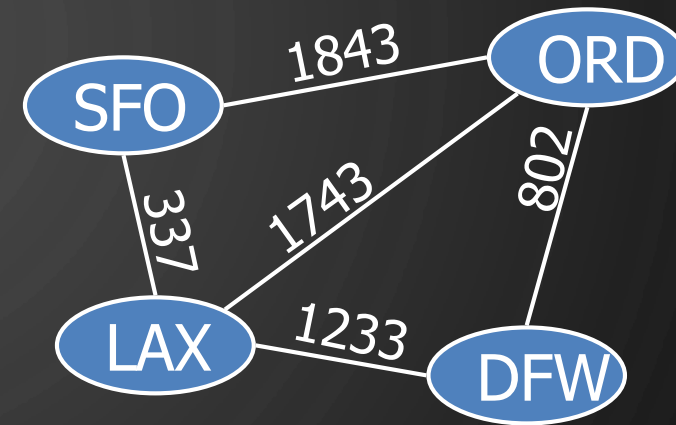


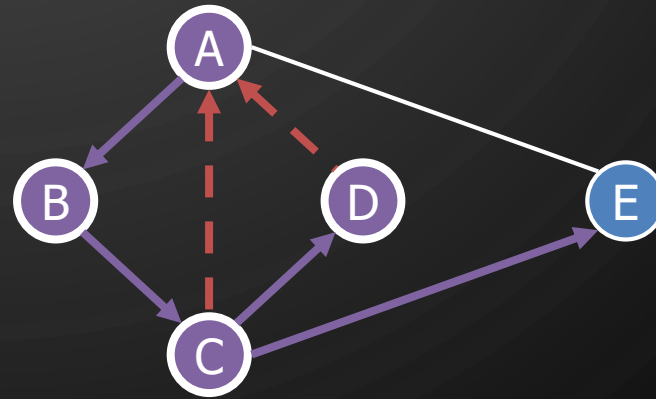
# CHAPTER 14

## GRAPH ALGORITHMS

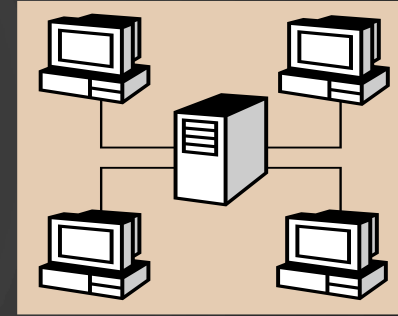
ACKNOWLEDGEMENT: THESE SLIDES ARE ADAPTED FROM SLIDES PROVIDED WITH DATA STRUCTURES AND ALGORITHMS IN JAVA, GOODRICH, TAMASSIA AND GOLDWASSER (WILEY 2016)



# DEPTH-FIRST SEARCH



# DEPTH-FIRST SEARCH



- **Depth-first search (DFS)** is a general technique for traversing a graph
- A DFS traversal of a graph  $G$ 
  - Visits all the vertices and edges of  $G$
  - Determines whether  $G$  is connected
  - Computes the connected components of  $G$
  - Computes a spanning forest of  $G$
- DFS on a graph with  $n$  vertices and  $m$  edges takes  $O(n + m)$  time
- DFS can be further extended to solve other graph problems
  - Find and report a path between two given vertices
  - Find a cycle in the graph
- Depth-first search is to graphs as what Euler tour is to binary trees

# DFS ALGORITHM FROM A VERTEX






Algorithm DFS( $G, u$ )

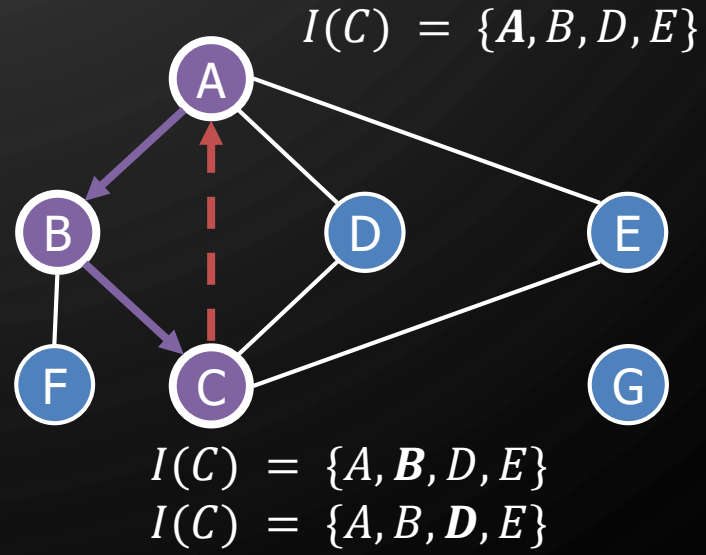
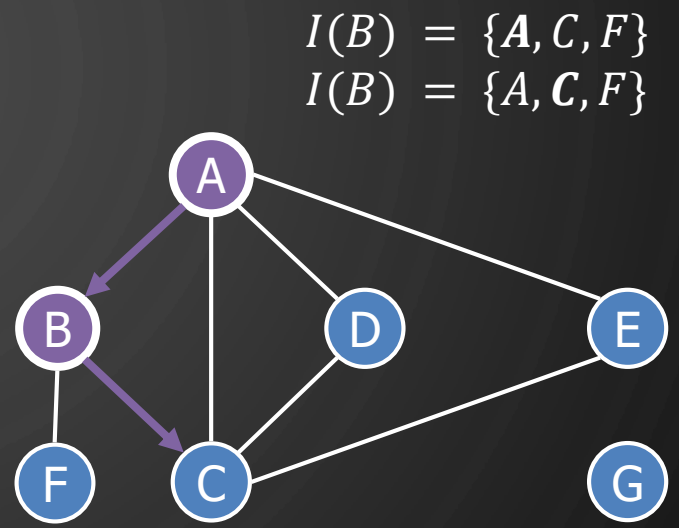
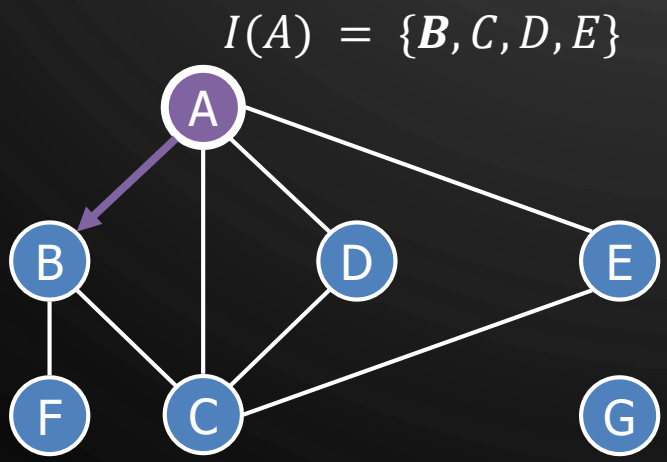
**Input:** A graph  $G$  and a vertex  $u$  of  $G$

**Output:** A collection of vertices reachable from  $u$ ,  
with their discovery edges

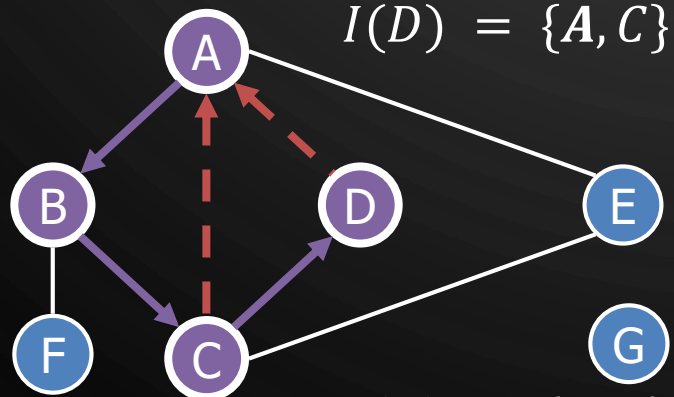
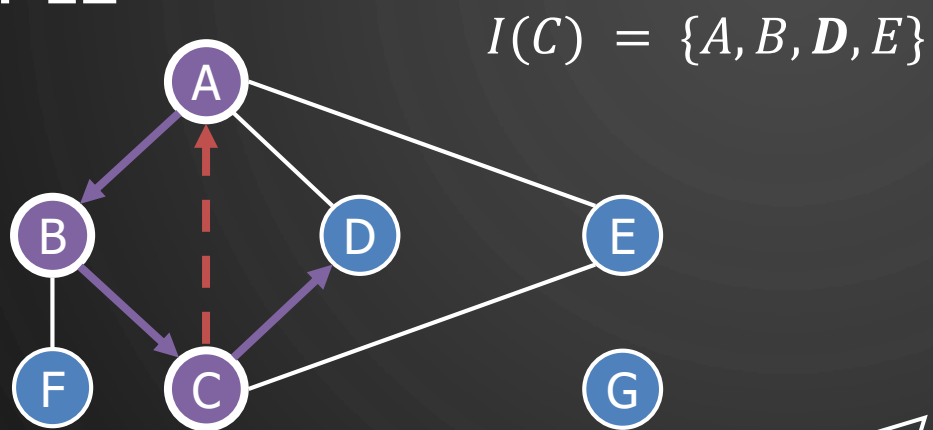
1. Mark  $u$  as visited
2. **for each** edge  $e = (u, v) \in G.outgoingEdges(u)$  **do**
3.     **if**  $v$  has not been visited **then**
4.         Record  $e$  as a discovery edge for  $v$
5.         DFS( $G, v$ )

