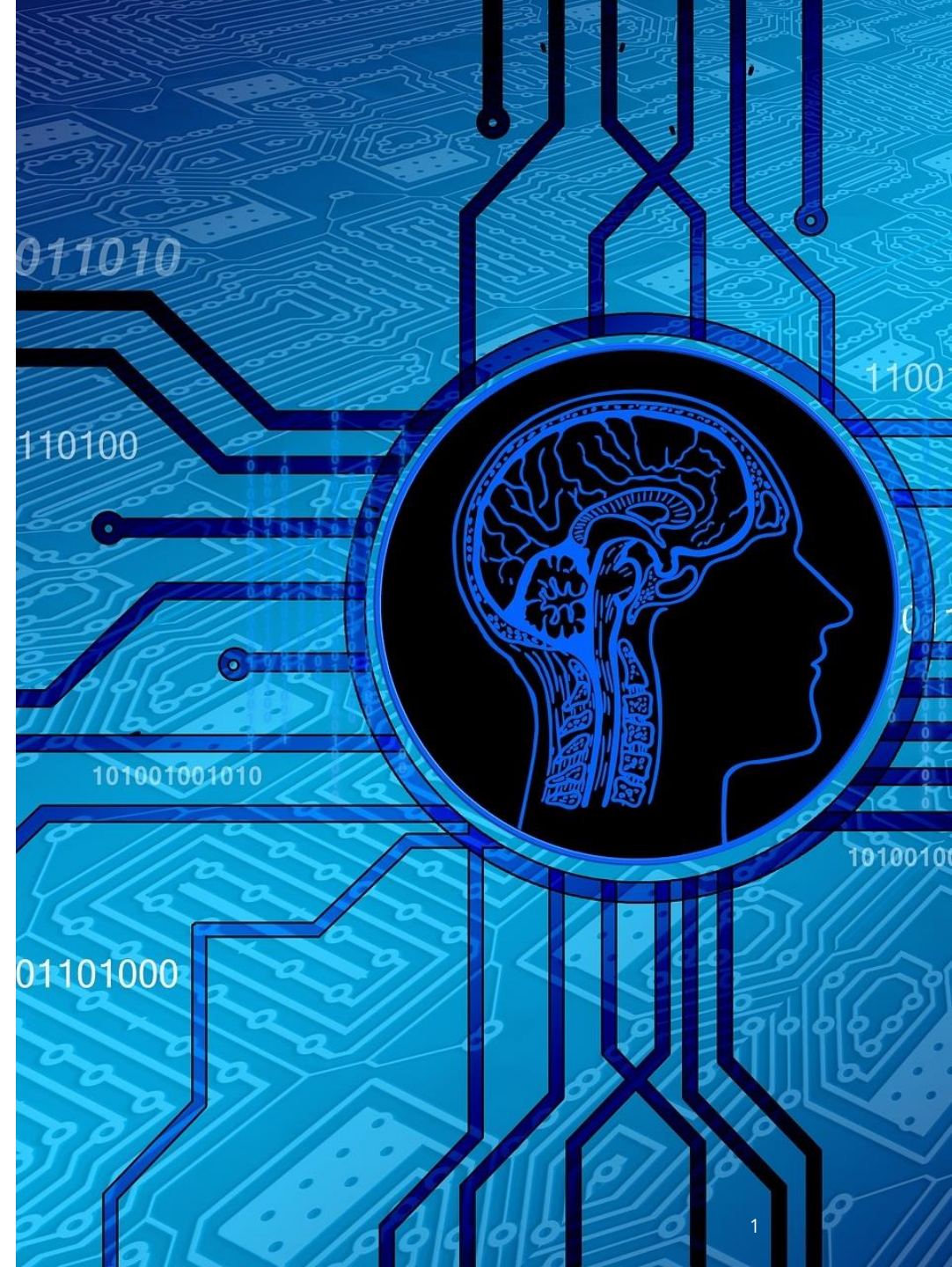


Informed Search

Informatics 2D: Reasoning and Agents
Lecture 4



Criterion	Breadth-First	Uniform-Cost	Depth-First	Depth-Limited	Iterative Deepening	Bidirectional (if applicable)
Complete?	Yes ^a	Yes ^{a,b}	No	No	Yes ^a	Yes ^{a,d}
Time	$O(b^d)$	$O(b^{1+\lceil C^*/\epsilon \rceil})$	$O(b^m)$	$O(b^l)$	$O(b^d)$	$O(b^{d/2})$
Space	$O(b^d)$	$O(b^{1+\lceil C^*/\epsilon \rceil})$	$O(bm)$	$O(bl)$	$O(bd)$	$O(b^{d/2})$
Optimal?	Yes ^c	Yes	No	No	Yes ^c	Yes ^{c,d}

Review: Summary of algorithms

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**.

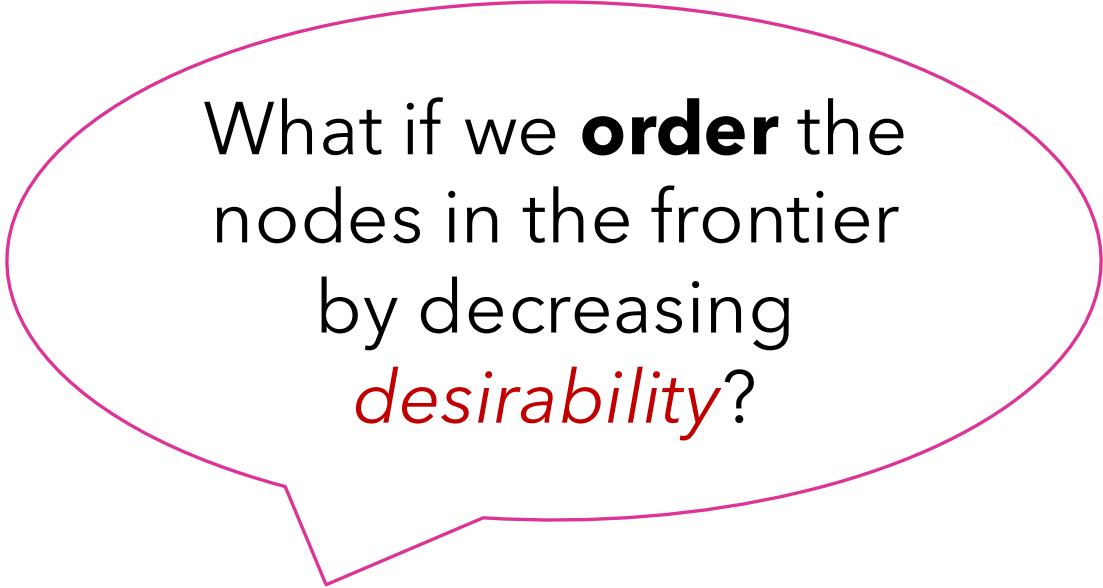
Review: Graph search

```
function GRAPH-SEARCH(problem) returns a solution, or failure
  initialize the frontier using the initial state of problem
  initialize the explored set to be empty
  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
    add the node to the explored set
    expand the chosen node, adding the resulting nodes to the frontier
      only if not in the frontier or explored set
```

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

Making search 'informed'

- Tree-Search
- Graph Search



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**.

