

CS3100 DSA2

Fall 2023

Warm up:

Show that the sum of degrees of all nodes in any undirected graph is even

Show that for any graph $G = (V, E)$,
 $\sum_{v \in V} \deg(v)$ is even

$\sum_{v \in V} \deg(v)$ is even

CS 3100

Data Structures and Algorithms 2

Lecture 4: Depth First Search, Topological Sort

Co-instructors: Robbie Hott and Tom Horton
Fall 2023

Readings in CLRS 4th edition:

- Chapter 20: Sections 20-3, 20-4, and 20-5

Announcements

- Course website and schedule updated
 - PS1 available, due Sept 8 (Friday) at 11:59pm
 - Try to work on the first half through this weekend
 - PA1 available, due Sept 17 (Sunday) at 11:59pm
 - Assignment deadlines have been added to the calendar
- Office Hours
 - Prof Hott: 3-5pm Monday, 4-5pm Thursday
 - Prof Horton: 2-3:30 Mon, 3:30-5 Tue, 2:30-4 Thu, 2-3 Fri
 - TA office hours posted soon, check our website

Breadth First Search

Traversing Graphs

“Traversing” means processing each vertex edge in some organized fashion by following edges between vertices

- We speak of *visiting* a vertex. Might do something while there.

Recall traversal of binary trees:

- Several strategies: In-order, pre-order, post-order
- Traversal strategy implies an order of visits
- We used recursion to describe and implement these

Graphs can be used to model interesting, complex relationships

- Often traversal used just to process the set of vertices or edges
- Sometimes traversal can identify interesting properties of the graph
- Sometimes traversal (perhaps modified, enhanced) can answer interesting questions about the problem-instance that the graph models

BFS: Specific Input/Output

Input:

- A graph \underline{G}
- single start vertex \underline{s}

Output:

- Distance from \underline{s} to each node in \underline{G} (distance = number of edges)
- Breadth-First Tree of \underline{G} with root \underline{s}

Strategy:

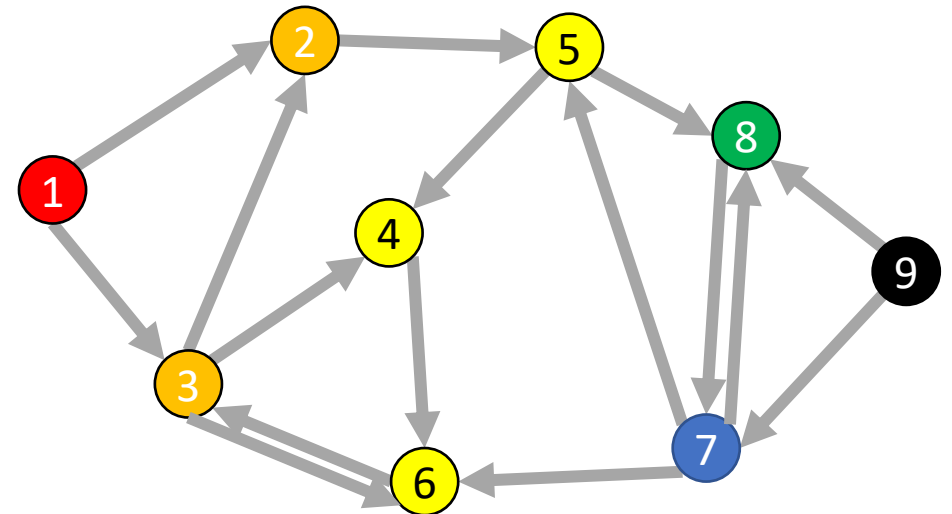
Start with node \underline{s} , visit all neighbors of \underline{s} , then all neighbors of neighbors of \underline{s} , ...

Important: The paths in this BFS tree represent the **shortest paths** from s to each node in G

- But edge weight's (if any) not used, so “short” is in terms of number of edges in path

BFS

```
def bfs(graph, s):  
    toVisit.enqueue(s)  
    mark s as "seen"  
    While toVisit is not empty:  
        current = toVisit.dequeue()  
        for v in neighbors(current):  
            if v not seen:  
                mark v as seen  
                toVisit.enqueue(v)
```



BFS: Shortest Path

```
def shortest_path(graph, s, t):
```

```
    toVisit.enqueue(s)  
    mark s as "seen"
```

```
    While toVisit is not empty:
```

```
        current = toVisit.dequeue()
```

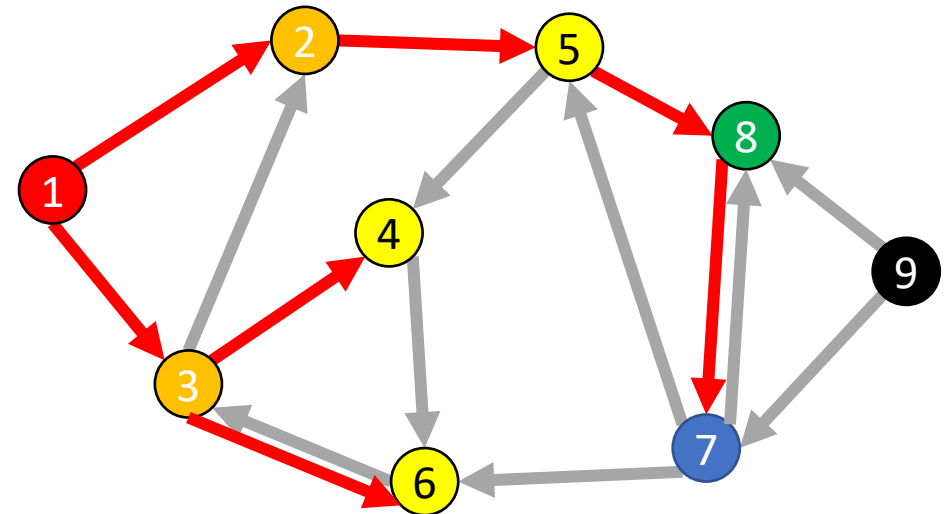
```
        for v in neighbors(current):
```

```
            if v not seen:
```

```
                mark v as seen
```

```
                toVisit.enqueue(v)
```

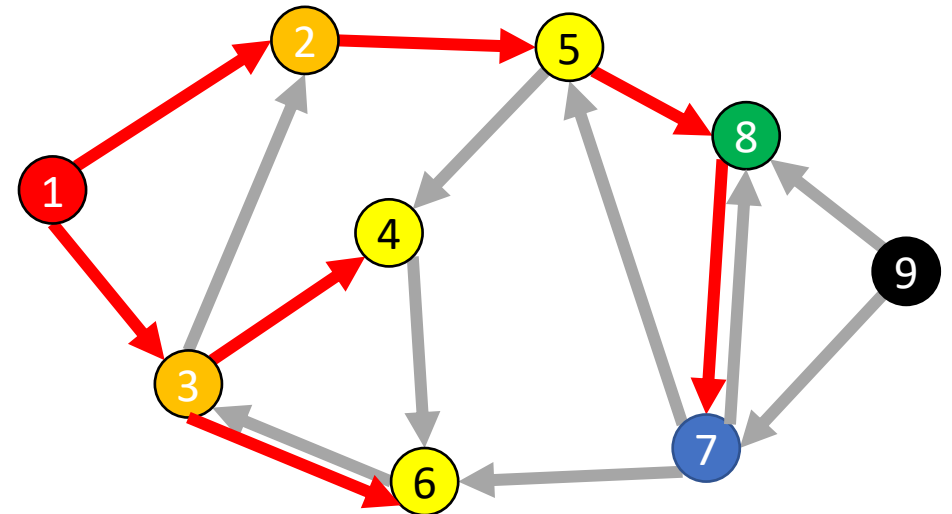
Idea: when it's seen, remember its "layer" depth!



