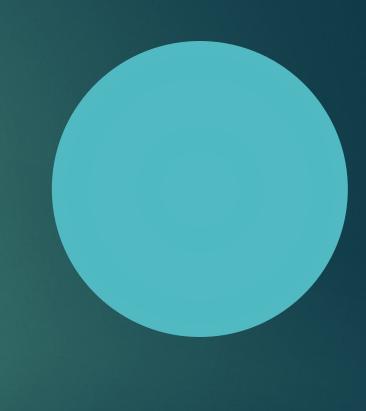
DEPTH FIRST SEARCH



Depth-first search

- Depth-first search is a widely used graph traversal algorithm besides breadthfirst search
- It was investigated as strategy for solving mazes by Trémaux in the 19th century
- It explores as tar as possible along each branch before backtracking // BFS was a layer-by-layer algorithm
- Time complexity of traversing a graph with DFS: O(V+E)
- Memory complexity a bit better than that of BFS !!!
- By itself the DFS isn't all that useful, but when augmented to perform other tasks such as count connected components, determine connectivity, or find bridges/articulation points then DFS really shines.

Depth-first search

dfs(vertex)

vertex set visited true print vertex

for v in vertex neighbours
if v is not visited
dfs(v)

dfs(vertex)

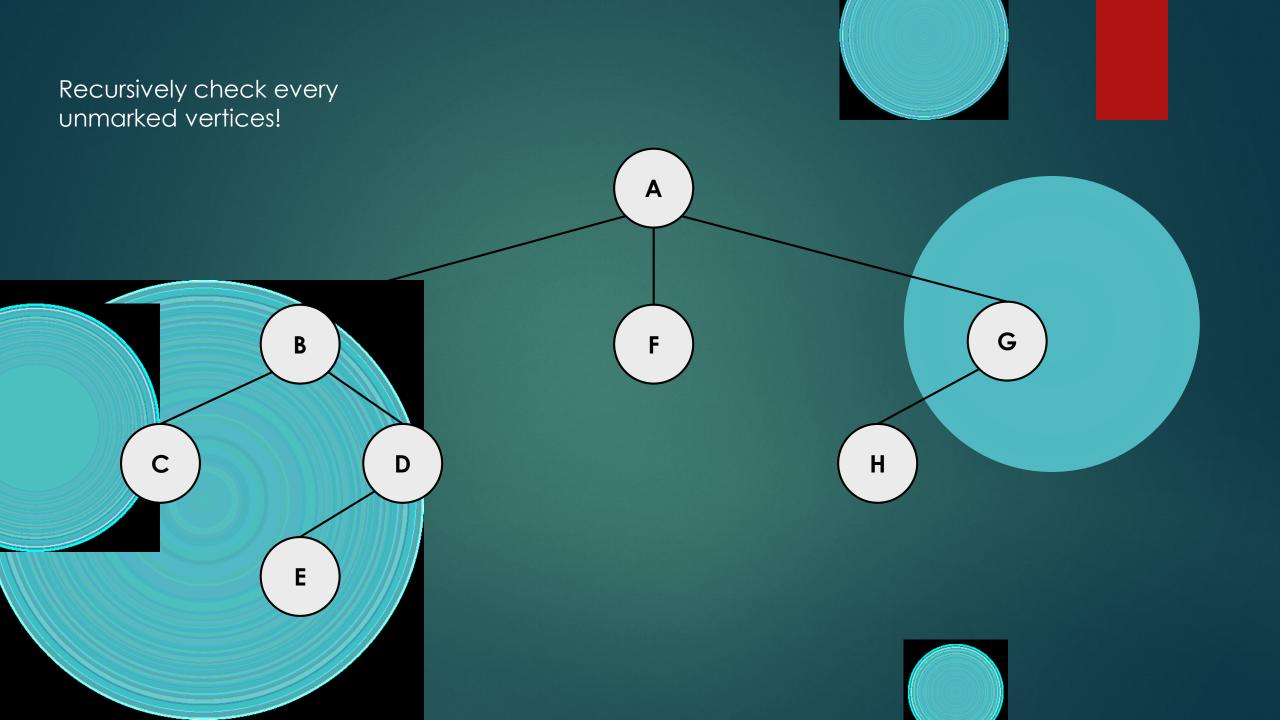
Stack stack vertex set visited true stack.push(vertex)

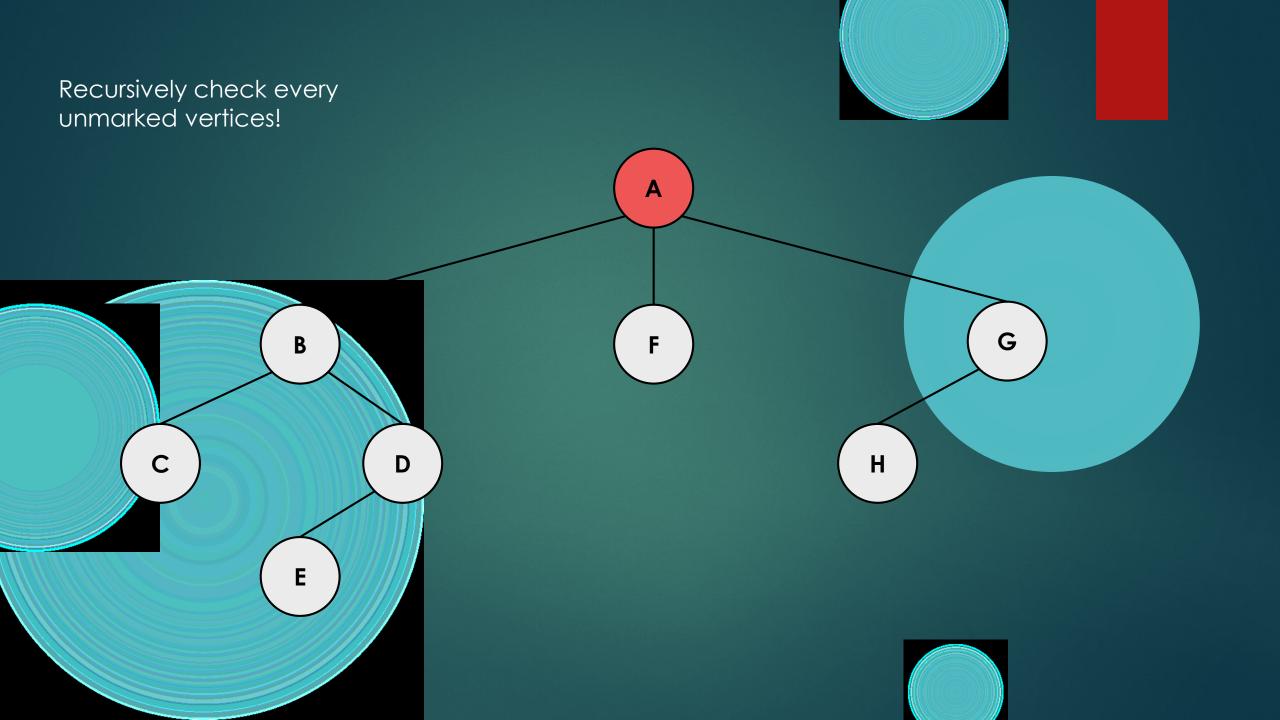
while stack not empty actual = stack.pop()

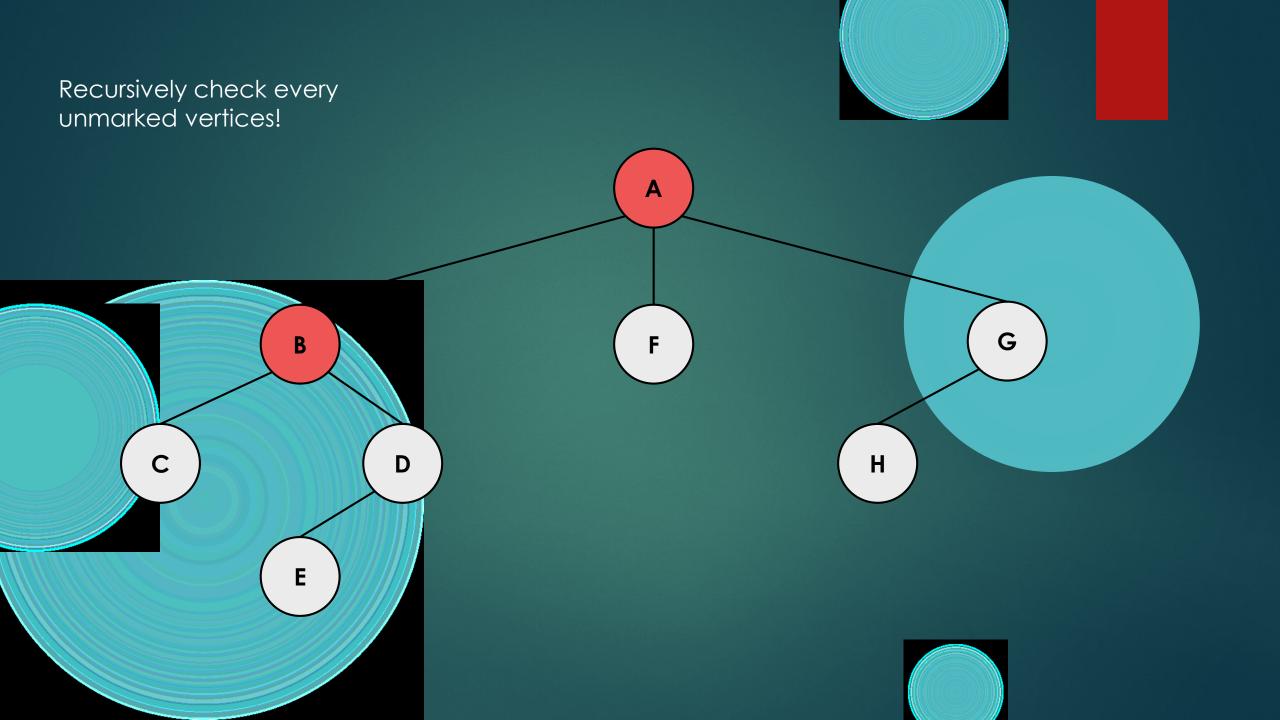
for v in actual neighbours
if v is not visited
v set visited true
stack.push(v)

