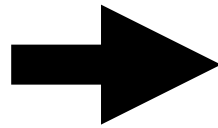
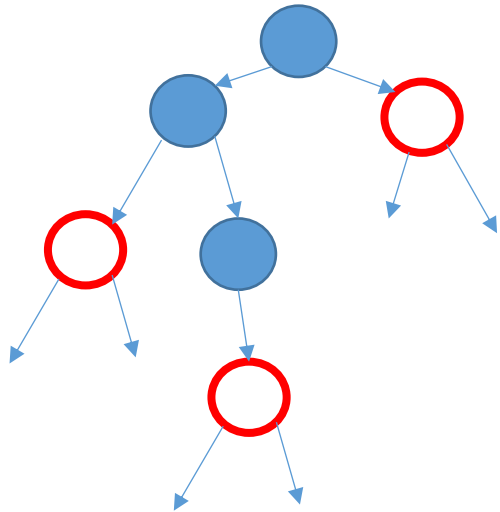


# CS440/ECE448 Lecture 3: Search Order

Slides by Mark Hasegawa-Johnson, 1/2020

Including some slides written by Svetlana Lazebnik, 9/2016

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# Outline

- Uniform cost search (UCS) = slightly different implementation of Dijkstra's Algorithm
- Breadth-first search (BFS) = special case of UCS
- Depth-first search (DFS)

# Dijkstra's Shortest Path Algorithm

- Initialize:
  - $d_{nl}$  = distance from n to l
  - $V_n = \infty$  for all vertices n
  - Unvisited = {all nodes but start}
  - k = Start Node
- While Goal  $\in$  Unvisited
  - For n  $\in$  Neighbor(k)
    - $V_n = \min(V_n, V_k + d_{kn})$
  - $k \leftarrow \operatorname{argmin}_{l \in \text{Unvisited}} V_l$



# Uniform Cost Search

- Initialize:
  - $d_{nl}$  = distance from n to l
  - $V_k = 0$  for start\_node k, only
  - Frontier = {}
  - k = Start Node
- While Goal  $\neq k$ 
  - For n  $\in$  Neighbor(k)
    - Frontier  $\leftarrow n: \min(V_n, V_k + d_{kn})$
  - $k \leftarrow \underset{l \in \text{Frontier}}{\text{argmin}} V_l$



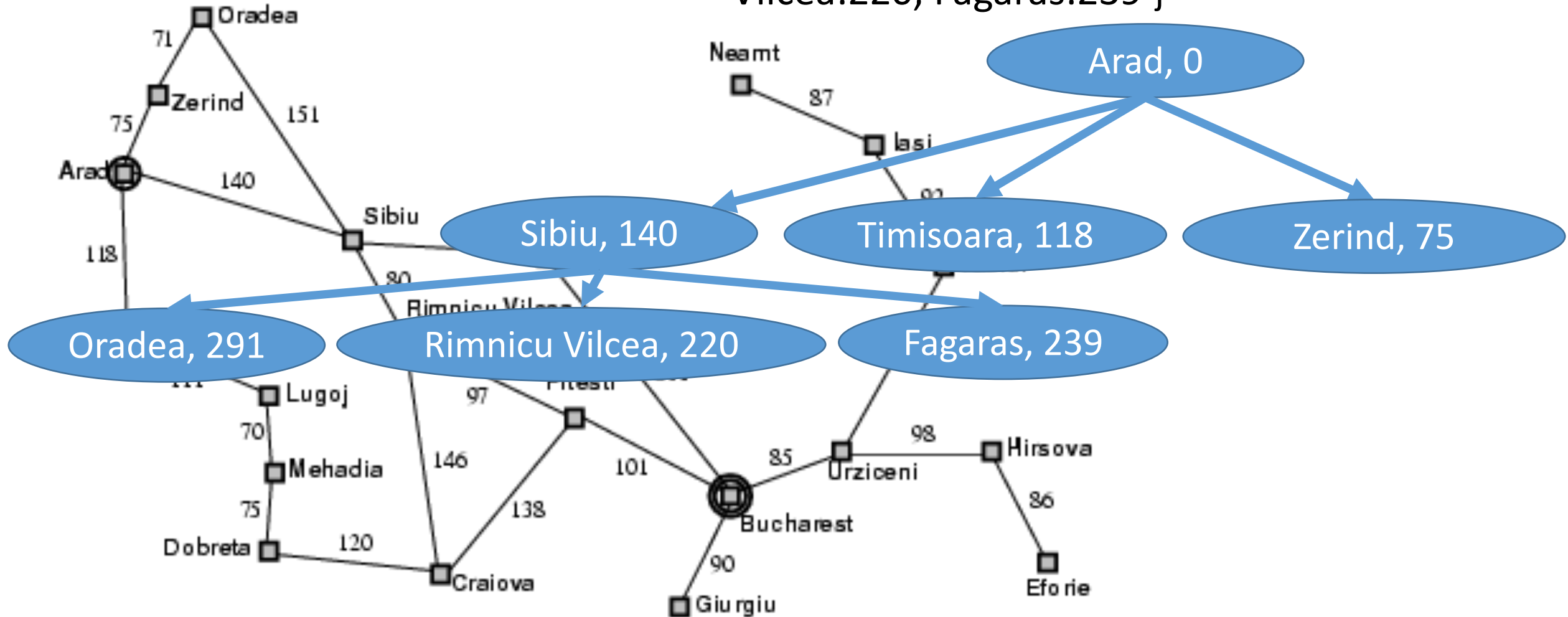
# Dijkstra's algorithm vs. Uniform Cost Search

- They evaluate the same nodes  $k$ , in exactly the same order.
- They give the same (minimum-cost) path as a result.
- The only difference:
  - Dijkstra's algorithm keeps track of  $V_n$  for all nodes in the search space
  - UCS keeps track of  $V_n$  only for nodes you've explored

# Example: Romania

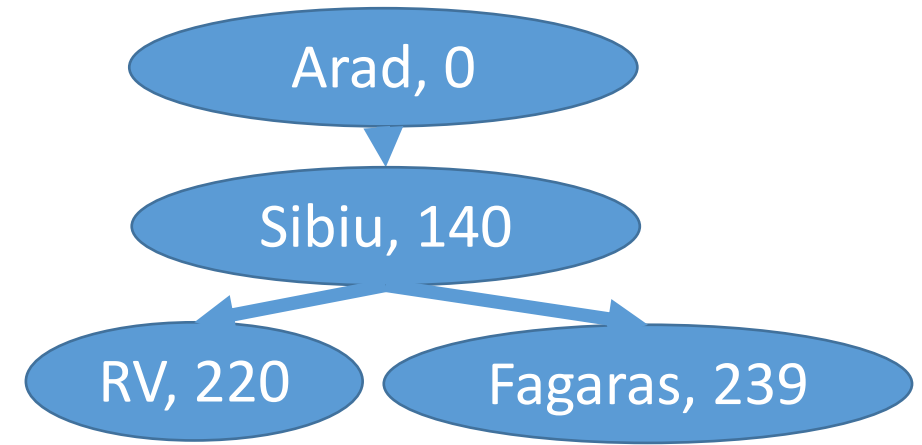
Frontier: { Zerind:75, Timisoara:118, Oradea:291, Rimnicu Vilcea:220, Fagaras:239 }

Explored: { Arad:0, Sibiu:140, Zerind:75, Timisoara:118, Oradea:291, Rimnicu Vilcea:220, Fagaras:239 }



