



# RECITATION 2

# ANALYSIS OF ALGORITHMS

## II

2016 SPRING

# Outline

- Problem – 1 BFS Example
- Problem – 2 BFS and Shortest Path Problem
- Problem – 3 DFS Example
- Problem – 4 Bipartite Graph Example
- Problem – 5 Greedy Alg, *Coffee Shop Problem*

# Graph Traversal

- Application examples
  - Given a graph representation and a vertex  $s$  in the graph
  - Find all paths from  $s$  to the other vertices
- Two common graph traversal algorithms
  - **Breadth-First Search (BFS)**
  - **Depth-First Search (DFS)**

# Data Structures for DFS and BFS Implementation

- What is the main data structures employed when implementing DFS and BFS?
- ✓ BFS → QUEUE (FIFO Queue can be used)
  - We trace the graph layer by layer by considering all of the children of a node before starting to trace nodes further away
- ✓ DFS → STACK
  - Algorithm considers the immediate unexplored children before considering the other children of a node while moving from parent node to child node.
  - It goes deeper through the branches of the graph, before tracing other branches.

# BFS Algorithm using Queue

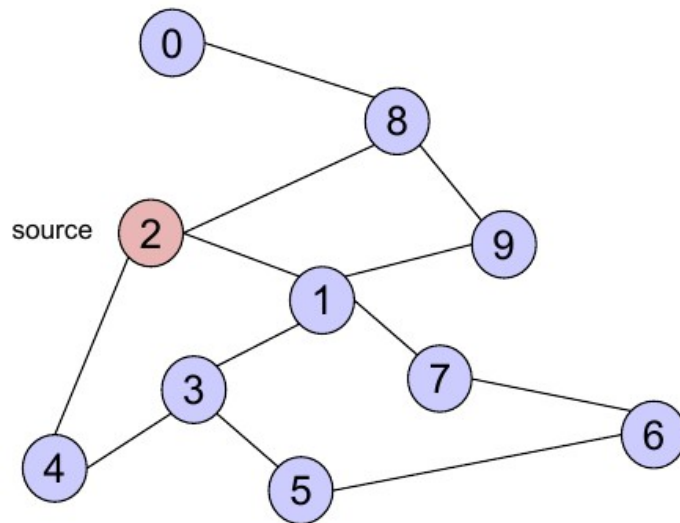
**Algorithm**  $BFS(s)$

**Input:**  $s$  is the source vertex

**Output:** Mark all vertices that can be visited from  $s$ .

1. **for** each vertex  $v$
2.     **do**  $flag[v] := \text{false}$ ;   // **flag[ ]: visited or not**
3.      $Q = \text{empty queue}$ ;       **Why use queue? Need FIFO**
4.      $flag[s] := \text{true}$ ;
5.      $enqueue(Q, s)$ ;
6.     **while**  $Q$  is not empty
7.         **do**  $v := dequeue(Q)$ ;
8.         **for** each  $w$  adjacent to  $v$
9.             **do if**  $flag[w] = \text{false}$
10.                 **then**  $flag[w] := \text{true}$ ;
11.                  $enqueue(Q, w)$

# Problem 1 - BFS Example



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

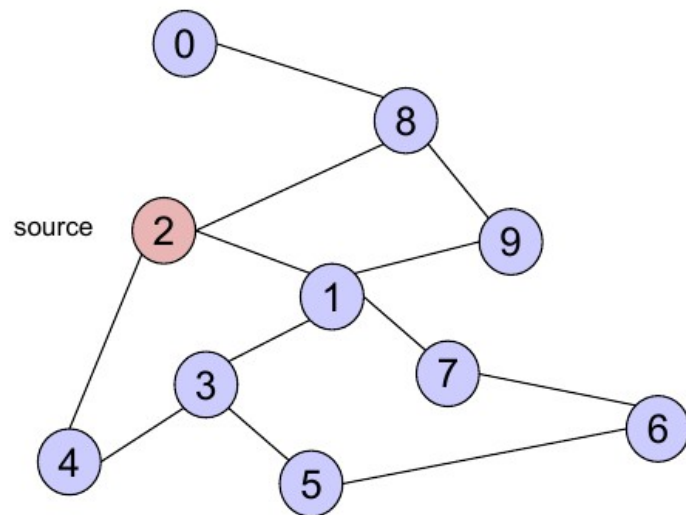
0	F
1	F
2	F
3	F
4	F
5	F
6	F
7	F
8	F
9	F

Initialize "visited"  
table (all False)

$Q = \{ \}$

Initialize **Q** to be empty

# BFS Example



$Q = \{ 2 \}$

Place source 2 on the queue

Adjacency List

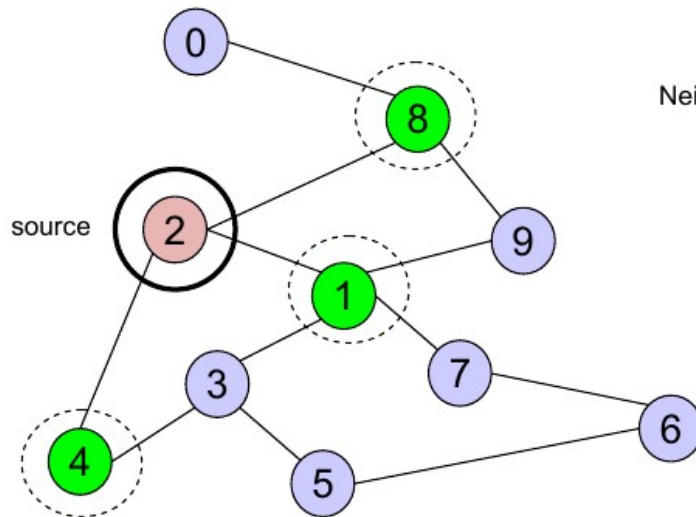
0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	F
1	F
2	T
3	F
4	F
5	F
6	F
7	F
8	F
9	F

Flag that 2 has been visited

# BFS Example



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Neighbors →

Visited Table (T/F)

