

# **FIT2004**

## **Algorithms and Data Structures**

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Referencing materials by  
Nathan Compane, Aamir Cheema, Arun Konagurthu and Lloyd Allison



# Faculty of Information Technology, Monash University

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Ready?

# Agenda

- The Graph data structure
- Graph Traversal algorithms

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- The Graph data structure
  - Introduction
  - Representation
- Graph Traversal algorithms

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  - Representation
- Graph Traversal algorithms
  - Breadth First Search (BFS)
  - Depth First Search (DFS)
  - Dijkstra's shortest distance

# Agenda

- The Graph data structure
    - Introduction
    - Representation
  - Graph Traversal algorithms
    - Breadth First Search (BFS)
    - Depth First Search (DFS)
- } Basic for many graph-algorithms

Let us begin...



- Master race of all data structure

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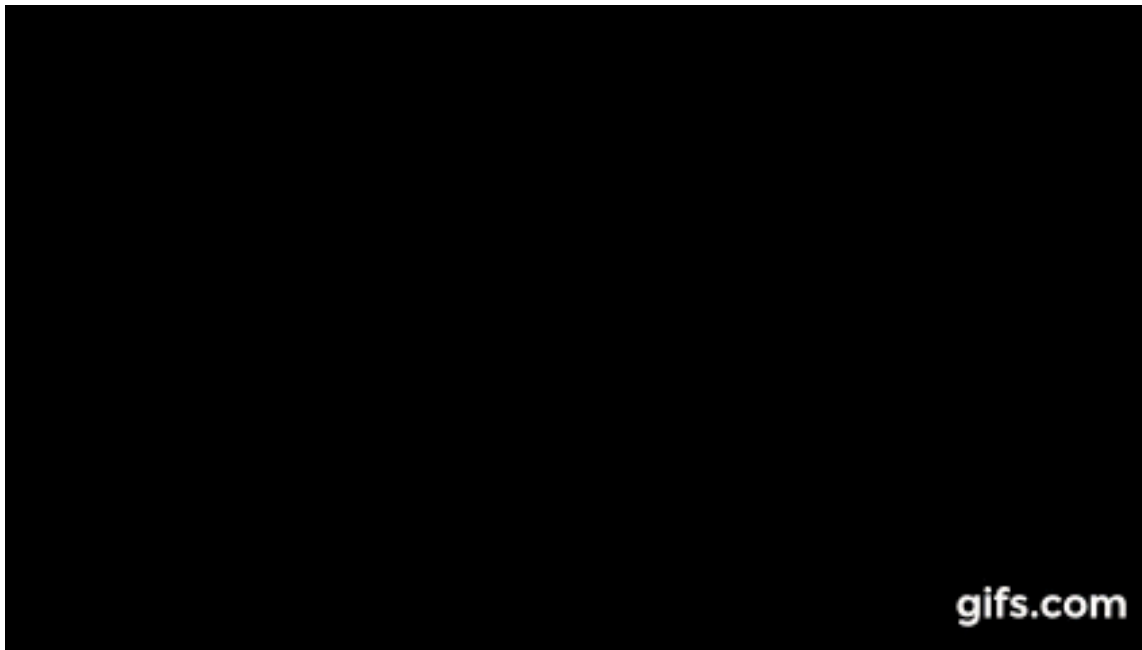
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  - Have you seen it before?



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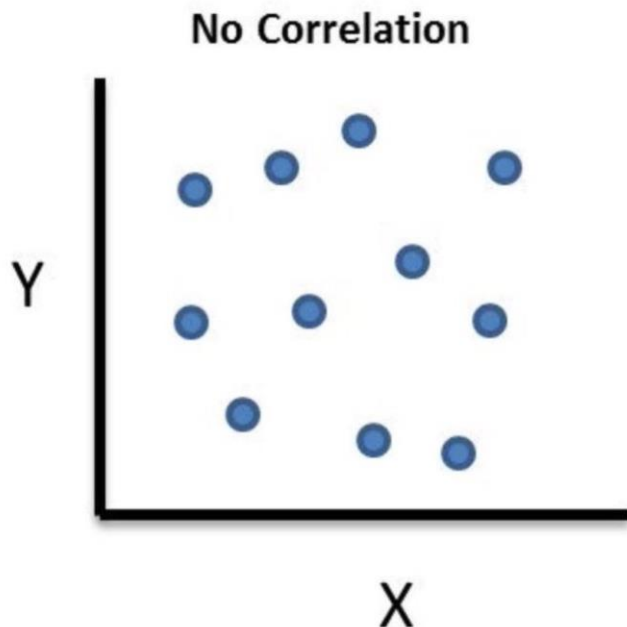


- So what is a graph?
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  - Was Nickleback right?

# Graph

## Introduction

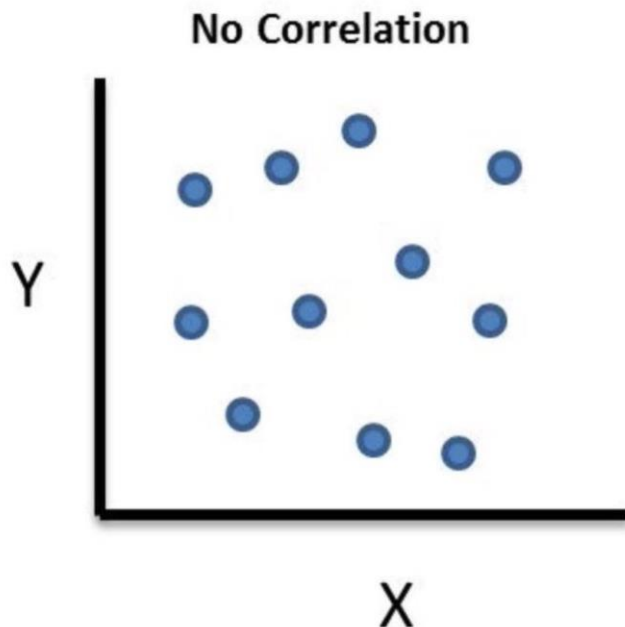
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# Graph

## Introduction

- So what is a graph?
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Here's a graph of my life thus far

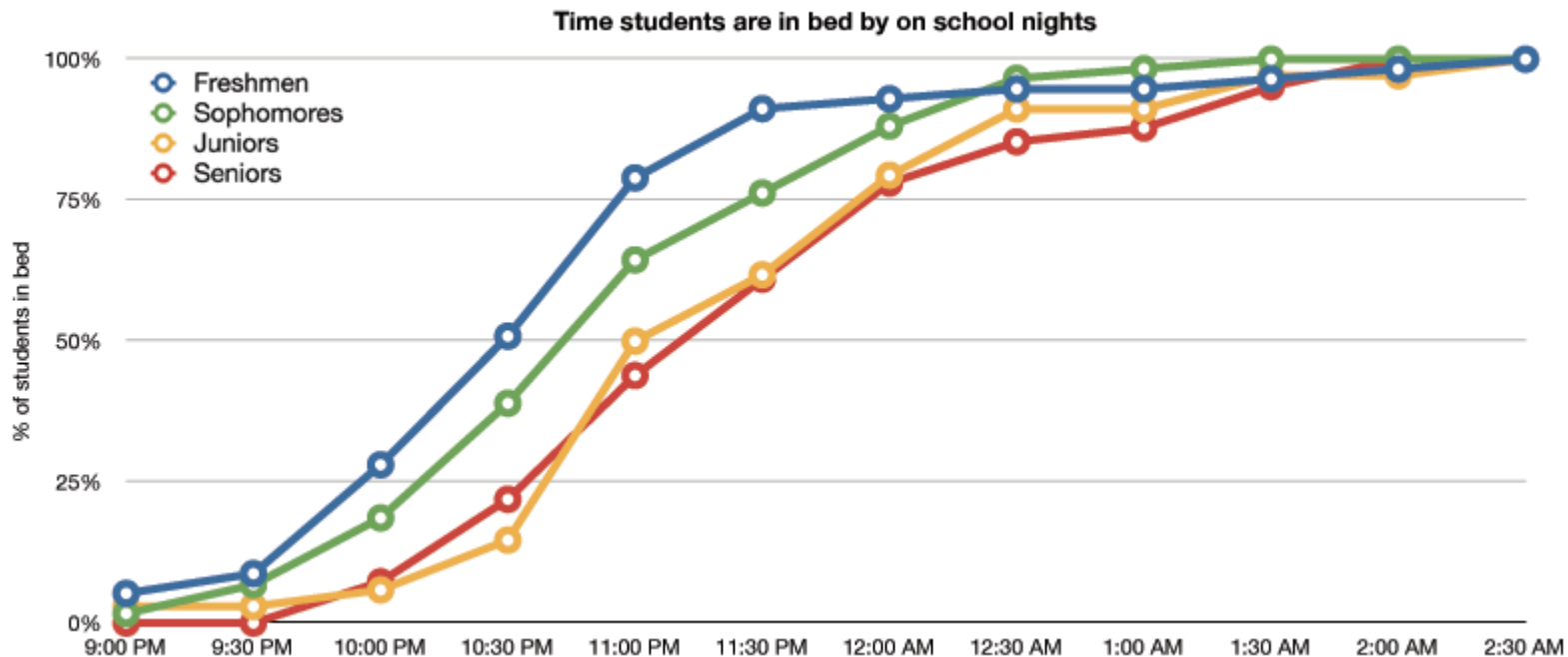


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  - Edge (Edges)

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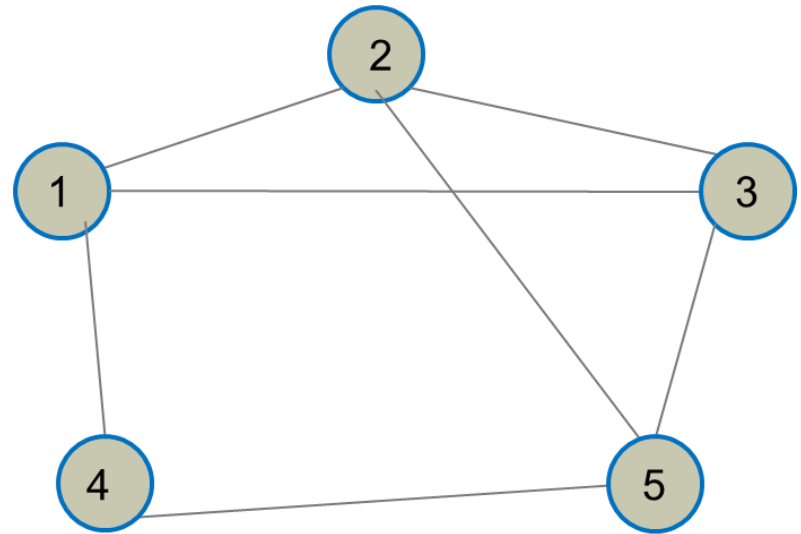
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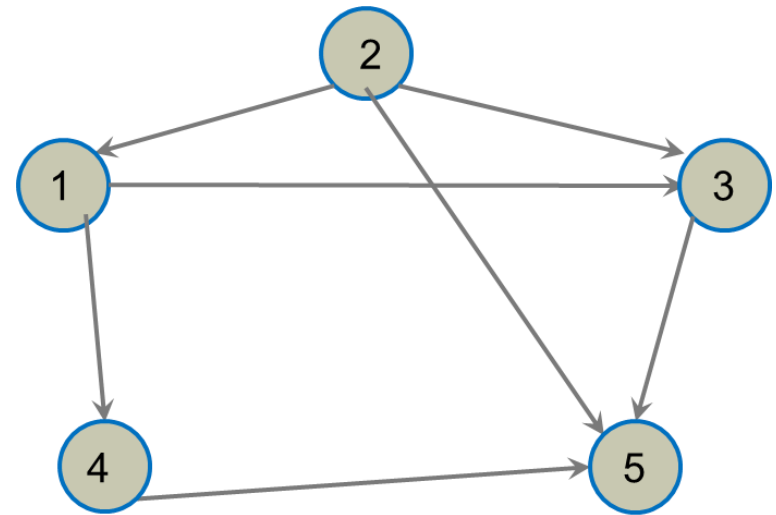
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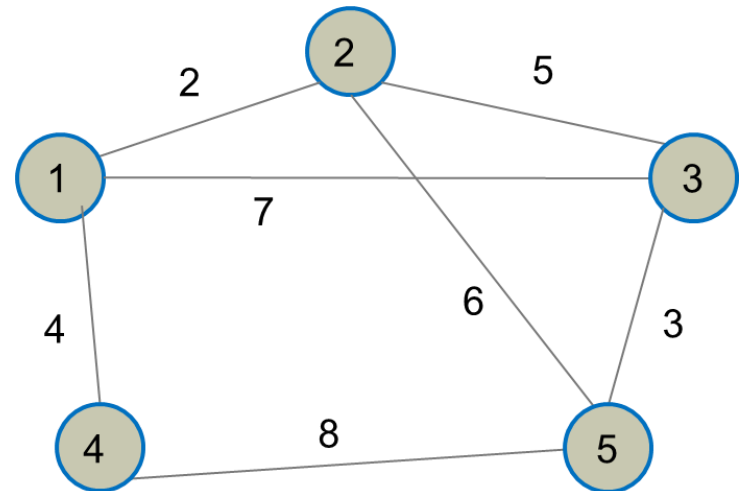
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    - Also known as link
    - These links can be directed or undirected
    - These links can be **weighted** or **unweighted**



Questions?

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- Simple graph
  - No self-edges arrow from itself to itself
  - No multi-edges between vertices no  $a$  to  $b$ , then  $b$  to  $a$

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# Graph

## Properties

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  - $|V|$  is the number of vertices
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 $[V] * ([V] - 1)$

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undirected graph  $|V| = |E|$

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- A graph is called sparse if  $E \ll V^2$
- A graph is called dense if  $E \approx V^2$

Questions?

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# Graph Importance

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  - Trees are in fact graphs.



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    - Each **webpage** is a **vertex**
    - Each **hyperlink** is an **edge**

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    - **Google's PageRank** is a graph algorithm
      - **Traversal** through **webpages** and **propagate authority**
      - You can code it yourself, it is easy!



Questions?

- How do we represent graphs?

- How do we represent graphs? 2 possible way!

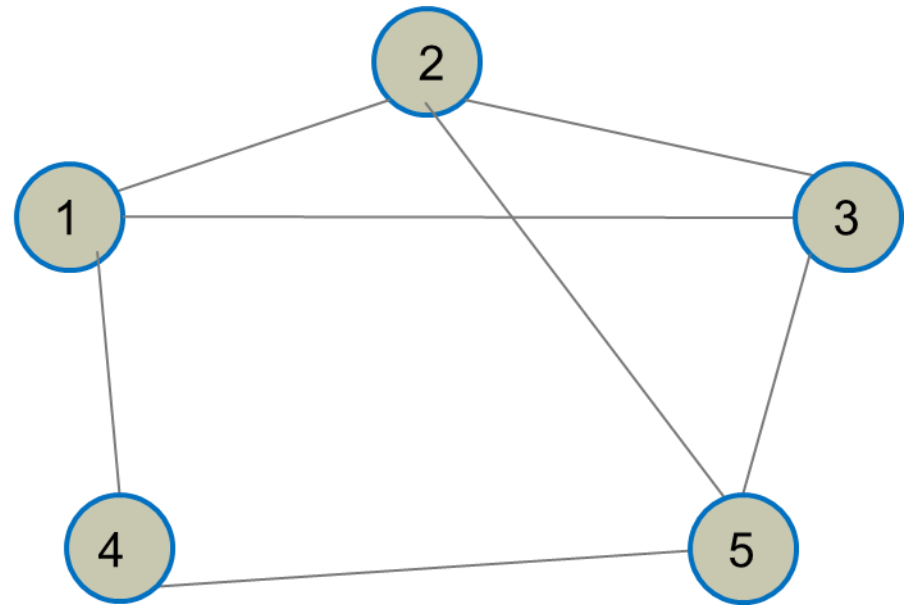
- How do we represent graphs? **2 possible** way!
  - Adjacency matrix
  - Adjacency list prefer this a lot of time

- Adjacency matrix
  - Store edge information in a matrix

# Graph Representation

- Adjacency matrix
  - Store edge information in a matrix
    - True/ False or 1/0 for unweighted

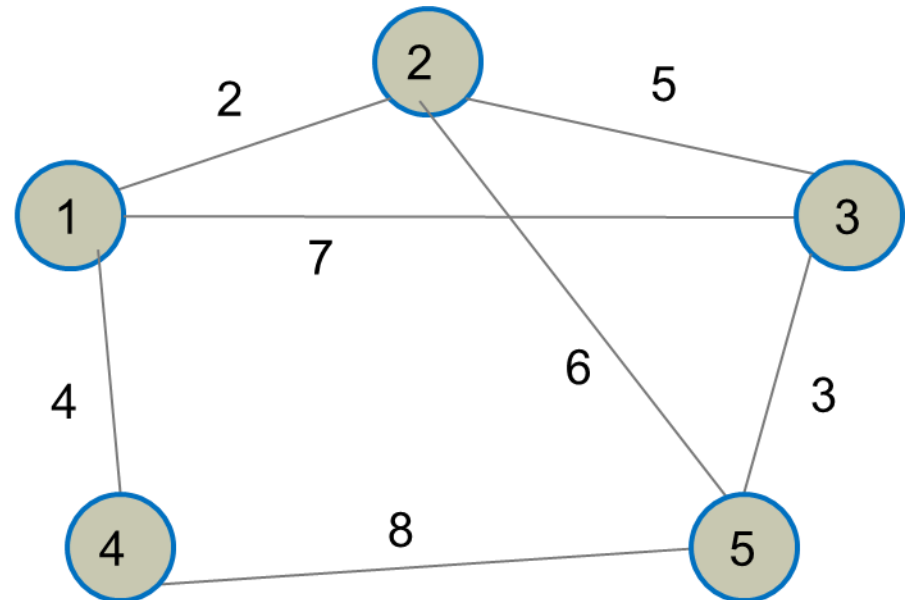
	1	2	3	4	5
1	F	T	T	T	F
2	T	F	T	F	T
3	T	T	F	F	T
4	T	F	F	F	T
5	F	T	T	T	F



# Graph Representation

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1		2	7	4	
2	2		5		6
3	7	5			3
4	4				8
5		6	3	8	



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good-dense graph, bad-sparse graph
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    - $O(V)$  to check/ traverse all adjacent vertices
      - Adjacent = neighbour have a edge

- Adjacency list

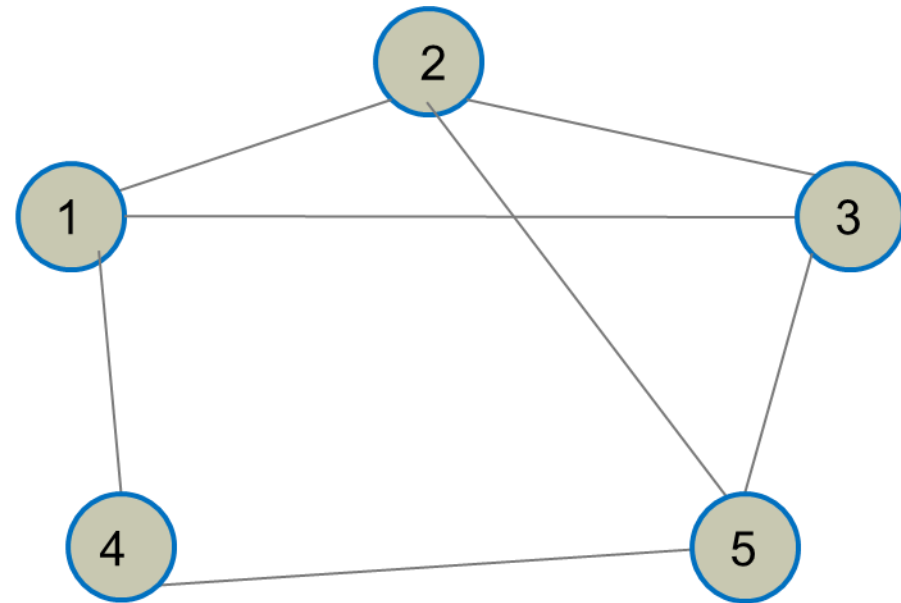
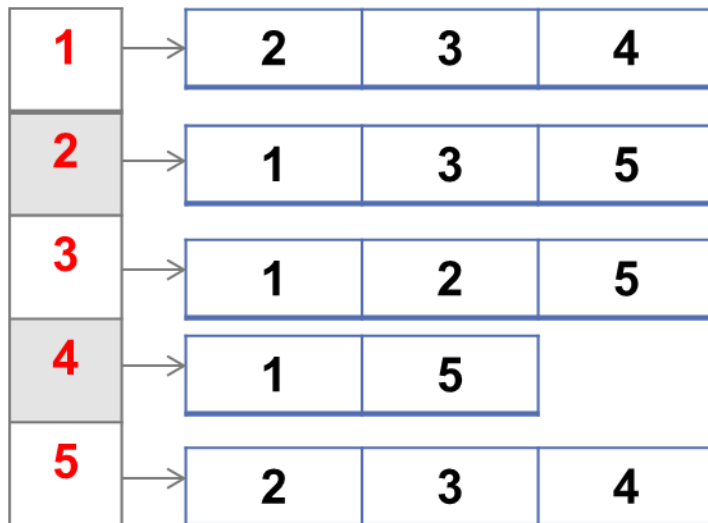
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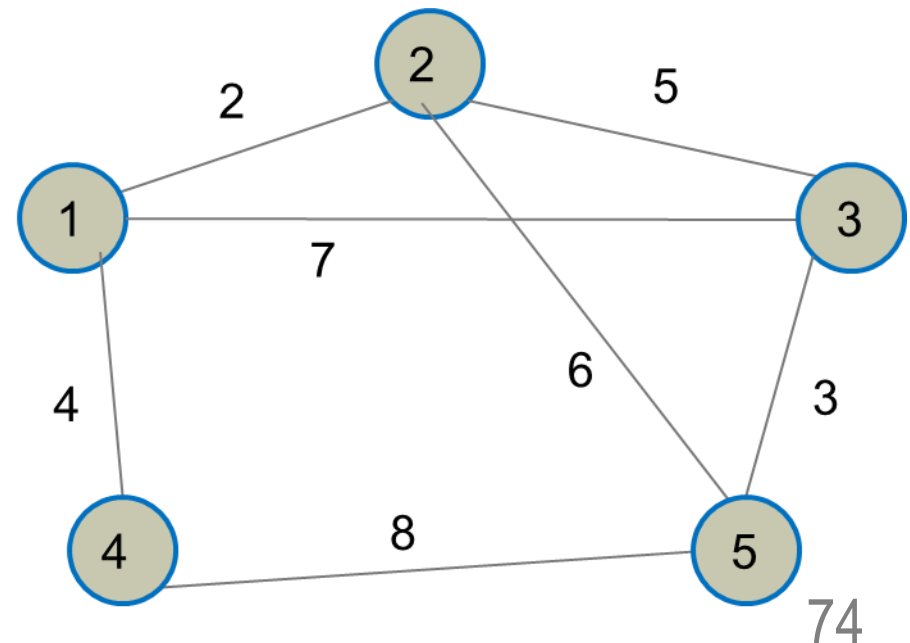
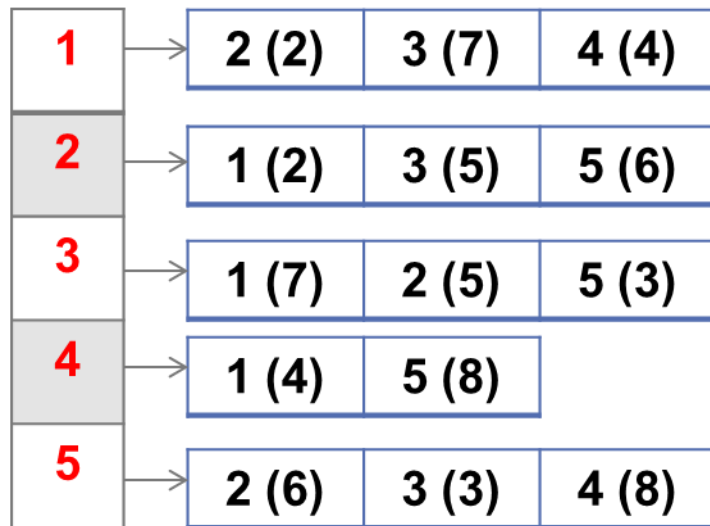
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not  $O(VE)$   $E = V[V-1]$ , only link to every other vertices
  - Time complexity?

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    - Time complexity?
      - $O(\log V)$  to check if an edge exist if the edges are sorted binary search
      - $O(X)$  to retrieve all of the adjacent vertices of a vertex
- when traverse use adjacency list

since graph can be sparse, adjacency matrix would need to loop through until biggest vertex  
but Adjacency List has length of list, number of links to other vertices

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  - Space complexity?
    - $O(V+E)$ . Storing  $V$  vertices (as an array) and then total of  $E$  edges
  - Time complexity?
    - $O(\log V)$  to check if an edge exist if the edges are sorted
      - But you can't use binary search on linked list!
      - So this is still  $O(X)$  but you can terminate earlier once you reach a bigger vertex
    - $O(X)$  to retrieve all of the adjacent vertices of a vertex
      - Where  $X$  = number of adjacent vertices (output-sensitive complexity)

Questions?

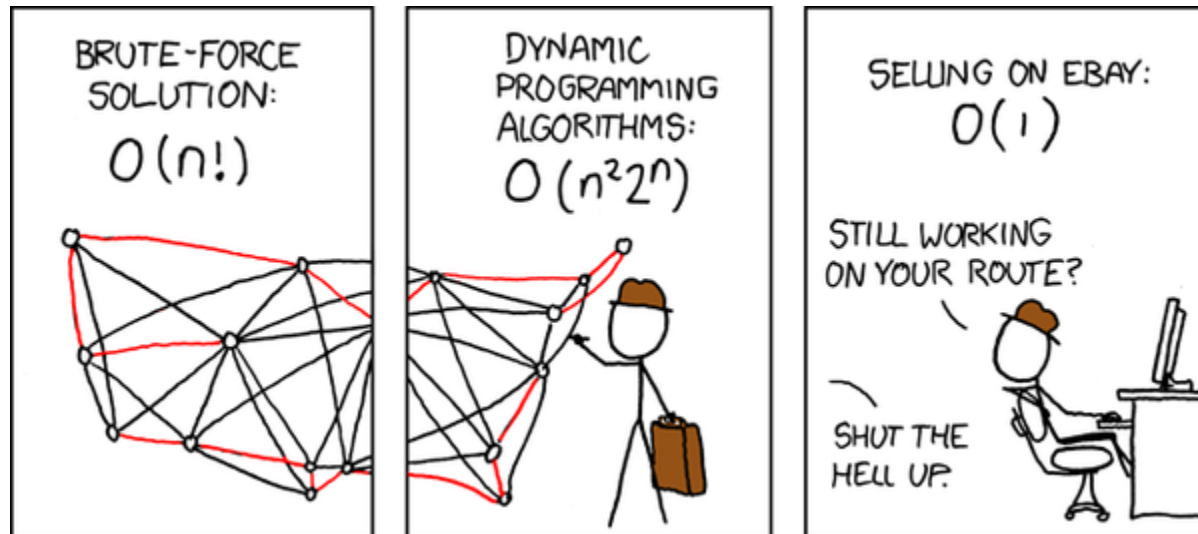


- Going from a place to another

- Going from a place (source vertex) to another

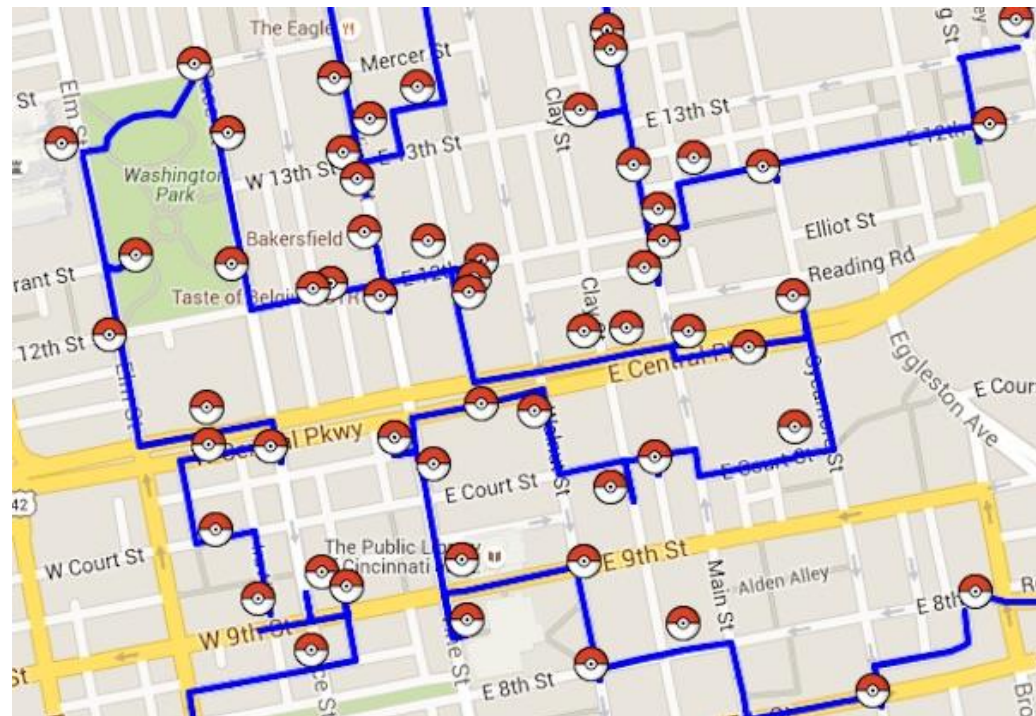
# Graph Traversal

- Going from a place (**source vertex**) to another or everywhere!



# Graph Traversal

- Going from a place (**source vertex**) to another or everywhere!