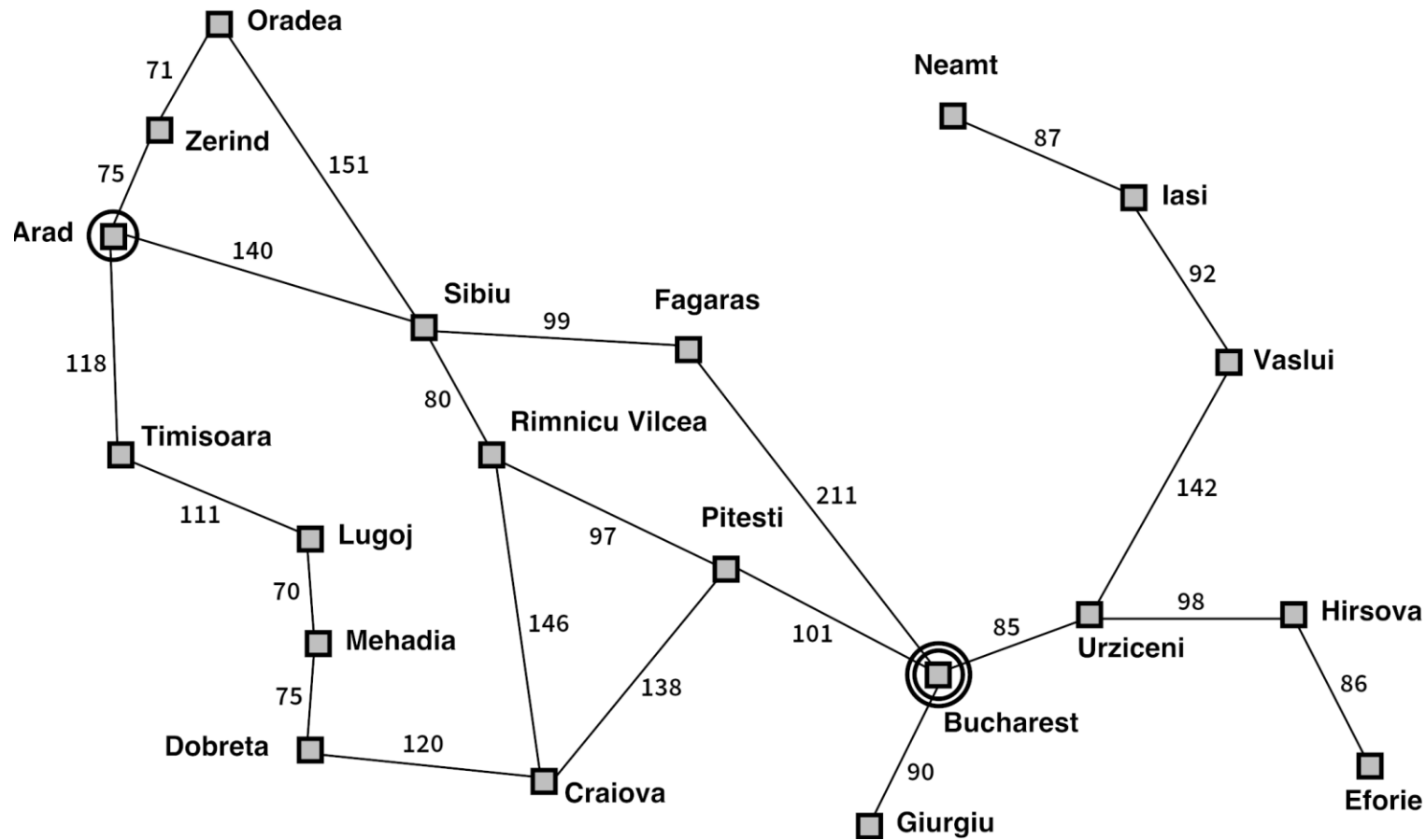
Best-first search

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

- 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 search**.
 - This may not change the worst-case complexity of the algorithm, but can help in the average case.
- We will introduce **admissibility, consistency** conditions to identify good heuristics.



Straight-line distance to Bucharest

Arad	366
Bucharest	0
Craiova	160
Dobreta	242
Eforie	161
Fagaras	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

Romania Example

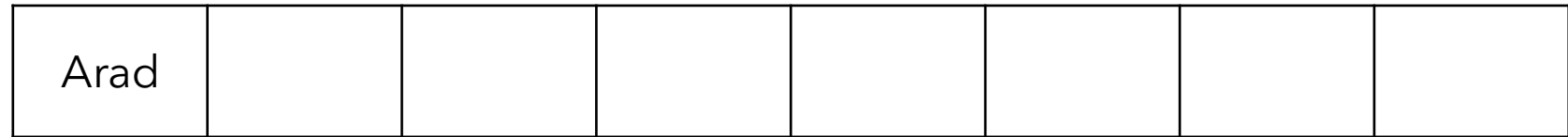
Greedy best-first search



Greedy best-first search

- 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.

Greedy best-first search



Straight-line distance

to Bucharest

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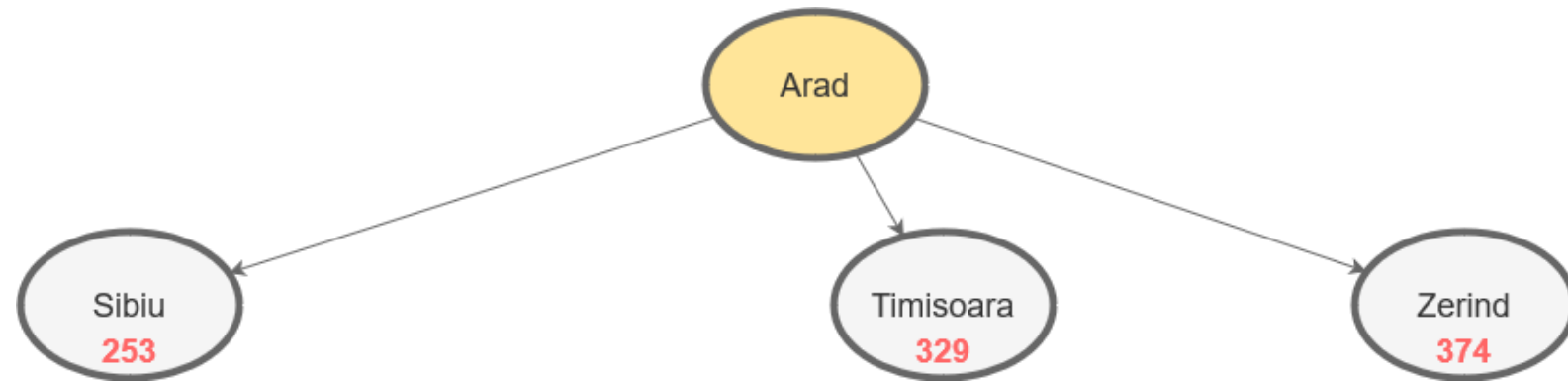
Greedy best-first search

Sibiu	Timiso-ara	Zerind					
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Straight-line distance

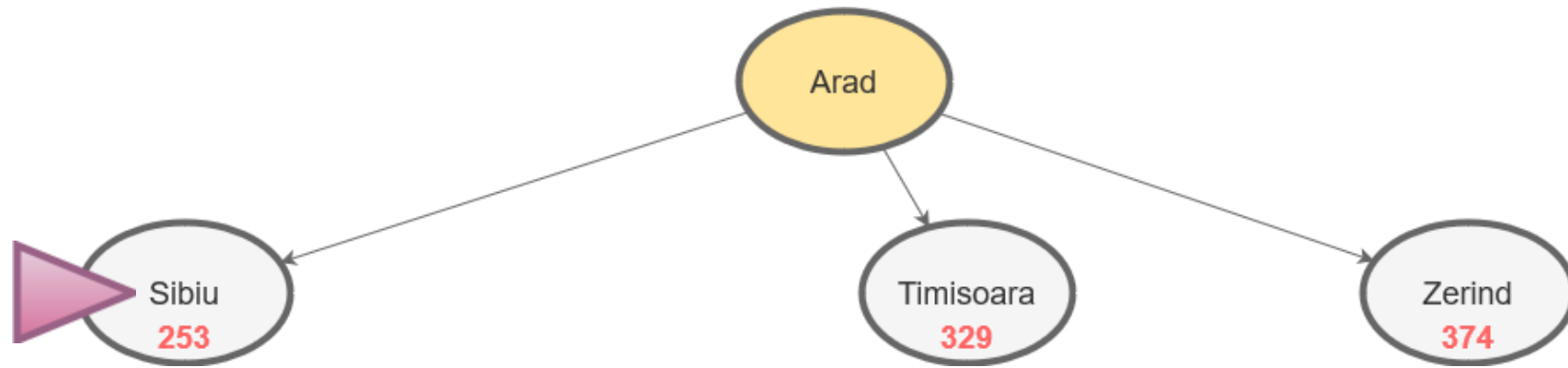
to Bucharest

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Urziceni	80
Vaslui	199
Zerind	374



