

UIT2504 Artificial Intelligence

Heuristics-Based Search Strategies

C. Aravindan
<AravindanC@ssn.edu.in>

Professor of Computer Science
SSN College of Engineering

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Searching for a solution

- Start with initial state in the working set
- Iterate:
 - Return failure if the working set is empty
 - Choose and remove a state x from the working set
 - If it is a goal state, return solution
 - Else, expand x and add the successor states $S(x)$ to the working set

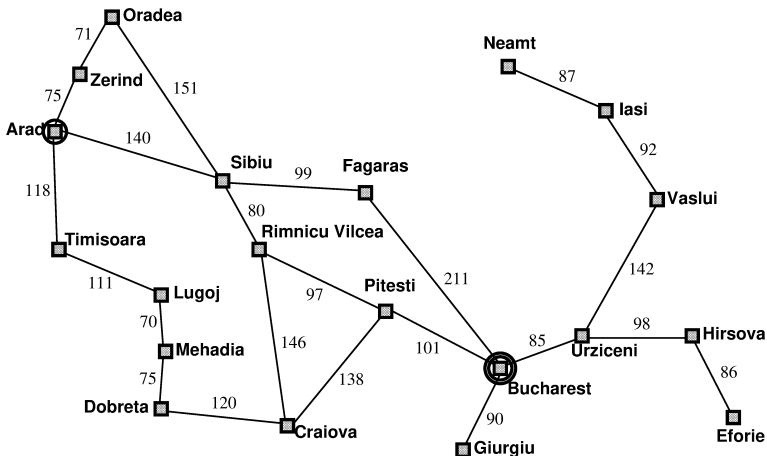
- Uninformed:
 - Breadth-First
 - Depth-First
 - Iterative Deepening
 - Bi-Directional Search
- Informed (Heuristics):
 - Best-first Greedy
 - A^*
 - Local Search Strategies
- Constraint Satisfaction

Performance Measures

- Completeness
- Time Complexity
- Space Complexity
- Optimality

Exercise

- Practice all the basic search strategies to find a route from Arad to Bucharest in the following state graph



Questions?

Evaluating a state

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- Working set is maintained as a priority queue based on the evaluation function f
- Obviously, the quality of search depends on the evaluation function f

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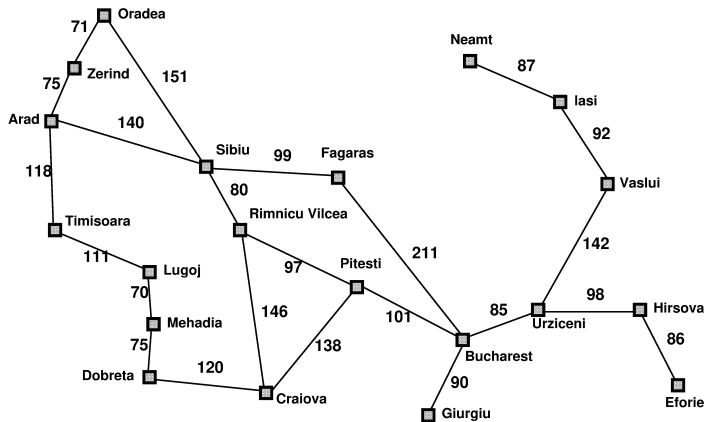
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- For example, can you think of a heuristics for the route finding problem in a map? — Straight line distance (SLD) from the current city to the destination city
- Heuristics should be an easy function to compute!
- $h(s^*)$ should be 0 for any goal state s^*

Example: Route finding problem



Straight-line distance to Bucharest

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Find route from Arad to Bucharest

Example: Sliding puzzle

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5	8	
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- An **admissible heuristics** is one which does not overestimate — in our example, both $h_1(x)$ and $h_2(x)$ are admissible

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18	12	14	13	13	12	14	14
14	16	13	15	12	14	12	16
14	12	18	13	15	12	14	14
15	14	14	♔	13	16	13	16
♔	14	17	15	♔	14	16	16
17	♔	16	18	15	♔	15	♔
18	14	♔	15	15	14	♔	16
14	14	13	17	12	14	12	18

Questions?

