

Informed search methods

Uninformed vs. Informed Search

- Uninformed search strategies
 - Find solutions to problems by systematically generating new states and testing for goal
 - Most are incredibly inefficient

Uninformed vs. Informed Search

- Uninformed search strategies
 - Find solutions to problems by systematically generating new states and testing for goal
 - Most are incredibly inefficient
- Informed search strategies
 - Use problem-specific knowledge
 - Find solutions more efficiently

Informed Search

- Informed search strategy
 - One that uses problem-specific knowledge beyond the problem definition itself
- General approach is called “Best First” search
 - Node selected for expansion based on an evaluation function, $f(n)$
 - Measuring a distance to goal
 - Node with lowest evaluation is selected

Techniques

- Best-first search
 - Expand “minimum cost” nodes first
 - Greedy search
 - A* search

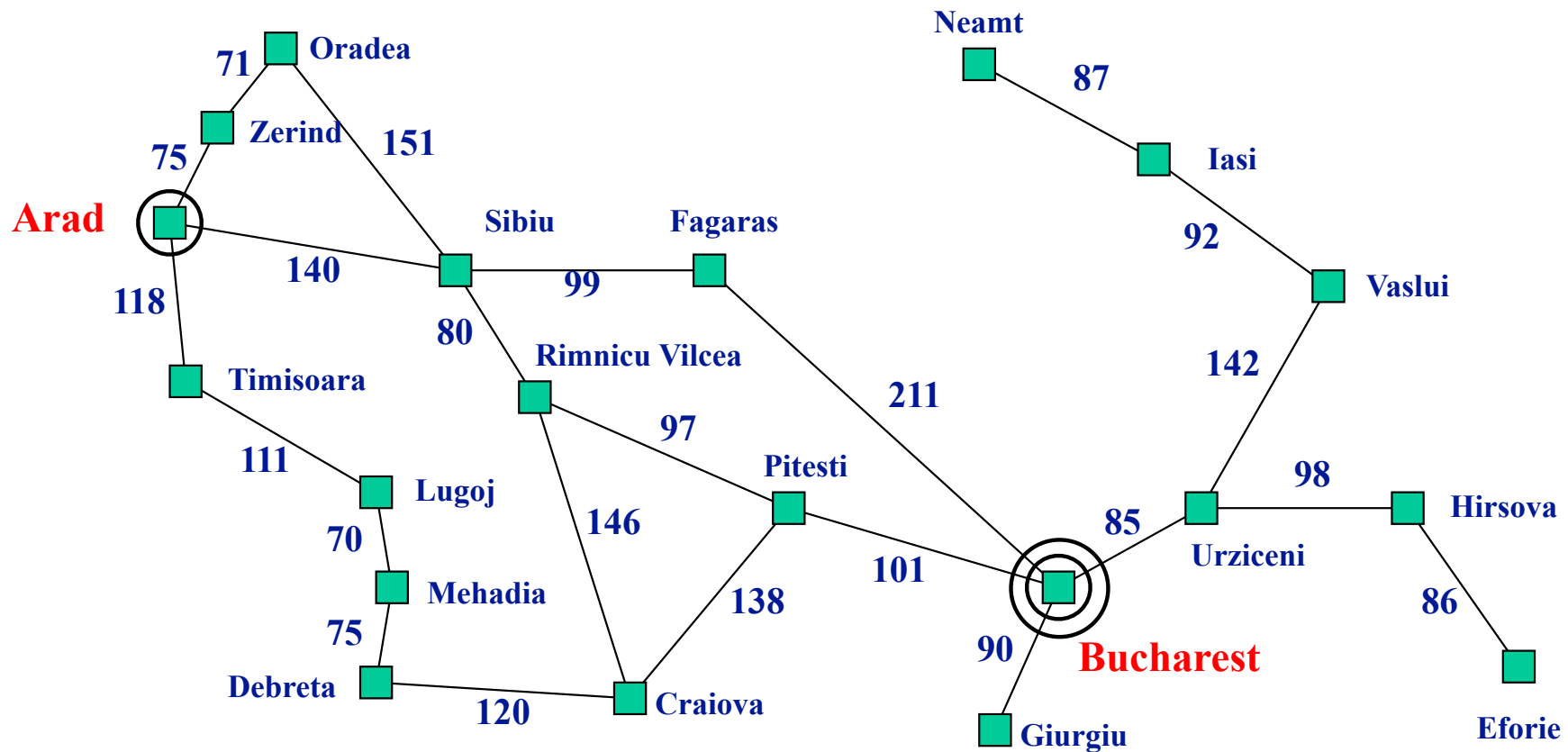
Best-First Search

- Expand “minimum cost” nodes first
 - Greedy search
 - A^* search
- Open set
 - Priority queue implementation
 - Insert expanded nodes in decreasing order of desirability (most desirable first)

Greedy Search

- Simplest best-first search strategy
- Minimize estimated cost to reach goal
- Heuristic function $h(n)$
 - **Estimate of cost from node n to goal**
 - Require $h(n) = 0$ if n is goal
- Greedy search expands the node that *appears* to be closest to the goal
 - Each step tries to get as close to goal as possible
 - Though, will not consider if its action will be the best in the long run