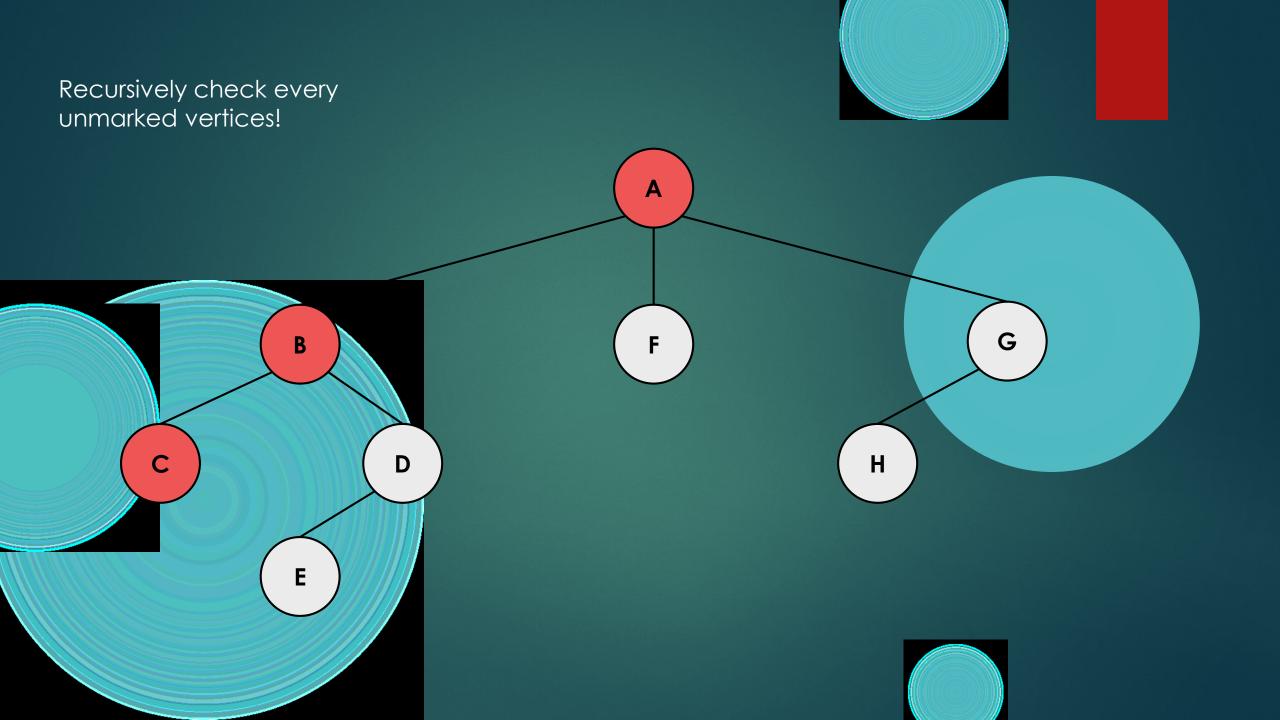
RECURSION

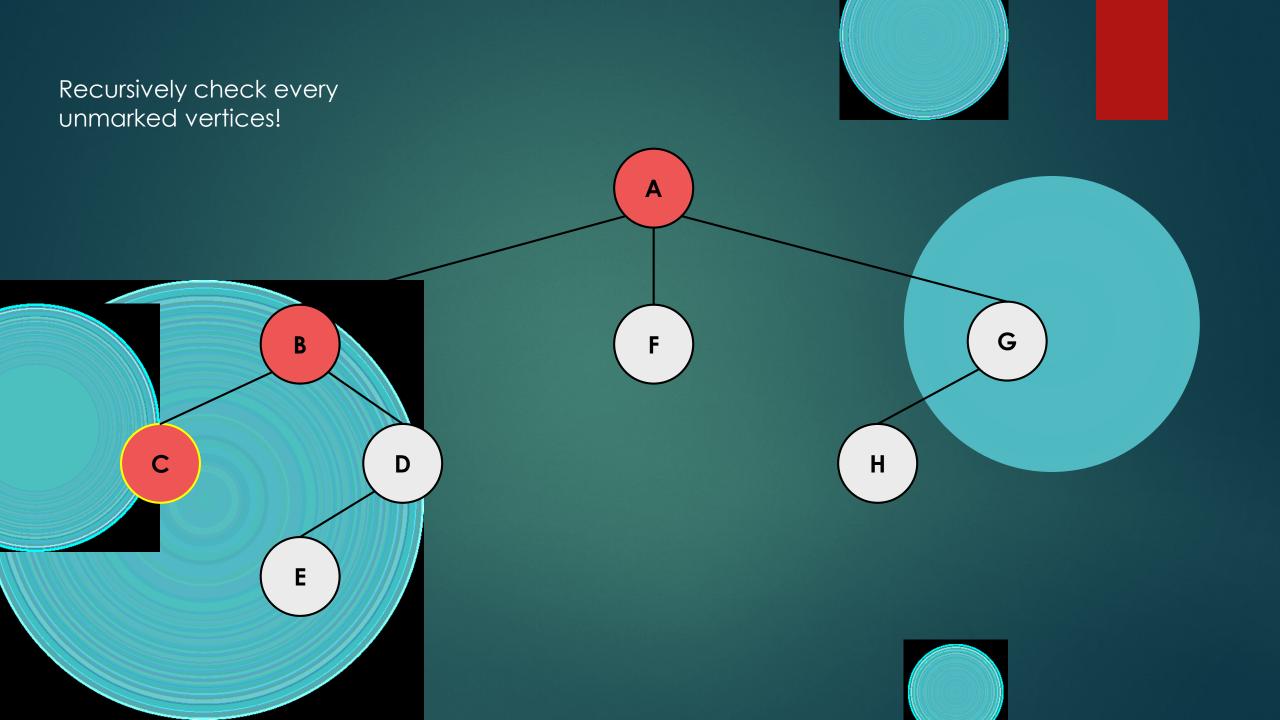
ITERATION

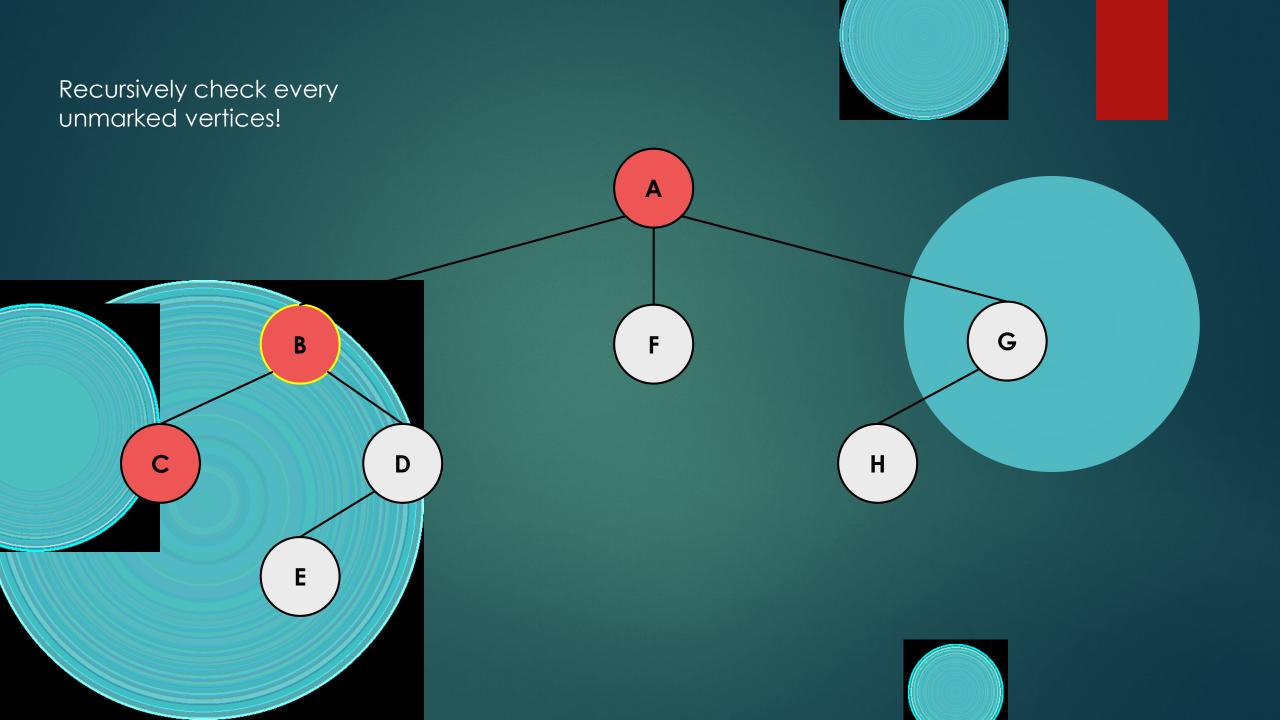


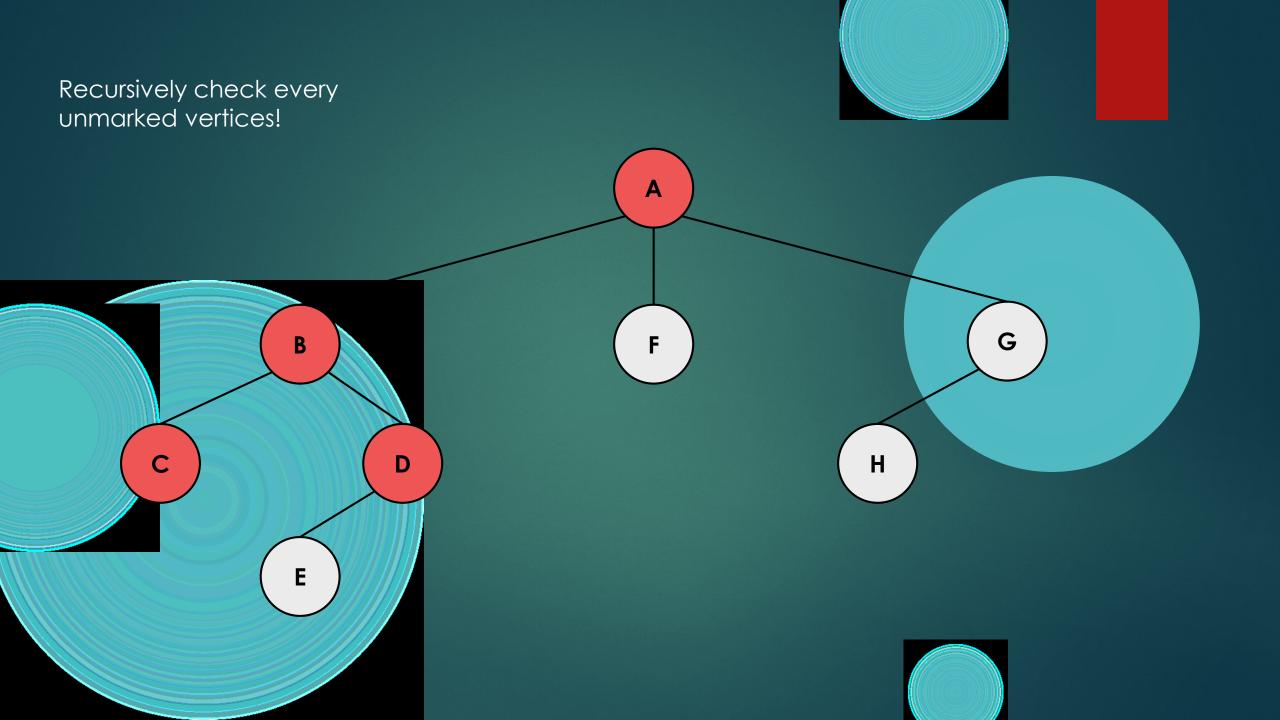


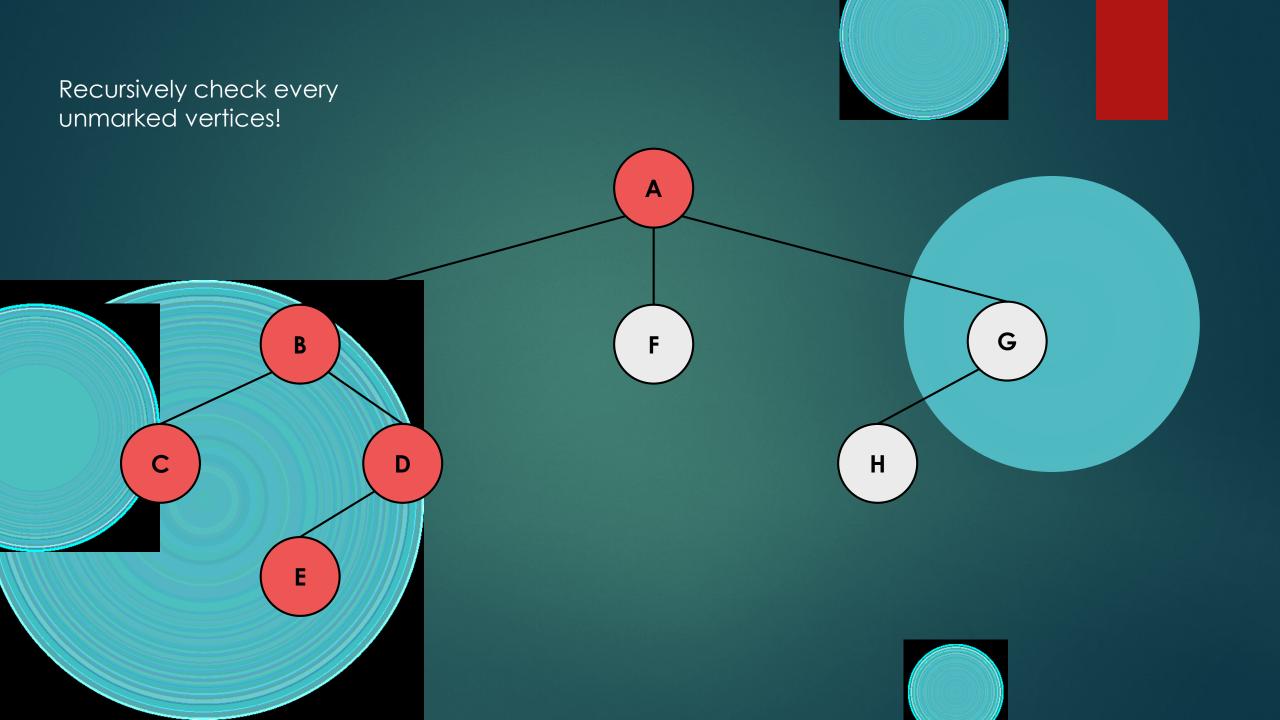


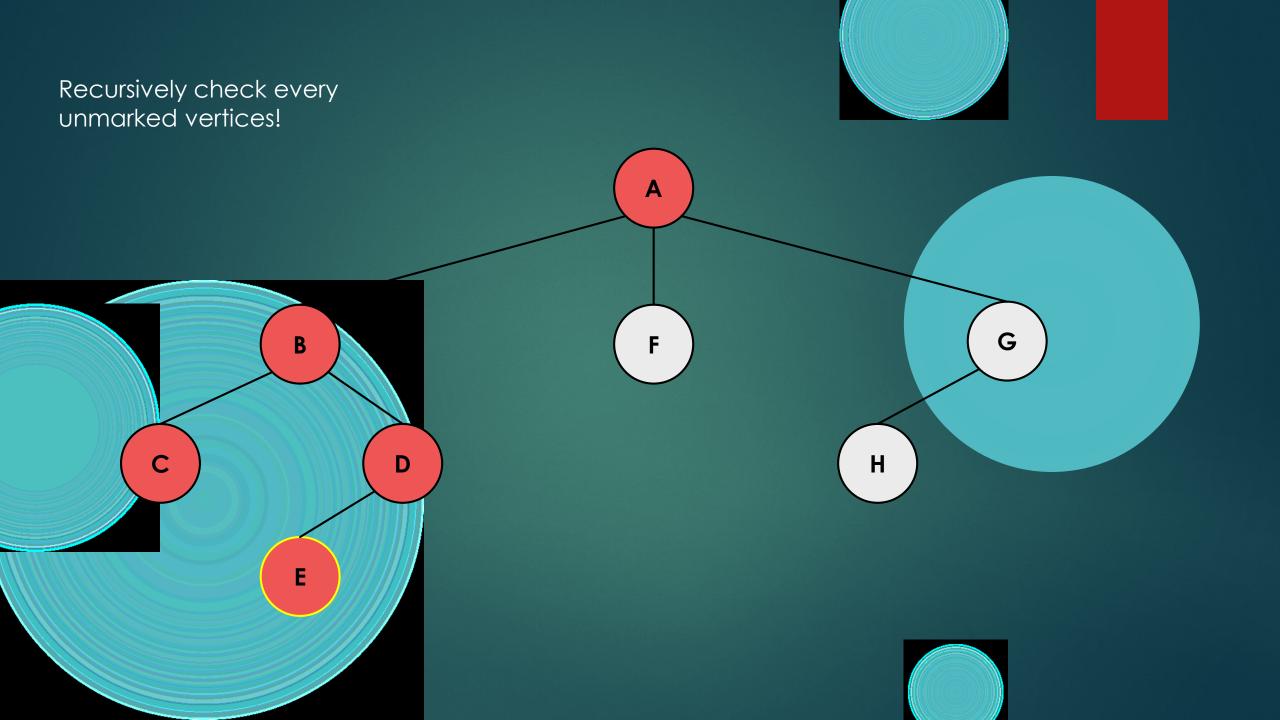


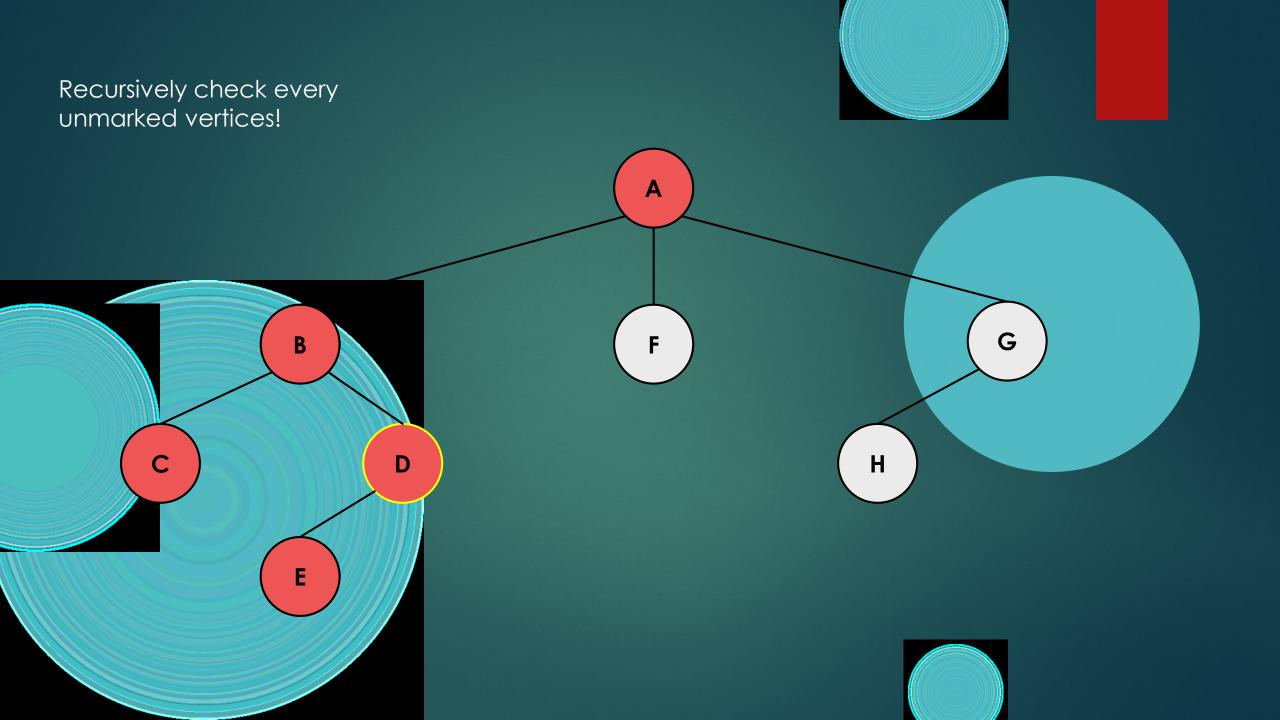


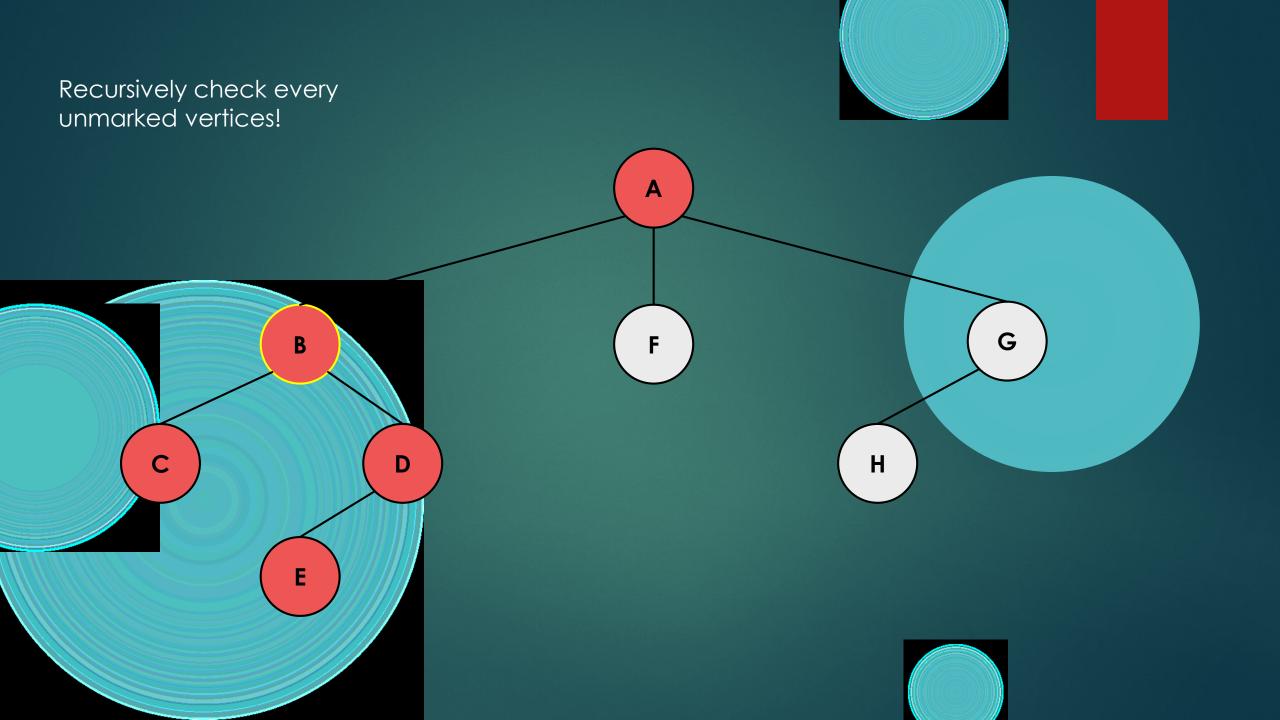


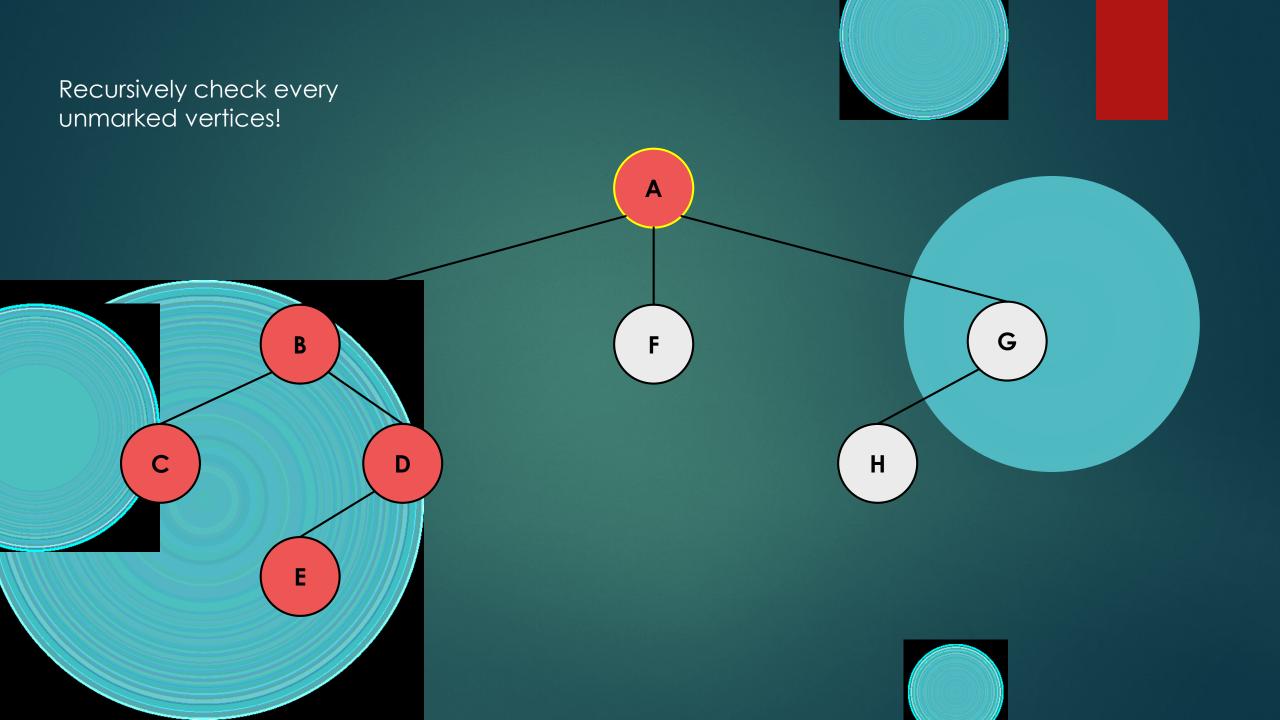


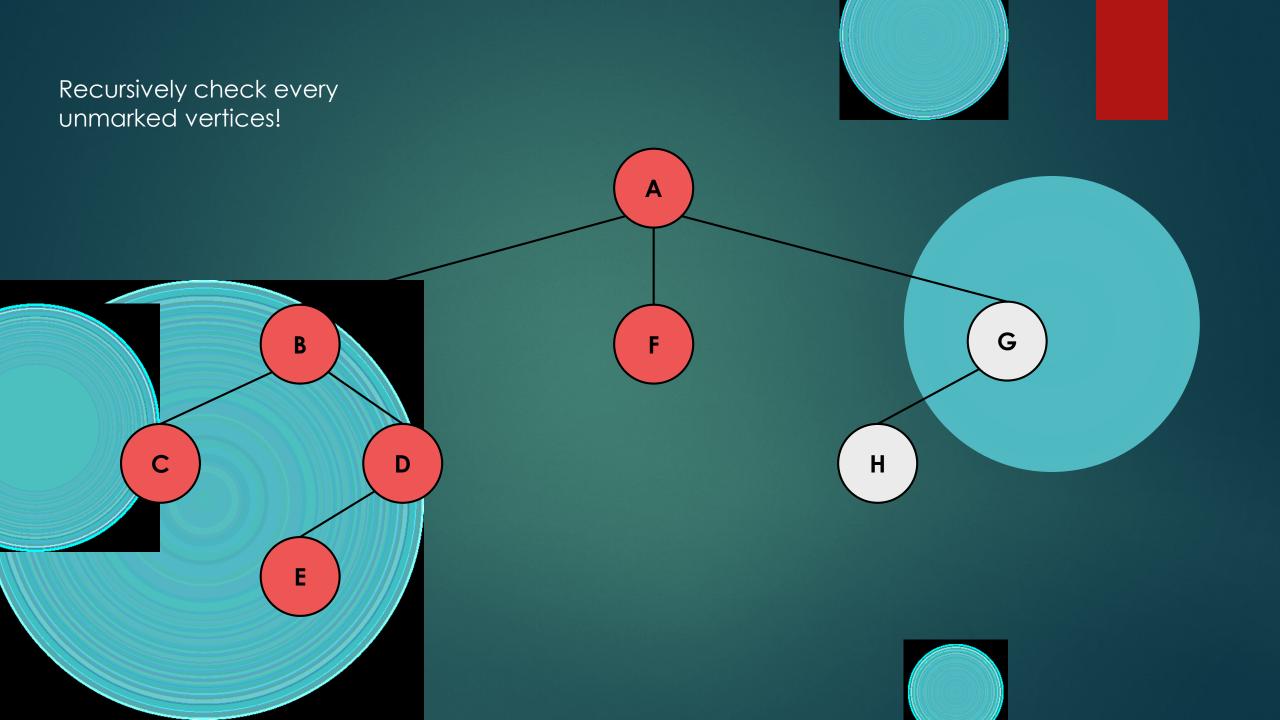


