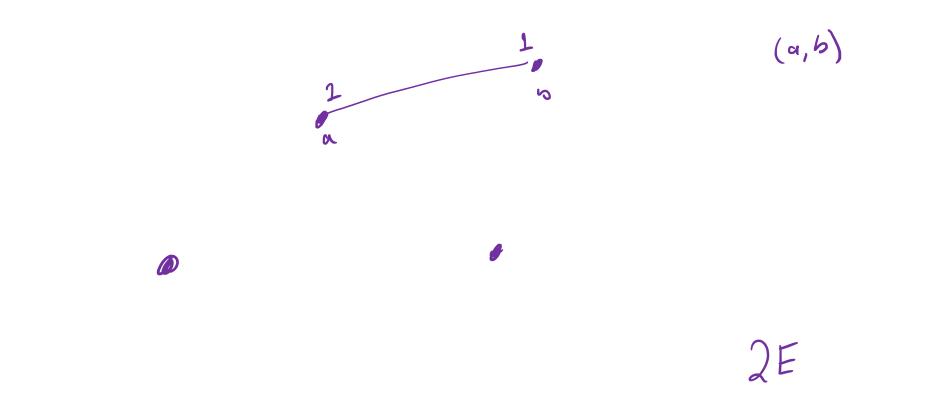
CS3100 DSA2 Spring 2024

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

Co-instructors: Robbie Hott and Ray Pettit Spring 2024

Readings in CLRS 4th edition:

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

Announcements

- PS1 due tomorrow
- PA1 available, Gradescope submission coming later this week
- Office hours
 - Prof Hott Office Hours: Mondays 11a-12p, Fridays 10-11a and 2-3p
 - Prof Pettit Office Hours: Mondays and Wednesdays 2:30-4:00p
 - TA office hours posted on 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 <u>order</u> 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 <u>**G**</u>
- single start vertex <u>s</u>

Output:

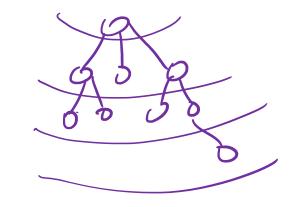
- Distance from <u>s</u> to each node in <u>G</u> (distance = number of edges)
- Breadth-First Tree of $\underline{\mathbf{G}}$ with root $\underline{\mathbf{s}}$

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

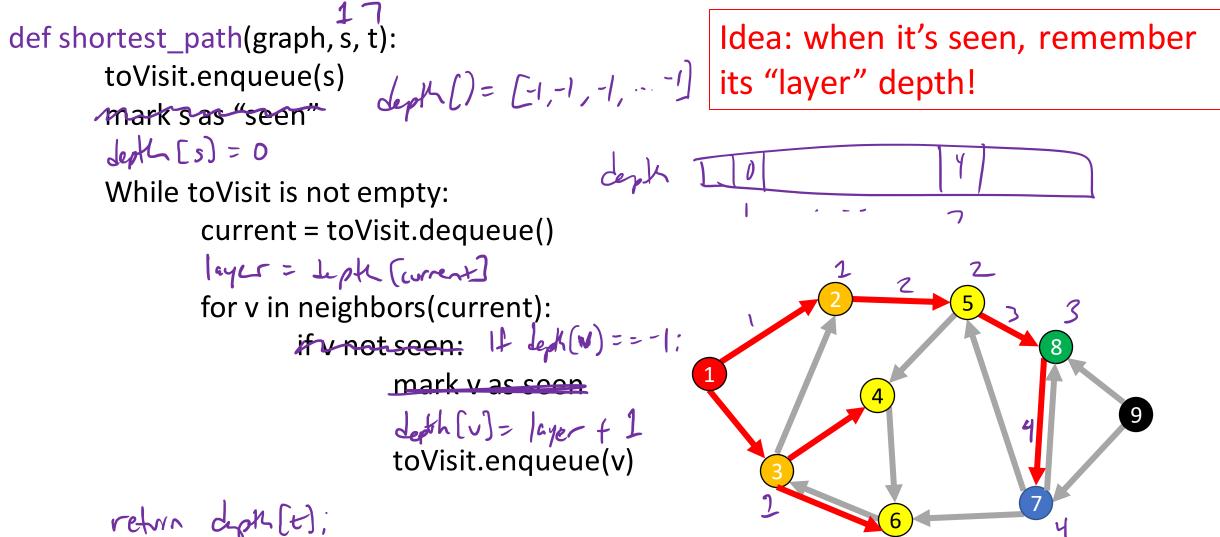
• But edge weight's (if any) not used, so "short" is in terms of number of edges in path



