



# A\* (A-Star)

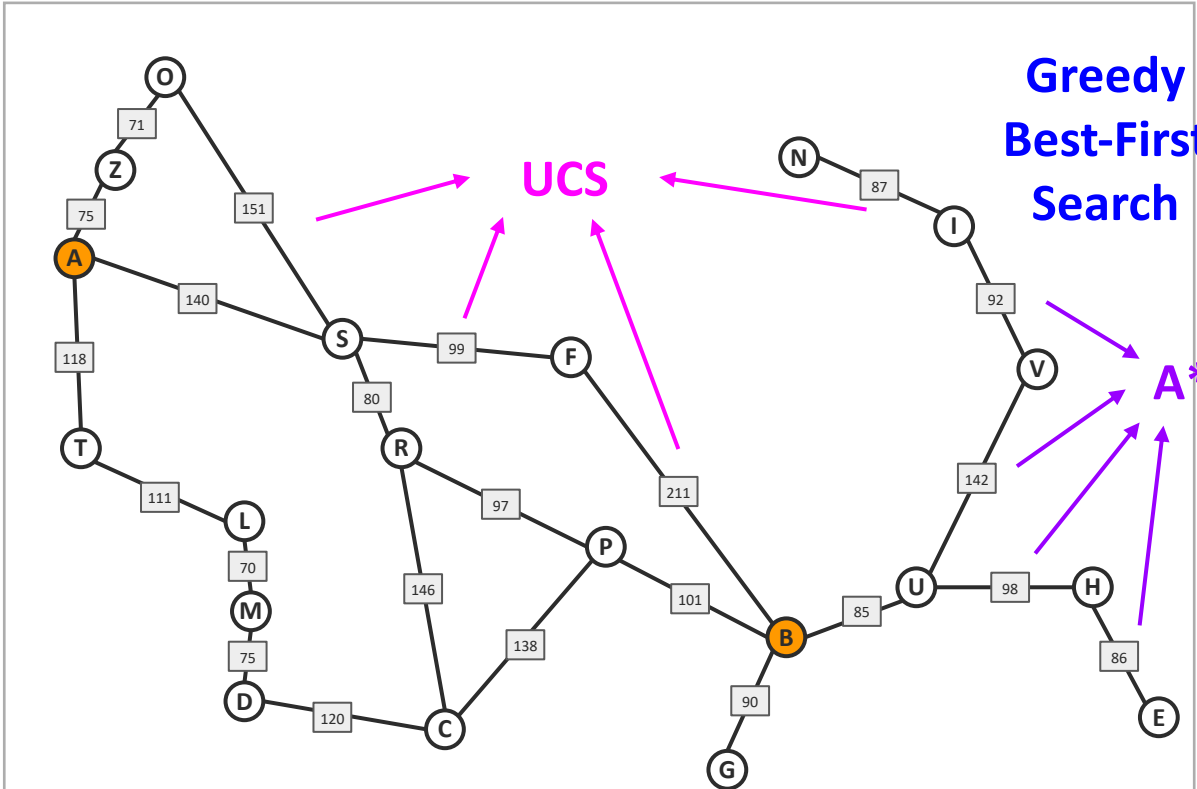
# A\* Search

- Combination of uniform cost search and greedy best-first search
- Assume:
  - **$g(n)$**  is the path cost from the initial state to node  **$n$**
  - **$h(n)$**  is the estimated cost of the shortest path from  **$n$**  to a **goal state** (heuristic function)
- Expand the leaf node that minimizes:

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

- **$f(n)$**  finds the most promising paths from the initial state to a goal

# Traveling from Arad to Bucharest



**Figure 1.** A simplified road map of part of Romania, with road distances in miles.

Arad	366
Bucharest	0
Craiova	160
Drobeta	242
Eforie	161
Fagaras	176
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**Table 1.** Heuristic function: straight-line distances to Bucharest.

Greedy  
Best-First  
Search

UCS

A\*

# A\* search

- Strategy:
  - Expand the node with the lowest  $f(n)$
  - $f(n) = g(n) + h(n)$
- Implementation
  - Fringe (set of leaf nodes) is a priority queue
    - Priority is given to the nodes with lowest  $f(n)$