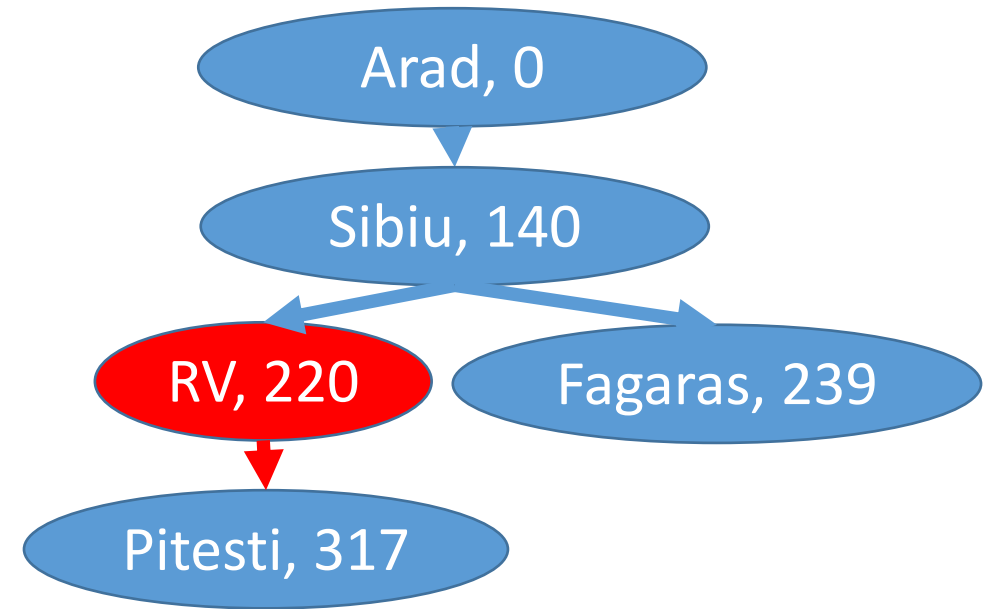
# Why it Works

- $d_{nl} \geq 0$  means that every node on the best path to the Goal,  $G$ , has a cost,  $V_n$ , less than or equal the cost of the goal,  $V_n \leq V_G$



# Why it Works

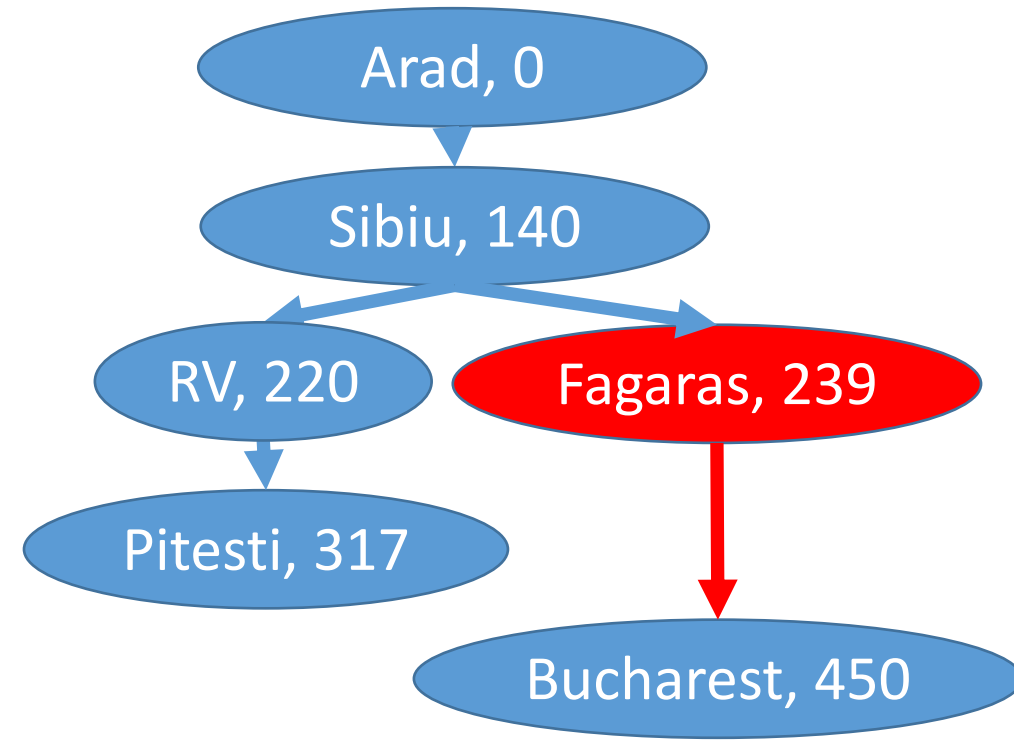
- $k \leftarrow \underset{l \in \text{Frontier}}{\text{argmin}} V_l$  means that the lowest-cost nodes are expanded first





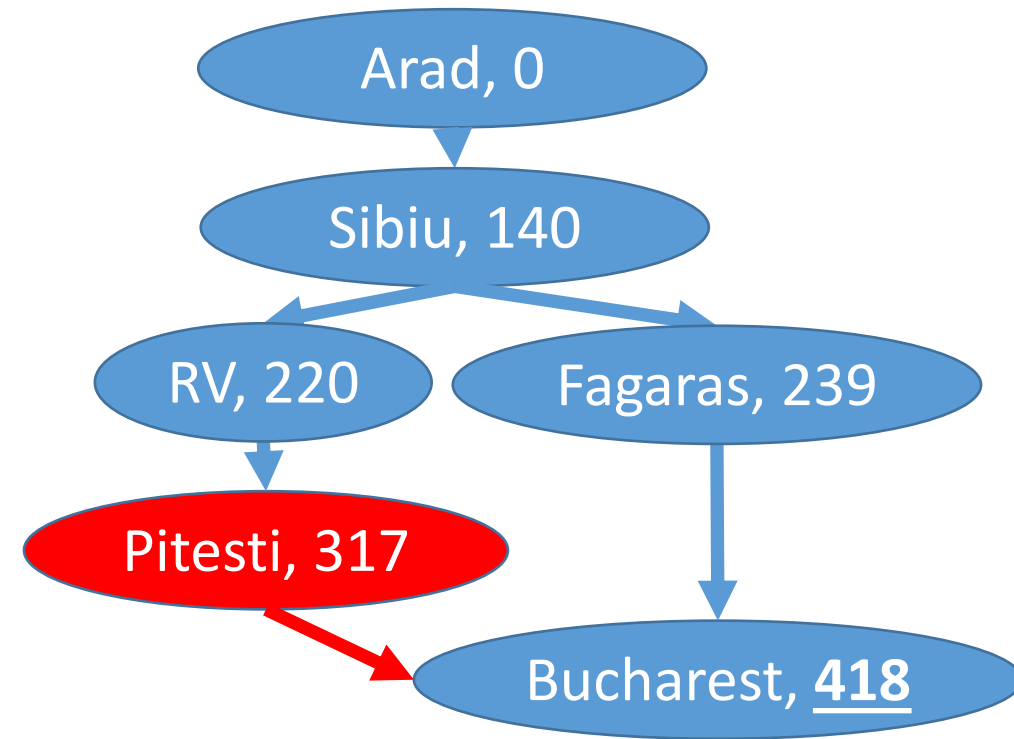
# Why it Works

- We don't end when Goal is placed on the **Frontier**, we only end when Goal is **expanded**



