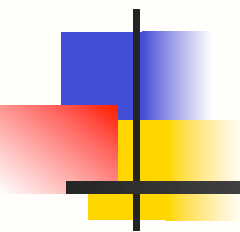


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# A\* Search



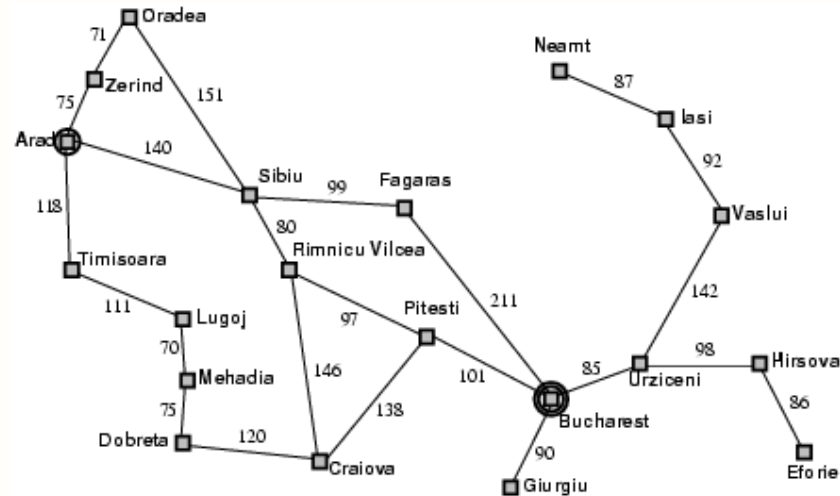
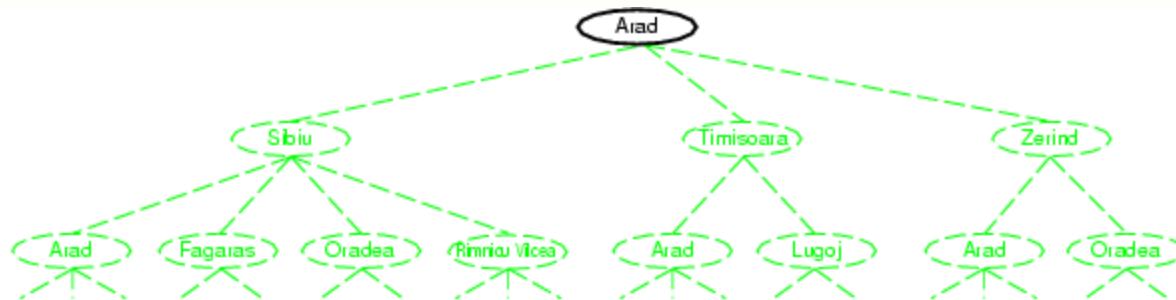


# Tree search algorithms

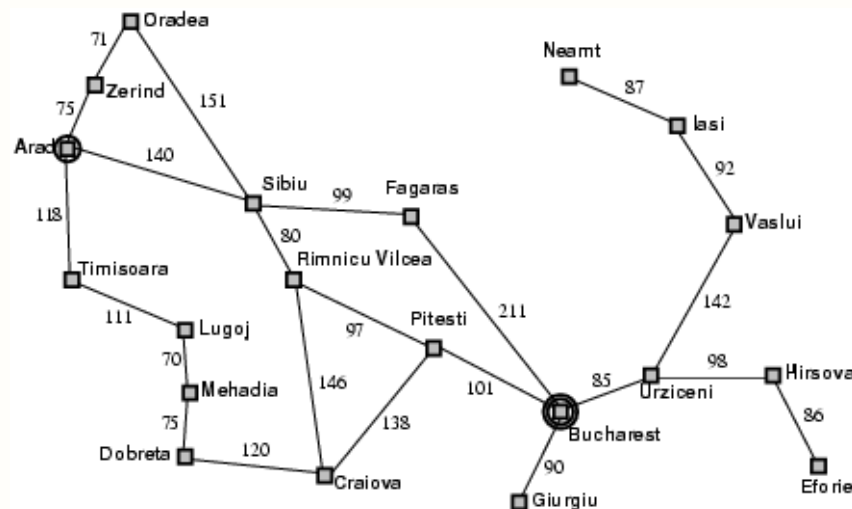
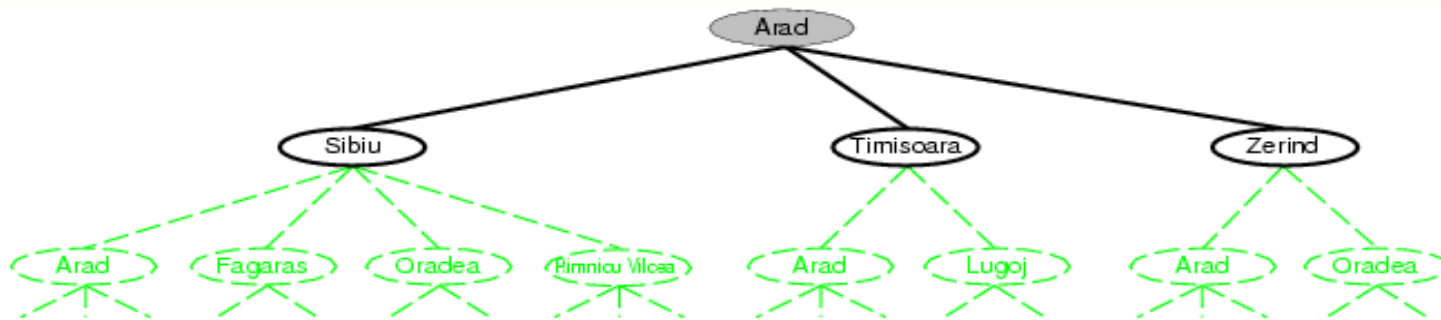
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- Basic idea:
  - Exploration of state space by generating successors of already-explored states (a.k.a. ~**expanding** states).
  - Every states is evaluated: *is it a goal state?*