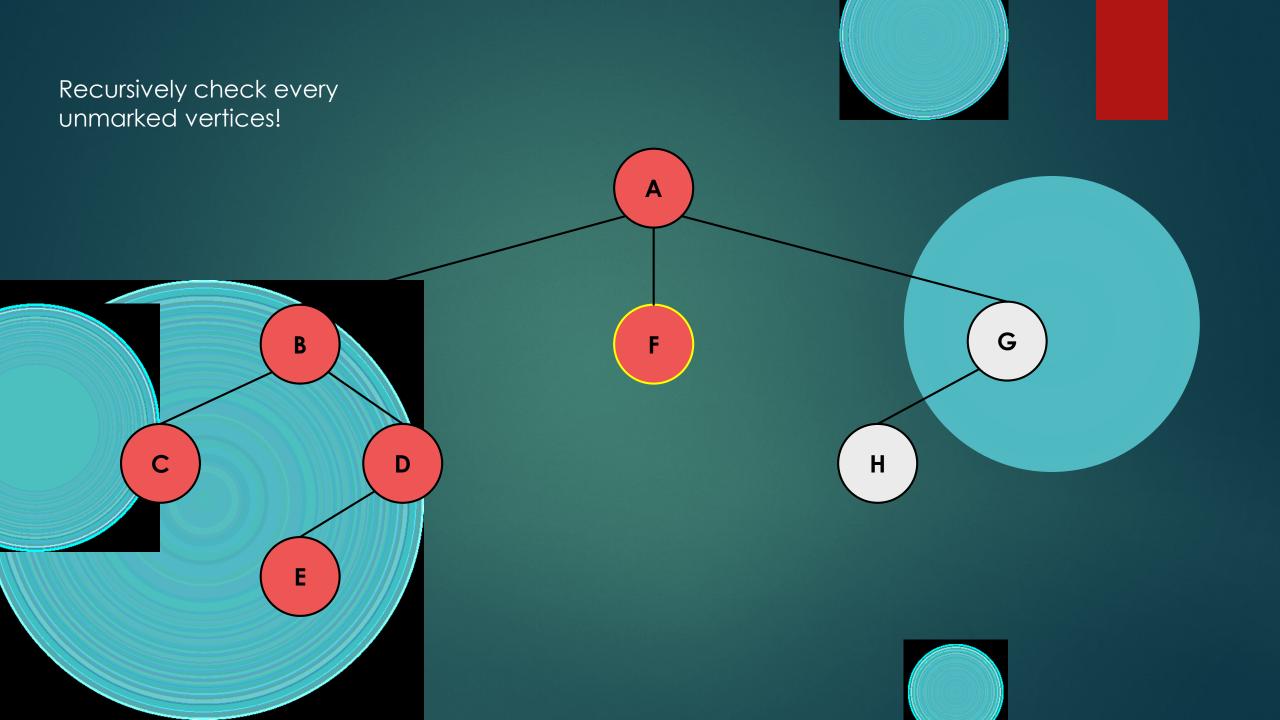


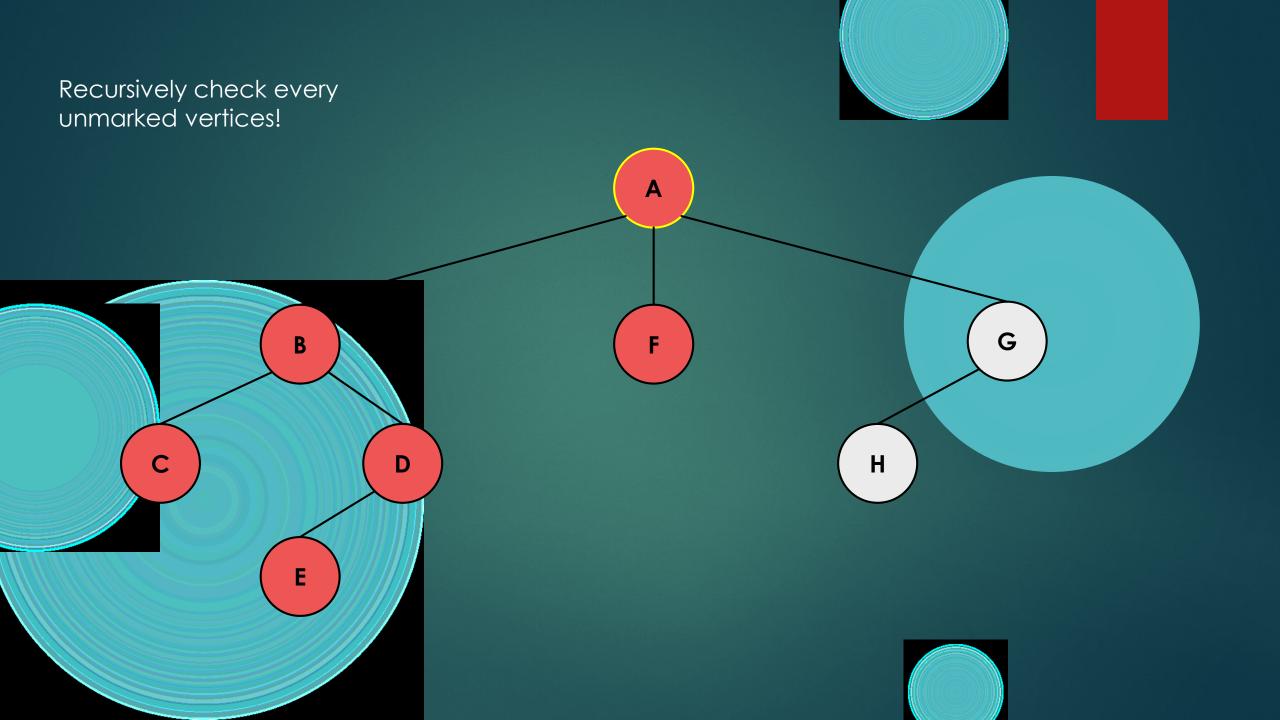


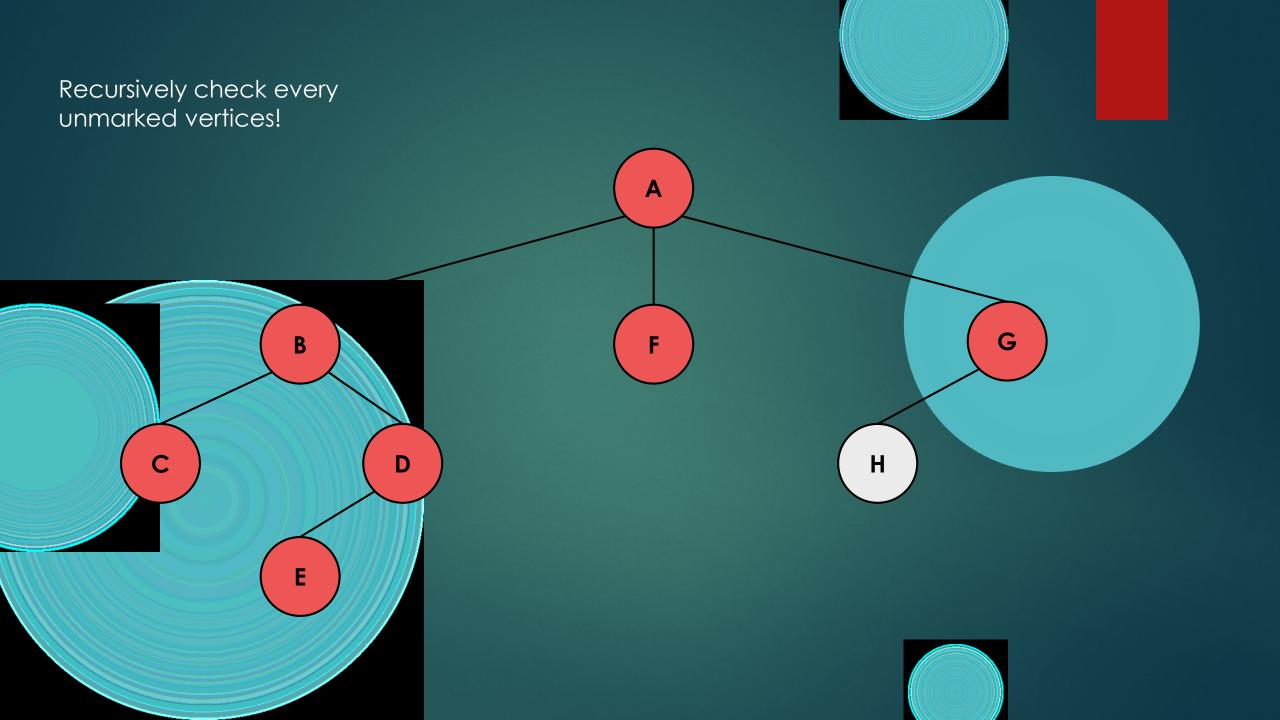


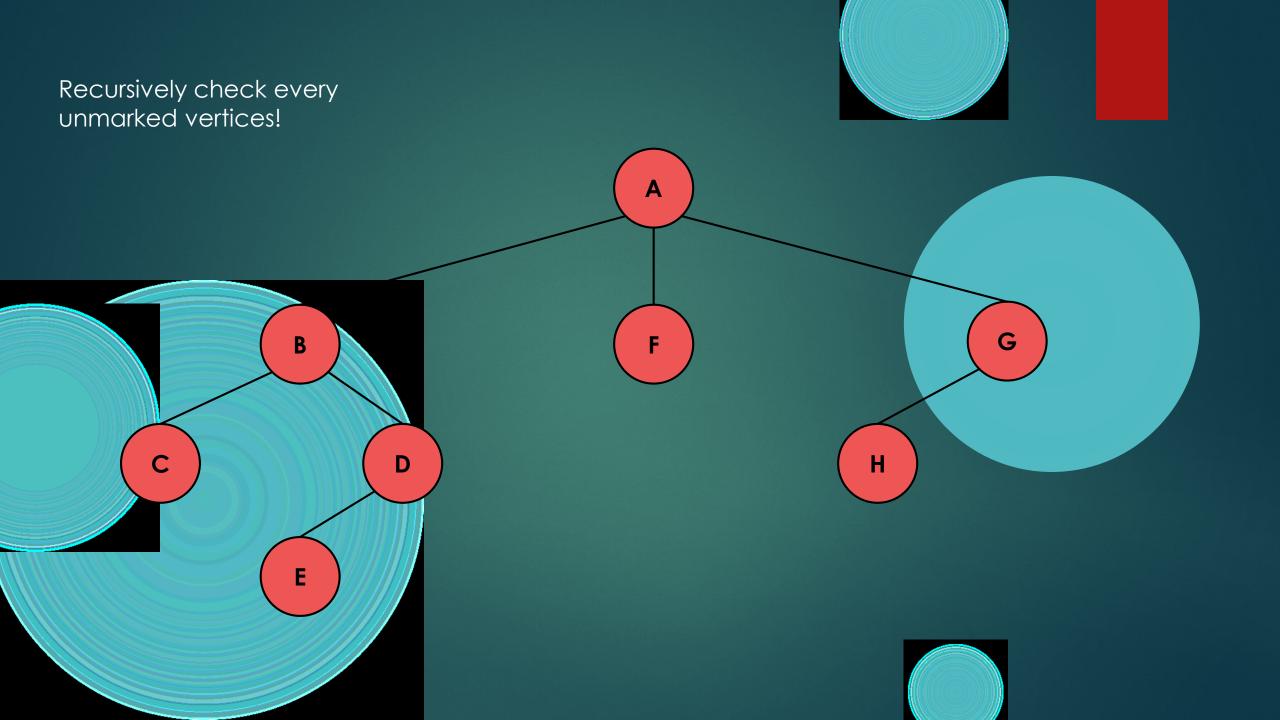


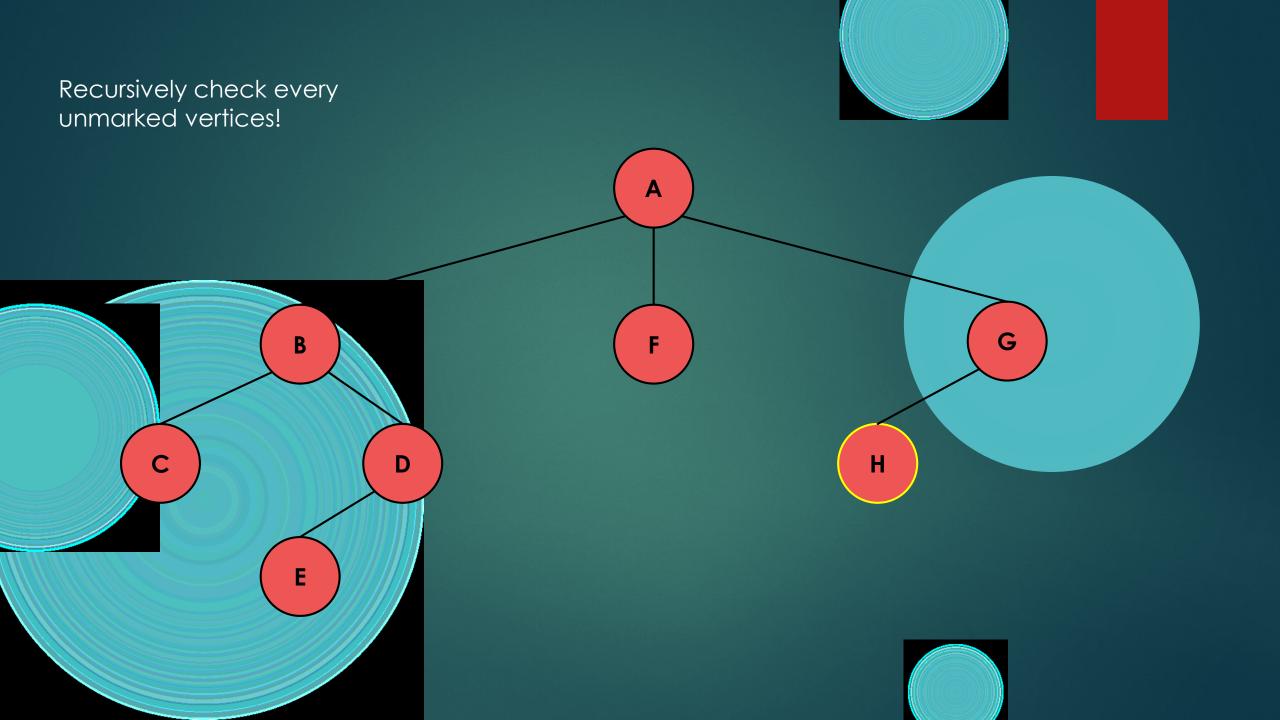


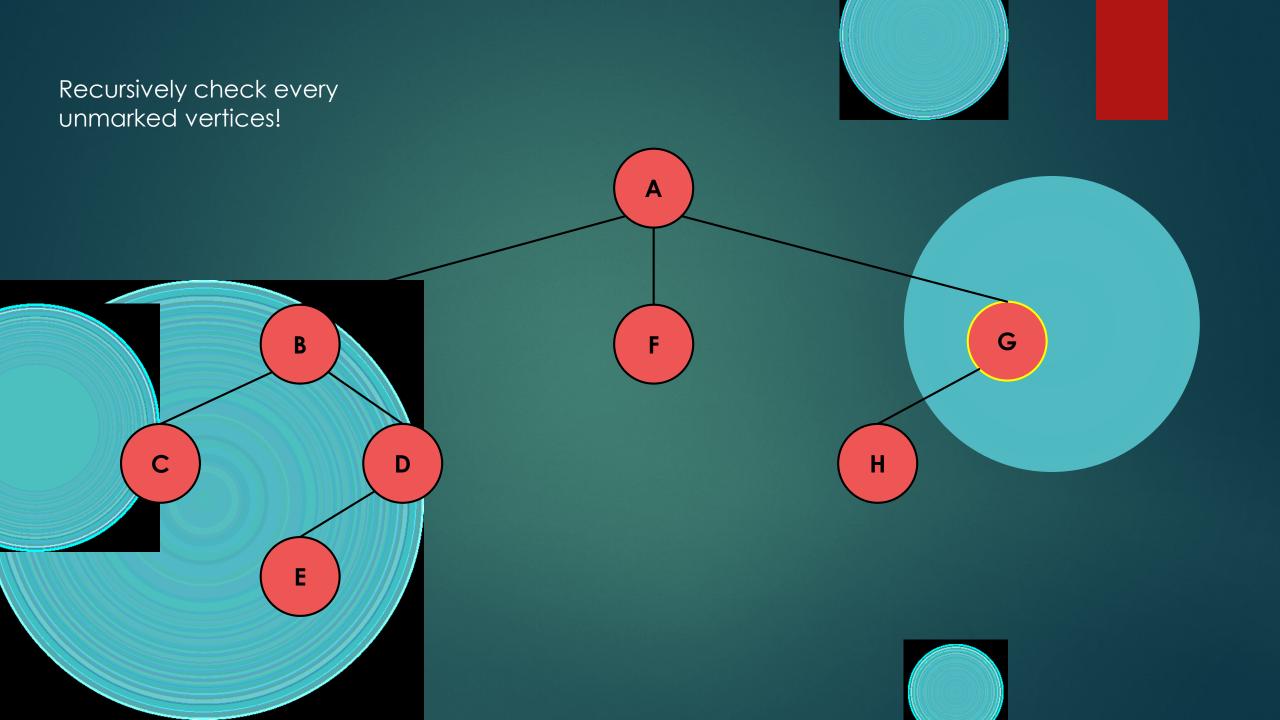


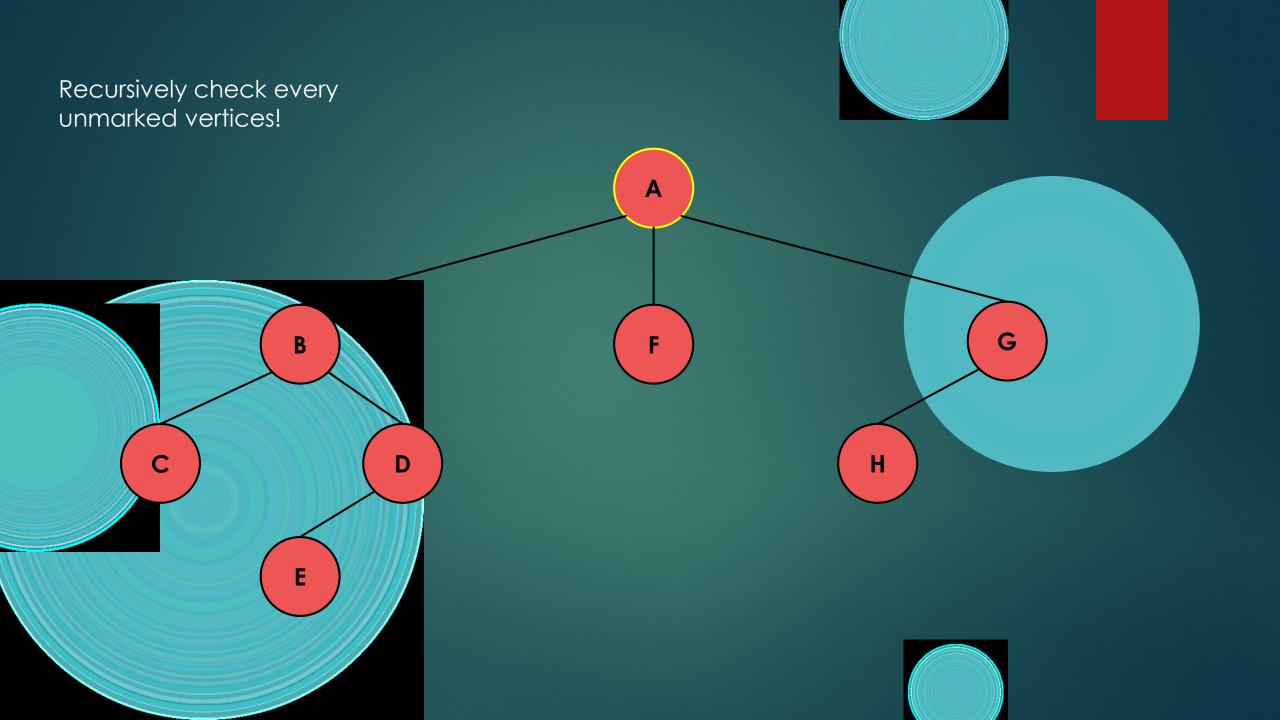


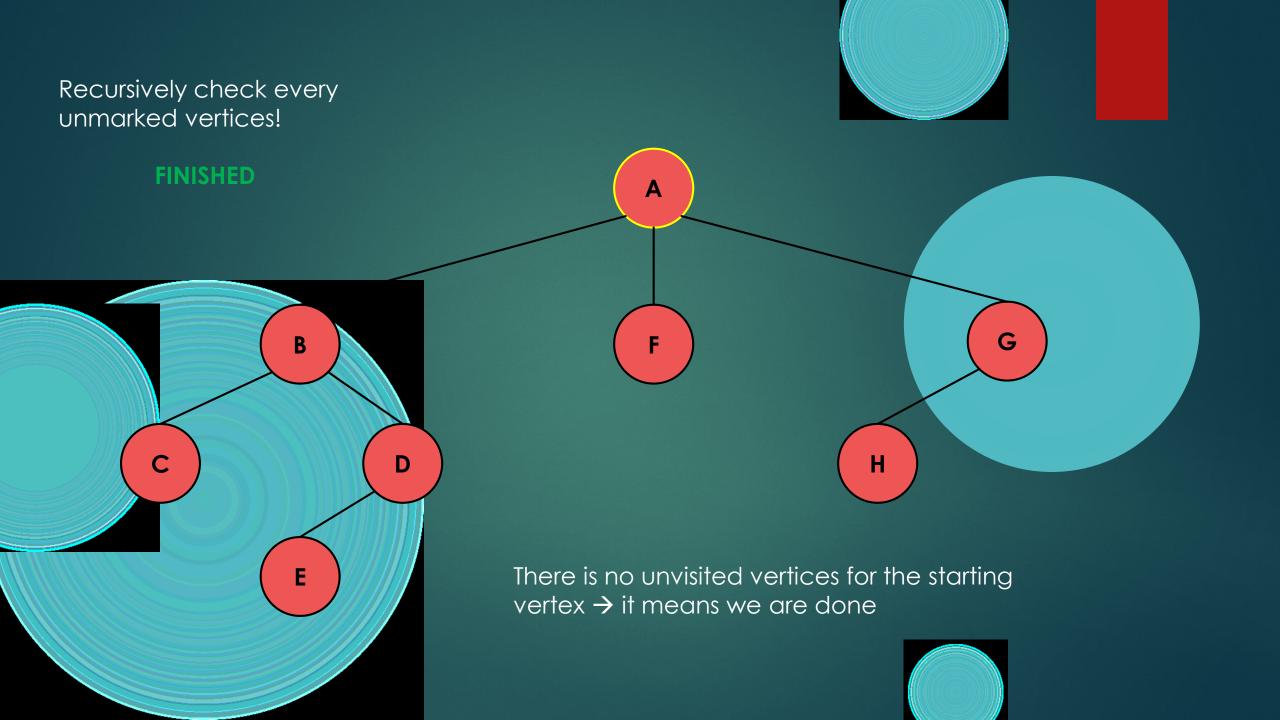






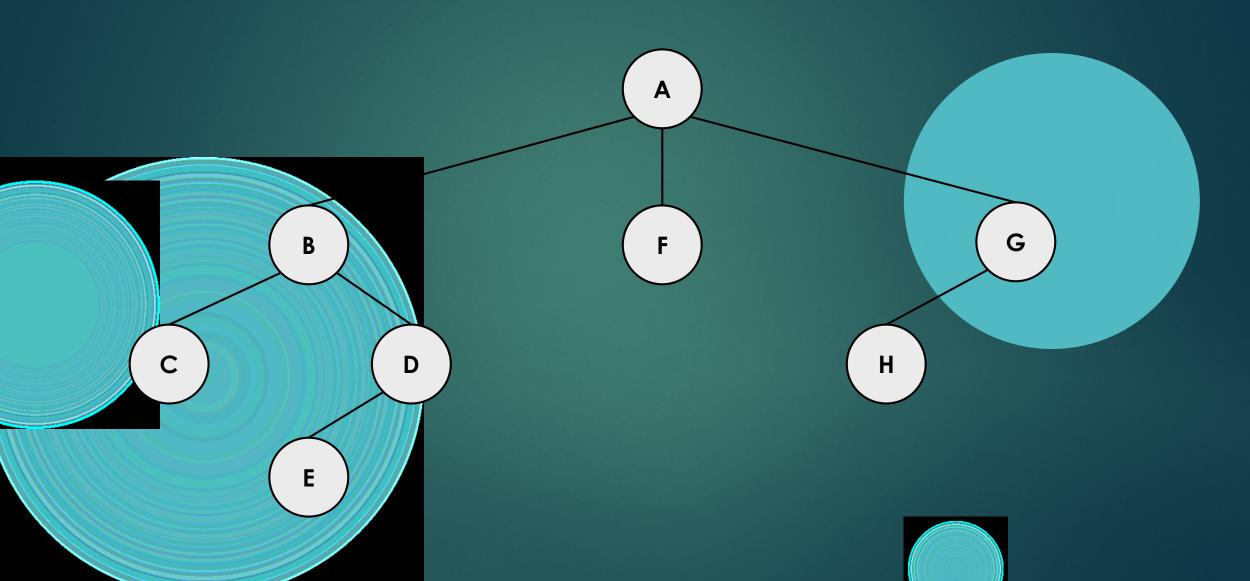


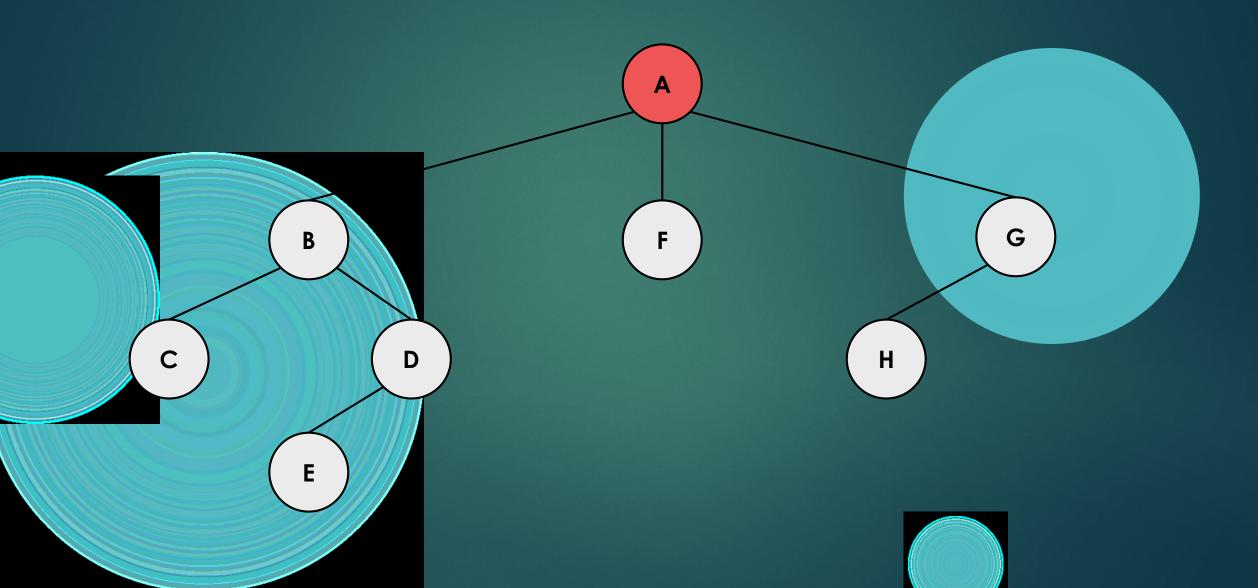


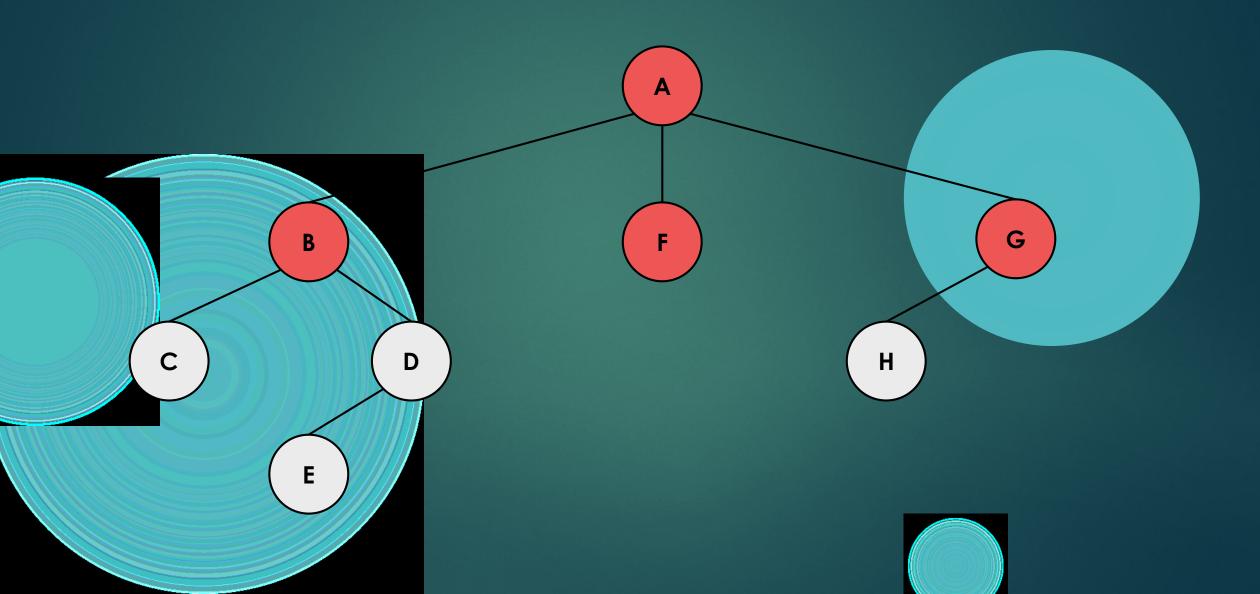