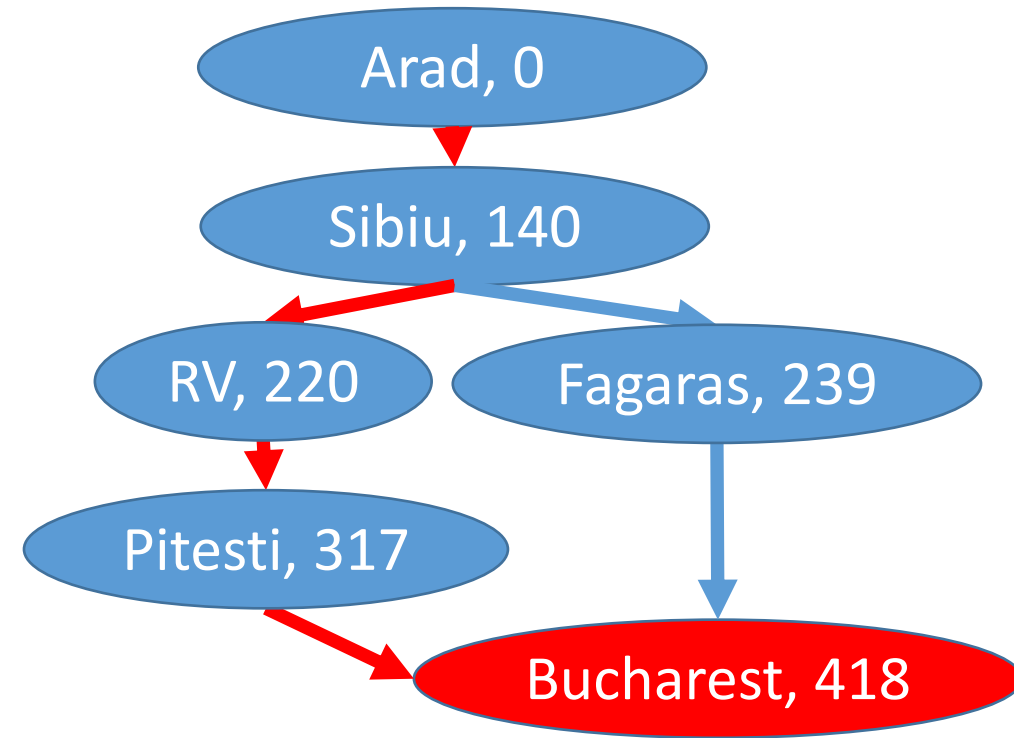
# Why it Works

- $k \leftarrow \underset{l \in \text{Frontier}}{\text{argmin}} V_l$  means that every predecessor with a cost less than  $V_G$  is expanded before the goal is expanded



# Why it Works

- Therefore we always find the shortest path



# Computational Considerations

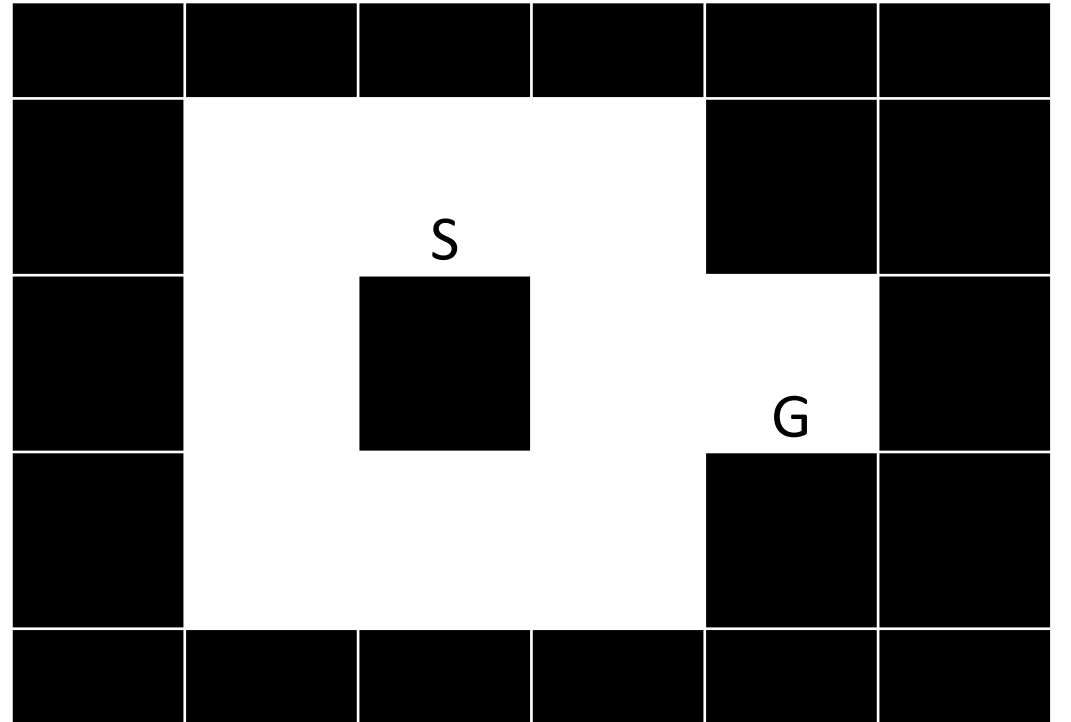
- Suppose there are  $N$  world states
- $k \leftarrow \operatorname{argmin}_{l \in \text{Frontier}} V_l$ 
  - Naïve implementation: requires you to sort through the whole frontier, to find the smallest. Complexity:  $O\{N\}$  per search step!!
  - Better implementation: keep the frontier sorted, as a **priority queue**. Then complexity is  $O\{\log N\}$  to insert a node into the frontier, and  $O\{1\}$  to retrieve the minimum.
- $\text{Frontier} \leftarrow n: \min(V_n, V_k + d_{kn})$ 
  - “Explored list” is a **hash table** (python: a dict), so that, given a world state  $n$ , you can immediately tell ( $O\{1\}$ ) whether or not that state has been explored.
  - Each state in the “Explored list” has a pointer to corresponding state in the Frontier, if that state is still in the frontier ( $O\{1\}$ ).

# Outline

- Uniform cost search (UCS) = slightly different implementation of Dijkstra's Algorithm
- **Breadth-first search (BFS) = special case of UCS**
- **Depth-first search (DFS)**

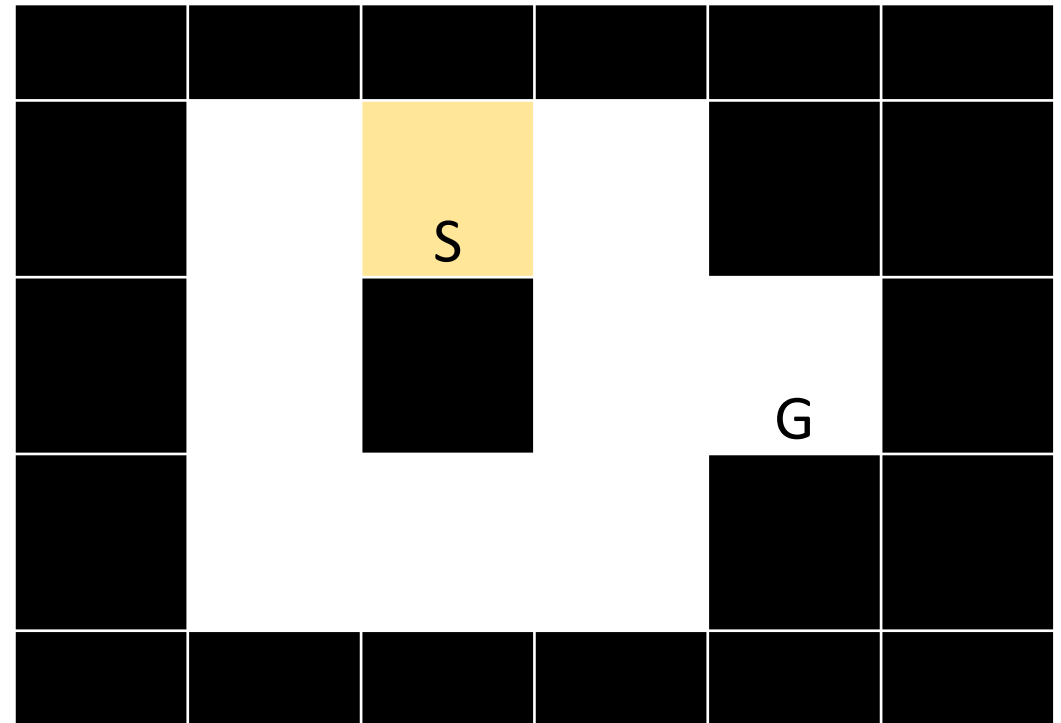
# Breadth-first search (BFS) = special case of UCS