0	F
1	T
2	T
3	F
4	T
5	F
6	F
7	F
8	T
9	F

Mark neighbors  
as visited 1, 4, 8

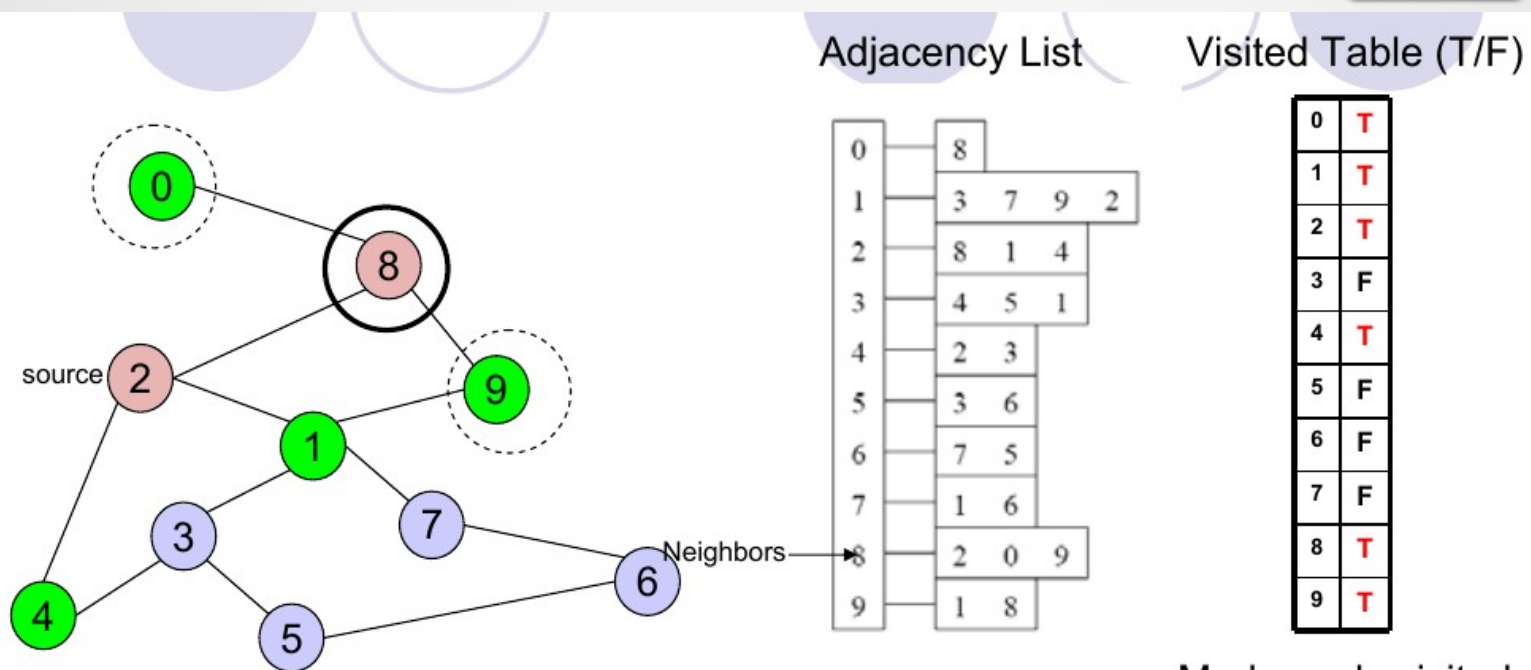
$Q = \{2\} \rightarrow \{8, 1, 4\}$

Dequeue 2.

Place all unvisited neighbors of 2 on the queue



# BFS Example



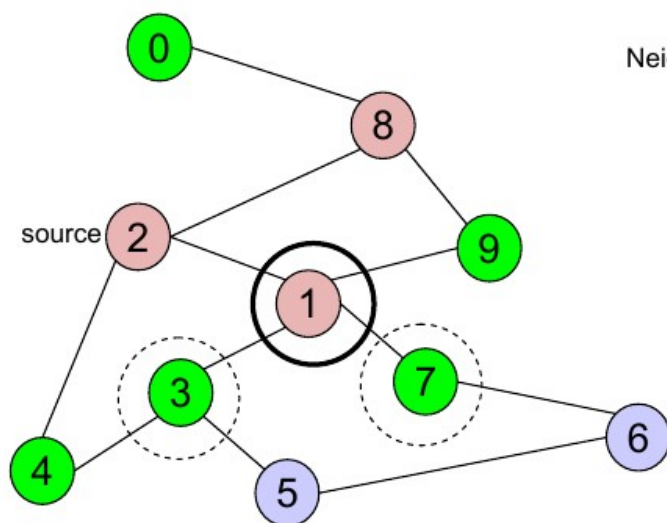
Mark newly visited neighbors 0, 9

$Q = \{ 8, 1, 4 \} \rightarrow \{ 1, 4, 0, 9 \}$

Dequeue 8.

- Place all unvisited neighbors of 8 on the queue.
- Notice that 2 is not placed on the queue again, it has been visited!

# BFS Example



Adjacency List

Neighbors →	0	8
	1	3 7 9 2
	2	8 1 4
	3	4 5 1
	4	2 3
	5	3 6
	6	7 5
	7	1 6
	8	2 0 9
	9	1 8

Visited Table (T/F)

0	T
1	T
2	T
3	T
4	T
5	F
6	F
7	T
8	T
9	T

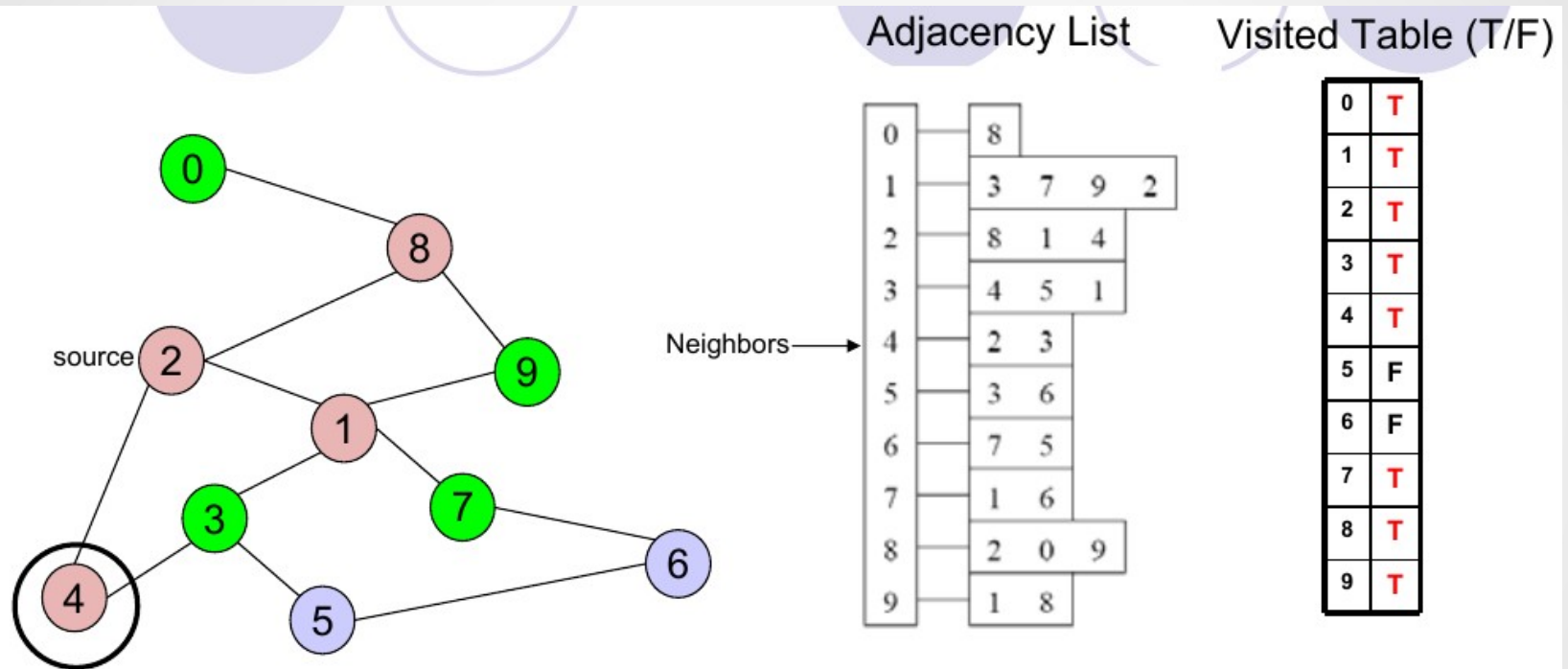
Mark newly visited neighbors 3, 7

$Q = \{ 1, 4, 0, 9 \} \rightarrow \{ 4, 0, 9, 3, 7 \}$

Dequeue 1.

- Place all unvisited neighbors of 1 on the queue.
- Only nodes 3 and 7 haven't been visited yet.

# BFS Example

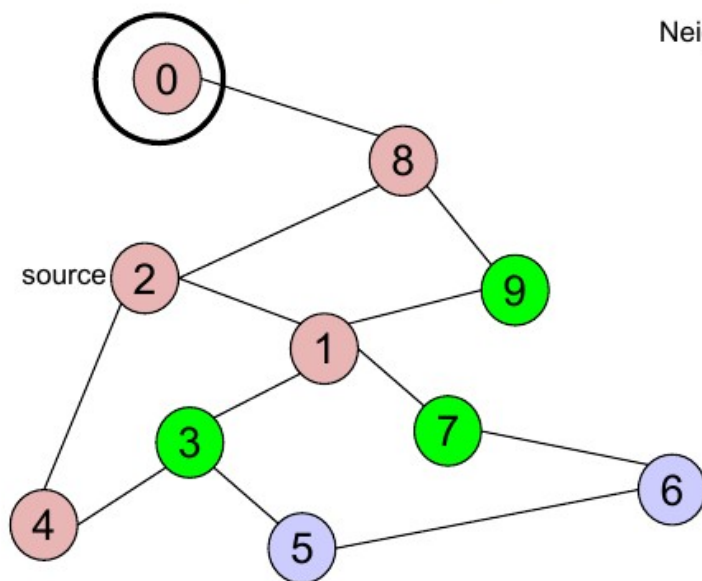


$Q = \{4, 0, 9, 3, 7\} \rightarrow \{0, 9, 3, 7\}$

Dequeue 4.