# EXAMPLE

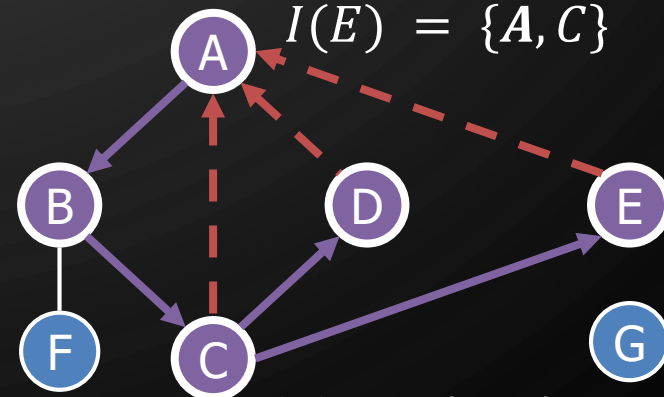
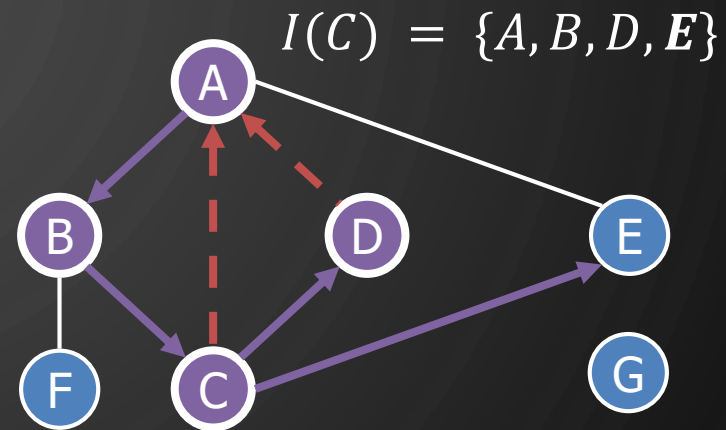
-  unexplored vertex
-  visited vertex
-  unexplored edge
-  discovery edge
-  back edge



# EXAMPLE

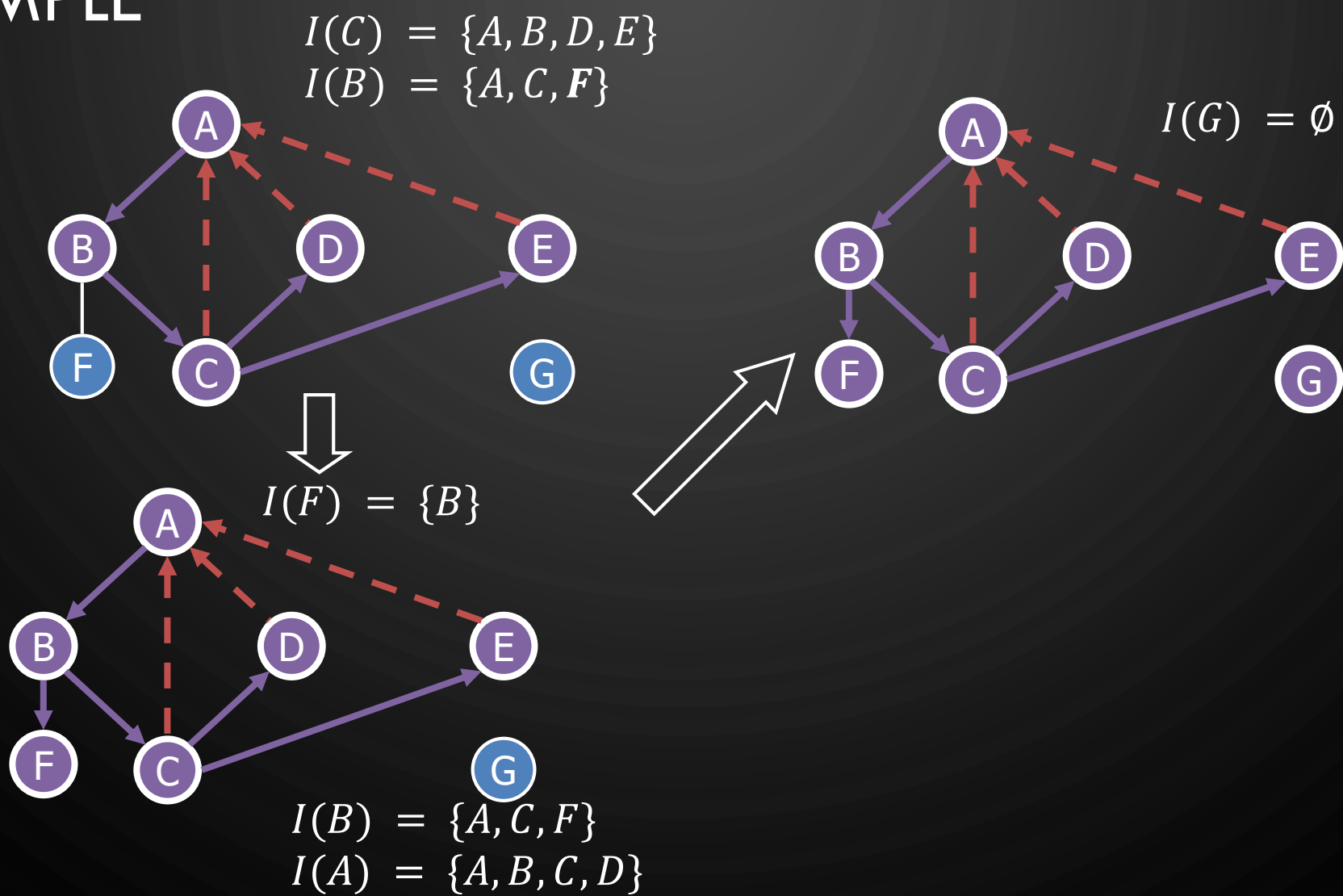


$$I(D) = \{A, C\}$$
$$I(D) = \{A, C\}$$



$$I(E) = \{A, C\}$$
$$I(E) = \{A, C\}$$

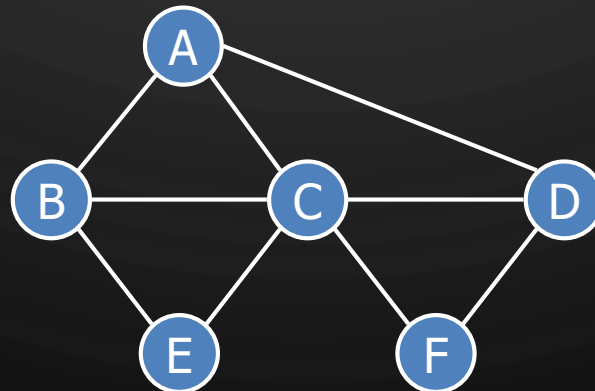
# EXAMPLE



# EXERCISE

## DFS ALGORITHM

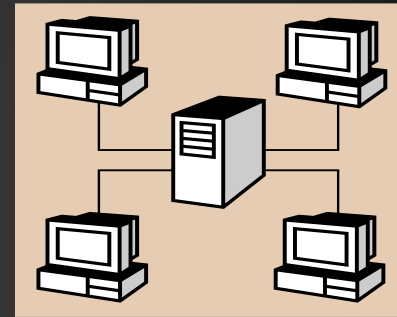
- Perform DFS of the following graph, start from vertex A
  - Assume adjacent edges are processed in alphabetical order
  - Number vertices in the order they are visited
  - Label edges as discovery or back edges







# DFS ALGORITHM



- The algorithm uses a mechanism for setting and getting “labels” of vertices and edges

Algorithm DFS( $G$ )