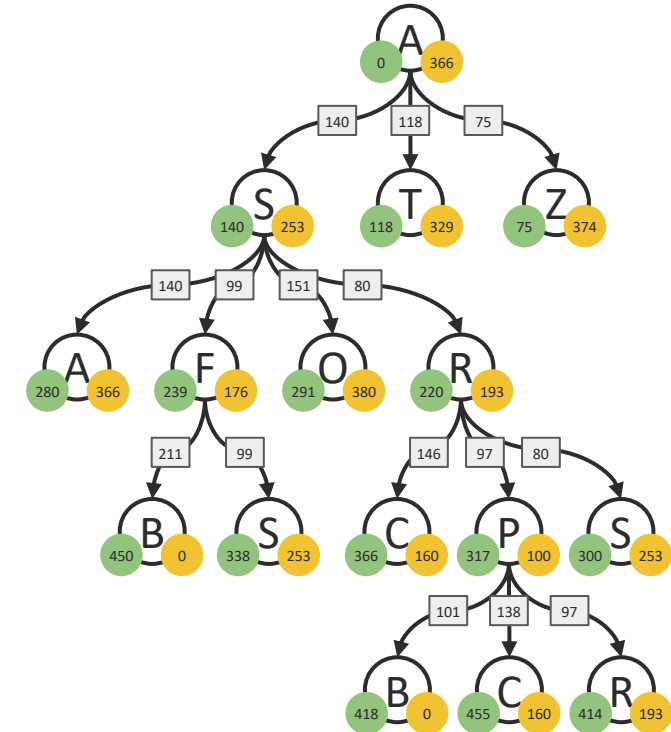


Figure 2. Search tree.

# A\* search

```
set initial state of problem as the tree root
```

```
while True:
```

```
    choose node with lowest  $f(n)$  value for expansion
```

```
    if there are no candidates for expansion
```

```
        return failure
```

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    if chosen node is a goal state
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```
        return path from root to node
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```
    else
```

```
        expand node and add neighbors to the tree
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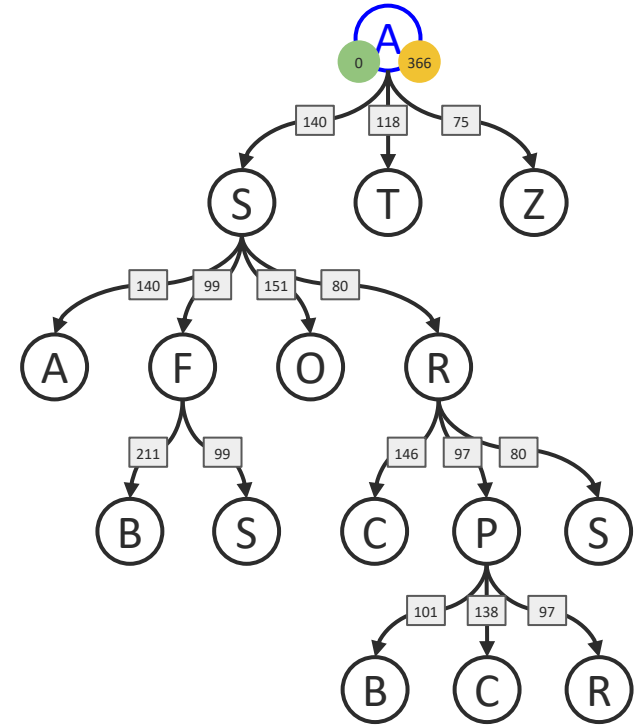


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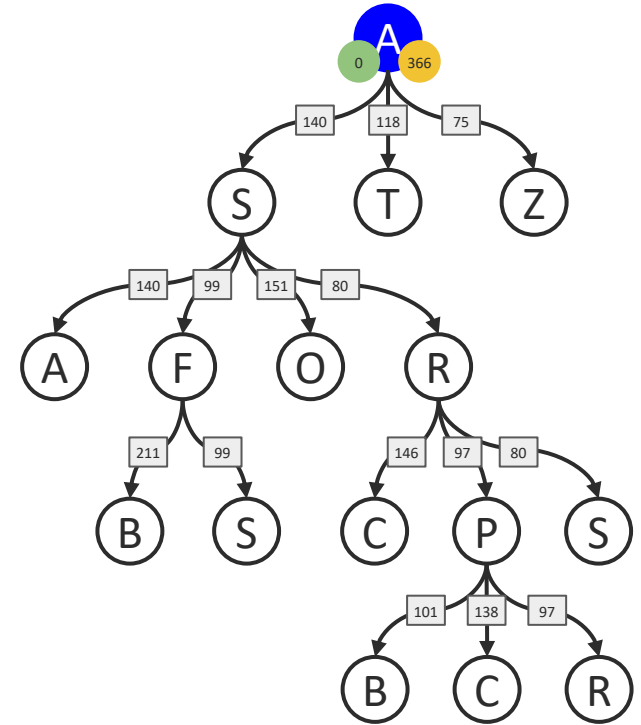
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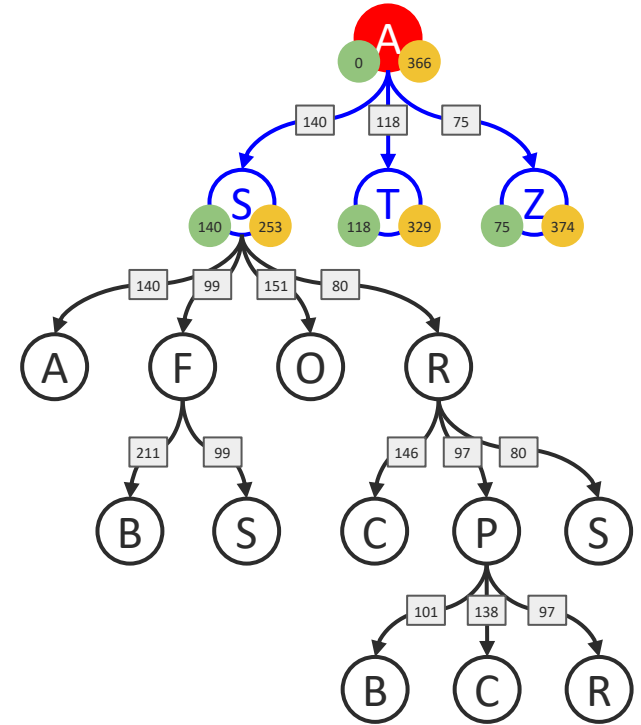