Romania Step Costs in km



Straight-Line Distance to Bucharest

H_{SLD}

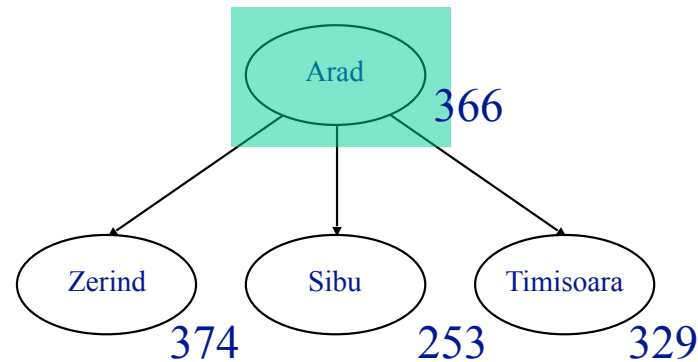
Arad	366	Mehadia	241
Bucharest	0	Neamt	234
Craiova	160	Oradea	380
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Iasi	226	Vaslui	199
Lugoj	244	Zerind	374

Greedy Search Example: Romania



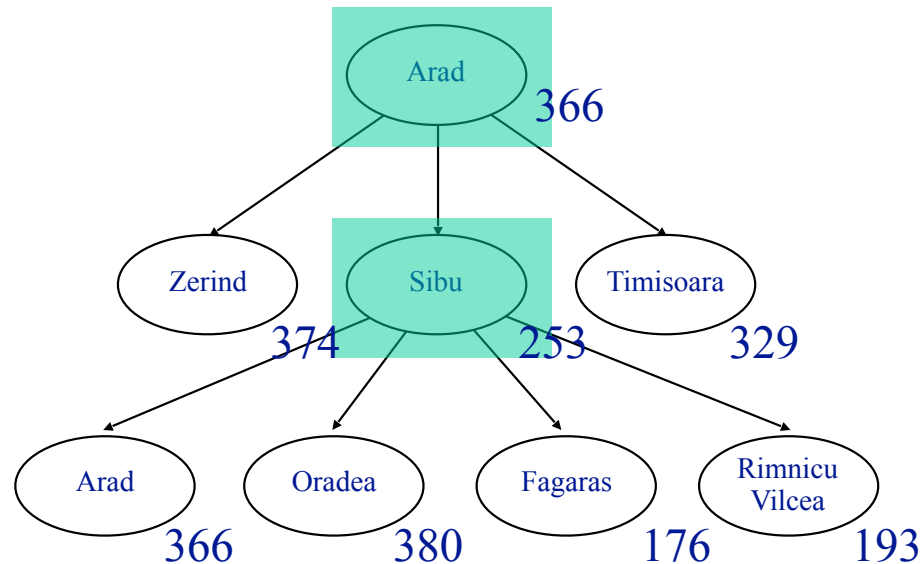
Open = { Arad(366) }

Greedy Search Example: Romania



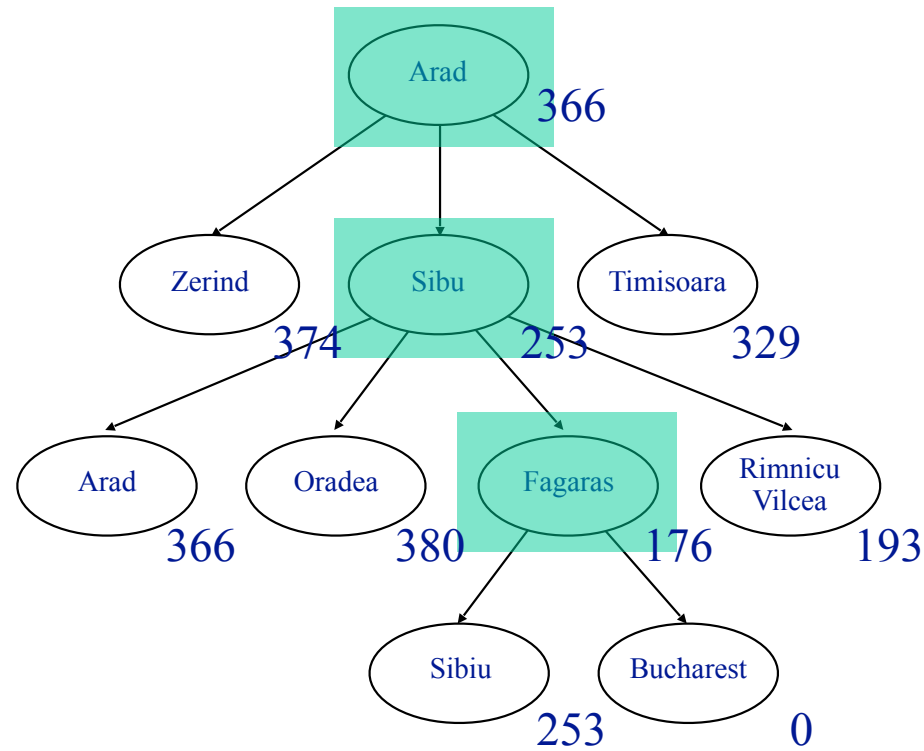
Open = { Sibü(253), Timisoara(329), Zerind(374) }