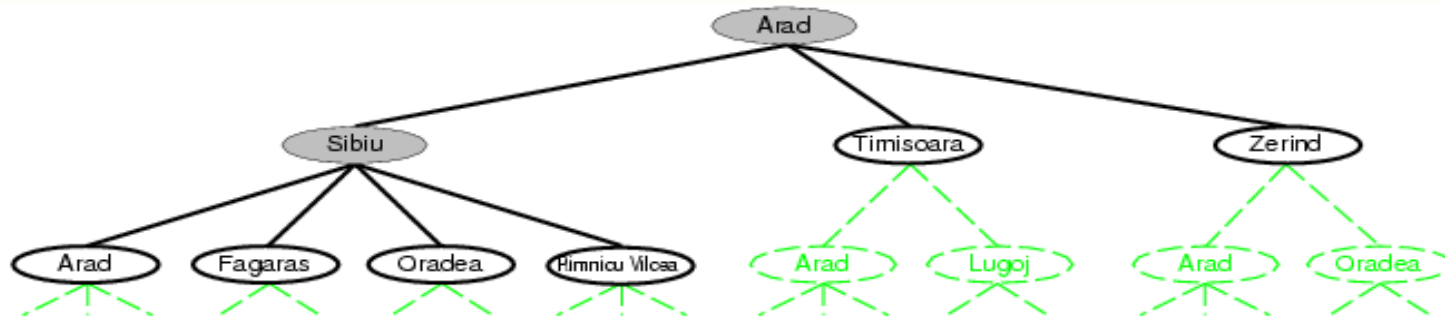
# Tree search example



# Tree search example



# Tree search example



```
function TREE-SEARCH(problem, strategy) returns a solution, or failure
  initialize the search tree using the initial state of problem
  loop do
    if there are no candidates for expansion then return failure
    choose a leaf node for expansion according to strategy
    if the node contains a goal state then return the corresponding solution
    else expand the node and add the resulting nodes to the search tree
```