BFS

to Visit Q: X X X X X X X X X 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) 9 Secr Se 8

BFS: Shortest Path



BFS: Shortest Path

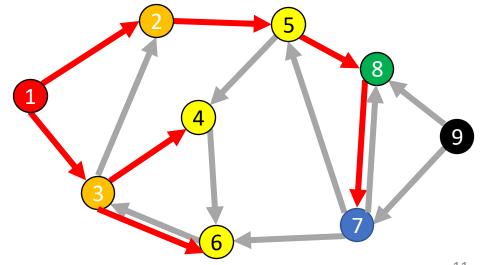
```
def shortest_path(graph, s, t):
                                                     Idea: when it's seen, remember
                                                    its "layer" depth!
      toVisit.enqueue(s)
      depth[s] = 0
                                                                  dept([5] = 2 (dept)
parent[5] = 2 (node)
      While toVisit is not empty:
             current = toVisit.dequeue()
             layer = depth [current]
                                                                       2
             for v in neighbors(current):
                    if v does not have a depth:
                           depth[v]=layer+1
                           toVisit.enqueue(v)
      return depth[t]
```

9

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)

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

From CLRS

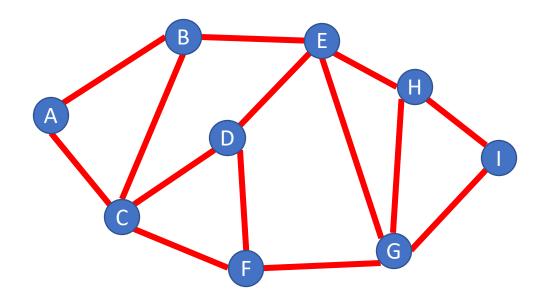
Vertices here have some properties:

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

Color meanings here:

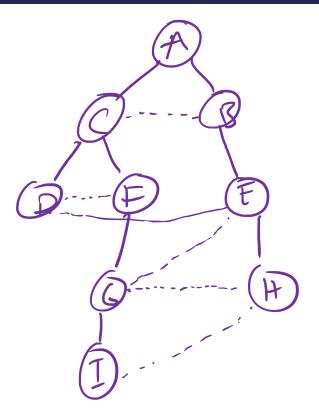
- White: haven't seen this vertex yet
- Gray: vertex has been seen and added to the queue for processing later
- Black: vertex has been removed from queue and its neighbors seen and added to the queue