Greedy Search Example: Romania



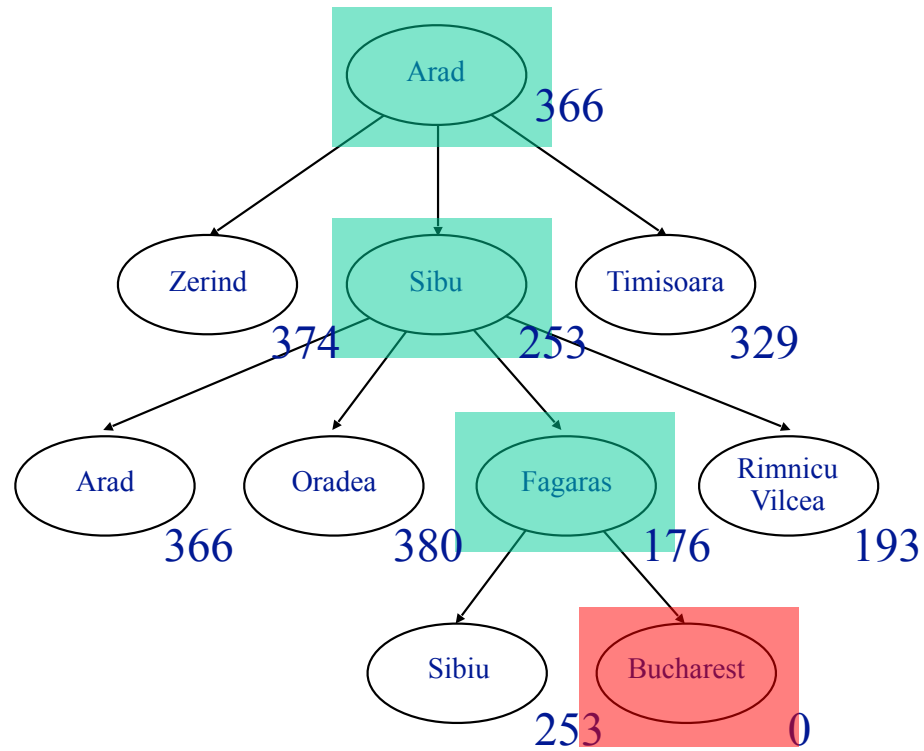
Open = { Fagaras(176), RimnicuVilcea(193), Timisoara(329),
Arad(366), Zerind(374), Oradea(380) }

Greedy Search Example: Romania

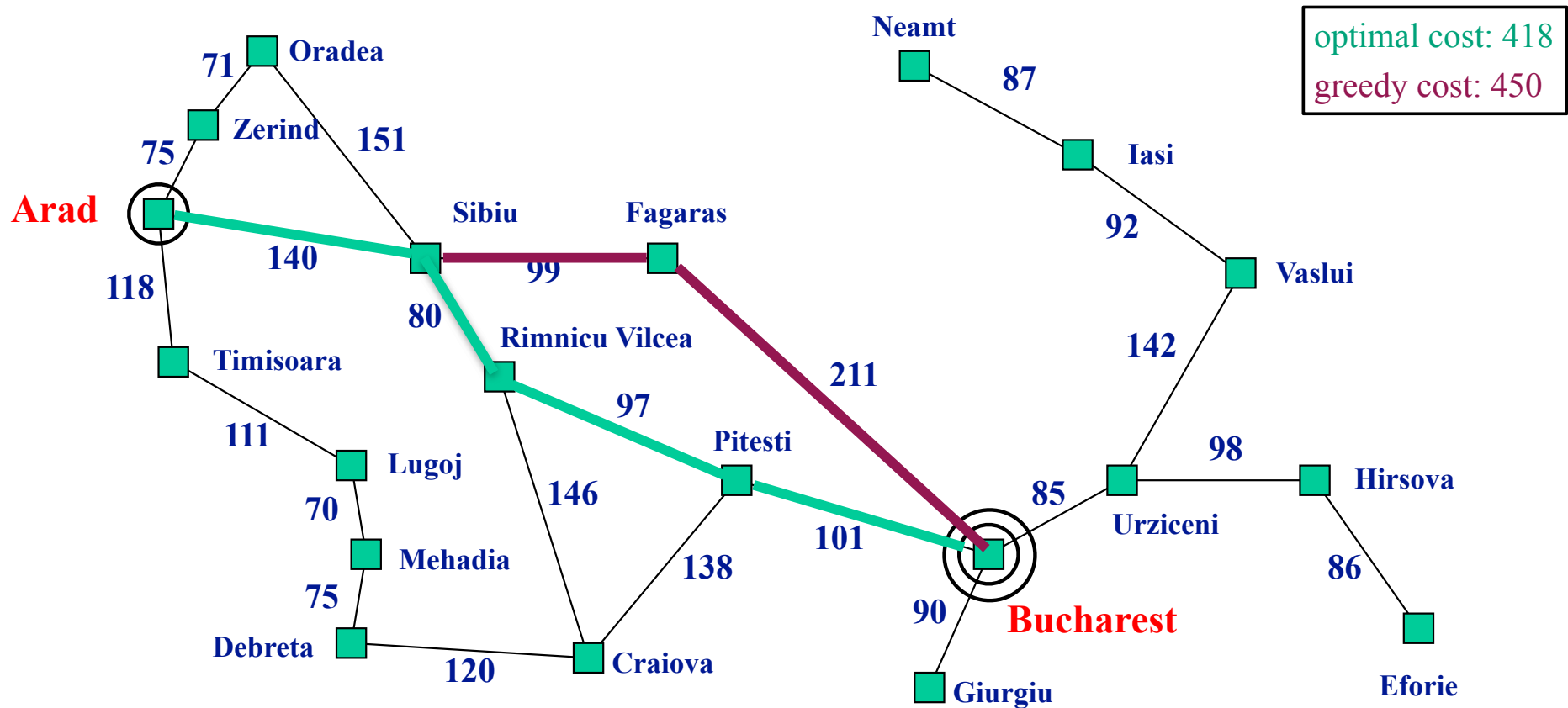


Open = { **Bucharest(0)**, Fagaras(176), RimnicuVilcea(193), **Sibiu(253)**, Timisoara(329), Arad(366), Zerind(374), Oradea(380) }

Greedy Search Example: Romania



Romania Step Costs in km



Properties of Greedy Search

- Not complete (**can be bad**)
 - Can get stuck in loops
 - Complete if check for repeated states
- Time (**can be bad**)
 - $O(b^m)$, m =maximum depth (worst case, like DFS)
- Space (**can be bad**)
 - $O(b^m)$, keeps all nodes in memory (worst case)
- Not optimal (**can be bad**)
 - Heuristic is an estimate

But a good heuristic can give dramatic improvement!