BFS: Shortest Path

```
def shortest_path(graph, s, t):  
    layer = 0  
    toVisit.enqueue(s)  
    depth[s] = layer  
    While toVisit is not empty:  
        current = toVisit.dequeue()  
        layer = depth [current]  
        for v in neighbors(current):  
            if v does not have a depth:  
                depth[v]=layer+1  
                toVisit.enqueue(v)  
    return depth[t]
```

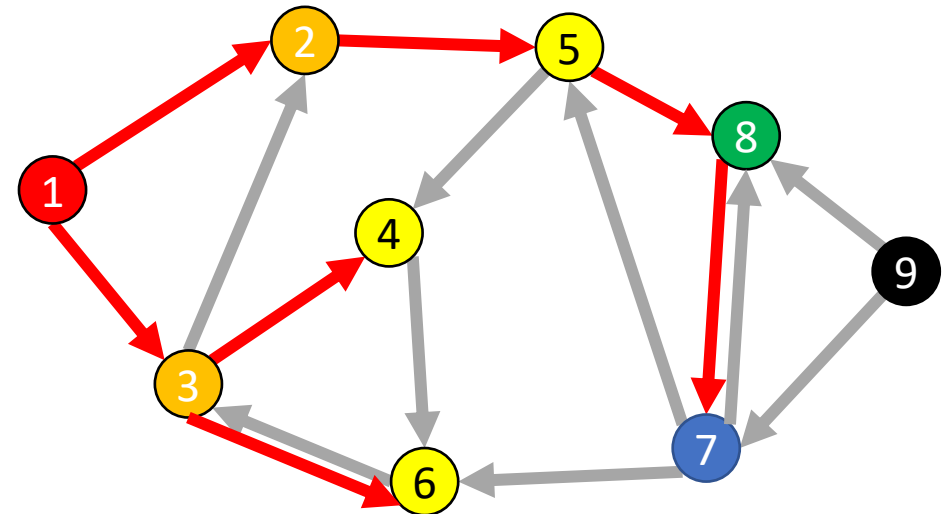
Idea: when it's seen, remember its "layer" depth!



BFS: Shortest Path

```
def shortest_path(graph, s, t):  
    layer = 0  
    depth = [-1,-1,-1,...] # Length matches |V|  
    toVisit.enqueue(s)  
    mark a as "seen"  
    depth[s] = 0  
    While toVisit is not empty:  
        current = toVisit.dequeue()  
        layer = depth[current]  
        if current == t:  
            return layer  
        for v in neighbors(current):  
            if v not seen:  
                mark v as seen  
                toVisit.enqueue(v)  
                depth[v] = layer + 1
```

Idea: when it's seen, remember its "layer" depth!



Breadth-first search from CLRS 20.2

BFS(G, s)

```
1  for each vertex  $u \in G.V - \{s\}$ 
2       $u.color = \text{WHITE}$ 
3       $u.d = \infty$ 
4       $u.\pi = \text{NIL}$ 
5   $s.color = \text{GRAY}$ 
6   $s.d = 0$ 
7   $s.\pi = \text{NIL}$ 
8   $Q = \emptyset$ 
9  ENQUEUE( $Q, s$ )
10 while  $Q \neq \emptyset$ 
11      $u = \text{DEQUEUE}(Q)$ 
12     for each  $v \in G.Adj[u]$ 
13         if  $v.color == \text{WHITE}$ 
14              $v.color = \text{GRAY}$ 
15              $v.d = u.d + 1$ 
16              $v.\pi = u$ 
17             ENQUEUE( $Q, v$ )
18      $u.color = \text{BLACK}$ 
```

From CLRS

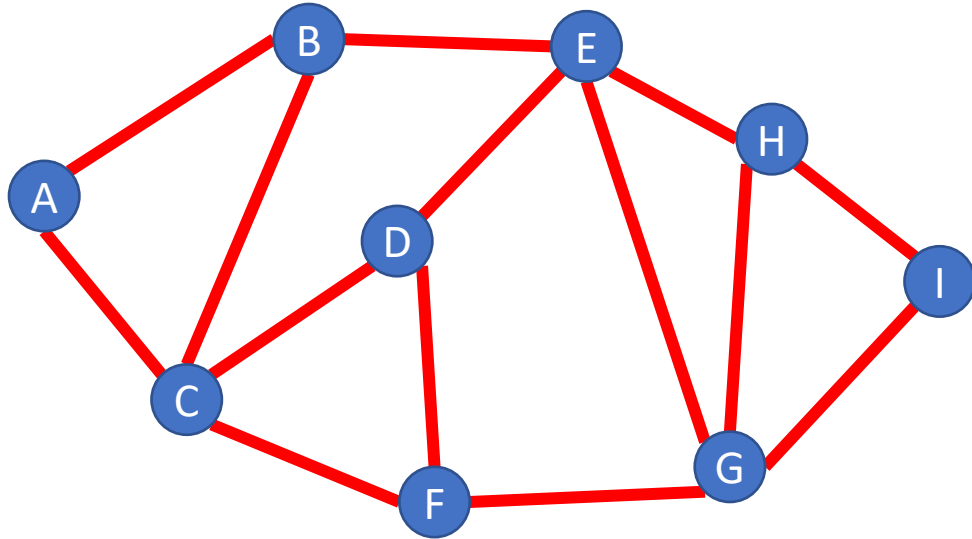
Vertices here have some properties:

- $color = \text{white/gray/black}$
- $d = \text{distance from start node}$
- $pi = \text{parent in tree, i.e. } v.pi \text{ is vertex by which } v \text{ was connected to BFS tree}$

Color meanings here:

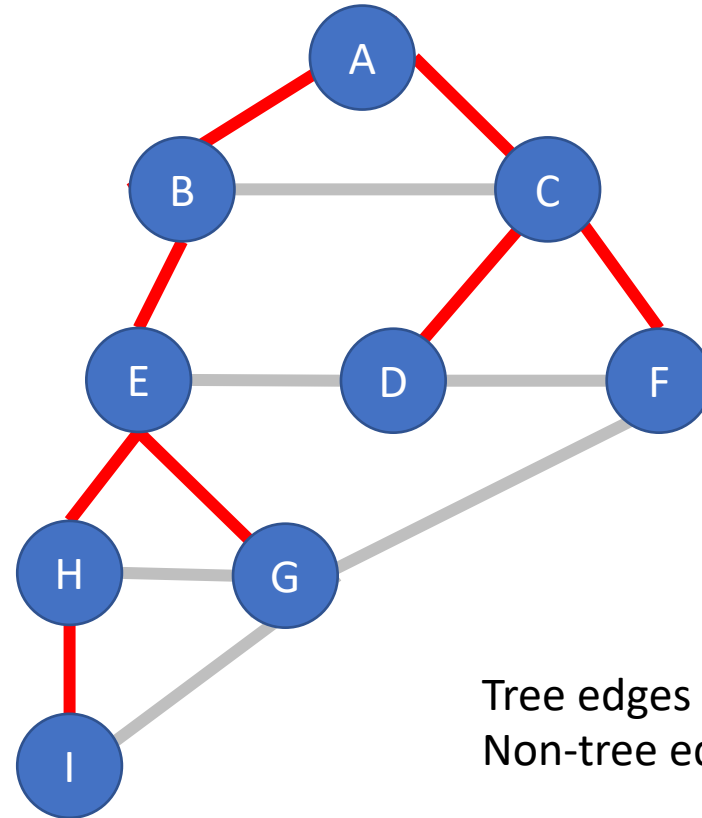
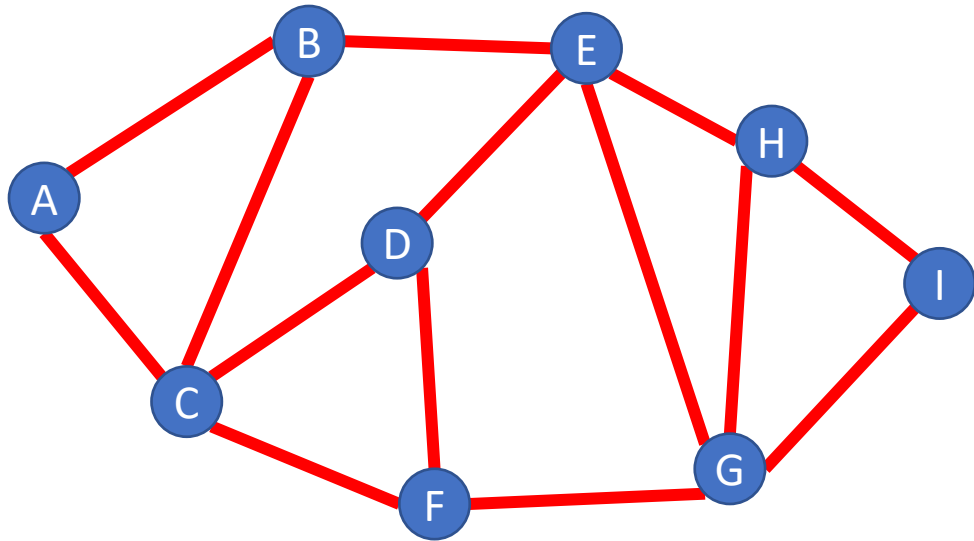
- White: haven't seen this vertex yet
- Gray: vertex has been seen and added to the queue for processing later
- Black: vertex has been removed from queue and its neighbors seen and added to the queue