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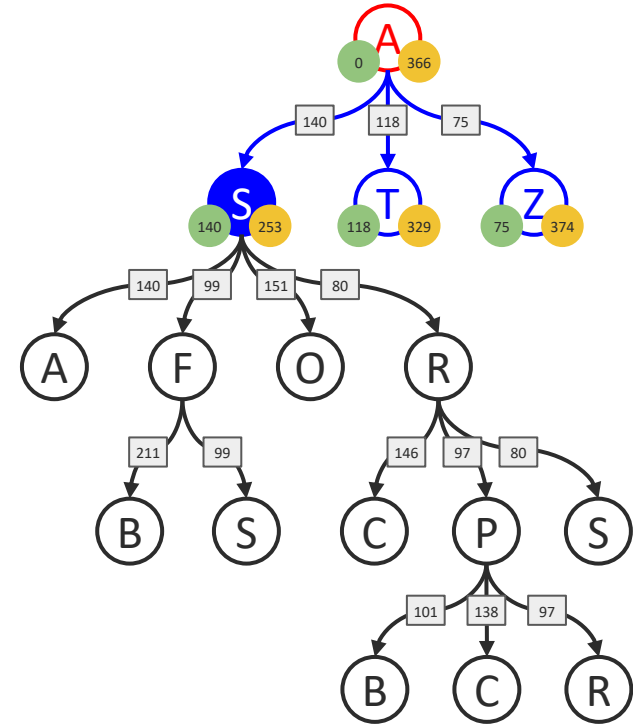


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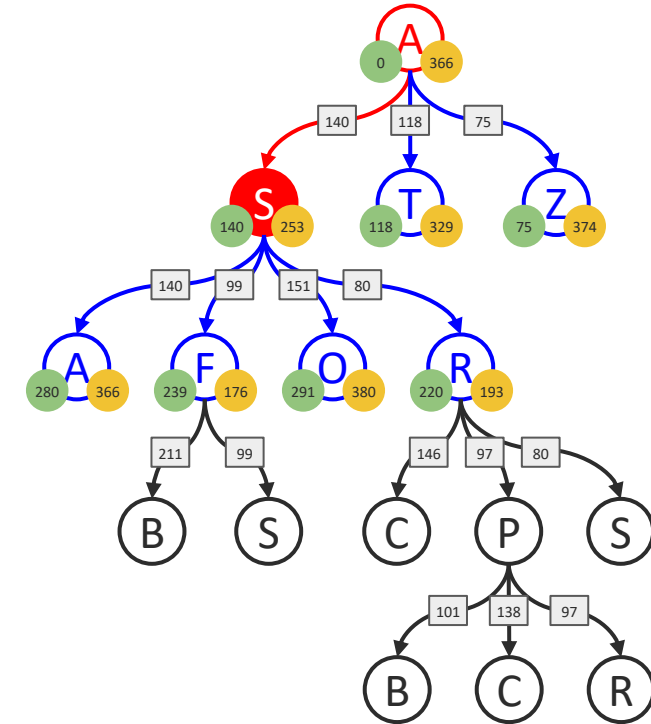