Greedy best-first search

Sibiu	Timiso- ara	Zerind					
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Straight-line distance

to Bucharest

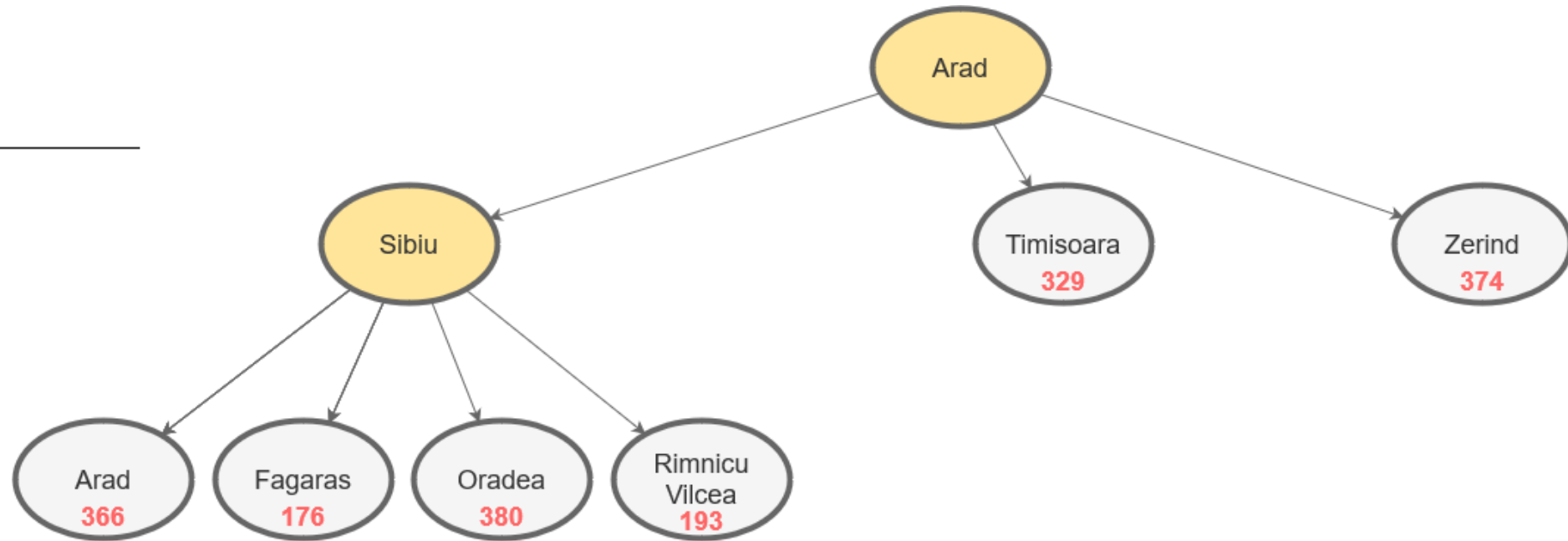
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Greedy best-first search

Timiso-ara	Zerind	Arad	Fagaras	Oradea	RV		
------------	--------	------	---------	--------	----	--	--

Straight-line distance

to Bucharest	
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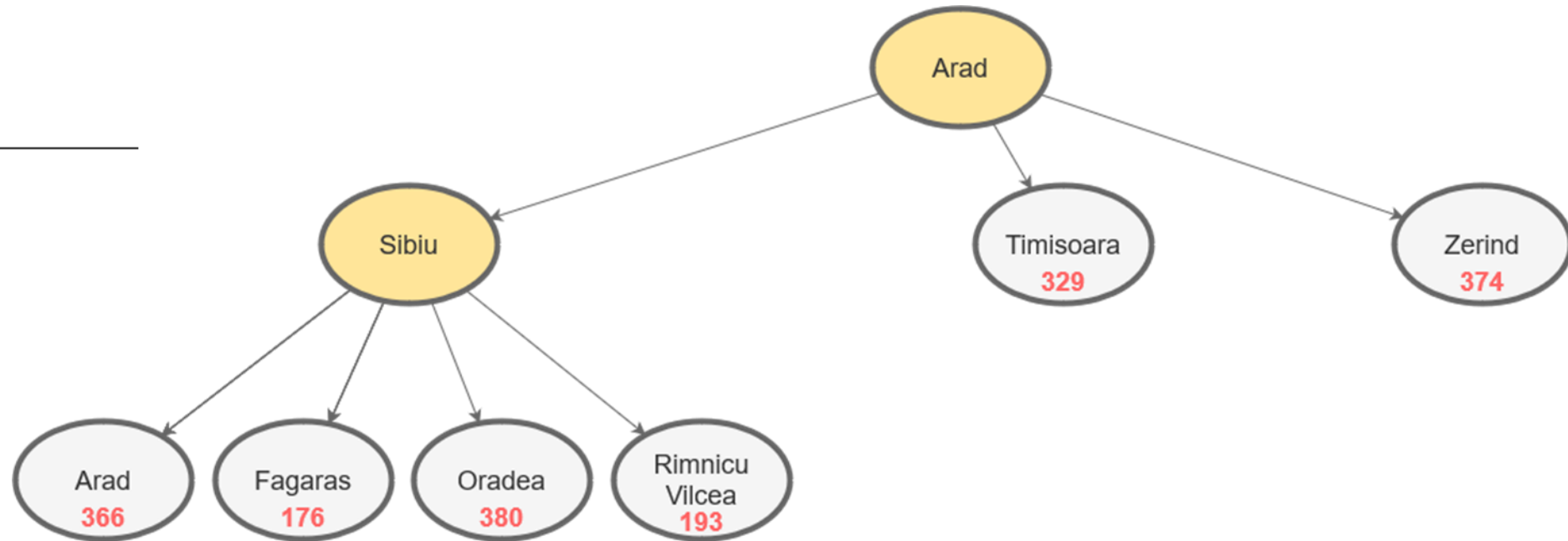


Greedy best-first search

Fagaras	RV	Timiso- ara	Arad	Zerind	Oradea		
---------	----	----------------	------	--------	--------	--	--