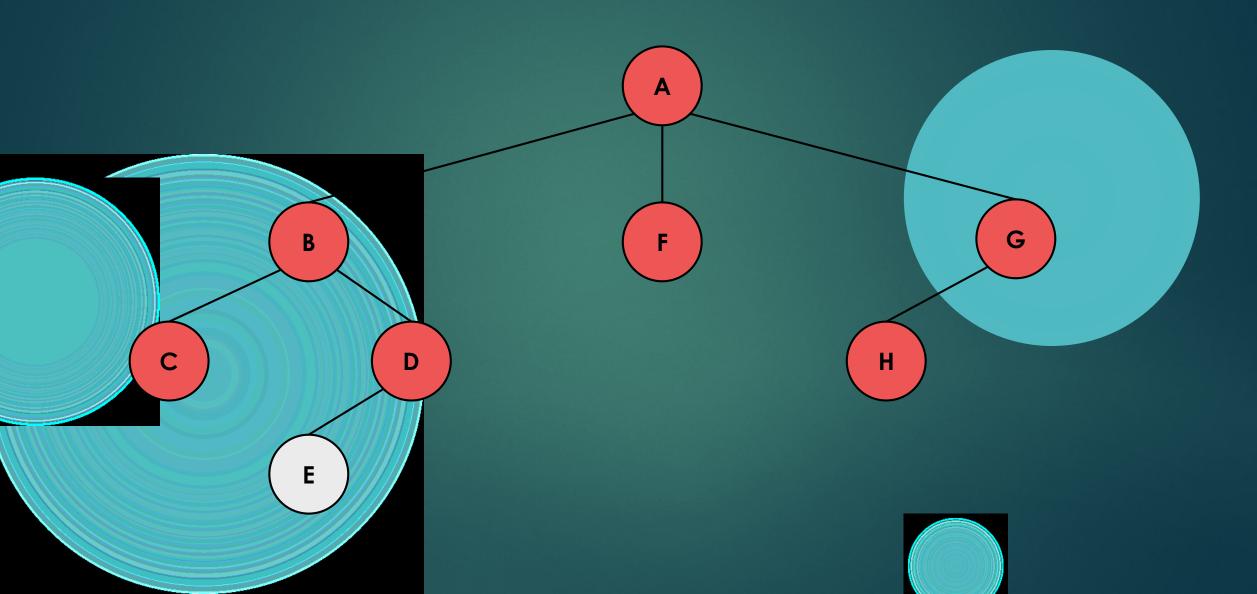
<u>Applications</u>

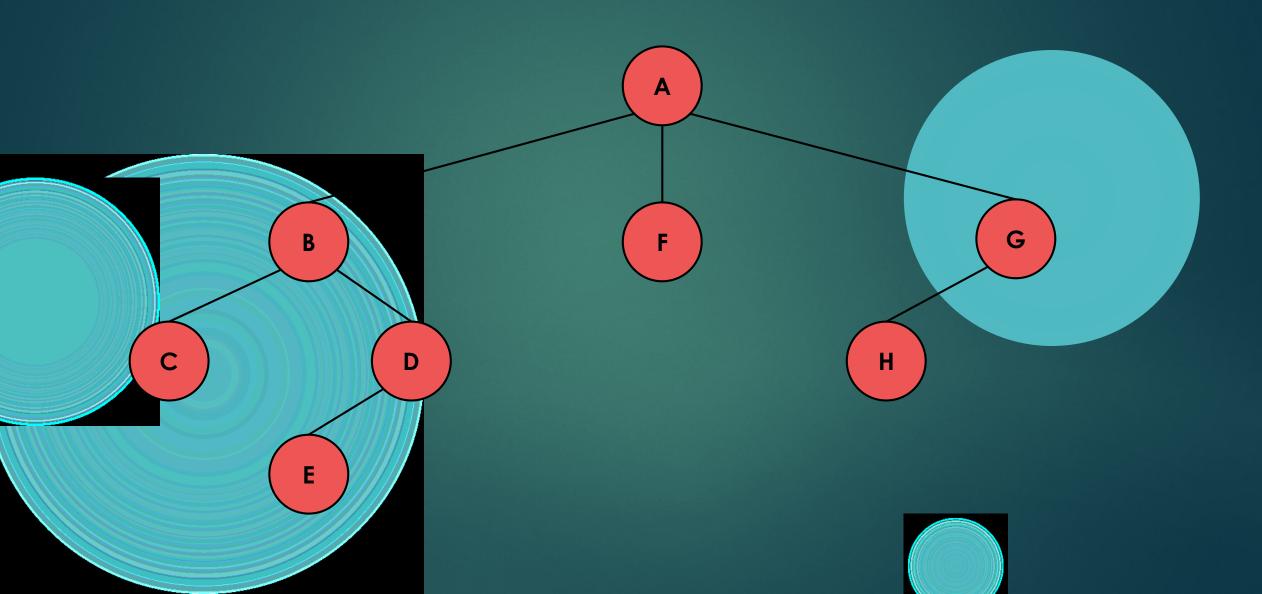
- Topological ordering
- Kosaraju algorithm for finding strongly connected components in a graph which can be proved to be very important in recommendation systems (youtube)
- Detecting cycles (checking whether a graph is a DAG or not)
 - ▶ Processes waiting for each other → this is a cycle
 - Generating mazes OR finding way out of a maze

