

Princess Nora University
Faculty of Computer & Information Systems



جامعة الأميرة نورة بنت عبد الرحمن
Princess Nora Bint Abdul Rahman University



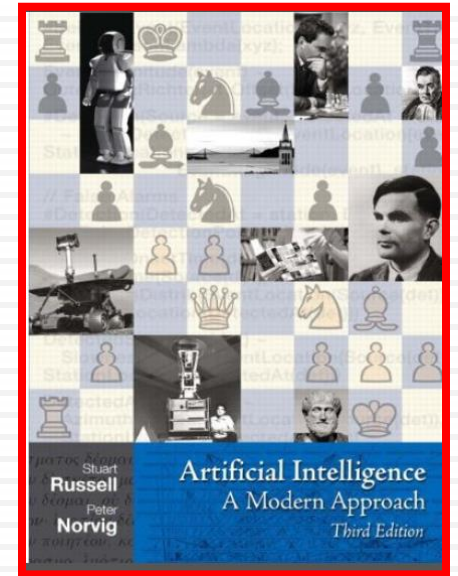
ARTIFICIAL INTELLIGENCE

(CS 370D)

Computer Science
Department



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(CHAPTER-3-PART3)

PROBLEM SOLVING AND SEARCH



Searching algorithm

Uninformed Search Algorithms(Blind Search)

- 3.1 Breadth first Search
- 3.2 Depth First Search
- 3.3 Depth limited Search
- 3.4 Iterative Deeping Search
- 3.5 Bidirectional Search

Informed Search (Heuristic Search)

- Best First Search
- Greedy Search
- Perfect Information Search
- A* Search
- Iterative Deepening A* Search
- A* with PathMax





informed Search Algorithms (Heuristic Search)

1. Best First Search
2. Greedy Search
3. Perfect Information Search
4. A* Search
5. Iterative Deepening A* Search
6. A* with PathMax





Uninformed **versus** Informed

Uninformed search

- does not have any additional information on the **quality** of states.
- So, it is impossible to determine **which state is the better than others**. As a result, search efficiency depends only on the structure of a state space

Informed search

- heuristically informed search uses a certain kind of information about states in order to guide search along **promising** branches within a state space.
- Using problem specific knowledge as hints to guide the search.





Uninformed **versus** Informed (cont)

Uninformed search

➤ look for solutions by **systematically** generating new states and checking each of them against the goal.

1. It is very **inefficient** in most cases.
2. Most successor states are “obviously” a bad choice.
3. Such strategies do not use problem-specific knowledge

Informed search

1. They are almost always **more efficient** than uninformed strategies.
2. May reduce time and space complexities.
3. Evaluation function $f(n)$ measures distance to the goal.
4. Order nodes in **Frontier** according to $f(n)$ and decide which node to expand next.



Informed search & Exploration

- *Modified version from blind search algorithm*
 1. *Greedy best first search*
 2. *A* and its relatives*

- *The family of local search includes methods*
 1. inspired by statistical physics [*simulated annealing*]
 2. evolutionary biology [*genetic algorithms*]
 3. *online search* [in which agent is faced with state space that is completely unknown]





Informed search & Exploration

Best first search

Main idea: use an evaluation function $f(n)$ for each node

Implementation:

- Order the nodes in **Frontier** in decreasing order of desirability (from low $f(n)$ which means high desirability to high $f(n)$ which means low desirability.)
- There is a whole family of best-first search strategies, each with a different evaluation function.

Special cases:

- Greedy best-first search.
- A* search.





Best-first search Algorithm

1-greedy best first search

➤ Tries to expand the node that is closest to the goal

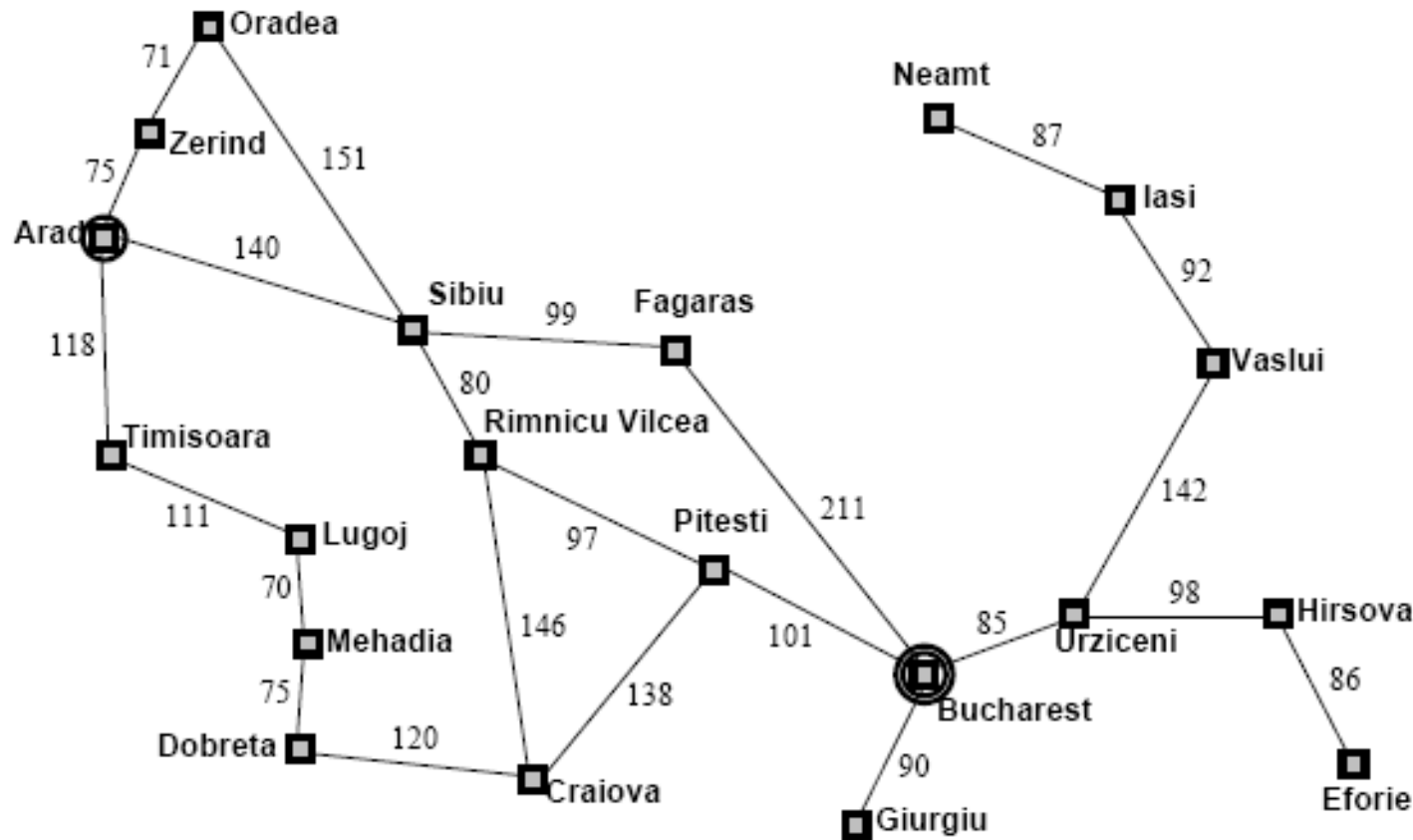
- Use straight line distance **ex : $h_{SLD}(IN(Arad))=366$**