-- 4 has no unvisited neighbors!

# BFS Example



Adjacency List

Neighbors →

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

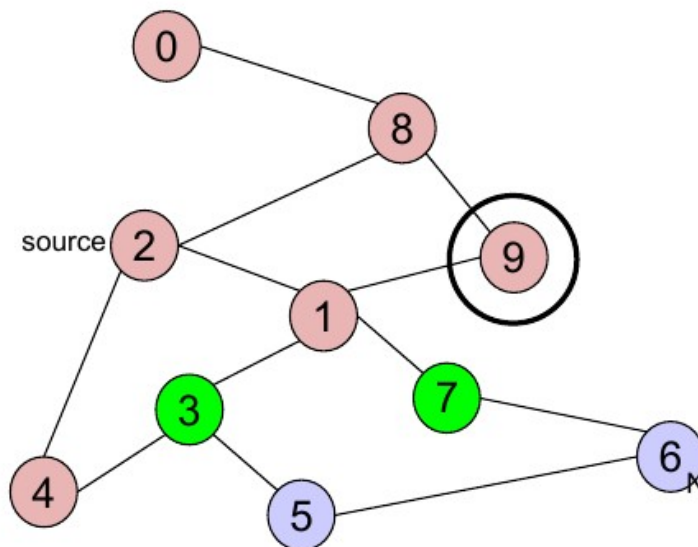
0	T
1	T
2	T
3	T
4	T
5	F
6	F
7	T
8	T
9	T

$Q = \{0, 9, 3, 7\} \rightarrow \{9, 3, 7\}$

Dequeue 0.

-- 0 has no unvisited neighbors!

# BFS Example



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

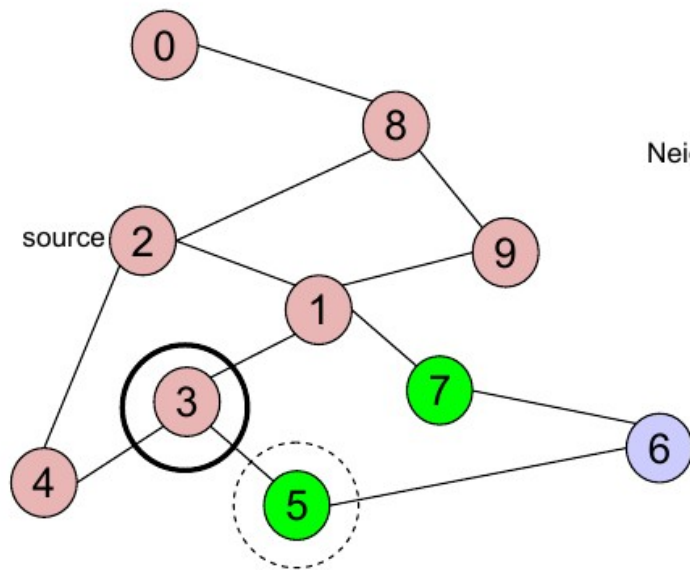
0	T
1	T
2	T
3	T
4	T
5	F
6	F
7	T
8	T
9	T

$Q = \{9, 3, 7\} \rightarrow \{3, 7\}$

Dequeue 9.

-- 9 has no unvisited neighbors!

# BFS Example



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Neighbors →

Visited Table (T/F)

0	T
1	T
2	T
3	T
4	T
5	T
6	F
7	T
8	T
9	T

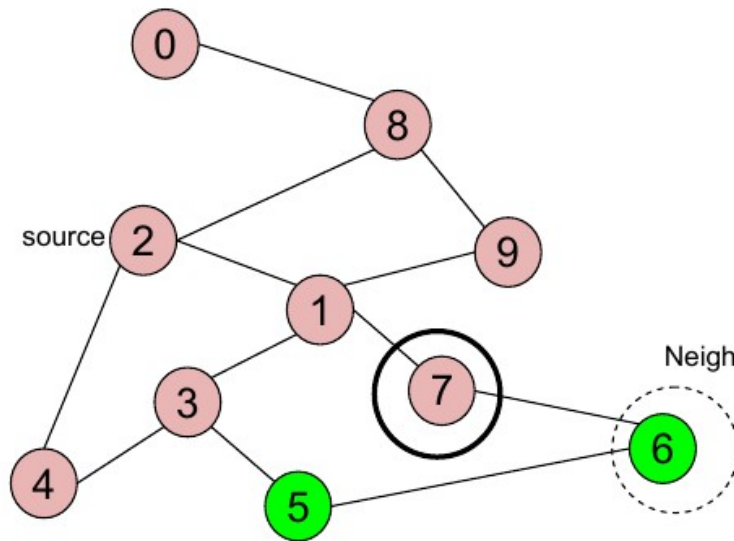
Mark new visited  
Vertex 5

$Q = \{3, 7\} \rightarrow \{7, 5\}$

Dequeue 3.

-- place neighbor 5 on the queue.

# BFS Example



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T
1	T
2	T
3	T
4	T
5	T
6	T
7	T
8	T
9	T

Mark new visited  
Vertex 6

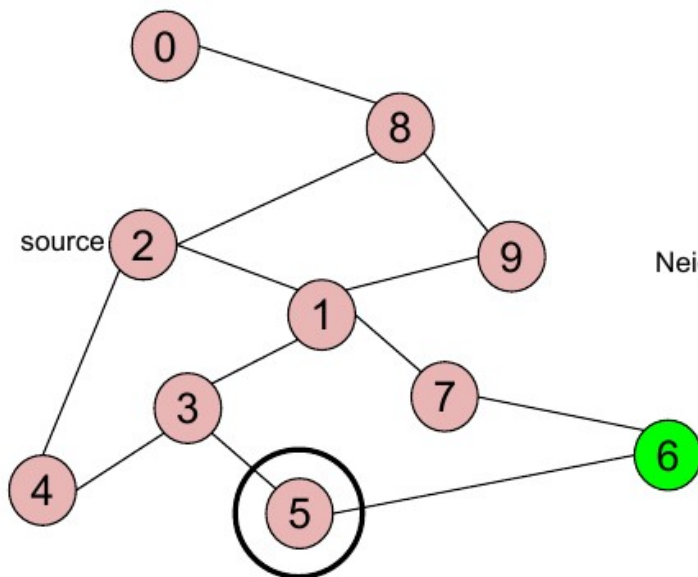
$Q = \{7, 5\} \rightarrow \{5, 6\}$

Dequeue 7.

-- place neighbor 6 on the queue



# BFS Example



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T
1	T
2	T
3	T
4	T
5	T
6	T
7	T
8	T
9	T

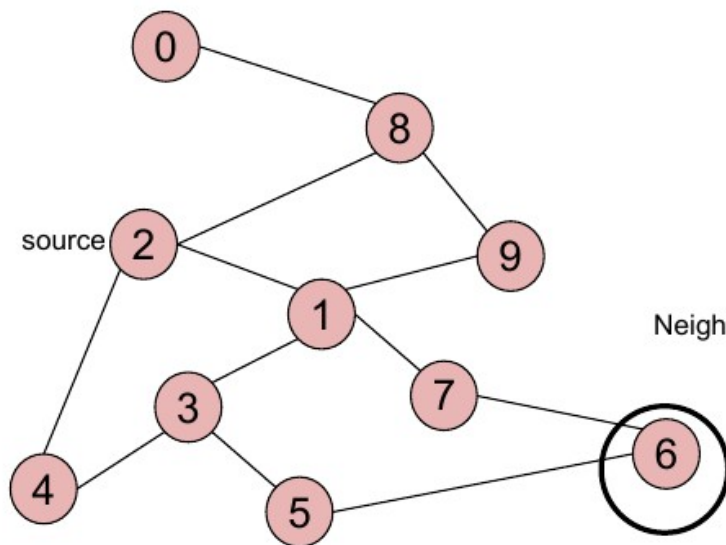
$Q = \{5, 6\} \rightarrow \{6\}$

Dequeue 5.