Straight-line distance to Bucharest

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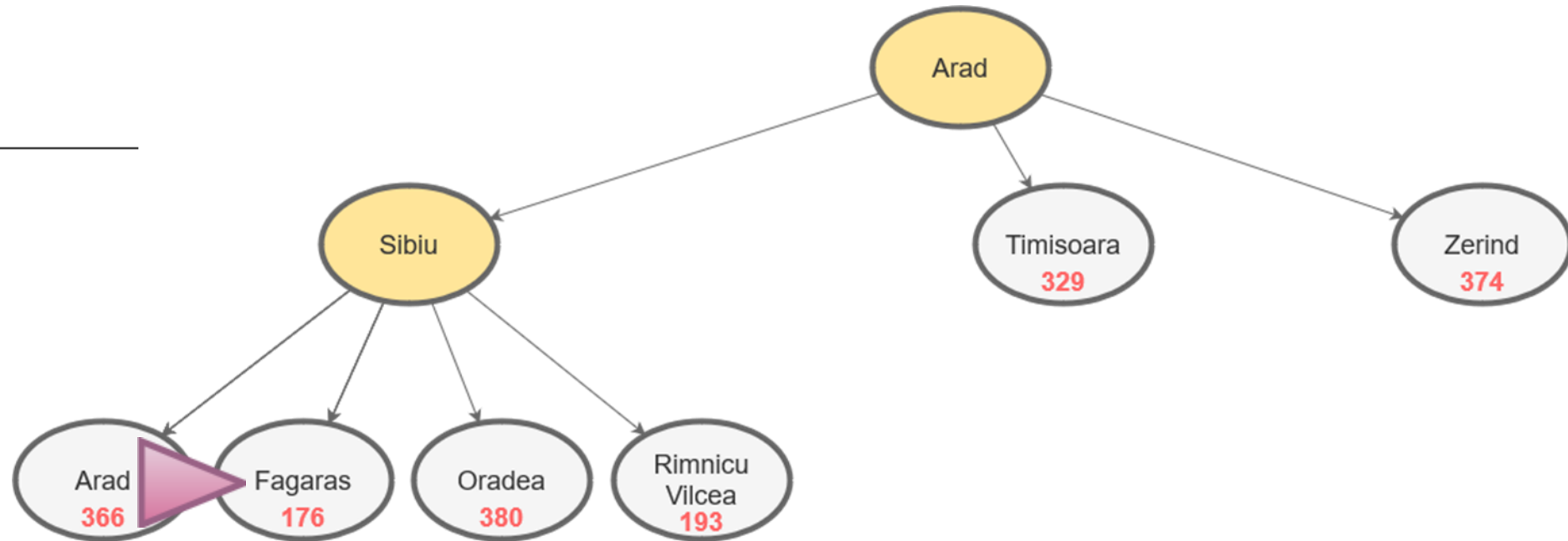
Greedy best-first search

Fagaras	RV	Timiso- ara	Arad	Zerind	Oradea		
---------	----	----------------	------	--------	--------	--	--

Straight-line distance

to Bucharest

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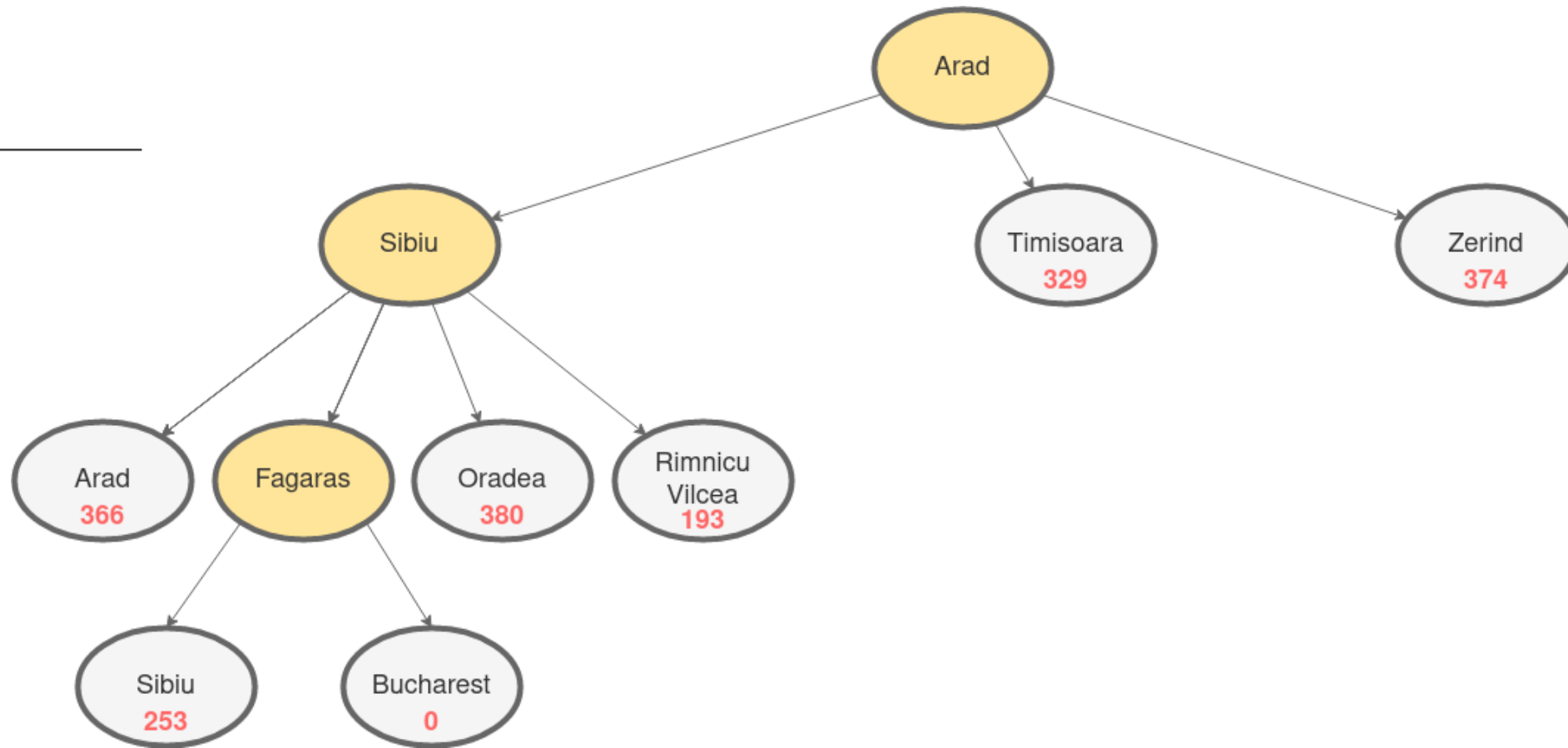
Greedy best-first search

Fagaras	RV	Timiso- ara	Arad	Zerind	Oradea	Sibiu	Bucha- rest
---------	----	----------------	------	--------	--------	-------	----------------

Straight-line distance

to Bucharest

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Properties of best-first search



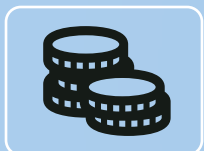
Complete?



Time complexity?



Space complexity?



Optimal?

Properties of best-first search



Complete?

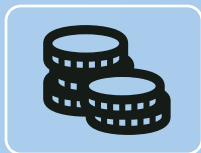
No! Can get stuck in loops.



Time complexity?



Space complexity?



Optimal?

Properties of best-first search



Complete?

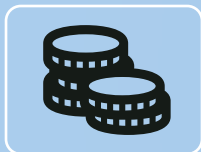
No! Can get stuck in loops.



Time complexity?



Space complexity?



Optimal?

GRAPH-SEARCH
is complete in
finite spaces.

Properties of best-first search



Complete?

No! Can get stuck in loops.

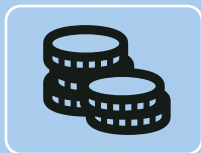


Time complexity?

$O(b^m)$ for tree version



Space complexity?



Optimal?

Properties of best-first search



Complete?

No! Can get stuck in loops.

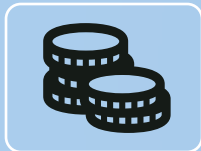


Time complexity?

$O(b^m)$ for tree version



Space complexity?



Optimal?

A good heuristic
can lead to
dramatic
improvement!

Properties of best-first search



Complete?

No! Can get stuck in loops.



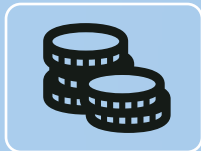
Time complexity?

$O(b^m)$ 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.