[note that the values of **fn** **h_{SLD}** cannot be computed from the problem description itself]



Greedy Search Example 1

Straight line distances between cities which are additionally provided

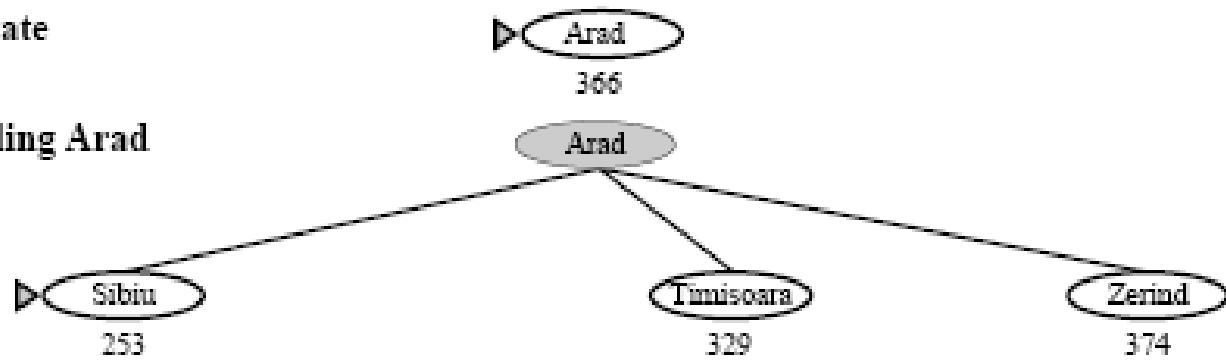


Arad	366
Bucharest	0
Craiova	160
Drobeta	242
Eforie	161
Fagaras	176
Giurgiu	77
Hirsova	151
Iasi	226
Lugoj	244
Mehadia	241
Neamt	234
Oradea	380
Pitesti	100
Rimnicu Vilcea	193
Sibiu	253
Timisoara	329
Urziceni	80
Vaslui	199
Zerind	374