- Breadth-First Search (BFS)
- Depth-First Search (DFS)

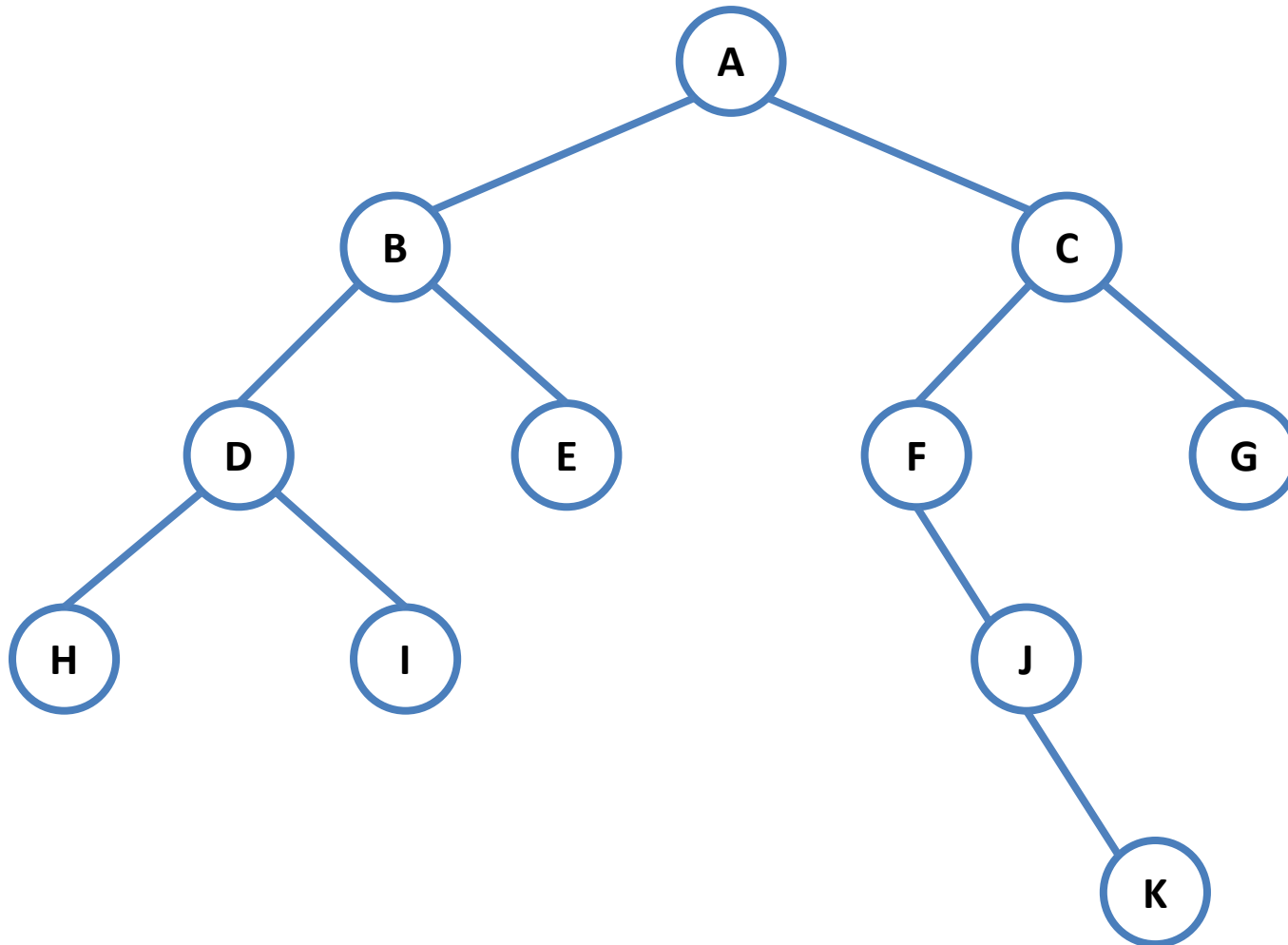
- Breadth-First Search (BFS)
  - Going wide
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  - Going deep

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- Let us begin with a tree first

- Breadth-First Search (BFS)
  - Going wide
- Depth-First Search (DFS)
  - Going deep
- Let us begin with a tree first
  - Recall a tree is a graph without cycles

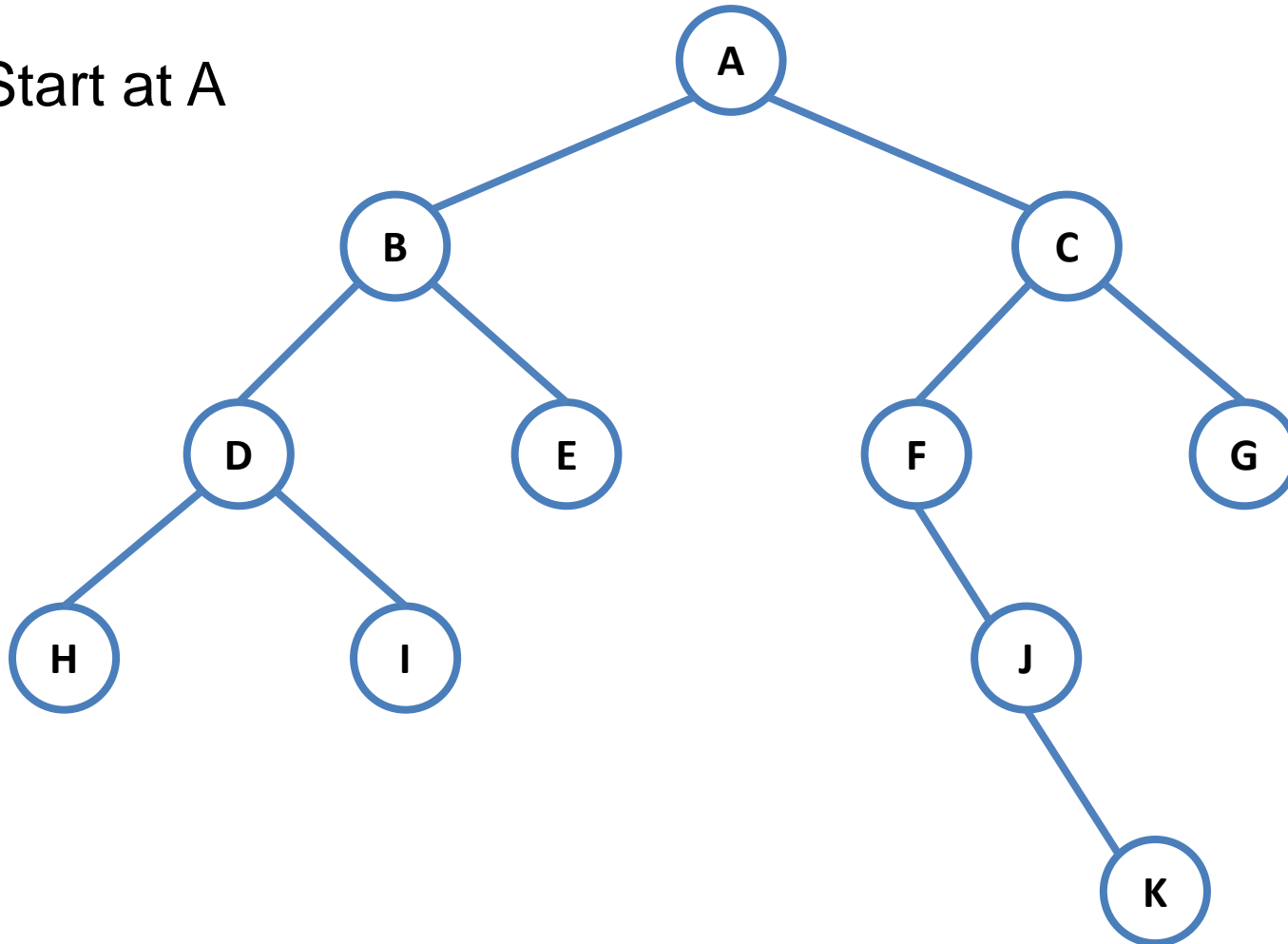


# Graph Traversal



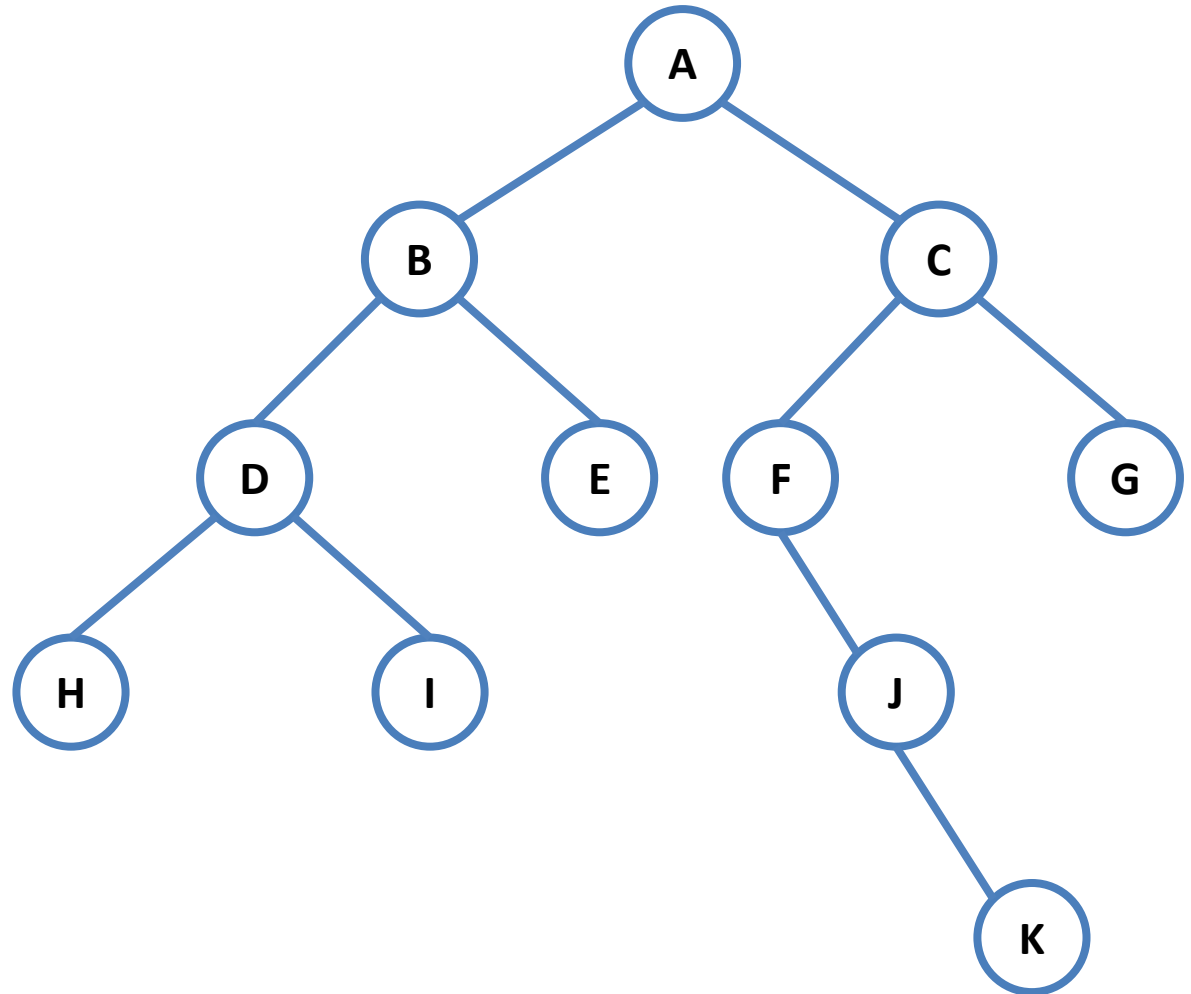
# Graph Traversal

- Start at A

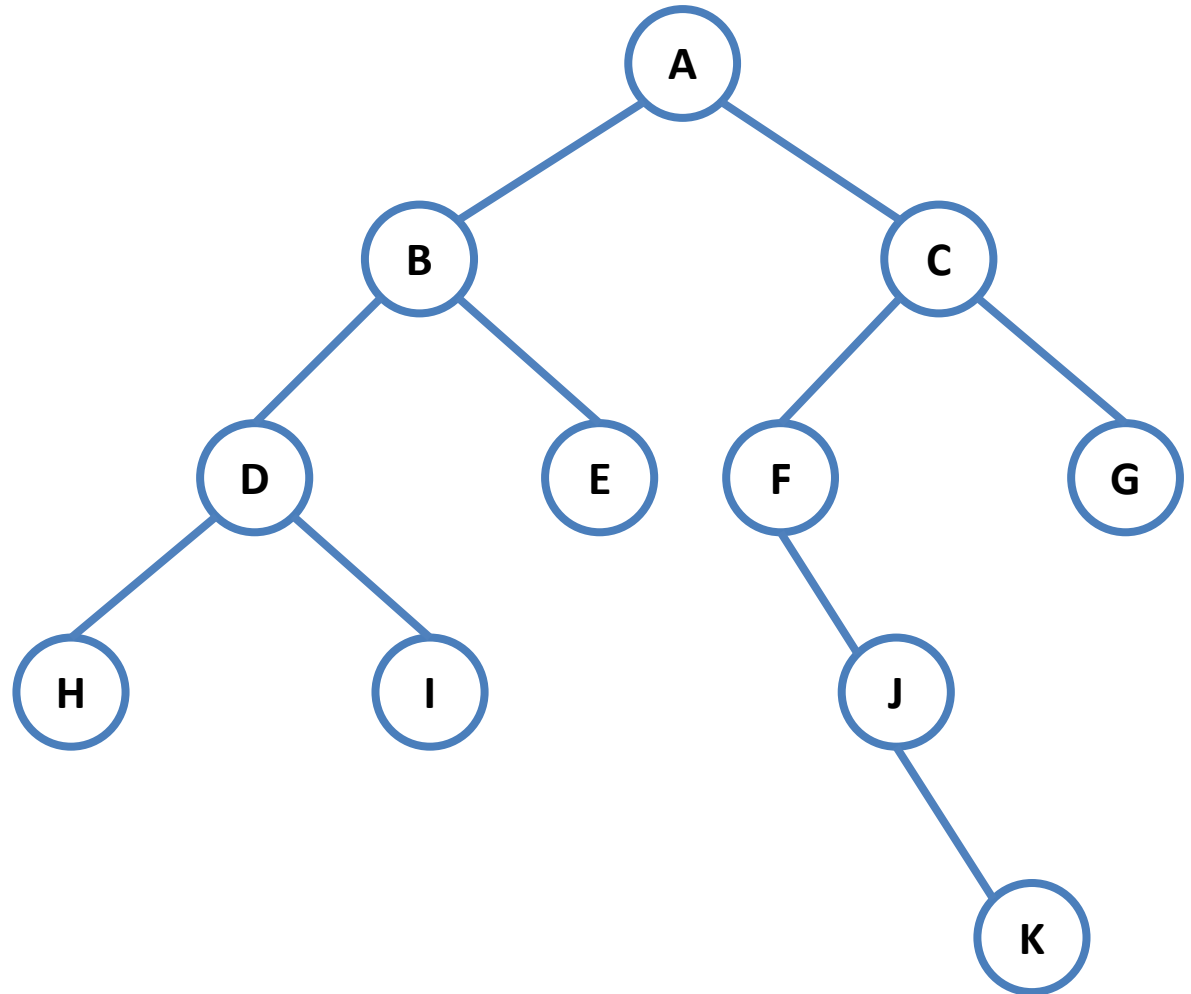


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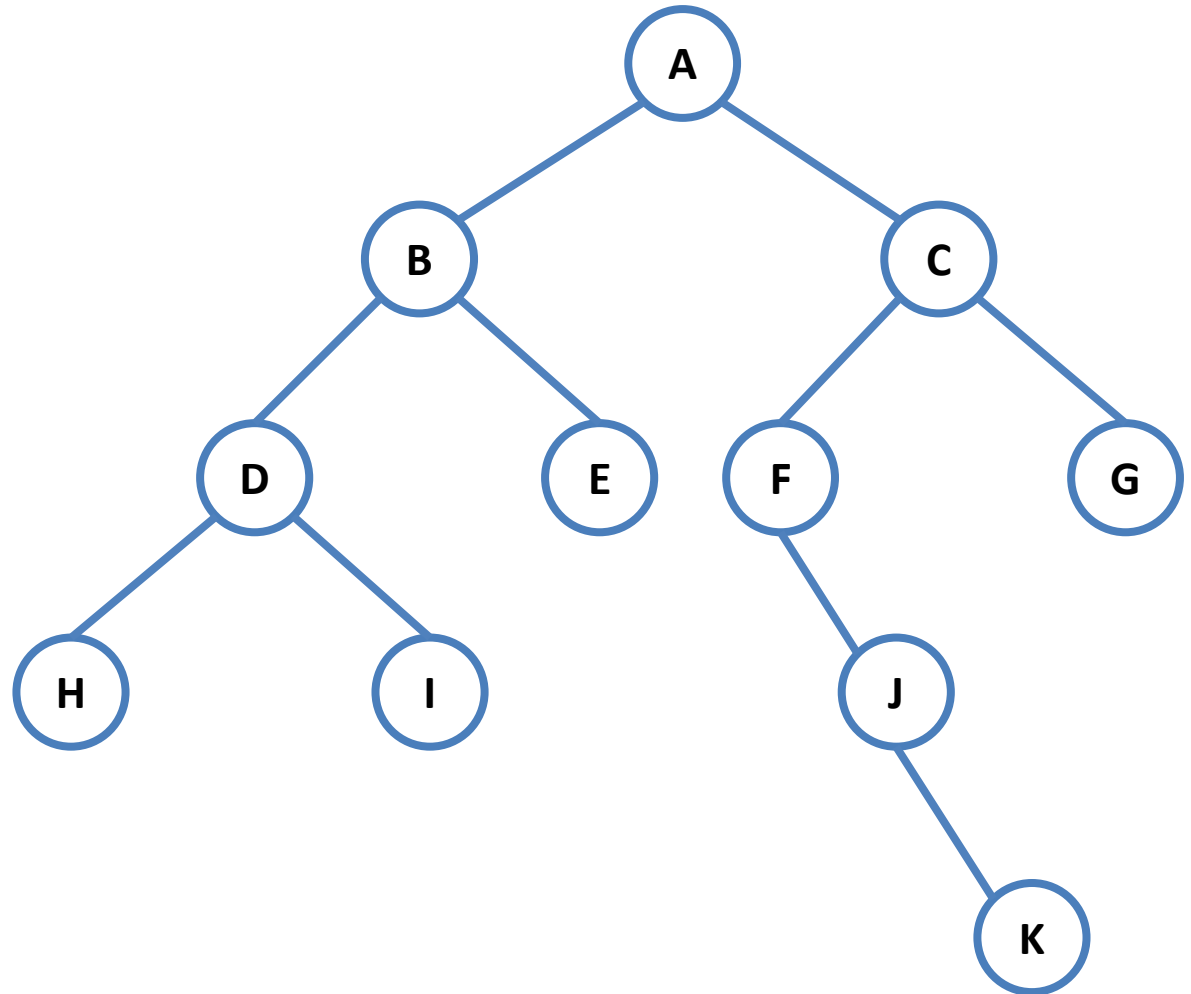


- Start at A, BFS



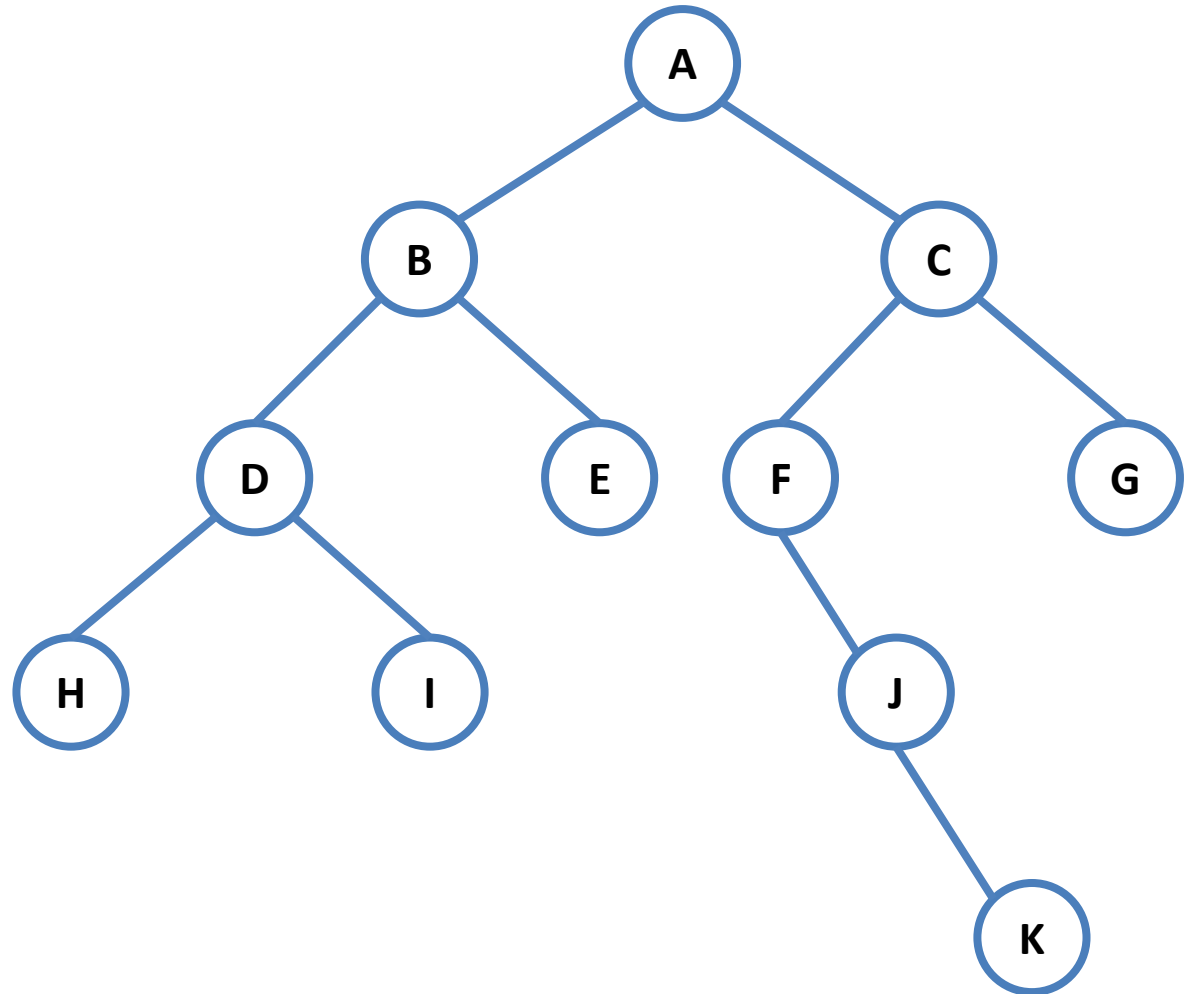
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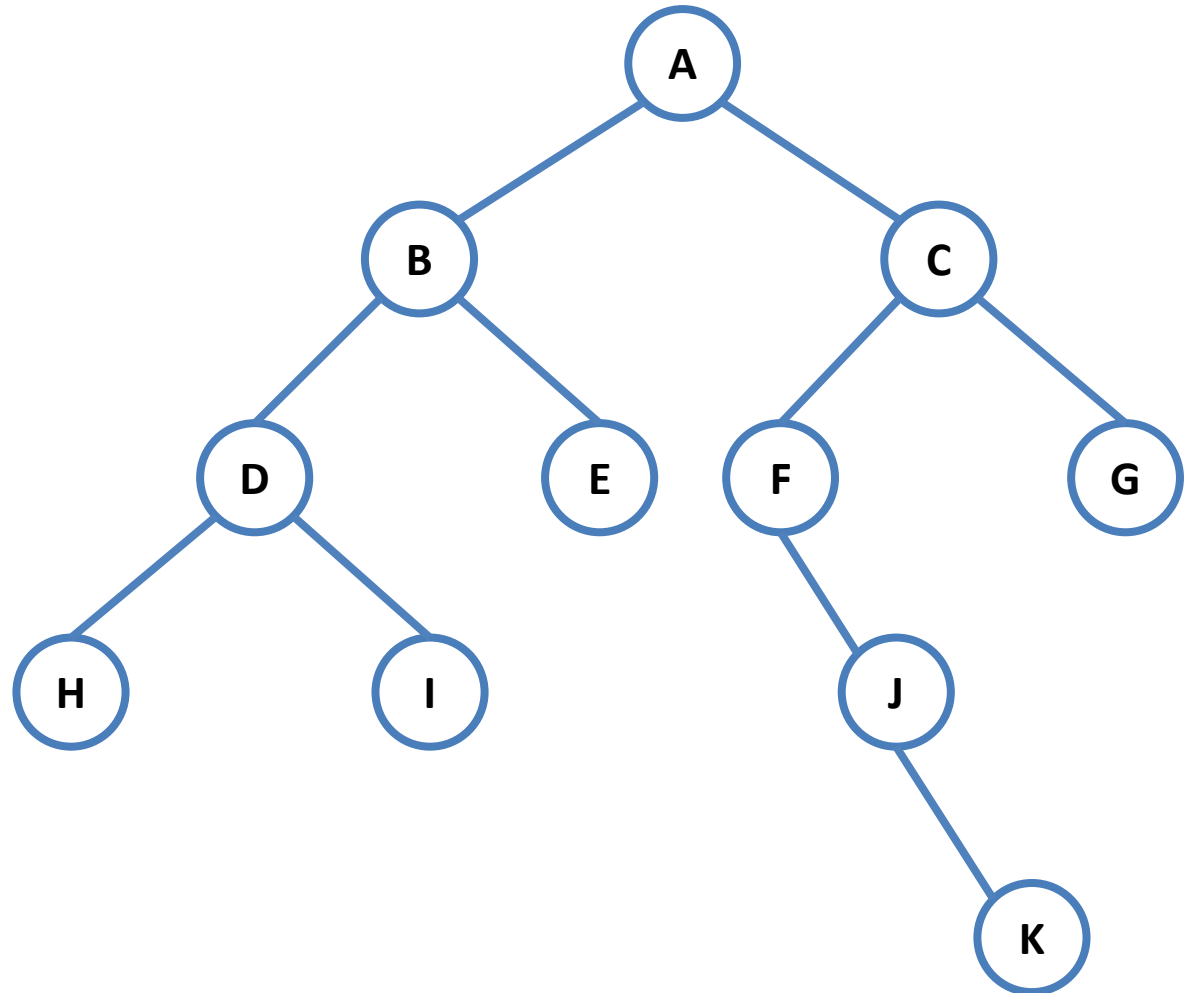
# Graph Traversal