Tree View of BFS Search Results

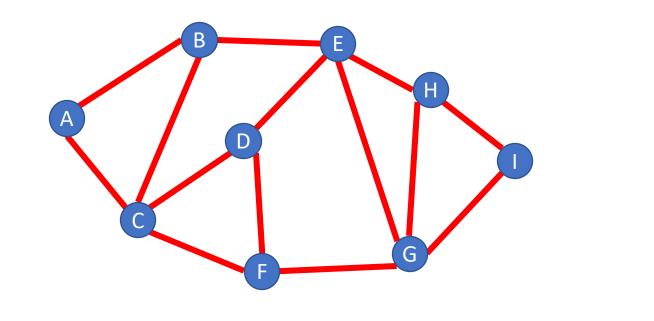


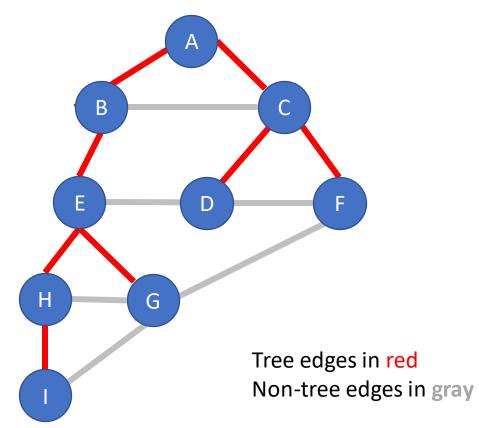
Draw BFS tree starting at A





Tree View of BFS Search Results





Analysis for Breadth-first search

For a graph having V vertices and E edges

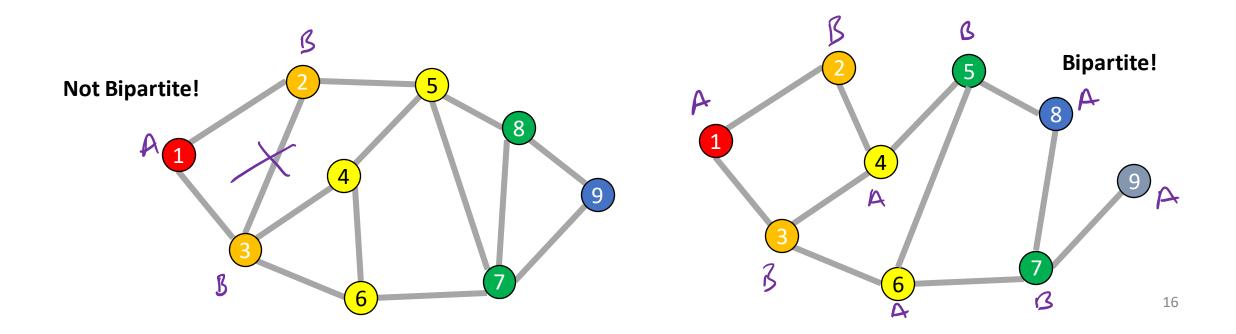
- Each edge is processed once in the while loop for a cost of $\Theta(E)$
- Each vertex is put into the queue once and removed from the queue and processed once, for a cost $\Theta(V)$
 - Also, cost of initializing colors or depth arrays is $\Theta(V)$
- Total **time-complexity**: $\Theta(V + E)$
 -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

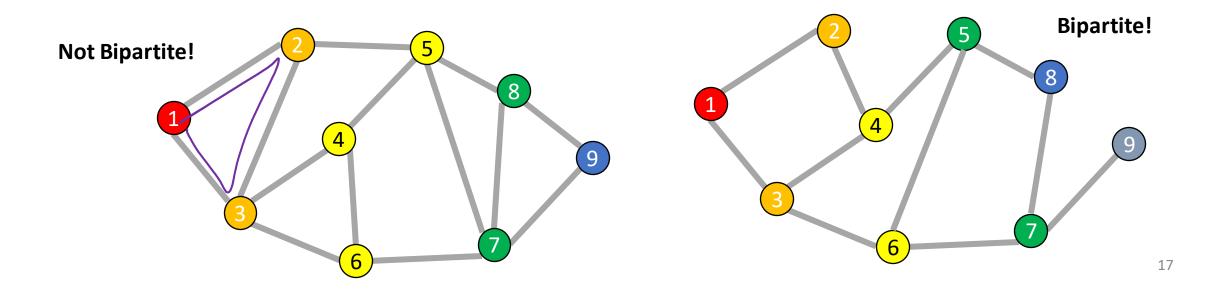
A (undirected) graph is **Bipartite** provided every vertex can be assigned to one of two teams such that every edge "crosses" teams