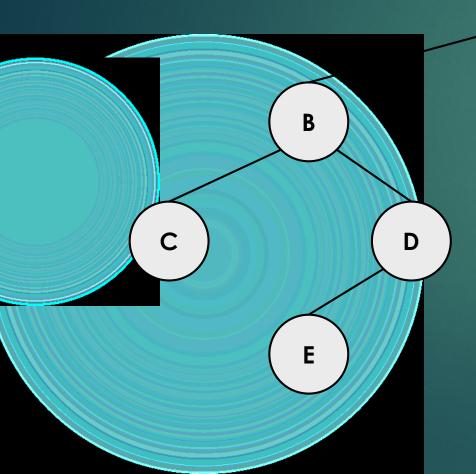


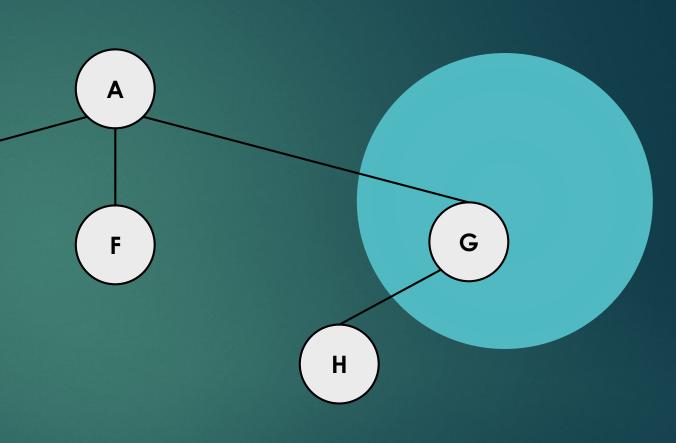


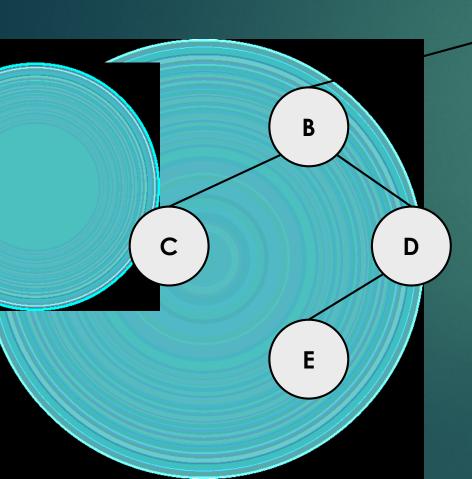


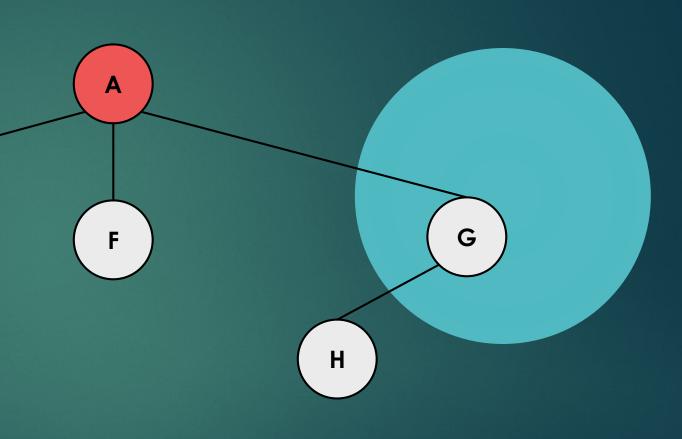


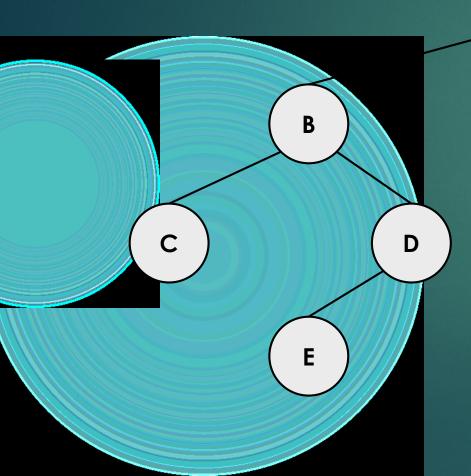


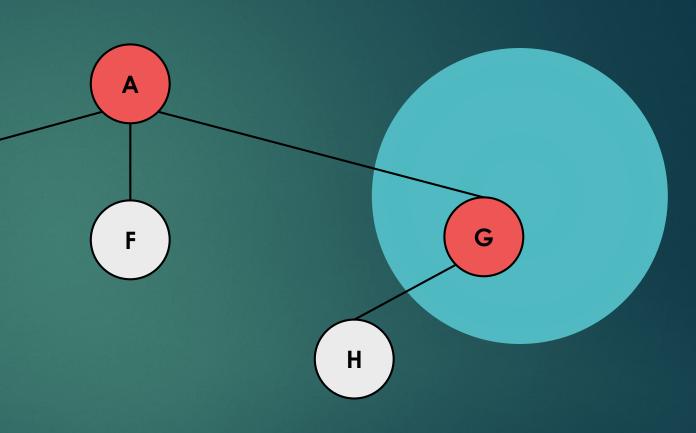


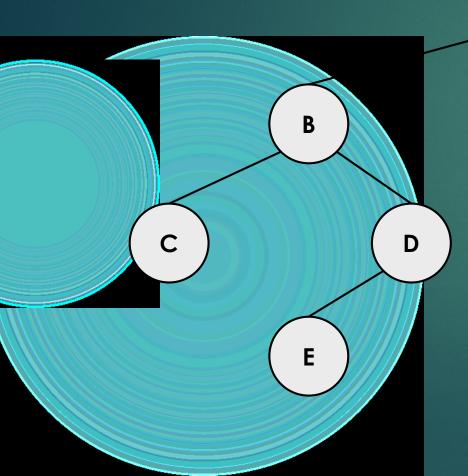


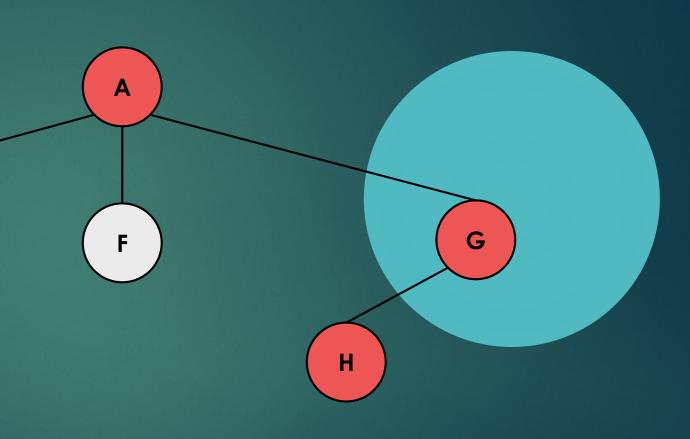


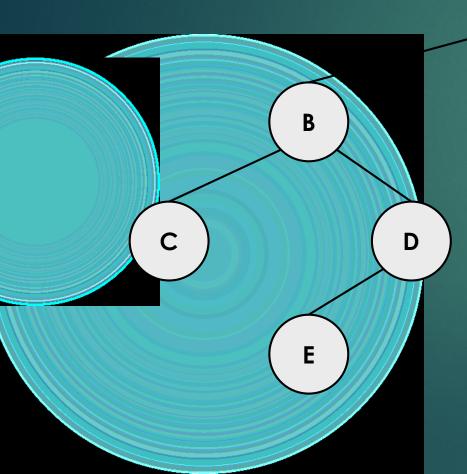


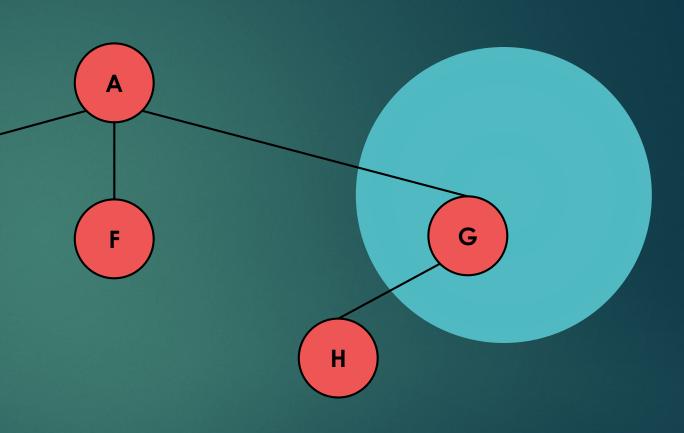


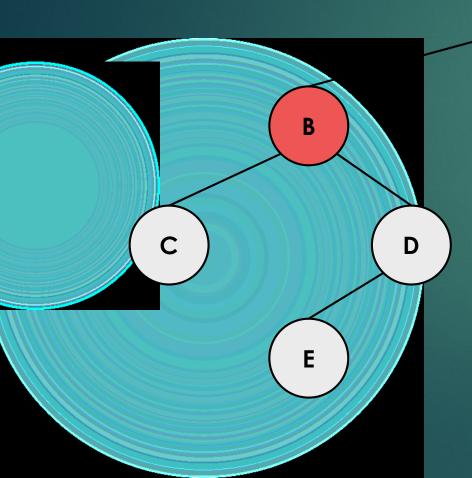


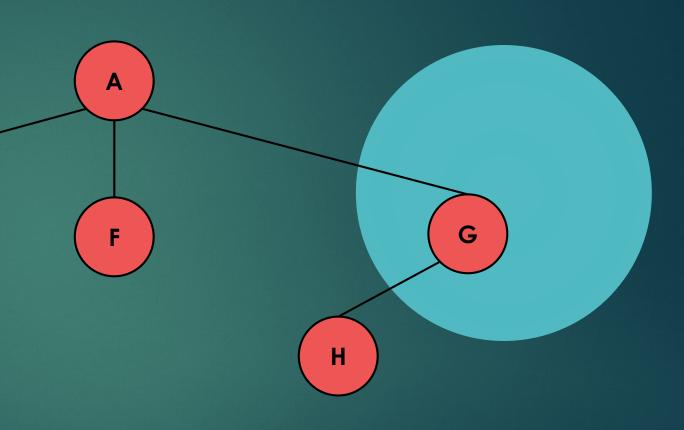






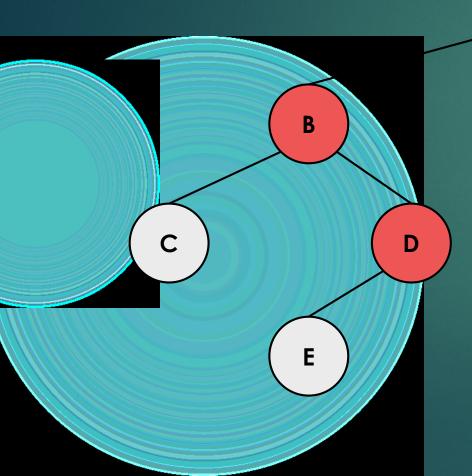


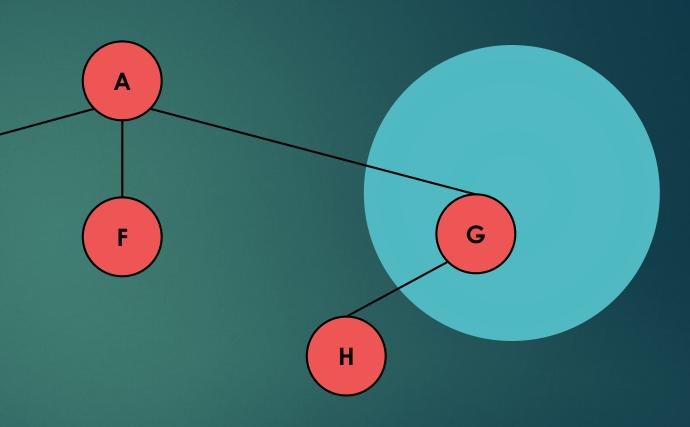




Symmetry in DFS

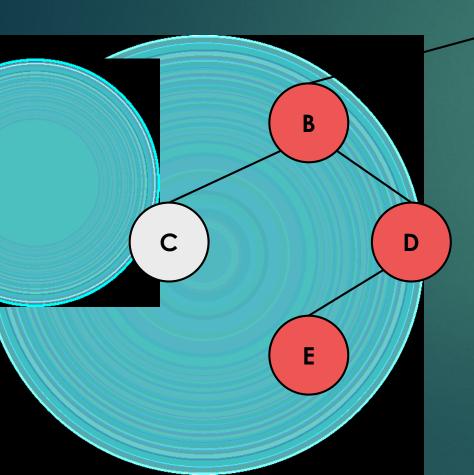
We can go to the opposite direction, it is going to be a valid DFS as well !!!

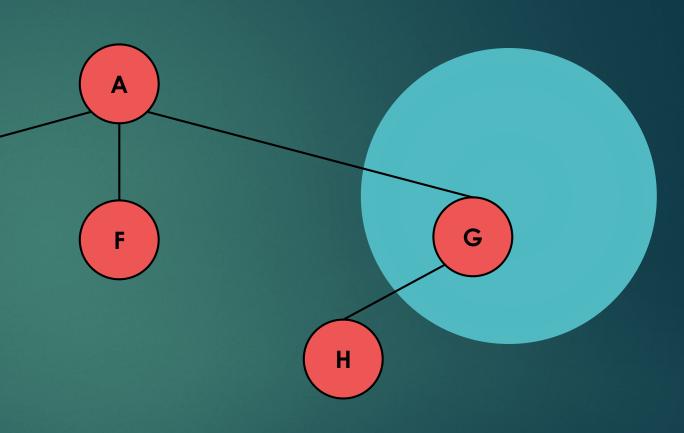




Symmetry in DFS

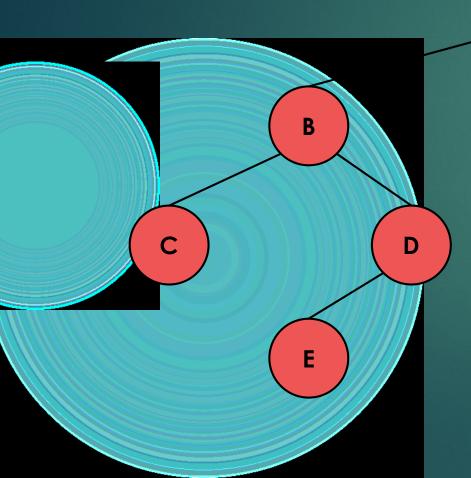
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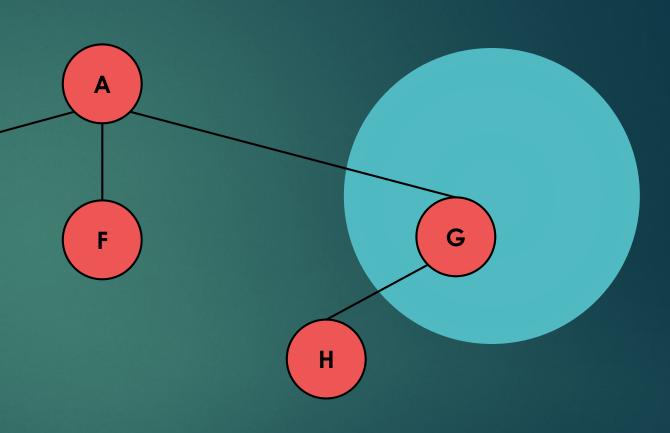




Symmetry in DFS

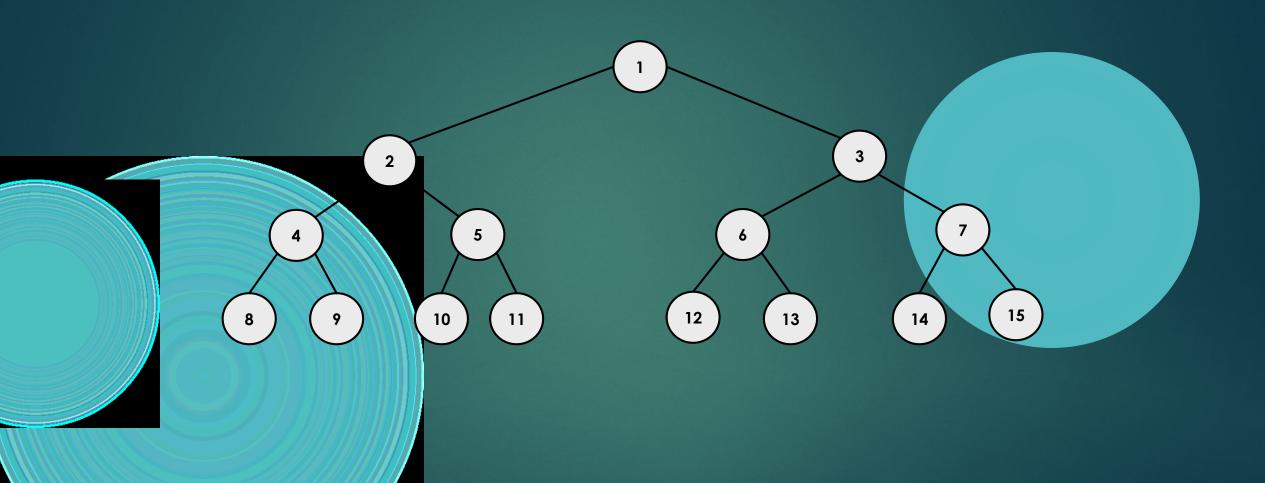
We can go to the opposite direction, it is going to be a valid DFS as well !!!