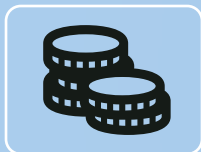
Time complexity?

$O(b^m)$ for tree version



Space complexity?

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



Optimal?

No

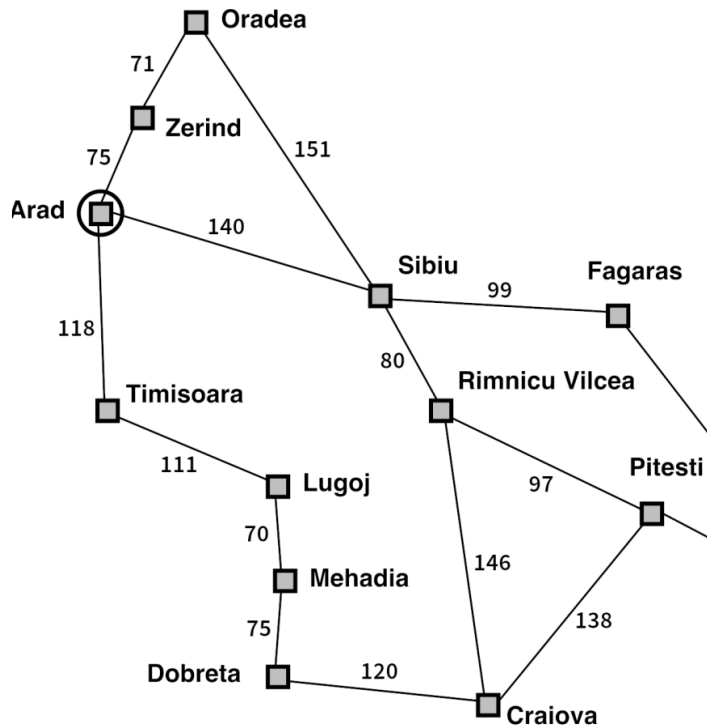
A* Search

A* search

- 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**

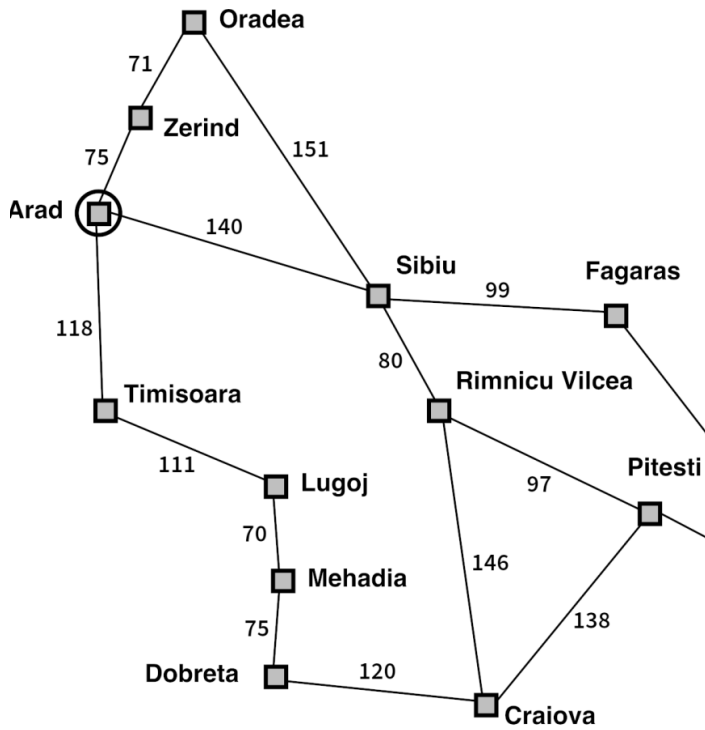
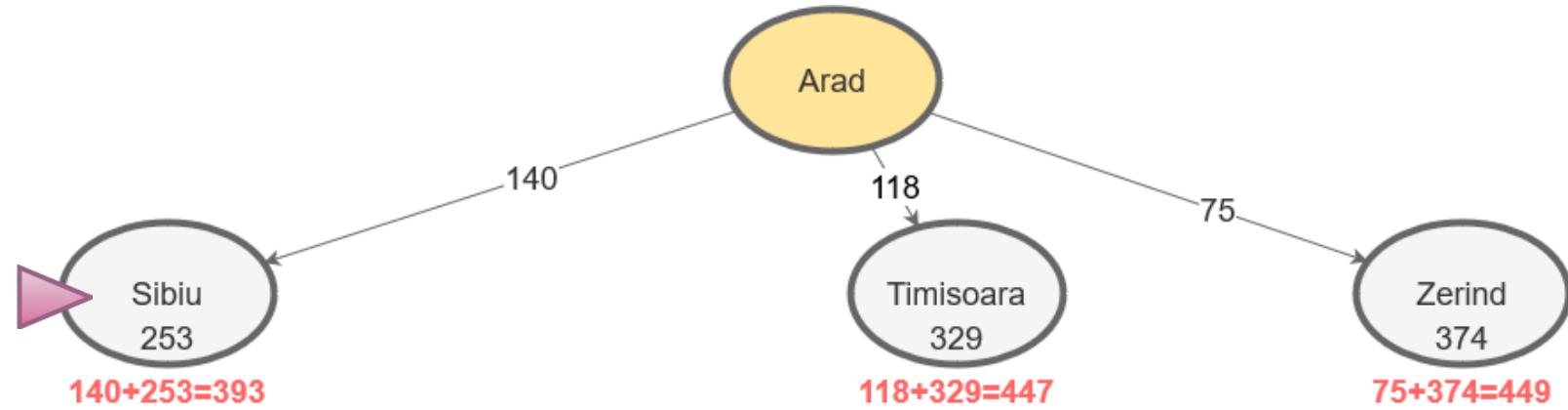


A* search



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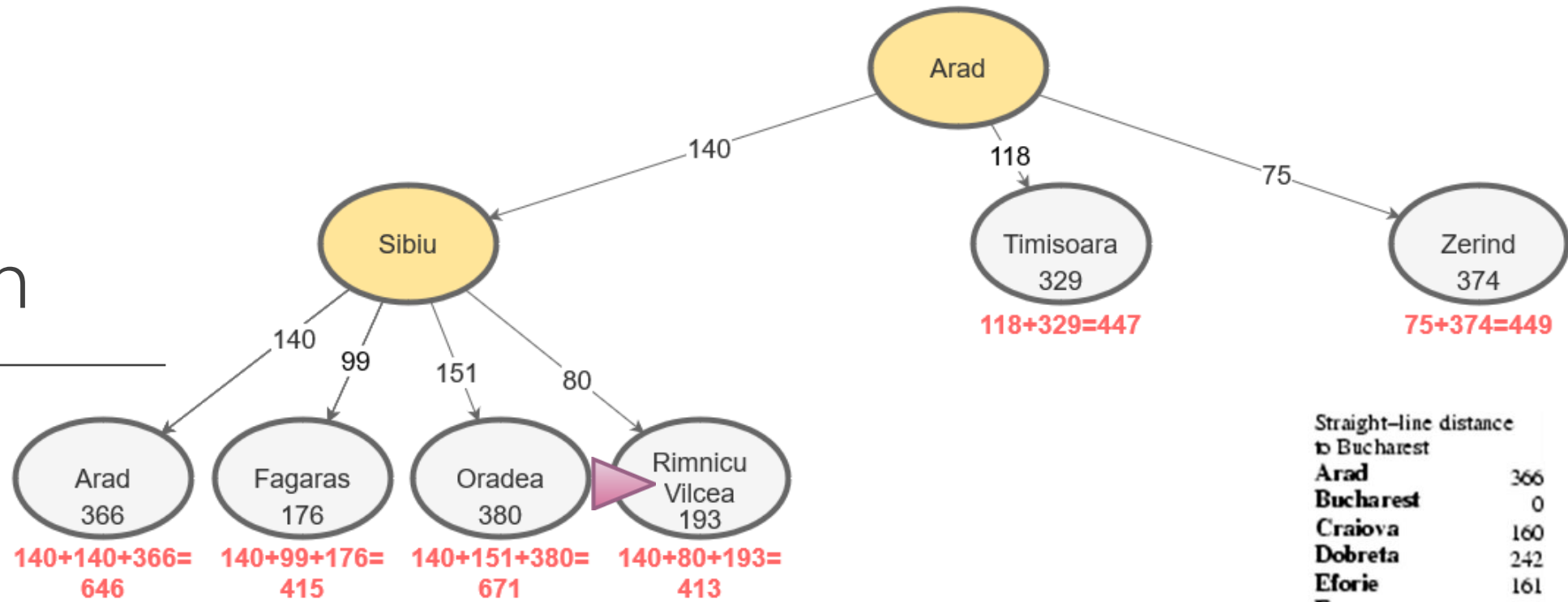
A* search