**Input:** Graph  $G$

**Output:** Labeling of the edges of  $G$  as discovery edges and back edges

```
1. for each  $v \in G.vertices()$  do
2.   setLabel( $v$ , UNEXPLORED)
3. for each  $e \in G.edges()$  do
4.   setLabel( $e$ , UNEXPLORED)
5. for each  $v \in G.vertices()$  do
6.   if getLabel( $v$ ) = UNEXPLORED then
7.     DFS( $G, v$ )
```

Algorithm DFS( $G, v$ )

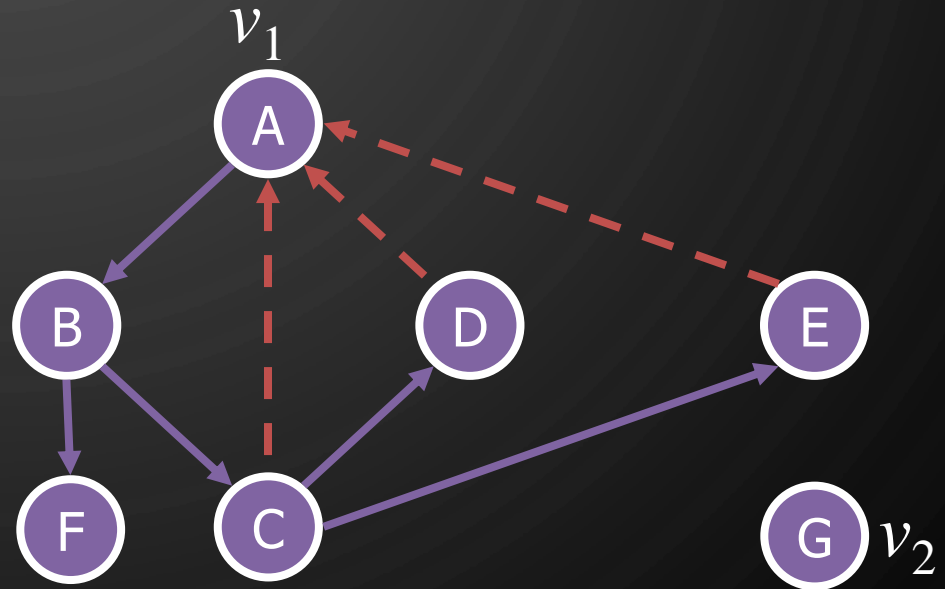
**Input:** Graph  $G$  and a start vertex  $v$

**Output:** Labeling of the edges of  $G$  in the connected component of  $v$  as discovery edges and back edges

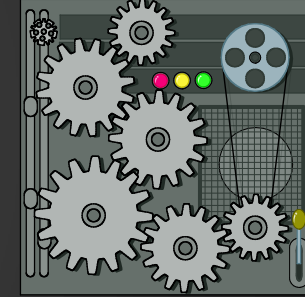
```
1. setLabel( $v$ , VISITED)
2. for each  $e \in G.outgoingEdges(v)$  do
3.   if getLabel( $e$ ) = UNEXPLORED)
4.      $w \leftarrow G.opposite(v, e)$ 
5.     if getLabel( $w$ ) = UNEXPLORED then
6.       setLabel( $e$ , DISCOVERY)
7.       DFS( $G, w$ )
8.   else
9.     setLabel( $e$ , BACK)
```

# PROPERTIES OF DFS

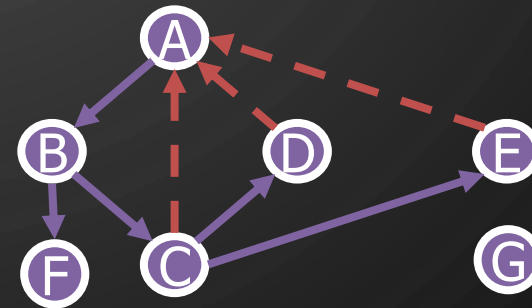
- Property 1
  - $\text{DFS}(G, v)$  visits all the vertices and edges in the connected component of  $v$
- Property 2
  - The discovery edges labeled by  $\text{DFS}(G, v)$  form a spanning tree of the connected component of  $v$



# ANALYSIS OF DFS



- Setting/getting a vertex/edge label takes  $O(1)$  time
- Each vertex is labeled twice
  - once as *UNEXPLORED*
  - once as *VISITED*
- Each edge is labeled twice
  - once as *UNEXPLORED*
  - once as *DISCOVERY* or *BACK*
- Function  $\text{DFS}(G, v)$  and the method  $\text{outgoingEdges}()$  are called once for each vertex
- DFS runs in  $O(n + m)$  time provided the graph is represented by the adjacency list structure
  - Recall that  $\sum_v \text{deg}(v) = 2m$



# APPLICATION PATH FINDING

- We can specialize the DFS algorithm to find a path between two given vertices  $u$  and  $z$  using the template method pattern
- We call  $\text{DFS}(G, u)$  with  $u$  as the start vertex
- We use a stack  $S$  to keep track of the path between the start vertex and the current vertex
- As soon as destination vertex  $z$  is encountered, we return the path as the contents of the stack



Algorithm  $\text{pathDFS}(G, v, z)$

**Input:** Graph  $G$ , a start vertex  $v$ ,  
a goal vertex  $z$

**Output:** Path between  $v$  and  $z$

```
1. setLabel( $v$ , VISITED)
2.  $S$ .push( $v$ )
3. if  $v = z$  then
4.   return  $S$ .elements()
5. for each  $e \in G$ .outgoingEdges( $v$ ) do
6.   if getLabel( $e$ ) = UNEXPLORED then
7.      $w \leftarrow G$ .opposite( $v, e$ )
8.     if getLabel( $w$ ) = UNEXPLORED then
9.       setLabel( $e$ , DISCOVERY)
10.     $S$ .push( $e$ )
11.    pathDFS( $G, w$ )
12.     $S$ .pop()
13.   else
14.     setLabel( $e$ , BACK)
15.  $S$ .pop()
```

# APPLICATION CYCLE FINDING

- We can specialize the DFS algorithm to find a simple cycle using the template method pattern
- We use a stack  $S$  to keep track of the path between the start vertex and the current vertex
- As soon as a back edge  $(v, w)$  is encountered, we return the cycle as the portion of the stack from the top to vertex  $w$



**Algorithm** cycleDFS( $G, v$ )

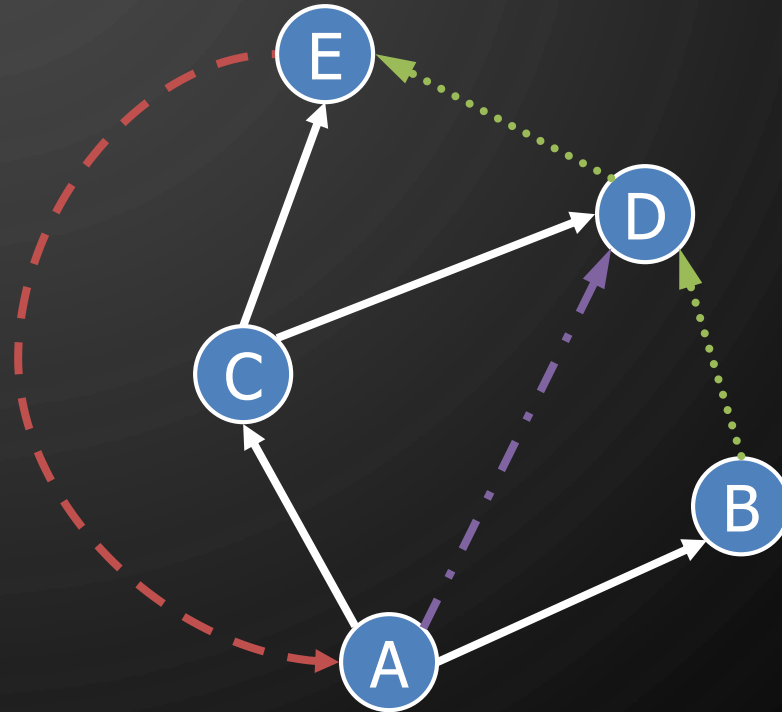
**Input:** Graph  $G$ , a start vertex  $v$

**Output:** Cycle containing  $v$

```
1. setLabel( $v$ , VISITED)
2.  $S$ .push( $v$ )
3. for each  $e \in G$ .outgoingEdges( $v$ ) do
4.   if getLabel( $e$ ) = UNEXPLORED then
5.      $w \leftarrow G$ .opposite( $v, e$ )
6.      $S$ .push( $e$ )
7.     if getLabel( $w$ ) = UNEXPLORED then
8.       setLabel( $e$ , DISCOVERY)
9.       cycleDFS( $G, w$ )
10.     $S$ .pop()
11.   else
12.     Stack  $T \leftarrow \emptyset$ 
13.     repeat
14.        $T$ .push( $S$ .pop())
15.     until  $T$ .top() =  $w$ 
16.     return  $T$ .elements()
17.  $S$ .pop()
```