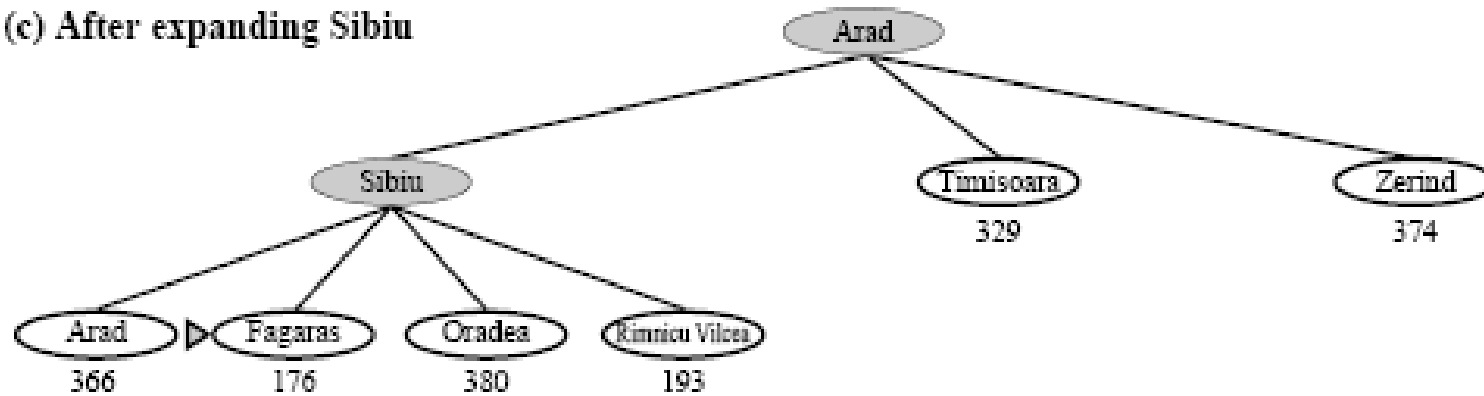
(a) The initial state



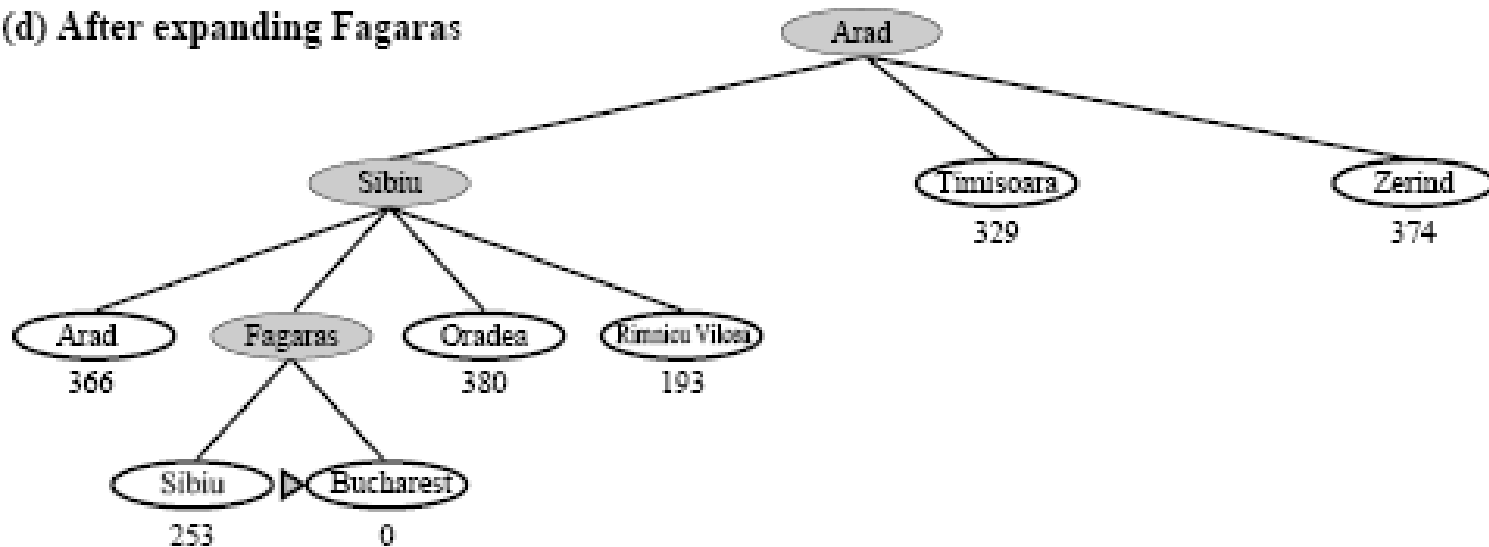
(b) After expanding Arad



(c) After expanding Sibiu



(d) After expanding Fagaras





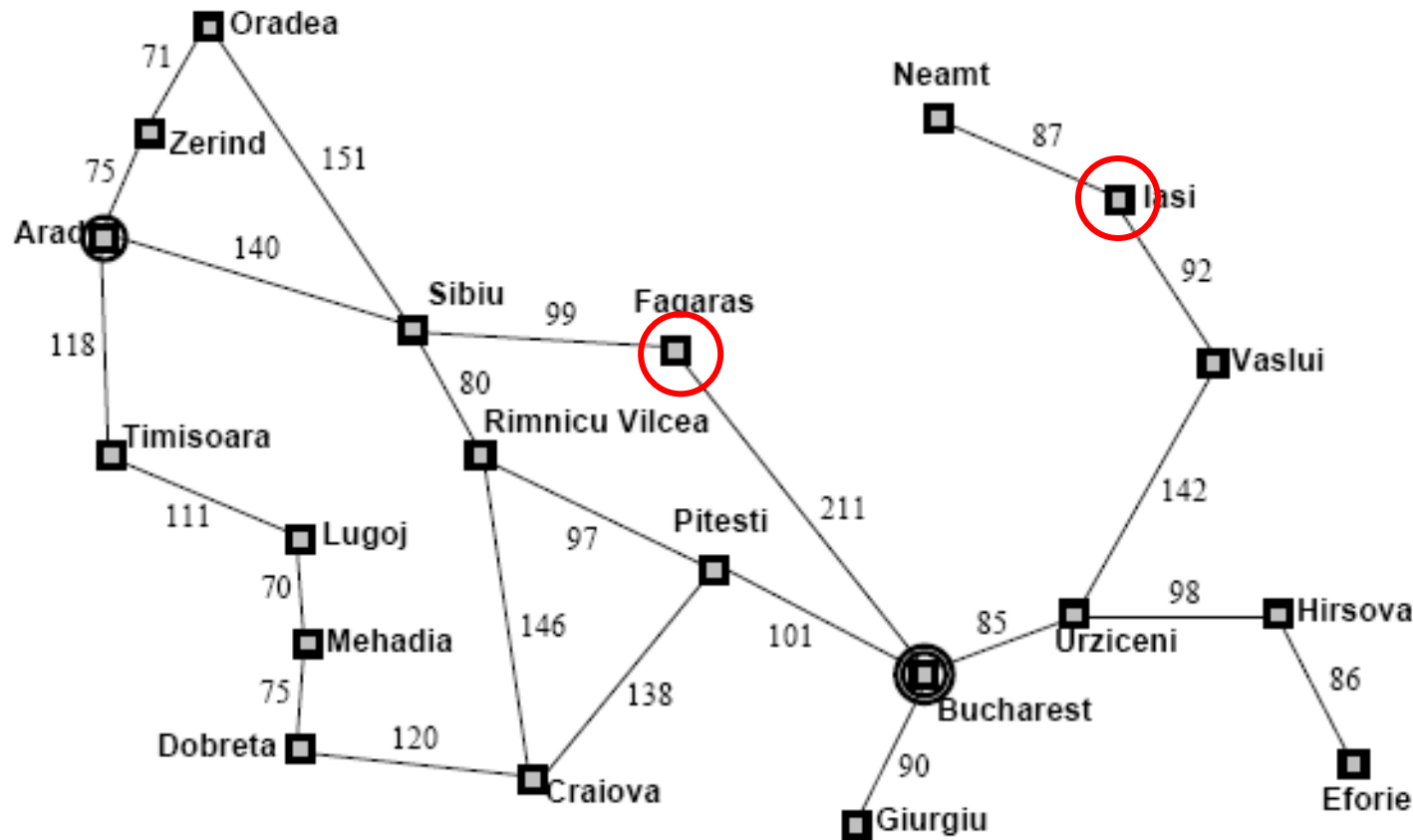
The greedy best first search using **hSLD** finds a solution without ever expanding a node that is not on solution path, hence its cost is **minimal**