Best-first greedy search

- As a simple strategy, we may let the evaluation function f to be the same as the heuristics function h

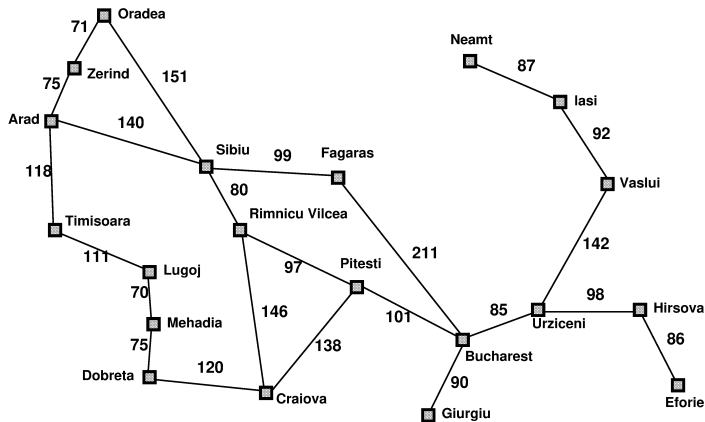
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- Nodes in the working set (priority queue) are organized based on estimated cost and the one with the least cost is given preference
- This is a generalization of **greedy design strategy**, that you have learnt in the previous semester

Best-first greedy search



Straight-line distance
to Bucharest

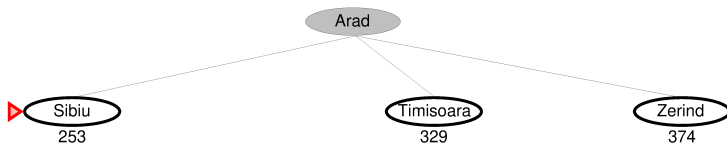
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Find the best route from Arad to Bucharest

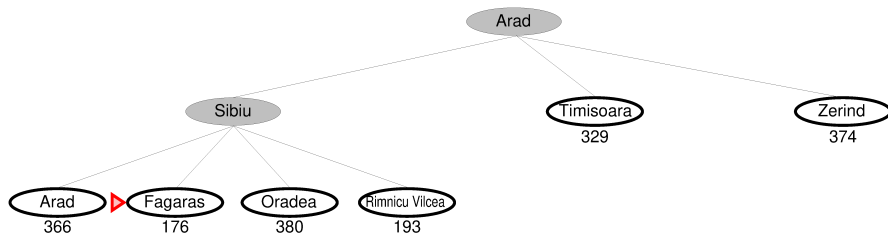
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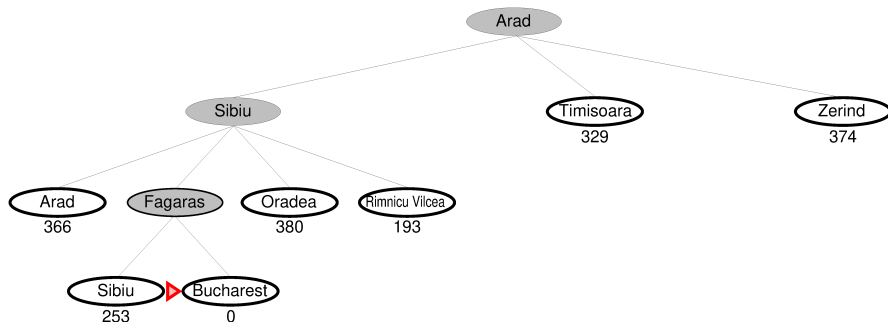
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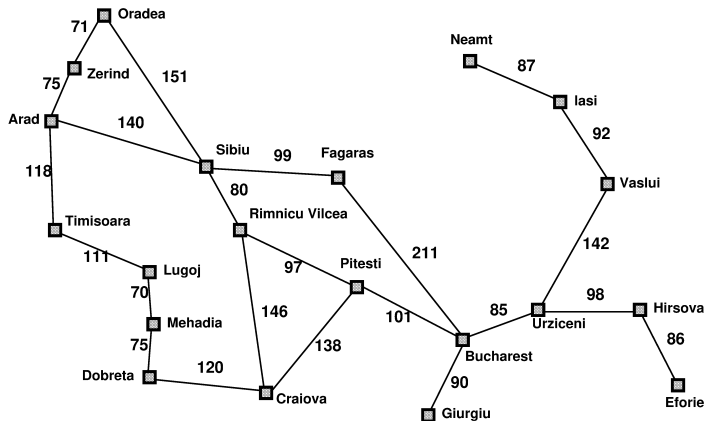
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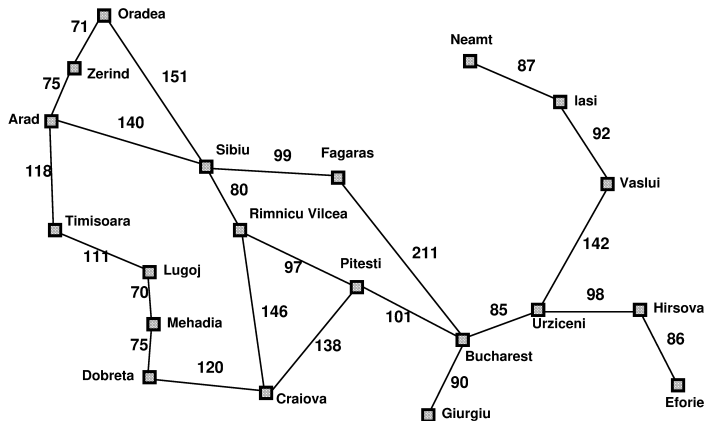
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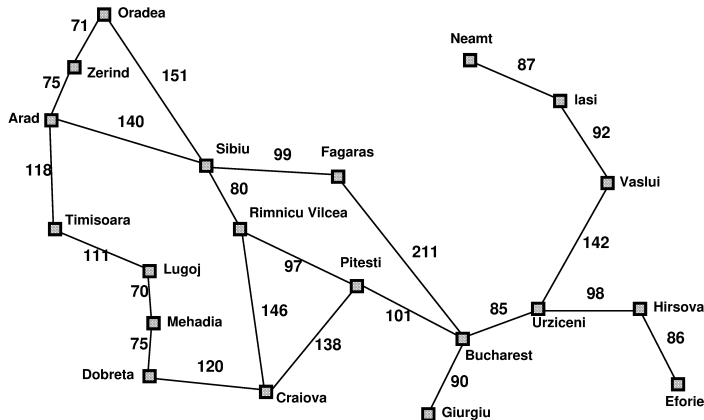
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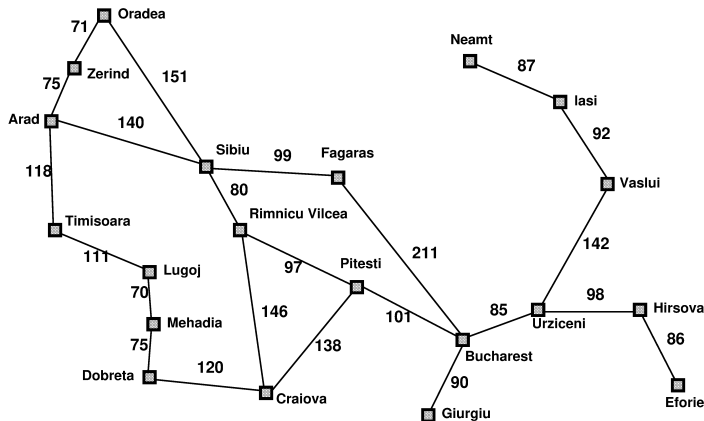
Solution found: Arad → Sibiu → Fagaras → Bucharest
with total cost