# DIRECTED DFS

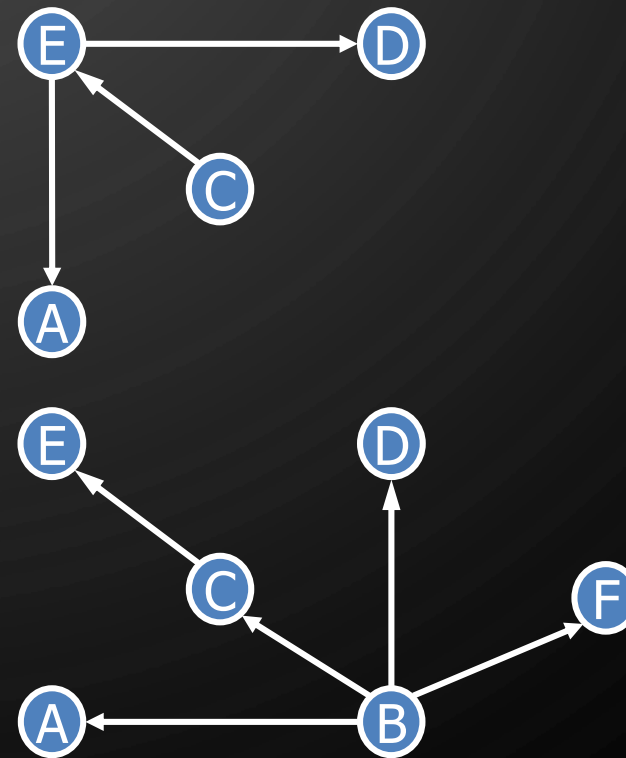
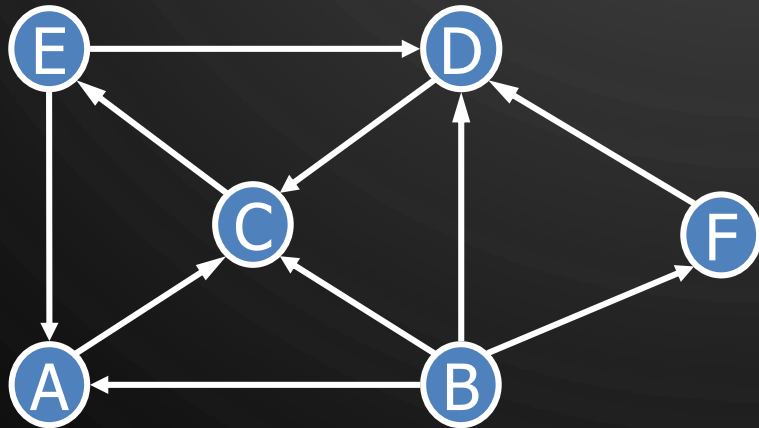
- We can specialize the traversal algorithms (DFS and BFS) to digraphs by traversing edges only along their direction
- In the directed DFS algorithm, we have four types of edges
  - discovery edges
  - back edges
  - forward edges
  - cross edges
- A directed DFS starting at a vertex  $s$  determines the vertices reachable from  $s$



# REACHABILITY



- DFS tree rooted at  $v$ : vertices reachable from  $v$  via directed paths

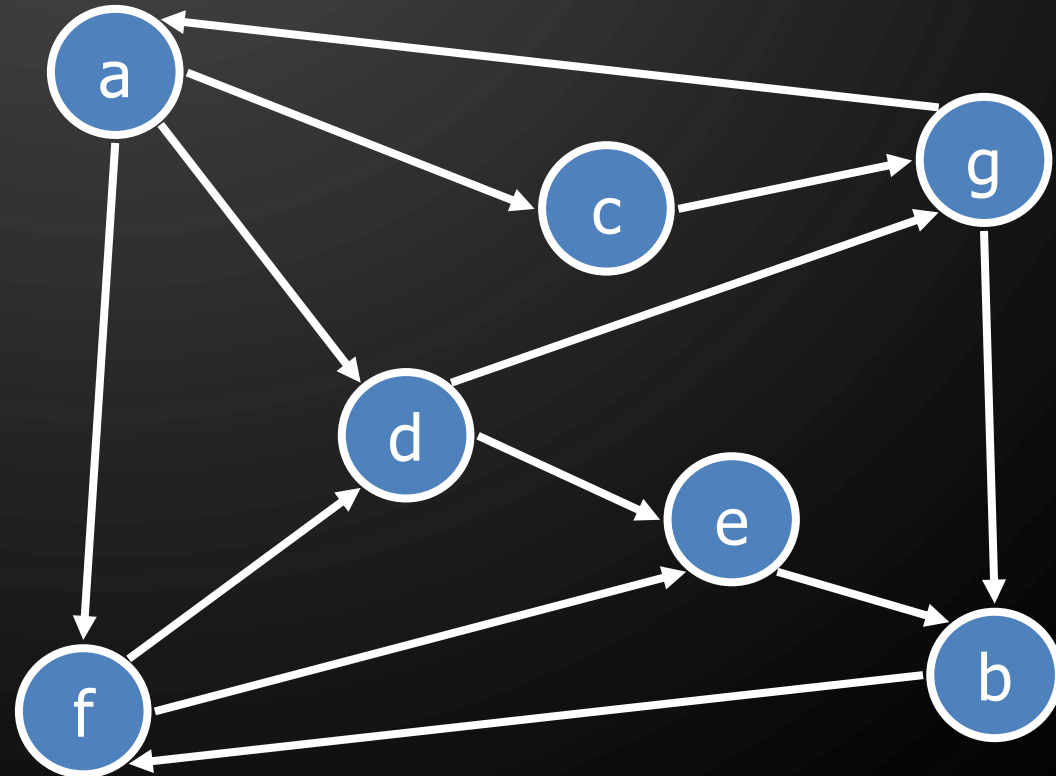




# STRONG CONNECTIVITY



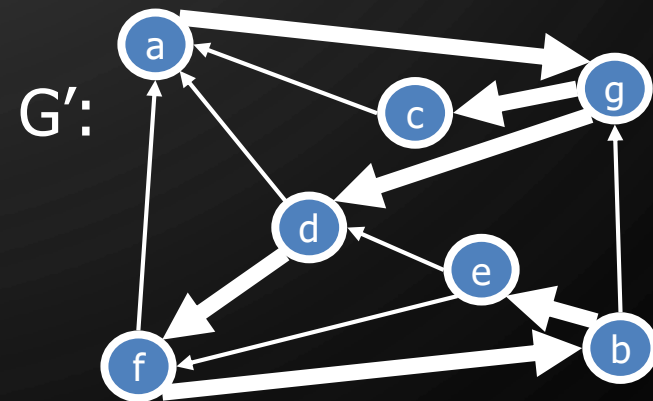
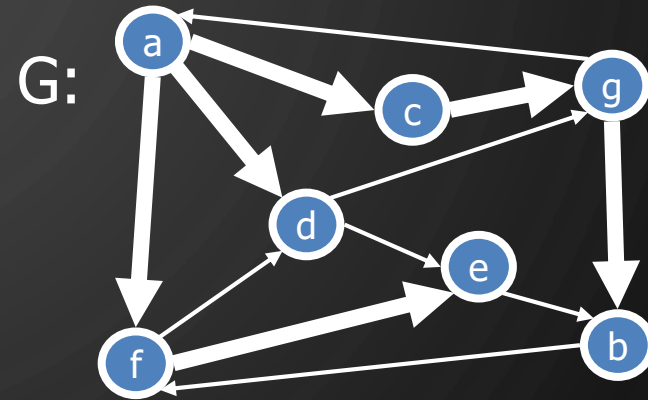
- Each vertex can reach all other vertices