This show why the algorithm is called greedy [at each step it tries to get as close to goal as it can]



Greedy Search Example 2

Straight line distances between cities which are additionally provided



Arad	366
Bucharest	0
Craiova	160
Drobeta	242
Eforie	161
Fagaras	176
Giurgiu	77
Hirsova	151
Iasi	226
Lugoj	244
Mehadia	241
Neamt	234
Oradea	380
Pitesti	100
Rimnicu Vilcea	193
Sibiu	253
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- Consider the problem of getting from **Iasi** to **Fagras**
- The heuristic suggests that **Neamt** be expanded first because it is closest to **Fagaras** but it is like dead end
- The solution is to go first to **Vaslui** a step that is actually farther from the goal according to the heuristic & then continue to **Urzicent**, **Bucharest** and **Fagaras**.
- In this case , then heuristic causes unnecessary needs to be expanded

Greedy best first search

- Resembles depth first search in the way it prefers to follow a single path all the way to goal but it will back up when it hits a dead end
- It is not optimal (greedy) and incomplete (because of backtracking)

Dr. Abeer Mahmoud
(course coordinator)





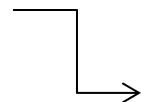
Best-first search Algorithm

2-A* best first search

- **Main idea:** avoid expanding paths that are already expensive.
- Minimizing the total estimated solution cost
- It evaluate a node by

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

Cost so far to reach n



Estimated Cost to get from n to goal

- Path cost is **g** and heuristic function is **h**
 - $f(\text{state}) = g(\text{state}) + h(\text{state})$
 - Choose smallest overall path cost (known + estimate)



A*
Search
Example

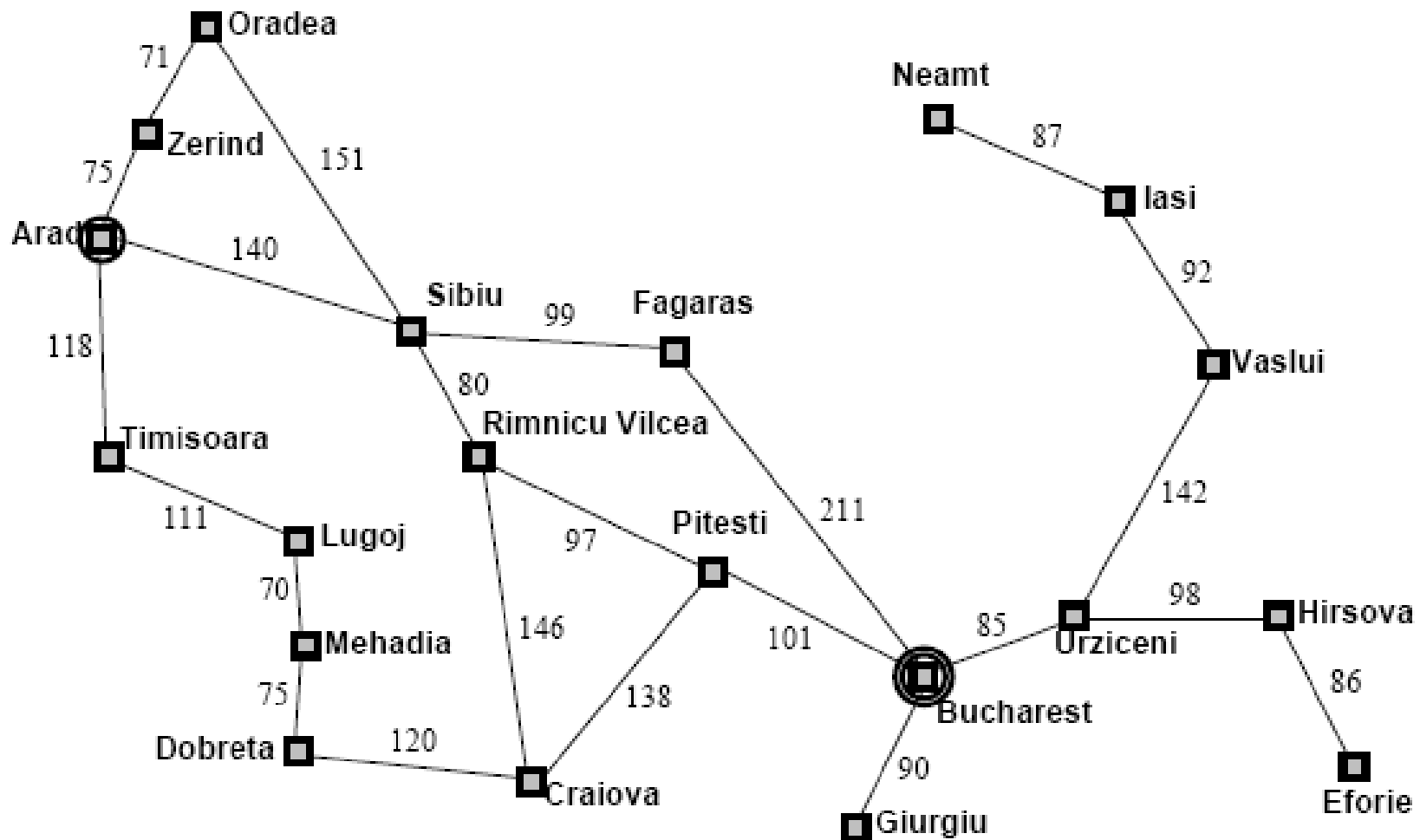
Distances to Bucharest

Town	SLD
Arad	366
Bucharest	0
Craiova	160
Dobreta	242
Eforie	161
Fagaras	178
Giurgiu	77
Hirsova	151
Iasi	226
Lugoj	244