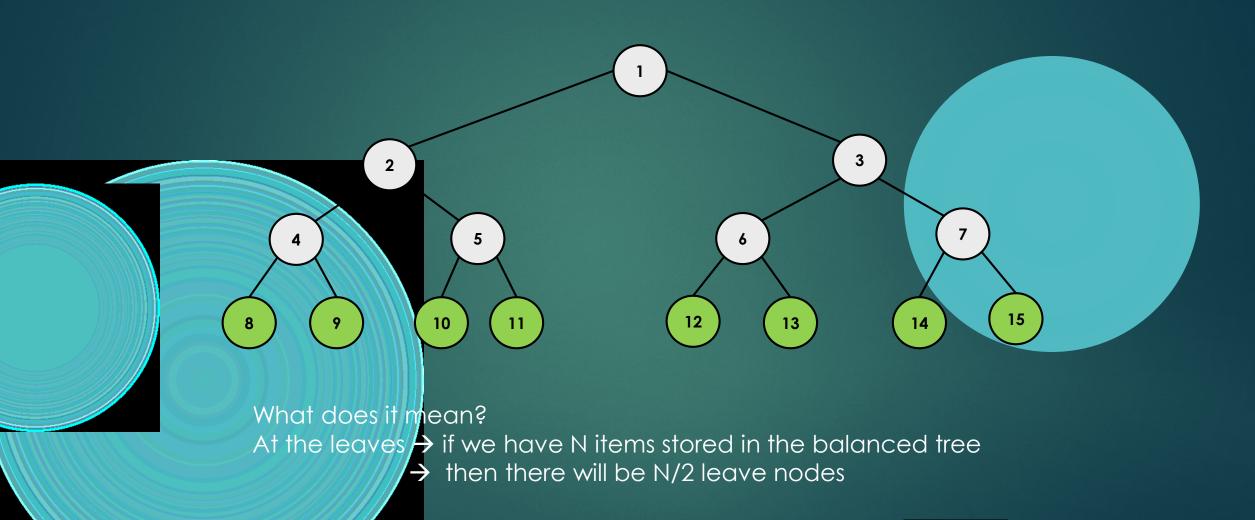


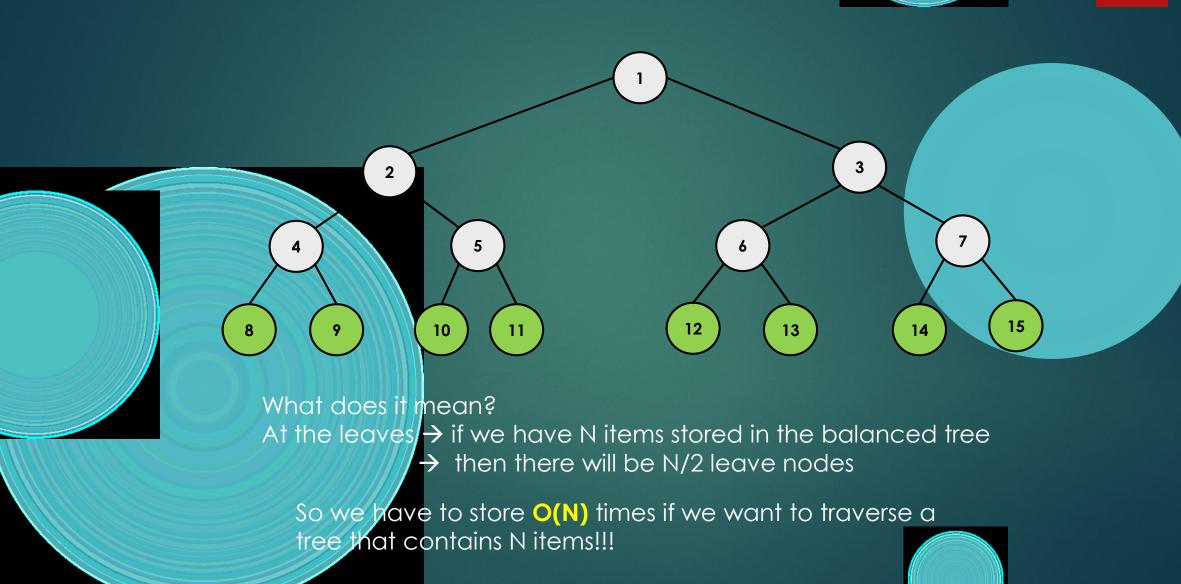


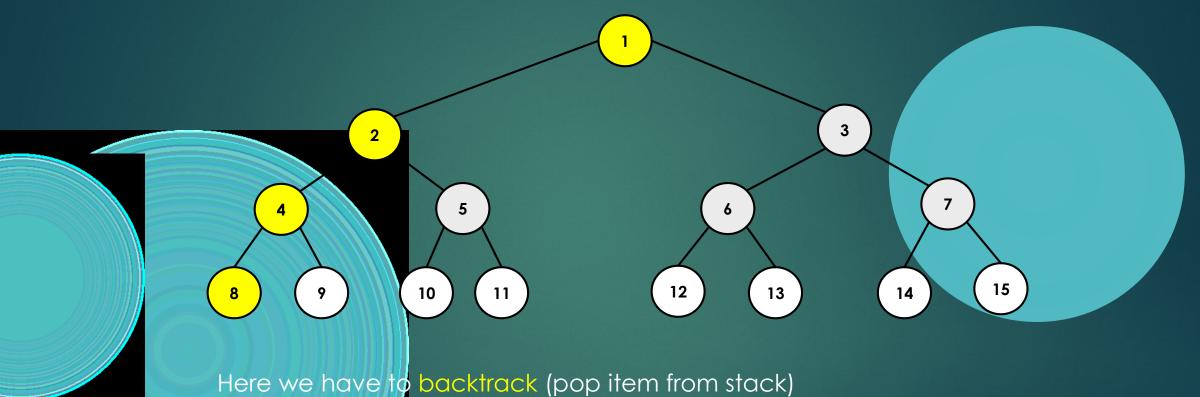
DFS IMPLEMENTATION

Memory Complexity







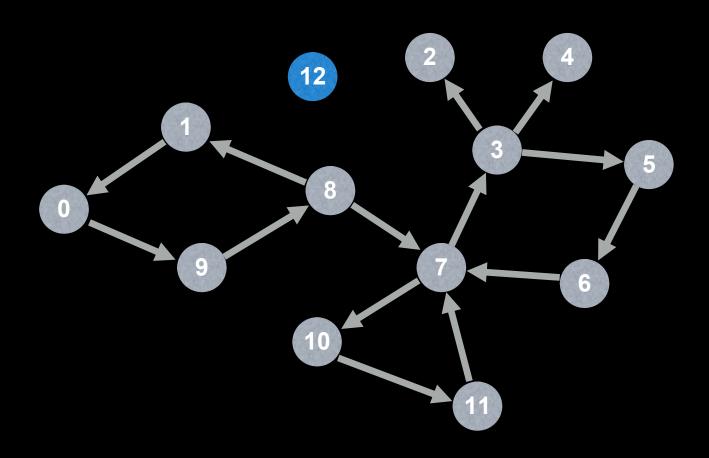


- We just have to store as many items in the stack as the height of the tree
- Which is log/N!!!
- The memory complexity will be O(logN)

- Breadth-First Search: O(N)
- Depth-First Search: O(logN)
 - Depth-First Search is **preferred** most of the time
 There may be some situations where BFS is better
 - Artificial intelligence
 - Robot movements



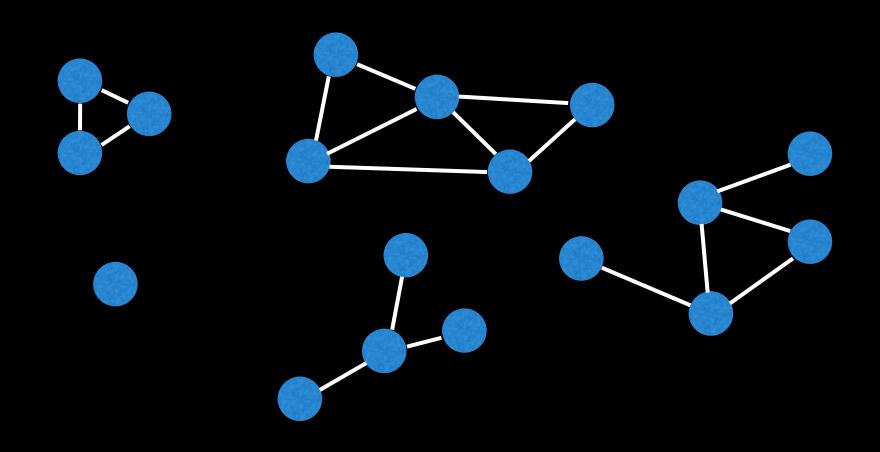
Basic DFS



```
# Global or class scope variables
n = number of nodes in the graph
g = adjacency list representing graph
visited = [false, ..., false] # size n
function dfs(at):
 if visited[at]: return
 visited[at] = true
 neighbours = graph[at]
 for next in neighbours:
   dfs(next)
# Start DFS at node zero
start node = 0
dfs(start node)
```

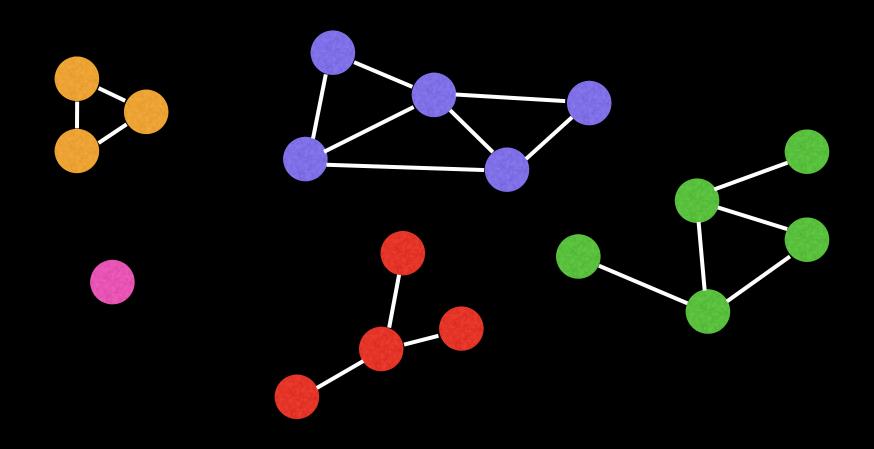
Connected Components

Sometimes a graph is split into multiple components. It's useful to be able to identify and count these components.



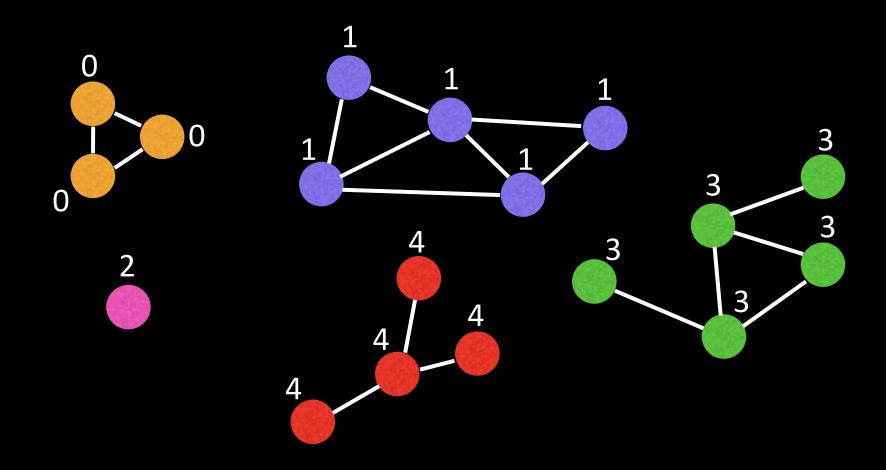
Connected Components

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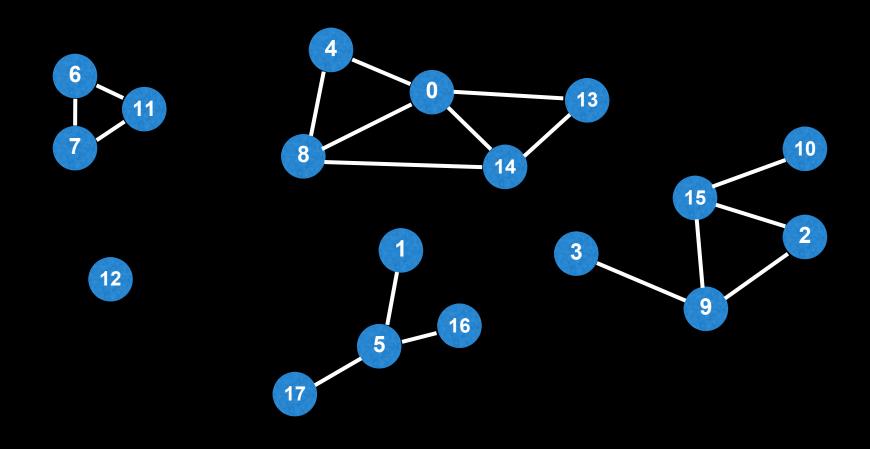


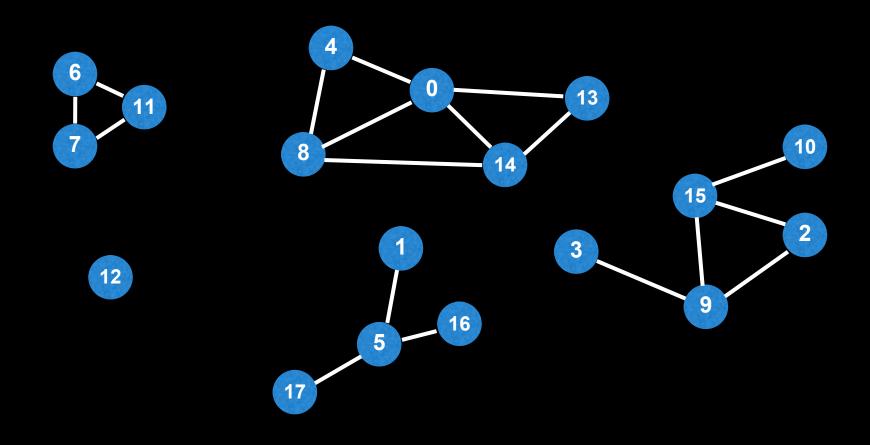
Connected Components

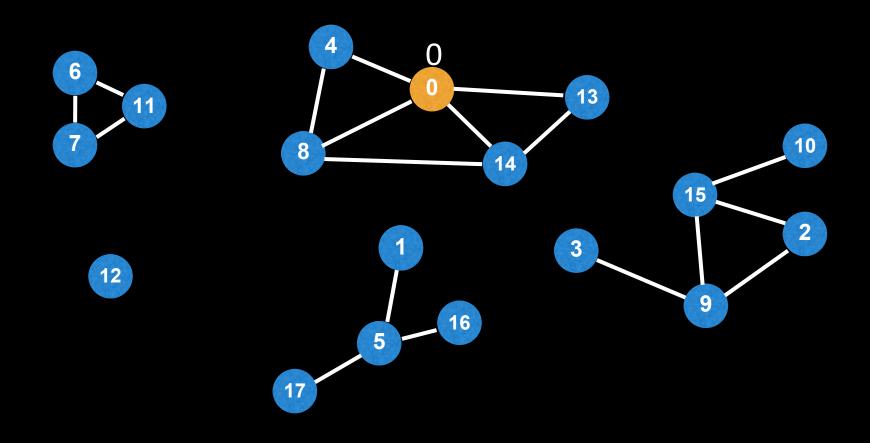
Assign an integer value to each group to be able to tell them apart.

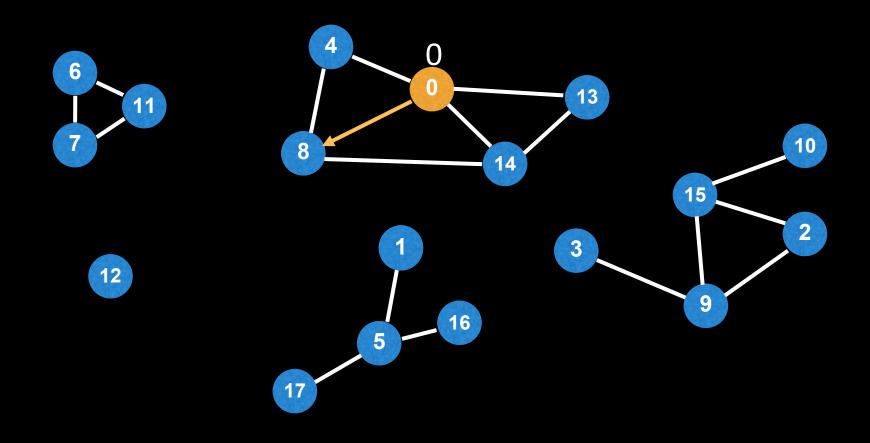


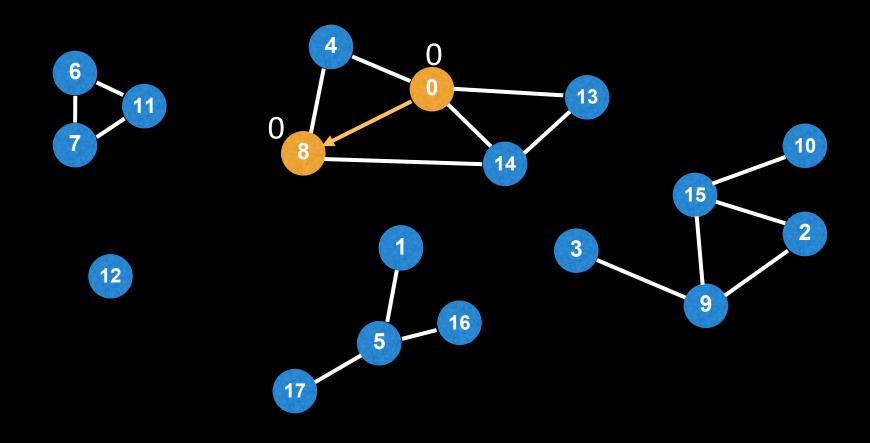
We can use a DFS to identify components. First, make sure all the nodes are labeled from [0, n) where n is the number of nodes.