- Start at A, BFS
- A
- B



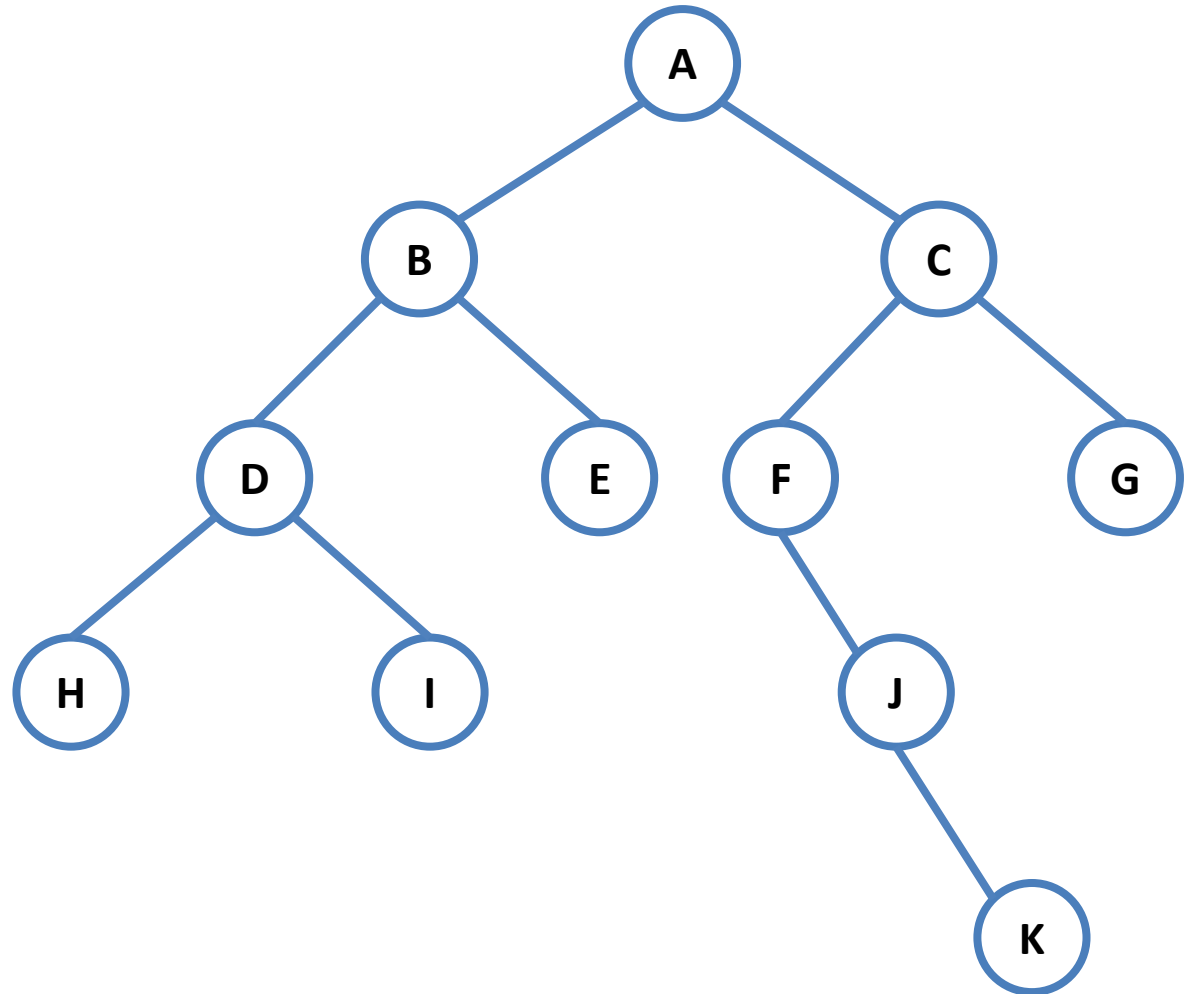
# Graph Traversal

- Start at A, BFS
- A
- B
- C



# Graph Traversal

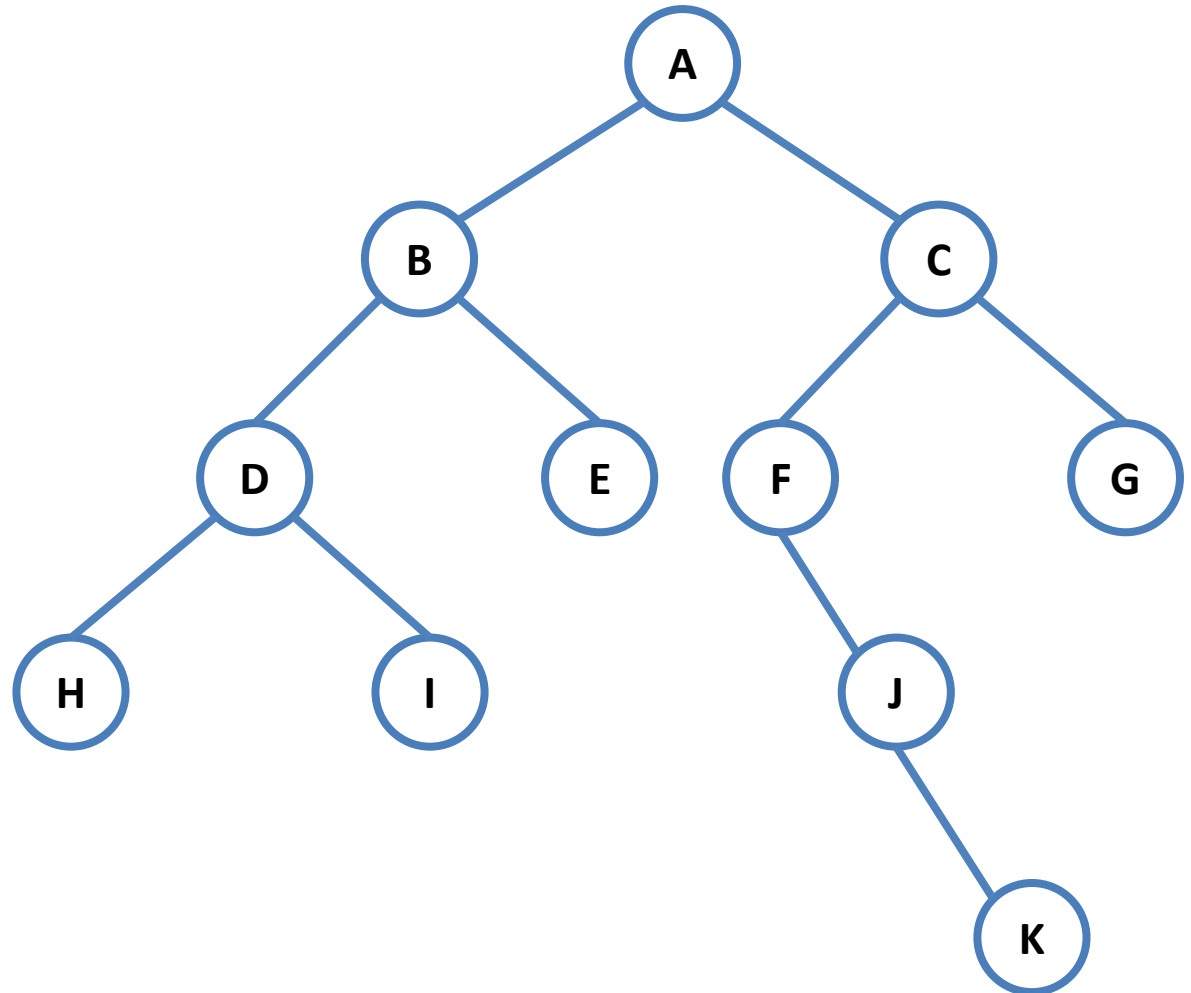
- Start at A, BFS
- A
- B
- C
- D





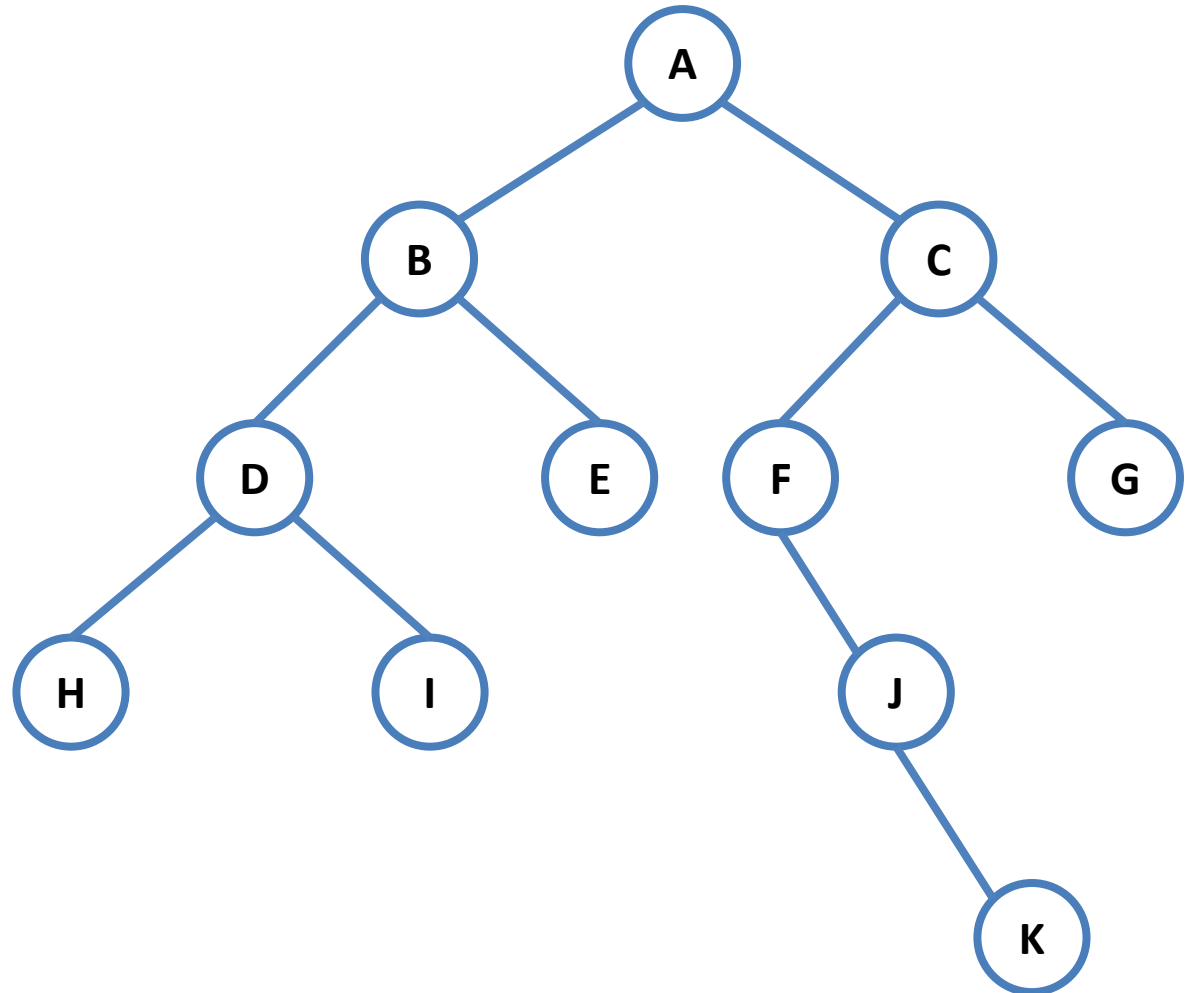
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- Start at A, BFS
- A
- B
- C
- D
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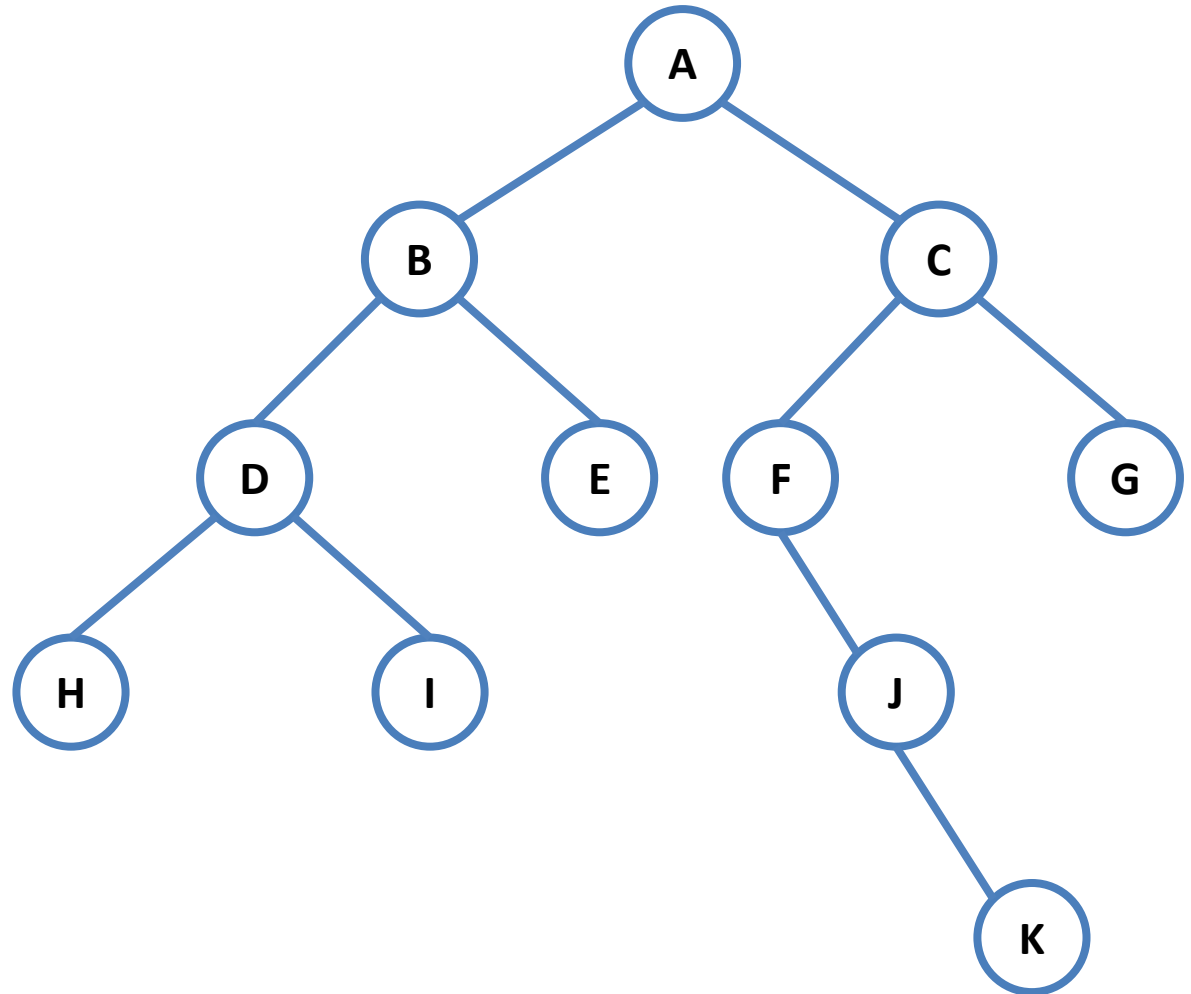
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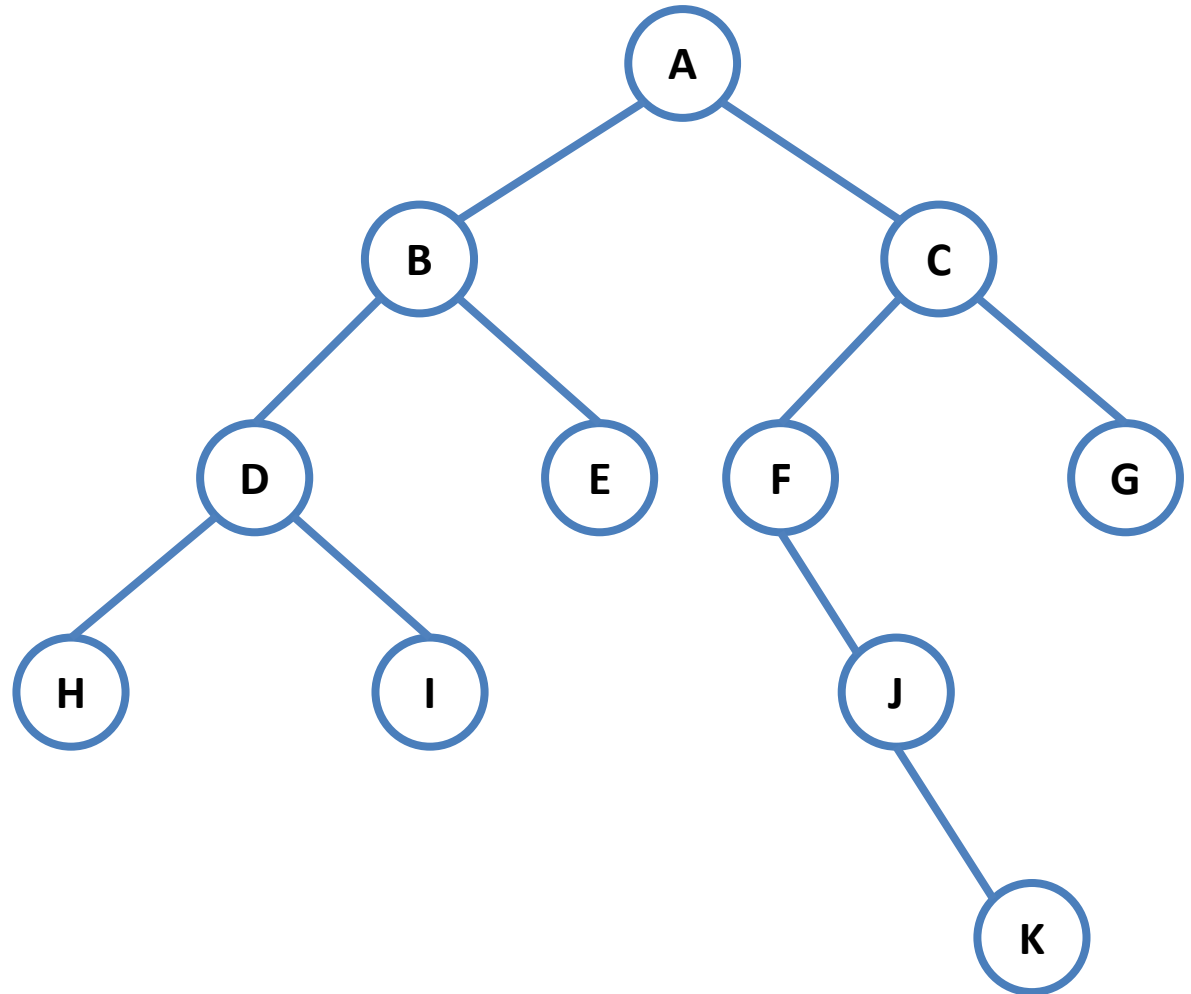
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- D
- E
- F
- G



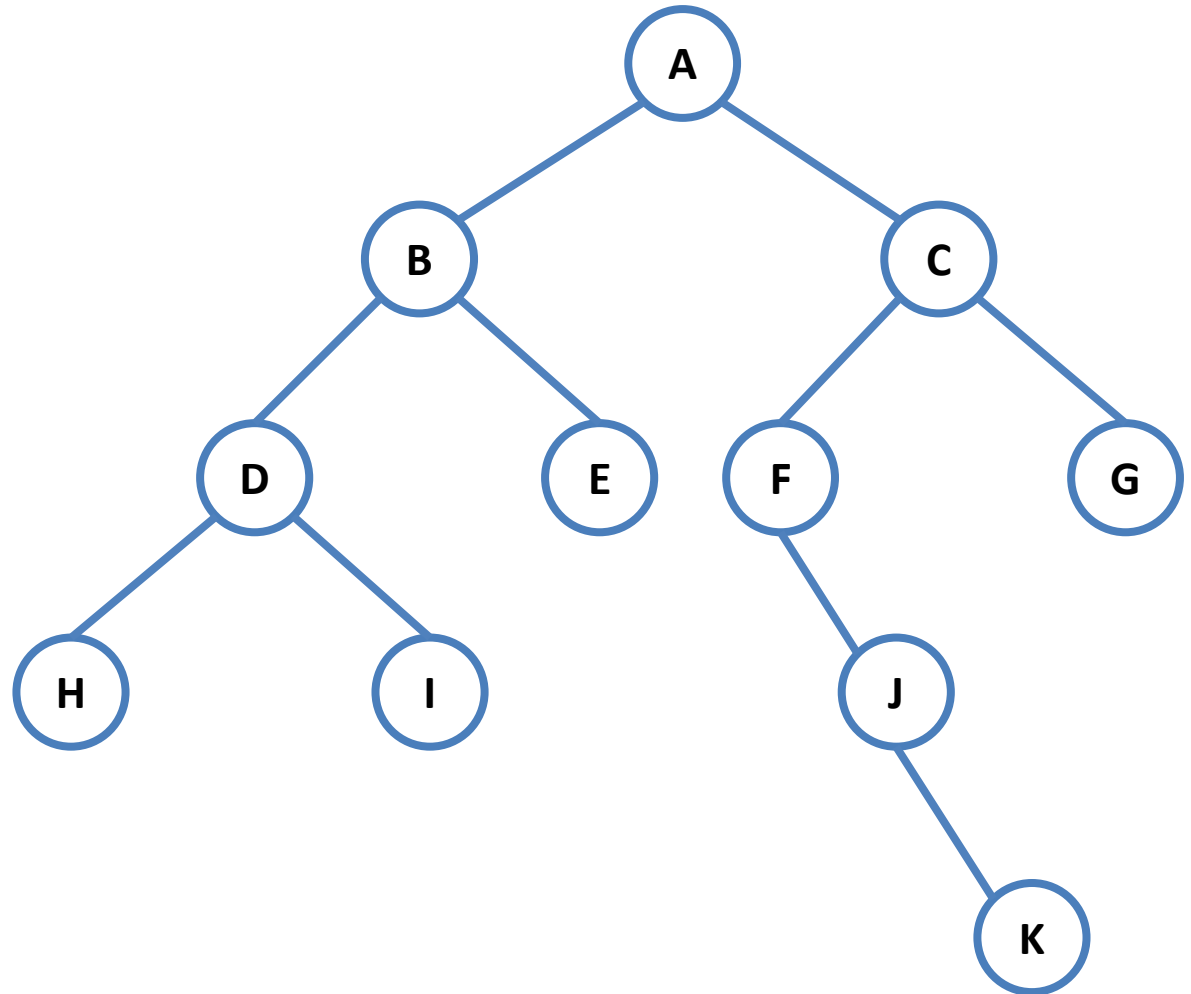
# Graph Traversal

- Start at A, BFS
- A
- B
- C
- D
- E
- F
- G
- ... and so on...