Tree View of BFS Search Results



Draw BFS tree starting at A

Tree View of BFS Search Results



Tree edges in red
Non-tree edges in gray

Analysis for Breadth-first search

For a graph having V vertices and E edges

- Each edge is processed once in the while loop for a cost of $\Theta(E)$
- Each vertex is put into the queue once and removed from the queue and processed once, for a cost $\Theta(V)$
 - Also, cost of initializing colors or depth arrays is $\Theta(V)$

Total **time-complexity**: $\Theta(V + E)$

- For graph algorithms this is called “linear”

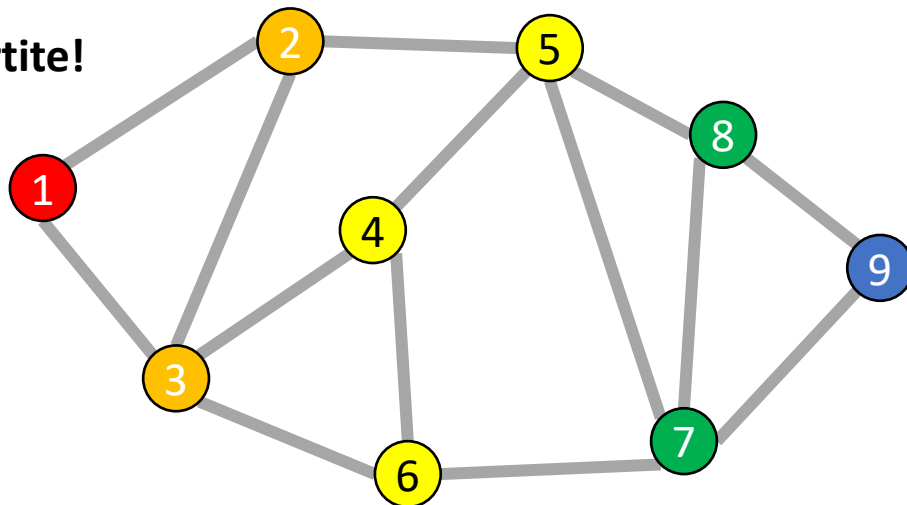
Space complexity: extra space is used for queue and also depth/color arrays, so $\Theta(V)$

Definition: Bipartite

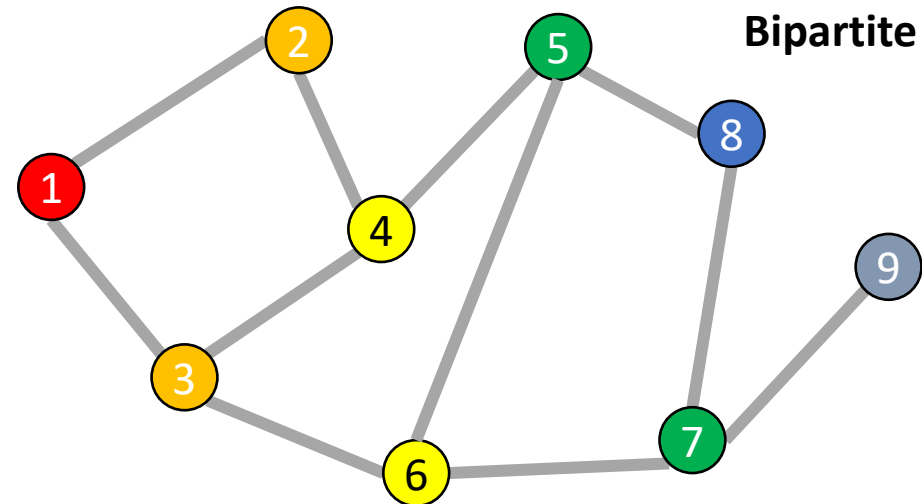
A (undirected) graph is **Bipartite** provided every vertex can be assigned to one of two teams such that every edge “crosses” teams

- Alternative: Every vertex can be given one of two colors such that no edges connect same-color nodes

Not Bipartite!



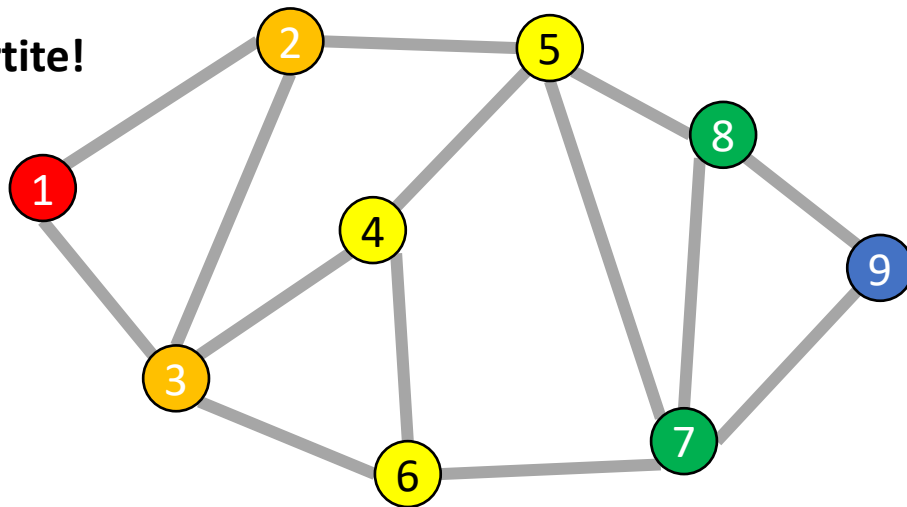
Bipartite!