- ... when every step has exactly the same cost,  $d_{nl} = 1$
- Example: solving a maze



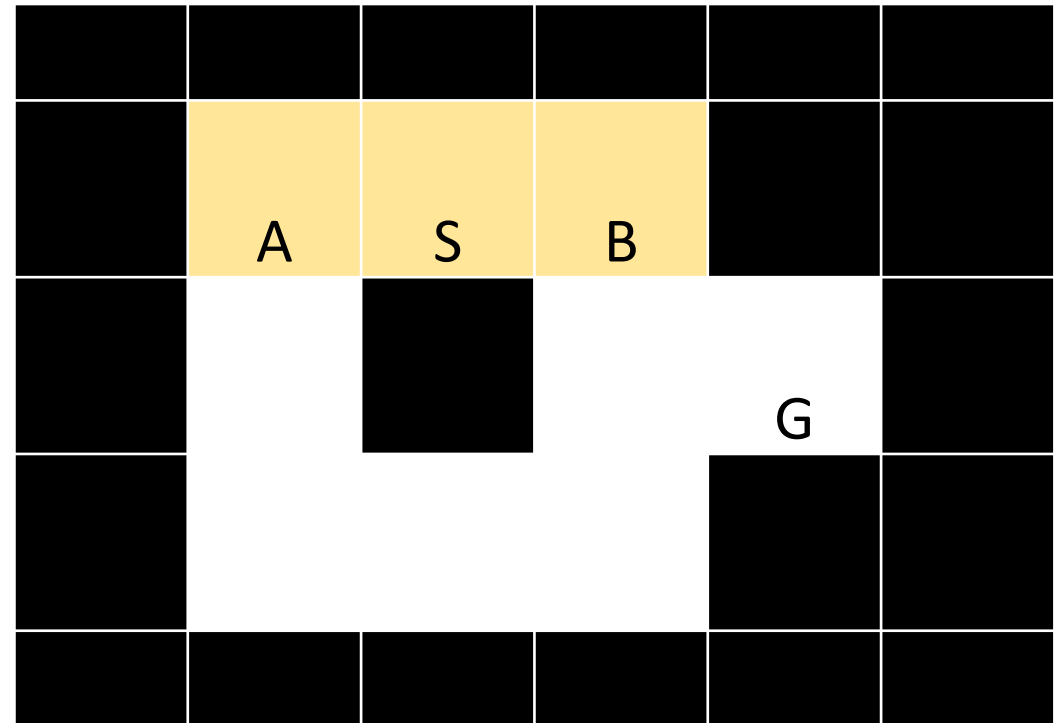
# BFS Computational Savings

- Frontier doesn't have to be a priority queue
- It can just be a regular first-in, first-out (FIFO) queue
- Example: Frontier = {S}



# BFS Computational Savings

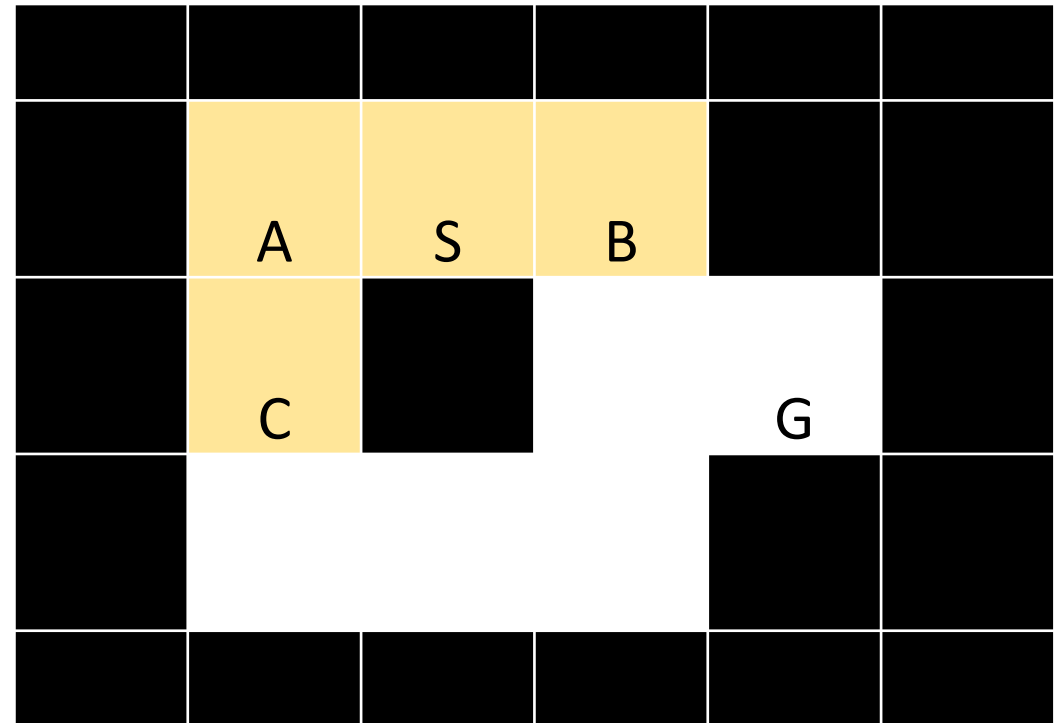
- Frontier doesn't have to be a priority queue
- It can just be a regular first-in, first-out (FIFO) queue
- Example: Frontier = {A,B}
- Pop A from the queue, expand it





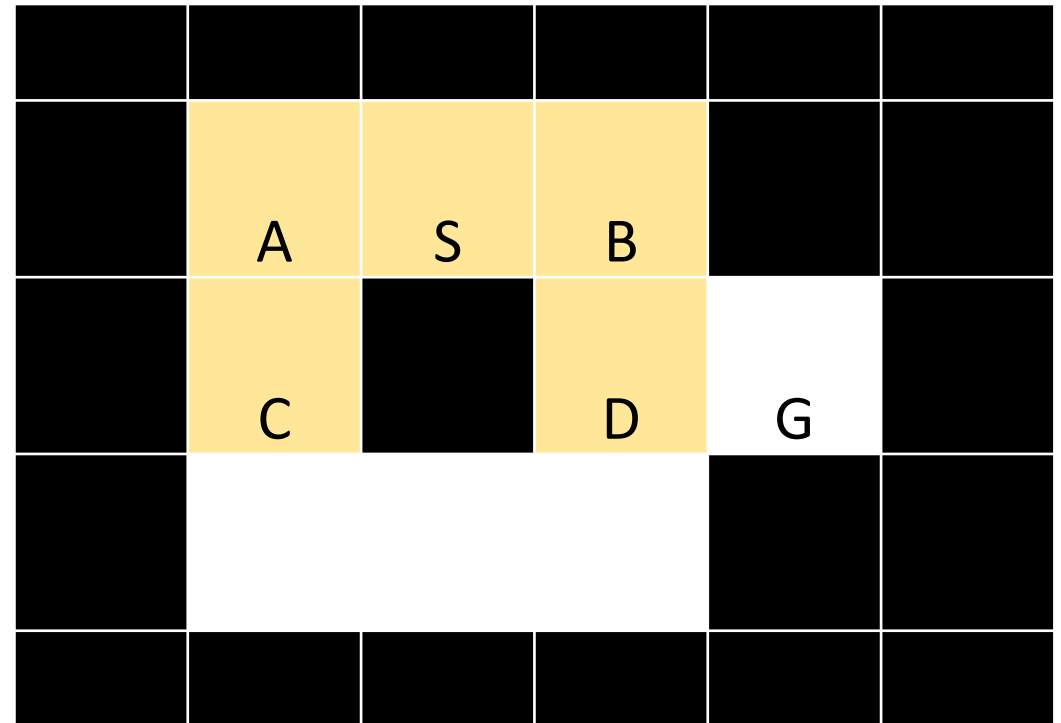
# BFS Computational Savings

- Frontier doesn't have to be a priority queue
- It can just be a regular first-in, first-out (FIFO) queue
- Example: Frontier = {B,C}
- Pop B from the queue, expand it