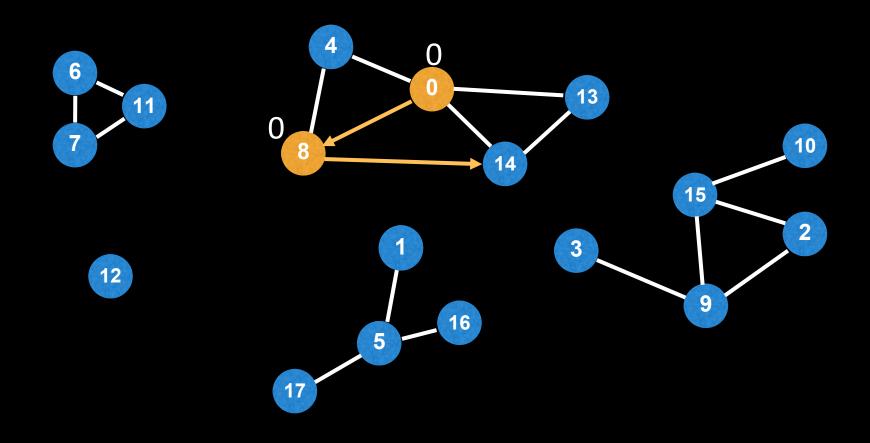


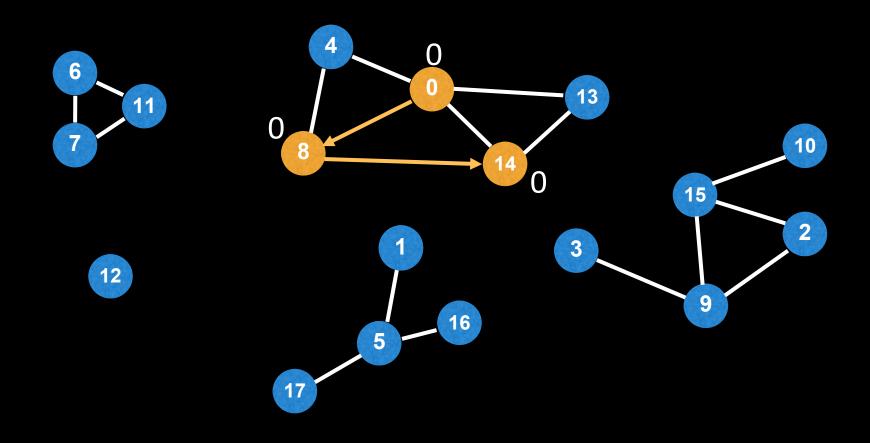


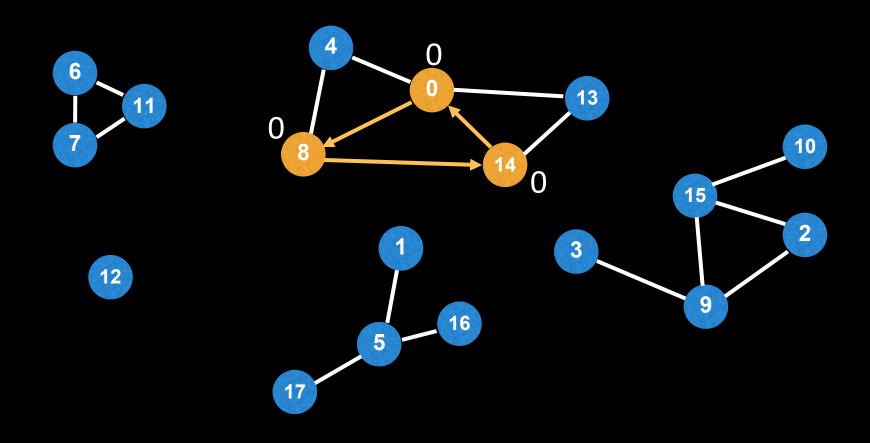


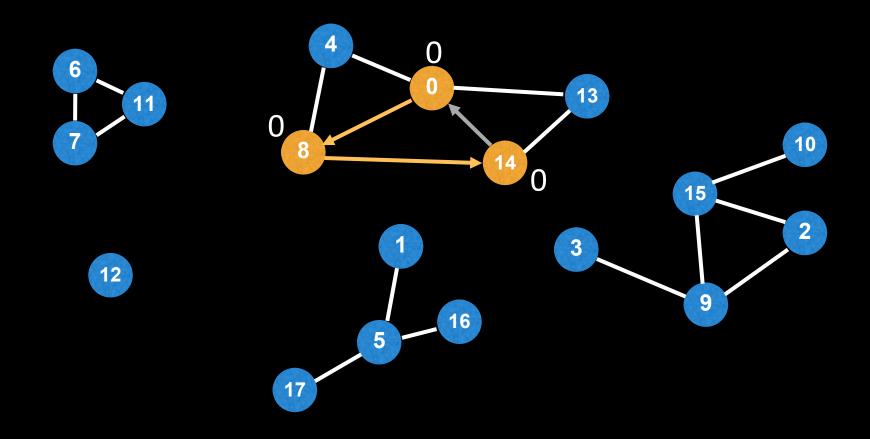


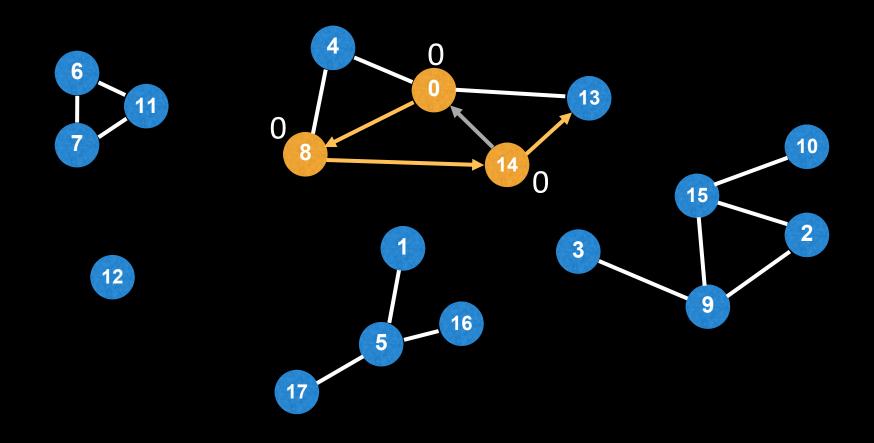


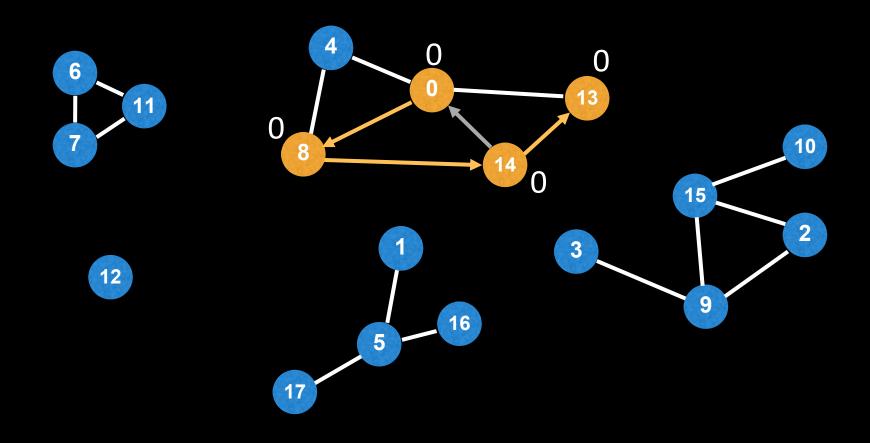


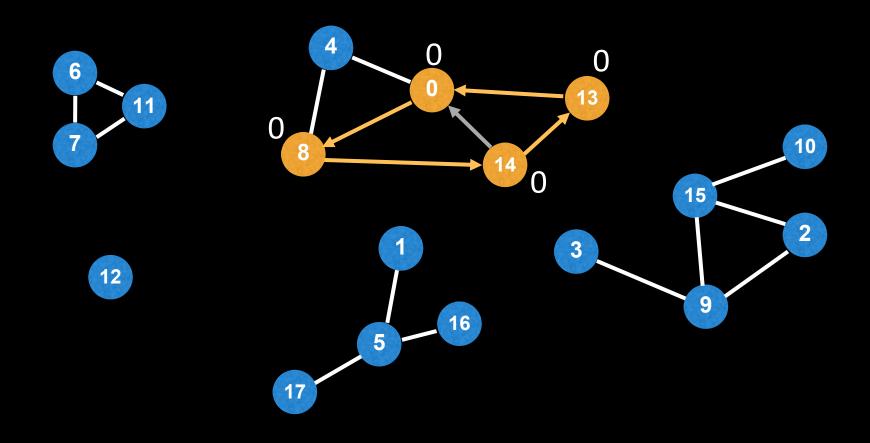


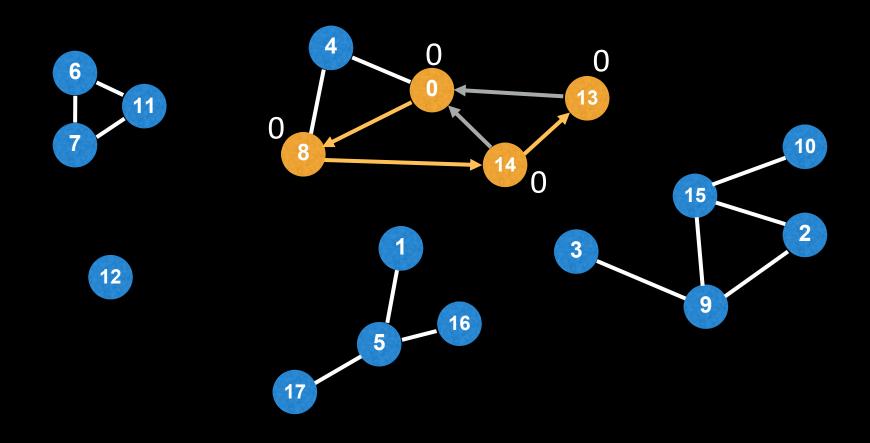


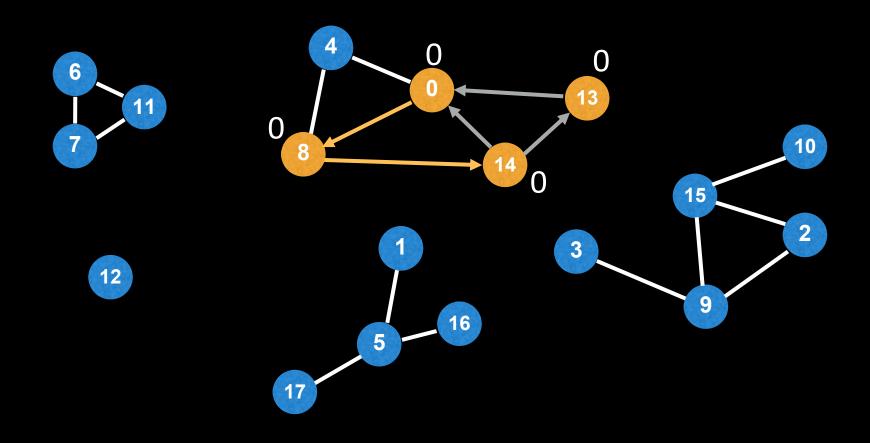


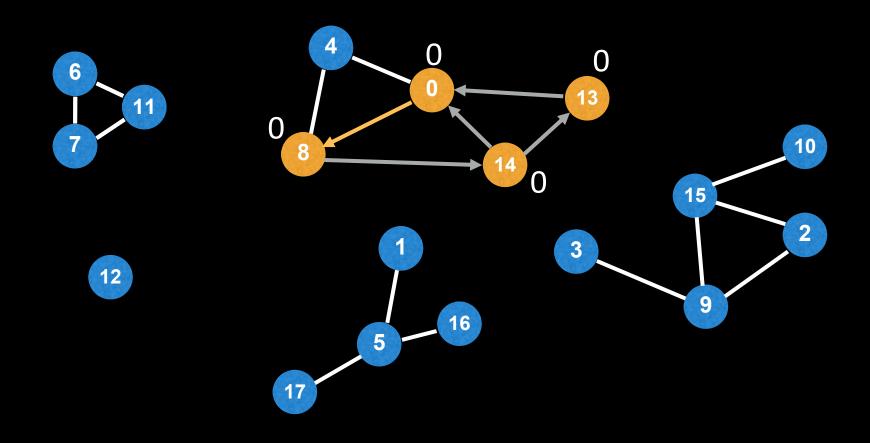


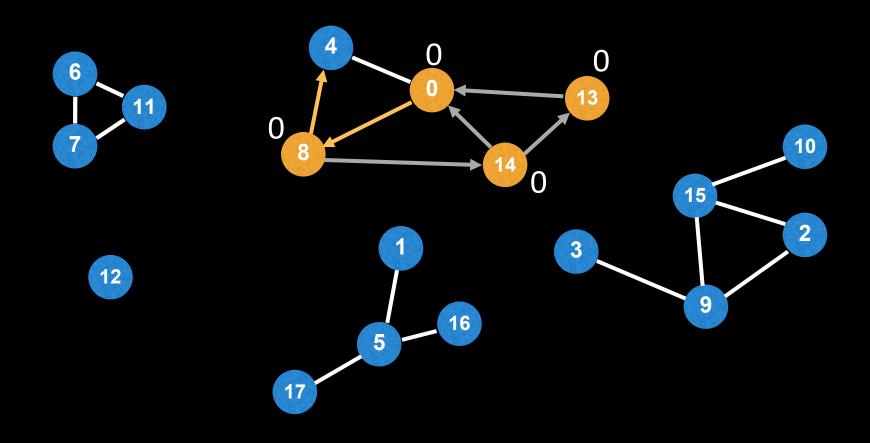


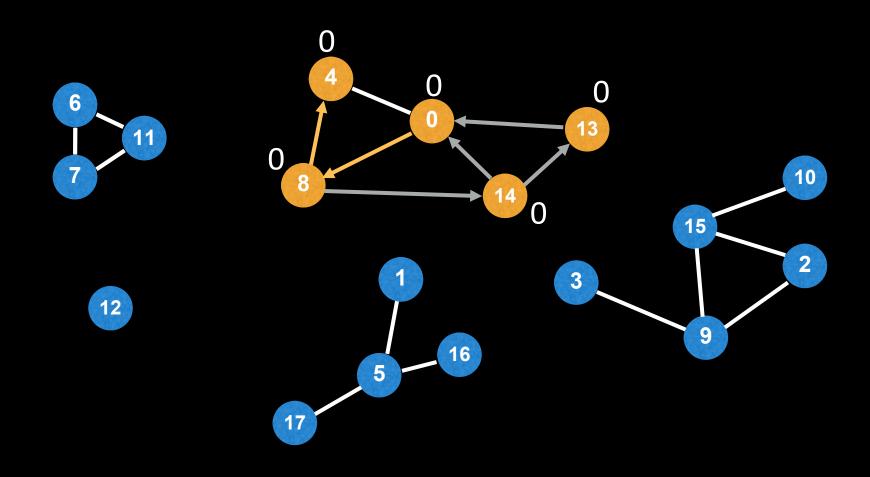


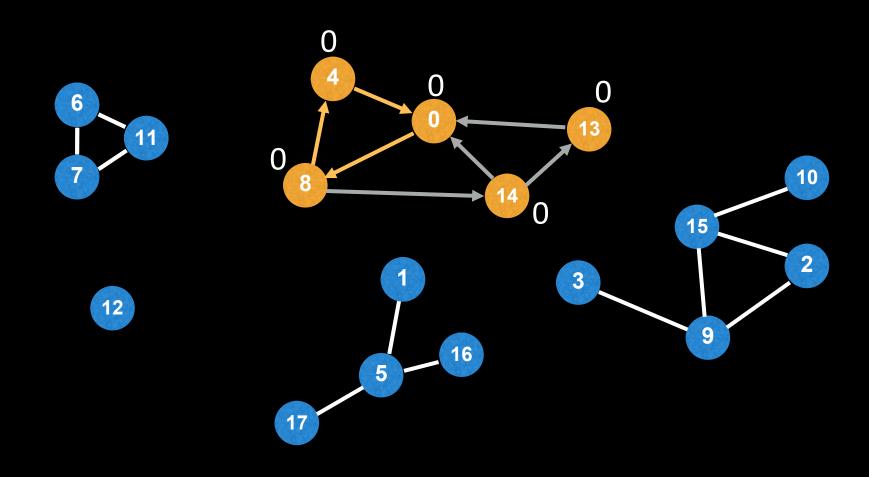


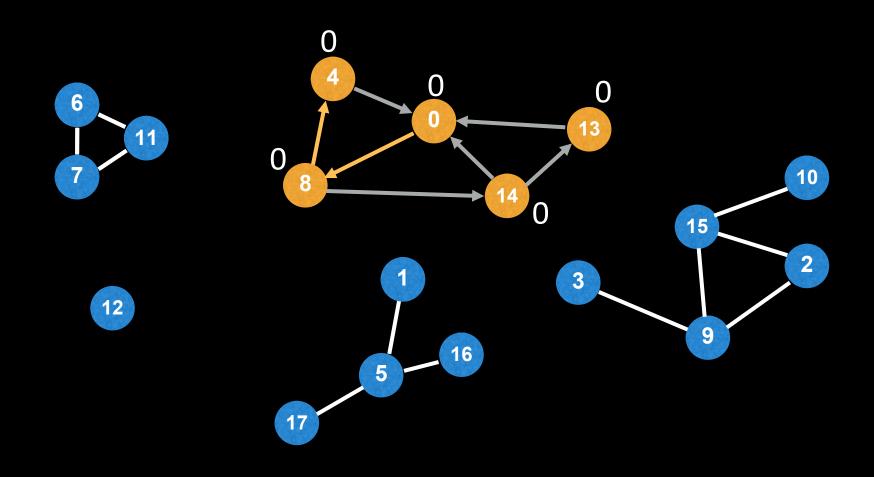


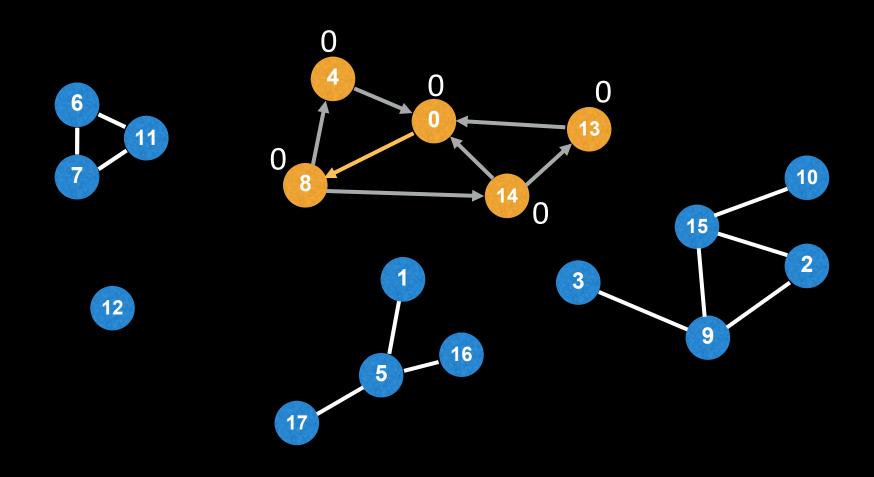


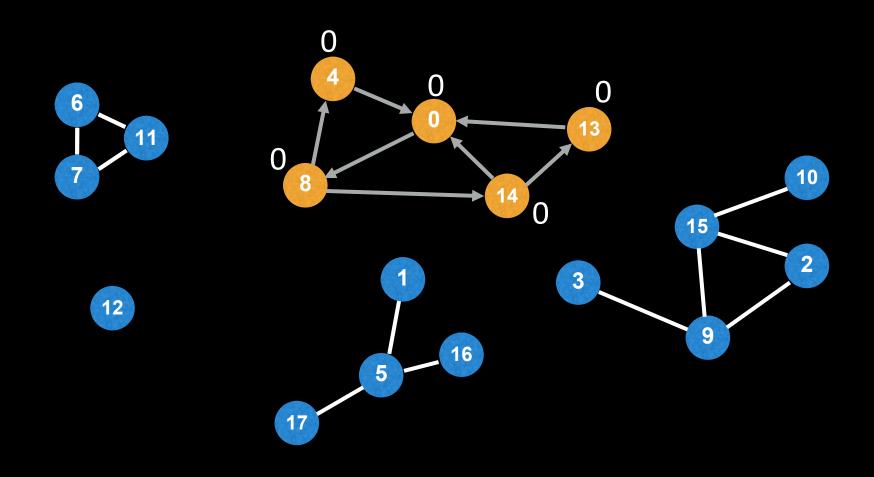


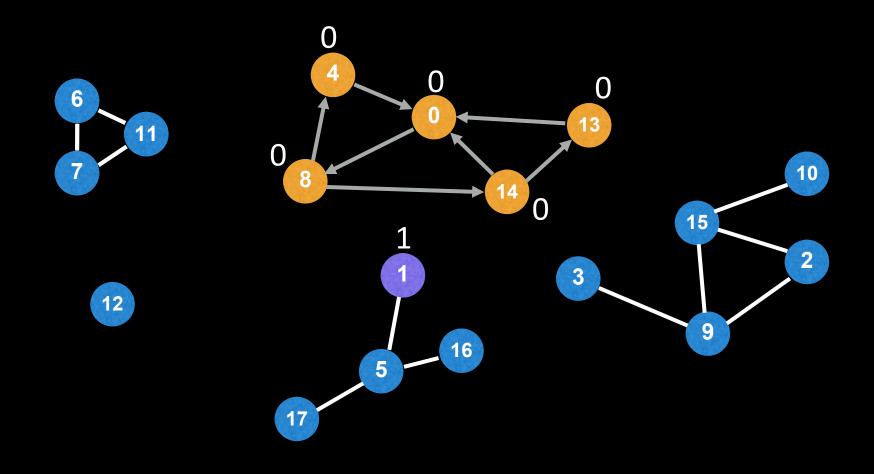


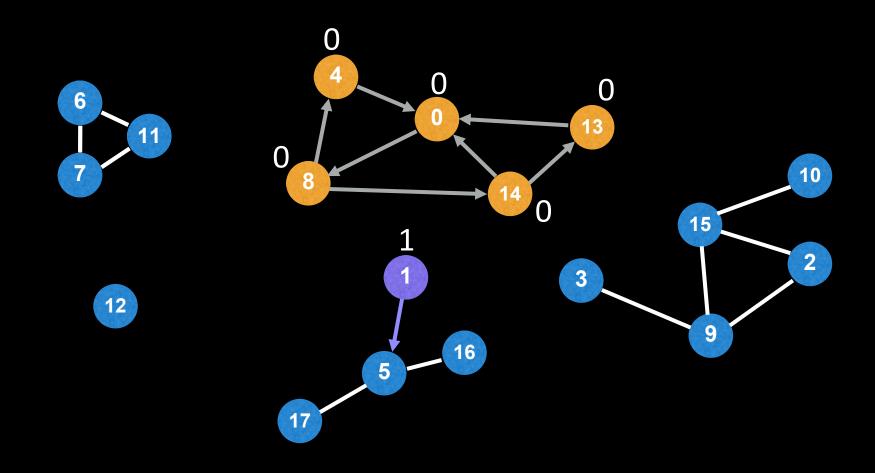


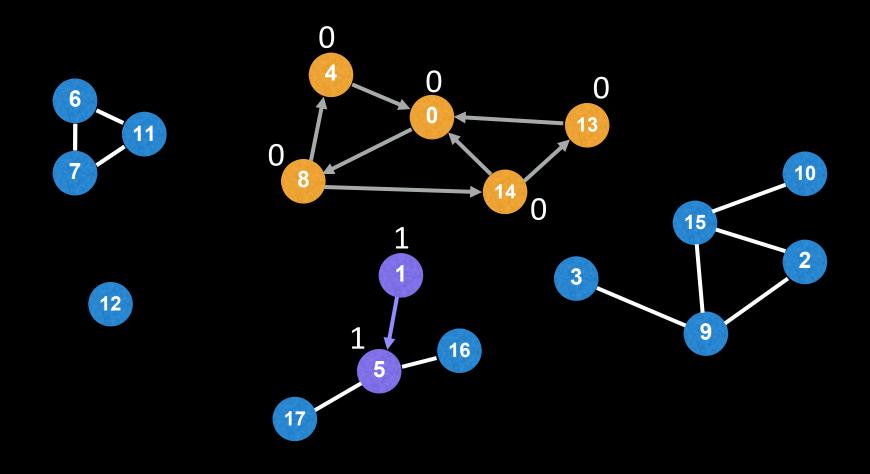


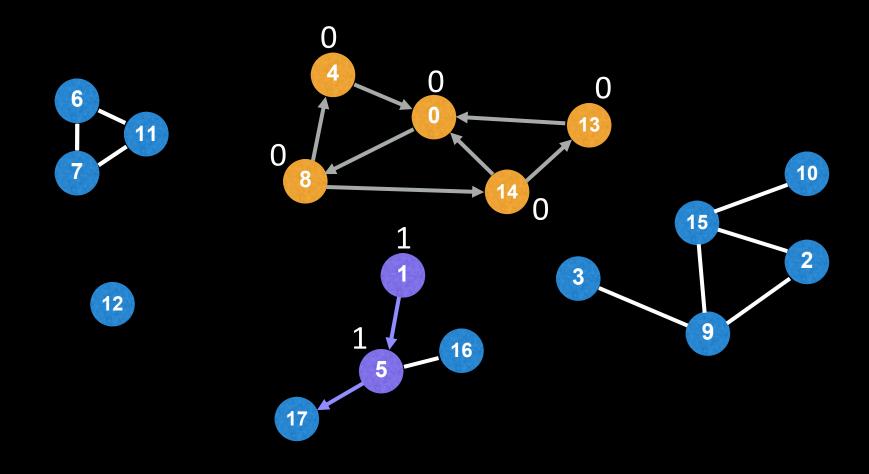


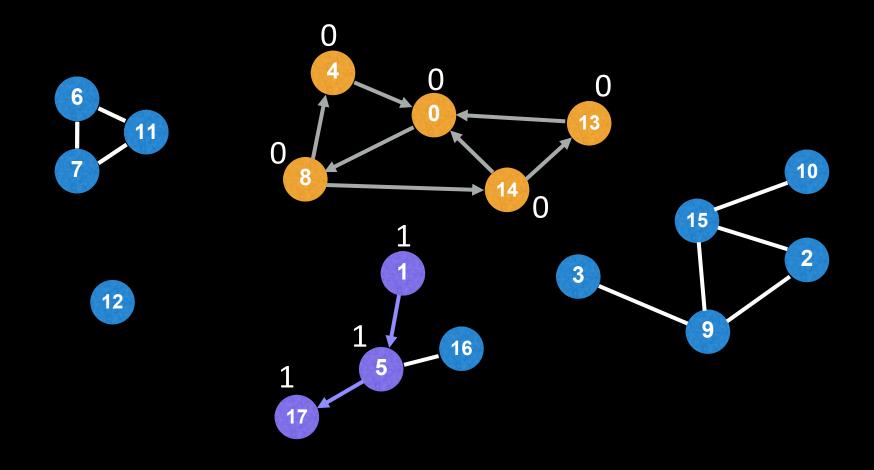


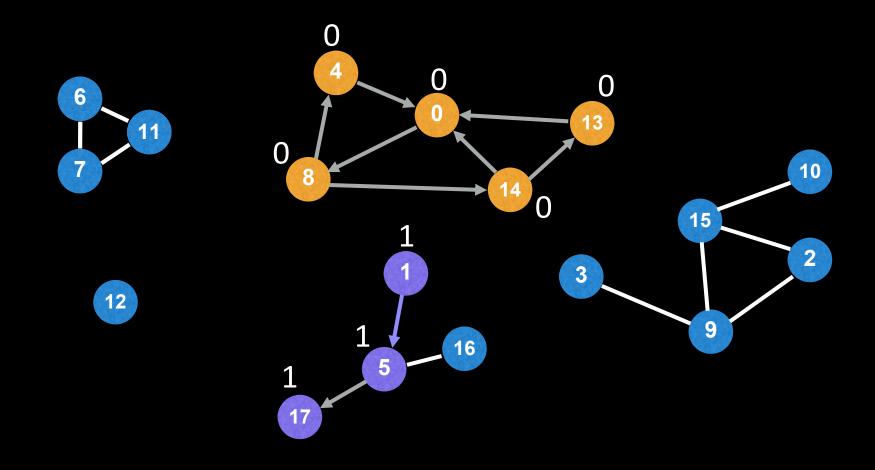


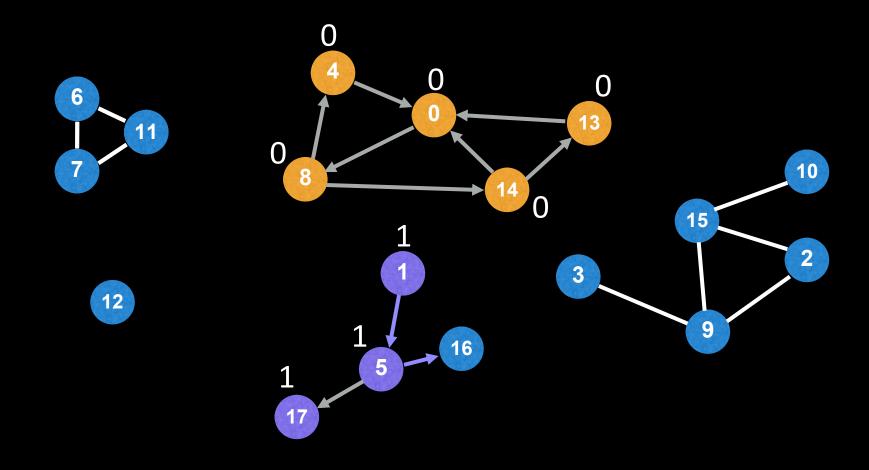