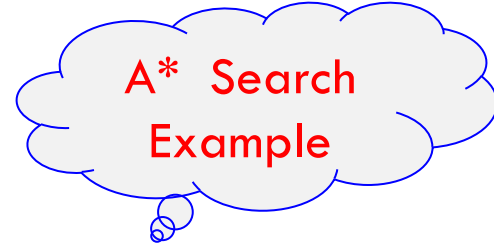
Town	SLD
Mehadai	241
Neamt	234
Oradea	380
Pitesti	98
Rimnicu	193
Sibiu	253
Timisoara	329
Urziceni	80
Vaslui	199
Zerind	374

We can use straight line distances as an admissible heuristic as they will never overestimate the cost to the goal. This is because there is no shorter distance between two cities than the straight line distance. Press space to continue with the slideshow.

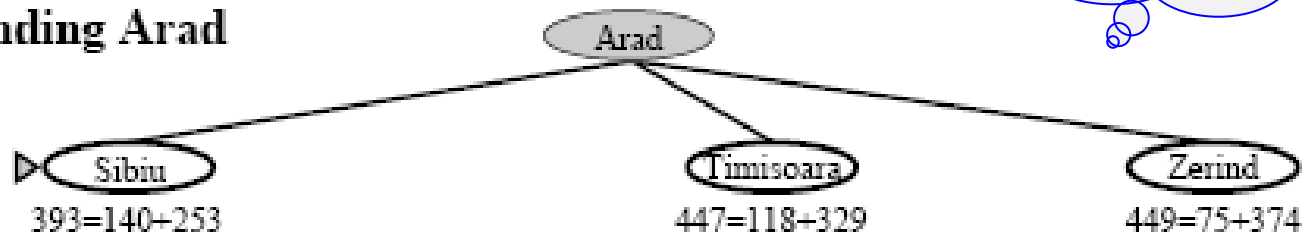
A* Search Example