Best-first greedy search



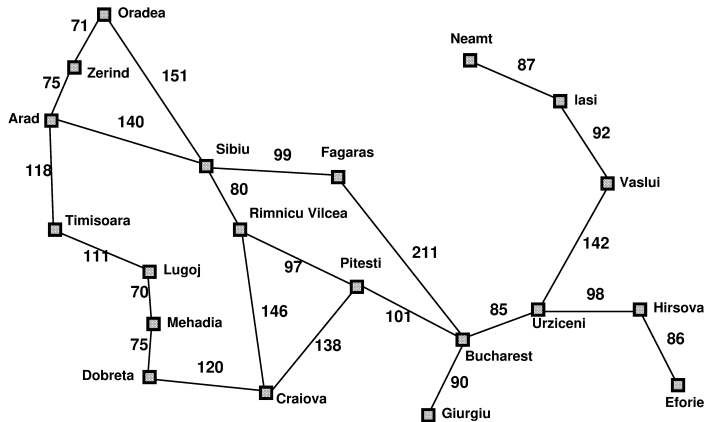
Solution found: Arad → Sibiu → Fagaras → Bucharest
with total cost 450
Is it optimal?

Best-first greedy search



Solution found: Arad → Sibiu → Fagaras → Bucharest
 with total cost 450
 Is it optimal? — No!

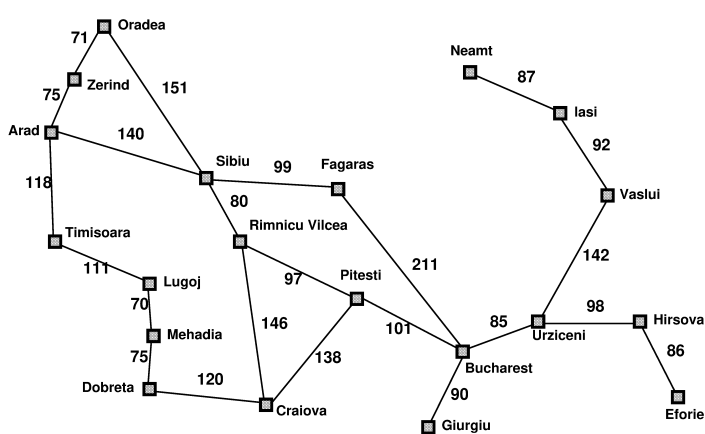
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Find a path from Iasi to Fagaras

Best-First Greedy: Complexities

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- Time complexity?

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