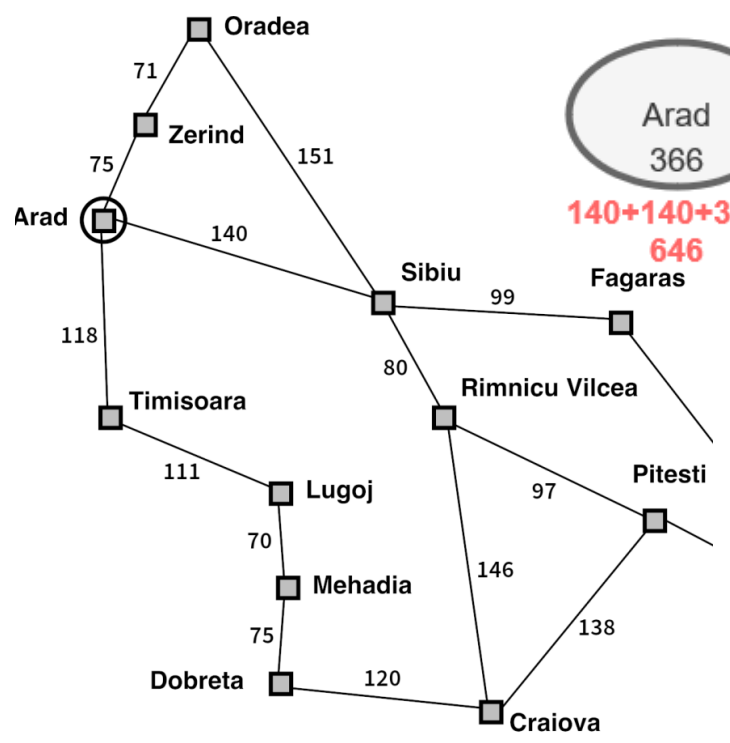
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A* search