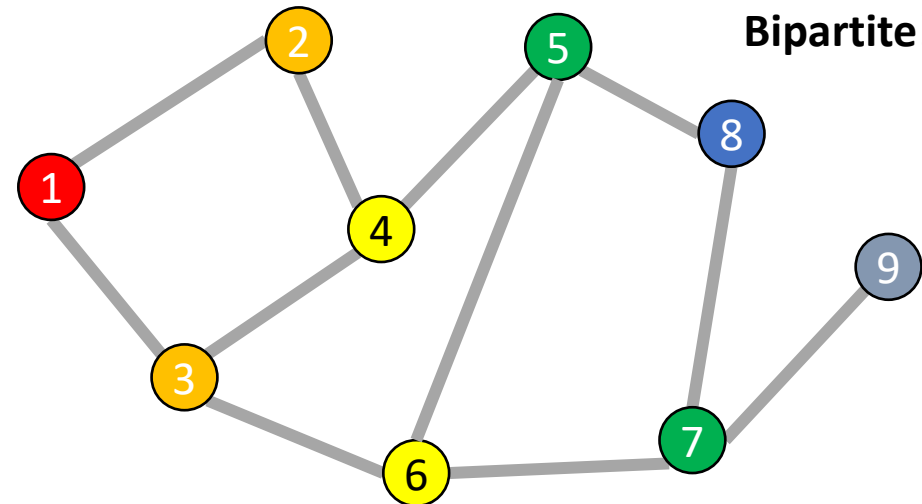
Odd Length Cycles

A graph is bipartite if and only if it has no odd length cycles

Not Bipartite!



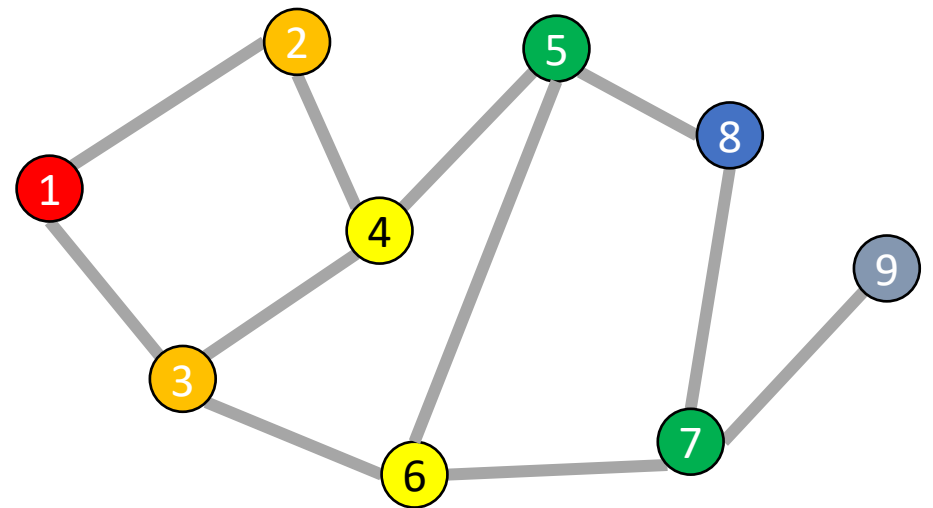
Bipartite!



BFS: Bipartite Graph?

```
def bfs(graph, s):  
    toVisit.enqueue(s)  
    mark s as "seen"  
    While toVisit is not empty:  
        current = toVisit.dequeue()  
        for v in neighbors(current):  
            if v not seen:  
                mark v as seen  
                toVisit.enqueue(v)
```

Idea: Check for edges in
the same layer!



BFS: Bipartite Graph?

```
def isBipartite(graph, s):
```

```
    toVisit.enqueue(s)
```

```
    mark s as "seen"
```

```
    While toVisit is not empty:
```

```
        current = toVisit.dequeue()
```

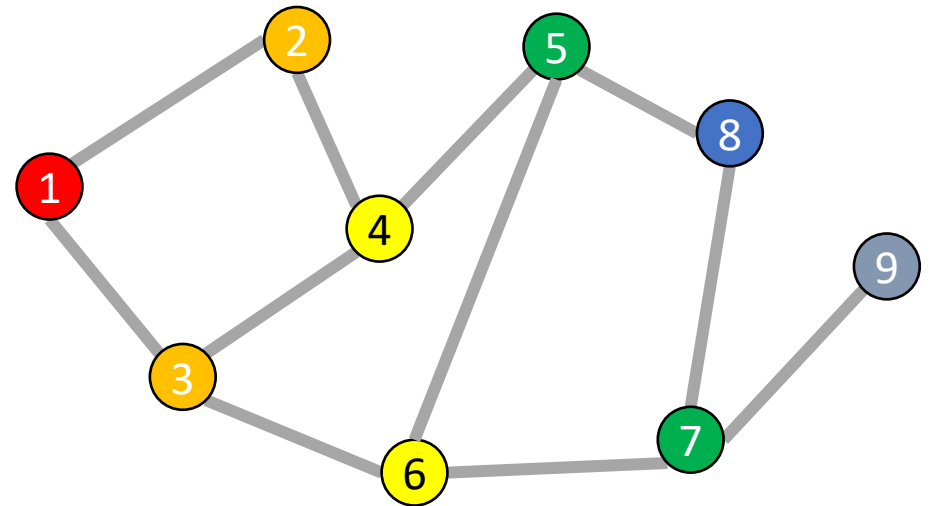
```
        for v in neighbors(current):
```

```
            if v not seen:
```

```
                mark v as seen
```

```
                toVisit.enqueue(v)
```

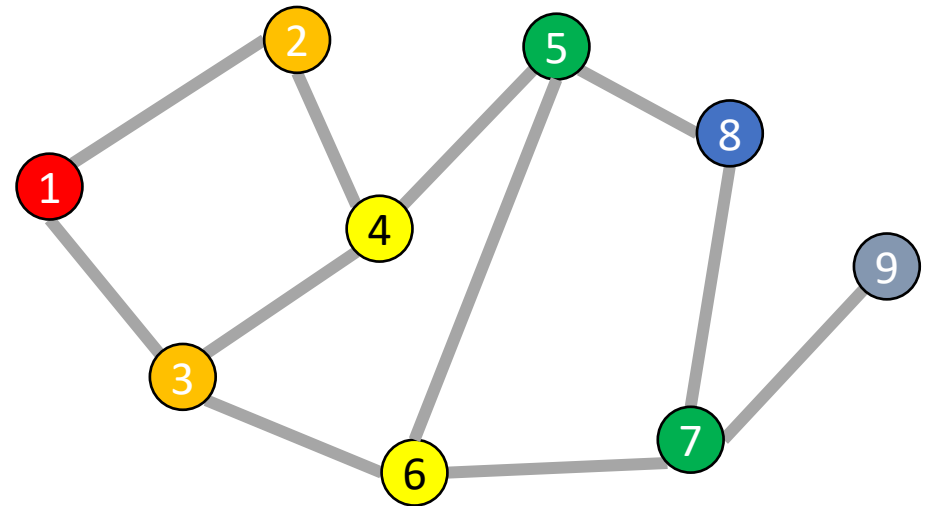
Idea: Check for edges in the same layer!