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



Odd Length Cycles

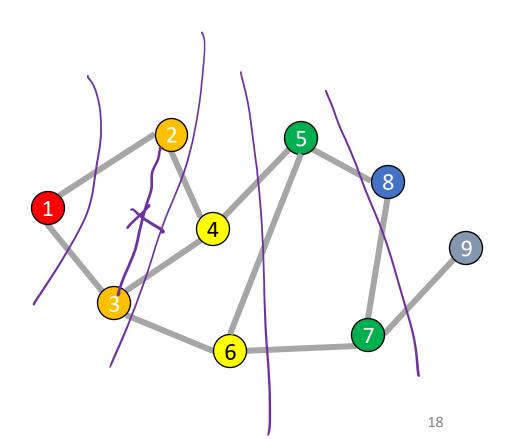
A graph is bipartite if and only if it has no odd length cycles



BFS: Bipartite Graph?

```
def bfs(graph, s):
       toVisit.enqueue(s)
       depth[s] = 0
       depth = [-1, -1, -1, ...] # Length matches |V|
       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)
```

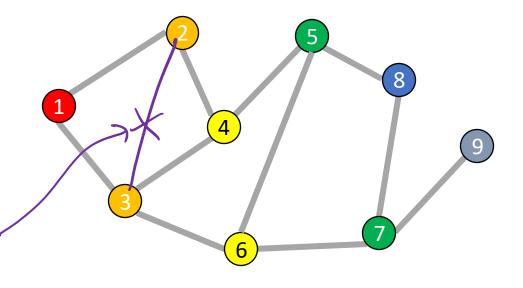
Idea: Check for edges in the same layer!



BFS: Bipartite Graph?