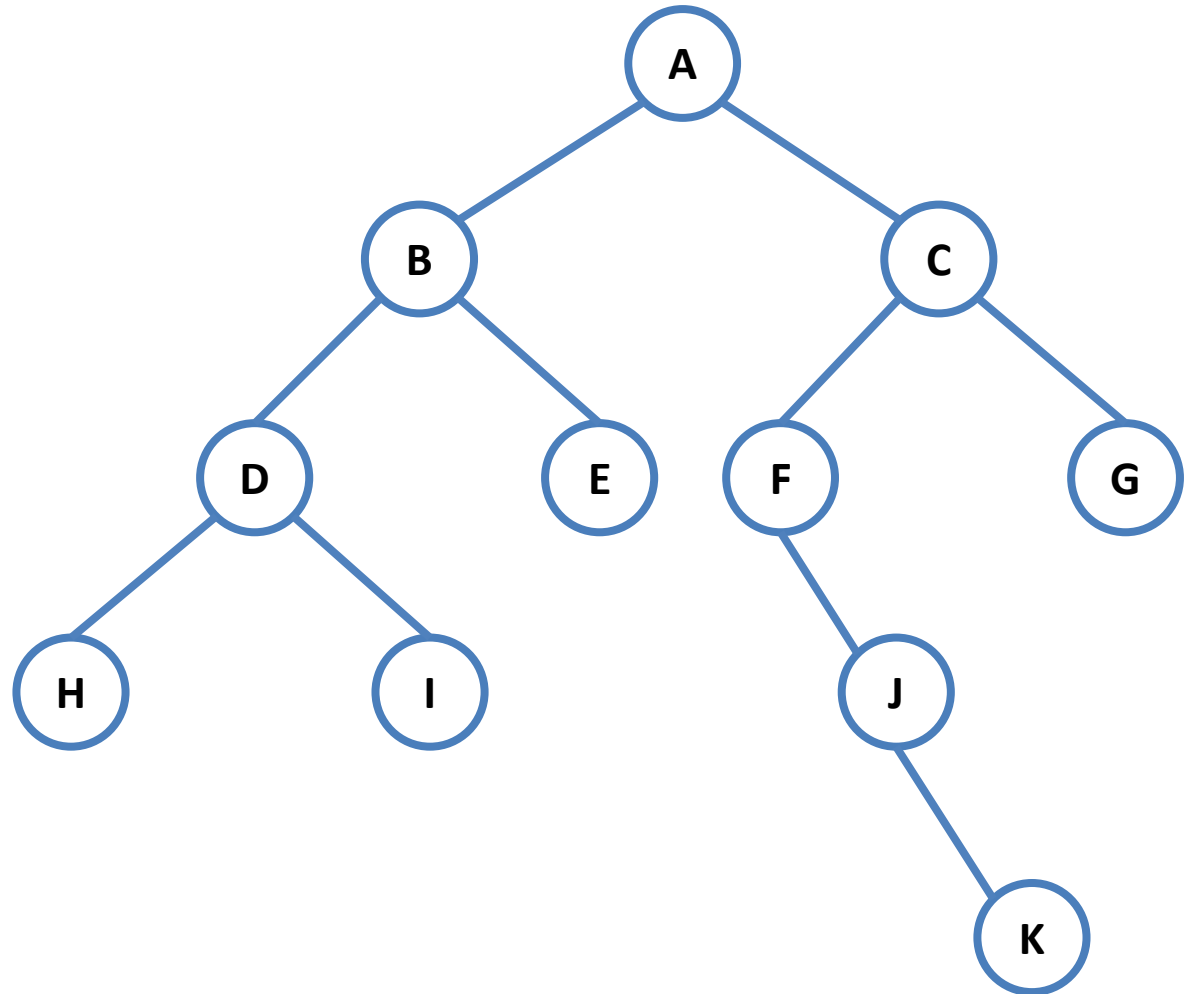


# Graph Traversal

- Start at A

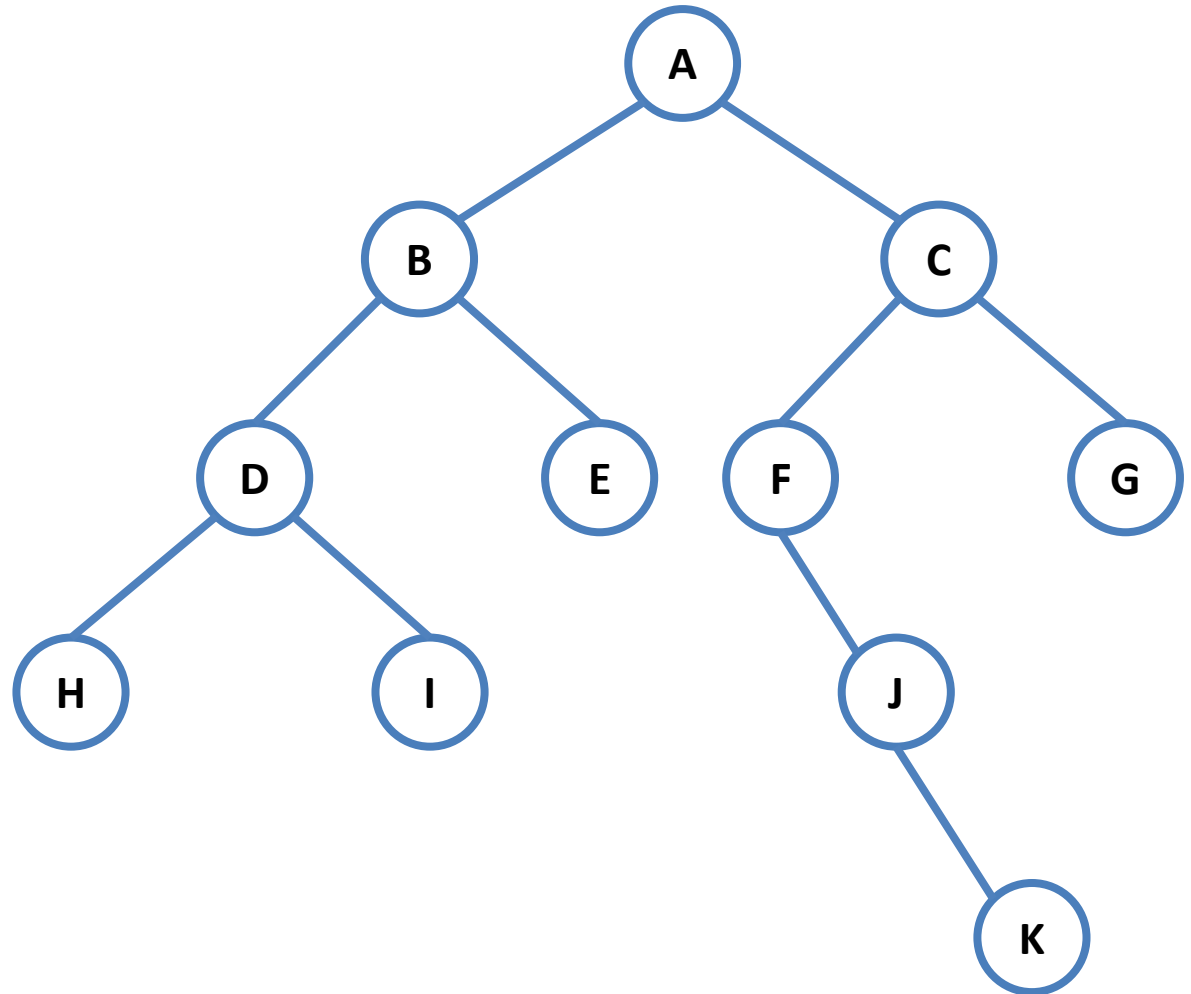


- Start at A, DFS



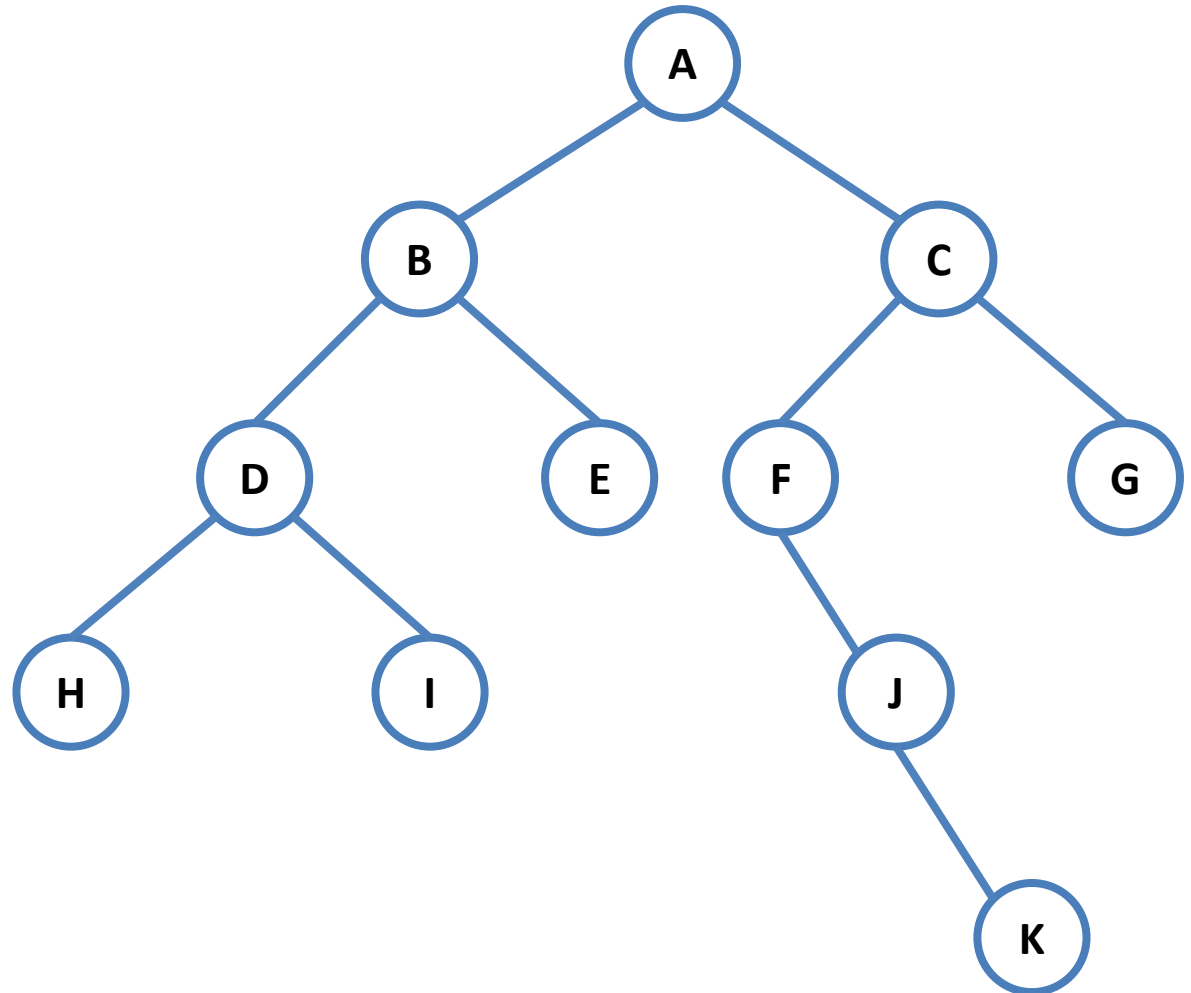
# Graph Traversal

- Start at A, DFS
- A



# Graph Traversal

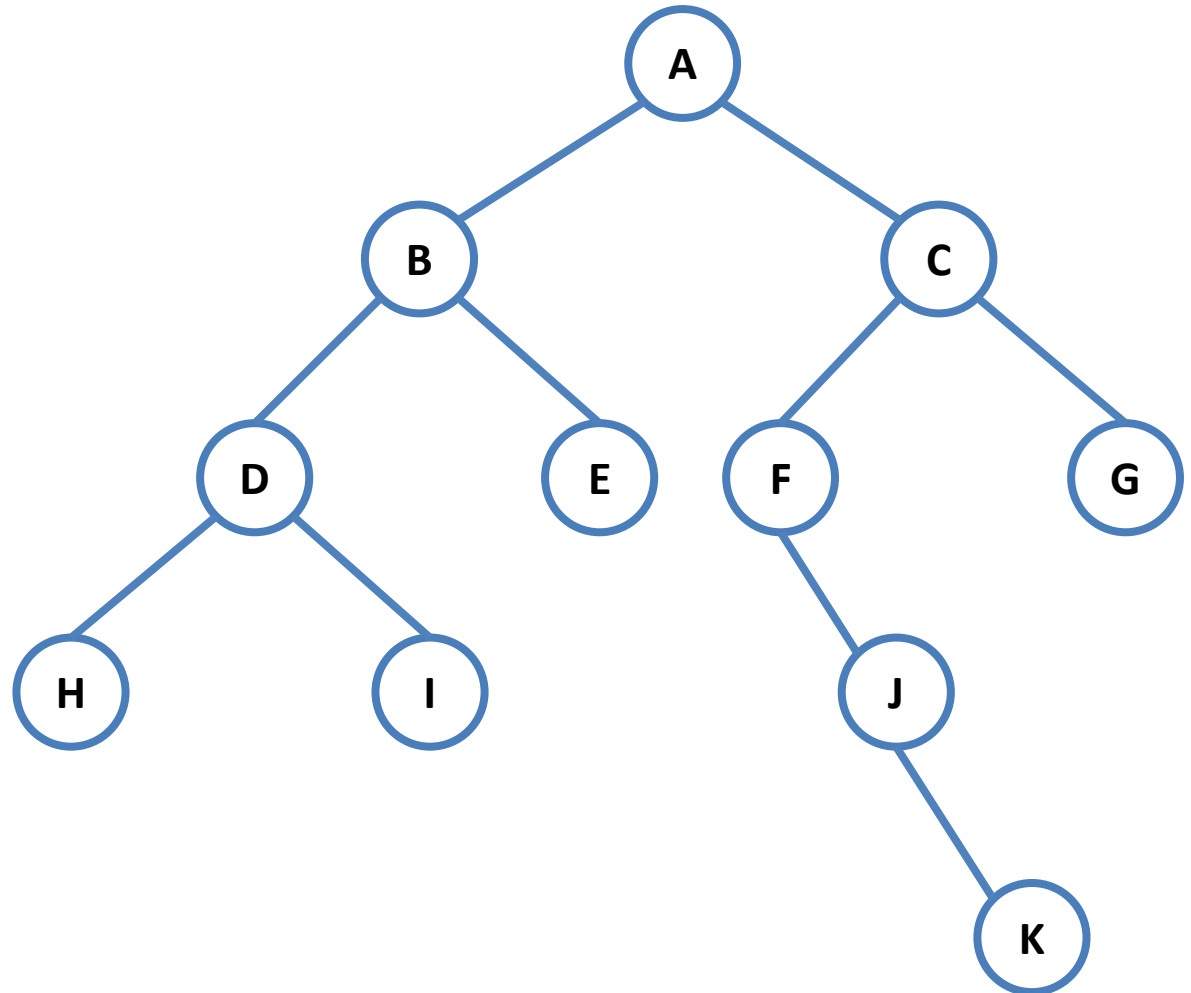
- Start at A, DFS
- A
- B





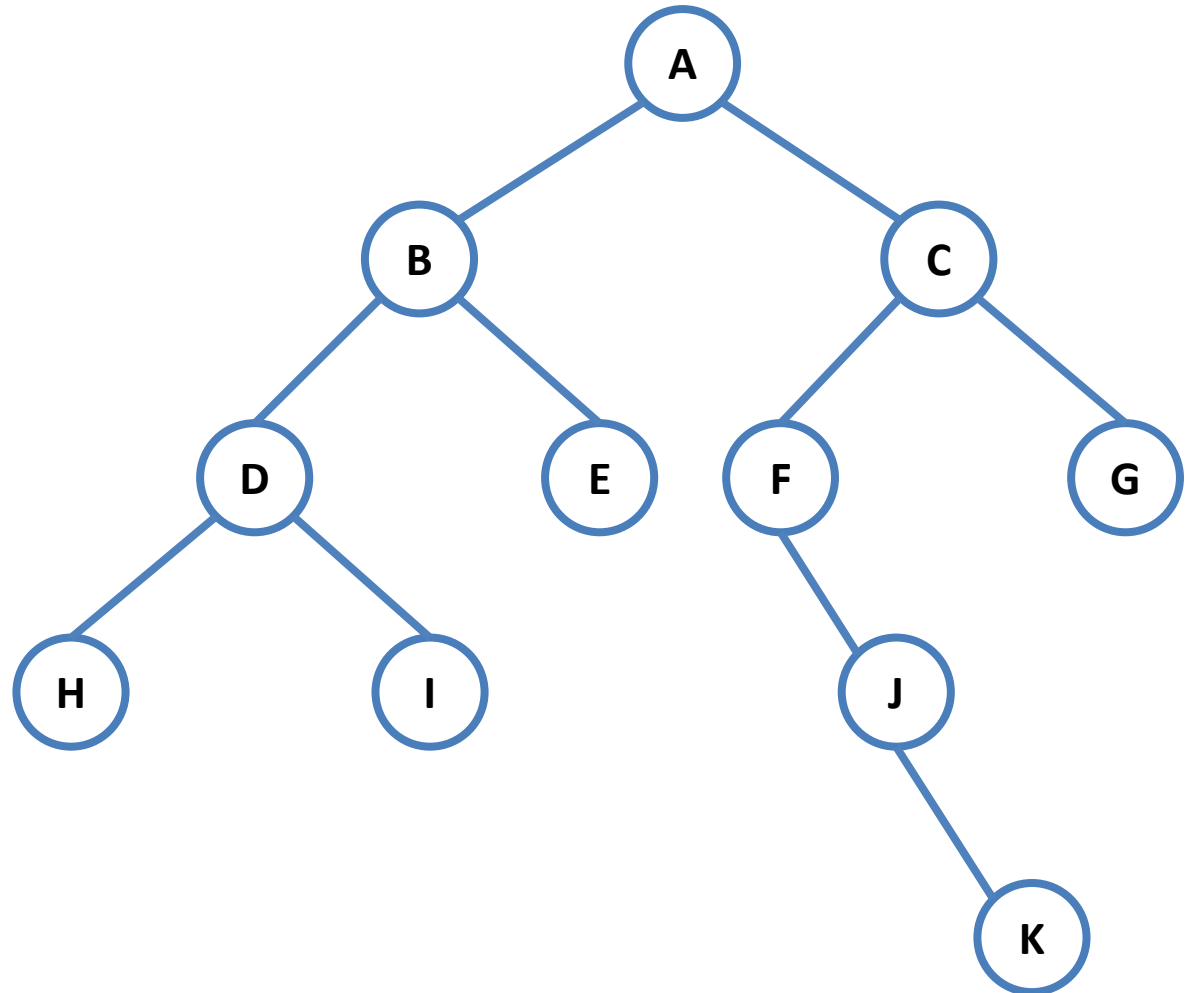
# Graph Traversal

- Start at A, DFS
- A
- B
- D, go deep!



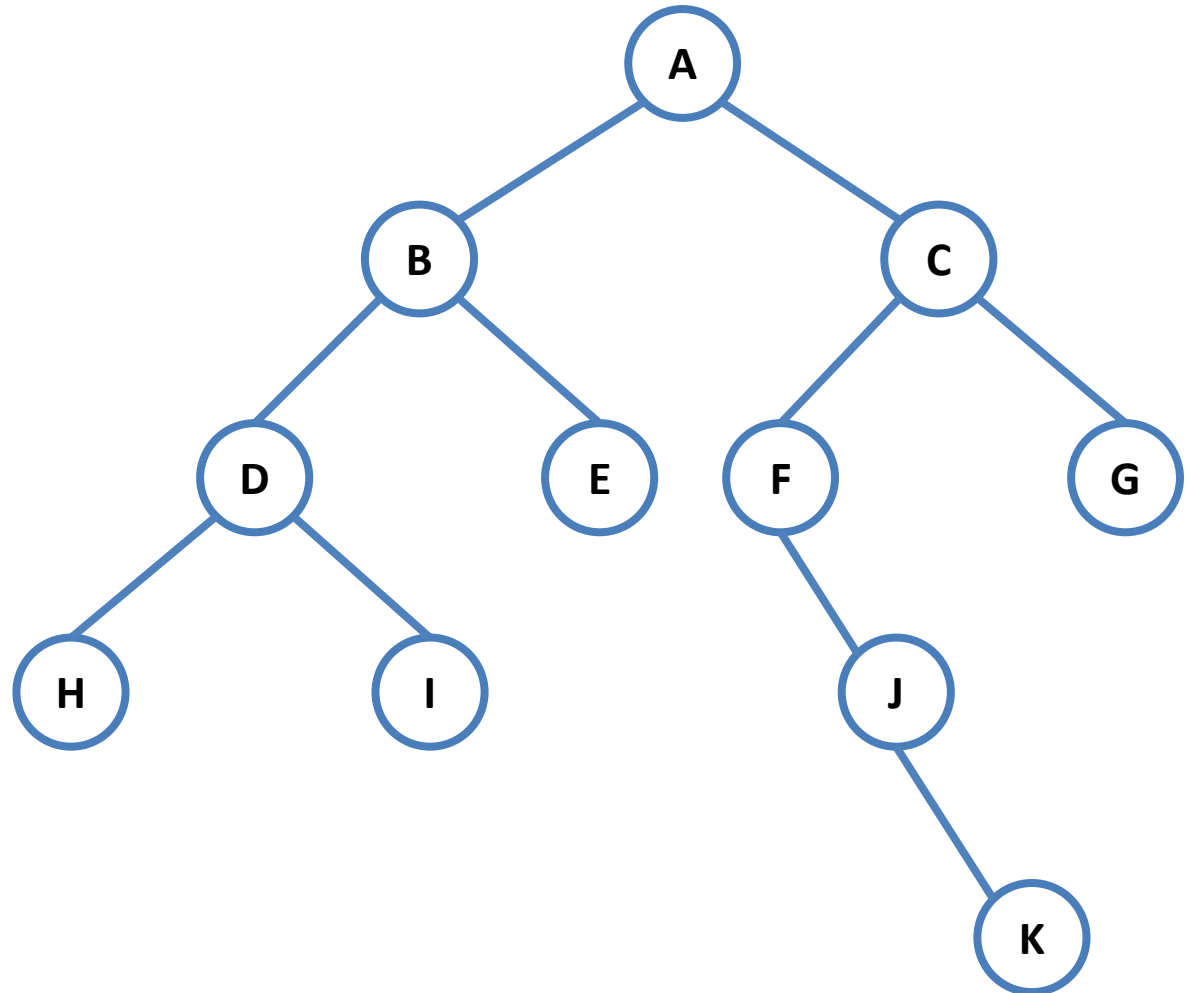
# Graph Traversal

- Start at A, DFS
- A
- B
- D
- H



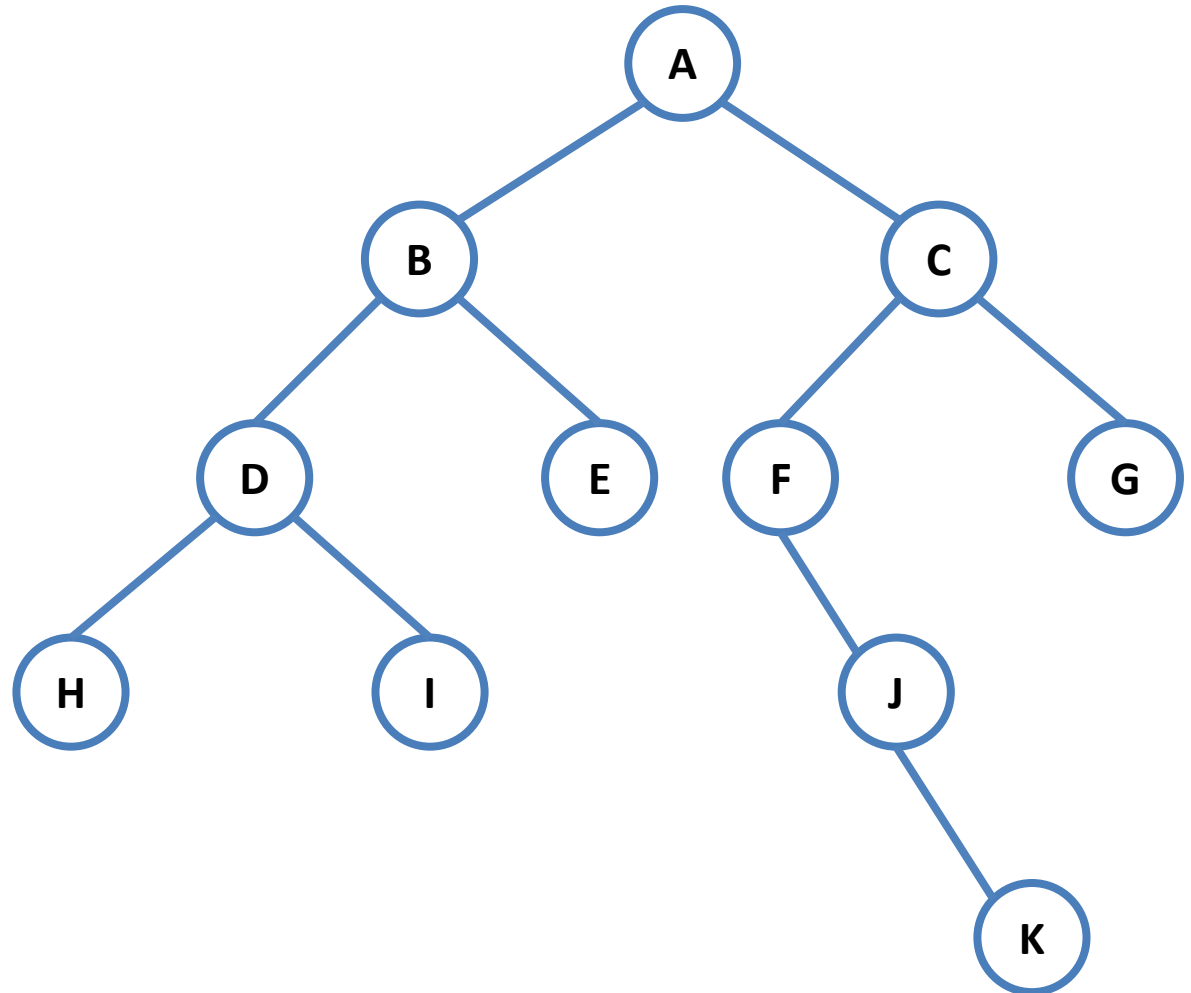
# Graph Traversal

- Start at A, DFS
- A
- B
- D
- H
- can't go deeper



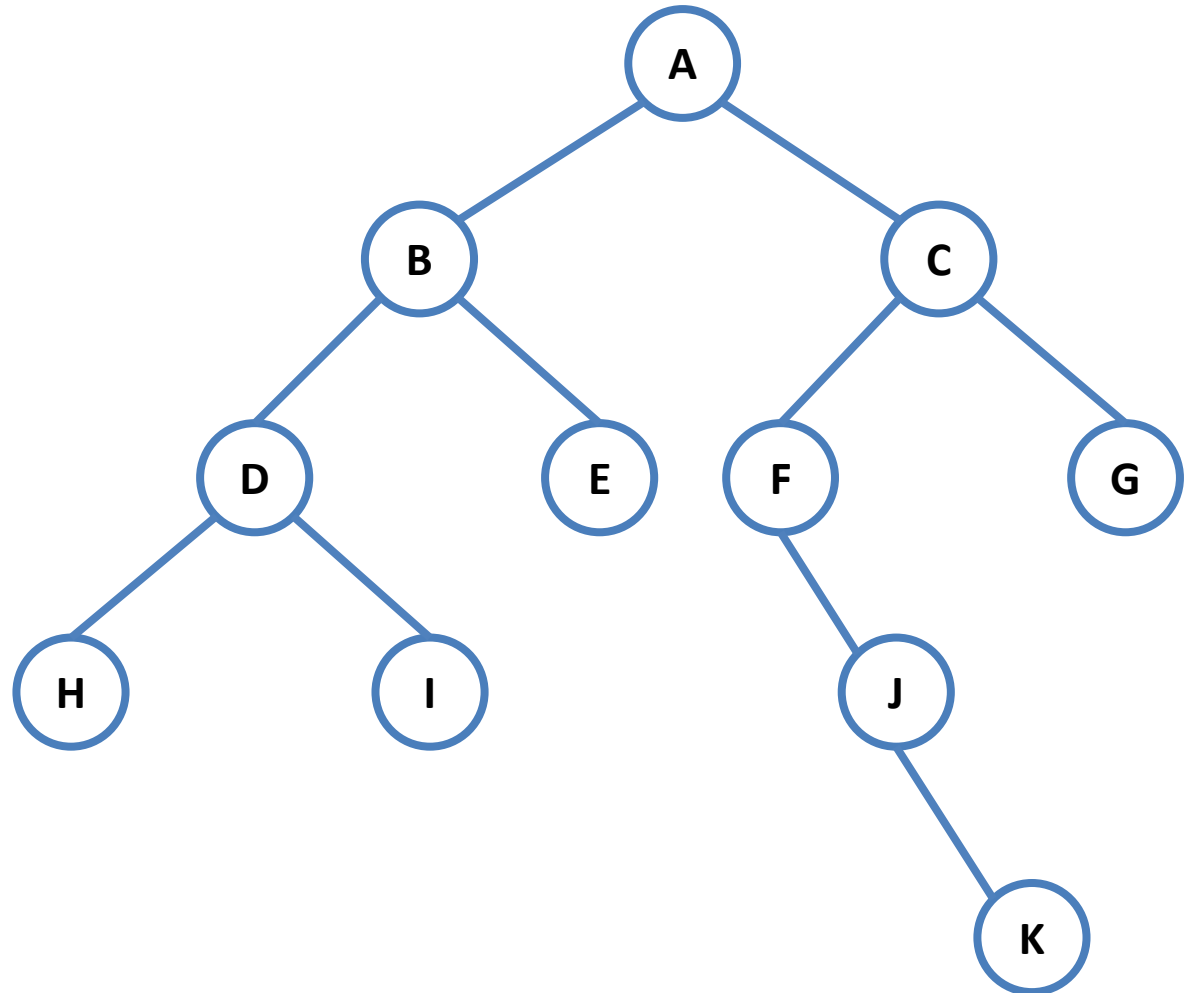
# Graph Traversal

- Start at A, DFS
- A
- B
- D
- H
- I



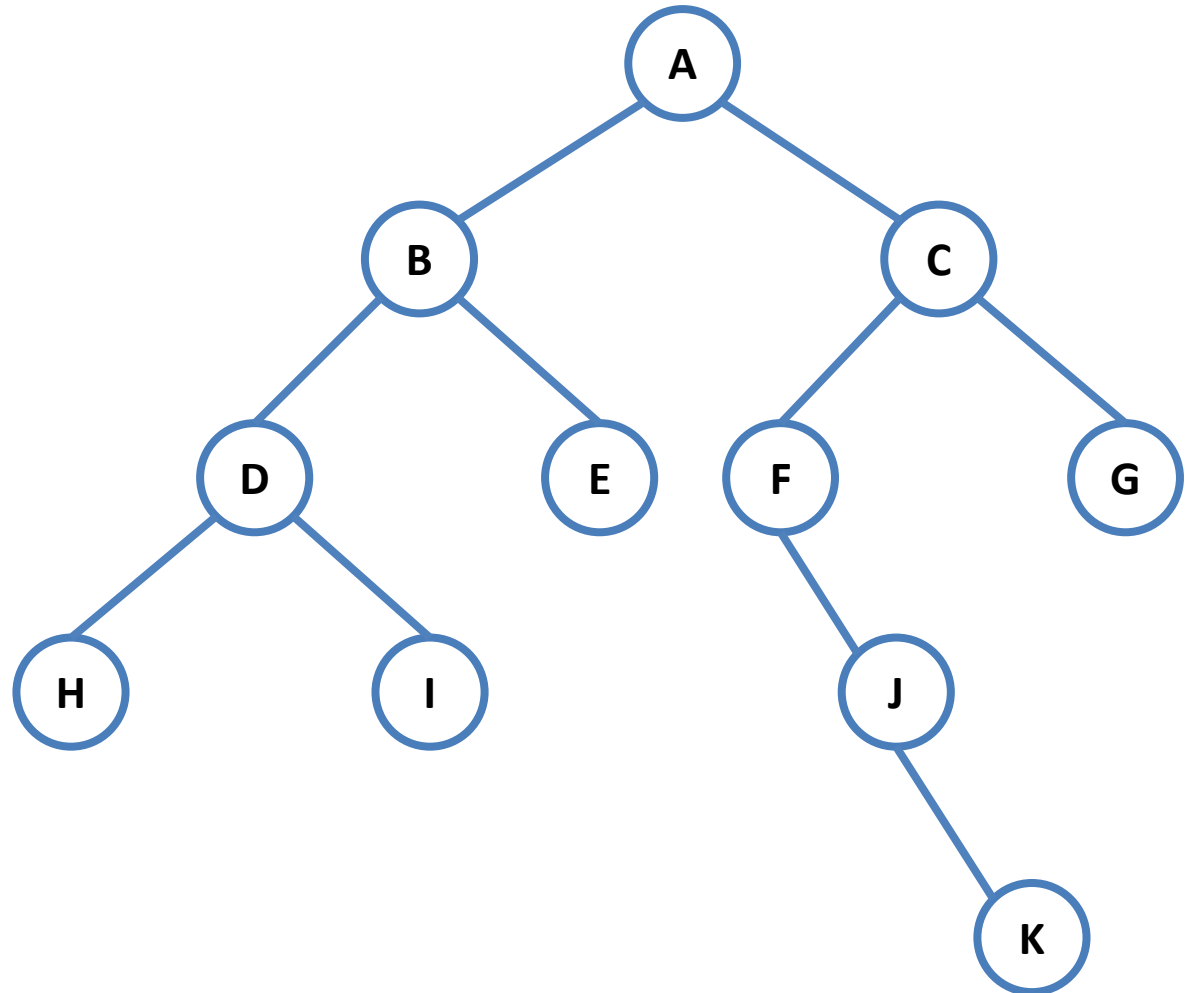
# Graph Traversal

- Start at A, DFS
- A
- B
- D
- H
- I
- E



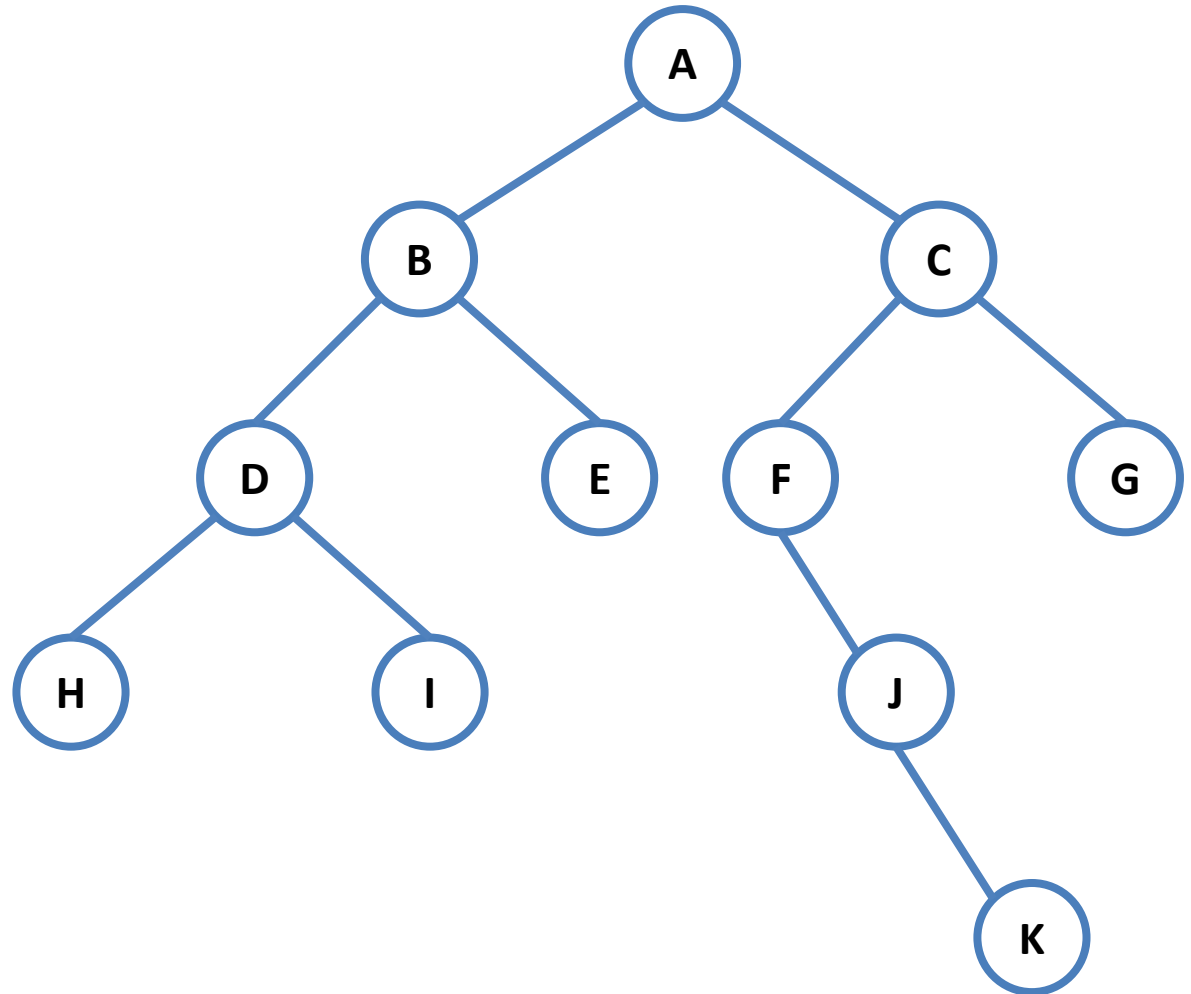
# Graph Traversal

- Start at A, DFS
- A
- B
- D
- H
- I
- E
- C



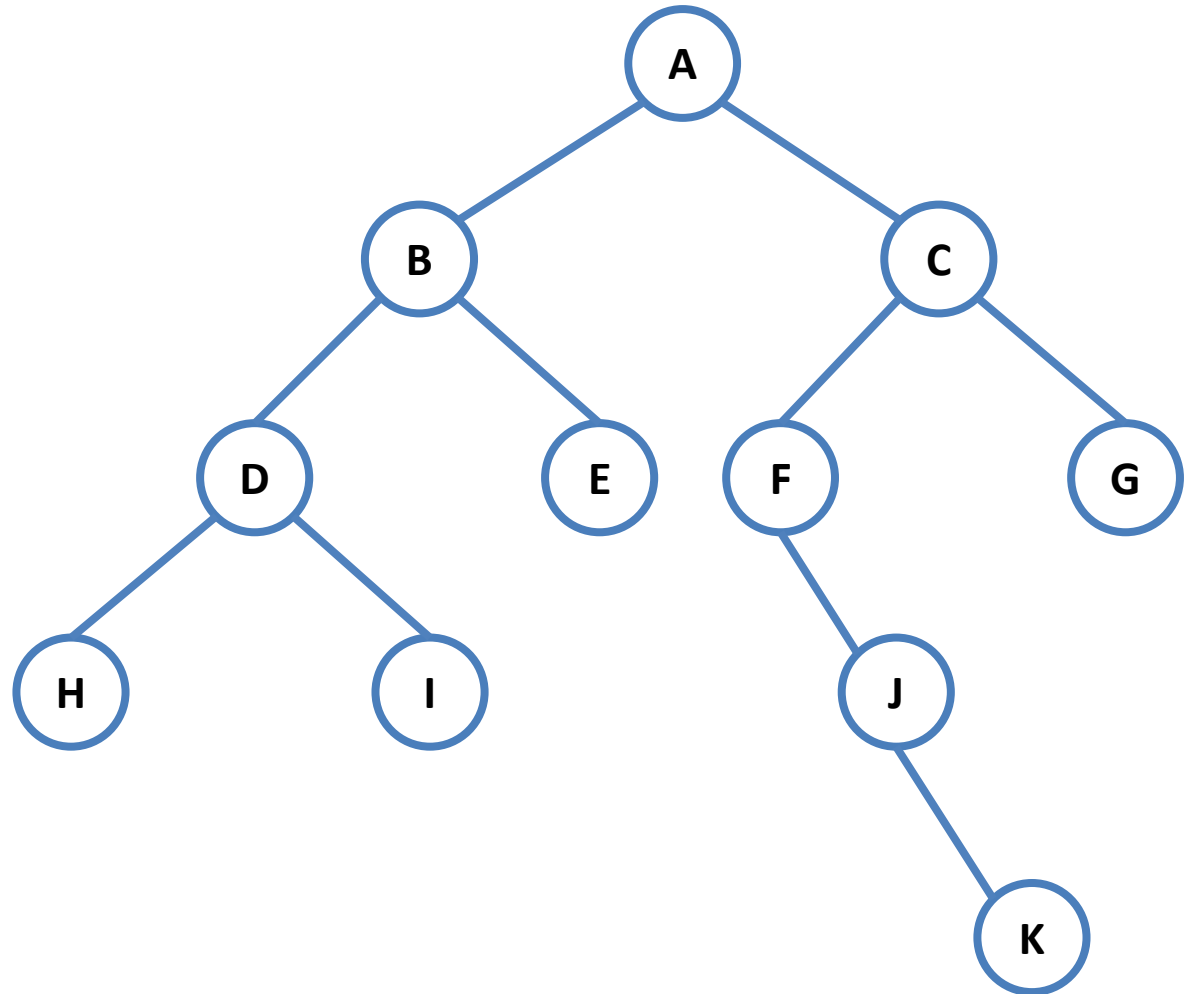
# Graph Traversal

- Start at A, DFS
- A
- B
- D
- H
- I
- E
- C
- F



# Graph Traversal

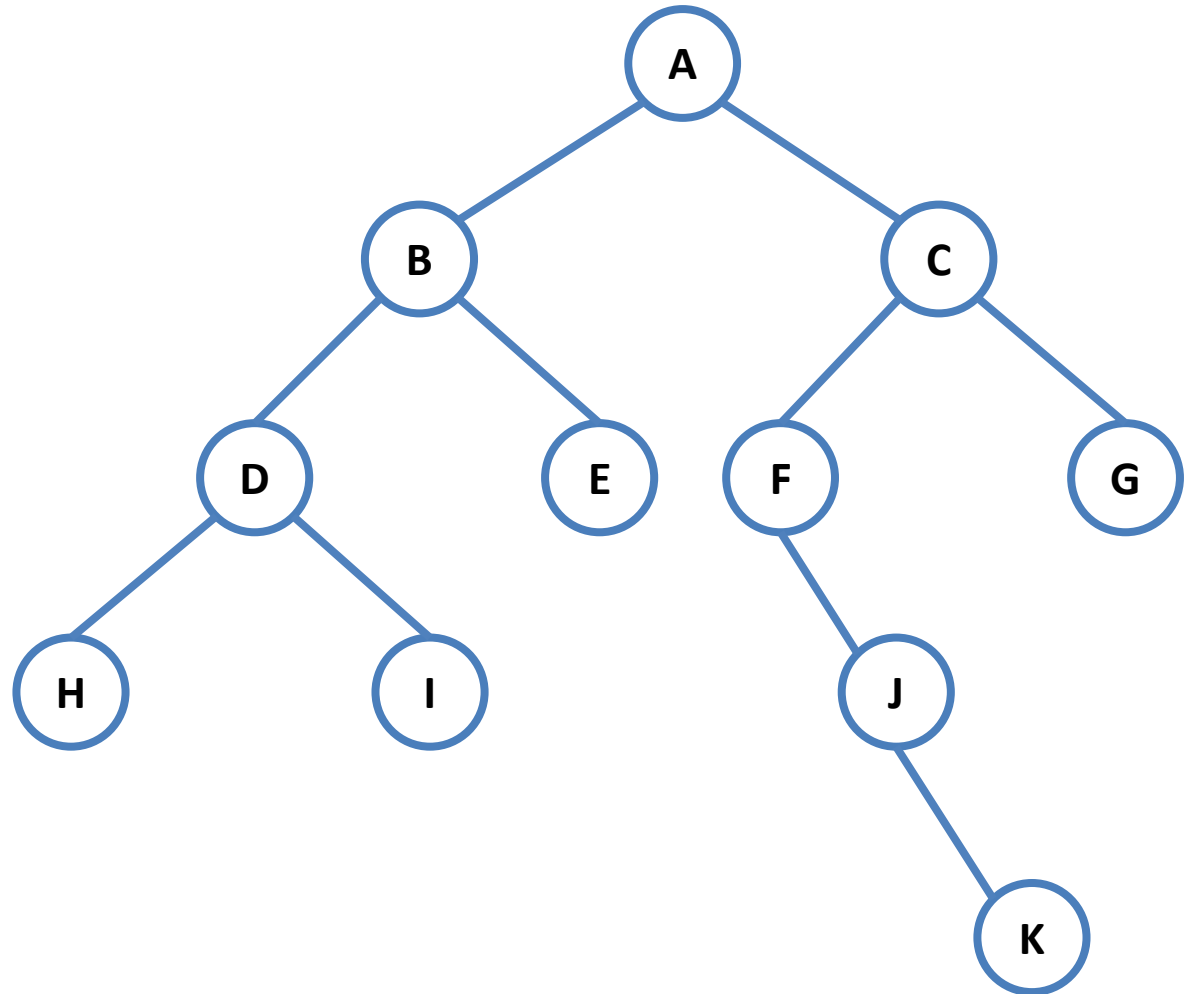
- Start at A, DFS
- A
- B
- D
- H
- I
- E
- C
- F
- J