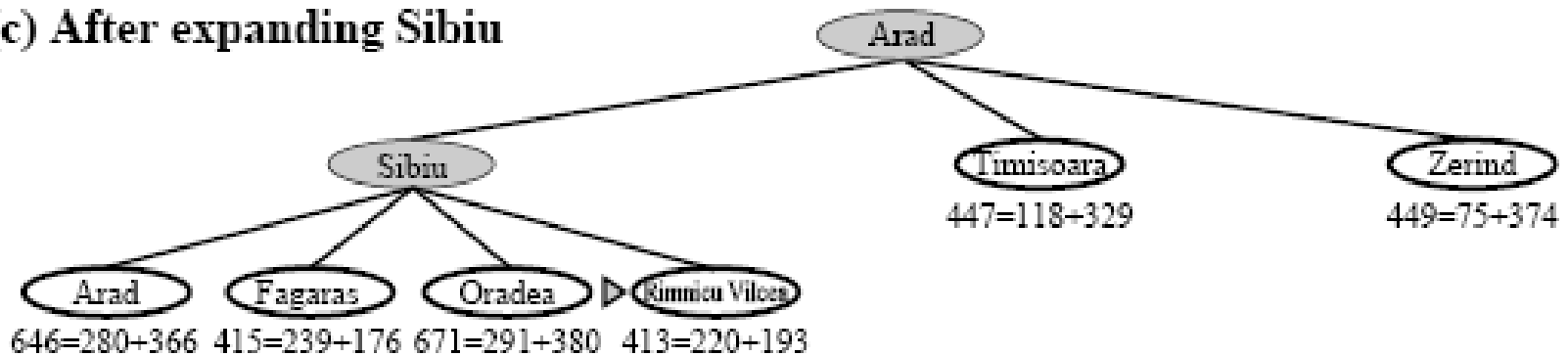
(a) The initial state



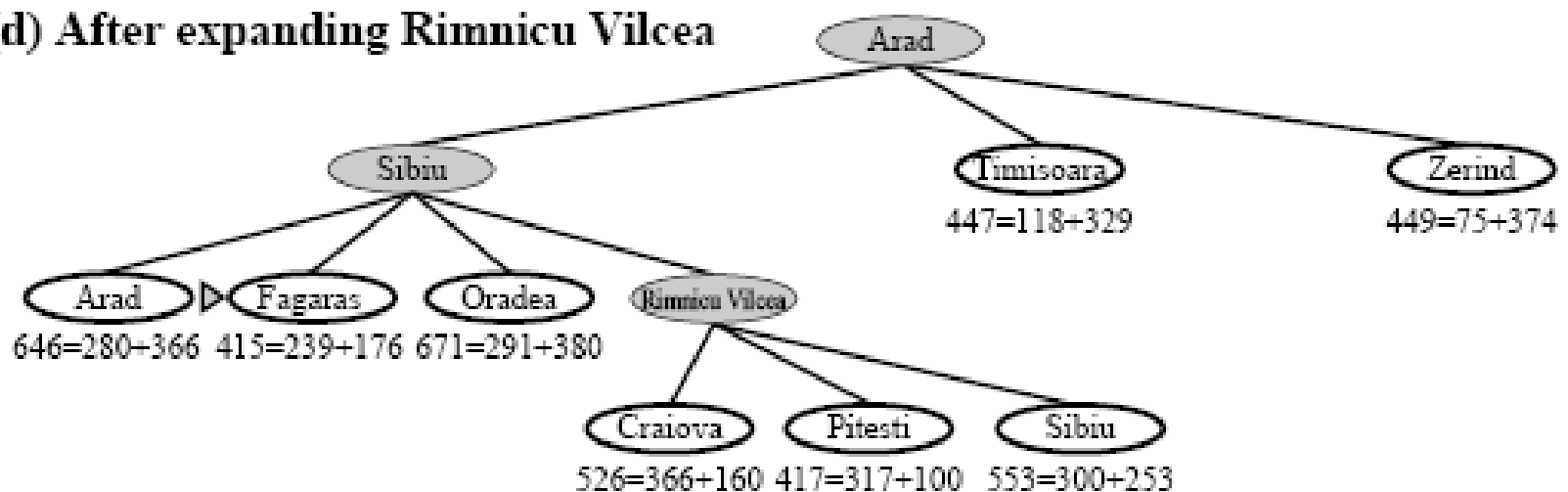
(b) After expanding Arad



(c) After expanding Sibiu

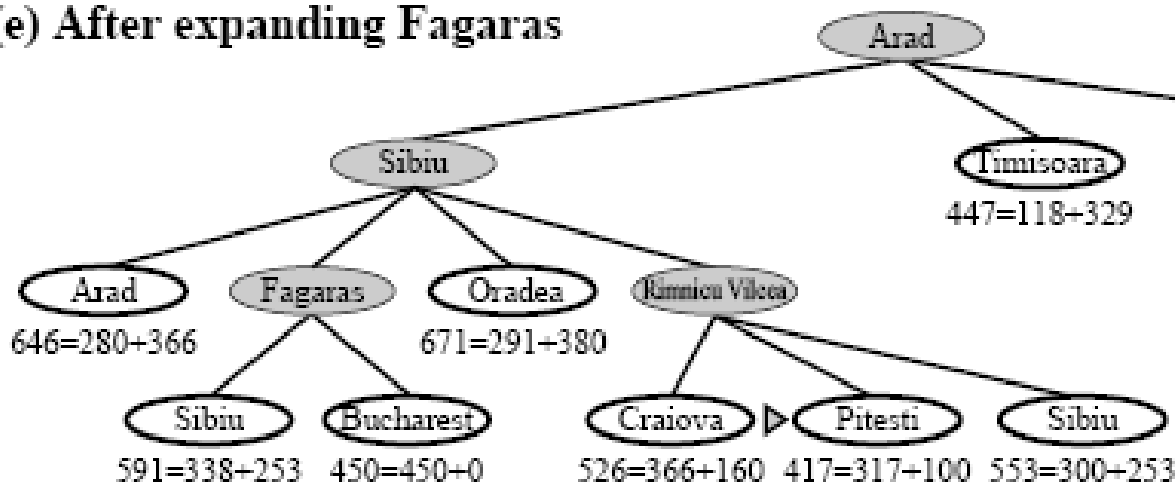


(d) After expanding Rimnicu Vilcea

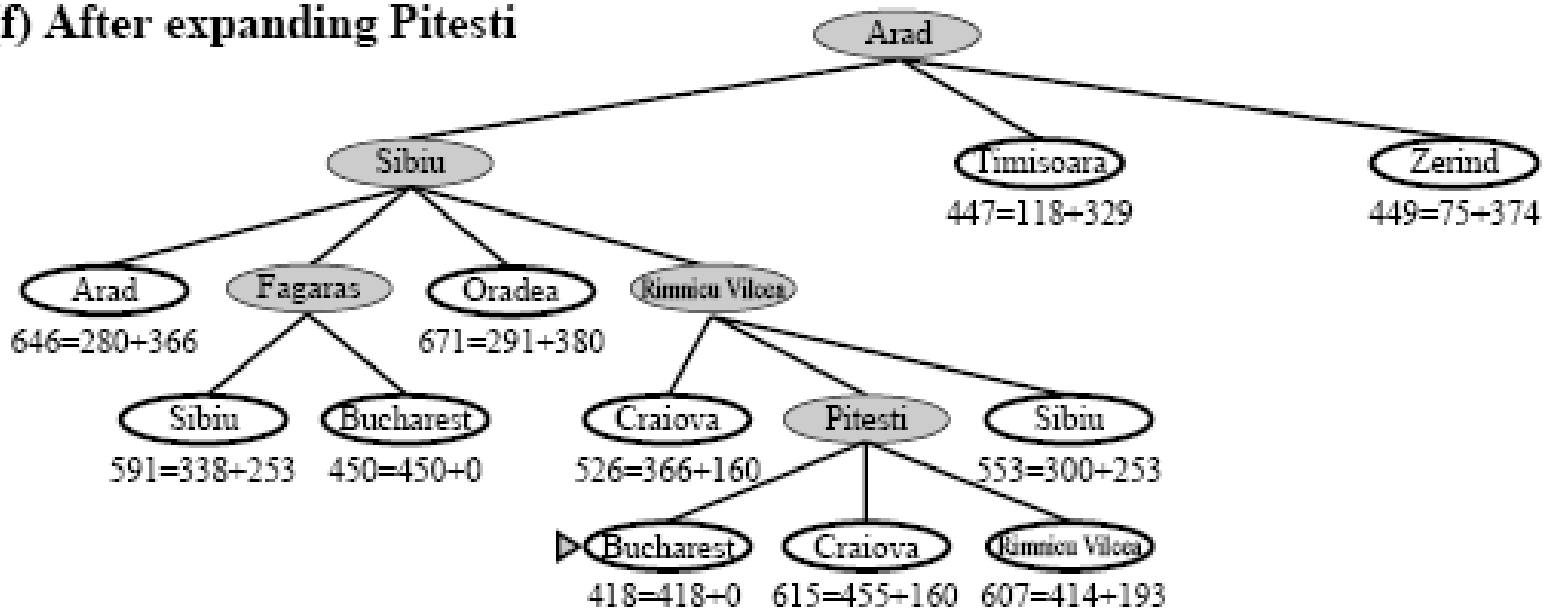


A* Search Example

(e) After expanding Fagaras



(f) After expanding Pitesti





Properties of A*

- Complete? Yes
- Time? Exponential
- Space? Keeps all nodes in memory
- Optimal? Yes