A* Search

- Minimizes total estimated path cost
- Avoids expanding paths already expensive
- Evaluation function $f(n) = g(n) + h(n)$
 - Actual cost to reach node n so far $\rightarrow g(n)$
 - Estimated cost from n to goal $\rightarrow h(n)$
 - Estimated total cost through n to goal
 $\rightarrow f(n) = g(n) + h(n)$
- A* search uses *admissible* heuristic
 - i.e., $h(n) \leq h^*(n)$ where $h^*(n)$ is true cost from n to goal
 - e.g., $h_{\text{SLD}}(n)$ never overestimates actual distance

A* Search Example: Romania

Open = { Arad(366) }

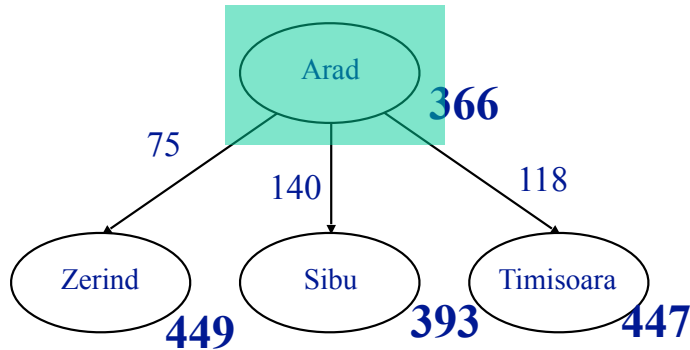


$g(n)$: cost so far

$f(n) = g(n) + h(n)$ = cost so far + estimated cost to goal

A* Search Example: Romania

Open = { Sibiu(393),
Timisoara(447), Zerind(449) }



$$f(\text{Zerind}) = 75 + 374 = 449$$

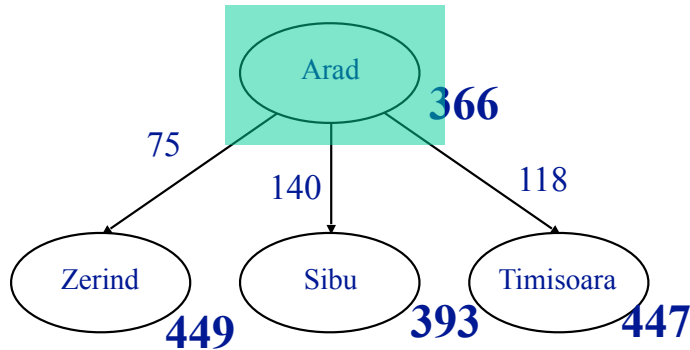
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$$f(\text{Zerind}) = 75 + 374 = 449$$

$$f(\text{Sibu}) = 140 + 253 = 393$$

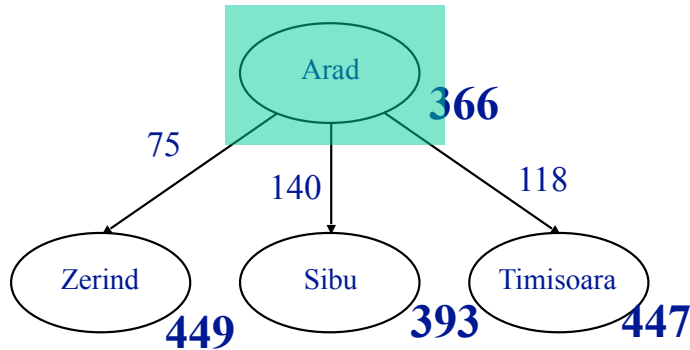
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$$f(\text{Zerind}) = 75 + 374 = 449$$