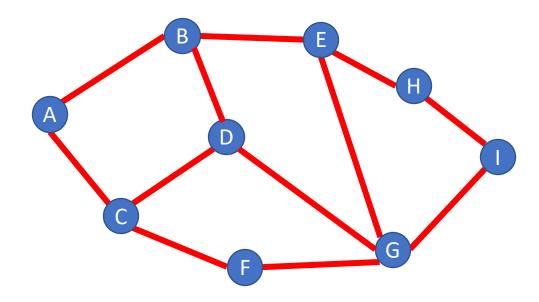
```
def bfs(graph, s):
       toVisit.enqueue(s)
       depth[s] = 0
       depth = [-1, -1, -1, ...] # Length matches |V|
       While toVisit is not empty:
               current = toVisit.dequeue()
               layer = depth [current]
               for v in neighbors(current):
                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

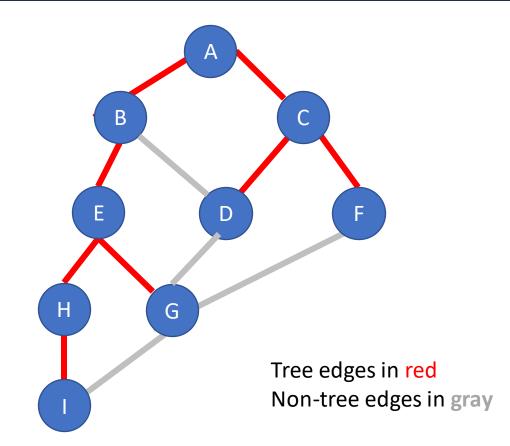
Idea: Check for edges in the same layer!



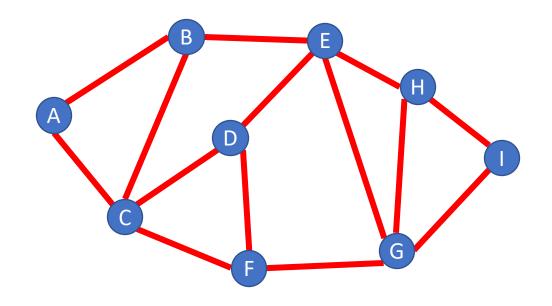
return True

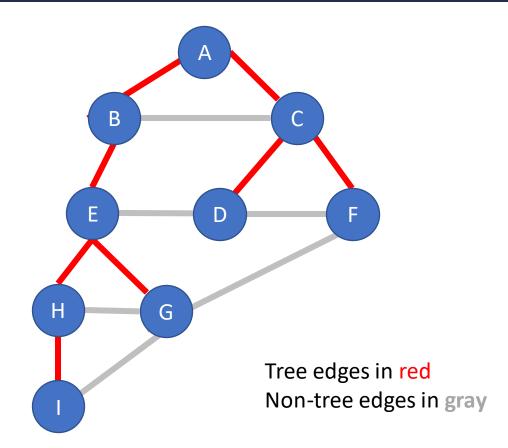
BFS Tree for a Bipartite Graph





BFS Tree for a Non-Bipartite Graph





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

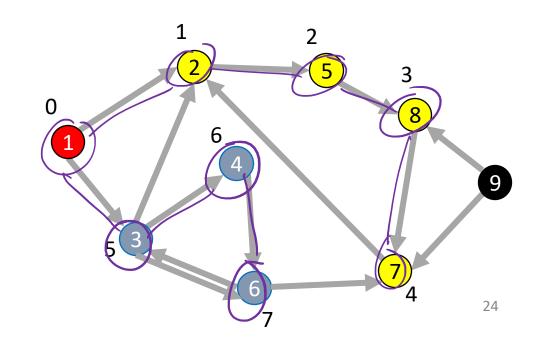
Depth-First Search

Input: a node s , Caph G

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

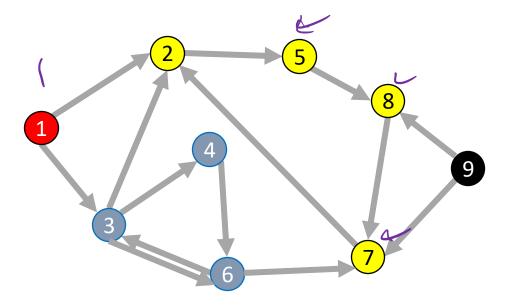
Output:

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



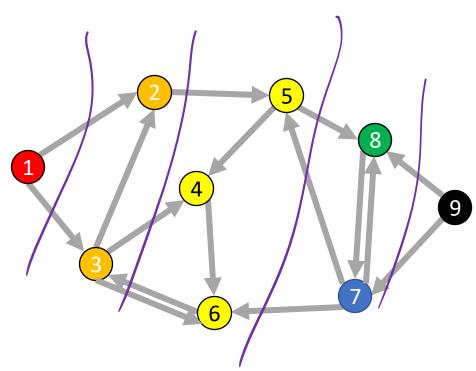
DFS: Non-recursively (less common)

```
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

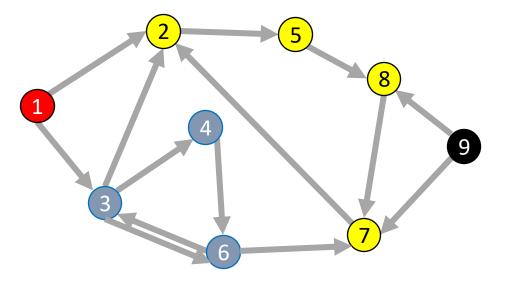
```
def bfs(graph, s):
      toVisit.enqueue(s)
      mark s as "seen"
      While toVisit is not empty:
             current = toVisit.dequeue()
             for v in neighbors(current):
                   if v not seen:
                          mark v as seen
                          toVisit.enqueue(v)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



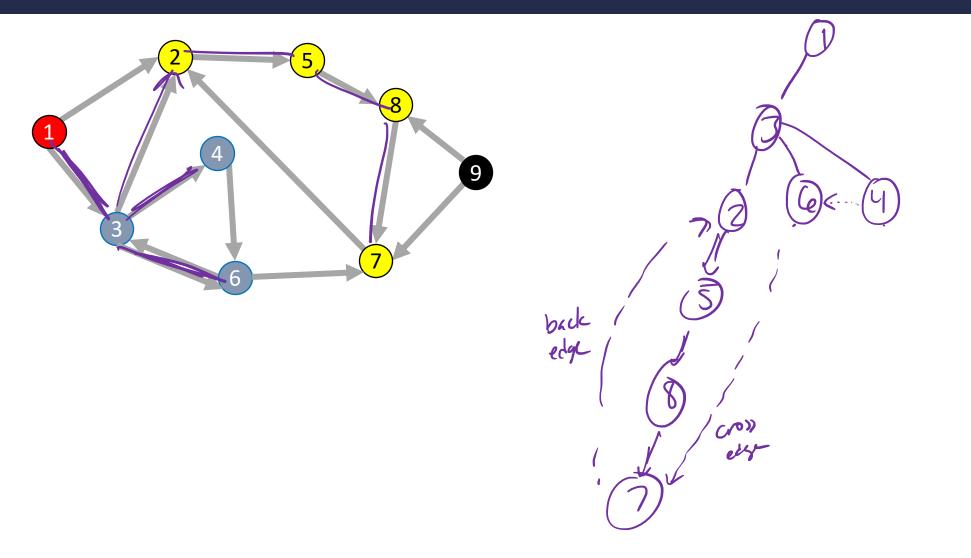
DFS: Recursively