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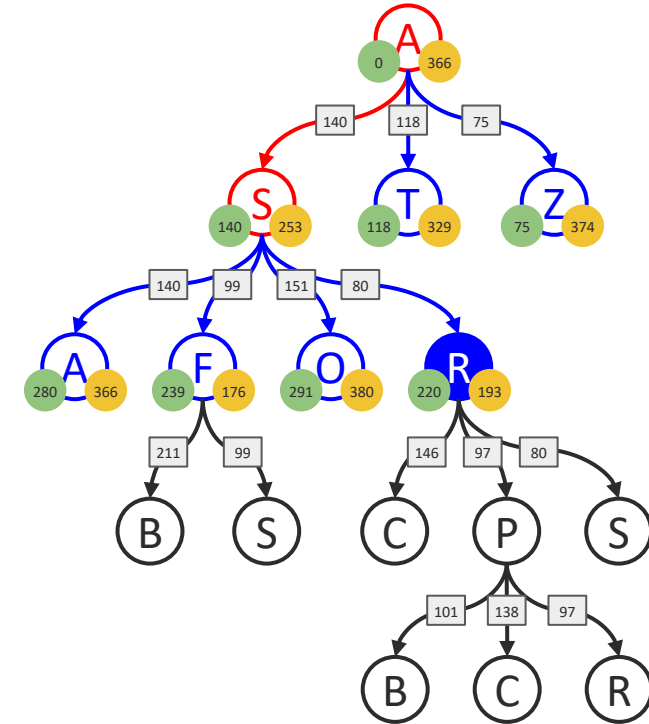


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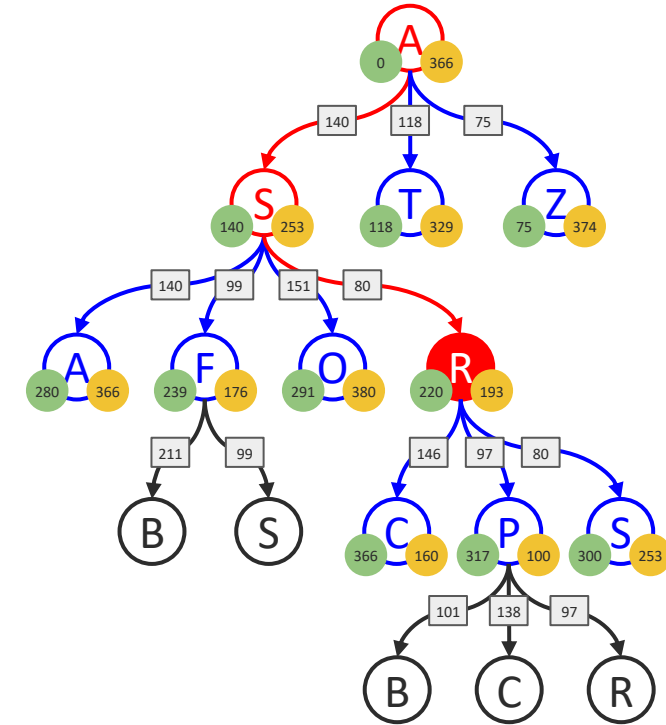


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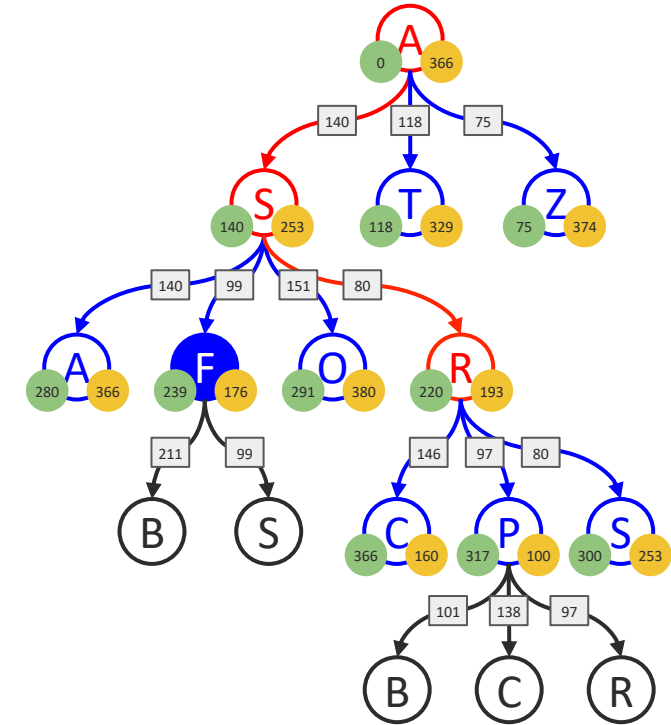


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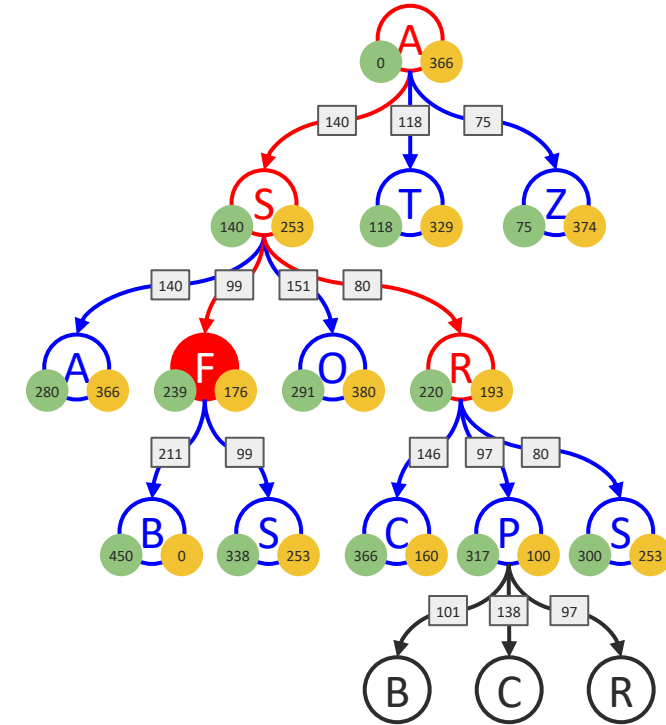


Figure 3. Search tree.