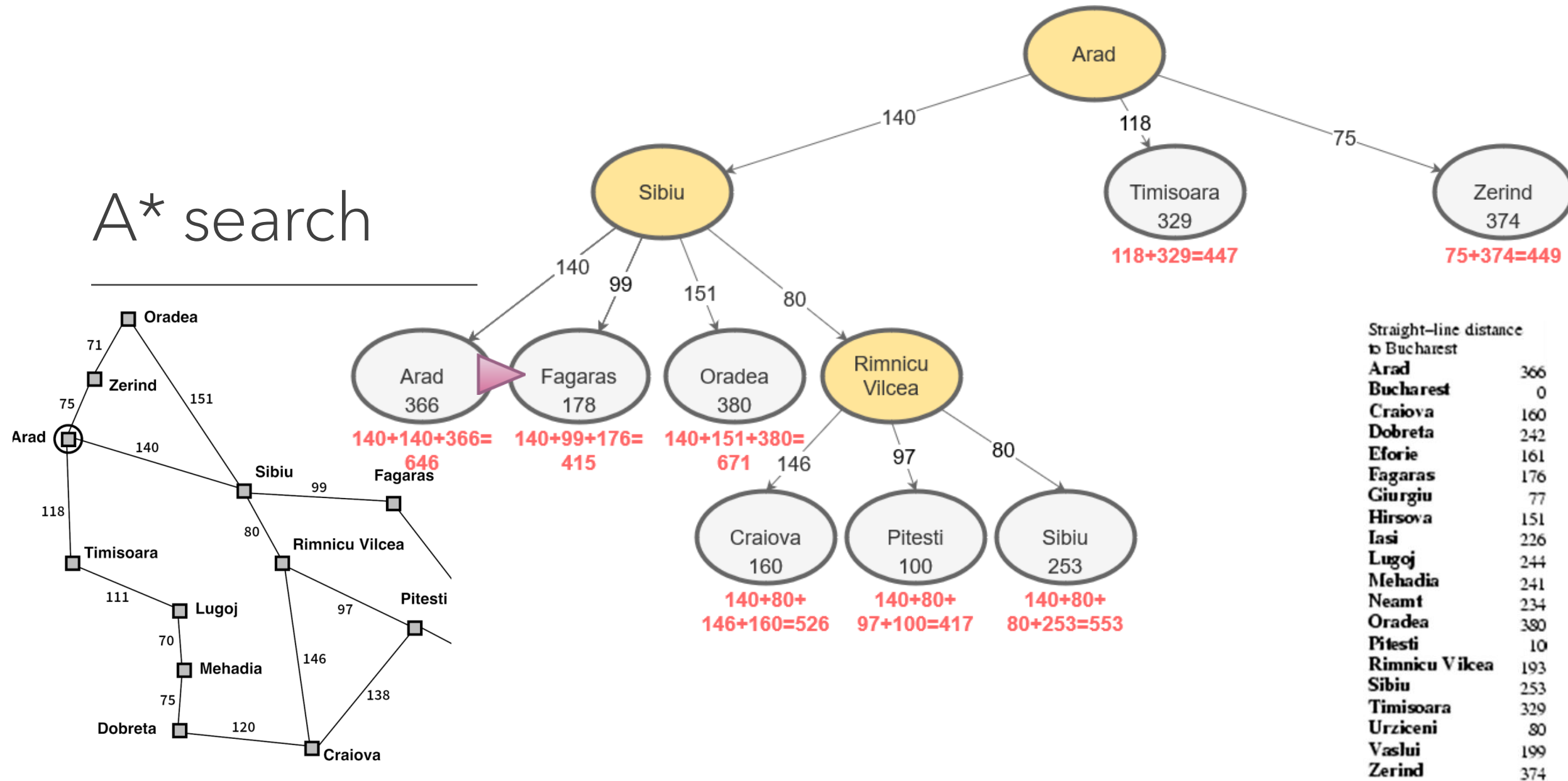


$140+140+366=646$
 $140+99+176=415$
 $140+151+380=671$
 $140+80+193=413$

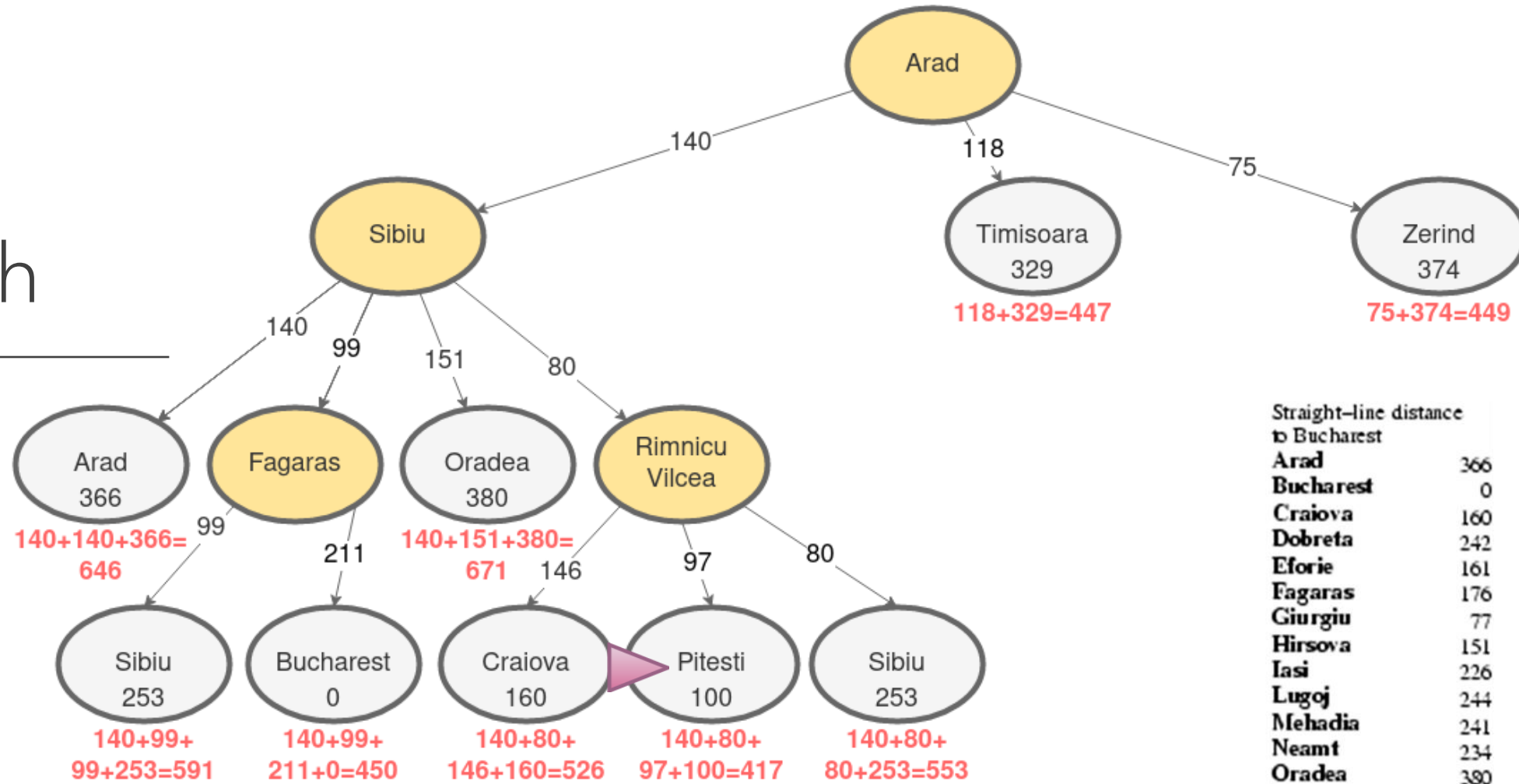


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A* search

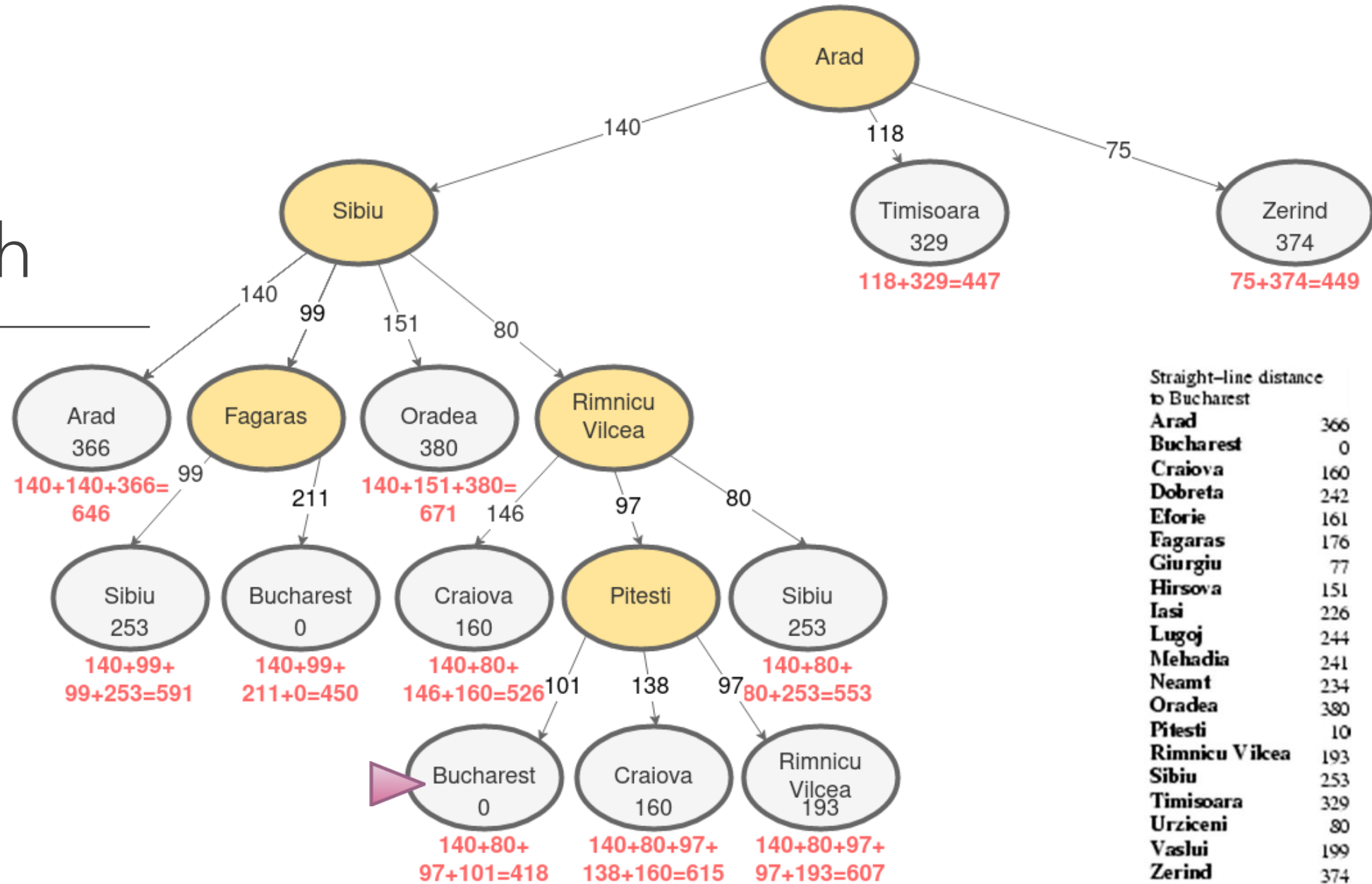


A* search



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A* search



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Heuristics

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*

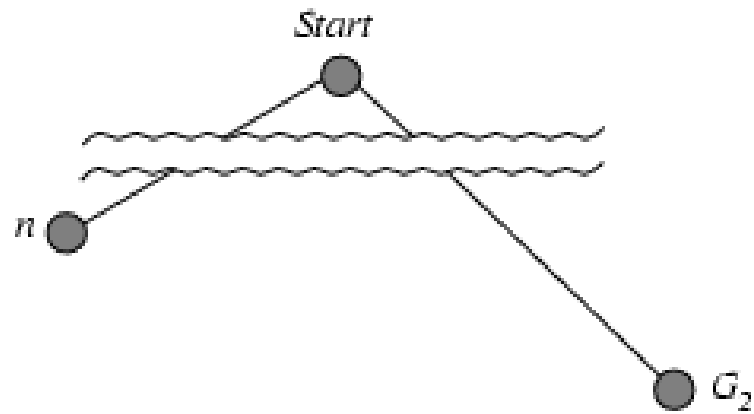
$h(n)$ never overestimates the cost to reach the goal

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 .