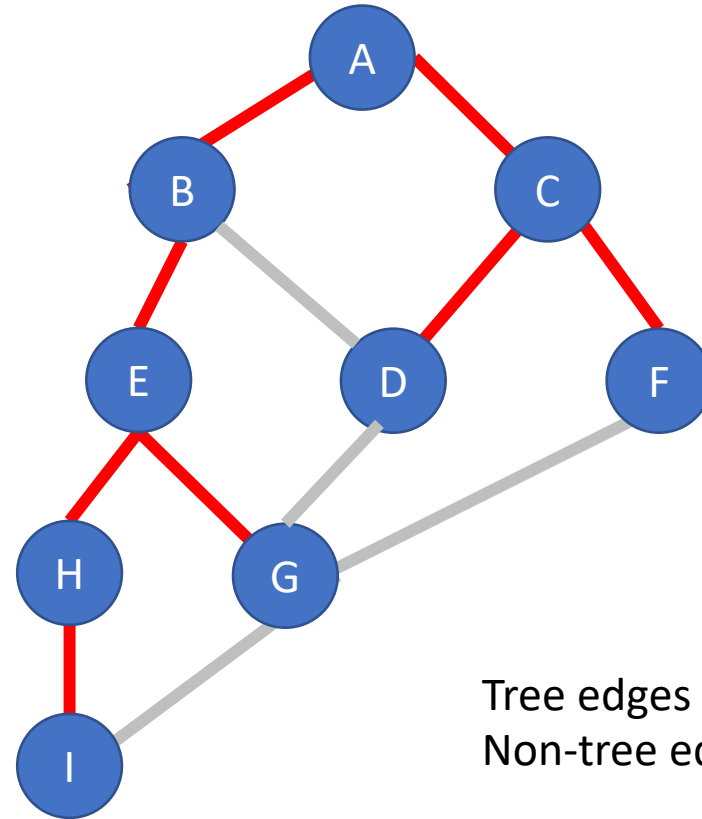
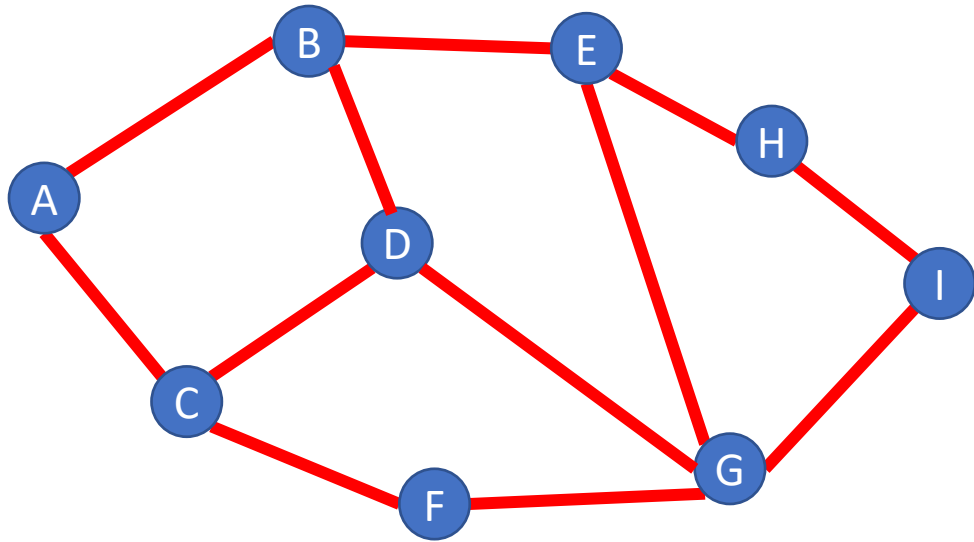
BFS: Bipartite Graph?

```
def isBipartite(graph, s):  
    layer = 0  
    depth = [-1,-1,-1,...] # Length matches |V|  
    toVisit.enqueue(s)  
    depth[s] = 0  
    While toVisit is not empty:  
        current = toVisit.dequeue()  
        layer = depth[current]  
        for v in neighbors(current):  
            if v not seen:  
                depth[v] = layer+1  
                toVisit.enqueue(v)  
            elif depth[v] == depth[current]:  
                return False  
    return True
```

Idea: Check for edges in the same layer!

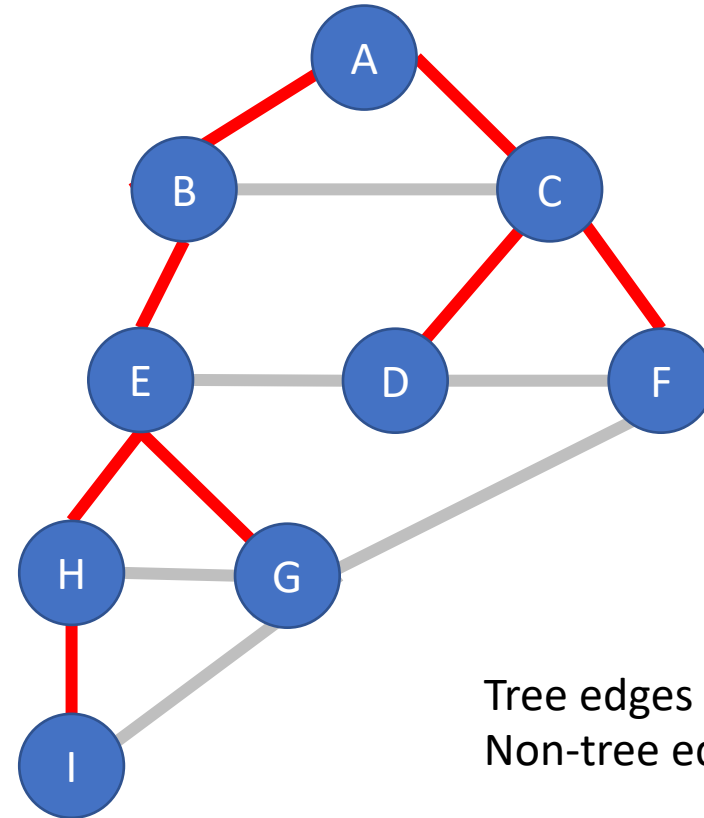
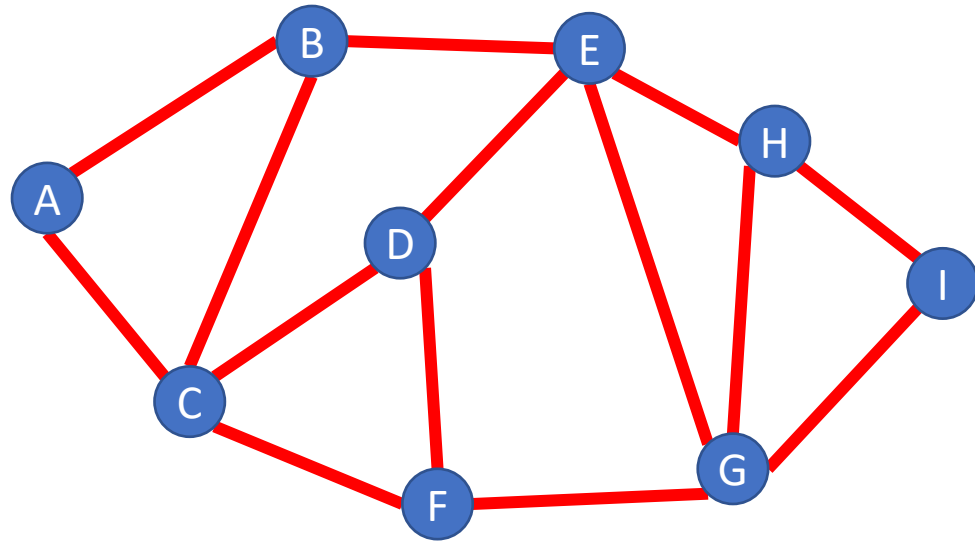


BFS Tree for a Bipartite Graph



Tree edges in red
Non-tree edges in gray

BFS Tree for a Non-Bipartite Graph



Tree edges in red
Non-tree edges in gray

Depth-First Search

DFS: the Strategy in Words

Depth-first search: Strategy

- Go as deep as can visiting un-visited nodes
 - Choose any un-visited vertex when you have a choice
- When stuck at a dead-end, backtrack as little as possible
 - Back up to where you could go to another unvisited vertex
- Then continue to go on from that point
- Eventually you'll return to where you started
 - Reach all vertices? Maybe, maybe not