# Knowledge Check 1



Can we stop A\* as soon as a goal state is added to the tree?

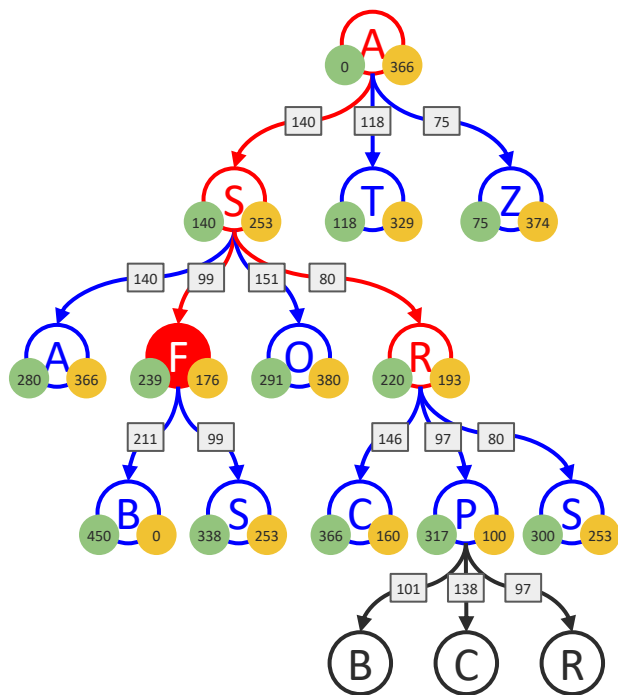


Figure 4. Search tree

A

Yes

B

No

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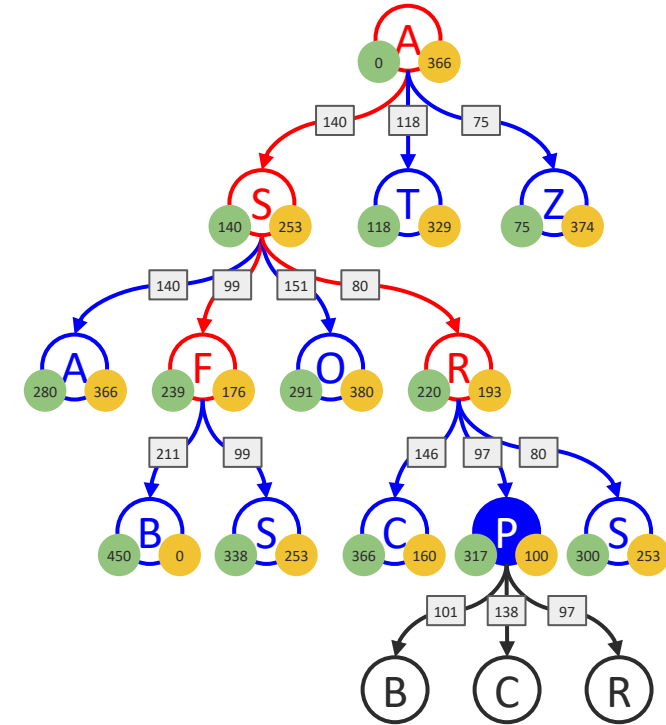


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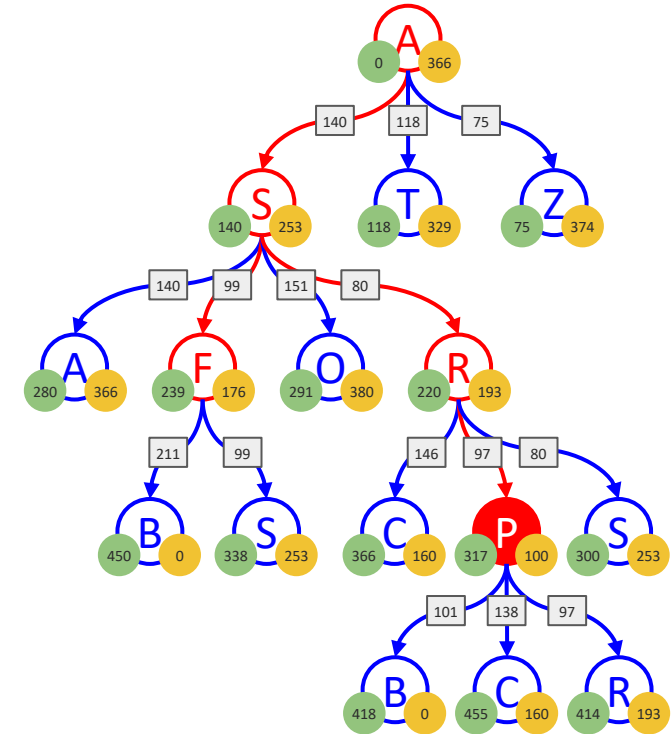