-- no unvisited neighbors of 5



# BFS Example



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

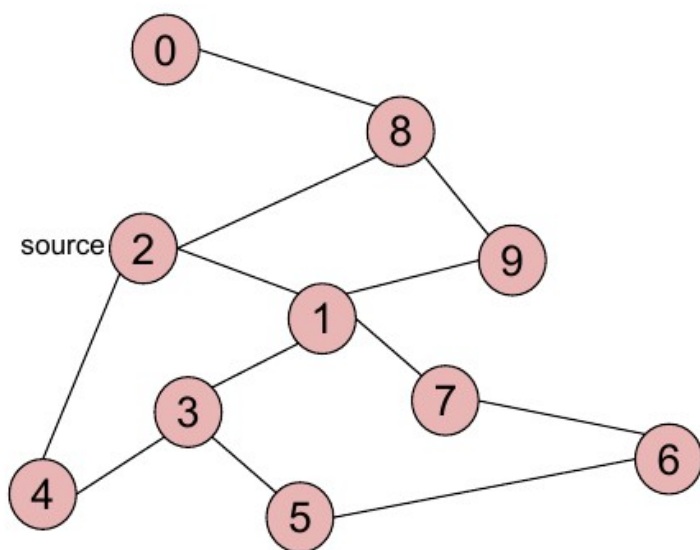
0	T
1	T
2	T
3	T
4	T
5	T
6	T
7	T
8	T
9	T

$Q = \{6\} \rightarrow \{\}$

Dequeue 6.

-- no unvisited neighbors of 6

# BFS Example



**Q = { } STOP!!! Q is empty!!!**

Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T
1	T
2	T
3	T
4	T
5	T
6	T
7	T
8	T
9	T

What did we discover?

Look at "visited" tables.

There exists a path from source vertex 2 to all vertices in the graph

# Applications of BFS

What can we do with the BFS code we just discussed?

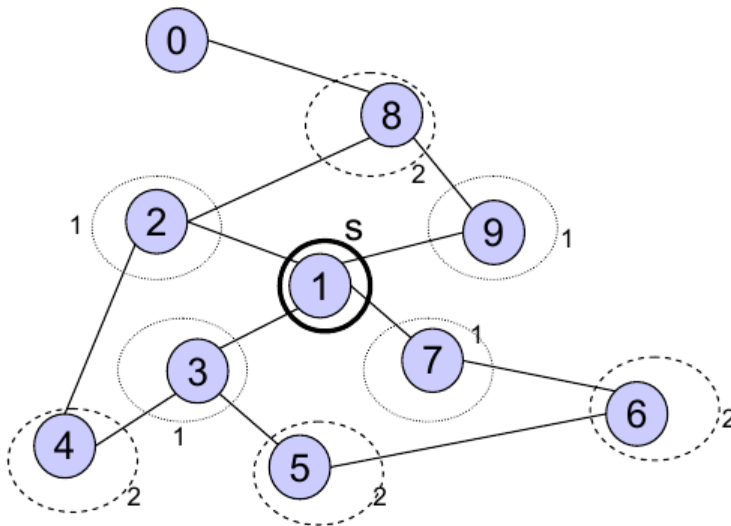
- Is there a **path** from source  $s$  to a vertex  $v$ ?
  - Check  $flag[v]$ .
- Is an undirected graph **connected**?
  - Scan array  $flag[ ]$ .
  - If there exists  $flag[u] = false$  then ...
- Is a directed graph **strongly connected**?
  - Scan array  $flag[ ]$ .
  - If there exists  $flag[u] = false$  then ...

# Other Applications of BFS

- To find the shortest path from a vertex  $s$  to a vertex  $v$  in an **unweighted graph**
- To find the **length** of such a path
- To find out if a graph contains **cycles**
- To find the **connected components** of a graph that is not connected
- To construct a **BSF tree/forest** from a graph

## Problem 2 - BFS and Shortest Path Problem

- Given any source vertex  $s$ , BFS visits the other vertices at increasing distances away from  $s$ . In doing so, BFS discovers shortest paths from  $s$  to the other vertices.
- What do we mean by “distance”? The number of edges on a path from  $s$  (unweighted graph)



Example

Consider  $s$ =vertex 1

Nodes at distance 1?  
2, 3, 7, 9

Nodes at distance 2?  
8, 6, 5, 4

Nodes at distance 3?  
0

# Finding Shortest Paths Using BFS

- The BFS code we have seen
  - find out if there exists a path from a vertex  $s$  to a vertex  $v$
  - **prints the vertices** of a graph (connected/strongly connected).
- What if we want to find
  - the **shortest path** from  $s$  to a vertex  $v$  (or to every other vertex)?
  - **the length of the shortest path** from  $s$  to a vertex  $v$ ?
- In addition to array *flag*[], use an array named **prev**[], one element per vertex.
  - $prev[w] = v$  means that vertex  $w$  was visited right after  $v$

# Finding Shortest Paths Using BFS

## Algorithm $BFS(s)$

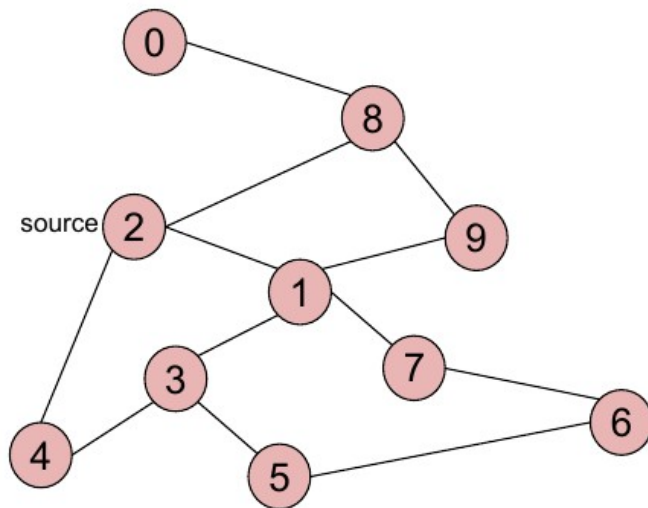
```
1.  for each vertex  $v$ 
2.      do  $flag(v) := \text{false};$ 
3.       $pred[v] := -1;$ 
4.   $Q = \text{empty queue};$ 
5.   $flag[s] := \text{true};$ 
6.   $enqueue(Q, s);$ 
7.  while  $Q$  is not empty
8.      do  $v := dequeue(Q);$ 
9.      for each  $w$  adjacent to  $v$ 
10.         do if  $flag[w] = \text{false}$ 
11.            then  $flag[w] := \text{true};$ 
12.                 $pred[w] := v;$ 
13.                 $enqueue(Q, w)$ 
```

← initialize all  $pred[v]$  to -1

← already got shortest path from  $s$  to  $v$

← record where you came from

# Example



**Q = { } STOP!!! Q is empty!!!**

Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	T	2
2	T	-
3	T	1
4	T	2
5	T	3
6	T	7
7	T	1
8	T	2
9	T	8