# STRONG CONNECTIVITY ALGORITHM



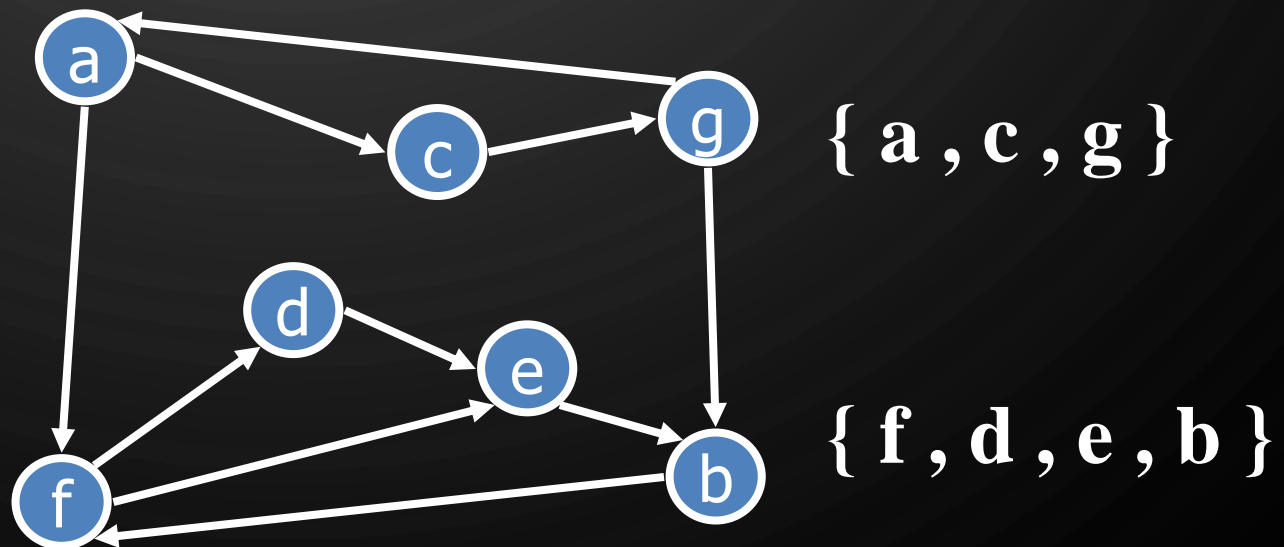
- Pick a vertex  $v$  in  $G$
- Perform a DFS from  $v$  in  $G$ 
  - If there's a  $w$  not visited, print "no"
- Let  $G'$  be  $G$  with edges reversed
- Perform a DFS from  $v$  in  $G'$ 
  - If there's a  $w$  not visited, print "no"
  - Else, print "yes"
- Running time:  $O(n + m)$



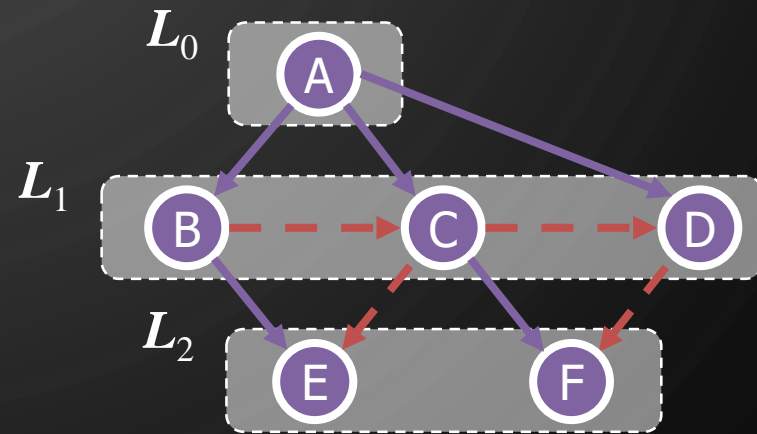
# STRONGLY CONNECTED COMPONENTS



- Maximal subgraphs such that each vertex can reach all other vertices in the subgraph
- Can also be done in  $O(n + m)$  time using DFS, but is more complicated (similar to biconnectivity).



# BREADTH-FIRST SEARCH



# BREADTH-FIRST SEARCH

- **Breadth-first search (BFS)** is a general technique for traversing a graph
- A BFS traversal of a graph  $G$ 
  - Visits all the vertices and edges of  $G$
  - Determines whether  $G$  is connected
  - Computes the connected components of  $G$
  - Computes a spanning forest of  $G$
- BFS on a graph with  $n$  vertices and  $m$  edges takes  $O(n + m)$  time
- BFS can be further extended to solve other graph problems
  - Find and report a path with the minimum number of edges between two given vertices
  - Find a simple cycle, if there is one

# BFS ALGORITHM

- The algorithm uses a mechanism for setting and getting “labels” of vertices and edges

## Algorithm BFS( $G$ )

**Input:** Graph  $G$

**Output:** Labeling of the edges and partition of the vertices of  $G$

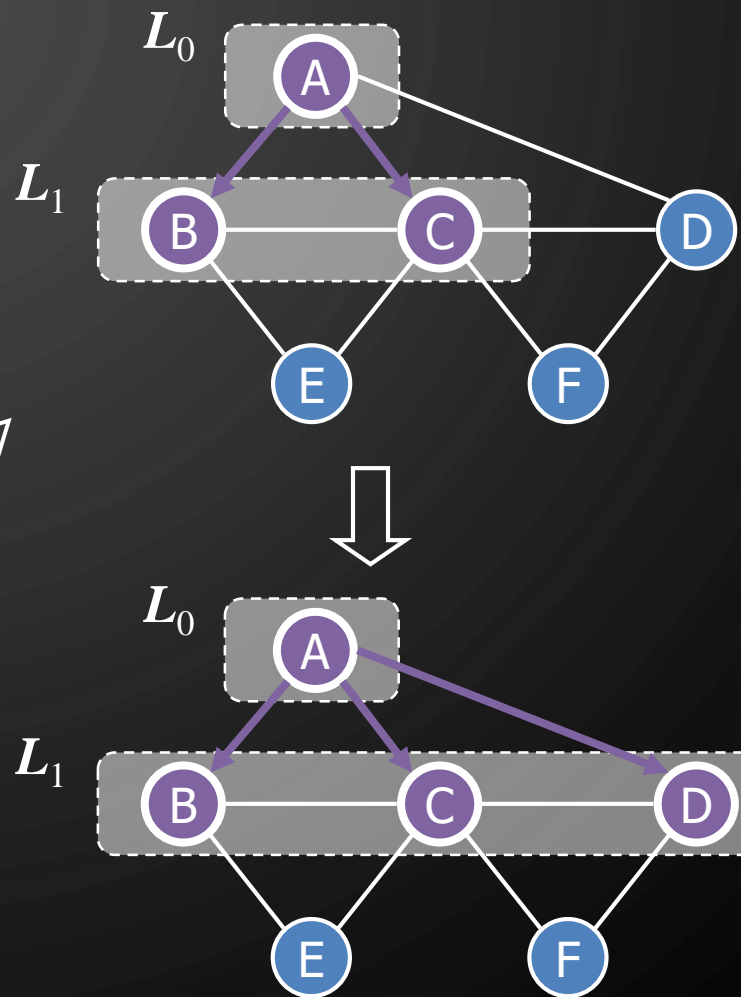
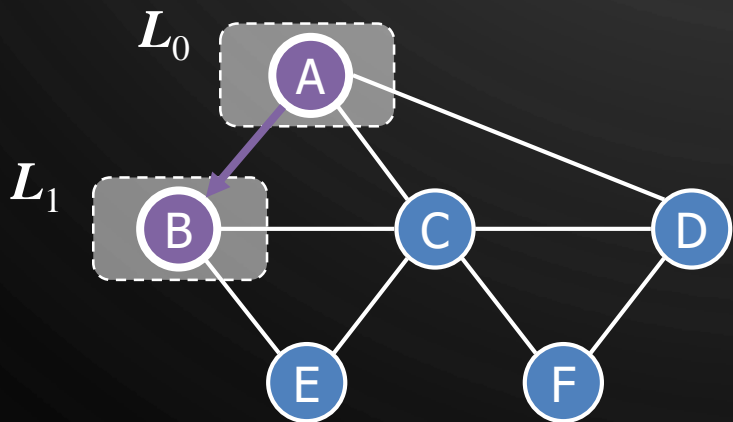
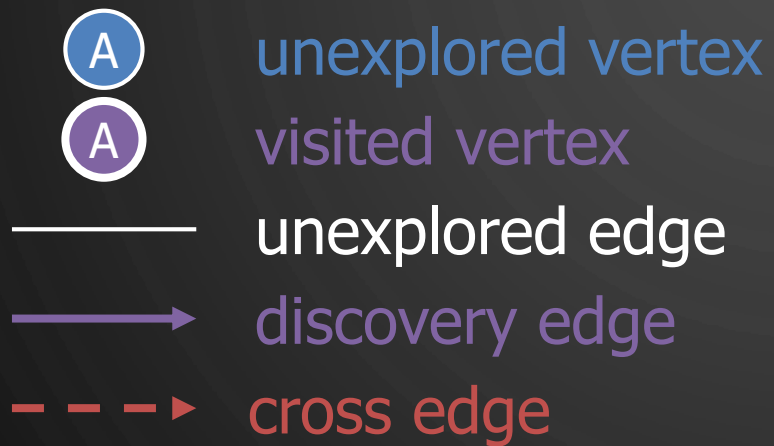
1. **for each**  $v \in G.vertices()$  **do**
2.   setLabel( $v$ , *UNEXPLORED*)
3. **for each**  $e \in G.edges()$  **do**
4.   setLabel( $e$ , *UNEXPLORED*)
5. **for each**  $v \in G.vertices()$  **do**
6.   **if** getLabel( $v$ ) = *UNEXPLORED* **then**
7.     BFS( $G, v$ )