```
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



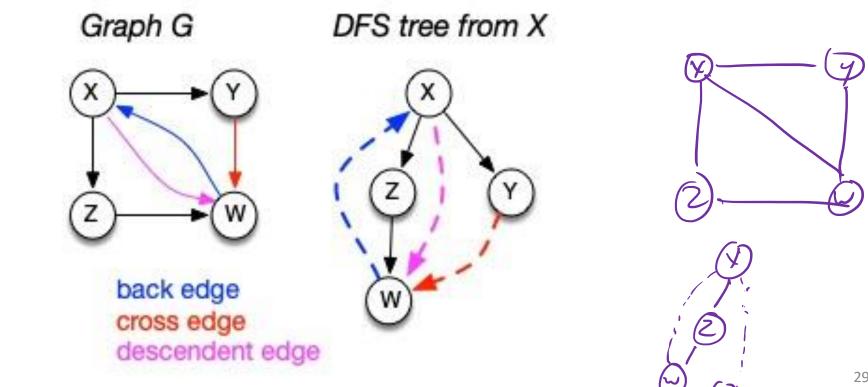
View of DFS Results as a Tree



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?

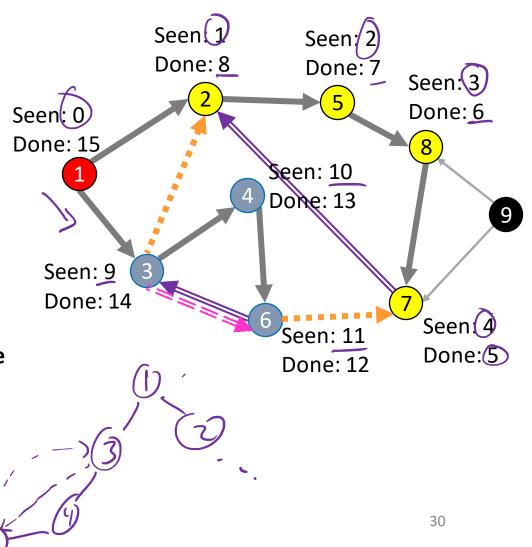


Using DFS

Consider the "seen times" and "done times"

Edges can be categorized:

- Tree Edge
 - (*a*, *b*) was followed when pushing
 - (*a*, *b*) when *b* was **unseen** when we were at *a*
- Back Edge
 - (*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 ====→
 - (*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
 - (*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



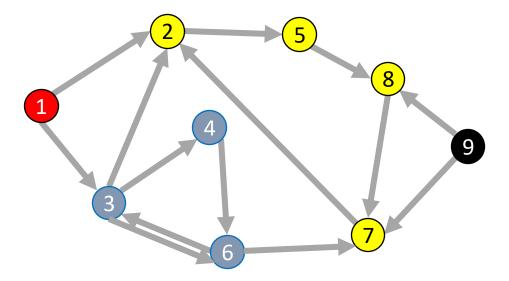
DFS: Cycle Detection

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

Idea: Look for a back edge!