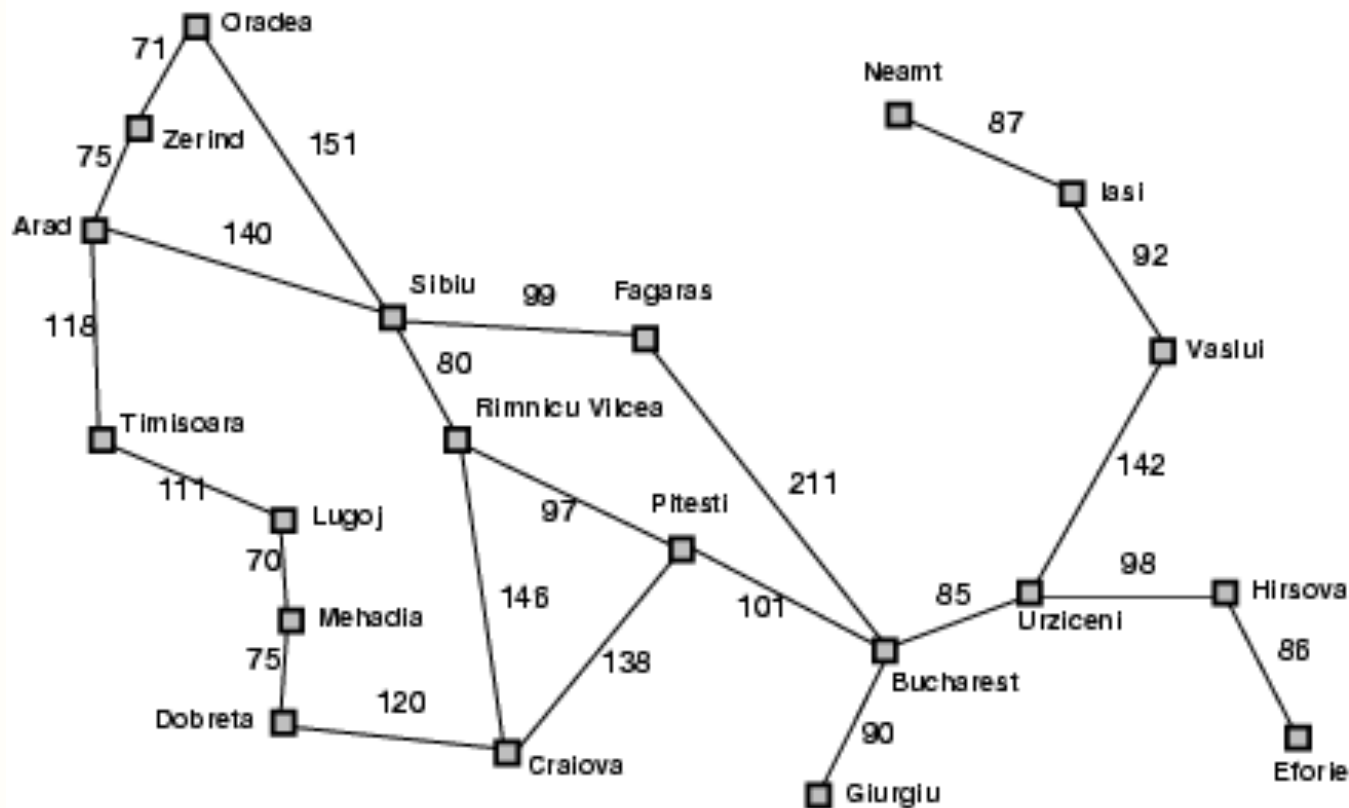


# Best-first search

---

- Idea: use an **evaluation function**  $f(n)$  for each node
  - $f(n)$  provides an estimate for the total cost.
    - Expand the node  $n$  with smallest  $f(n)$ .
- Implementation:  
Order the nodes in fringe increasing order of cost.

# Romania with straight-line dist.



Straight-line distance  
to Bucharest

<b>Arad</b>	366
<b>Bucharest</b>	0
<b>Craiova</b>	160
<b>Dobreta</b>	242
<b>Eforie</b>	161
<b>Fagaras</b>	176
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<b>Zerind</b>	374



# A\* search

---

- Idea: avoid expanding paths that are already expensive
- 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
- Best First search has  $f(n)=h(n)$
- Uniform Cost search has  $f(n)=g(n)$





# Admissible heuristics

---

- A heuristic  $h(n)$  is **admissible** if for every node  $n$ ,  $h(n) \leq h^*(n)$ , where  $h^*(n)$  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)
- **Theorem**: If  $h(n)$  is admissible,  $A^*$  using TREE-SEARCH is optimal



# 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: it is guaranteed to expand less or equal nr of nodes.
- 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