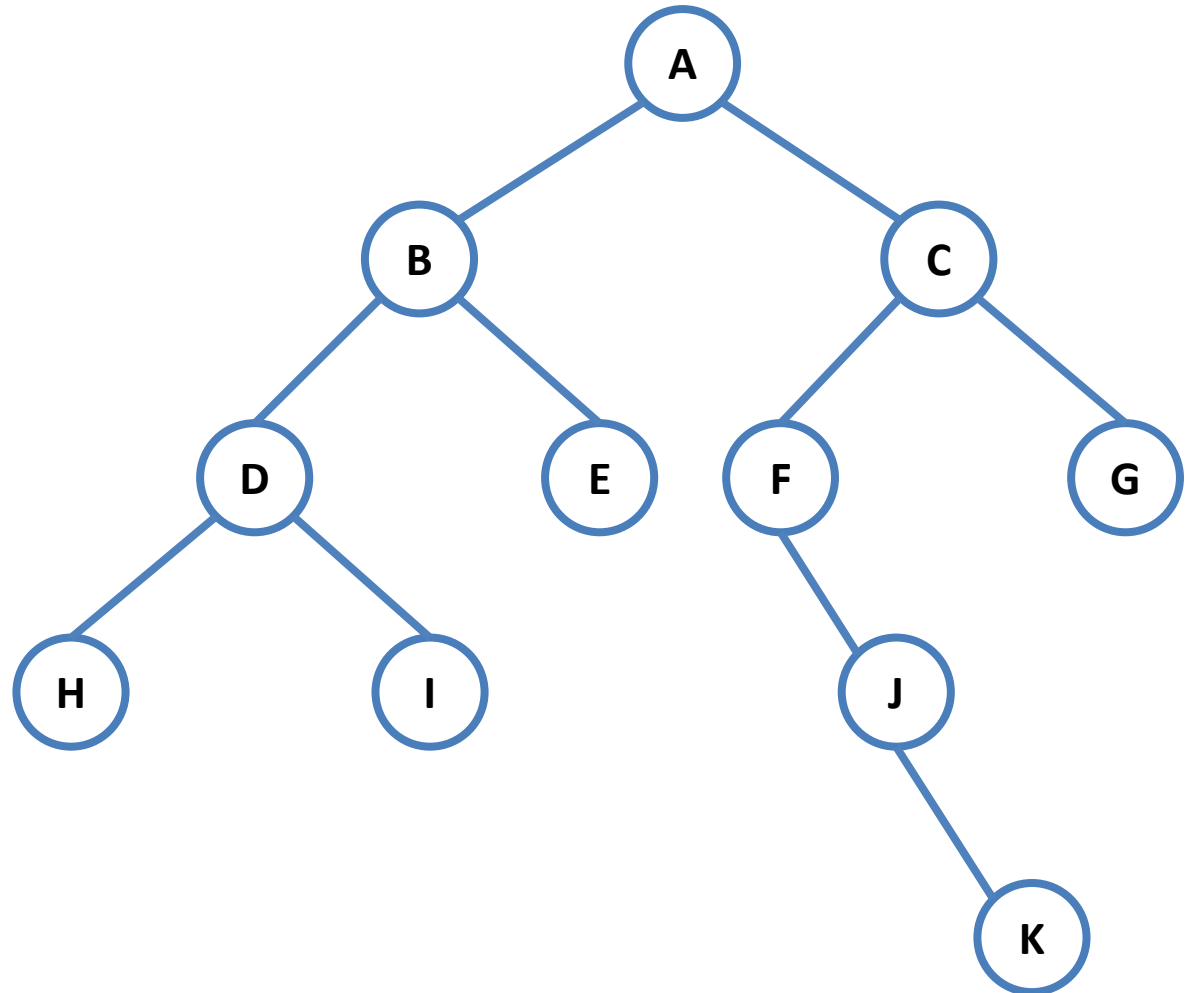
# Graph Traversal

- Start at A, DFS
- A
- B
- D
- H
- I
- E
- C
- F
- J
- K



# Graph Traversal

- Start at A, DFS
- B
- D
- H
- I
- E
- C
- F
- J
- K
- G, finally

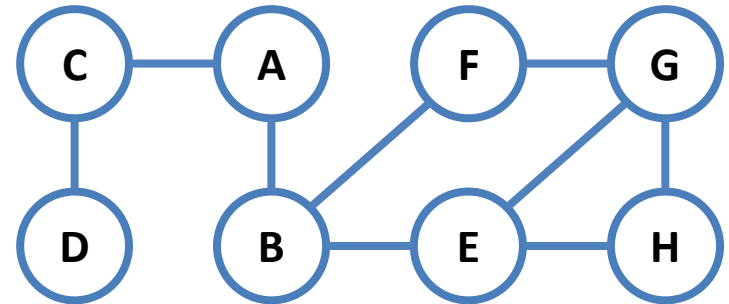


Questions?

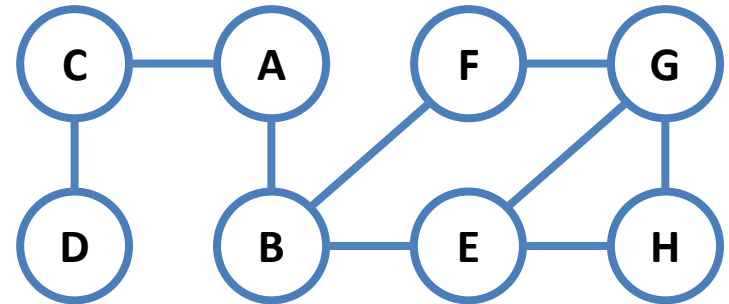
# Graph

## BFS Implementation

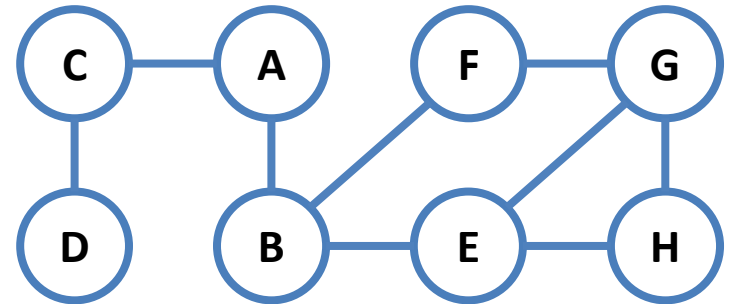
- How would you implement it?



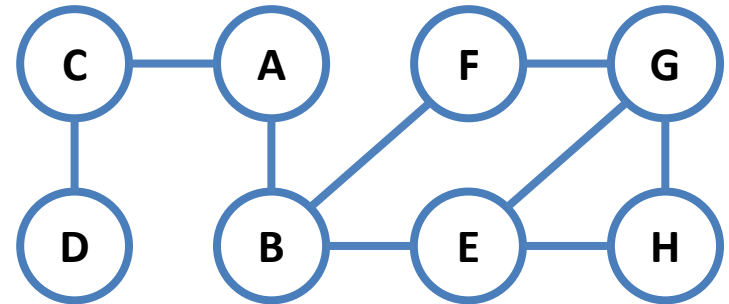
- How would you implement it?
  - Let say we begin from vertex A



- How would you implement it?
  - Let say we begin from vertex A
  - What is our traversal?

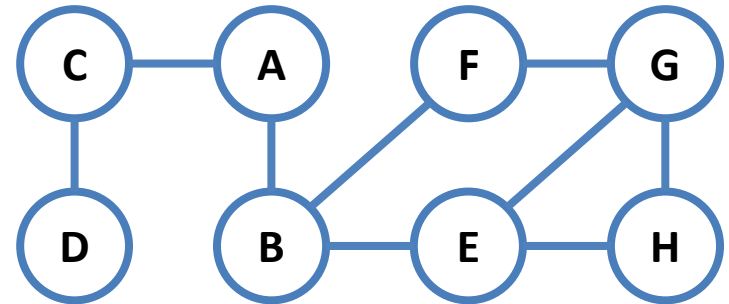


- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered



Discovered									
Visited									

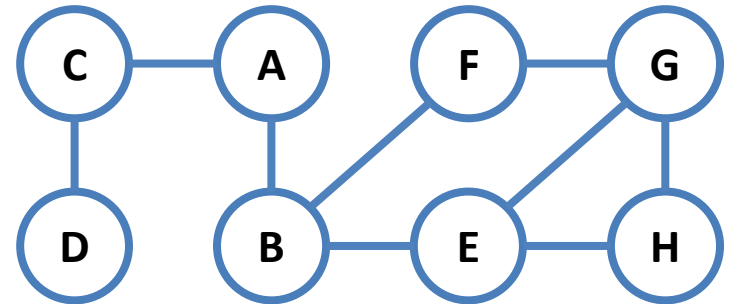
- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Start with A



Discovered	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>
Visited	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>

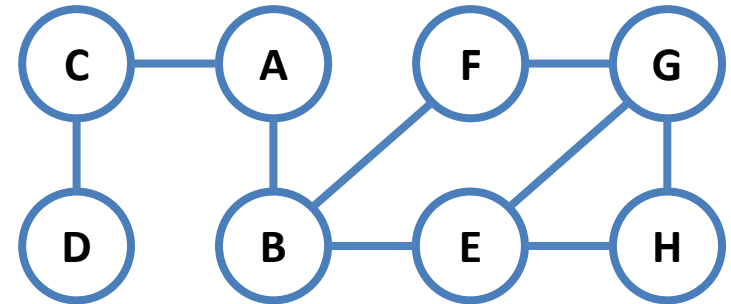


- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it



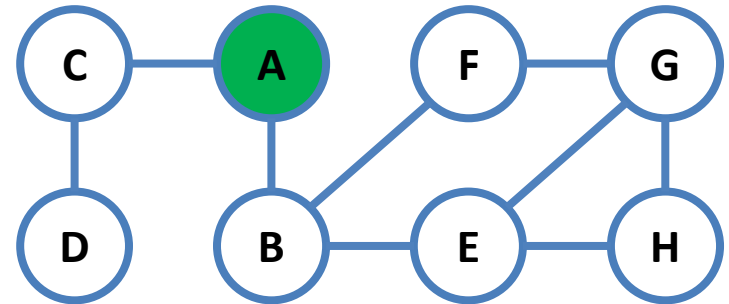
Discovered	A								
Visited									

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty



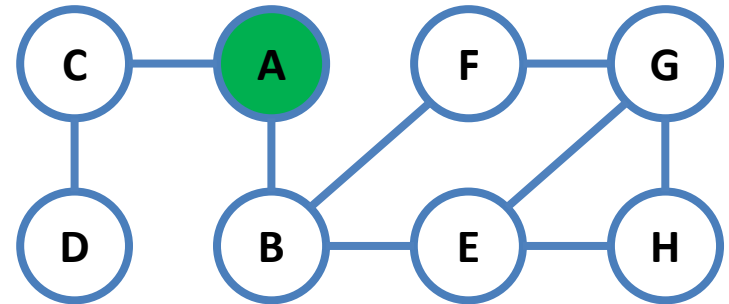
Discovered	A								
Visited									

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered



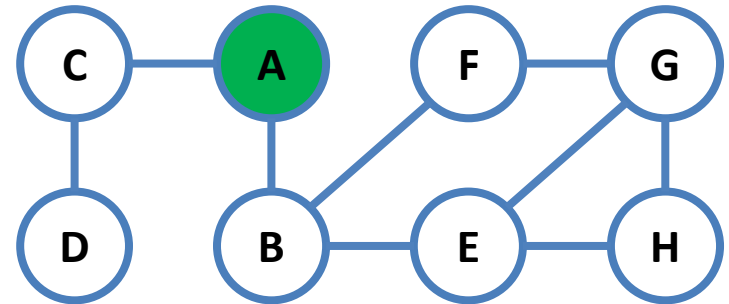
Discovered	A								
Visited									

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited



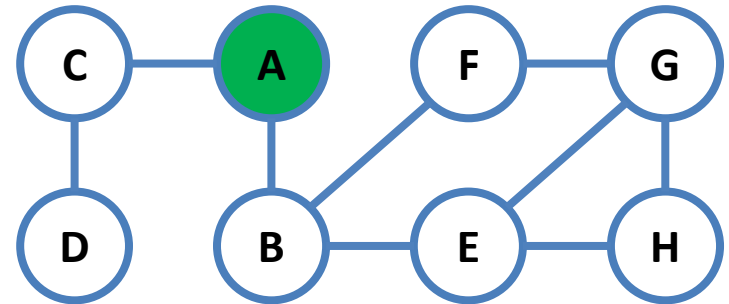
Discovered									
Visited	A								

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served



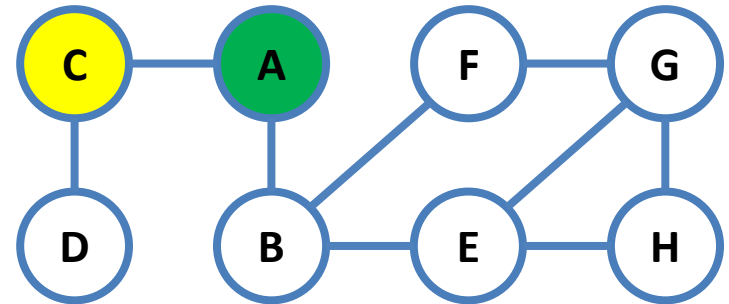
Discovered									
Visited	A								

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



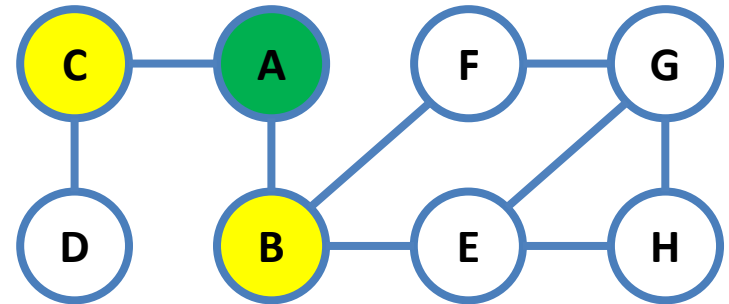
Discovered									
Visited	A								

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



Discovered	C								
Visited	A								

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



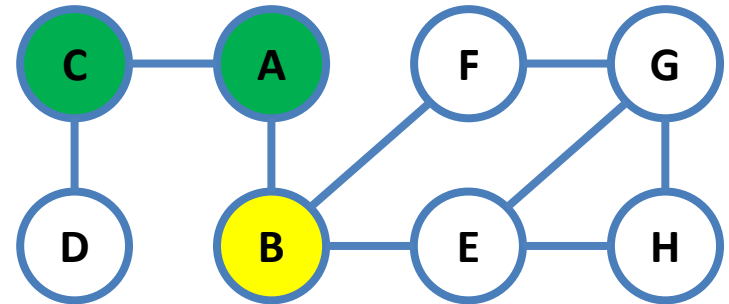
Discovered	C	B							
Visited	A								



# Graph

## BFS Implementation

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue

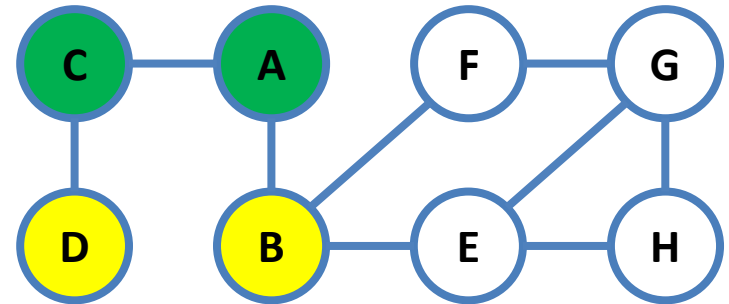


Discovered		B							
Visited	A	C							

# Graph

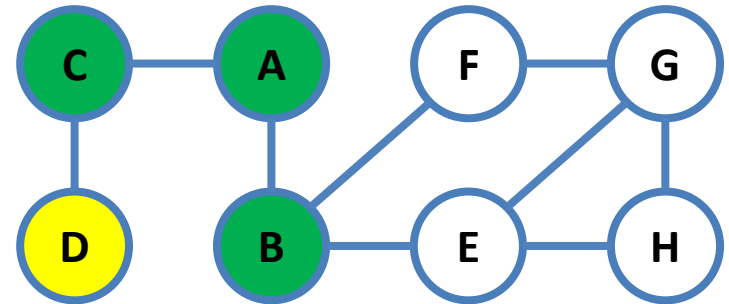
## BFS Implementation

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



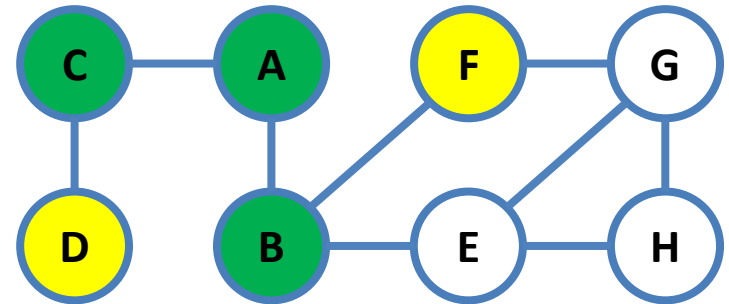
Discovered		B	D						
Visited	A	C							

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



Discovered			D						
Visited	A	C	B						

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue

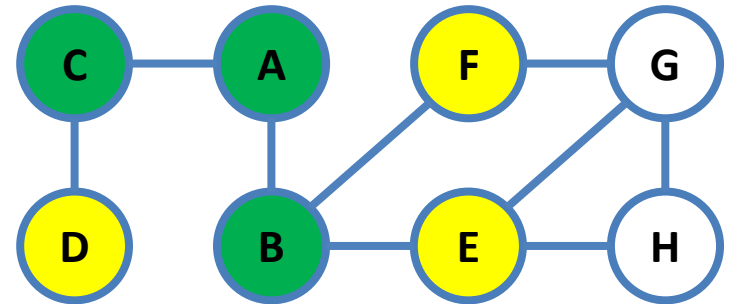


Discovered			D	F						
Visited	A	C	B							

# Graph

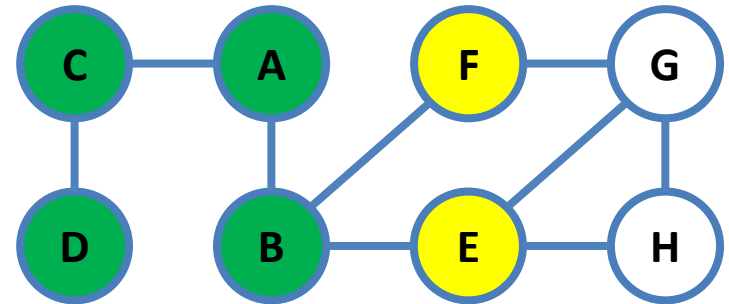
## BFS Implementation

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



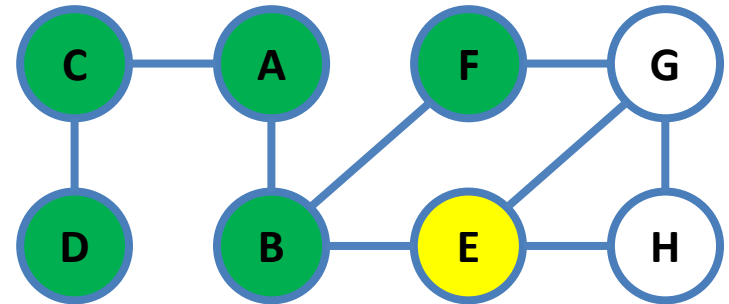
Discovered			D	F	E				
Visited	A	C	B						

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



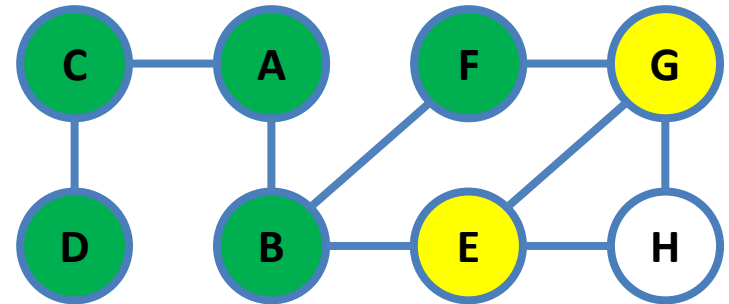
Discovered				F	E					
Visited	A	C	B	D						

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



Discovered					E					
Visited	A	C	B	D	F					

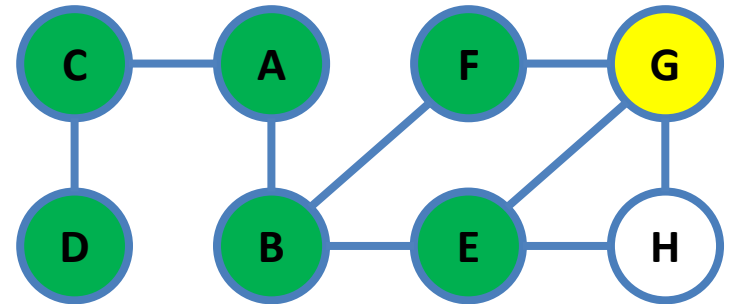
- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



Discovered					E	G				
Visited	A	C	B	D	F					

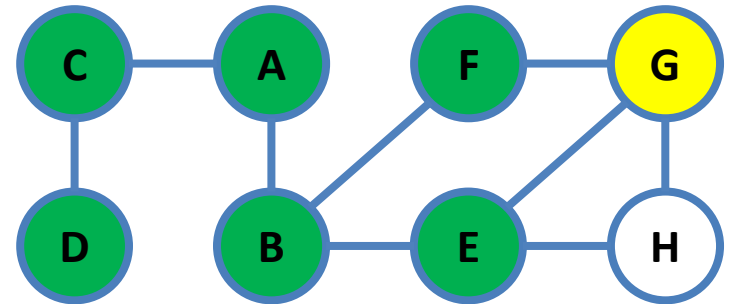


- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



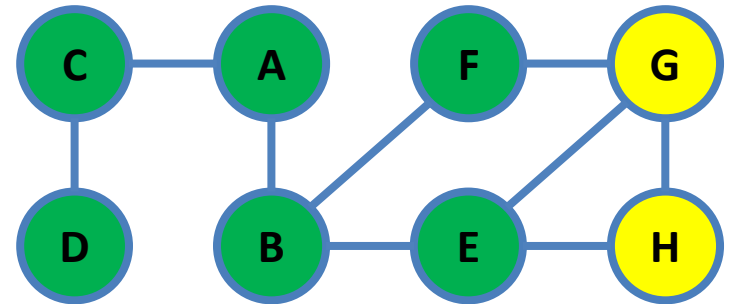
Discovered						G				
Visited	A	C	B	D	F	E				

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



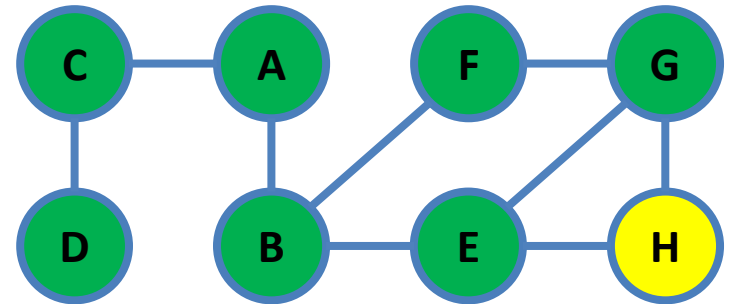
Discovered						G	G?			
Visited	A	C	B	D	F	E				

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



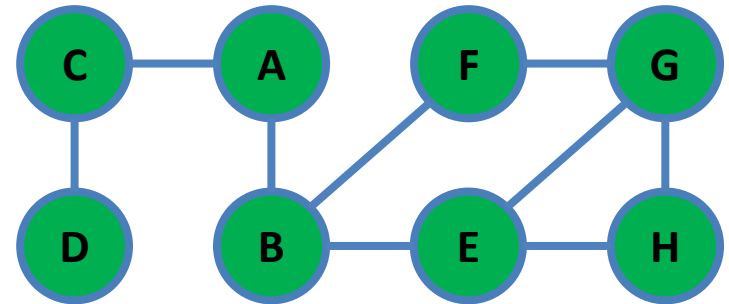
Discovered						G	H			
Visited	A	C	B	D	F	E				

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



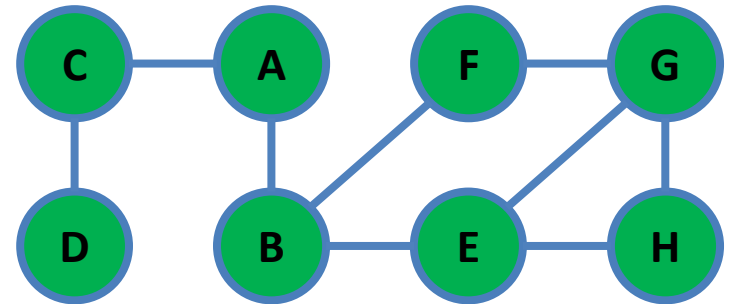
Discovered							H			
Visited	A	C	B	D	F	E	G			

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



Discovered									
Visited	A	C	B	D	F	E	G	H	

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue
- The traversal answer is not unique



Discovered									
Visited	A	C	B	D	F	E	G	H	

- Complexity?

# Graph

## BFS Implementation

- Complexity?
  - Time is  $O(V+E)$



- Complexity?
  - Time is  $O(V+E)$ 
    - Each vertex is visited once

# Graph

## BFS Implementation

- Complexity?
  - Time is  $O(V+E)$ 
    - Each vertex is visited once
    - Each edge is visited twice

# Graph

## BFS Implementation

- Complexity?
  - Time is  $O(V+E)$ 
    - Each vertex is visited once
    - Each edge is visited twice
      - For each  $\langle u, v \rangle$  we visit from  $u$  and also from  $v$   
directed only visit once

# Graph

## BFS Implementation

- Complexity?
  - Time is  $O(V+E)$ 
    - Each vertex is visited once
    - Each edge is visited twice
      - For each  $\langle u, v \rangle$  we visit from  $u$  and also from  $v$
  - Space is  $O(V+E)$

# Graph

## BFS Implementation

- Complexity?
  - Time is  $O(V+E)$ 
    - Each vertex is visited once
    - Each edge is visited twice
      - For each  $\langle u, v \rangle$  we visit from  $u$  and also from  $v$
  - Space is  $O(V+E)$ 
    - $V$  maximum for the discovered queue

# Graph

## BFS Implementation

- Complexity?
  - Time is  $O(V+E)$ 
    - Each vertex is visited once
    - Each edge is visited twice
      - For each  $\langle u, v \rangle$  we visit from  $u$  and also from  $v$
  - Space is  $O(V+E)$ 
    - $V$  maximum for the discovered queue
    - $E$  to stored all of the edges (adjacency list)

# Graph

## BFS Implementation

- Complexity?
  - Time is  $O(V+E)$ 
    - Each vertex is visited once
    - Each edge is visited twice
      - For each  $\langle u, v \rangle$  we visit from  $u$  and also from  $v$
  - Space is  $O(V+E)$ 
    - $V$  maximum for the discovered queue
    - $E$  to stored all of the edges (adjacency list)
  - But don't we need to check the discovered queue for each vertex  $v$ ?
    - $O(V)$  search through the queue?

# Graph

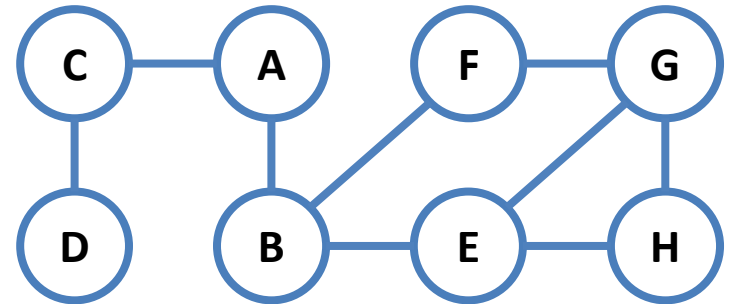
## BFS Implementation

- Complexity?
  - Time is  $O(V+E)$ 
    - Each vertex is visited once
    - Each edge is visited twice
      - For each  $\langle u, v \rangle$  we visit from  $u$  and also from  $v$
  - Space is  $O(V+E)$ 
    - $V$  maximum for the discovered queue
    - $E$  to stored all of the edges (adjacency list)
  - But don't we need to check the discovered queue for each vertex  $v$ ?
    - $O(V)$  search through the queue?
    - **NO!** Implement a Node class with `self.discovered = True/ False`  
in individual vertex

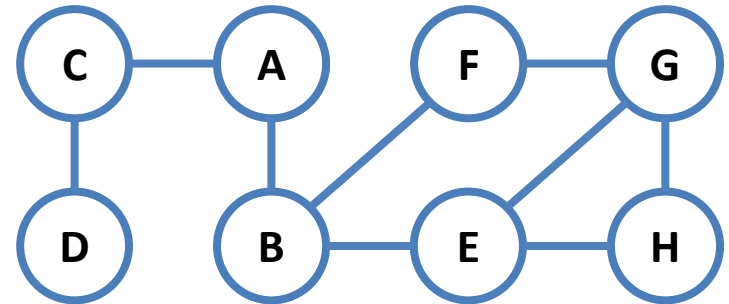


Questions?

- How would you implement it?
  - Let say we begin from vertex A
  - What is our DFS traversal?