DFS: Cycle Detection

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)

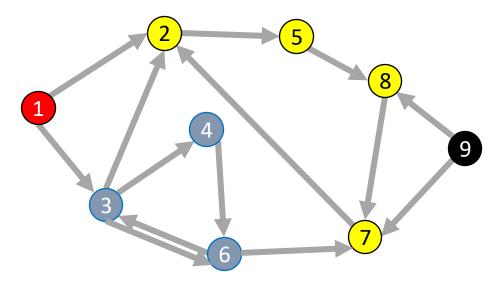
mark curr as seen for v in neighbors(current):

if v not seen:

dfs_rec(graph, v, seen, done)

mark curr as done

Idea: Look for a back edge!



DFS: Cycle Detection

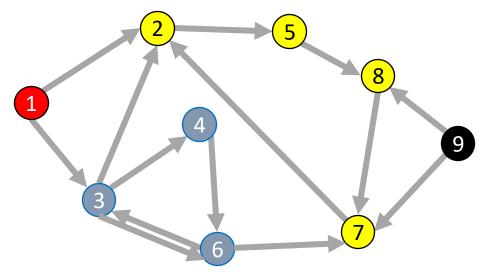
def hasCycle(graph, s):

```
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
```

Idea: Look for a back edge!



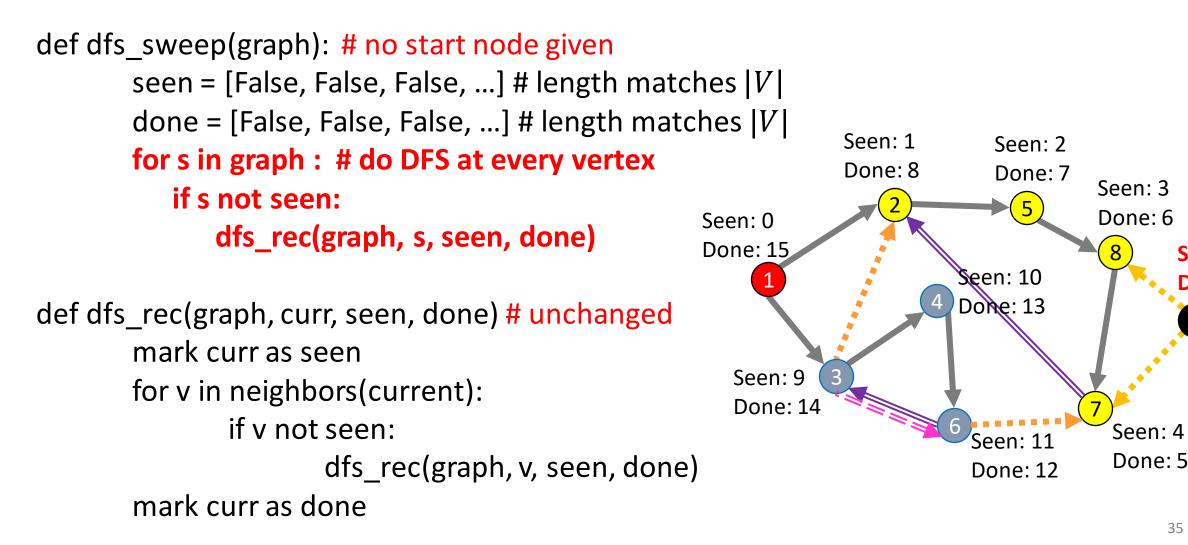
Back Edges in Undirected Graphs

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?

DFS "Sweep" to Process All Nodes



Seen: 16

Done: 17

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 $\Theta(V)$