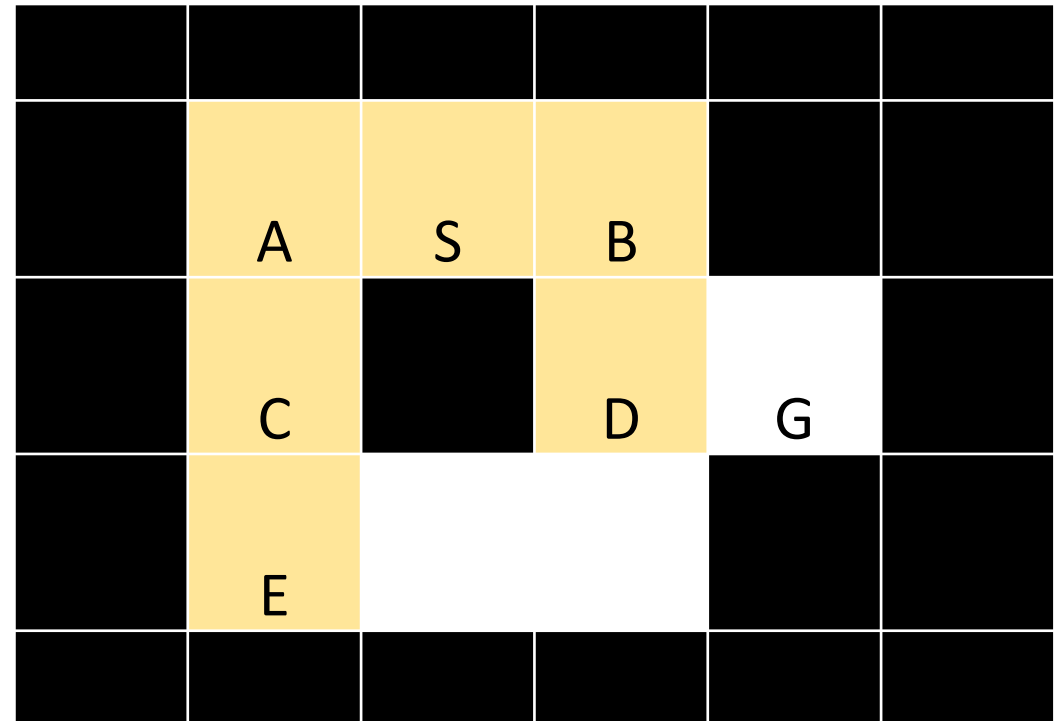
# BFS Computational Savings

- Frontier doesn't have to be a priority queue
- It can just be a regular first-in, first-out (FIFO) queue
- Example: Frontier = {C,D}
- Pop C from the queue, expand it



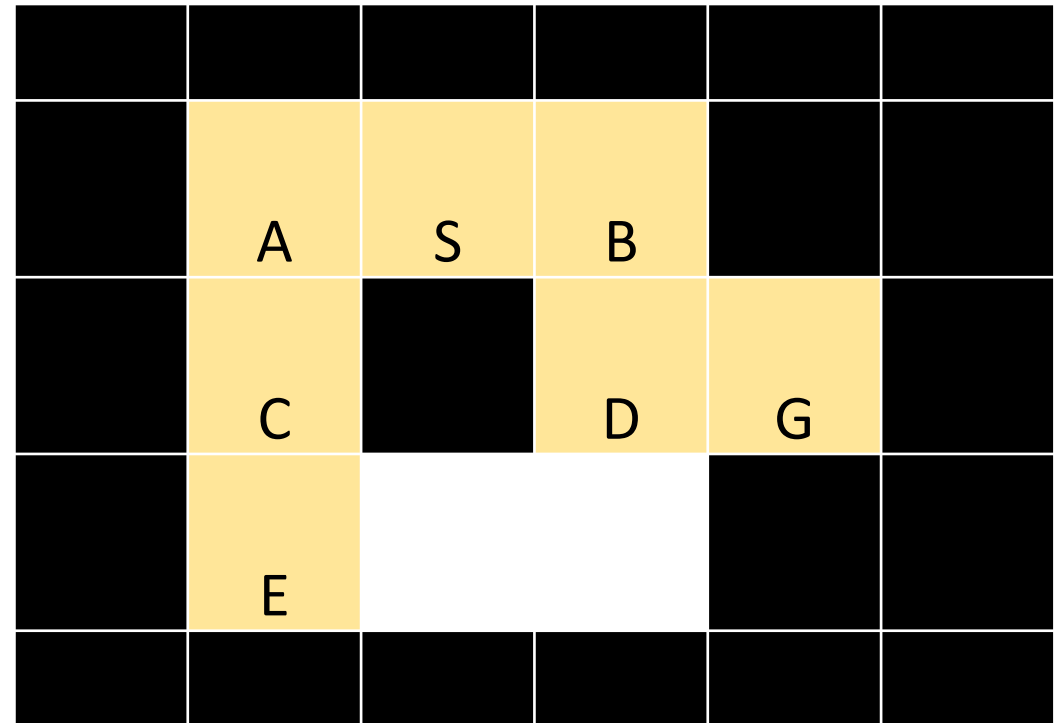
# BFS Computational Savings

- Frontier doesn't have to be a priority queue
- It can just be a regular first-in, first-out (FIFO) queue
- Example: Frontier = {D,E}
- Pop D from the queue, expand it