## Algorithm BFS( $G, s$ )






**Input:** Graph  $G$ , a start vertex  $s$

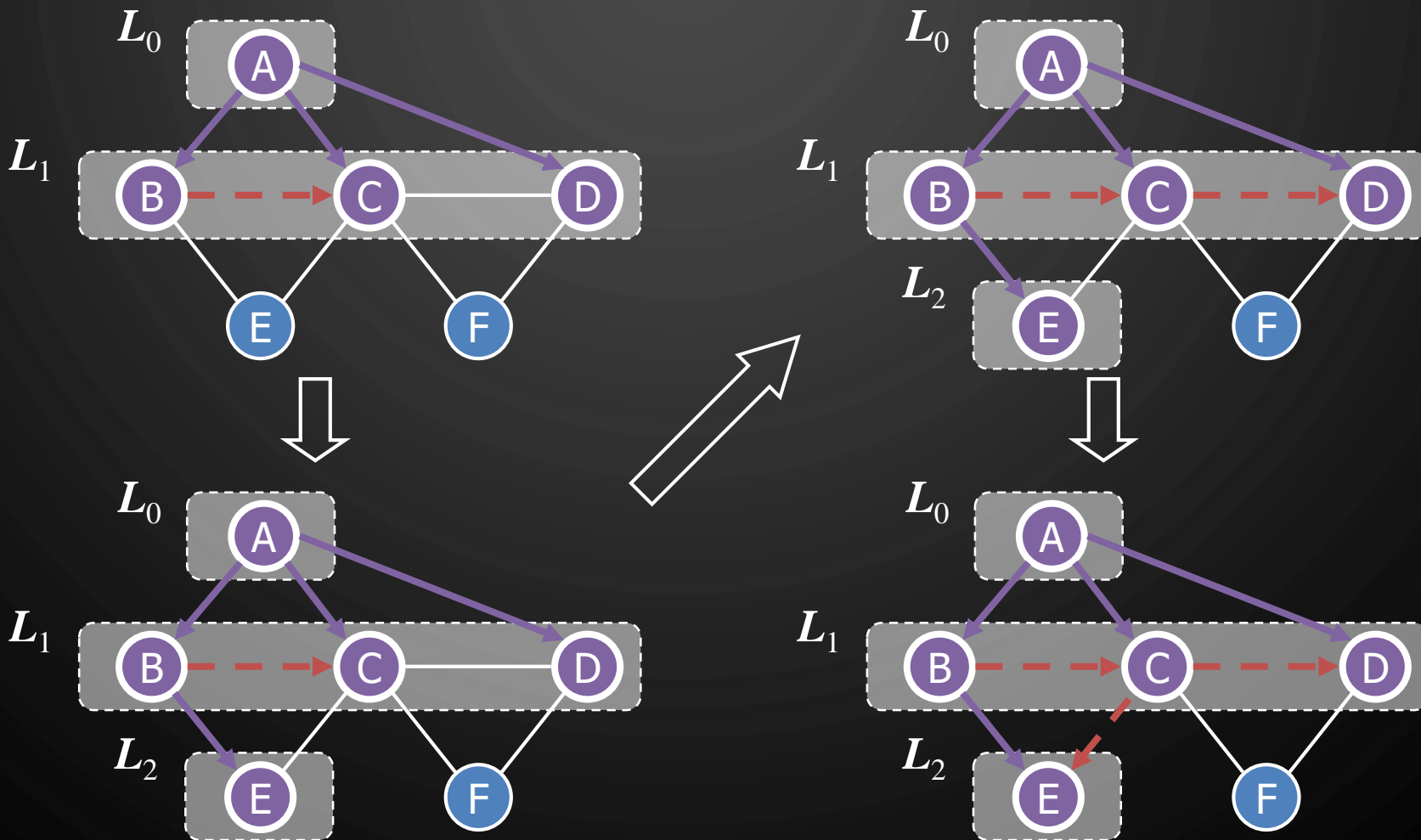
1. List  $L_0 \leftarrow \{s\}$
2. setLabel( $s$ , *VISITED*)
3.  $i \leftarrow 0$
4. **while**  $\neg L_i.isEmpty()$  **do**
5.   List  $L_{i+1} \leftarrow \emptyset$
6.   **for each**  $v \in L_i$  **do**
7.     **for each**  $e \in G.outgoingEdges(v)$  **do**
8.       **if** getLabel( $e$ ) = *UNEXPLORED* **then**
9.          $w \leftarrow G.opposite(v, e)$
10.        **if** getLabel( $w$ ) = *UNEXPLORED* **then**
11.         setLabel( $e$ , *DISCOVERY*)
12.         setLabel( $w$ , *VISITED*)
13.          $L_{i+1} \leftarrow L_{i+1} \cup \{w\}$
14.        **else**
15.         setLabel( $e$ , *CROSS*)
16.     $i \leftarrow i + 1$

# EXAMPLE








# EXAMPLE

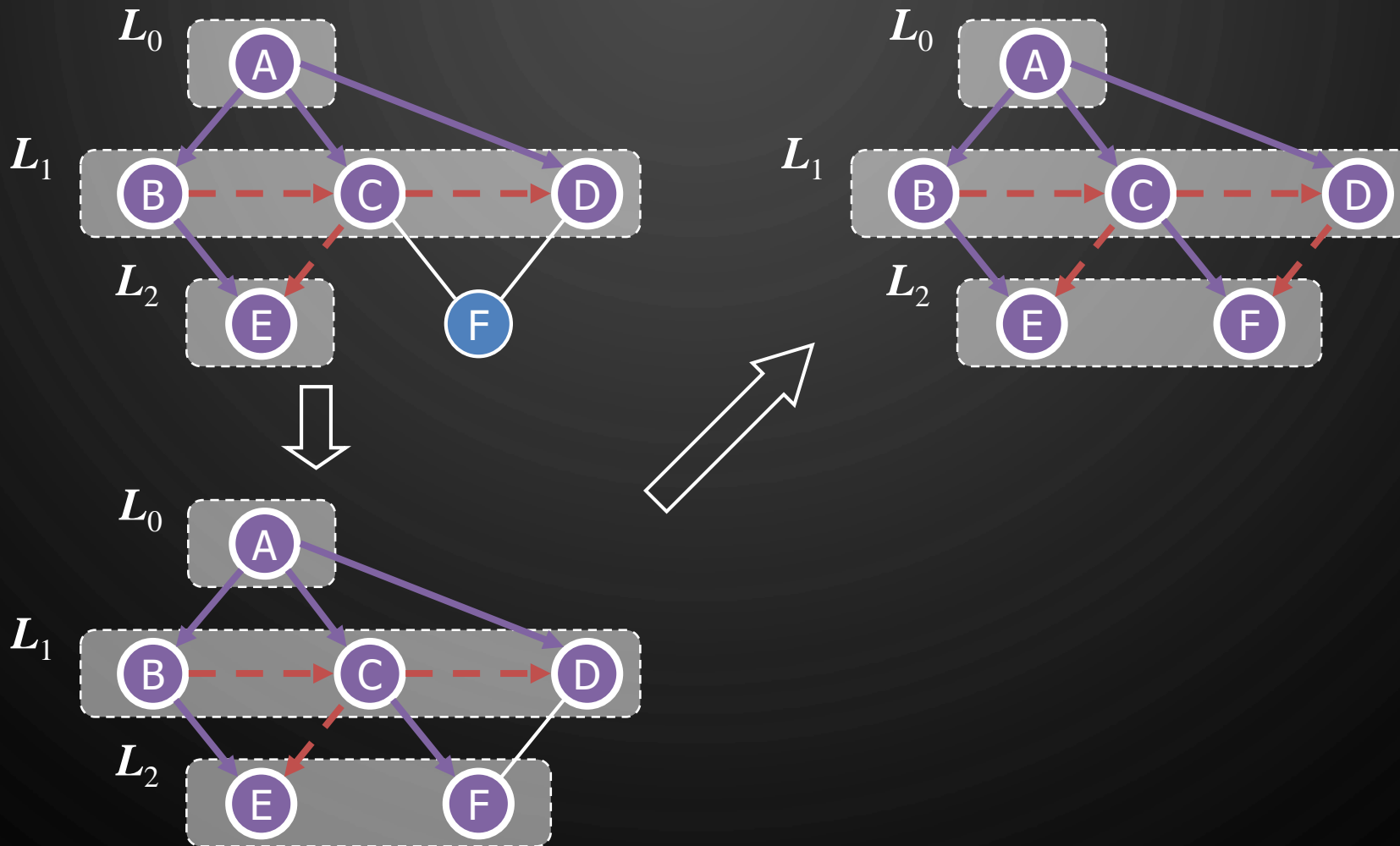
-  unexplored vertex
-  visited vertex
-  unexplored edge
-  discovery edge
-  cross edge