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

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

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

since G_2 is suboptimal

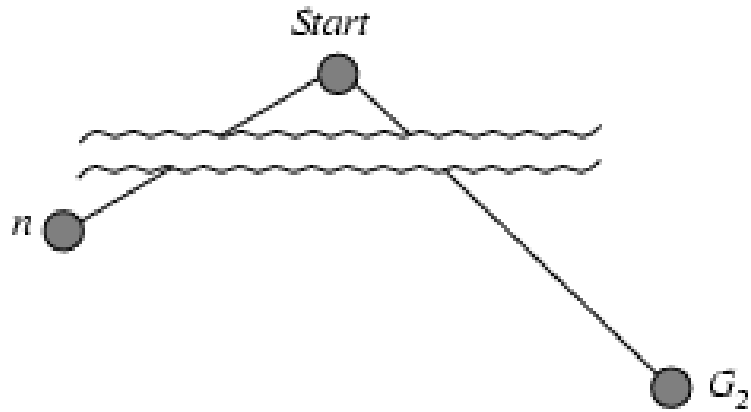
$$f(G) = g(G)$$

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

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

from above

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 .

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

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

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

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

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

Consistent heuristics

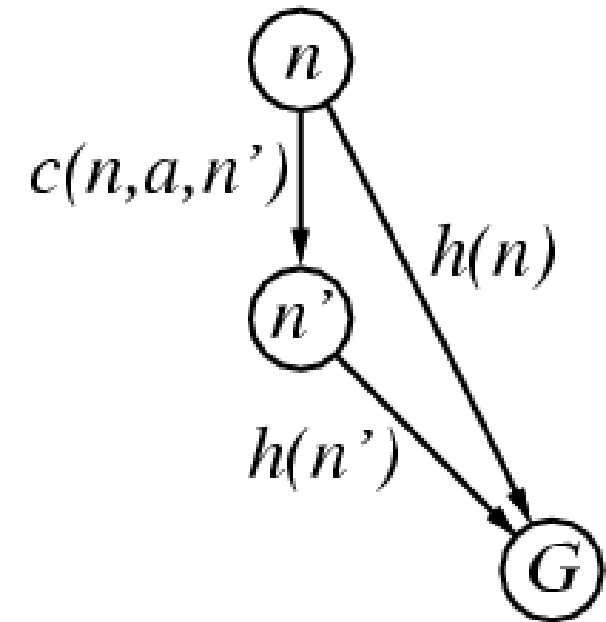
A heuristic $h(n)$ is **consistent** if for every node n , every successor n' of n generated by any action a ,

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

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

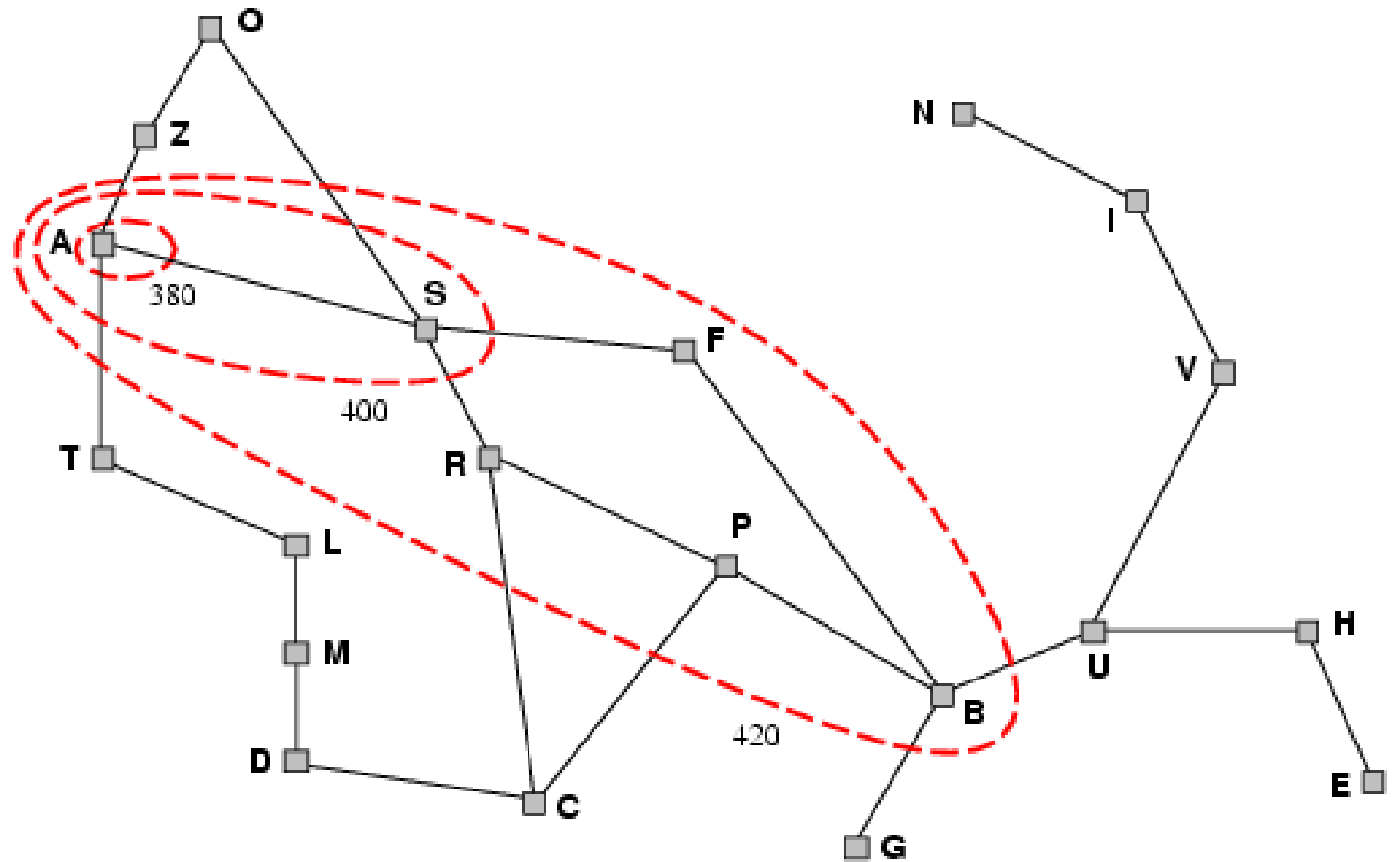
If $h(n)$ is consistent, A^* using **GRAPH-SEARCH** is **optimal**

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^*



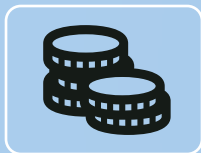
Complete?



Time complexity?



Space complexity?



Optimal?

Properties of A*



Complete?

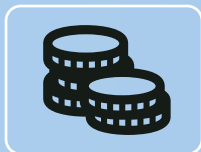
Yes (unless there are infinitely many nodes with $f \leq f(G)$)



Time complexity?



Space complexity?



Optimal?

Properties of A*



Complete?

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



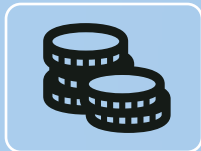
Time complexity?

Exponential



Space complexity?

Keeps all nodes in memory



Optimal?

Properties of A*



Complete?

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



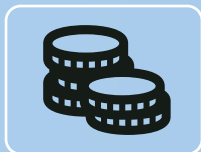
Time complexity?

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

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!



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

Summary

Smart search based on **heuristic scores**.

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

Why?

- 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).