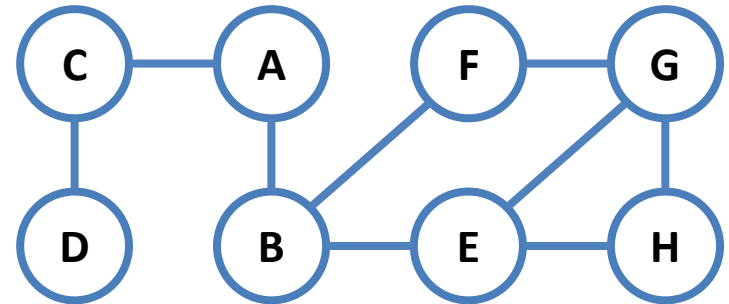


- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered



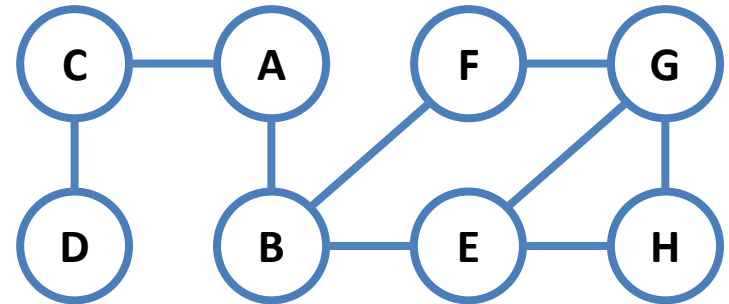
Discovered									
Visited									

- How would you implement it?
  - Let say we begin from vertex A
  - Have a ~~queue~~ **stack** for discovered



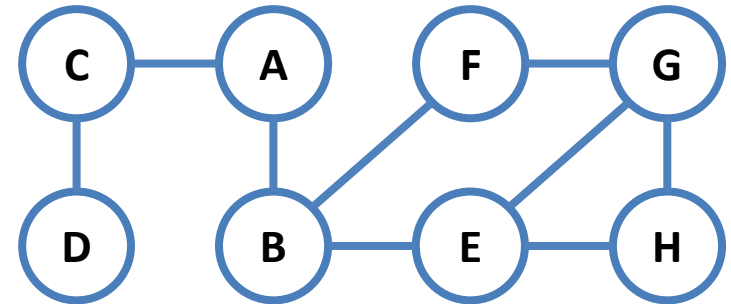
Discovered	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>
Visited	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>	<div></div>

- How would you implement it?
  - Let say we begin from vertex A
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    - Push source (A) into it



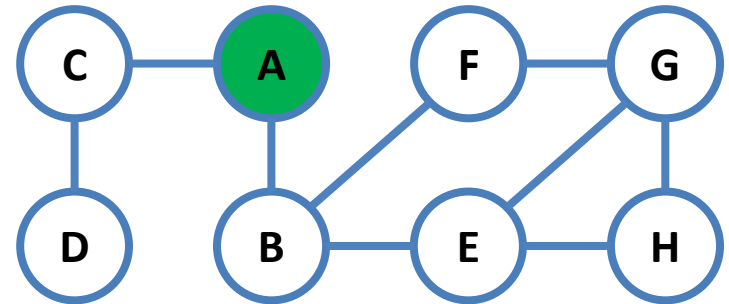
Discovered	A								
Visited									

- How would you implement it?
  - Let say we begin from vertex A
  - Have a stack for discovered
    - Push source (A) into it
  - While discovered is not empty
    - Pop from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue



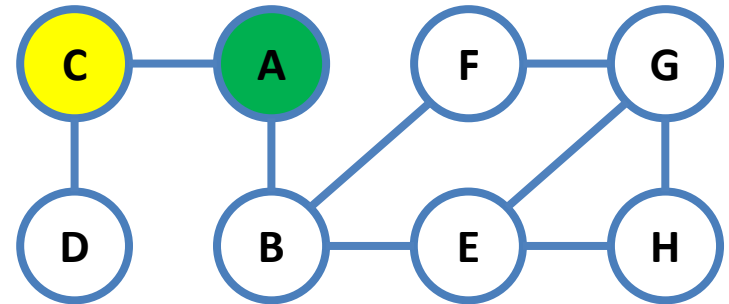
Discovered	A								
Visited									

- How would you implement it?
  - Let say we begin from vertex A
  - Have a **stack for discovered**
    - Push source (A) into it
  - While discovered is not empty
    - **Pop from discovered, to visited**
    - For **each edge  $\langle u, v \rangle$**  where **u** is the **served**
      - If vertex v is **not discovered** or **visited**, **add to discovered queue**



Discovered									
Visited	A								

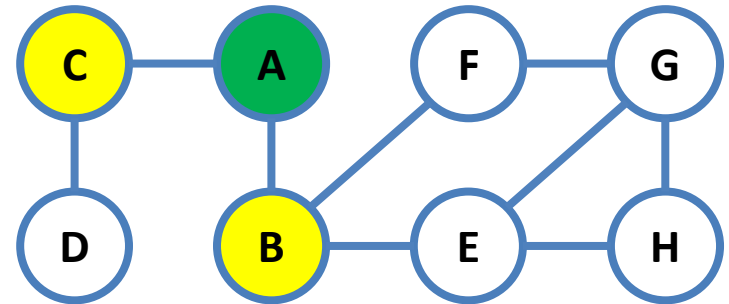
- How would you implement it?
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Discovered	C								
Visited	A								

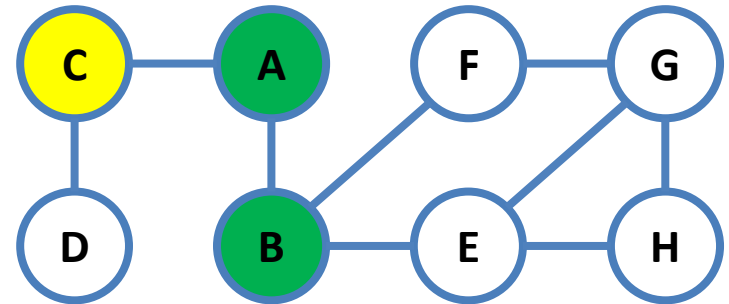


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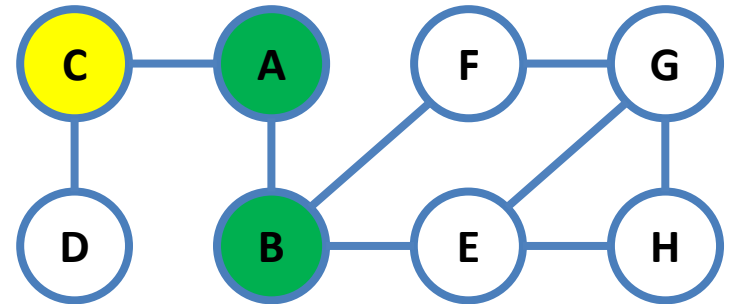
Discovered	C	B							
Visited	A								

- How would you implement it?
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    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
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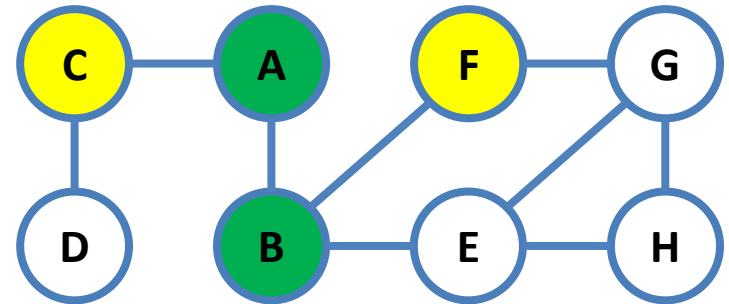
Discovered	C								
Visited	A	B							

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Discovered	C	A?							
Visited	A	B							

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  - Let say we begin from vertex A
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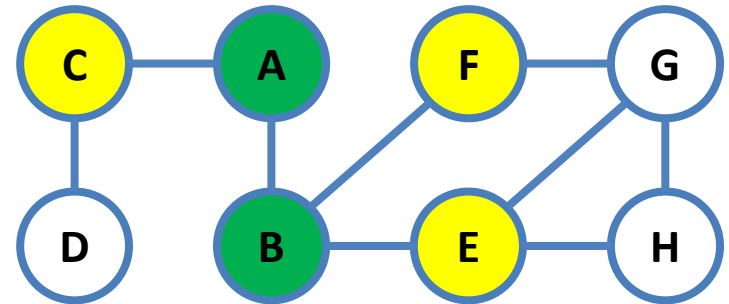


Discovered	C	F							
Visited	A	B							

# Graph

## DFS Implementation

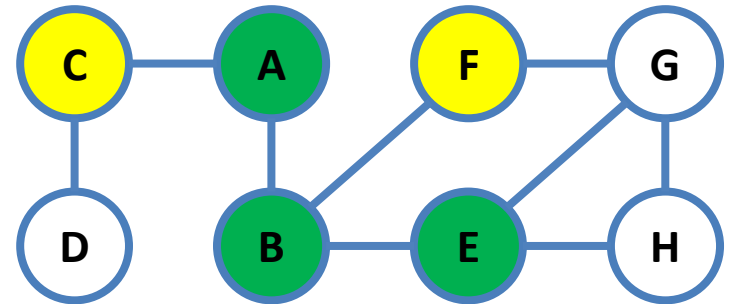
- How would you implement it?
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DFS pop B and push to visited (last in first out)  
BFS serve C and append to visited (first in first out)

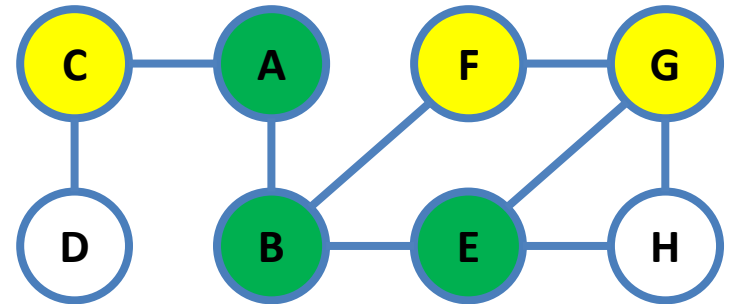
Discovered	C	F	E						
Visited	A	B							

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Discovered	C	F							
Visited	A	B	E						

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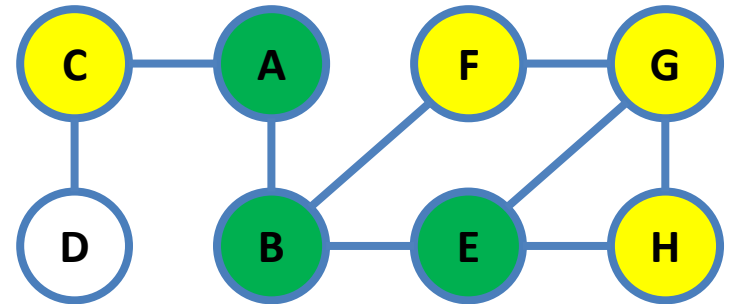


Discovered	C	F	G						
Visited	A	B	E						

# Graph

## DFS Implementation

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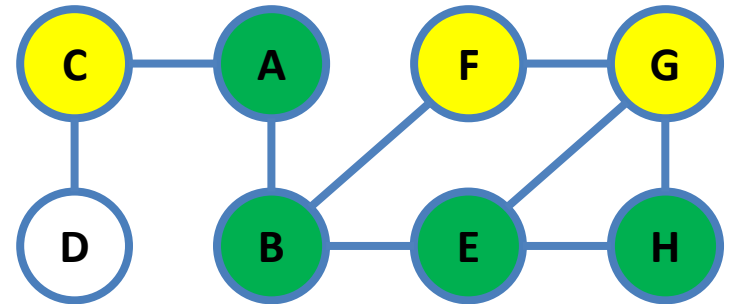
Discovered	C	F	G	H						
Visited	A	B	E							



# Graph

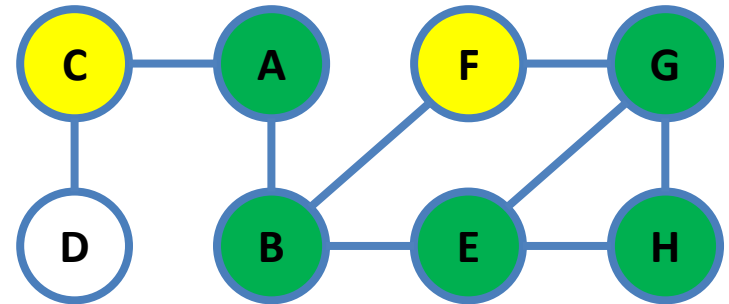
## DFS Implementation

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Discovered	C	F	G						
Visited	A	B	E	H					

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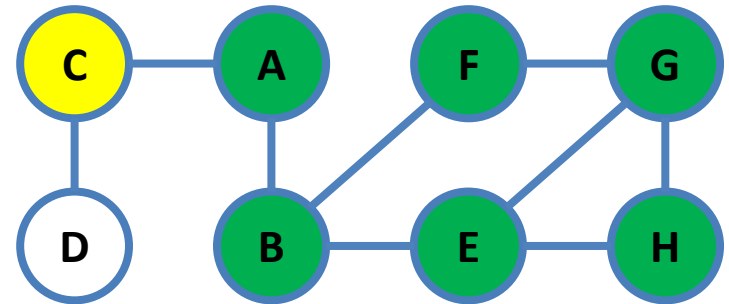


Discovered	C	F							
Visited	A	B	E	H	G				

# Graph

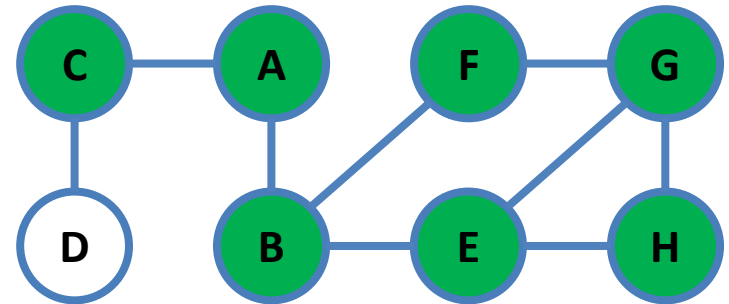
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Discovered	C								
Visited	A	B	E	H	G	F			

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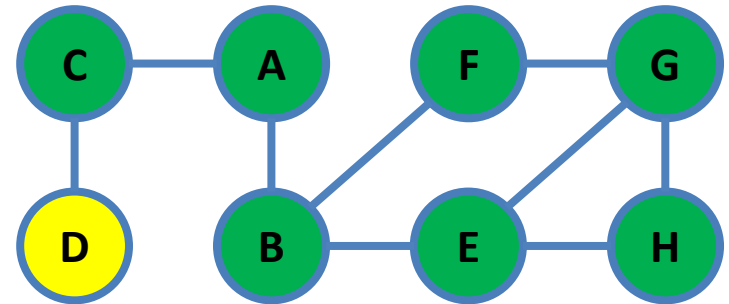


Discovered									
Visited	A	B	E	H	G	F	C		

# Graph

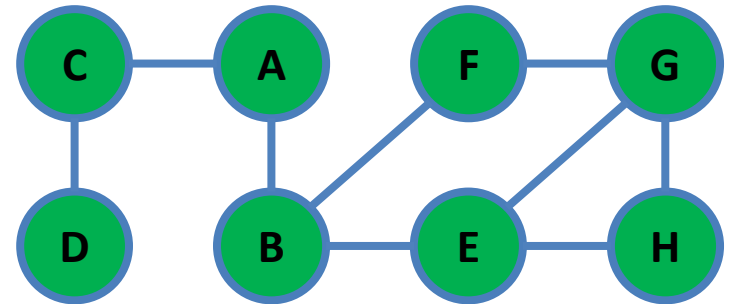
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Discovered	<b>D</b>								
Visited	A	B	E	H	G	F	C		

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Discovered									
Visited	A	B	E	H	G	F	C	D	

# Graph

## DFS Implementation

- Complexity?

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  - Time is  $O(V+E)$ 
    - Explanation same as BFS



# Graph

## DFS Implementation

- Complexity?
  - Time is  $O(V+E)$ 
    - Explanation same as BFS
  - Space is  $O(V+E)$ 
    - Explanation same as DFS

# Graph

## DFS Implementation

- Can you think of another way to implement DFS?

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- Recursion!
  - Best way to implement any form of traversal
    - Just like when you implemented tree/ trie traversal

# Graph

## DFS Implementation

- Can you think of another way to implement DFS?
- Recursion!
  - Best way to implement any form of traversal
    - Just like when you implemented tree/ trie traversal
- Let us just write them all out as a live coding session!

Questions?

- Can you think of another way to implement DFS?
- Recursion!
  - Best way to implement any form of traversal
    - Just like when you implemented tree/ trie traversal  
Recursion stack is the discovered stack in DFS

```
1 def dfs(current_vertex):
2     current_vertex.visited = True
3     for next_vertex in current_vertex.adjacent:
4         if next_vertex.visited == False:
5             dfs(next_vertex)
6
7     source_vertex = A
8     dfs(source_vertex)
```