# 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

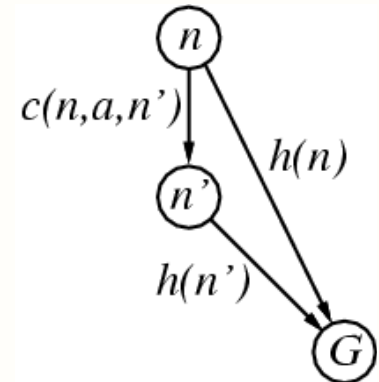
# Consistent heuristics

- A heuristic 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

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



**It's the triangle inequality !**

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

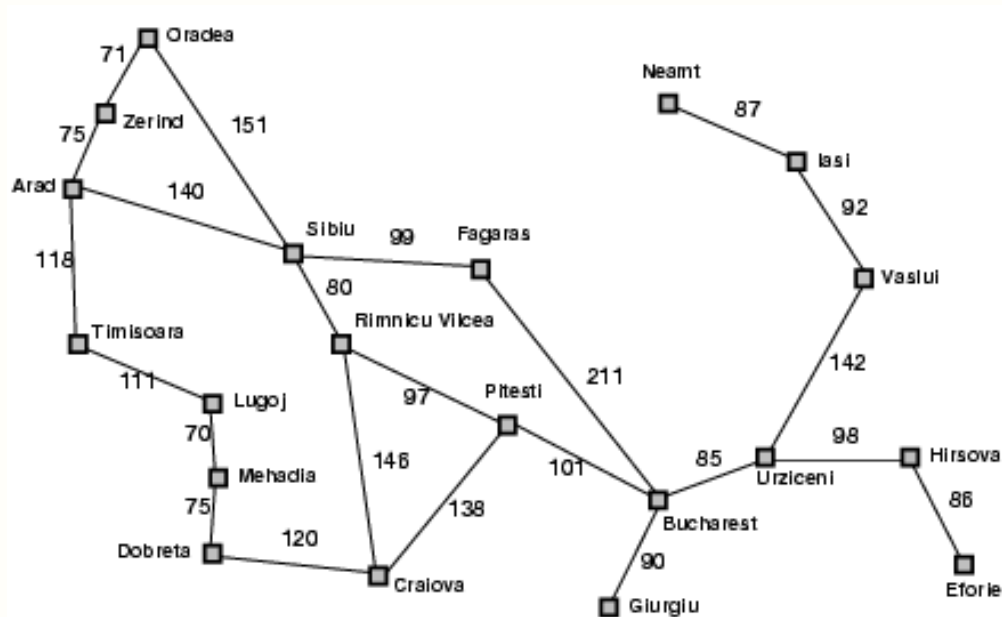
- Theorem:**

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