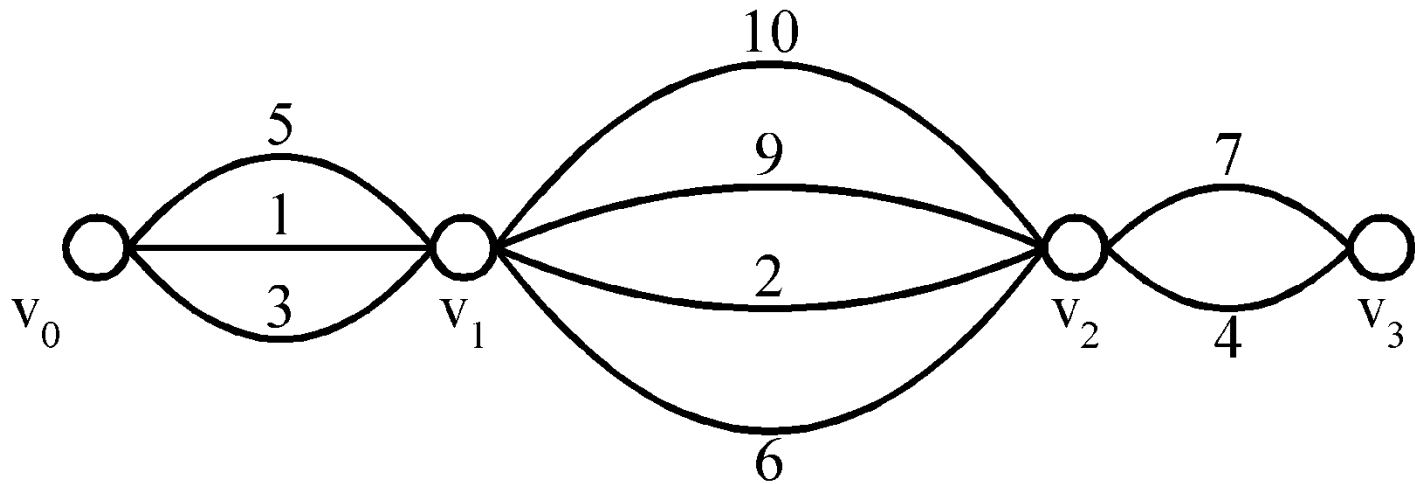
Examples





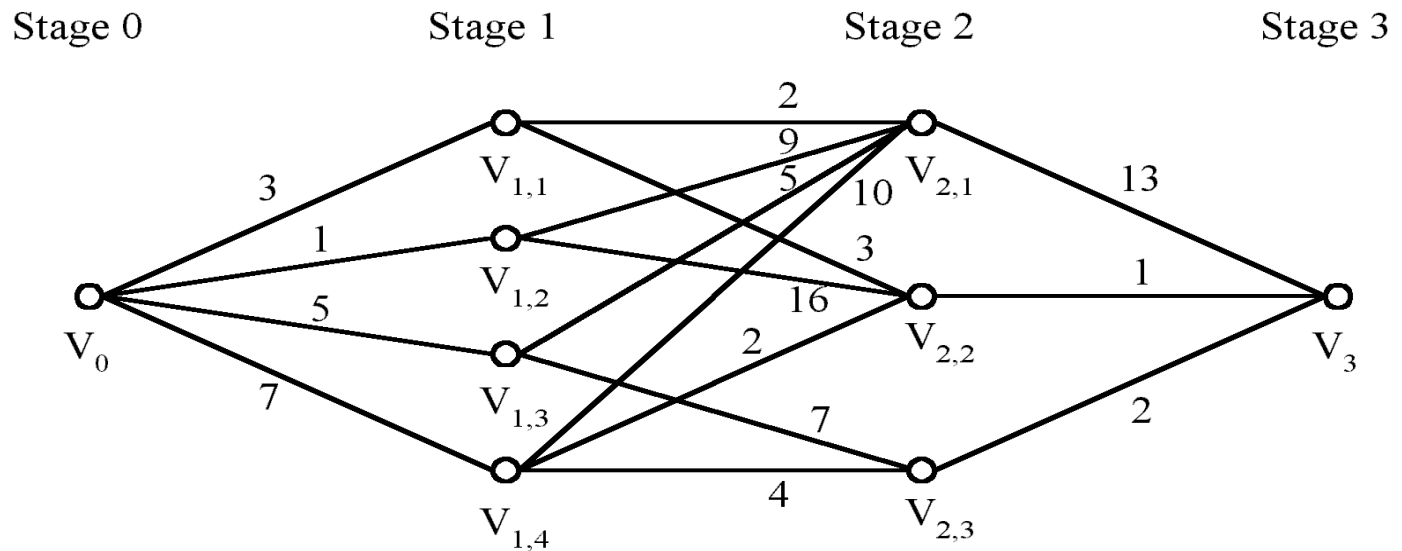
Shortest paths

- Find a shortest path from v_0 to v_3 ???
- Can the **greedy** method solve this problem???
- The shortest path: $1 + 2 + 4 = 7$.



Shortest paths on a multi-stage graph

- Find a shortest path from v_0 to v_3 in the multi-stage graph.



- Greedy method: $v_0 v_{1,2} v_{2,1} v_3 = 23$
- Optimal: $v_0 v_{1,1} v_{2,2} v_3 = 7$
- The greedy method does not work.