- Can you think of another way to implement DFS?
- Recursion!
  - Best way to implement any form of traversal
    - Just like when you implemented tree/ trie traversal
  - Make sense because we are going depth-first like how recursion does it!

```
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Questions?



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  - Finding all connected components
  - Finding cycles
  - Shortest path (brute force)
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  - Topological sort (later on)
  - ... and many more!

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  - Shortest path (brute force)
  - Shortest path (non-brute force) on unweighted graph
  - Topological sort (later on)
  - ... and many more!
- We will see more in unit notes and tutorials

Questions?

Break!

- Classical problem

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  - Given a set of locations
  - Given routes between locations
  - Given the distance between locations

- Classical problem
  - Given a set of locations
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  - Given routes between locations
  - Given the distance between locations
  - Can you find the shortest distance from a source to a destination?
  - What is the path?

- Classical problem
  - Given a set of locations as **vertices  $V$**
  - Given routes between locations as **edges  $E$**
  - Given the distance between locations as **weights  $W$**
  - Can you find the shortest distance from a source to a destination?
  - What is the path?

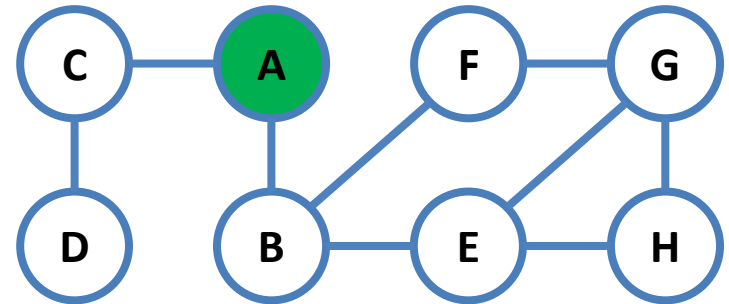
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  - Can you find the shortest distance from a source to a destination?
  - What is the path? This would require **backtracking**
- If the graph is **unweighted**?
  - Use **BFS** from the **source**!

- Classical problem
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  - Can you find the shortest distance from a source to a destination?
  - What is the path? This would require **backtracking**
  
- If the graph is unweighted?
  - Use BFS from the source!
  - Look back our BFS example
    - tree going down level by level
    - so give shortest distance

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it with a distance 0
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    - Serve from discovered, to visited
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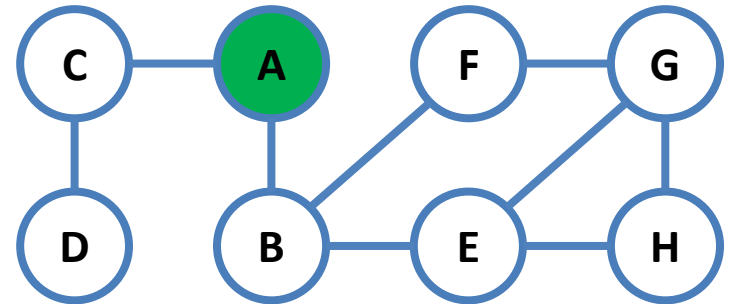


Discovered	A,0								
Visited									

# Graph

## Shortest distance with BFS

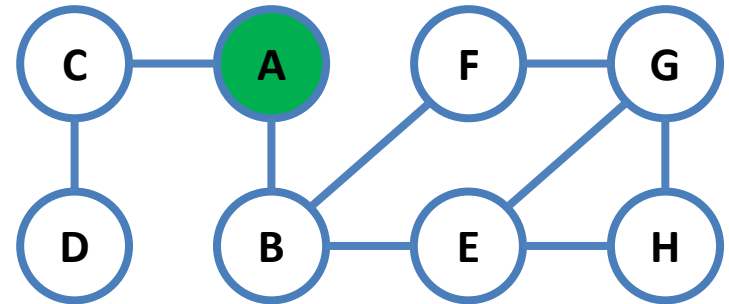
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Discovered									
Visited	A,0								

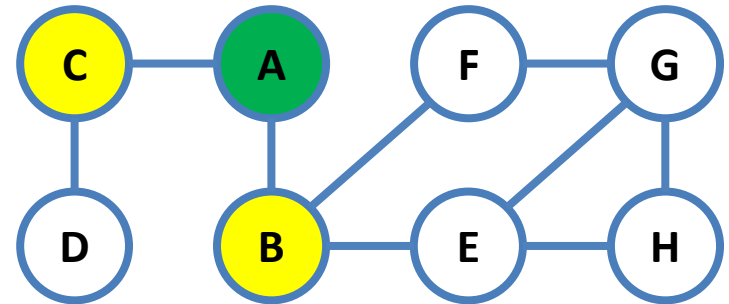


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      - $v.\text{distance} = u.\text{distance} + 1$



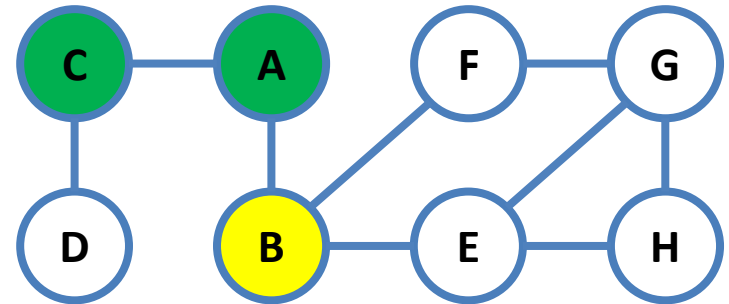
Discovered									
Visited	A,0								

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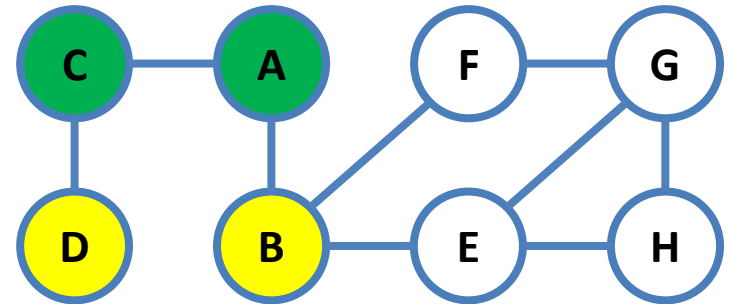
Discovered	C,1	B,1							
Visited	A,0								

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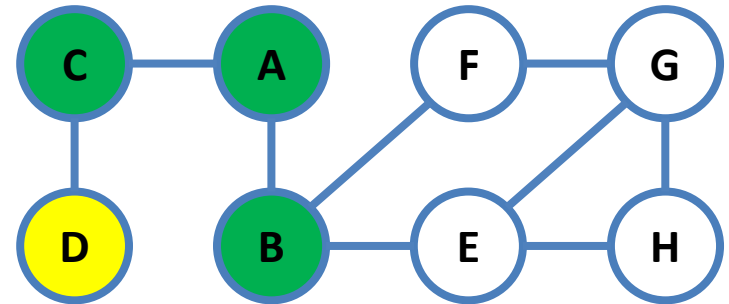
Discovered		B,1							
Visited	A,0	C,1							

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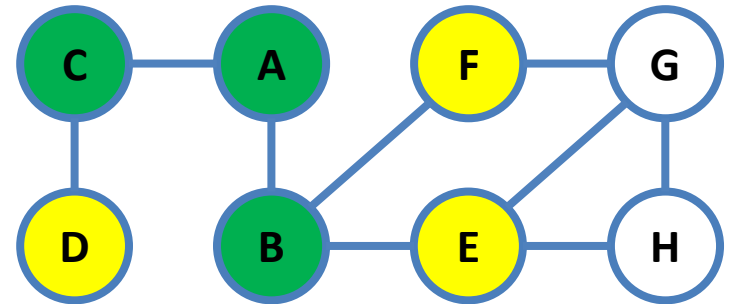
Discovered		B,1	D,2						
Visited	A,0	C,1							

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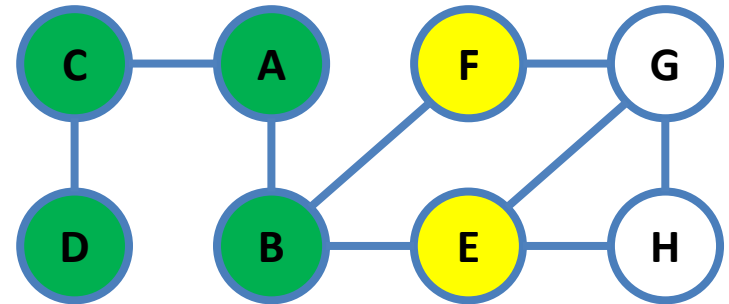
Discovered			D,2						
Visited	A,0	C,1	B,1						

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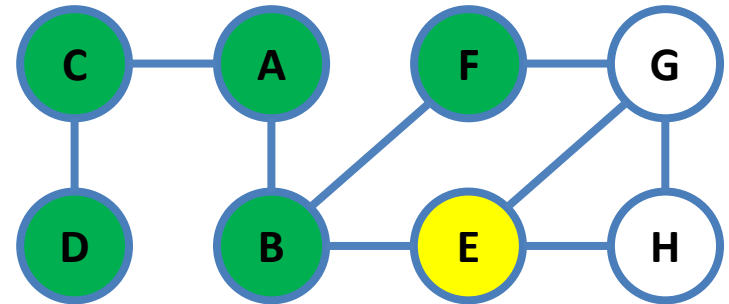
Discovered			D,2	F,2	E,2					
Visited	A,0	C,1	B,1							

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Discovered				F,2	E,2					
Visited	A,0	C,1	B,1	D,2						

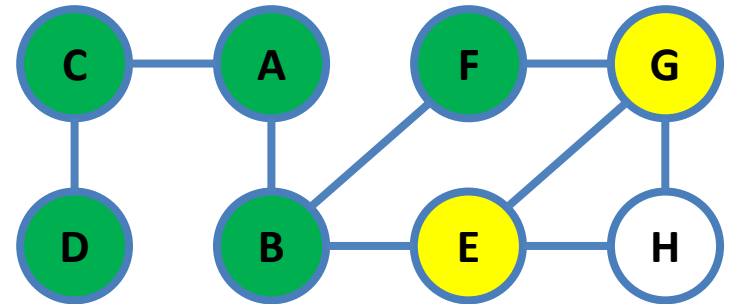
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Discovered					E,2					
Visited	A,0	C,1	B,1	D,2	F,2					

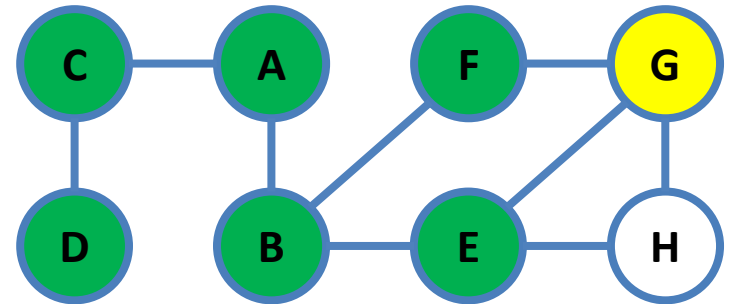


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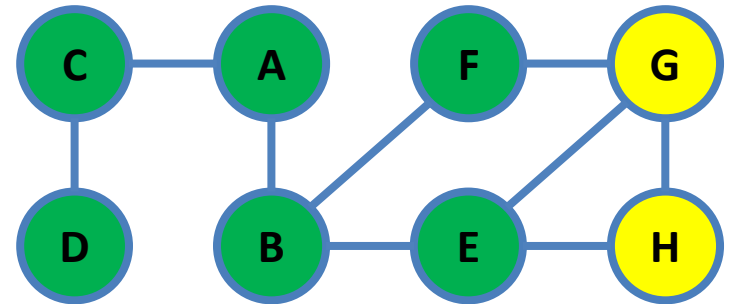
Discovered					E,2	G,3				
Visited	A,0	C,1	B,1	D,2	F,2					

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it with a distance 0
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue
      - $v.\text{distance} = u.\text{distance} + 1$



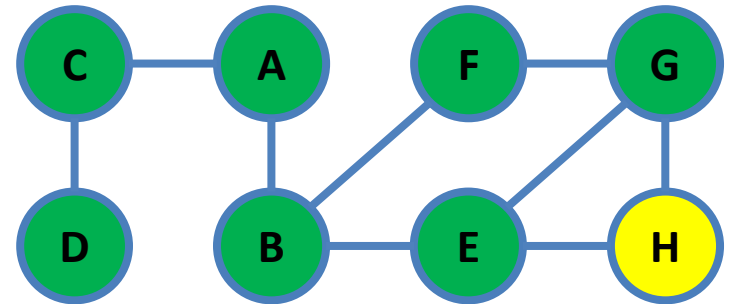
Discovered						G,3				
Visited	A,0	C,1	B,1	D,2	F,2	E,2				

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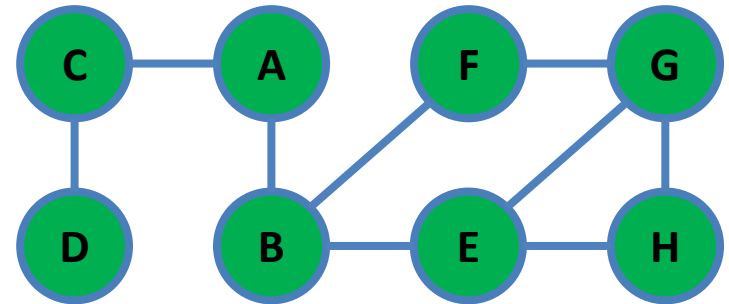
Discovered						G,3	H,3			
Visited	A,0	C,1	B,1	D,2	F,2	E,2				

- How would you implement it?
  - Let say we begin from vertex A
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Discovered							H,3			
Visited	A,0	C,1	B,1	D,2	F,2	E,2	G,3			

- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it with a distance 0
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue
      - $v.\text{distance} = u.\text{distance} + 1$
      - if  $v = E$  ( $v$  = target destination)  
break (to stop early)



Discovered									
Visited	A,0	C,1	B,1	D,2	F,2	E,2	G,3	H,3	

Questions?

- How would you modify the following to find the path?
- How would you implement it?
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it with a distance 0
  - While discovered is not empty
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  - Let say we begin from vertex A
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    - Put source (A) into it with a distance 0
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where u is the served
      - If vertex v is not discovered or visited, add to discovered queue
      - $v.\text{distance} = u.\text{distance} + 1$
      - $v.\text{previous} = u$  # enable backtracking



Questions?

# Graph

## Shortest path with Dijkstra

distance calculated from source

- What if the graph is weighted?

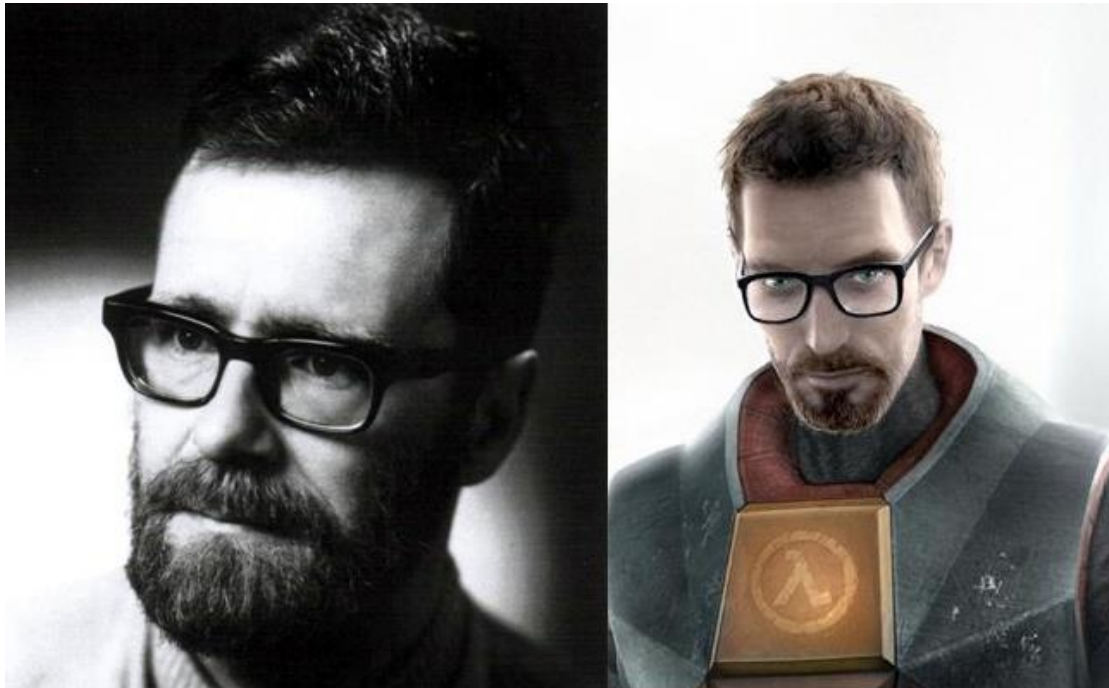
- What if the graph is weighted?
  - BFS is not able to do it anymore

- What if the graph is weighted?
  - BFS is not able to do it anymore since need Vertex object to store additional info self.weight
  - Sooo, we call in Dijkstra

# Graph

## Shortest path with Dijkstra

- What if the graph is weighted?
  - BFS is not able to do it anymore
  - Sooo, we call in Dijkstra (the left one)



# Graph

## Shortest path with Dijkstra

- What if the graph is weighted?
  - BFS is not able to do it anymore
  - Sooo, we call in Dijkstra (the left one)
- So Dijkstra came up with the shortest distance algorithm
  - Recall we can backtrack (previous) to get the path

Bae: Come over

Dijkstra: But there are so many routes to take and  
I don't know which one's the fastest

Bae: My parents aren't home

Dijkstra:

### Dijkstra's algorithm

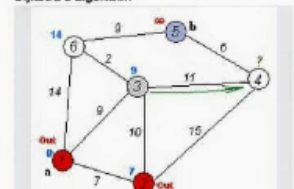
Graph search algorithm

*Not to be confused with Dykstra's projection algorithm.*

**Dijkstra's algorithm** is an algorithm for finding the shortest paths between nodes in a graph, which may represent, for example, road networks. It was conceived by computer scientist Edsger W. Dijkstra in 1956 and published three years later.<sup>[1][2]</sup>

The algorithm exists in many variants; Dijkstra's original variant found the shortest path between two nodes,<sup>[2]</sup> but a more common variant fixes a single node as the "source" node and finds shortest paths from the source to all other nodes in the graph, producing a **shortest-path tree**.

Dijkstra's algorithm



How Dijkstra came up with his algorithm

# Graph

## Shortest path with Dijkstra



most popular algorithms



All

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About 10,100,000 results (0.72 seconds)

Here I've put together a little list, in no particular order.

- Merge Sort, Quick Sort and Heap Sort. ...
- Fourier Transform and Fast Fourier Transform. ...
- Dijkstra's algorithm. ...
- RSA algorithm. ...
- Secure Hash Algorithm. ...
- Integer factorization. ...
- Link Analysis. ...
- Proportional Integral Derivative Algorithm.

More items...

The real 10 algorithms that dominate our world – Marcos Otero - Medium  
<https://medium.com/@.../the-real-10-algorithms-that-dominate-our-world-e95fa9f16c04>

Name	Best	Average	Worst	Memory	Stable
Bubble sort	$O(n^2)$	$O(n^2)$	$O(n^2)$	$O(1)$	Yes
Selection sort	$O(n^2)$	$O(n^2)$	$O(n^2)$	$O(1)$	No
Insertion sort	$O(n^2)$	$O(n^2)$	$O(n^2)$	$O(1)$	Yes
Merge sort	$O(n \log n)$	$O(n \log n)$	$O(n \log n)$	worst case is $O(n^2)$	Yes
Heap Sort	$O(n \log n)$	$O(n \log n)$	$O(n \log n)$	$O(1)$	Yes
Quick Sort	$O(n \log n)$	$O(n \log n)$	$O(n^2)$	$O(n)$	Yes
Radix Sort	$O(n \log n)$	$O(n \log n)$	$O(n \log n)$	$O(n)$	Yes

Bae: Come over

Dijkstra: But there are so many routes to take and  
I don't know which one's the fastest

Bae: My parents aren't home

Dijkstra:

## Dijkstra's algorithm



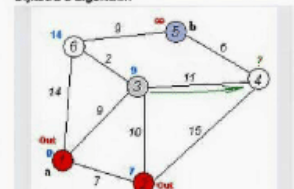
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Dijkstra's algorithm



## How Dijkstra came up with his algorithm

# Graph

## Shortest path with Dijkstra

- It is a combination of 2 algorithms



- It is a combination of 2 algorithms
  - Dynamic programming
  - Greedy

change queue in BFS into Heap (re-arrange itself with priority)  
to become Dijkstra

- It is a combination of 2 algorithms
  - Dynamic programming

The minimum distance from A to C can be the minimum of A to B (which we know) and minimum of B to C (which we know as well).
  - Greedy

- It is a combination of 2 algorithms

- Dynamic programming

The minimum distance from A to C can be the minimum of A to B (which we know) and minimum of B to C (which we know as well).

- Greedy

If I am at A, I can reach B and C. B is the closest, so I go to B and this is the shortest from A to B. I do not need to check if A to C then C to B (A->C->B) is the shortest anymore.

- It is a combination of 2 algorithms

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but GREED IS NOT GOOD... when will this fail?

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- Thus, Dijkstra doesn't work for negative edges

- It is a combination of 2 algorithms

- Dynamic programming

The minimum distance from A to C can be the minimum of A to B (which we know) and minimum of B to C (which we know as well).

- Greedy

If I am at A, I can reach B and C. B is the closest, so I go to B and this is the shortest from A to B. I do not need to check if A to C then C to B

(A→C→B) is the shortest anymore. since no checking, don't know the C or B combine with A to C can have shorter distance than A to B directly

but GREED IS NOT GOOD... when will this fail? When C to B is negative!

- Thus, Dijkstra doesn't work for negative edges

Note: might work at times when the negative edge isn't part of a cycle

Questions?



# Graph

## Shortest path with Dijkstra

- So how does Dijkstra work?

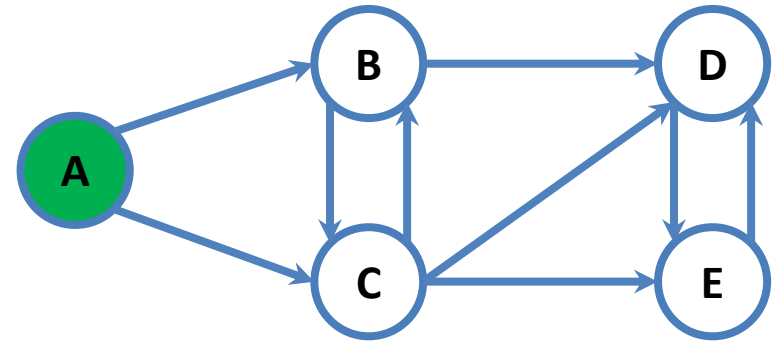
# Graph

## Shortest path with Dijkstra

- So how does Dijkstra work?

- Consider the following directed graph

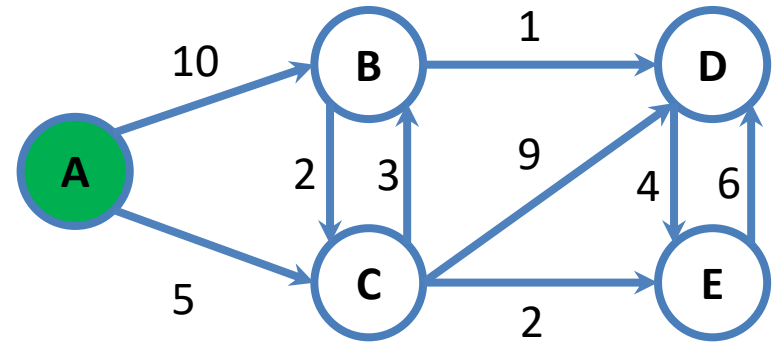
distance of visited vertex can not be changed



# Graph

## Shortest path with Dijkstra

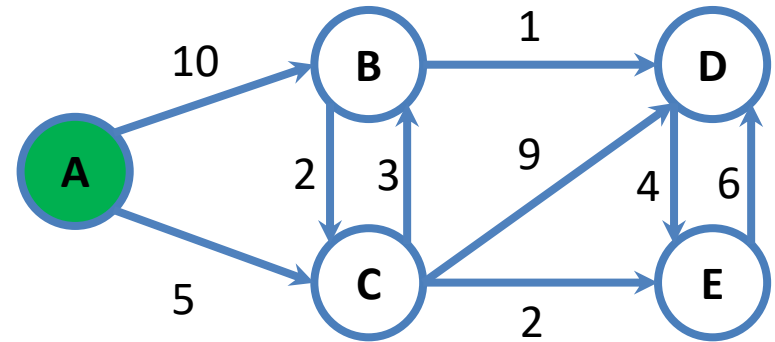
- So how does Dijkstra work?
  - Consider the following directed graph
    - Graph is weighted



# Graph

## Shortest path with Dijkstra

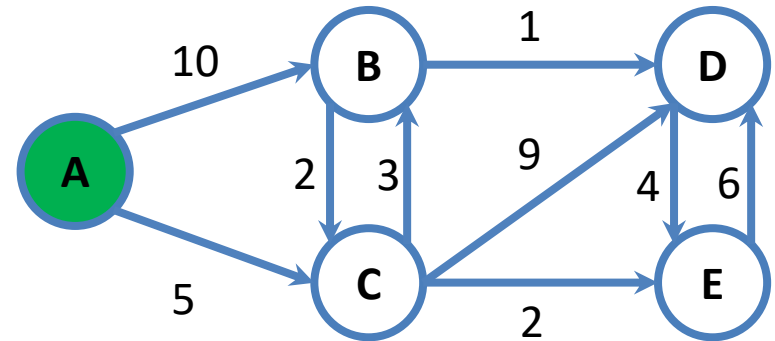
- So how does Dijkstra work?
  - Consider the following directed graph
    - Graph is weighted
  - So let us begin the algorithm...



# Graph

## Shortest path with Dijkstra

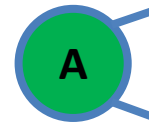
- So how does Dijkstra work?
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  - So let us begin the algorithm...
  - We are at A (source), and



# Graph

## Shortest path with Dijkstra

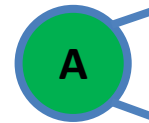
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**FOG OF  
WAR**

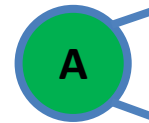


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    - Graph is weighted
  - So let us begin the algorithm...
  - So what happen is we will slowly wander to the closest point (from A)



**FOG OF  
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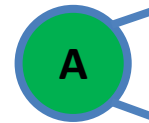
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- So how does Dijkstra work?
  - Consider the following directed graph
    - Graph is weighted
  - So let us begin the algorithm...
  - So what happen is we will slowly wander to the closest point (from A)
    - $A = 0$
    - $B = \text{infinity}$
    - $C = \text{infinity}$
    - $D = \text{infinity}$
    - $E = \text{infinity}$

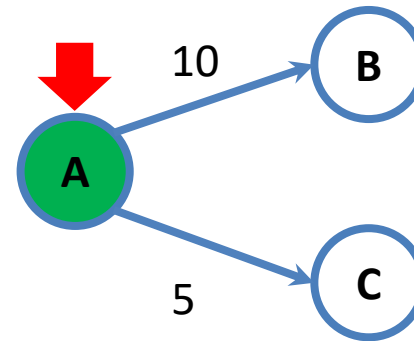


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- So how does Dijkstra work?

- Consider the following directed graph

- Graph is weighted



- So let us begin the algorithm...

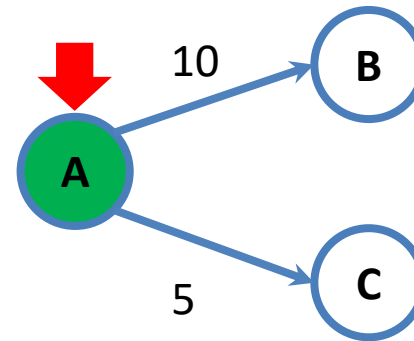
- So what happen is we will slowly wander to the closest point (from A)

- $A = 0$ , from here, we can see B and C (edges from A)
  - $B = \text{infinity}$
  - $C = \text{infinity}$
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- So how does Dijkstra work?

- Consider the following directed graph

- Graph is weighted

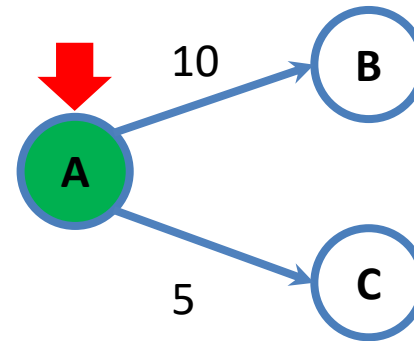


- So let us begin the algorithm...
- So what happen is we will slowly wander to the closest point (from A)
  - $A = 0$ , from here, we can see B and C (edges from A). Update distance
  - $B = 10$
  - $C = 5$
  - $D = \text{infinity}$
  - $E = \text{infinity}$

- So how does Dijkstra work?

- Consider the following directed graph

- Graph is weighted



- So let us begin the algorithm...

- So what happens if we will slowly wander to the closest point (from A)

- A = 0, from here, we can see B and C (edges from A). Update distance

- B = 10

- C = 5

- D = infinity

- E = infinity

- Closest is C, so we move to C

- So how does Dijkstra work?

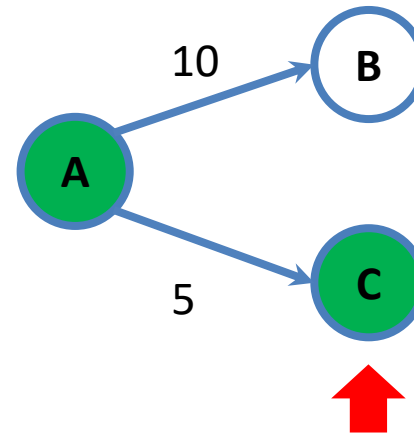
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- Graph is weighted

- So let us begin the algorithm...

- So what happens if we will slowly wander to the closest point (from A)

- $A = 0$ , from here, we can see B and C (edges from A). Update distance
  - $B = 10$
  - $C = 5$
  - $D = \text{infinity}$
  - $E = \text{infinity}$
  - Closest is C, so we move to C



- So how does Dijkstra work?

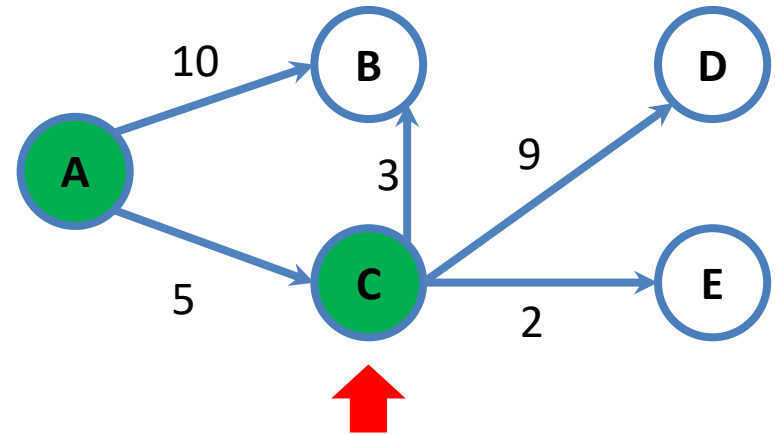
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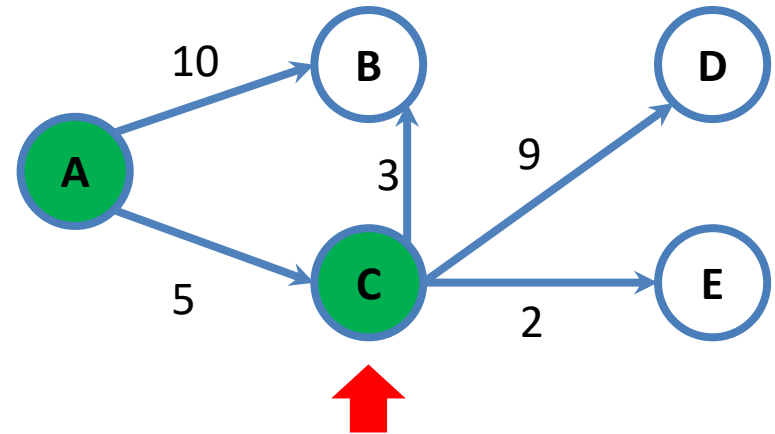
- So let us begin the algorithm...

- So what happens if we will slowly wander to the closest point (from A)

- $A = 0$ , from here, we can see B and C (edges from A). Update distance
- $B = 10$
- $C = 5$ , from here, we can see B, D and E
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- $E = \text{infinity}$



- So how does Dijkstra work?
  - Consider the following directed graph
    - Graph is weighted
  - So let us begin the algorithm...
  - So what happens we will slowly wander to the closest point (from A)
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    - $D = \text{infinity}$
    - $E = \text{infinity}$



- So how does Dijkstra work?

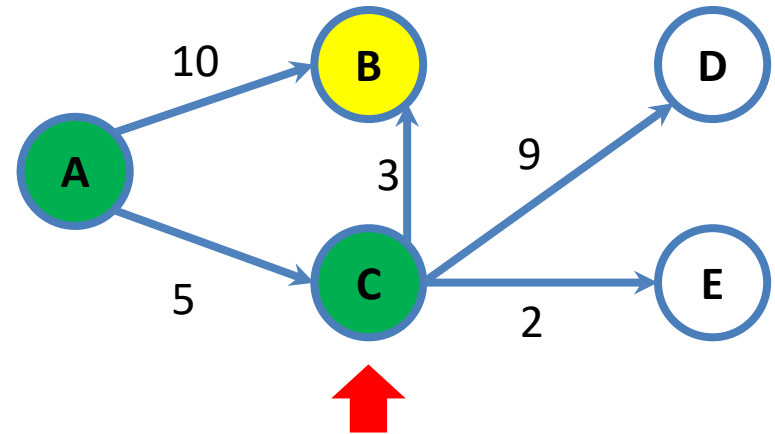
- Consider the following directed graph

- Graph is weighted

- So let us begin the algorithm...

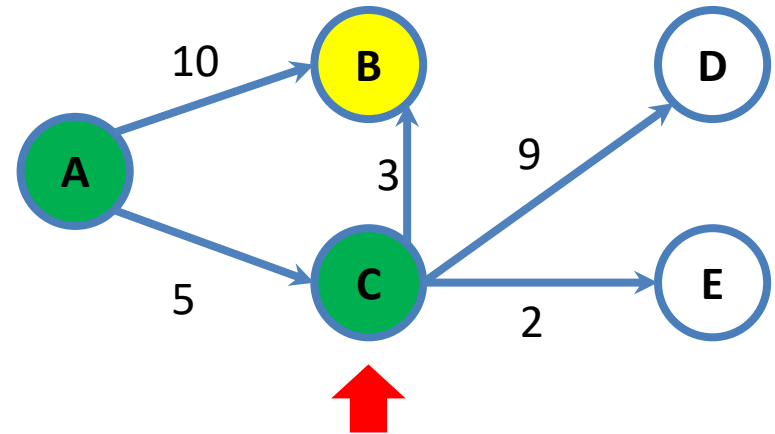
- So what happens if we will slowly wander to the closest point (from A)

- $A = 0$ , from here, we can see B and C (edges from A). Update distance
  - $B = 10$  (A→B) vs 8 (A→C→B)
  - $C = 5$ , from here, we can see B, D and E. Update the distance
  - $D = \text{infinity}$
  - $E = \text{infinity}$





- So how does Dijkstra work?
  - Consider the following directed graph
    - Graph is weighted
  - So let us begin the algorithm...
  - So what happens we will slowly wander to the closest point (from A)
    - $A = 0$ , from here, we can see B and C (edges from A). Update distance
    - $B = 8$
    - $C = 5$ , from here, we can see B, D and E. Update the distance
    - $D = \text{infinity}$
    - $E = \text{infinity}$



- So how does Dijkstra work?

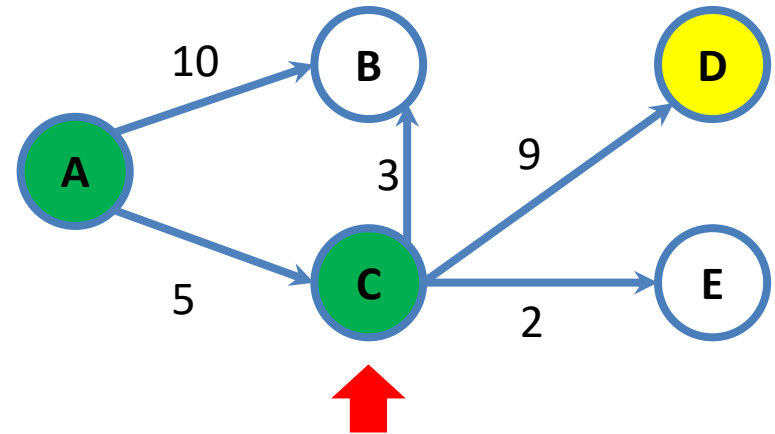
- Consider the following directed graph

- Graph is weighted

- So let us begin the algorithm...

- So what happens if we will slowly wander to the closest point (from A)

- $A = 0$ , from here, we can see B and C (edges from A). Update distance
  - $B = 8$
  - $C = 5$ , from here, we can see B, D and E. Update the distance
  - $D = 9$ ?
  - $E = \text{infinity}$



- So how does Dijkstra work?

- Consider the following directed graph

- Graph is weighted

- So let us begin the algorithm...

- So what happens if we will slowly wander to the closest point (from A)

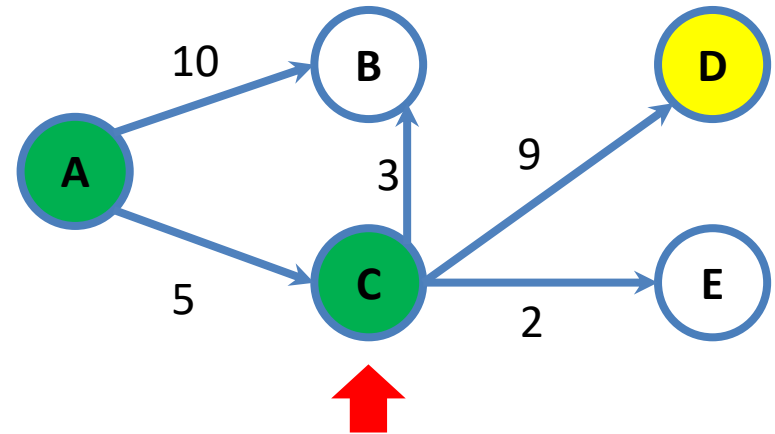
- A = 0, from here, we can see B and C (edges from A). Update distance

- B = 8

- C = 5, from here, we can see B, D and E. Update the distance

- D = 14 because distance is from A

- E = infinity comparing with infinity



- So how does Dijkstra work?

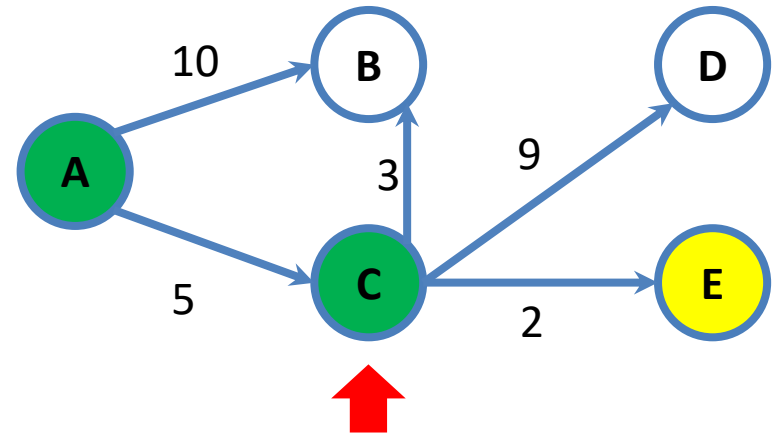
- Consider the following directed graph

- Graph is weighted

- So let us begin the algorithm...

- So what happens we will slowly wander to the closest point (from A)

- A = 0, from here, we can see B and C (edges from A). Update distance
- B = 8
- C = 5, from here, we can see B, D and E. Update the distance
- D = 14
- E = 7



- So how does Dijkstra work?

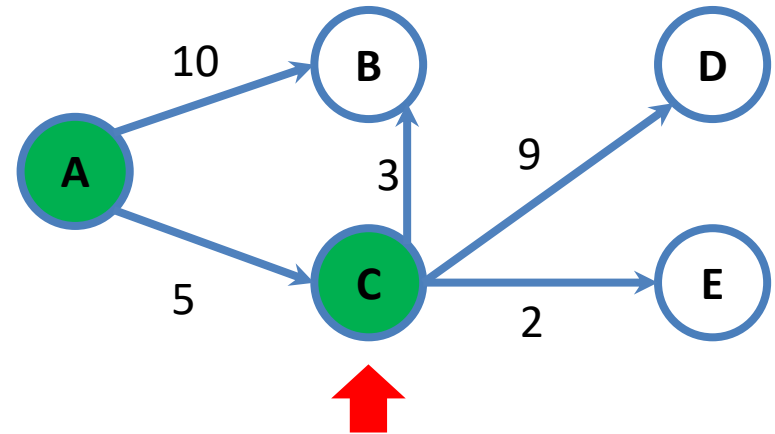
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- A = 0, from here, we can see B and C (edges from A). Update distance
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- D = 14
- E = 7
- Closest is E, so we go E



- So how does Dijkstra work?

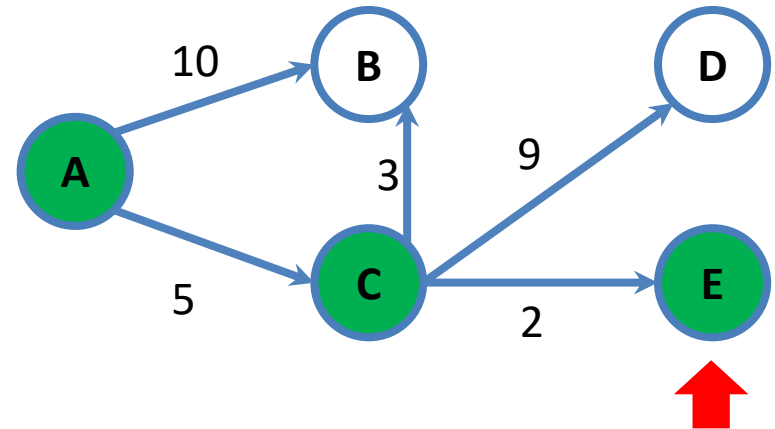
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- So how does Dijkstra work?

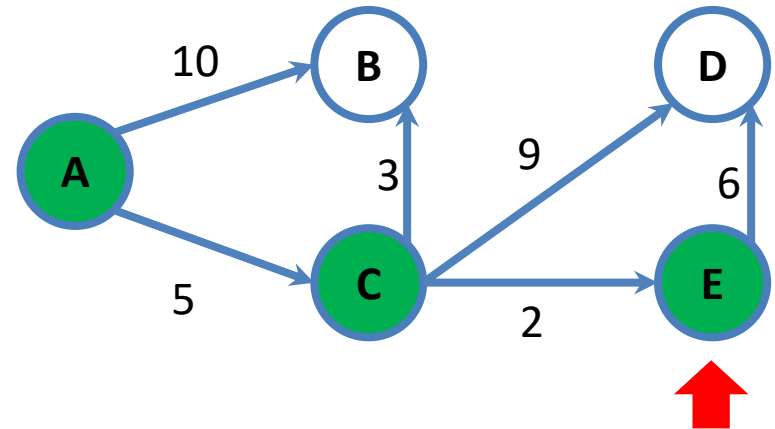
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- $A = 0$ , from here, we can see B and C (edges from A). Update distance
- $B = 8$
- $C = 5$ , from here, we can see B, D and E. Update the distance
- $D = 14$
- $E = 7$ , from here, we can see D.



- So how does Dijkstra work?

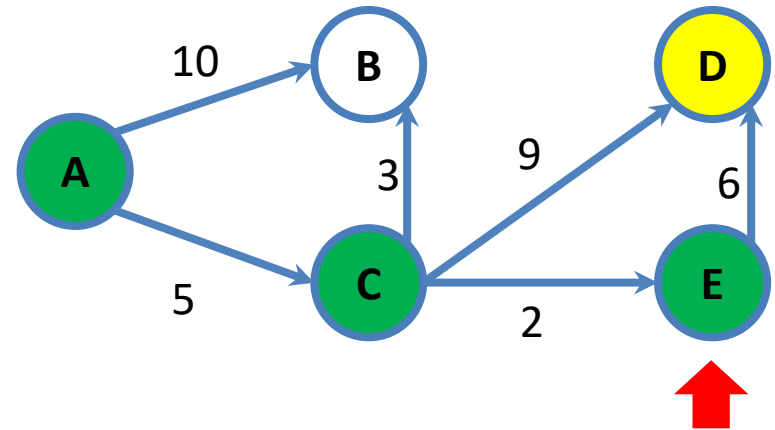
- Consider the following directed graph

- Graph is weighted

- So let us begin the algorithm...

- So what happens if we will slowly wander to the closest point (from A)

- $A = 0$ , from here, we can see B and C (edges from A). Update distance
- $B = 8$
- $C = 5$ , from here, we can see B, D and E. Update the distance
- $D = 14$  vs  $7+6$  (A→E→D)
- $E = 7$ , from here, we can see D. Update the distance

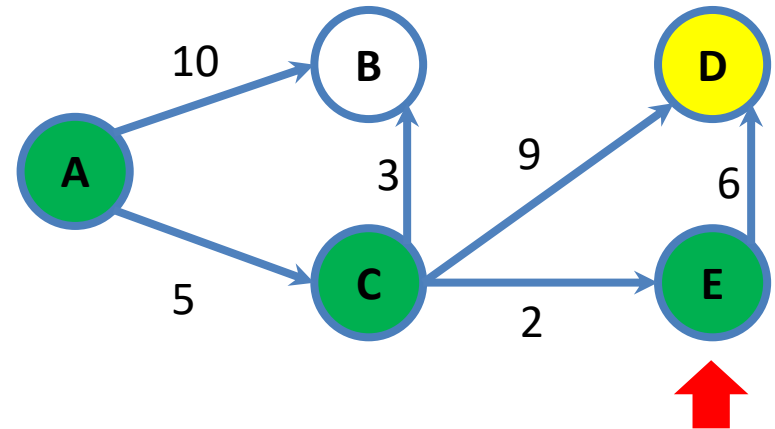




- So how does Dijkstra work?

- Consider the following directed graph

- Graph is weighted



- So let us begin the algorithm...

- So what happens if we will slowly wander to the closest point (from A)

- A = 0, from here, we can see B and C (edges from A). Update distance
- B = 8
- C = 5, from here, we can see B, D and E. Update the distance
- D = 13
- E = 7, from here, we can see D. Update the distance

- So how does Dijkstra work?

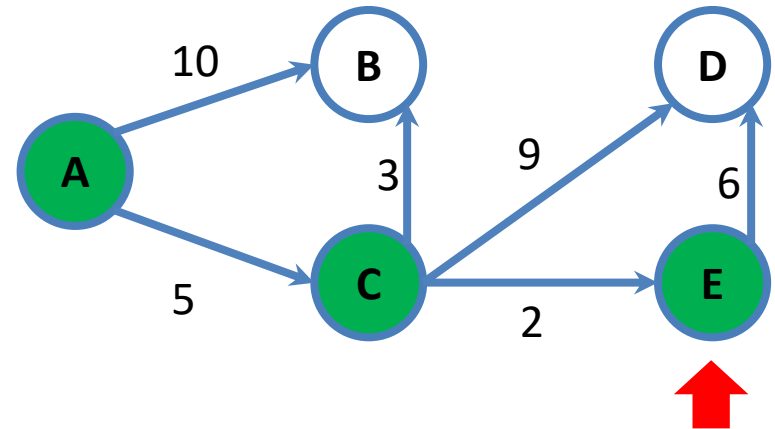
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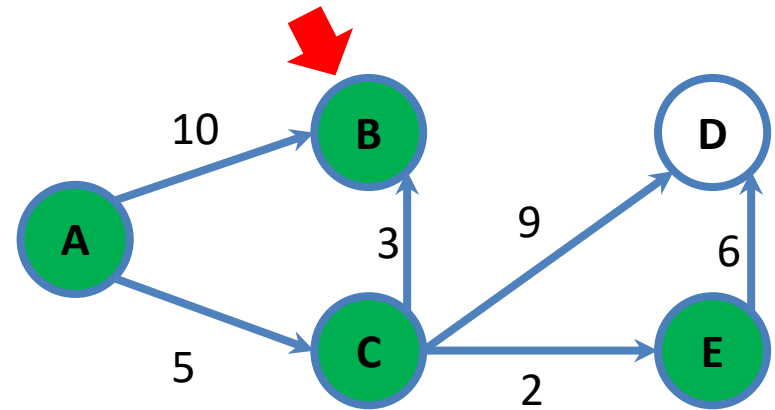
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- D = 13
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- Closest is B, so we go B



- So how does Dijkstra work?

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- Graph is weighted

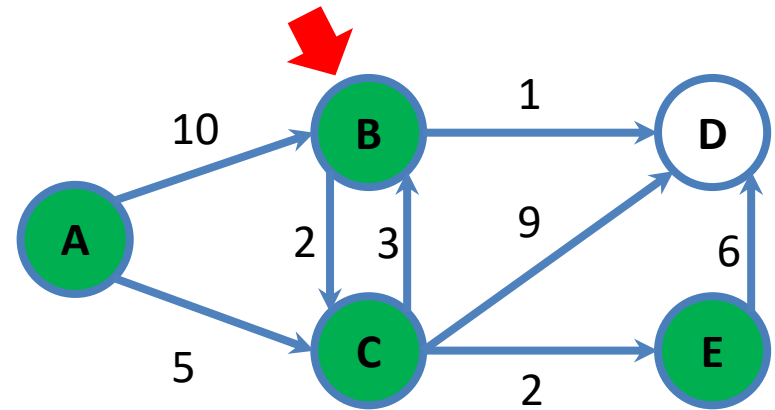


- So let us begin the algorithm...
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  - $C = 5$ , from here, we can see B, D and E. Update the distance
  - $D = 13$
  - $E = 7$ , from here, we can see D. Update the distance
  - Closest is B, so we go B

- So how does Dijkstra work?

- Consider the following directed graph

- Graph is weighted

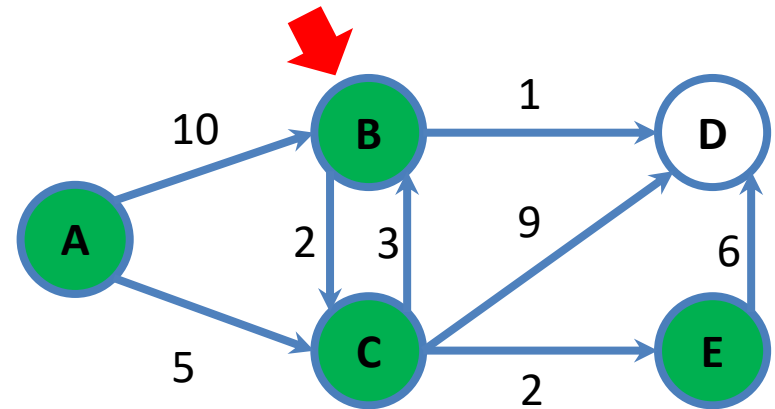


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  - C = 5, from here, we can see B, D and E. Update the distance
  - D = 13
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- So how does Dijkstra work?

- Consider the following directed graph

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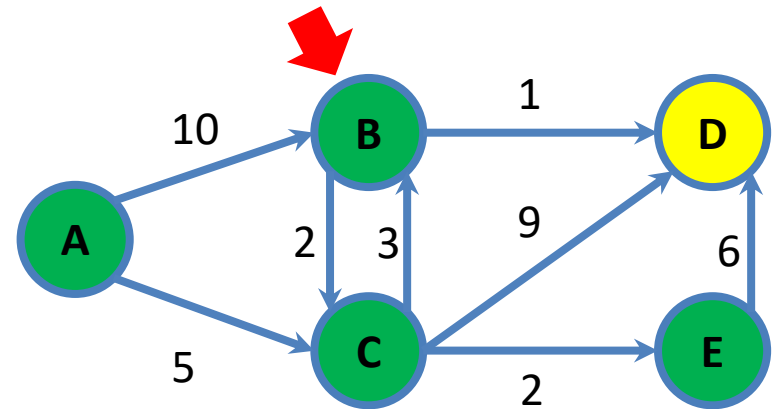


- So let us begin the algorithm...
- So what happens if we slowly wander to the closest point (from A)
  - A = 0, from here, we can see B and C (edges from A). Update distance
  - B = 8, from here, we can see C and D. Update distance for C?
  - C = 5, from here, we can see B, D and E. Update the distance
  - D = 13
  - E = 7, from here, we can see D. Update the distance

- So how does Dijkstra work?

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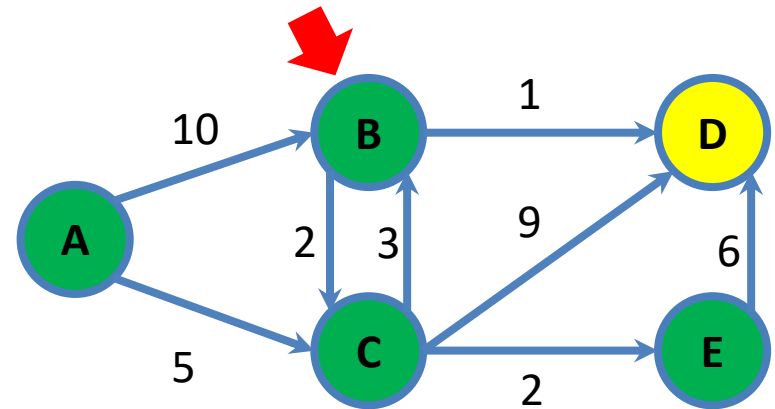


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  - D = 13
  - E = 7, from here, we can see D. Update the distance

- So how does Dijkstra work?

- Consider the following directed graph

- Graph is weighted



- So let us begin the algorithm...
- So what happens if we will slowly wander to the closest point (from A)
  - A = 0, from here, we can see B and C (edges from A). Update distance
  - B = 8, from here, we can see C and D. Update distance for D.
  - C = 5, from here, we can see B, D and E. Update the distance
  - D = 9 (8+1 via A->B->D)
  - E = 7, from here, we can see D. Update the distance

update distance from source A to the discovered vertex

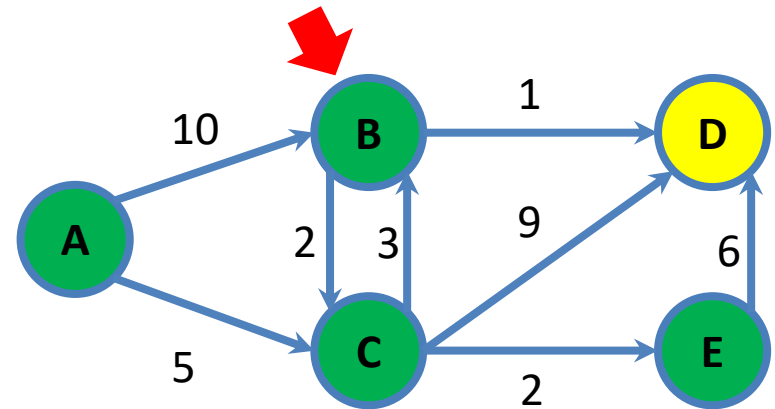
in A, compare distance to B and C, choose the smallest distance among the discovered vertices to visit

the visited vertex can not change its distance

- So how does Dijkstra work?

- Consider the following directed graph

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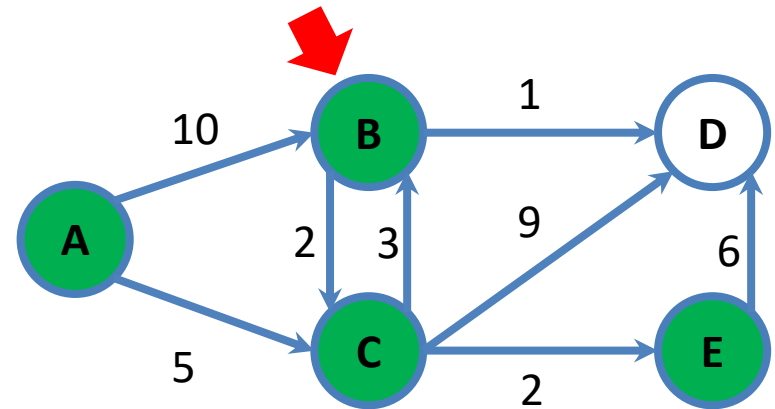
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  - $D = 9$
  - $E = 7$ , from here, we can see D. Update the distance



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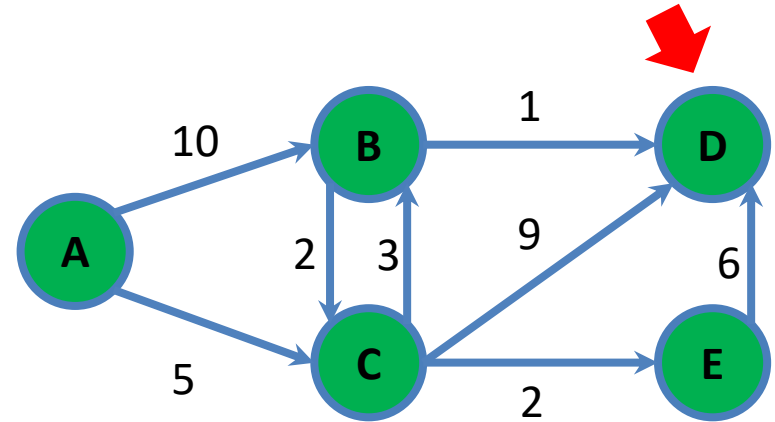


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  - C = 5, from here, we can see B, D and E. Update the distance
  - D = 9
  - E = 7, from here, we can see D. Update the distance
  - Closest is D, so we move there

- So how does Dijkstra work?

- Consider the following directed graph

- Graph is weighted

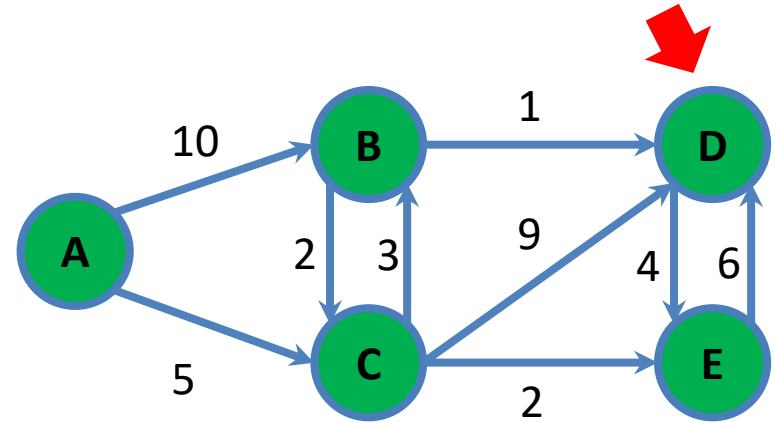


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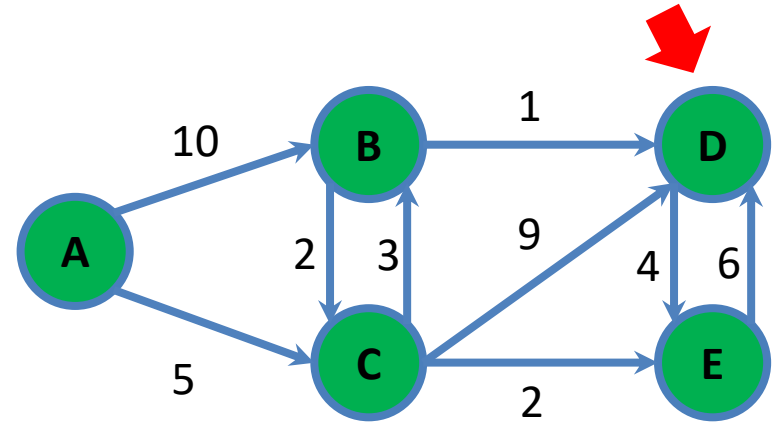


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  - C = 5, from here, we can see B, D and E. Update the distance
  - D = 9, from here, we can see E but E is already finalized
  - E = 7, from here, we can see D. Update the distance
  - Closest is D, so we move there

- So how does Dijkstra work?

- Consider the following directed graph

- Graph is weighted



- So let us begin the algorithm...
- So what happens if we will slowly wander to the closest point (from A)
  - A = 0, from here, we can see B and C (edges from A). Update distance
  - B = 8, from here, we can see C and D. Update distance for D.
  - C = 5, from here, we can see B, D and E. Update the distance
  - D = 9, from here, we can see E but E is already finalized
  - E = 7, from here, we can see D. Update the distance
  - And we are **done**!

Questions?

# Graph

## Shortest path with Dijkstra

- Algorithm?

- Algorithm? Very similar to the BFS except...

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  - Priority queue instead of a normal queue
    - Serve the closest vertex (not finalized)



- Algorithm? Very similar to the BFS except...
  - Priority queue instead of a normal queue
    - Serve the closest vertex (not finalized)
  - Update the distance if the neighbour vertex is visited but not finalized
    - To the shorter one

- This is the BFS algorithm, we change to Dijkstra now
  - Let say we begin from vertex A
  - Have a queue for discovered
    - Put source (A) into it with a distance 0
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue
      - $v.\text{distance} = u.\text{distance} + 1$

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- Try to modify this as part of the in-class activity

use MinHeap()

- This is the BFS algorithm, we change to Dijkstra now
    - Let say we begin from vertex A
    - Have a **priority queue** for discovered
      - Put source (A) into it with a distance 0
    - While discovered is not empty
      - Serve from discovered, to visited
      - For each edge  $\langle u, v, w \rangle$  where u is the served
        - If vertex v is not discovered or visited, add to discovered queue
          - » Set  $v.distance = u.distance + w$  v.previous = u, for the first path from the source to v has v.distance
        - If vertex v is discovered but not visited and  $v.distance > u.distance + w$ 
          - » Update  $v.distance = u.distance + w$  if new path found with lower distance from source to v, then v.previous = new u, to switch to second path that has lower v.distance
- edge realisation

discovered = MinHeap()

discovered.append(source.distance, source) (key, data)

discovered Heap always have lowest distance element at the beginning

- This is the BFS algorithm, we change to Dijkstra now
  - Let say we begin from vertex A
  - Have a **priority queue** for discovered
    - Put source (A) into it with a distance 0
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge **<u,v,w>** where u is the served
      - If vertex v is not discovered or visited, add to discovered queue
        - » Set **v.distance = u.distance + w**
      - If vertex v is discovered but not visited and **v.distance > u.distance + w**
        - » Update **v.distance = u.distance + w**
  - We use a min-heap for our priority queue!

- This is the BFS algorithm, we change to Dijkstra now
  - Let say we begin from vertex A
    - like Minheap, for discovered, to add in vertex with distance, Minheap would prioritise the one with lowest distance
  - Have a **priority queue** for discovered
    - Put source (A) into it with a distance 0
  - While discovered is not empty
    - Serve from discovered, to visited
    - For each edge  $\langle u, v, w \rangle$  where  $u$  is the served
      - If vertex  $v$  is not discovered or visited, add to discovered queue
        - » Set  $v.\text{distance} = u.\text{distance} + w$
      - If vertex  $v$  is discovered but not visited and  $v.\text{distance} > u.\text{distance} + w$ 
        - » Update  $v.\text{distance} = u.\text{distance} + w$
  - We use a min-heap for our priority queue!
    - Note that we need a pointer to the nodes to update distance in  $O(1)$

- Algorithm can be as follows (might differ):