$$f(\text{Sibiu}) = 140 + 253 = 393$$

$$f(\text{Timisoara}) = 118 + 329 = 447$$

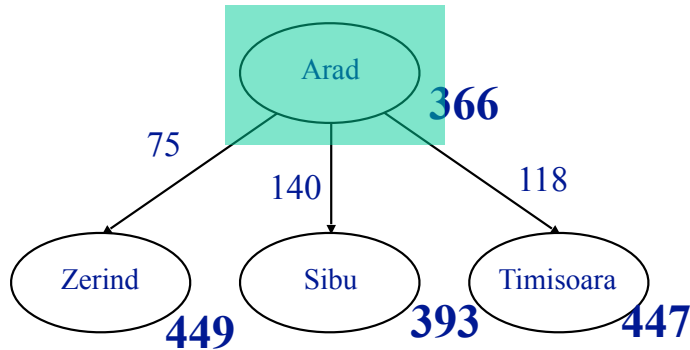
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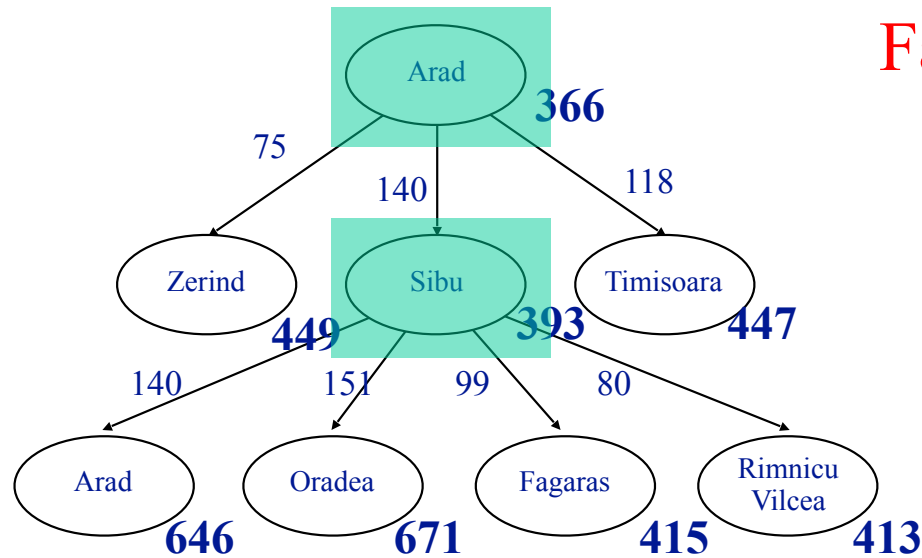
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A* Search Example: Romania



Open = { Rimnicu Vilcea(413),
Fagaras(415), Timisoara(447),
Zerind(449), Arad(646),
Oradea(671) }

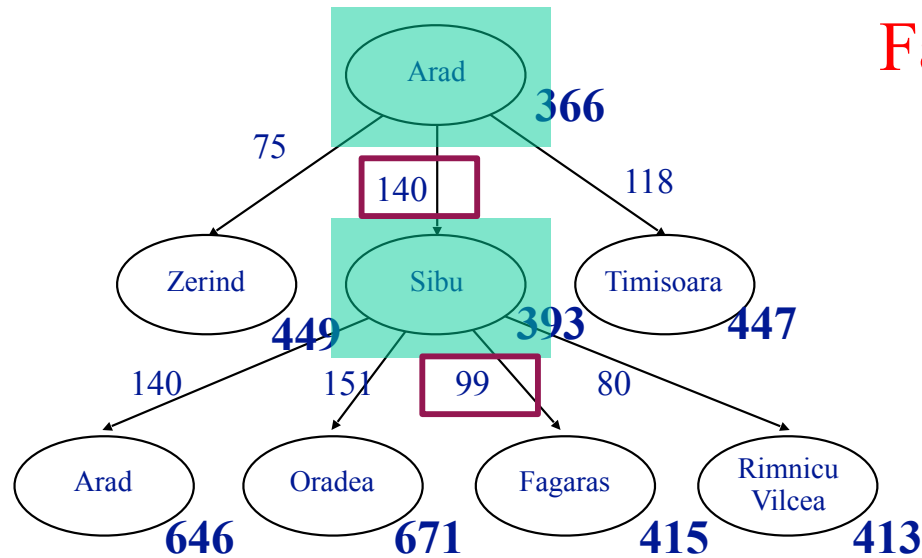