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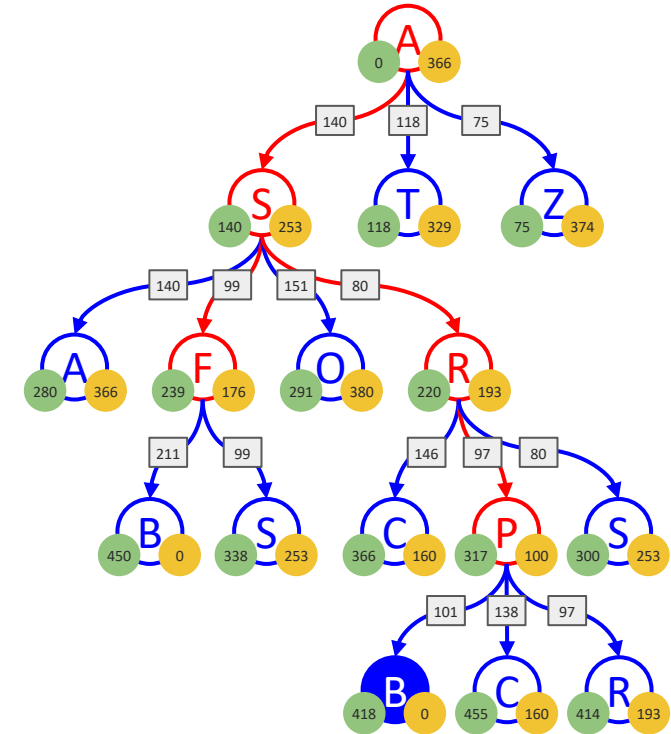