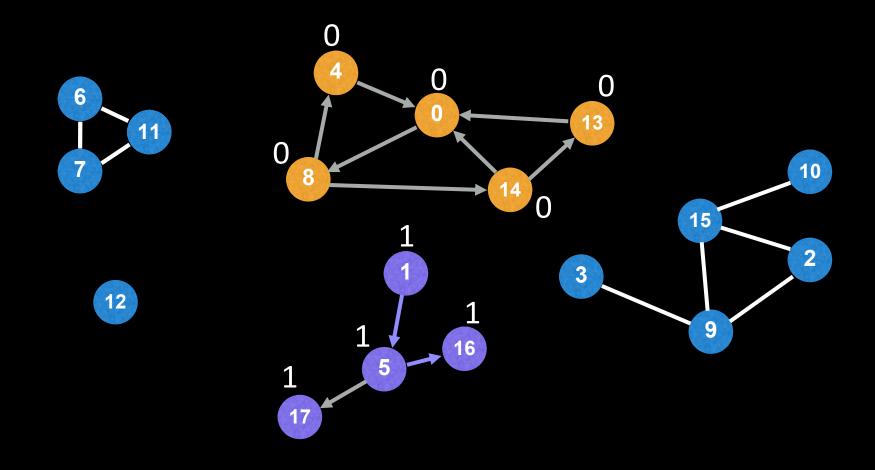


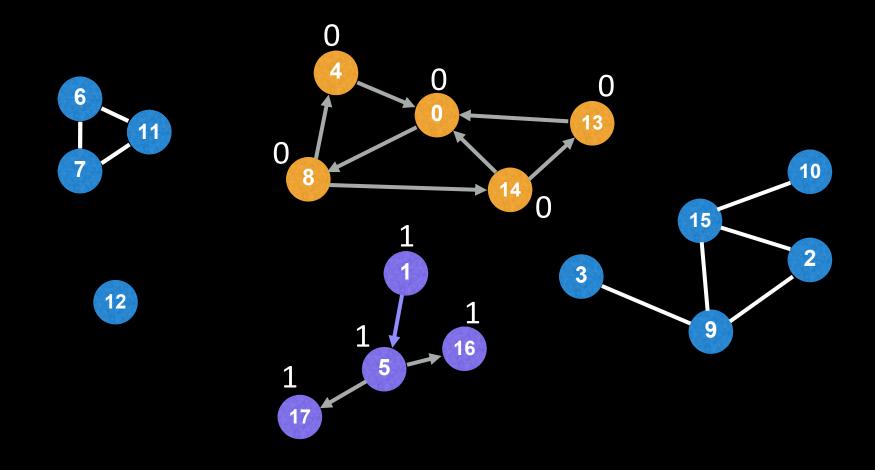


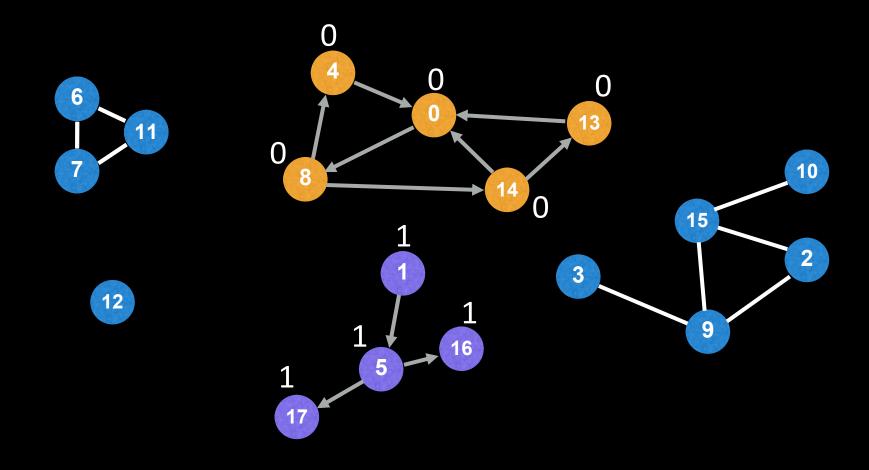




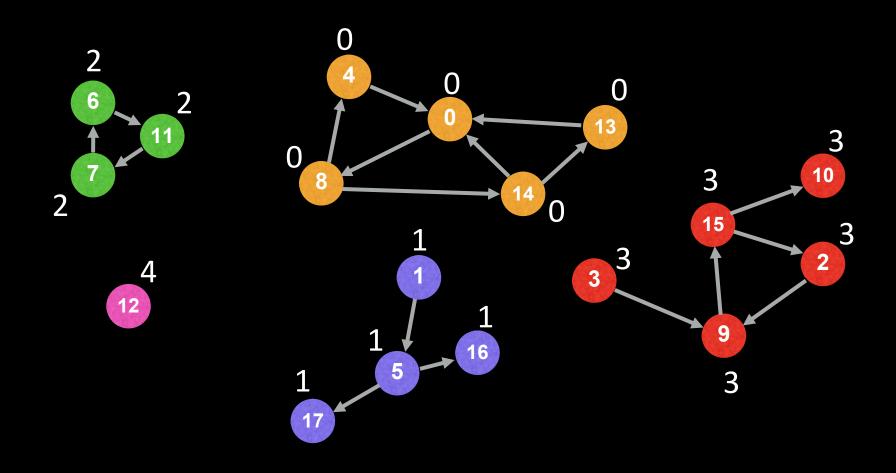








... and so on for the other components



```
# Global or class scope variables
n = number of nodes in the graph
g = adjacency list representing graph
count = 0
components = empty integer array # size n
visited = [false, ..., false] # size n
function findComponents():
 for (i = 0; i < n; i++):
  if !visited[i]:
   count++
   dfs(i)
 return (count, components)
function dfs(at):
 visited[at] = true
   components[at] = count
 for (next : g[at]):
  if !visited[next]:
   dfs(next)
```

What else can DFS do?

We can augment the DFS algorithm to:

- Compute a graph's minimum spanning tree.
- Detect and find cycles in a graph.
- Check if a graph is bipartite.
- Find strongly connected components.
- Topologically sort the nodes of a graph.
- Find bridges and articulation points.
- Find augmenting paths in a flow network.
- Generate mazes.