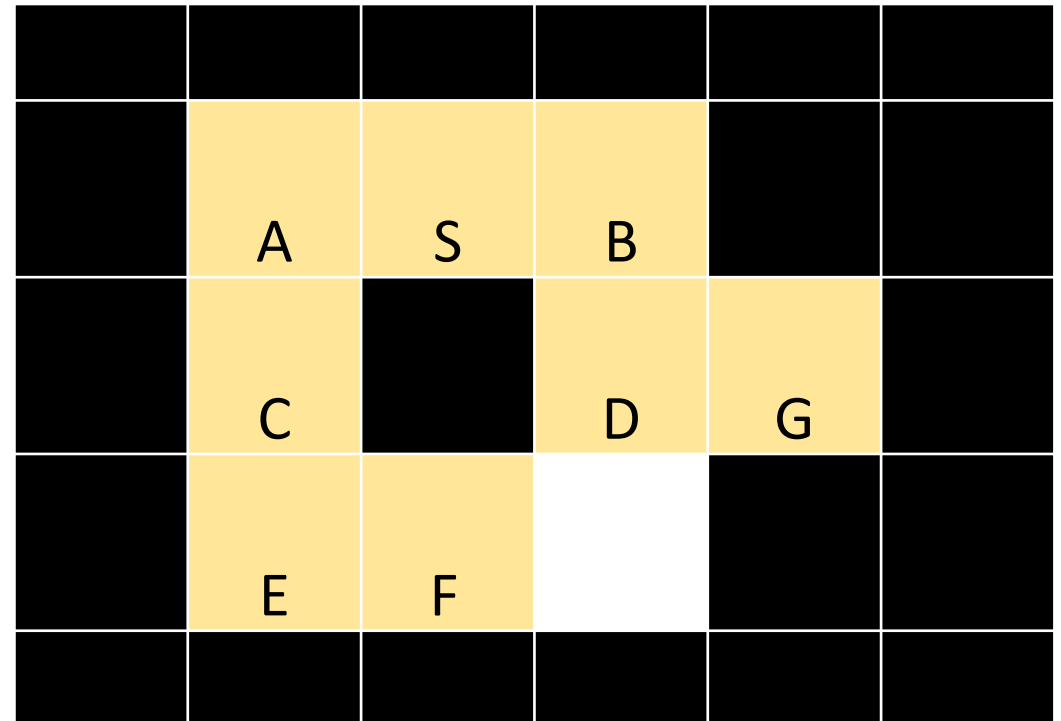
# BFS Computational Savings

- Frontier doesn't have to be a priority queue
- It can just be a regular first-in, first-out (FIFO) queue
- Example: Frontier = {E,G}
- Pop E from the queue, expand it



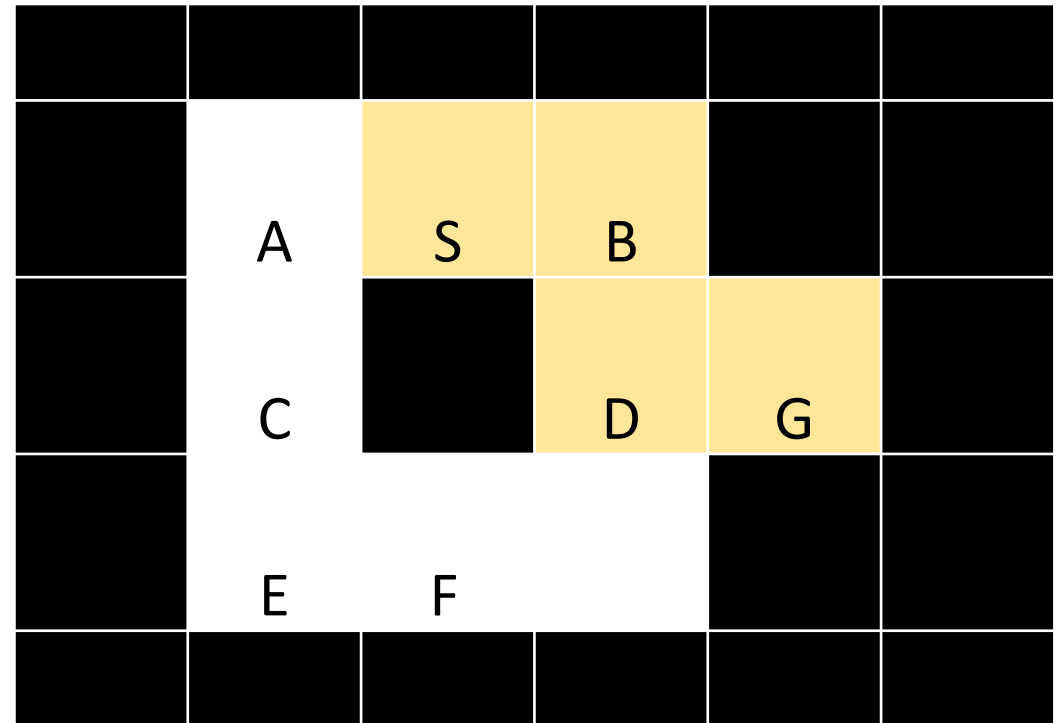
# BFS Computational Savings

- Frontier doesn't have to be a priority queue
- It can just be a regular first-in, first-out (FIFO) queue
- Example: Frontier = {G,F}
- Pop G from the queue, expand it



# BFS Computational Savings

- Frontier doesn't have to be a priority queue
- It can just be a regular first-in, first-out (FIFO) queue
- Example: G expanded, we learn that it's the goal, we found the best path



# Analysis of search strategies

- Strategies are evaluated along the following criteria:
  - **Completeness:** does it always find a solution if one exists?
  - **Optimality:** does it always find a least-cost solution?
  - **Time complexity:** number of nodes generated
  - **Space complexity:** maximum number of nodes in memory
- Time and space complexity are measured in terms of
  - **$b$ :** maximum branching factor of the search tree
  - **$d$ :** depth of the optimal solution
  - **$m$ :** maximum length of any path in the state space (may be infinite)

# Properties of breadth-first search

- **Complete?**

Yes (if branching factor  $b$  is finite).

Even w/o explored-set checking, it still works!

- **Optimal?**

Yes – if cost = 1 per step (uniform cost search will fix this)

- **Time?**

Number of nodes in a  $b$ -ary tree of depth  $d$ :  $O(b^d)$

( $d$  is the depth of the optimal solution)

- **Space?**

$O(b^d)$

- Space is the bigger problem (more than time)



# Properties of uniform-cost search

- **Complete?**

Yes (if branching factor  $b$  is finite).

Even w/o explored-set checking, it still works!

- **Optimal?**

Yes – even if cost  $\neq 1$  per step

- **Time?**

Number of nodes in a  $b$ -ary tree of depth  $d$ :  $O(b^d)$

( $d$  is the depth of the optimal solution)

- **Space?**

$O(b^d)$

- Space is the bigger problem (more than time)