```
1  discover_queue = MinHeap()
2  discover_queue.append([source,0])
3
4  while discover_queue is not empty:
5      u = discover_queue.serve()
6      u.visited = True
7      for each <u,v,w> in u.edges:
8          if v.visited = True:
9              pass
10         else:
11             if v.discovered = False:
12                 discover_queue.append([v, u.distance+w])
13                 v.discovered = True
14             else:
15                 if v.distance > u.distance+w:
16                     discover_queue.update(v, u.distance+w)
17                     v.discovered = True
```

Questions?



- Complexity?

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# Graph

## Shortest path with Dijkstra

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```

$O(V)$   
discovered-queue

# Graph

## Shortest path with Dijkstra

$O(V)$

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```

Serve:  $O(\log V)$

# Graph

## Shortest path with Dijkstra

$O(V)$

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16                    discover_queue.update(v, u.distance+w)
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```

if dense graph, maximum number of  
edges from each vertex  
=  $V$  (total number of vertices) - 1

upheap() method in  
MinHEAP

Serve:  $O(\log V)$

how many  $V$ , how many

edges of  $u$  only not entire graph

$O(V)$  bounded by  $V - 1$   
 $u$  can connect to any other vertices

# Graph

## Shortest path with Dijkstra

$O(V)$

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```

Serve:  $O(\log V)$

$O(V)$

Update:  $O(\log V)$

### Time Complexity?

$O(V)$  →

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```

MinHeap like Tree

Serve:  $O(\log V)$

$O(V)$

Update:  
 $O(\log V)$

search the vertex by key in tree

- Time Complexity?  $O(V^2 \log V)$

$O(V)$

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```

Serve:  $O(\log V)$

$O(V)$

Update:  $O(\log V)$



- Time Complexity?  $O(V^2 \log V) = O(E \log V)$

$O(V)$

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```

Serve:  $O(\log V)$

$O(V)$

Update:  $O(\log V)$

- Time Complexity?  $O(V^2 \log V) = O(E \log V)$

- Recall for dense graph,  $E \approx V^2$

$O(V)$

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```

downHeap  
tree, so  $\log(V)$

Serve:  $O(\log V)$

$O(V)$

Update:  
 $O(\log V)$

- Time Complexity?  $O(V^2 \log V) = O(\overset{V^2}{E} \log V)$ 
  - Recall for dense graph,  $E \approx V^2$
- Note that with Fibonacci heap instead of your binary heap, we can reduce the complexity further to  $O(E + V \log V)$

- Time Complexity?  $O(V^2 \log V) = O(E \log V)$ 
  - Recall for dense graph,  $E \approx V^2$
- Note that with Fibonacci heap instead of your binary heap, we can reduce the complexity further to  $O(E + V \log V) = O(V^2 + V \log V) = O(V^2)$ 
  - For dense graph

Questions?

- What if we have a single source
  - As usual
- But single target?

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- What if we have a single source
  - As usual
- But single target?
- We can terminate after we have move the target vertex to the visited portion!
  - We would have the shortest distance
  - We can backtrack for the shortest path
    - Via vertex.previous attribute

Questions?

# Graph

## Shortest path with Dijkstra

- Why does Dijkstra work?

- Why does Dijkstra work?
  - Let us use Nathan's slides

# Proof of Correctness

**Claim:** For every vertex  $v$  which has been removed from the queue,  $\text{dist}[v]$  is correct

- Notation:

- $V$  is the set of vertices
- $Q$  is the set of vertices in the queue
- $S = V / Q$  = the set of vertices who have been removed from the queue

## Base Case

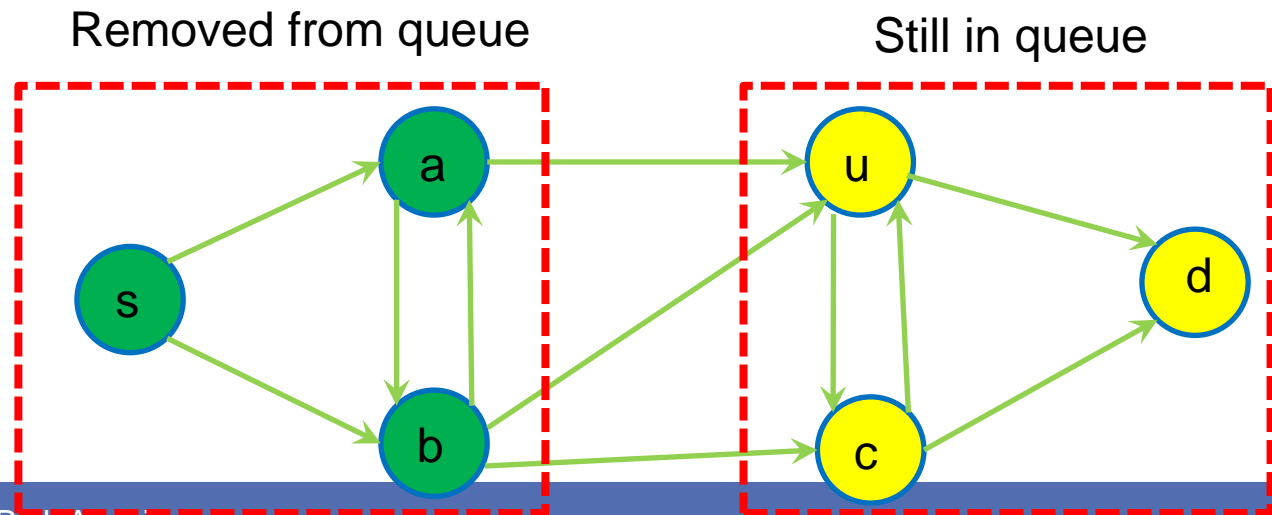
- $\text{Dist}[s]$  is initialised to 0, which is the shortest distance from  $s$  to  $s$  (since there are no negative weights)

# Proof of Correctness

**Claim:** For every vertex  $v$  which has been removed from the queue,  $\text{dist}[v]$  is correct

## Inductive Step:

- Assume that the claim holds for all vertices which have been removed from the queue (S)
- Let  $u$  be the next vertex which is removed from the queue
- We will show that  $\text{dist}[u]$  is correct



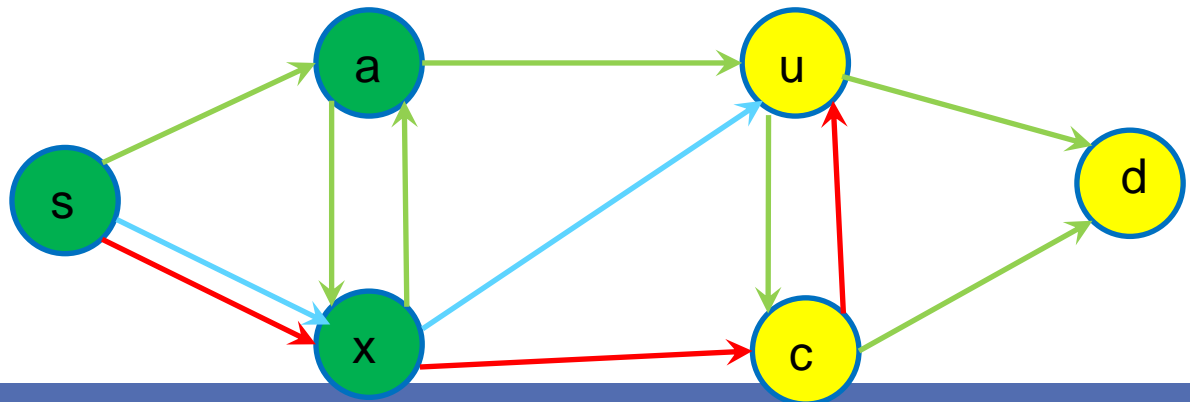
# Proof of Correctness

**Claim:** For every vertex  $v$  which has been removed from the queue,  $\text{dist}[v]$  is correct

## Inductive Step:

- Suppose (for contradiction) there is a shorter path  $P$ ,  $s \rightsquigarrow u$  with  $\text{len}(P) < \text{dist}[u]$
- Let  $x$  be the furthest vertex on  $P$  which is in  $S$  (i.e. has been finalised)
- By the inductive hypothesis,  $\text{dist}[x]$  is correct (since it is in  $S$ )

Current path  
Assumed  
shorter path ( $P$ )



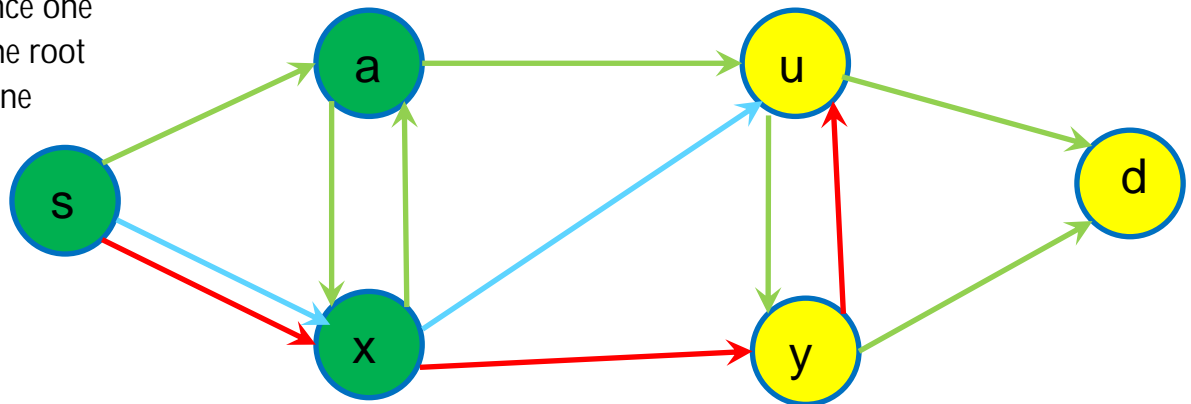
# Proof of Correctness

**Claim:** For every vertex  $v$  which has been removed from the queue,  $\text{dist}[v]$  is correct

## Inductive Step:

- By the inductive hypothesis,  $\text{dist}[x]$  is correct (since it is in  $S$ )
- Let  $y$  be the next vertex on  $P$  after  $x$
- $\text{len}(P) < \text{dist}[u]$  (by assumption)
- Edge weights are non-negative
- $\text{len}(s \rightsquigarrow y) \leq \text{len}(P) < \text{dist}[u]$

MinHeap() only serve the minimum distance one  
so, MinHeap() would serve the one on the root  
and compare its children roots and put one  
to be root, so  $O(\log(V))$



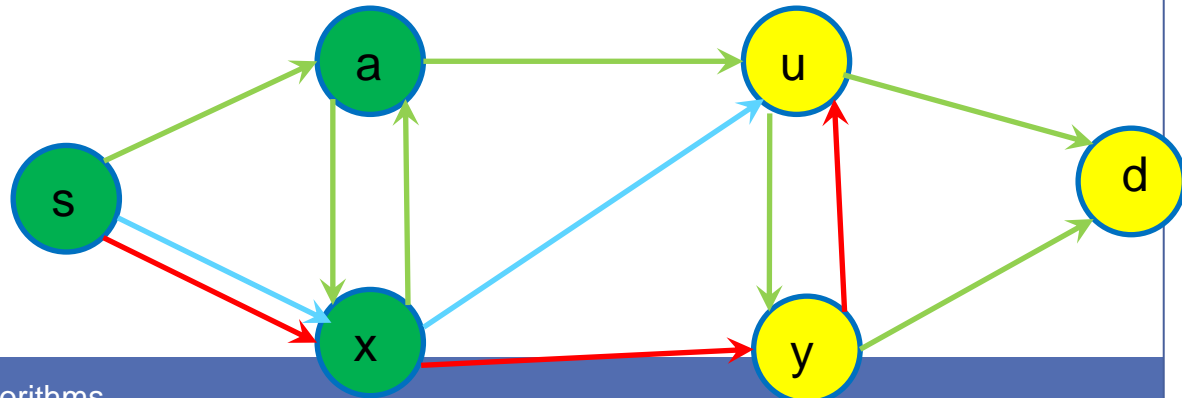


# Proof of Correctness

**Claim:** For every vertex  $v$  which has been removed from the queue,  $\text{dist}[v]$  is correct

## Inductive Step:

- $\text{Len}(s \rightsquigarrow y) \leq \text{len}(P) < \text{dist}[u]$
- Since we said that  $P$  (via  $x$  and  $y$ ) is a shortest path...
- $\text{dist}[y] = \text{len}(s \rightsquigarrow y) < \text{dist}[u]$
- So  $\text{dist}[y] < \text{dist}[u]$ ...
- If  $y \neq u$ , why didn't  $y$  get removed before  $u$ ???
- If  $y = u$ , how can  $\text{dist}[y] < \text{dist}[u]$ ???



- Why does Dijkstra work?
  - Let us use Nathan's slides
  - Or let me just explain it on the whiteboard...
  - Via proof by contradiction!

Questions?

- **Bellman-Ford**    not work for negative distances
- **Floyd-Warshall**    all pair

- Bellman-Ford
- Floyd-Warshall
  - With transitive closure

- Bellman-Ford
- Floyd-Warshall
  - With transitive closure
- We see it later in next lectures

- Bellman-Ford
  - Single source
  - Can know negative edges
- Floyd-Warshall
  - With transitive closure
  - Single or more sources
  - Can know negative edges
- We see it later in next lectures

Questions?



Thank You