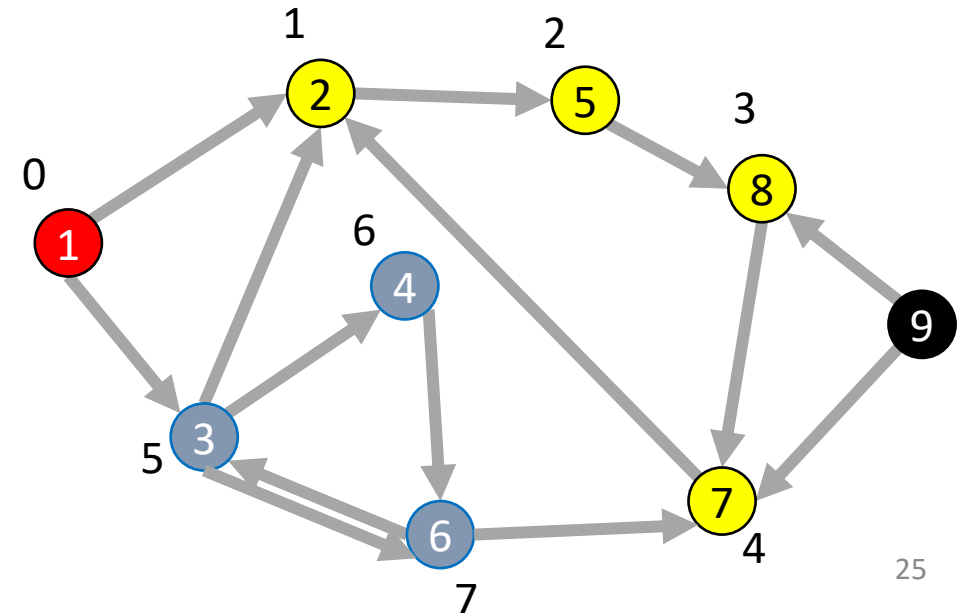
Depth-First Search

Input: a node s

Behavior: Start with node s , visit one neighbor of s , then all nodes reachable from that neighbor of s , then another neighbor of s ,...

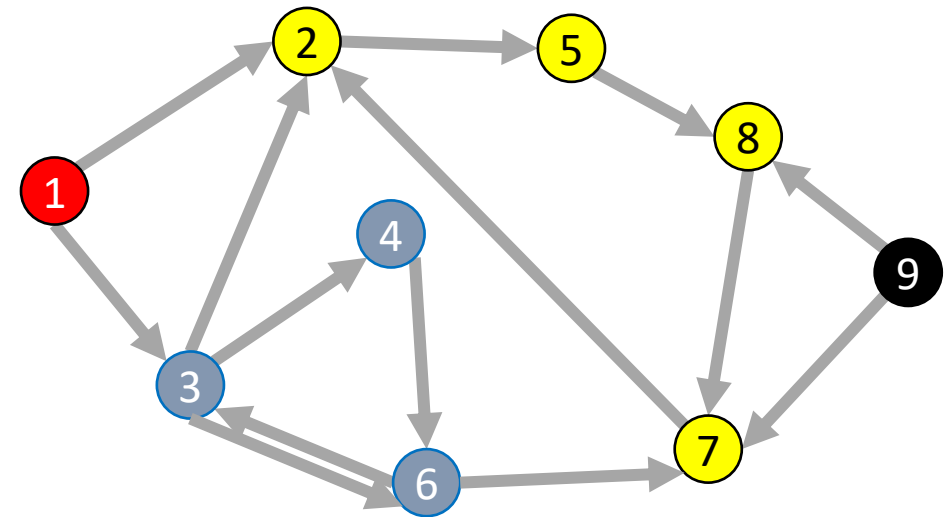
Output:

- Does the graph have a cycle?
- A topological sort of the graph.



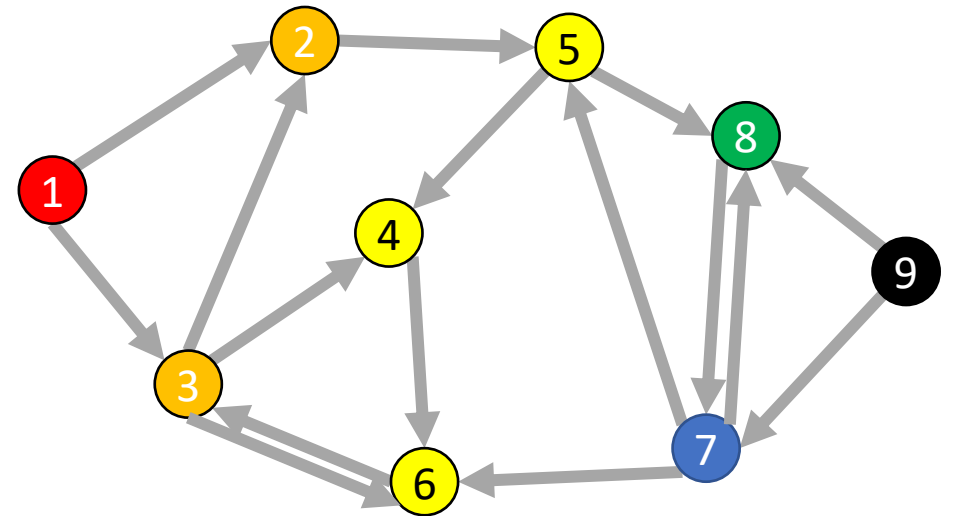
DFS: Non-recursively (less common)

```
def dfs(graph, s):  
    toVisit.push(s)  
    mark s as "seen"  
    While toVisit is not empty:  
        current = toVisit.pop()  
        for v in neighbors(current):  
            if v not seen:  
                mark v as seen  
                toVisit.push(v)
```



Remember: BFS

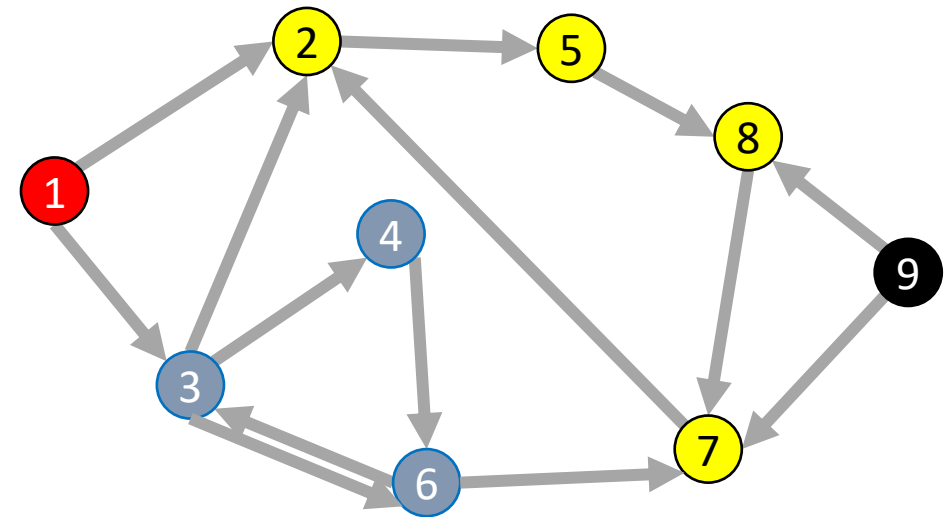
```
def bfs(graph, s):  
    toVisit.enqueue(s)  
    mark s as "seen"  
    While toVisit is not empty:  
        current = toVisit.dequeue()  
        for v in neighbors(current):  
            if v not seen:  
                mark v as seen  
                toVisit.enqueue(v)
```