keeps all checked nodes  
in memory to avoid repeated  
states

# A\* search example

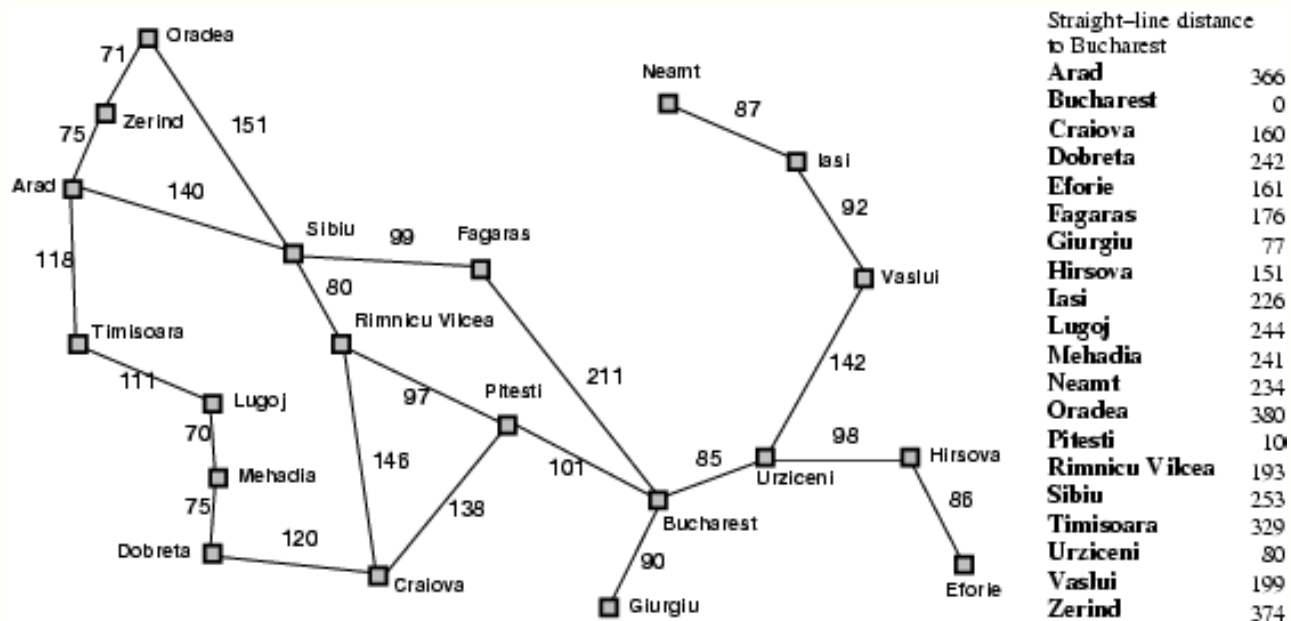
Arad  
366=0+366



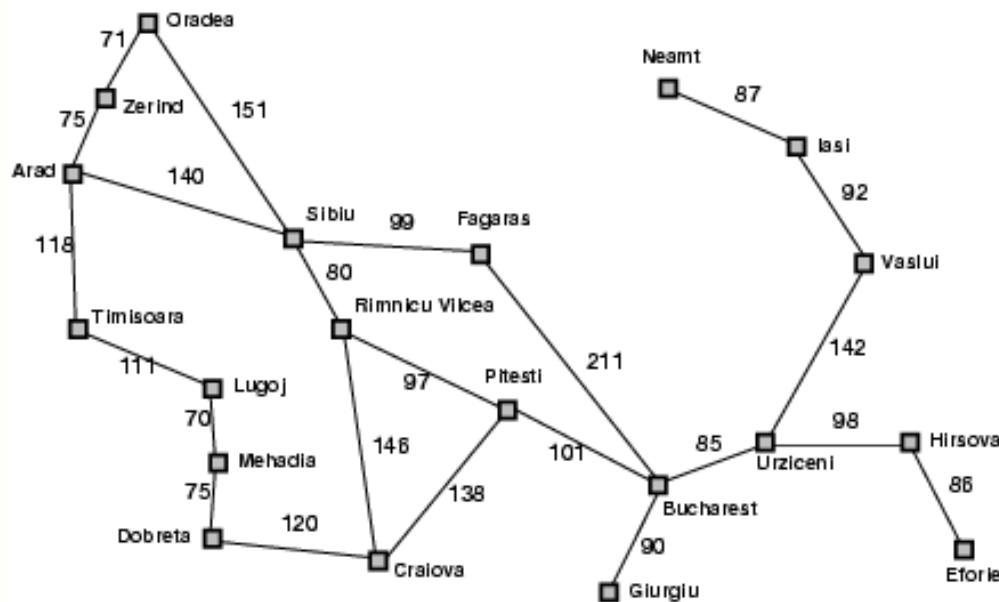
Straight-line distance  
to Bucharest

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# A\* search example



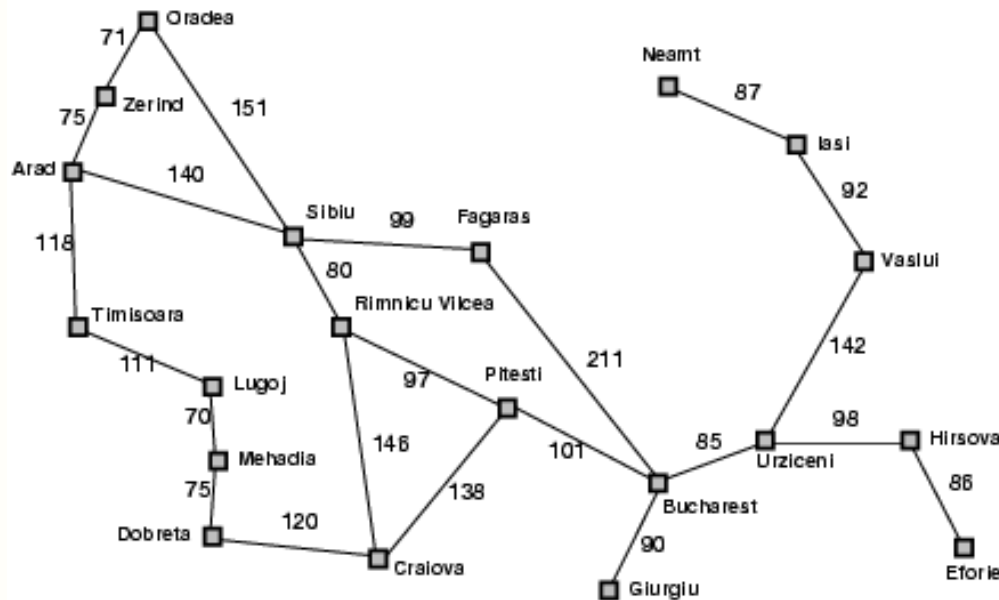
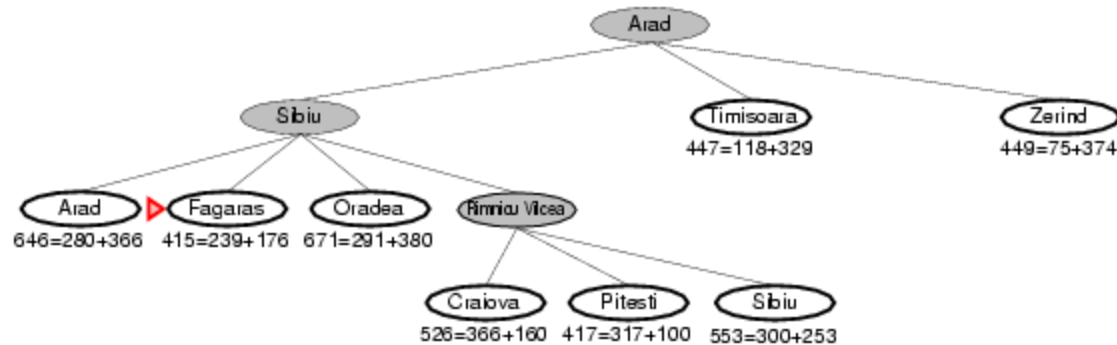
# A\* search example