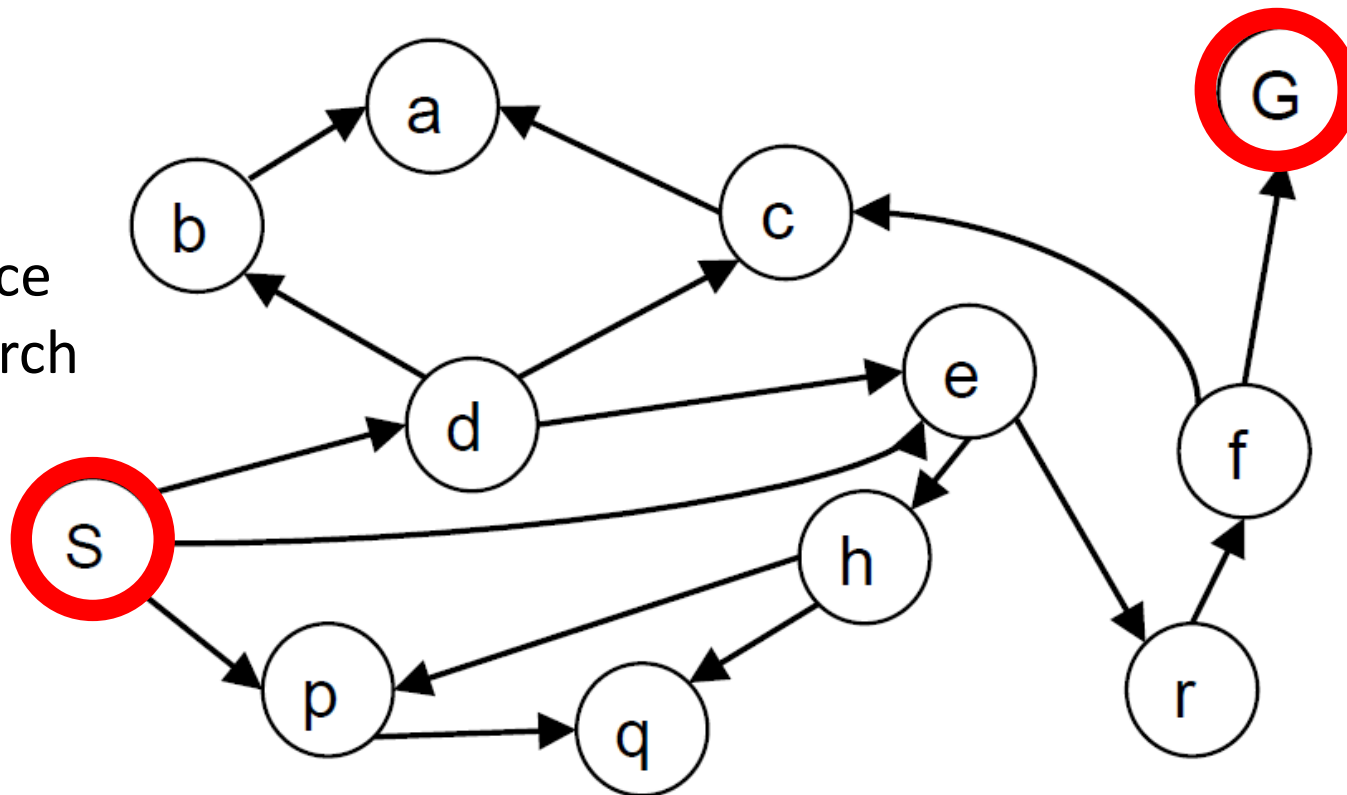
# Outline

- Uniform cost search (UCS) = slightly different implementation of Dijkstra's Algorithm
- Breadth-first search (BFS) = special case of UCS
- **Depth-first search (DFS)**

# Depth-first search

- Expand **deepest** unexpanded node (BFS: **shallowest**)
- Implementation: Frontier is a **last-in-first-out (LIFO)** stack (BFS:FIFO)

Example state space graph for a tiny search problem



# Depth-first search

Frontier:

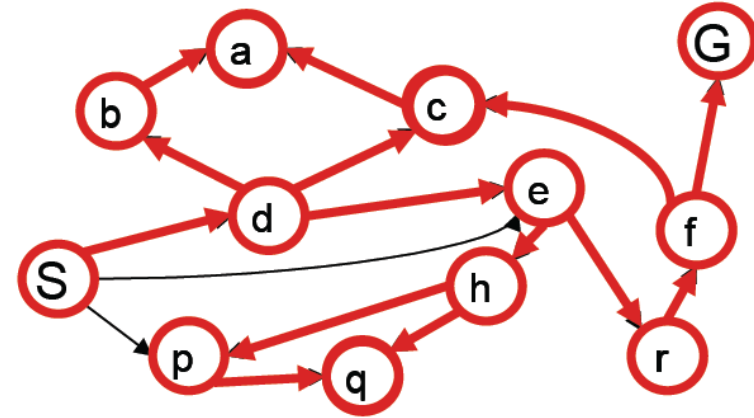
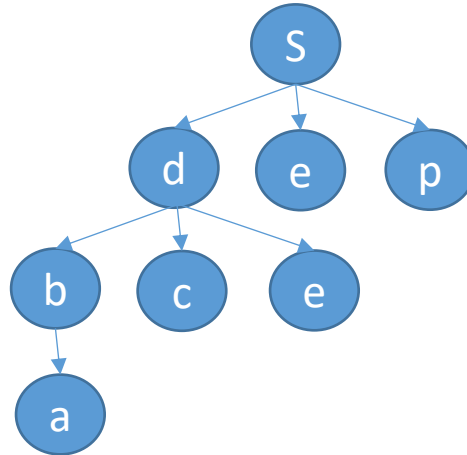
Step 0: {S}

Step 1: {d,e,p}

Step 2: {b,c,e,p} – FIFO

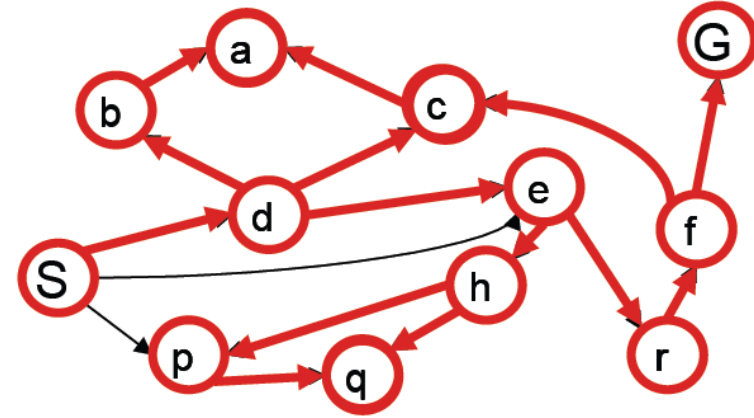
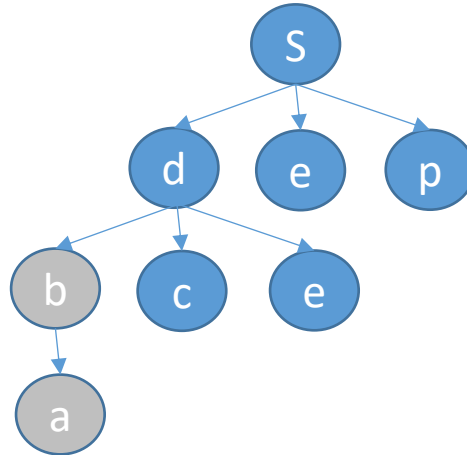
Step 3: {a,c,e,p}

Step 4: {c,e,p}



# The reason DFS is useful: Space

When we know that Sdba is a dead end, we can remove it from the tree!



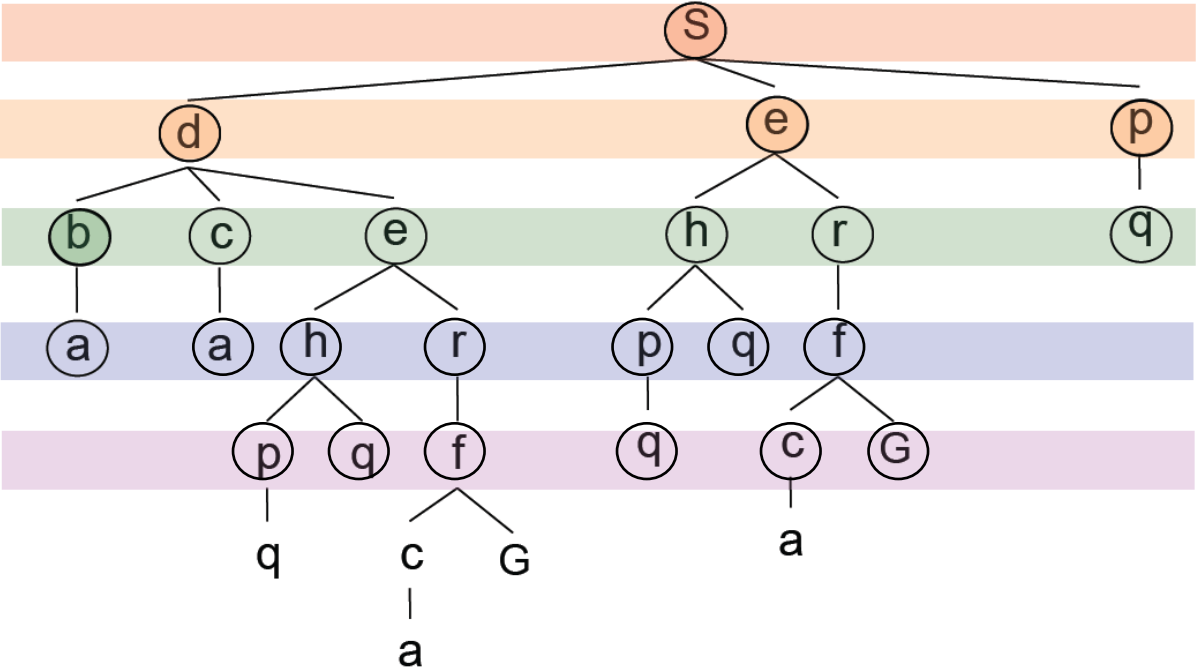
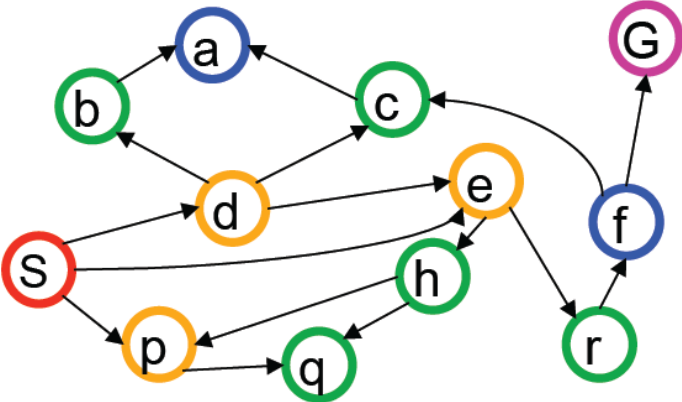
# Breadth-first search