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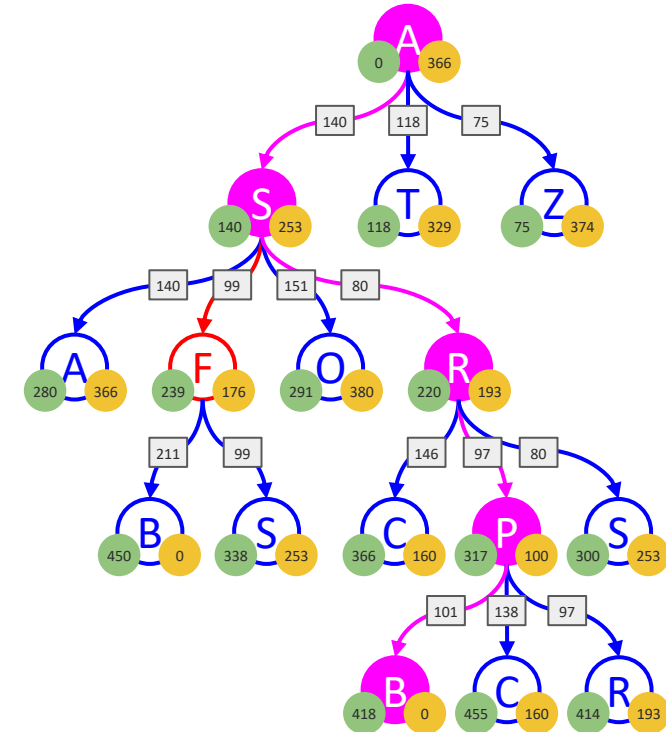
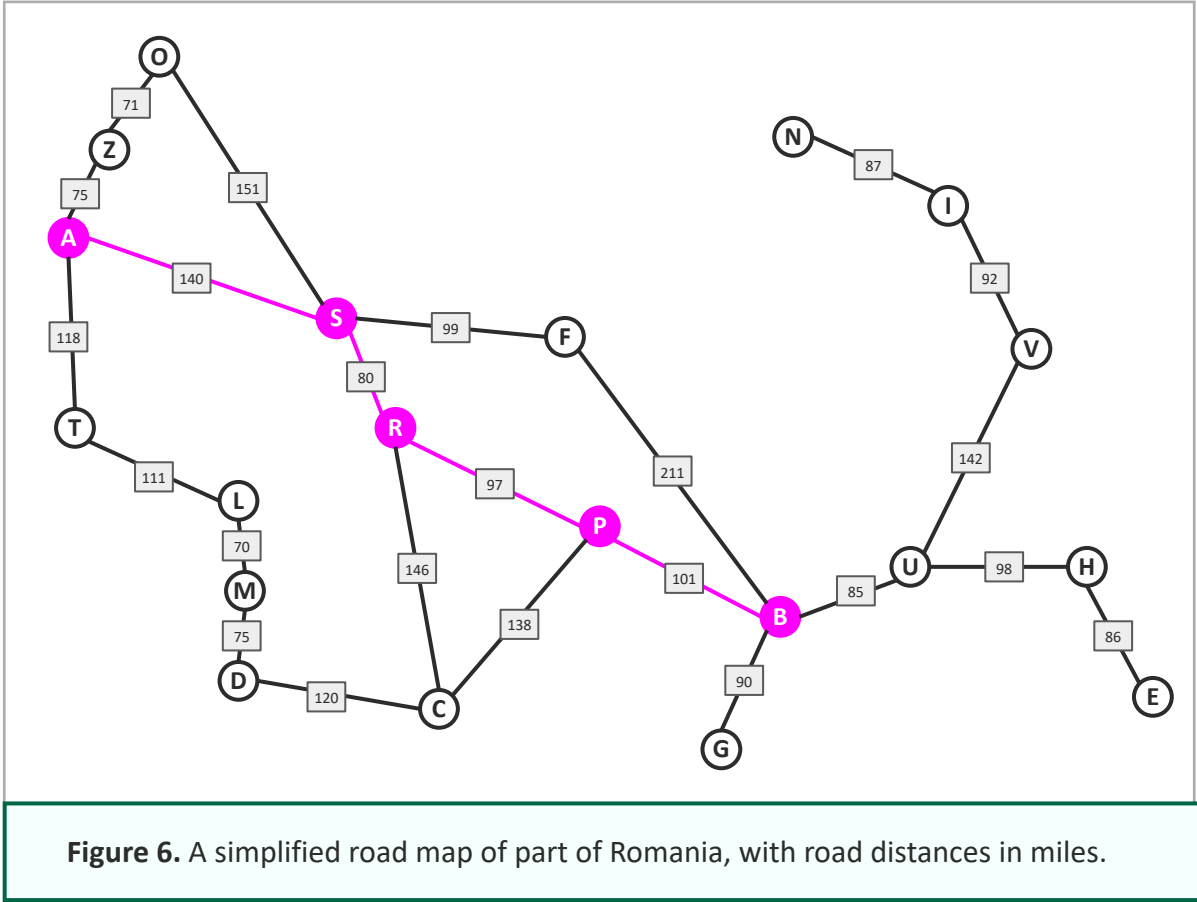


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# Knowledge Check 2



Is A\* complete?

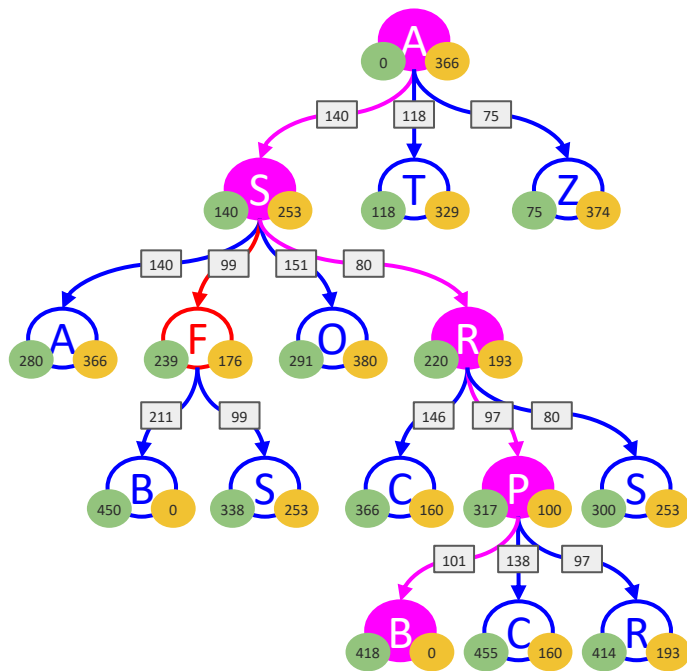


Figure 7. Search tree

A

Yes

B

No

# Admissible heuristics

- $A^*$  being cost-optimal depends on certain properties of the heuristic
- The key property is admissibility
  - An admissible heuristic is one that never overestimates the lowest cost to reach a goal
  - An admissible heuristic is therefore optimistic
- With an admissible heuristic,  $A^*$  is cost-optimal

# Knowledge Check 3



The heuristic  $h(n) = 0$  for all nodes is admissible, although not very useful. If we use this heuristic function, A\* will behave like which search algorithm?