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# A\* search example

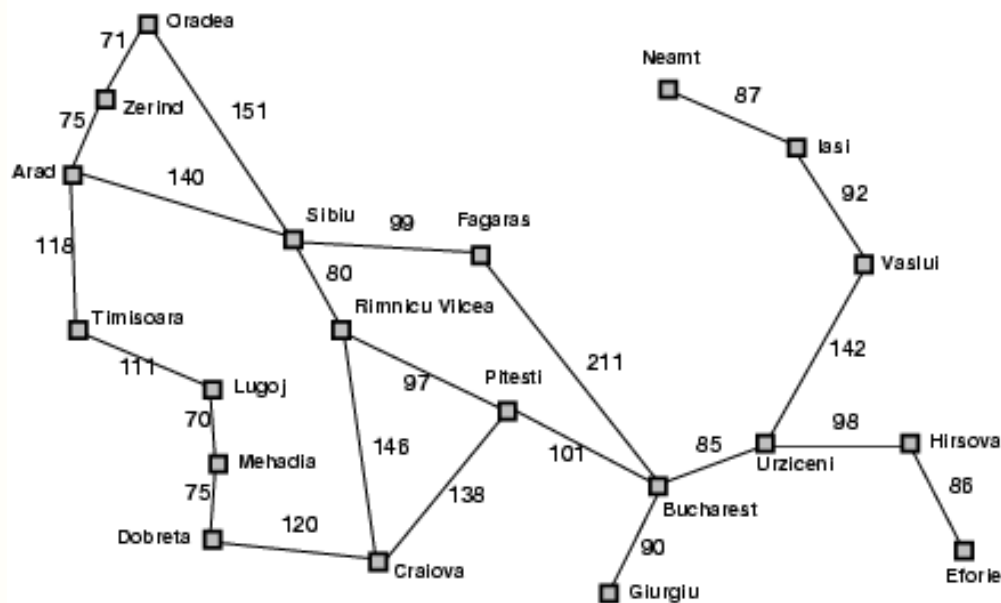
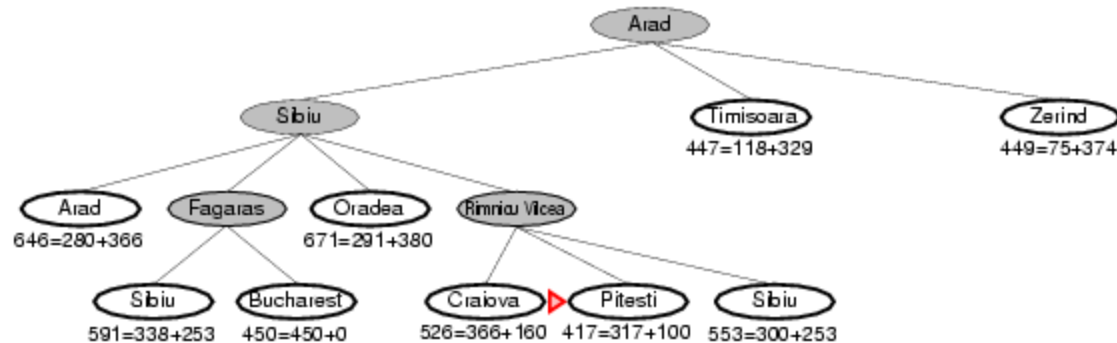