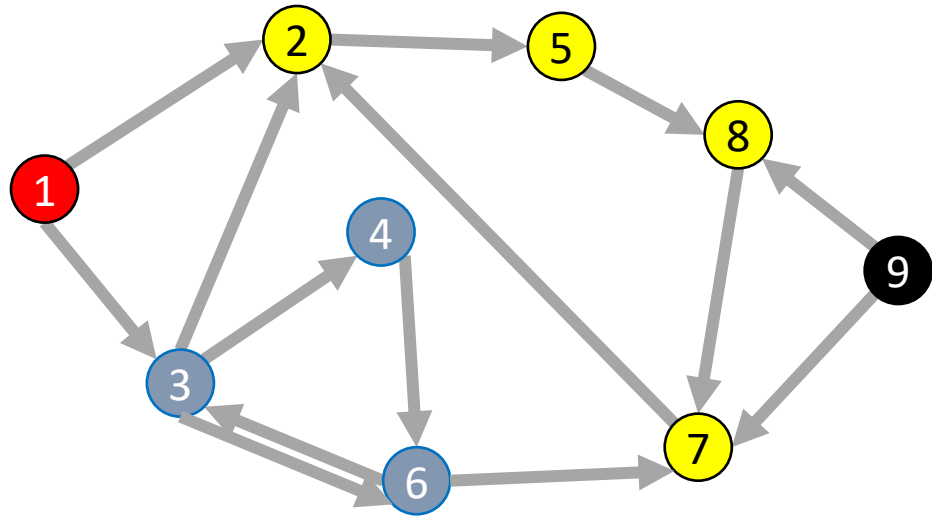
DFS: Recursively

```
def dfs(graph, s):  
    seen = [False, False, False, ...] # length matches |V|  
    done = [False, False, False, ...] # length matches |V|  
    dfs_rec(graph, s, seen, done)
```

```
def dfs_rec(graph, curr, seen, done)  
    mark curr as seen  
    for v in neighbors(current):  
        if v not seen:  
            dfs_rec(graph, v, seen, done)  
    mark curr as done
```



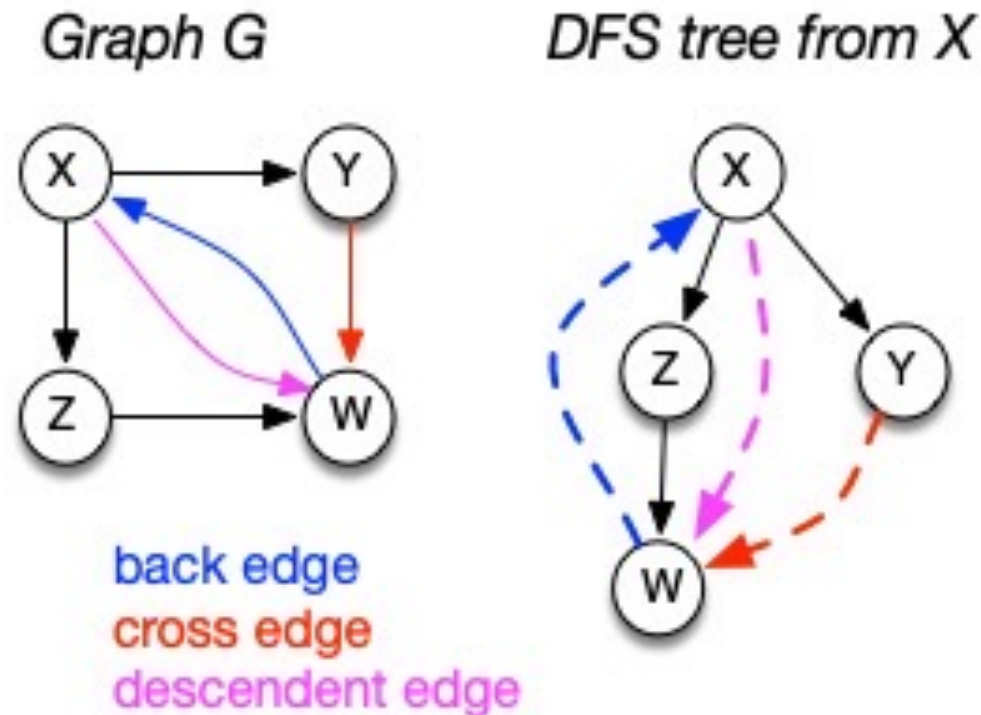
View of DFS Results as a Tree



Depth-first search tree

As DFS traverses a digraph, edges classified as:

- tree edge, back edge, descendant edge, or cross edge
- If graph undirected, do we have all 4 types?

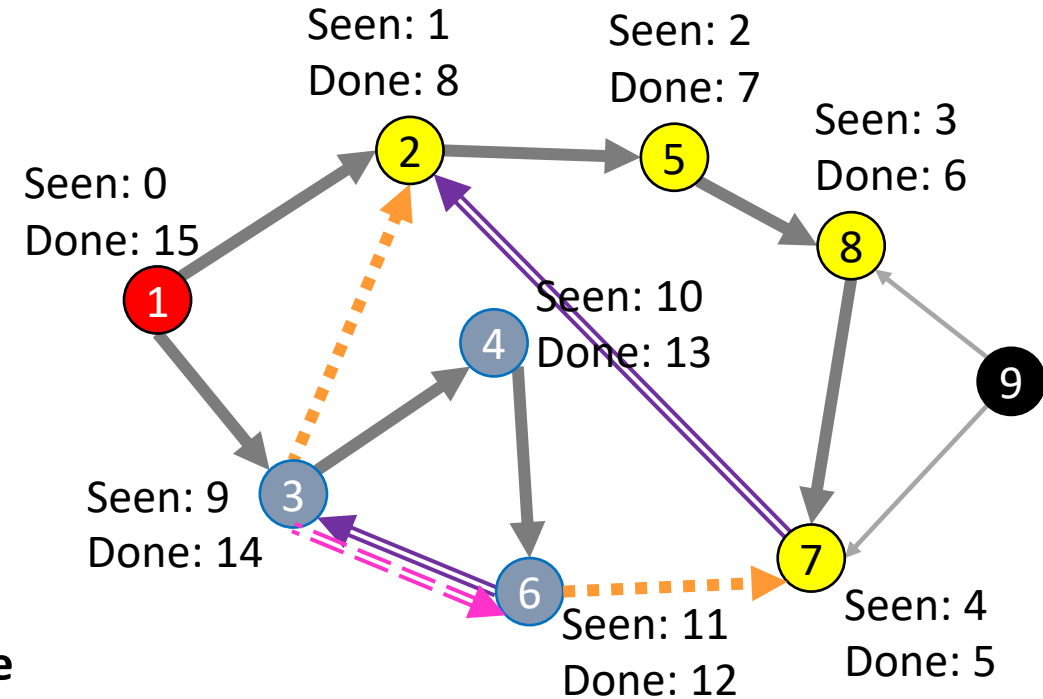


Using DFS

Consider the “seen times” and “done times”

Edges can be categorized:

- **Tree Edge** \longrightarrow
 - (a, b) was followed when pushing
 - (a, b) when b was **unseen** when we were at a
- **Back Edge** \Longrightarrow
 - (a, b) goes to an “ancestor”
 - a and b **seen** but not **done** when we saw (a, b)
 - $t_{seen}(b) < t_{seen}(a) < t_{done}(a) < t_{done}(b)$
- **Forward Edge** \dashrightarrow
 - (a, b) goes to a “descendent”
 - b was **seen** and **done** between when a was **seen** and **done**
 - $t_{seen}(a) < t_{seen}(b) < t_{done}(b) < t_{done}(a)$
- **Cross Edge** \dashrightarrow
 - (a, b) connects “branches” of the tree
 - b was **seen** and **done** before a was ever **seen**
 - (a, b) when $t_{done}(b) > t_{seen}(a)$ and