# EXAMPLE

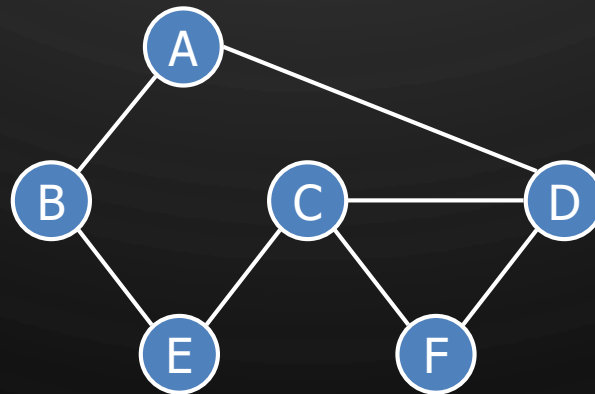
-  unexplored vertex
-  visited vertex
-  unexplored edge
-  discovery edge
-  cross edge



# EXERCISE

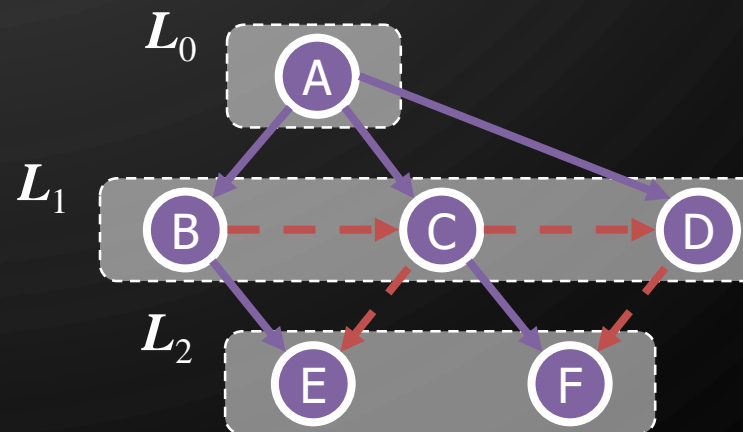
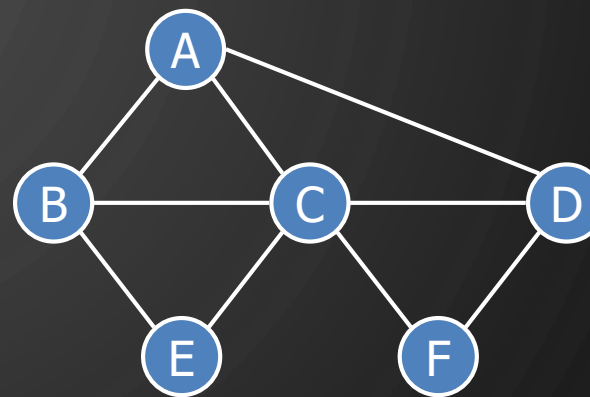
## BFS ALGORITHM

- Perform BFS of the following graph, start from vertex F
  - Assume adjacent edges are processed in alphabetical order
  - Number vertices in the order they are visited and note the level they are in
  - Label edges as discovery or cross edges



# PROPERTIES

- Notation
  - $G_s$ : connected component of  $s$
- Property 1
  - $\text{BFS}(G, s)$  visits all the vertices and edges of  $G_s$
- Property 2
  - The discovery edges labeled by  $\text{BFS}(G, s)$  form a spanning tree  $T_s$  of  $G_s$
- Property 3
  - For each vertex  $v \in L_i$ 
    - The path of  $T_s$  from  $s$  to  $v$  has  $i$  edges
    - Every path from  $s$  to  $v$  in  $G_s$  has at least  $i$  edges



# ANALYSIS

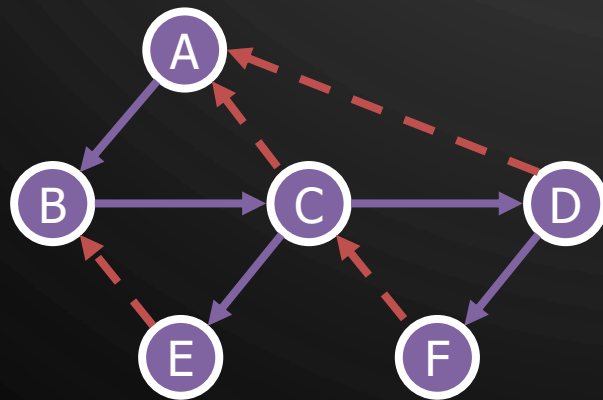
- Setting/getting a vertex/edge label takes  $O(1)$  time
- Each vertex is labeled twice
  - once as UNEXPLORED
  - once as VISITED
- Each edge is labeled twice
  - once as UNEXPLORED
  - once as DISCOVERY or CROSS
- Each vertex is inserted once into a sequence  $L_i$
- Method `outgoingEdges()` is called once for each vertex
- BFS runs in  $O(n + m)$  time provided the graph is represented by the adjacency list structure
  - Recall that  $\sum_v \deg(v) = 2m$

# APPLICATIONS

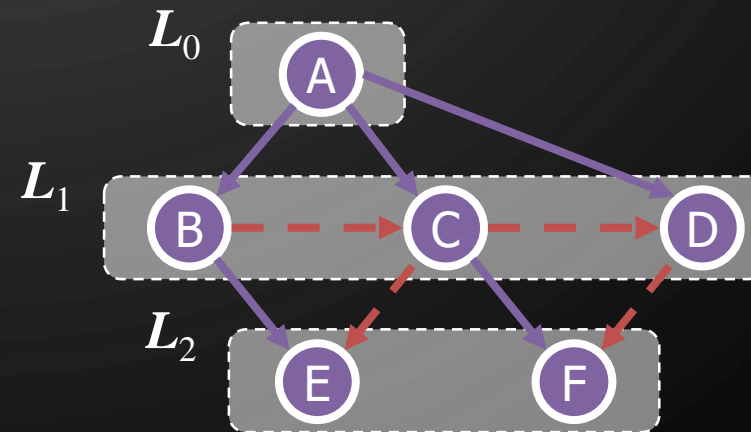
- Using the template method pattern, we can specialize the BFS traversal of a graph  $G$  to solve the following problems in  $O(n + m)$  time
  - Compute the connected components of  $G$
  - Compute a spanning forest of  $G$
  - Find a simple cycle in  $G$ , or report that  $G$  is a forest
  - Given two vertices of  $G$ , find a path in  $G$  between them with the minimum number of edges, or report that no such path exists

# DFS VS. BFS

Applications	DFS	BFS
Spanning forest, connected components, paths, cycles	✓	✓
Shortest paths		✓
Biconnected components	✓	



DFS

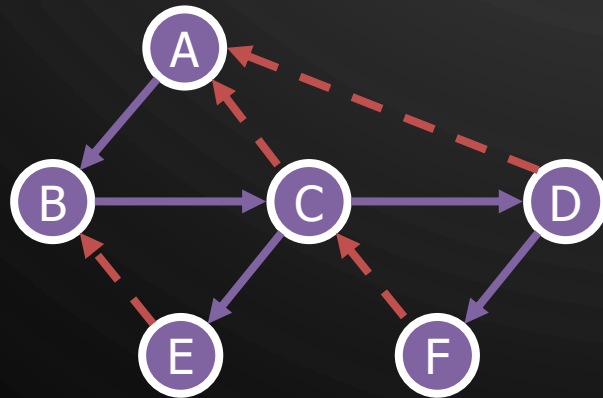


BFS

# DFS VS. BFS

## Back edge ( $v, w$ )

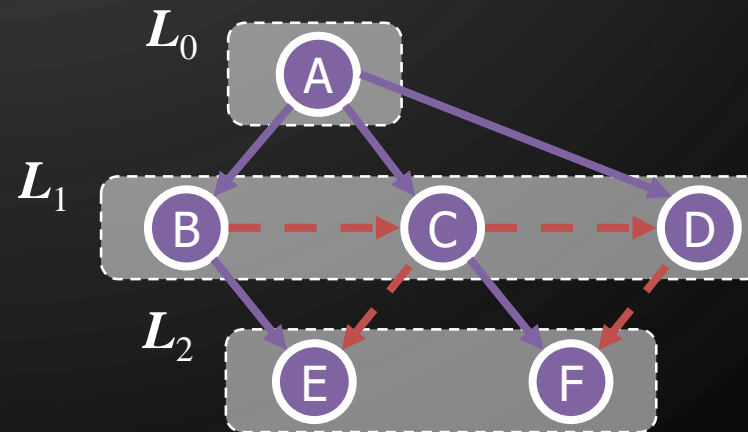
- $w$  is an ancestor of  $v$  in the tree of discovery edges