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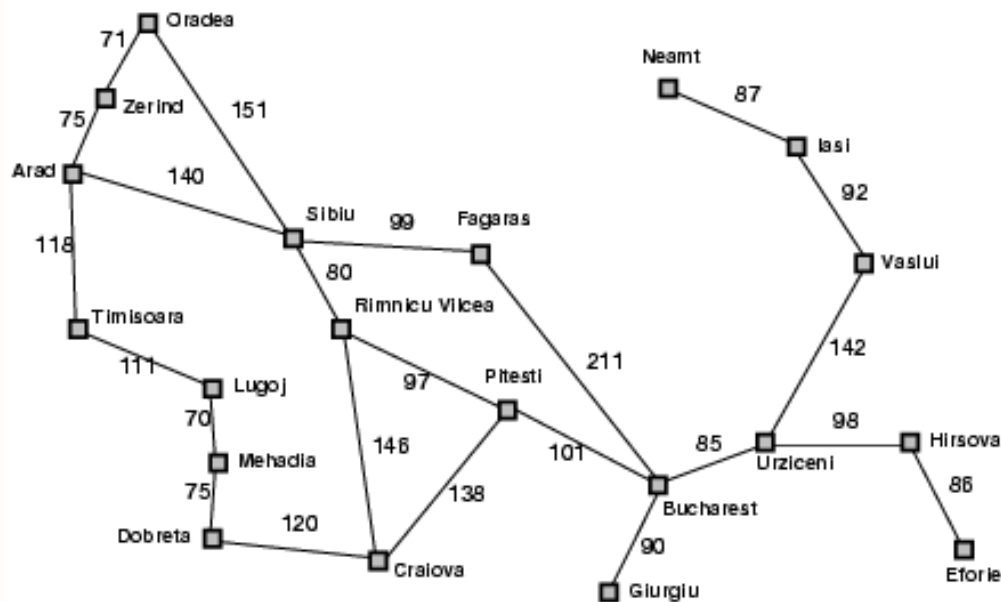
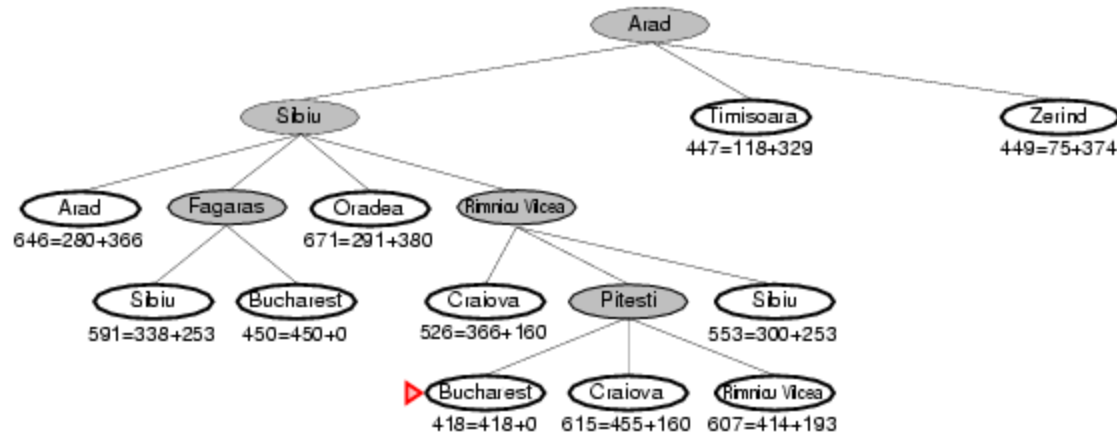
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# Properties of A\*

---

- Complete? Yes (unless there are infinitely many nodes with  $f \leq f(G)$ , i.e. step-cost  $> \epsilon$ )
- Time/Space? Exponential:  $b^d$   
except if:  $|h(n) - h^*(n)| \leq O(\log h^*(n))$
- Optimal? Yes
- Optimally Efficient: Yes (no algorithm with the same heuristic is guaranteed to expand fewer nodes)