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$$f(\text{Fagaras}) = (140 + 99) + 176 = 415$$

$g(n)$: cost so far

$f(n) = g(n) + h(n)$ = cost so far + estimated cost to goal

A* Search Example: Romania



Open = { Rimnicu Vilcea(413),
Fagaras(415), Timisoara(447),
Zerind(449), Arad(646),
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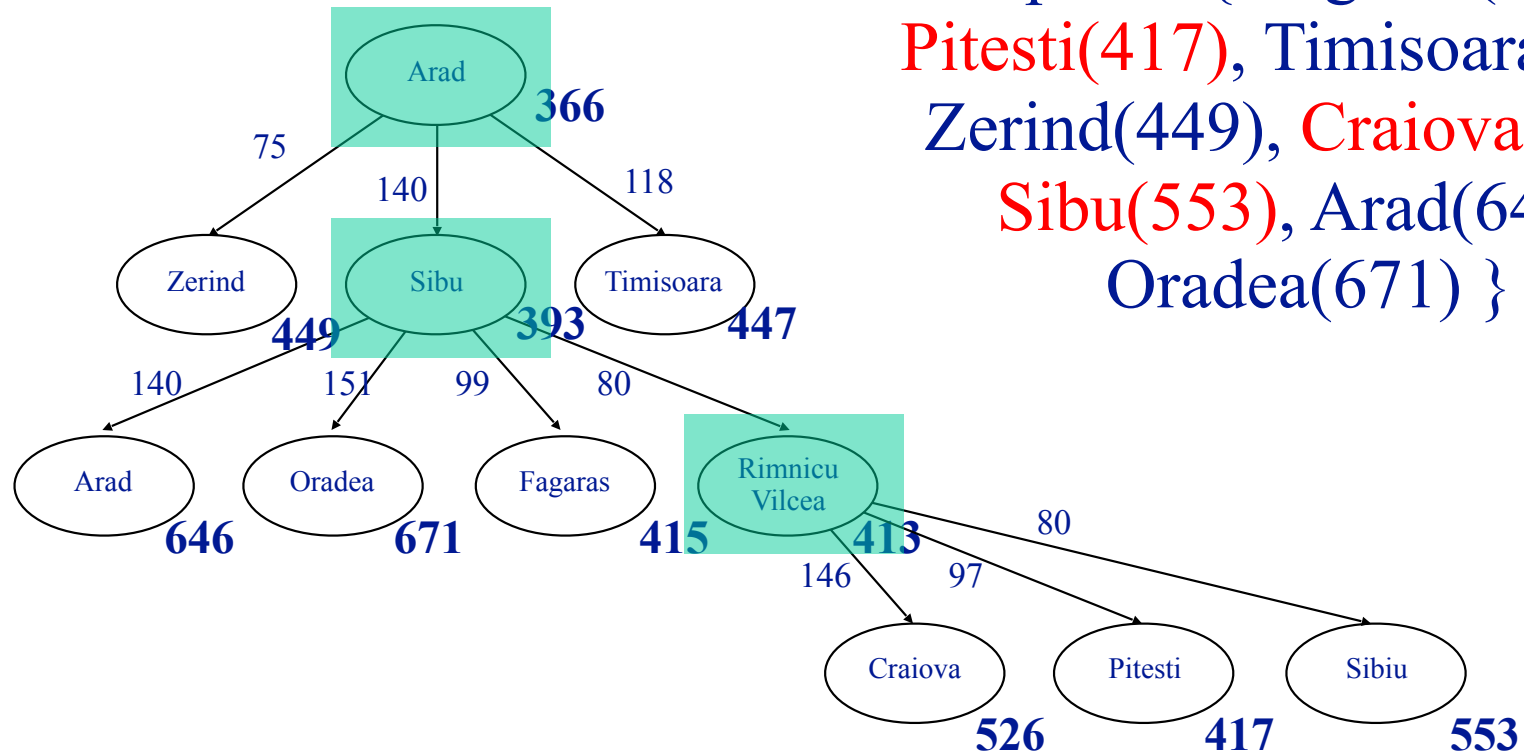
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A* Search Example: Romania