Computational complexity:

(s,  
d,e,p,  
b,c,e,h,r,q,  
a,a,h,r,p,q,f,  
p,q,f,q,c,G)

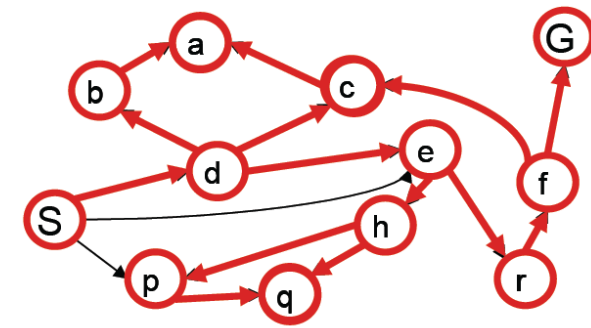
Space complexity:

We have to store the whole tree!

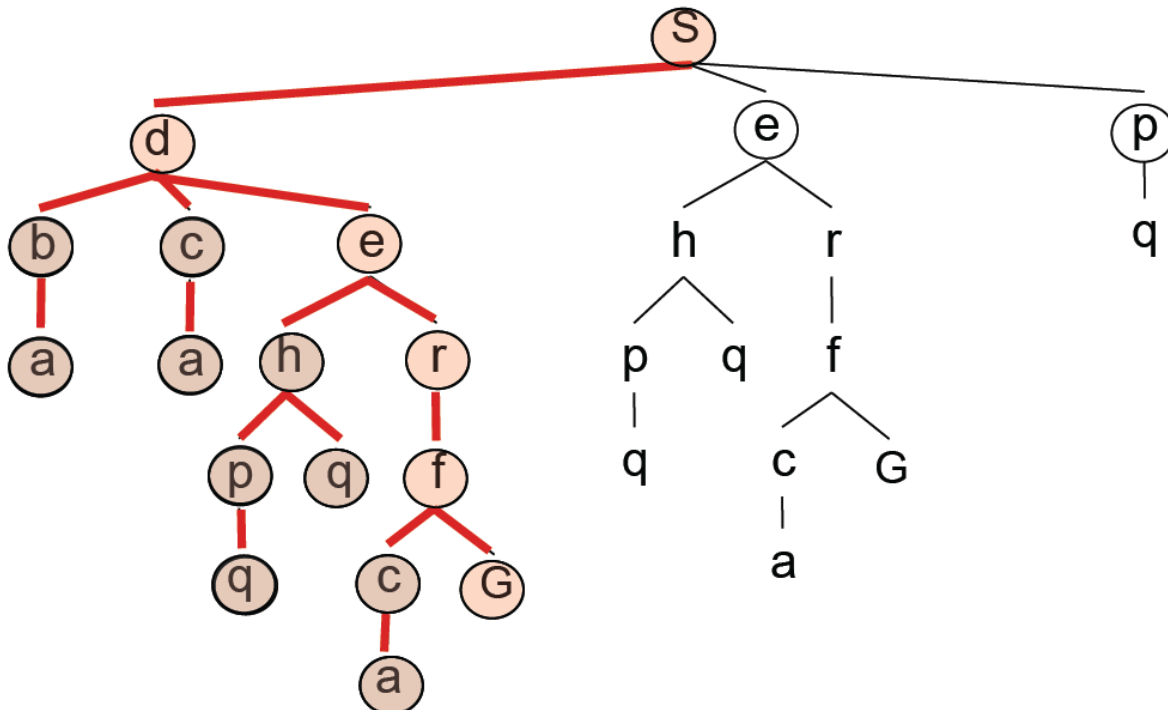


Example from P. Abbeel and D. Klein

# Depth-first search

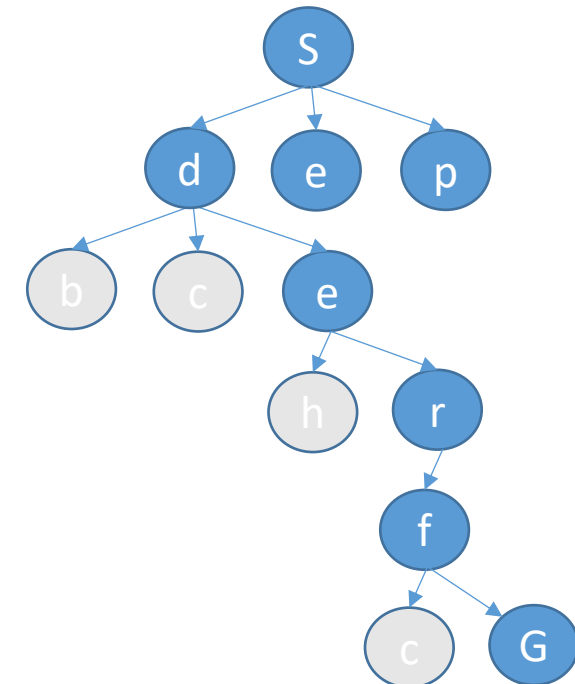


Computational complexity: here are the nodes we expanded



Example from P. Abbeel and D. Klein

Space complexity: here's the part of the tree that we still have in memory



# Analysis of search strategies

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# Properties of depth-first search

- **Complete? (always finds a solution if one exists?)**
  - Fails in infinite-depth spaces
  - Fails if there are loops (unless you keep an “Explored Set”)
- **Optimal? (always finds an optimal solution?)**
  - No – returns the first solution it finds
- **Time? (how long does it take, in terms of  $b$ ,  $d$ ,  $m$ ?)**
  - $O(b^m)$  (remember BFS was  $O(b^d)$ )
  - Terrible if  $m$  is much larger than  $d$
- **Space? (how much storage space?)**
  - $O(bm)$ , i.e., linear space!
  - The frontier doesn't need to keep track of failed paths, only the currently active path

# Comparison of Search Strategies

Algorithm	Complete?	Optimal?	Time complexity	Space complexity	Implement the Frontier as a...
<b>BFS</b>	Yes	If all step costs are equal	$O(b^d)$	$O(b^d)$	Queue
<b>DFS</b>	No	No	$O(b^m)$	$O(bm)$	Stack
<b>UCS</b>	Yes	Yes	Number of nodes w/ $V_n \leq V_G$	Number of nodes w/ $V_n \leq V_G$	Priority Queue sorted by $V_n$