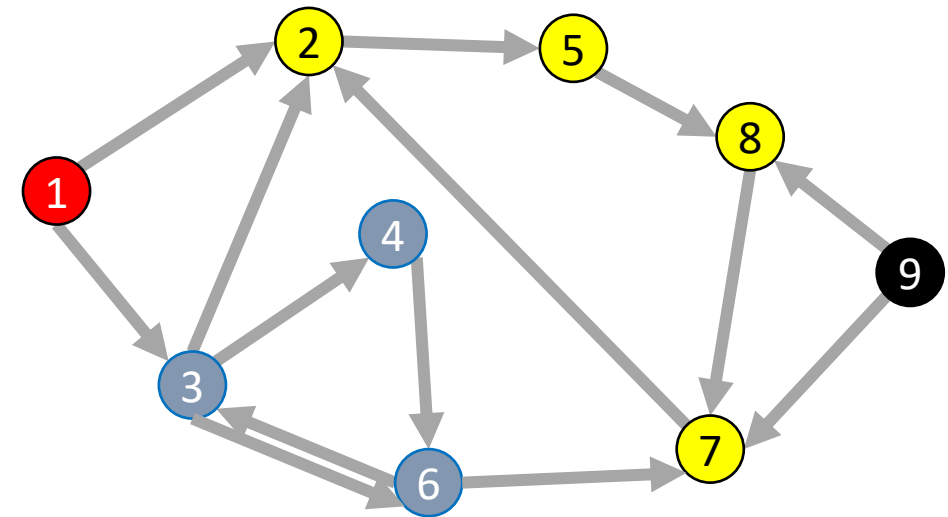


DFS: Cycle Detection

Idea: Look for a back edge!

```
def dfs(graph, s):  
    seen = [False, False, False, ...] # length matches |V|  
    done = [False, False, False, ...] # length matches |V|  
    dfs_rec(graph, s, seen, done)
```

```
def dfs_rec(graph, curr, seen, done)  
    mark curr as seen  
    for v in neighbors(current):  
        if v not seen:  
            dfs_rec(graph, v, seen, done)  
    mark curr as done
```



DFS: Cycle Detection

Idea: Look for a back edge!

```
def hasCycle(graph, s):  
    seen = [False, False, False, ...] # length matches |V|  
    done = [False, False, False, ...] # length matches |V|  
    dfs_rec(graph, s, seen, done)
```

```
def hasCycle_rec(graph, curr, seen, done)
```

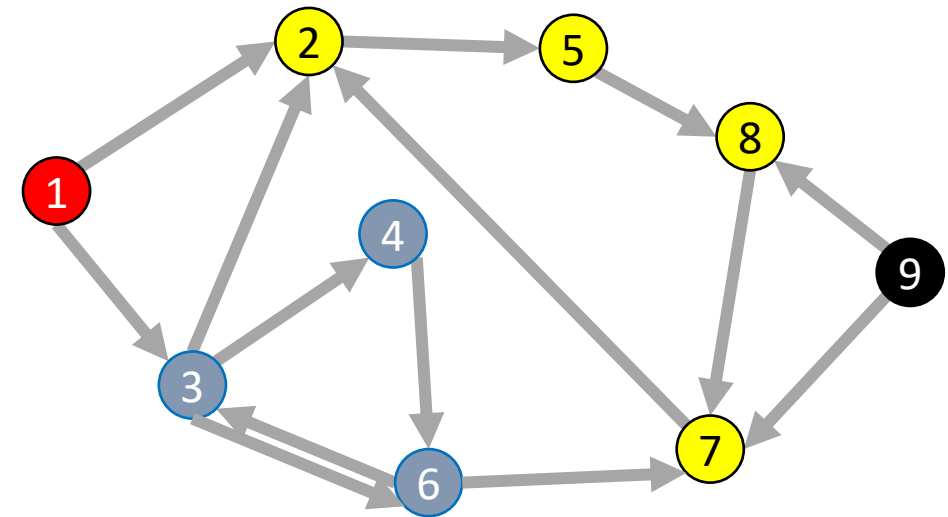
mark curr as seen

for v in neighbors(current):

if v not seen:

dfs_rec(graph, v, seen, done)

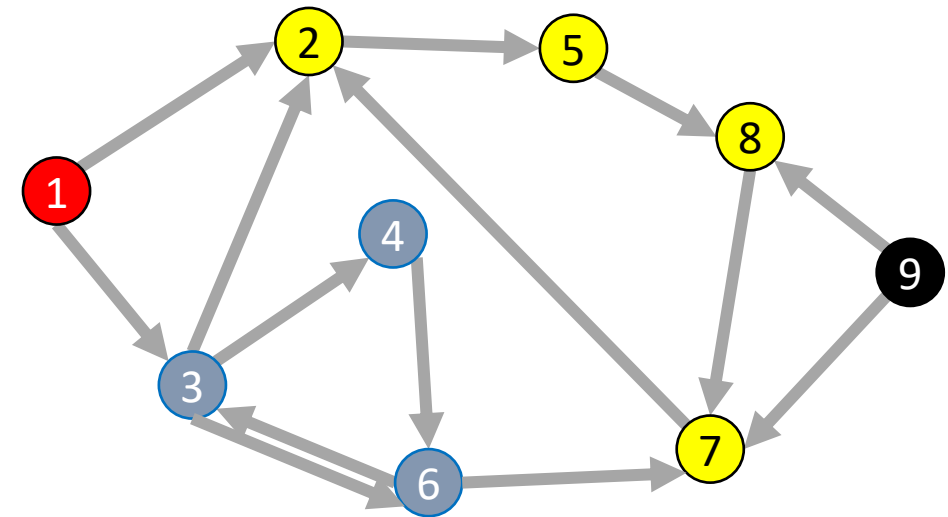
mark curr as done



DFS: Cycle Detection

Idea: Look for a back edge!

```
def hasCycle(graph, s):  
    seen = [False, False, False, ...] # length matches |V|  
    done = [False, False, False, ...] # length matches |V|  
    return hasCycle_rec(graph, s, seen, done)  
  
def hasCycle_rec(graph, curr, seen, done):  
    cycle = False  
    mark curr as seen  
    for v in neighbors(current):  
        if v seen and v not done:  
            cycle = True  
        elif v not seen:  
            cycle = dfs_rec(graph, v, seen, done) or cycle  
    mark curr as done  
    return cycle
```



Back Edges in Undirected Graphs

Finding back edges for an undirected graph is not **quite** this simple:

- The parent node of the current node is **seen** but not **done**
- Not a cycle, is it? It's the same edge you just traversed

Question: how would you modify our code to recognize this?

Time Complexity of DFS

For a digraph having V vertices and E edges

- Each edge is processed once in the while loop of `dfs_rec()` for a cost of $\Theta(E)$
 - Think about *adjacency list* data structure.
 - Traverse each list exactly once. (Never back up)
 - There are a total of E nodes in all the lists
- The non-recursive `dfs()` algorithm will do $\Theta(V)$ work even if there are no edges in the graph
- Thus over all time-complexity is $\Theta(V + E)$
 - Remember: this means the larger of the two values
 - Reminder: This is considered “linear” for graphs since there are two size parameters for graphs.
- Extra space is used for seen/done (or color) array.
 - Space complexity is $\Theta(V)$