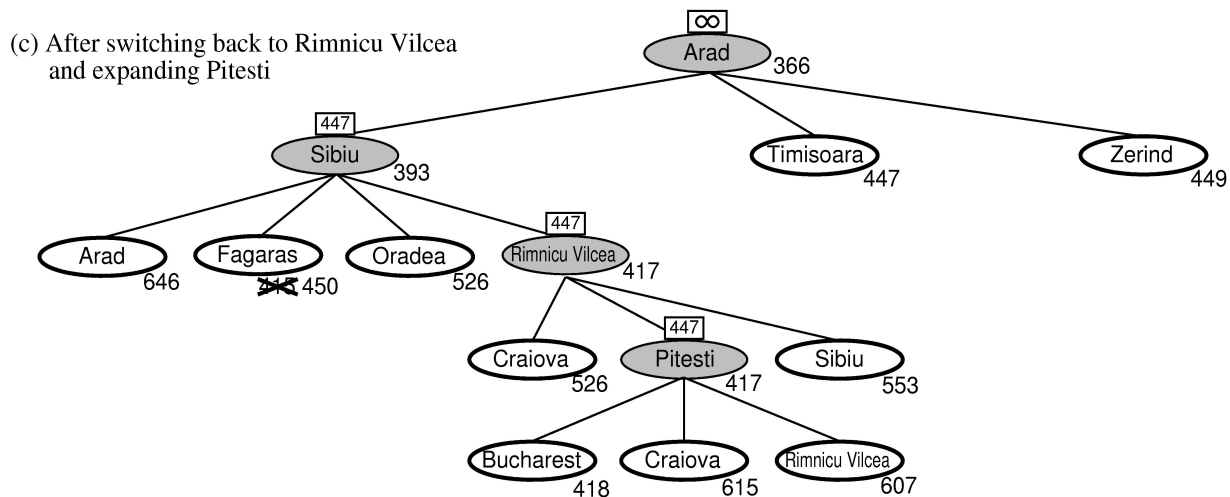
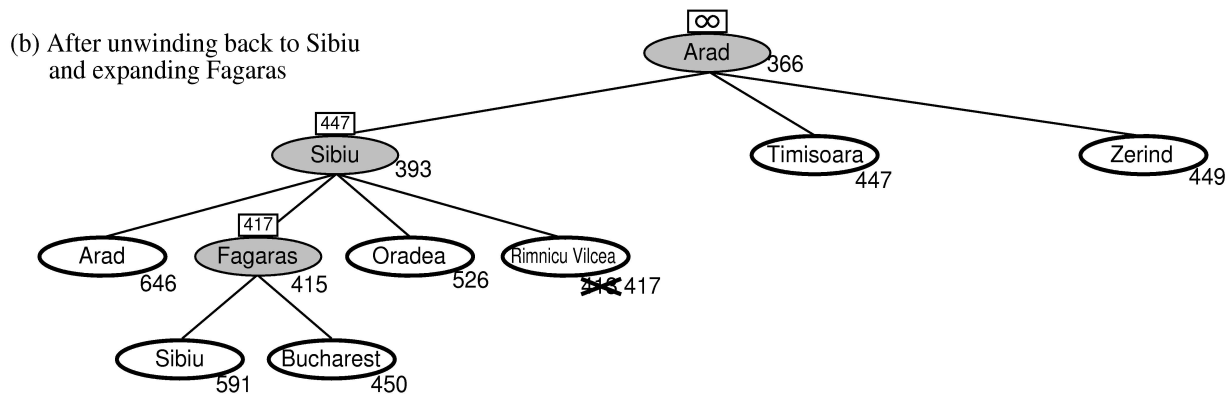
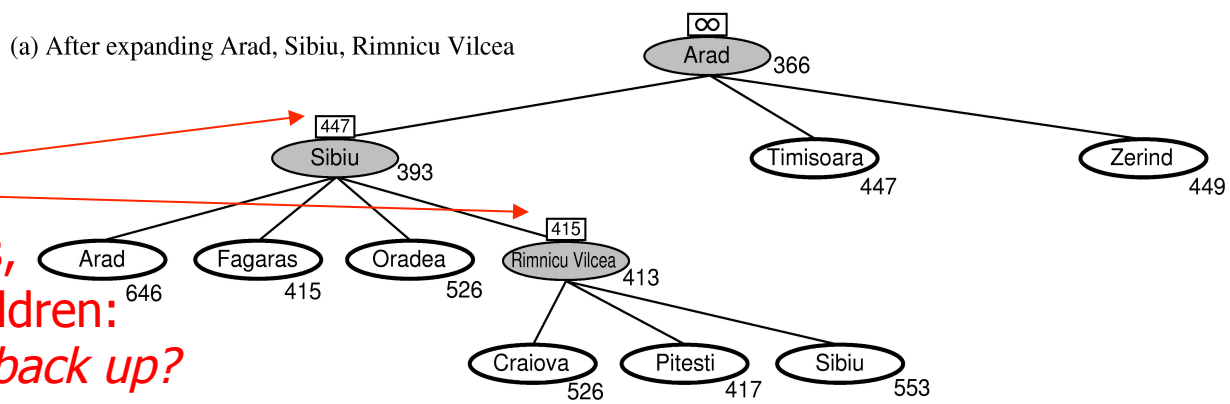
# Memory Bounded Heuristic Search: Recursive BFS

---

- How can we solve the memory problem for  $A^*$  search?
- Idea: Try something like depth first search, but let's not forget everything about the branches we have partially explored.
- *We remember the best  $f$ -value we have found so far in the branch we are deleting.*

# RBFS:

best alternative  
over fringe nodes,  
which are not children:  
*i.e. do I want to back up?*



RBFS changes its mind  
very often in practice.

This is because the  
 $f=g+h$  become more  
accurate (less optimistic)  
as we approach the goal.  
Hence, higher level nodes  
have smaller  $f$ -values and  
will be explored first.

Problem: We should keep  
in memory whatever we can.



# Simple Memory Bounded A\*

---

- This is like A\*, but when memory is full we delete the worst node (largest f-value).
- Like RBFS, we remember the best descendent in the branch we delete.
- If there is a tie (equal f-values) we delete the oldest nodes first.
- simple-MBA\* finds the optimal *reachable* solution given the memory constraint.
- Time can still be exponential.

A Solution is not reachable  
if a single path from root to goal  
does not fit into memory