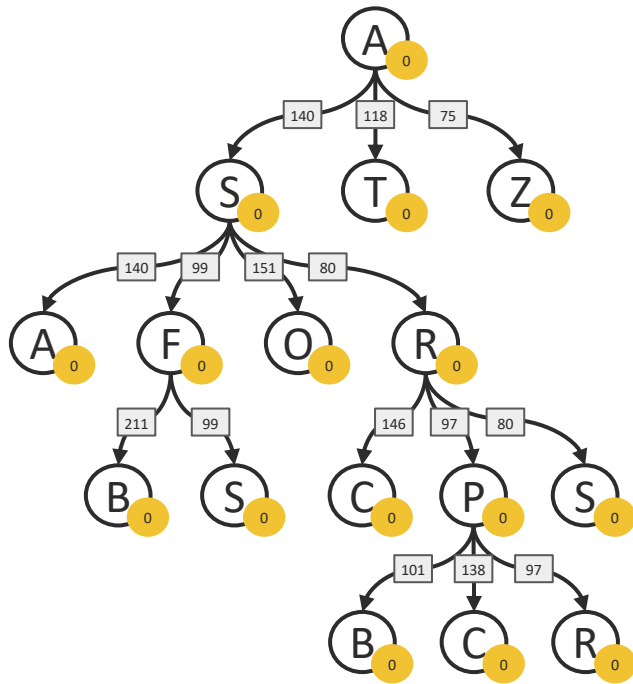


Figure 8. Search tree.

A

DFS

B

BFS

C

UCS

D

Iterative Deepening

# Consistent heuristics

- A heuristic  $h(n)$  is consistent if, for every node  $n$  and every successor  $n'$  of  $n$  generated by an action  $a$ , we have:

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

- Every consistent heuristic is admissible (but not vice versa)
- With a consistent heuristic, the first time we reach a state it will be on an optimal path (we can stop early)

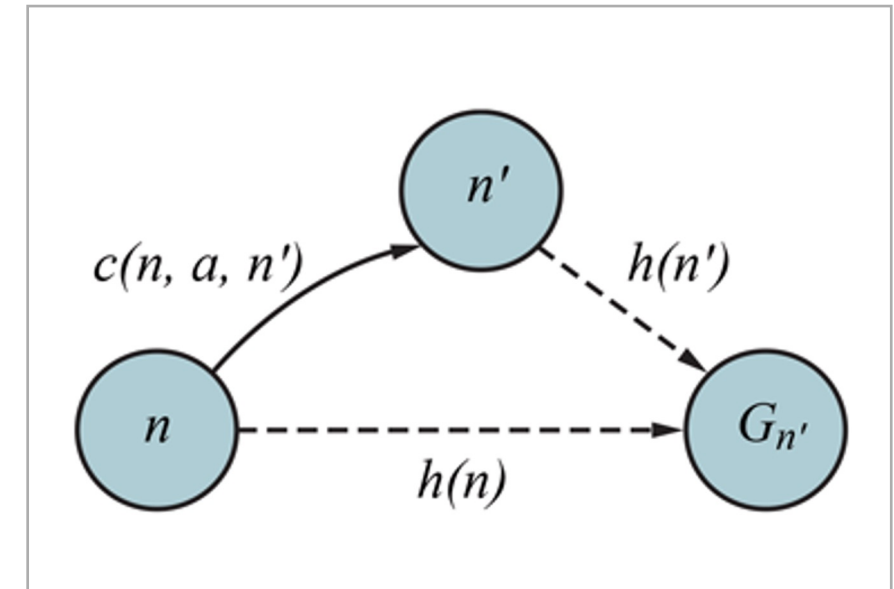


Figure 9. Triangle inequality.

# Inadmissible heuristics

- With an inadmissible heuristic,  $A^*$  may be cost-optimal
  - If there is even one cost-optimal path on which  $h(n)$  is admissible for all nodes  $n$  on the path
  - If the optimal solution has cost  $C^*$ , and the second-best has cost  $C_2$ , and if  $h(n)$  overestimates some costs, but never by more than  $C_2 - C^*$



# Weighted A\*

- A\* search has many good qualities, but it expands a lot of nodes
- We can explore fewer nodes if we are willing to accept solutions that are “good enough”
- If we use an inadmissible heuristic, we risk missing the optimal solution, but potentially reduce the number of nodes expanded
- We can weight the heuristic value more heavily

$$f(n) = g(n) + W \times h(n), \text{ for some } W > 1$$

# A\* Analysis



Is it complete? Is it guaranteed to find a solution if one exists?



What is its time complexity?



Is it optimal? Is it guaranteed to find the least cost path?



What is its space complexity?

# A\* Analysis



Is it complete? Is it guaranteed to find a solution if one exists?

- Yes, as long as all action costs are greater than zero and the state space either has a solution or is finite



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- In the worst case, is equivalent to UCS -  $O(b^{C/\epsilon})$



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You have reached the end  
of the lecture.



## Image/Figure References

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