*prev[ ]*

***prev[ ]* now can be traced backward to report the path!**

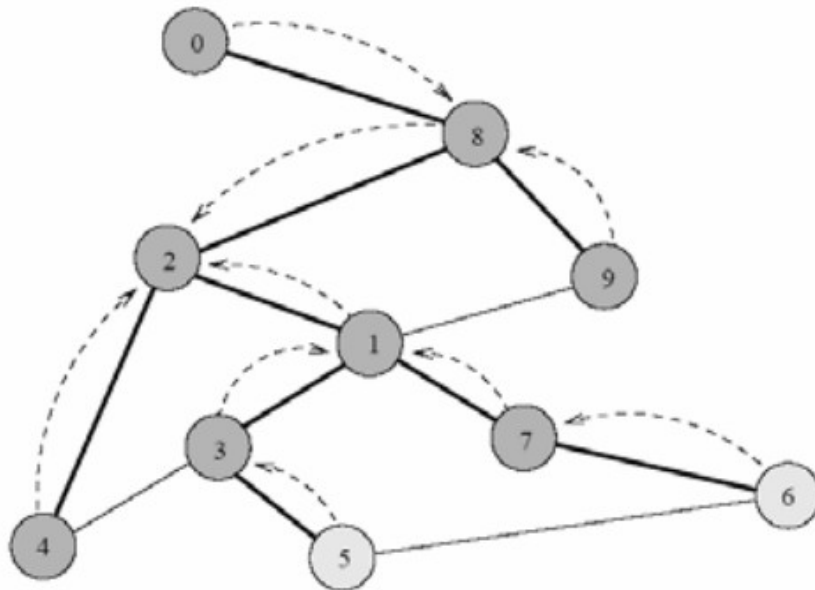


## Example

```
for each  $w$  adjacent to  $v$ 
    if  $flag[w] = \text{false}$  {
         $flag[w] = \text{true}$ ;
         $prev[w] = v$ ; // visited  $w$  right after  $v$ 
        enqueue( $w$ );
    }
```

- To print the **shortest path** from  $s$  to a vertex  $u$ , start with  $prev[u]$  and **backtrack** until reaching the source  $s$ .
  - Running time of backtracking = ?
- To find the length of the shortest path from  $s$  to  $u$ , start with  $prev[u]$ , backtrack and **increment a counter** until reaching  $s$ .
  - Running time = ?

# Example of Path Reporting



nodes      visited from

0	8
1	2
2	-
3	1
4	2
5	3
6	7
7	1
8	2
9	8

Try some examples; report path from s to v:

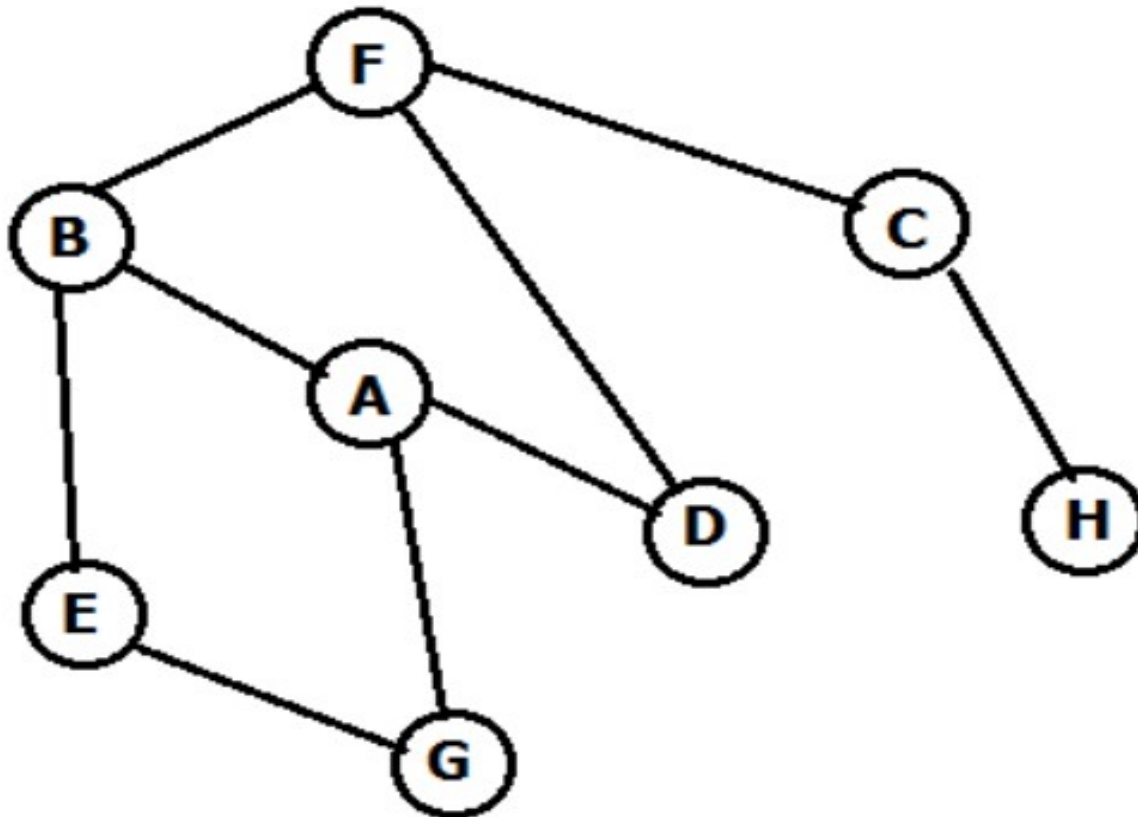
Path(2-0)  $\Rightarrow$

Path(2-6)  $\Rightarrow$

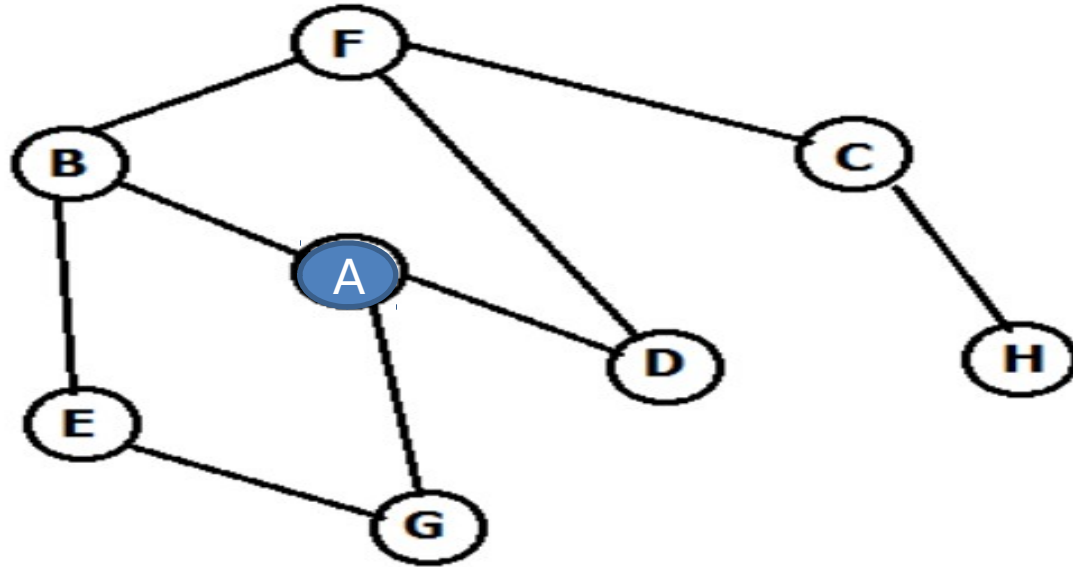
Path(2-1)  $\Rightarrow$

## PROBLEM 3

- Apply DFS to the following graph



## PROBLEM 3- DFS

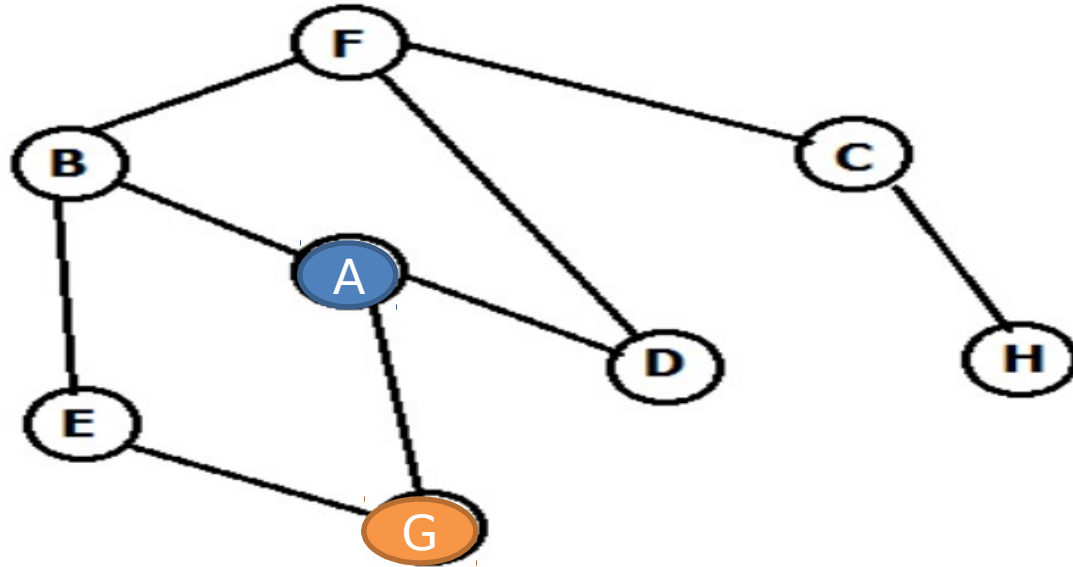


DFS Discovered Order= {A}

Stack={B,D,G}

↑  
head

## PROBLEM 3- DFS

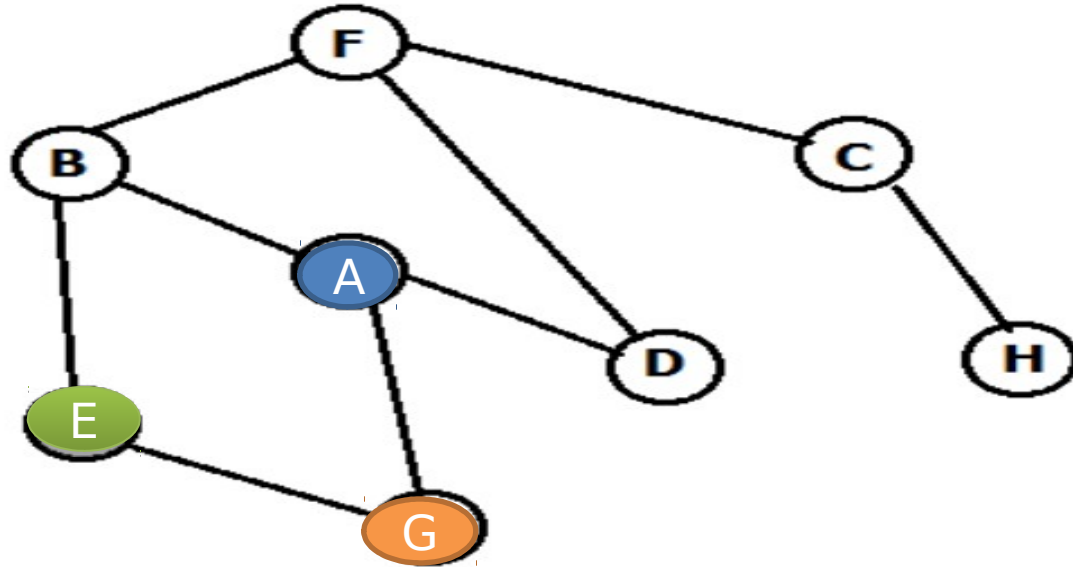


DFS Discovered Order= {A,G}

Stack={B,D,E}

↑  
head

## PROBLEM 3- DFS

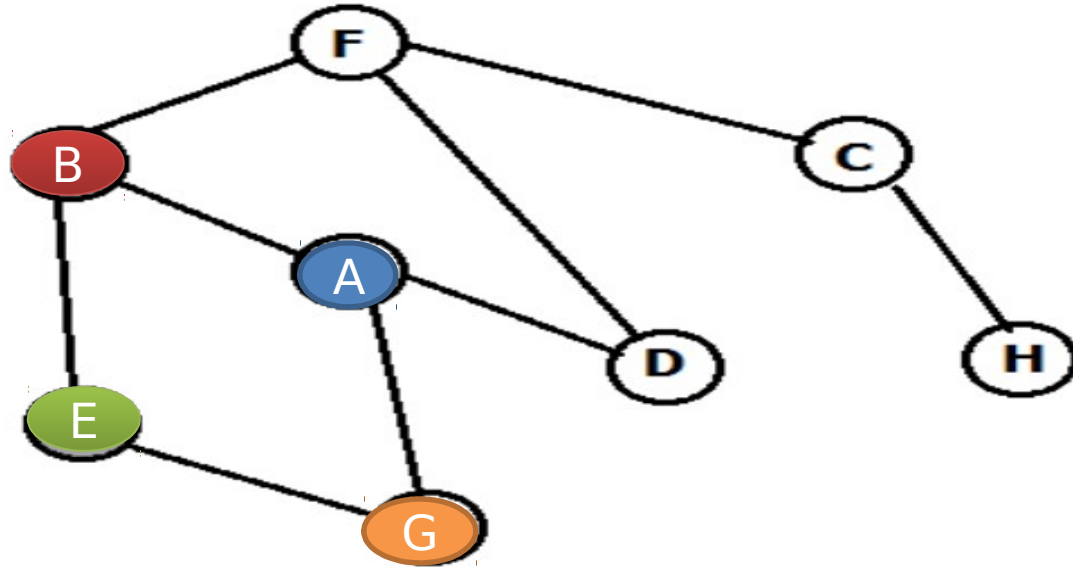


DFS Discovered Order= {A,G,E}

Stack={B,D,B}

↑  
head