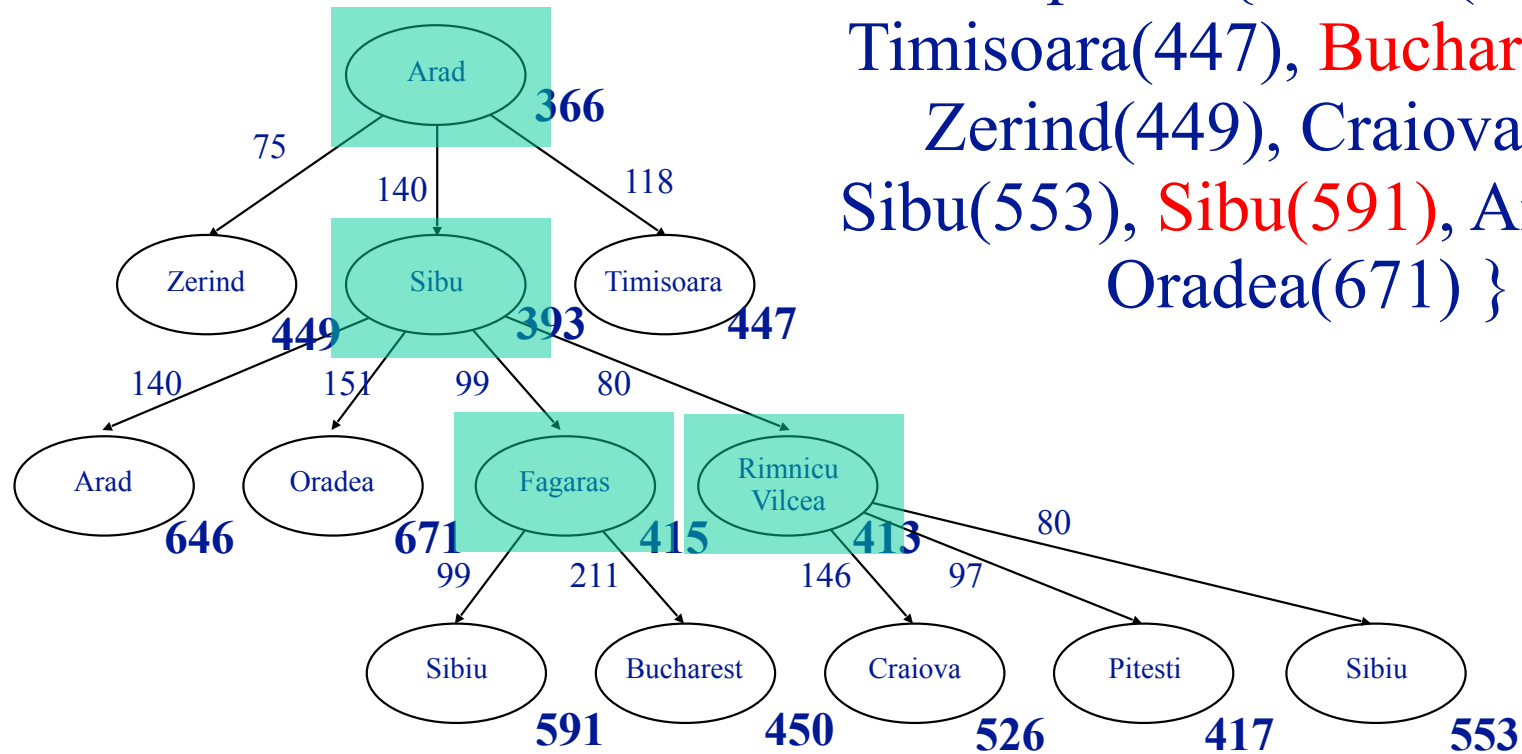


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A* Search Example: Romania

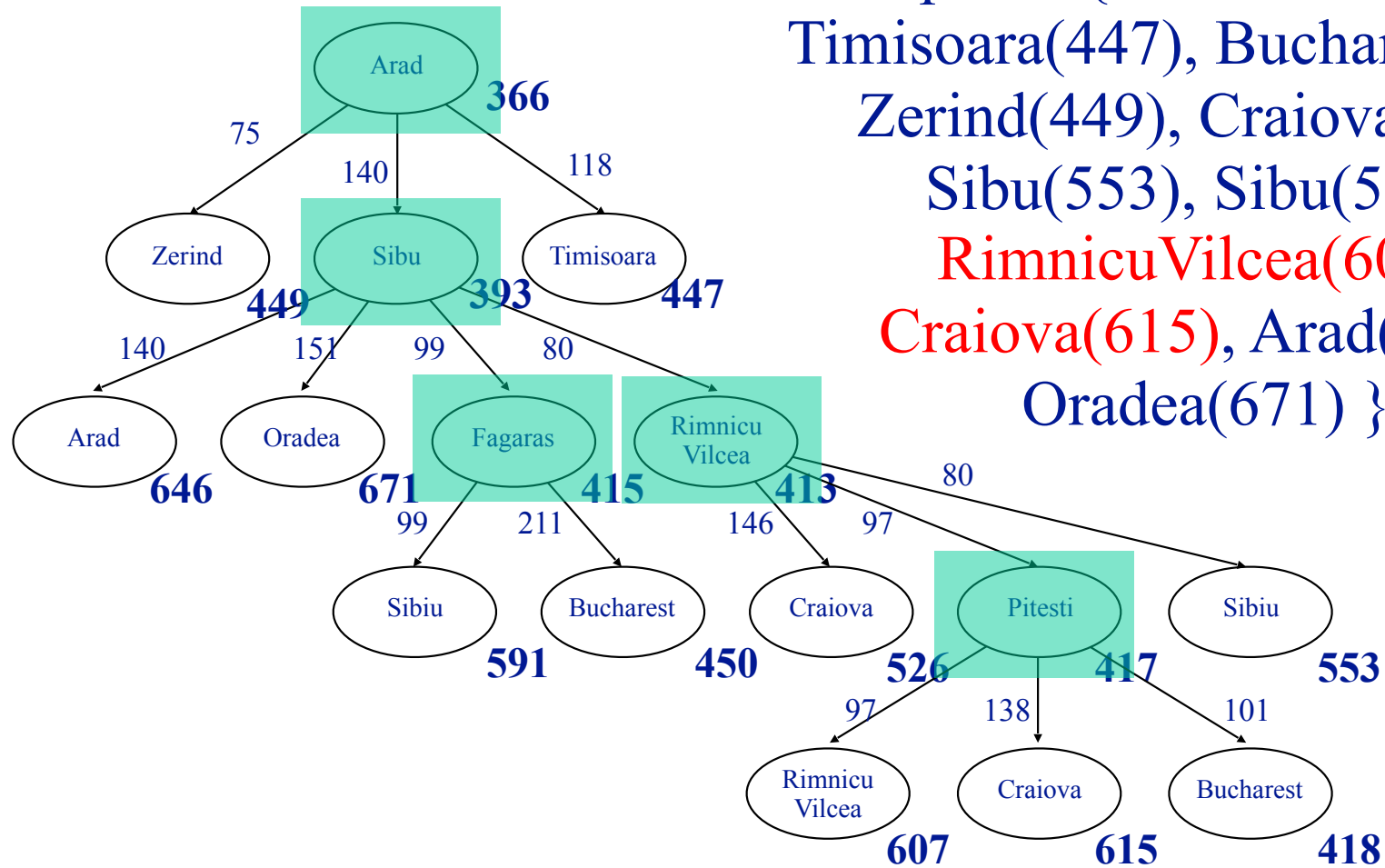
Open = { Pitesti(417),
Timisoara(447), **Bucharest(450)**,
Zerind(449), Craiova(526),
Sibu(553), **Sibu(591)**, Arad(646),
Oradea(671) }