Admissible heuristics (accepted evaluation function)

E.g., for the 8-puzzle:

$h_1(n)$ = number of misplaced tiles [tiles in wrong places]

□ $h_2(n)$ = total Manhattan distance [how many moves to reach right place]
(i.e., no. of squares from desired location of each tile)

7	2	4
5		6
8	3	1

Start state

	1	2
3	4	5
6	7	8

Goal state

- $h_1(S) = ?$
- $h_2(S) = ?$

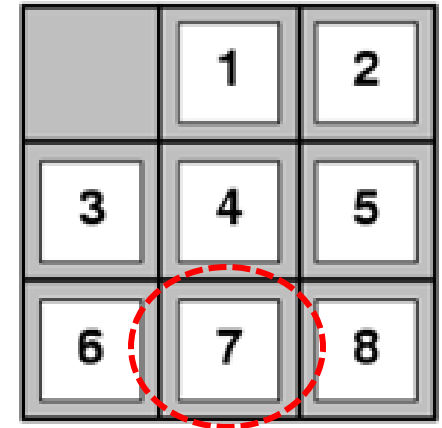
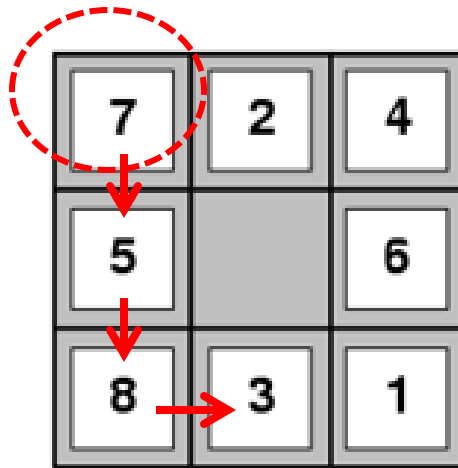
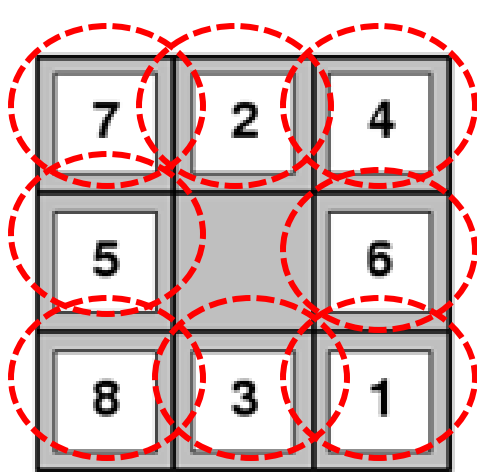


Admissible heuristics (accepted evaluation function)

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□ $h_2(n)$ = total Manhattan distance [how many moves to reach right place]
(i.e., no. of squares from desired location of each tile)



- $\underline{h_1(S)} = 8$

- $\underline{h_2(S)} = ?$

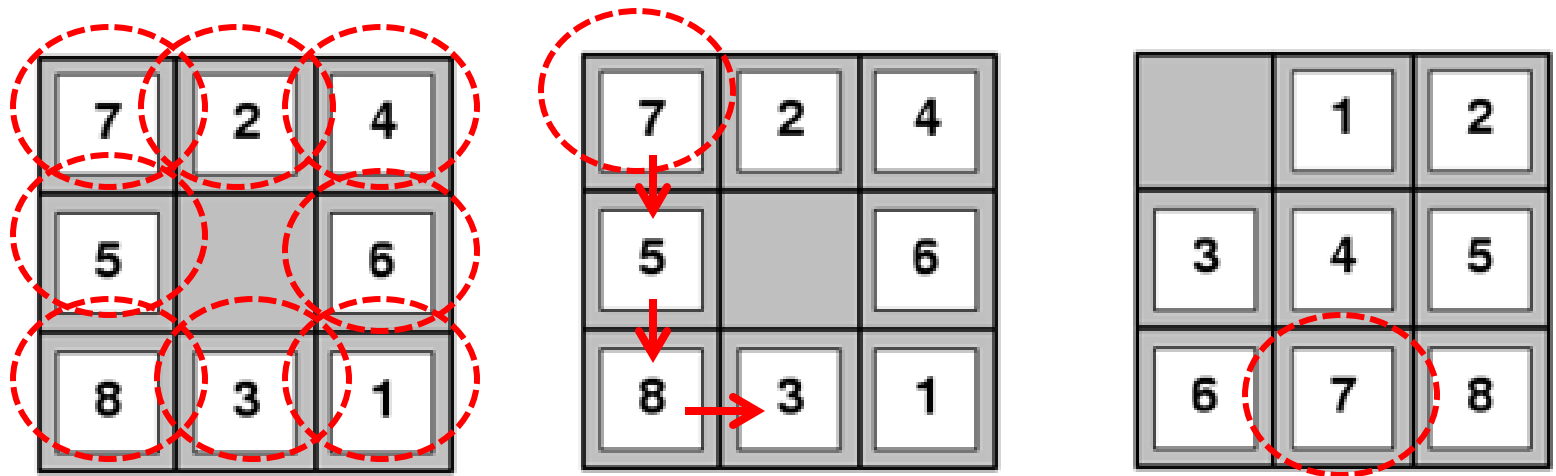




Admissible heuristics (accepted evaluation function)

E.g., for the 8-puzzle:

- $h_1(n)$ = number of misplaced tiles [tiles in wrong places]
- $h_2(n)$ = total Manhattan distance [how many moves to reach right place] (i.e., no. of squares from desired location of each tile)



- $h_1(S) = 8$

- $h_2(S) = 3+1+2+2+2+3+3+2 = 18$





Dominance (which is better)

- 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 = too many nodes
 $A^*(h_1) = 39,135$ nodes
 $A^*(h_2) = 1,641$ nodes





Thank you



**End of
Chapter 3-Part3**