## PROBLEM 3- DFS

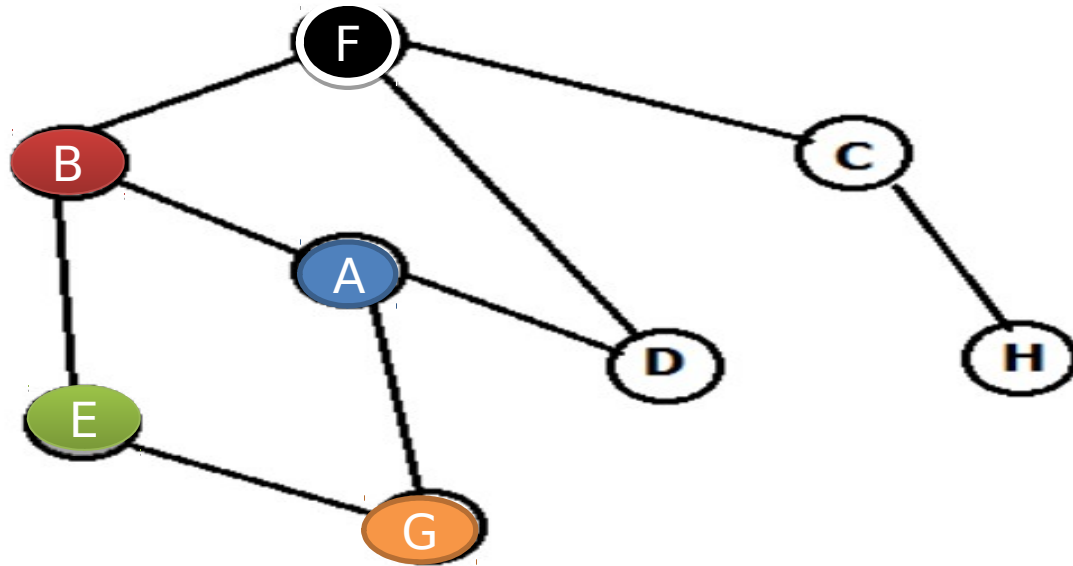


DFS Discovered Order= {A,G,E,B}

Stack={B,D,F}

↑  
head

## PROBLEM 3- DFS



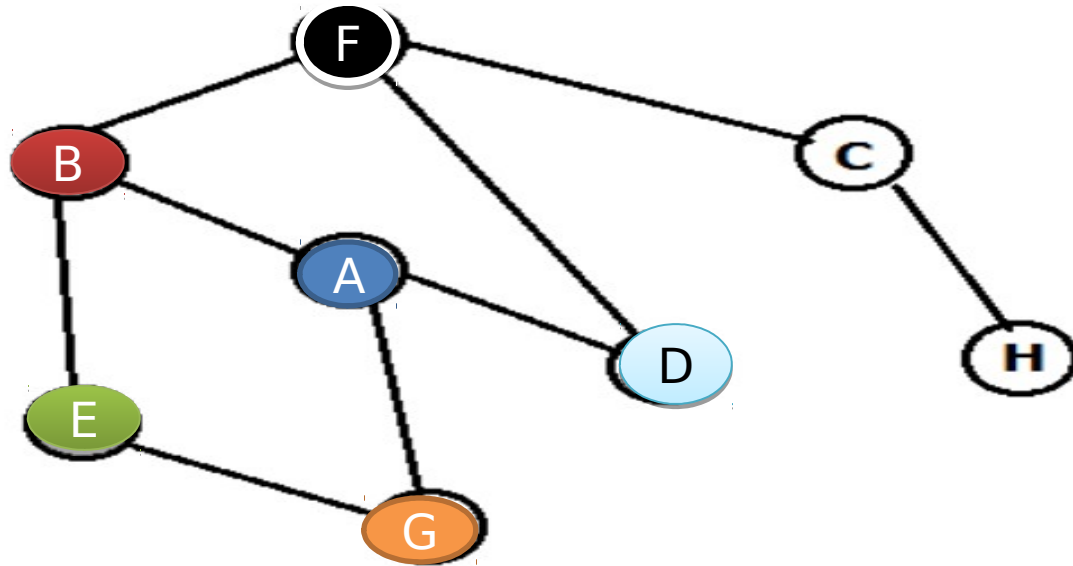
DFS Discovered Order= {A,G,E,B,F}

Stack={B,D,C,D}

↑  
head



## PROBLEM 3- DFS

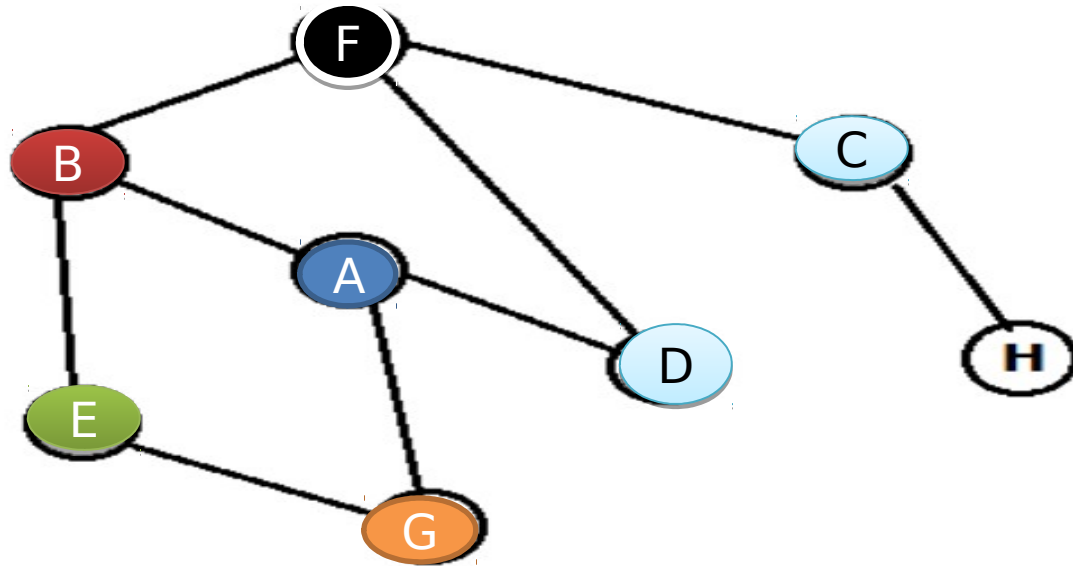


DFS Discovered Order= {A,G,E,B,F, D}

Stack={B,D,C}

↑  
head

## PROBLEM 3- DFS

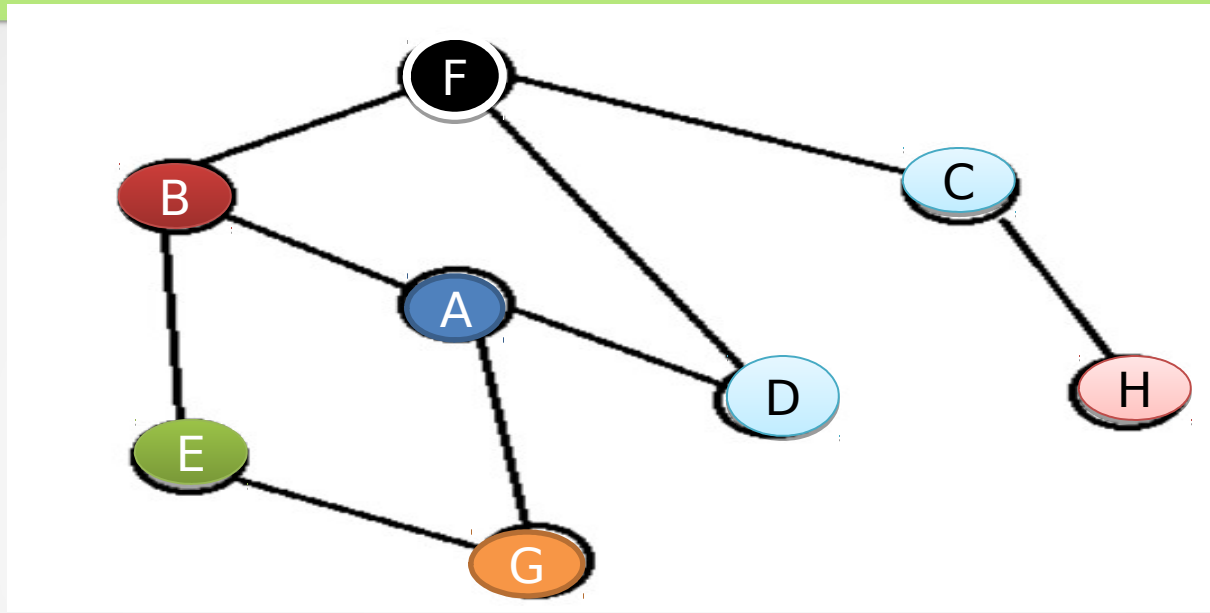


DFS Discovered Order= {A,G,E,B,F, D, C}

Stack={B,D,H}

↑  
head

## PROBLEM 3- DFS



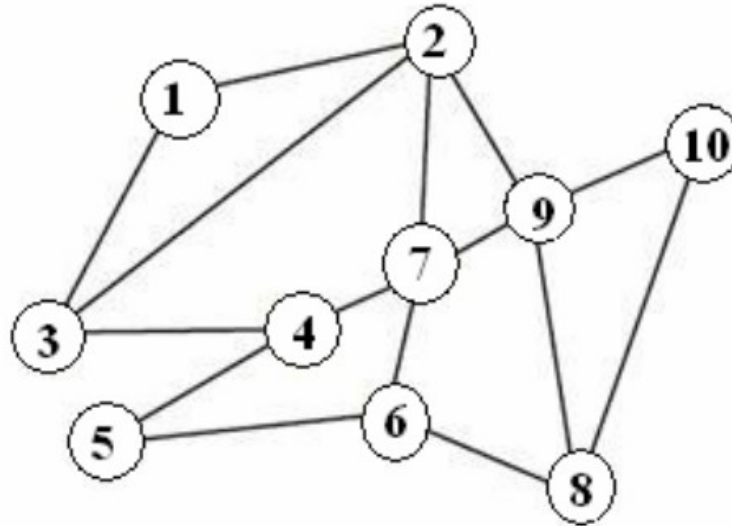
DFS Discovered Order= {A,G,E,B,F, D, C, H}