$g(n)$: cost so far

$f(n) = g(n) + h(n)$ = cost so far + estimated cost to goal

A* Search Example: Romania



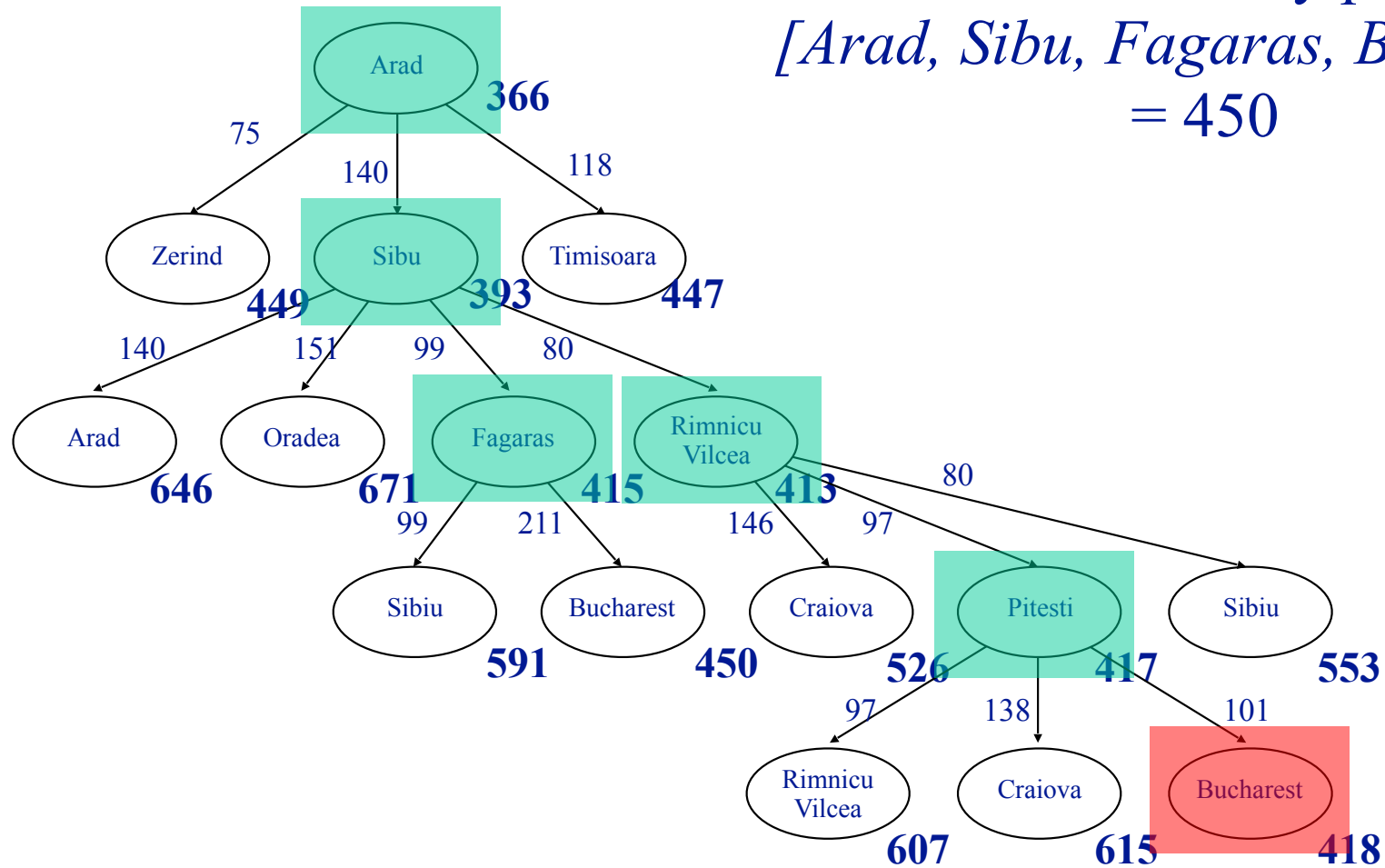
Open = { **Bucharest(418)**,
 Timisoara(447), Bucharest(450),
 Zerind(449), Craiova(526),
 Sibiu(553), Sibiu(591),
Rimnicu Vilcea(607),
Craiova(615), Arad(646),
 Oradea(671) }

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$f(n) = g(n) + h(n)$ = cost so far + estimated cost to goal

A* Search Example: Romania

Recall that Greedy path was
[Arad, Sibiu, Fagaras, Bucharest]
= 450

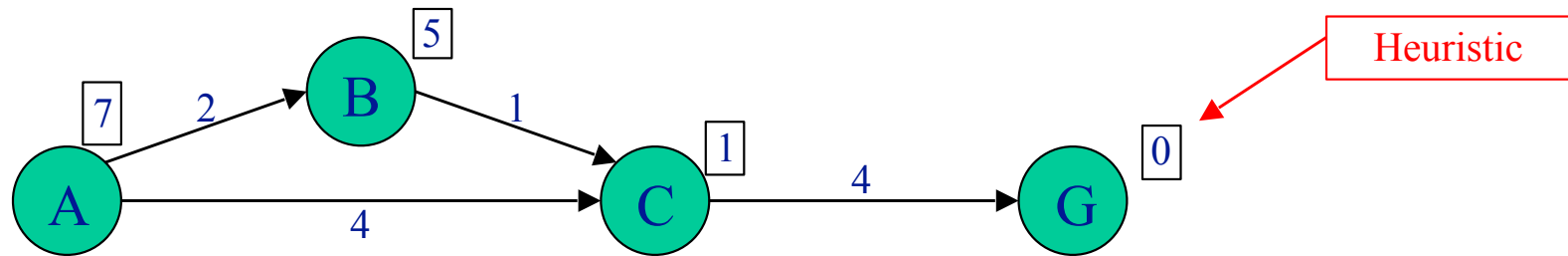


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Repeated states in A*

- Contrary to greedy search algorithms, avoiding repeated states can actually harm A*
- Consider:



$A(0+7=7)$ $B(2+5=7)$ $C(4+1=5)$ $G(4+4+0=8)$

- Goal is discovered before finding shortest path to it!
- Depending on algorithm construction, could return sub-optimal path
 - At best, devolves to BFS-like performance

Consistent Heuristics

- Consistency: The difference in heuristic values between two connected states is always smaller than (or equal to) the actual cost:

$$cost(x, y) \geq |h(x) - h(y)|$$

For two connected states x and y

Consistent Heuristics

- Consistency: The difference in heuristic values between two connected states is always smaller than (or equal to) the actual cost:

$$cost(x, y) \geq |h(x) - h(y)|$$

For two connected
states x and y

- Consistency ensures that the total estimated path cost always increases (or stays the same) as we get closer to a goal state
 - For this reason, sometimes also called a monotonic heuristic
 - Guarantees that shortest path is always expanded first

Properties of A*

- Complete (**good**)
 - Unless infinitely many nodes
- Time and space (**not good**)
 - Still exponential in worst case (keeps all nodes in memory)
 - Strongly depends on choice of heuristic
- Optimal (**good**)
 - Expands fewest nodes