Stack= { }

↑  
head

## PROBLEM 4

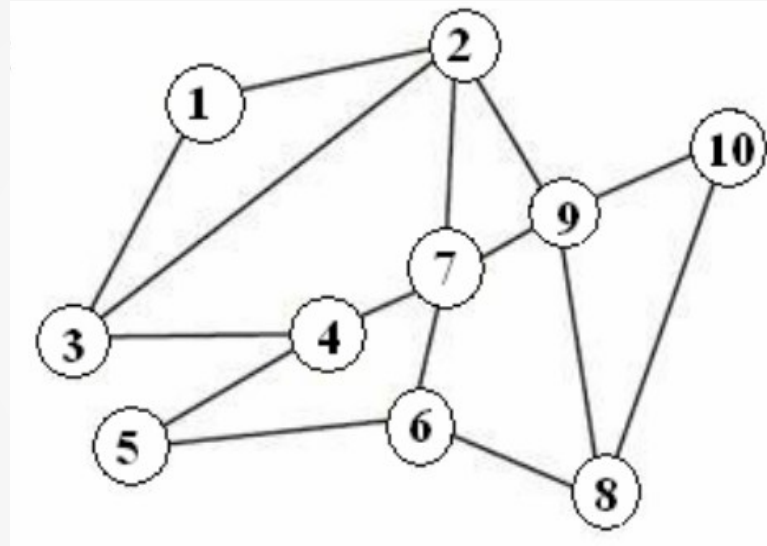
- What is a bipartite graph? Is the following graph is bipartite?



- A bipartite graph is a graph whose vertices can be grouped into two groups such that all the edges are between these two vertex group and there is no edge within a group.

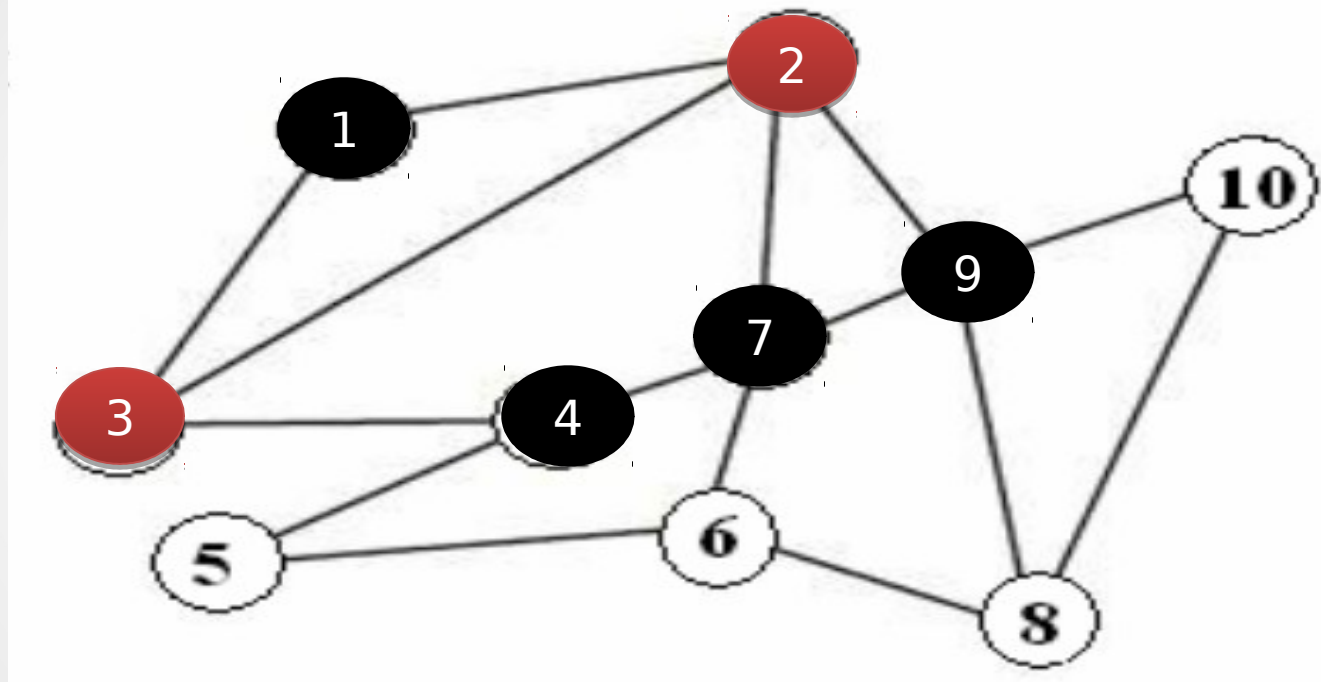
## PROBLEM 4

- What is a bipartite graph? Is the following graph is bipartite?



- If a graph can be two colored and there is no odd length cycles then the graph is bipartite.

## PROBLEM 4



- 4 and 7 has same color. So this graph is not bipartite.
- Also there many odd length cycles. One of them is (1,2,3)

# Greedy Algorithm:

*“An algorithm that builds a solution in small steps, choosing a decision at each step myopically [=locally, not considering what may happen ahead] to optimize some underlying criterion.”*

- Produces optimum solution for some problems.
  - Minimum spanning tree
  - Single-source shortest paths
  - Huffman trees
- Produces good approximate solutions for some other problems.
  - NP-Complete problems such as graph coloring

# Problem 5 - Coffee Shop

- You own a coffee shop that has  $n$  customers.
- It takes  $t_i$  minutes to prepare coffee for the  $i^{th}$  customer.
- $i^{th}$  customer's value for you (i.e. how frequent s/he comes to your shop) is  $v_i$
- If you start preparing coffee for the  $i^{th}$  customer at time  $s_i$  you finish at  $f_i = s_i + t_i$
- All customers arrive at the same time.
- You can prepare one coffee at a time.
- There is no gap after you finish one coffee and start another.



# Coffee Shop

- You are asked to design an algorithm.

**Input:**  $n, t_i, v_i$

**Output:** A schedule (i.e. ordering of customer requests)

**Aim:** Minimize wait time especially for valued customers

$$\text{Minimize : } \sum_{i=1}^n f_i * v_i$$

- What is the time complexity of your algorithm?
- Run your algorithm for a sample input.

# Algorithm

```
input :  $t[], v[], n$   
1 for  $i \leftarrow 1$  to  $n$  do  
2    $w[i, 1] \leftarrow v[i]/t[i];$            // weight of each customer:  
3    $w[i, 2] \leftarrow i;$   
4  $\text{sort}(w, \text{dec}, 1);$   
5  $t \leftarrow 0;$   
6  $\text{cost} \leftarrow 0;$   
7 for  $j \leftarrow 1$  to  $n$  do  
8    $\text{schedule}[j] \leftarrow w[j, 2];$   
9    $f[j] \leftarrow t + t[\text{schedule}[j]];$   
10   $t \leftarrow f[j];$   
11   $\text{cost} \leftarrow \text{cost} + f[\text{schedule}[j]] * v[\text{schedule}[j]];$   
12 return  $\text{schedule}, f, \text{cost}$ 
```

# Complexity of the algorithm

- Both for loops take  $O(n)$  time.
- Complexity of the algorithm depends on *sort method*.
- Typically  $O(n \log n)$

# A sample input

## Input:

- $t_1 = 2, t_2 = 3, t_3 = 1$
- $V_1 = 10, v_2 = 2, v_3 = 1$

## Output:

- Weights = 5, 0.67, 1
- Schedule: 1, 3, 2
- Finish times: 2, 3, 6
- Cost:  $2 \cdot 10 + 3 \cdot 1 + 6 \cdot 2 = 35$