DFS

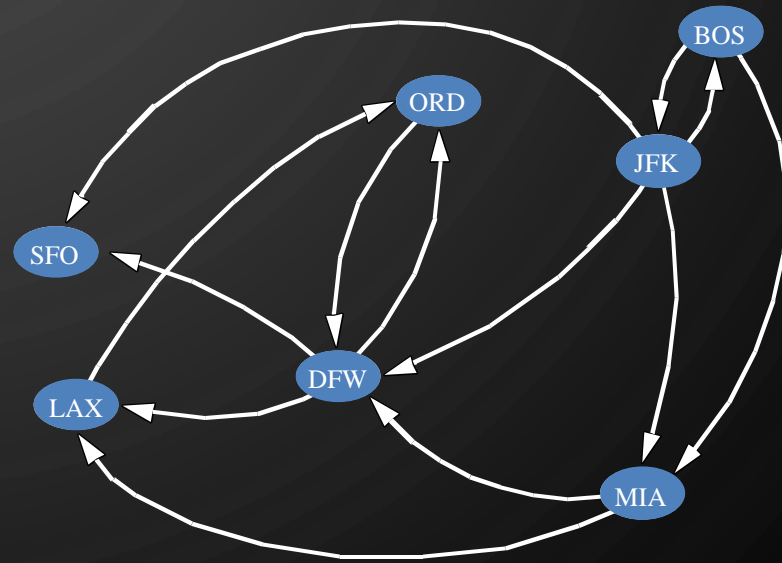
## Cross edge ( $v, w$ )

- $w$  is in the same level as  $v$  or in the next level in the tree of discovery edges



BFS

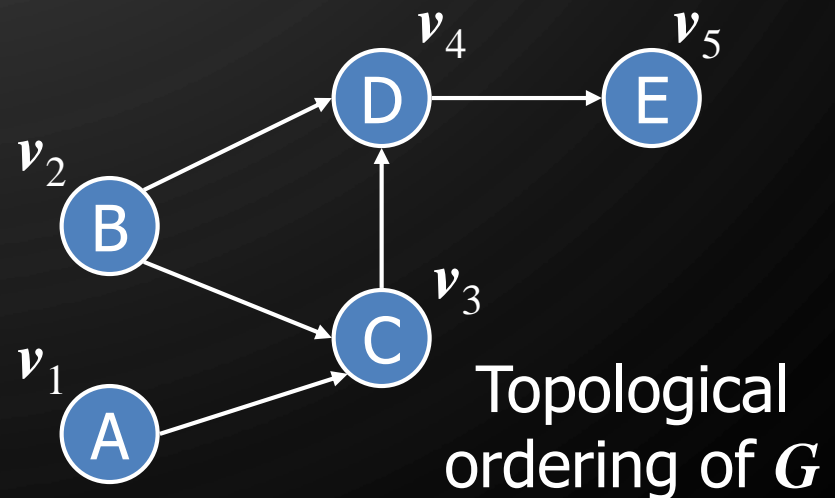
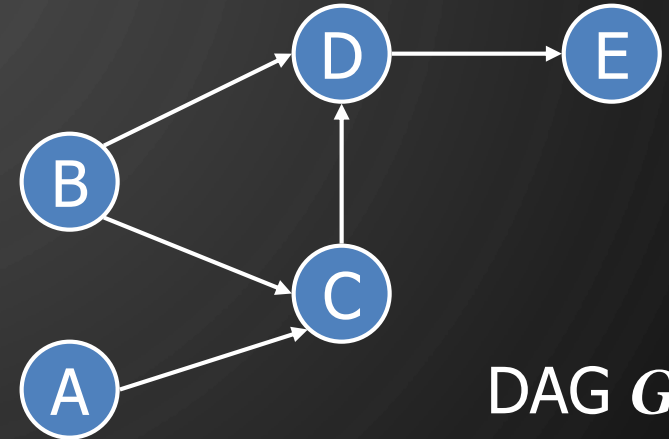
# TOPOLOGICAL ORDERING





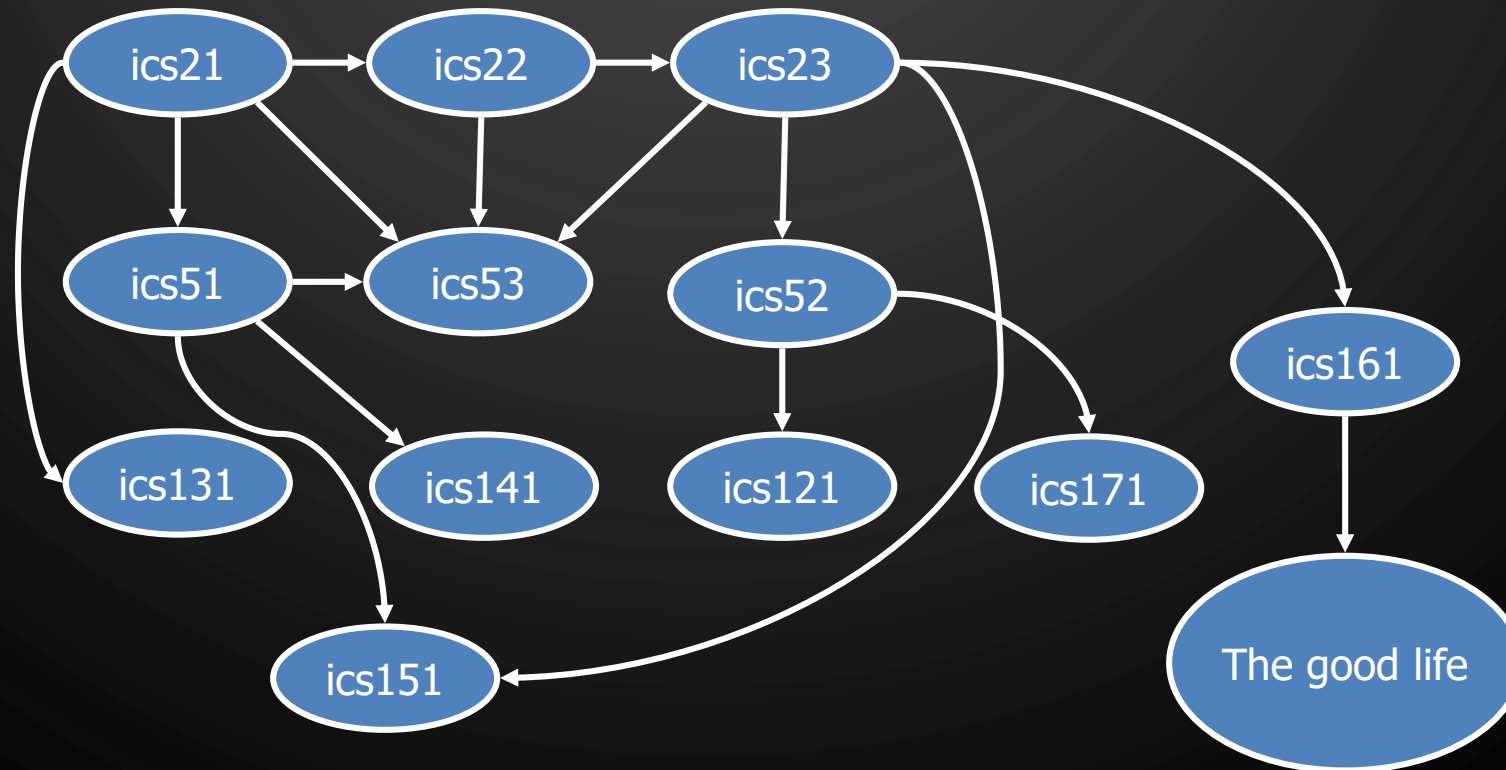
# DAGS AND TOPOLOGICAL ORDERING

- A **directed acyclic graph (DAG)** is a digraph that has no directed cycles
- A topological ordering of a digraph is a numbering
  - $v_1, \dots, v_n$
  - Of the vertices such that for every edge  $(v_i, v_j)$ , we have  $i < j$
- Example: in a task scheduling digraph, a topological ordering a task sequence that satisfies the precedence constraints
- Theorem - A digraph admits a topological ordering if and only if it is a DAG



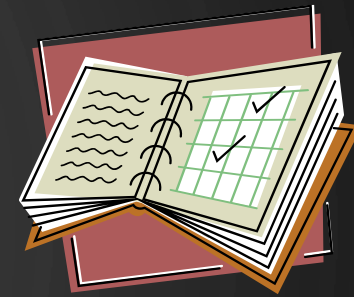
# APPLICATION

- Scheduling: edge  $(a, b)$  means task  $a$  must be completed before  $b$  can be started

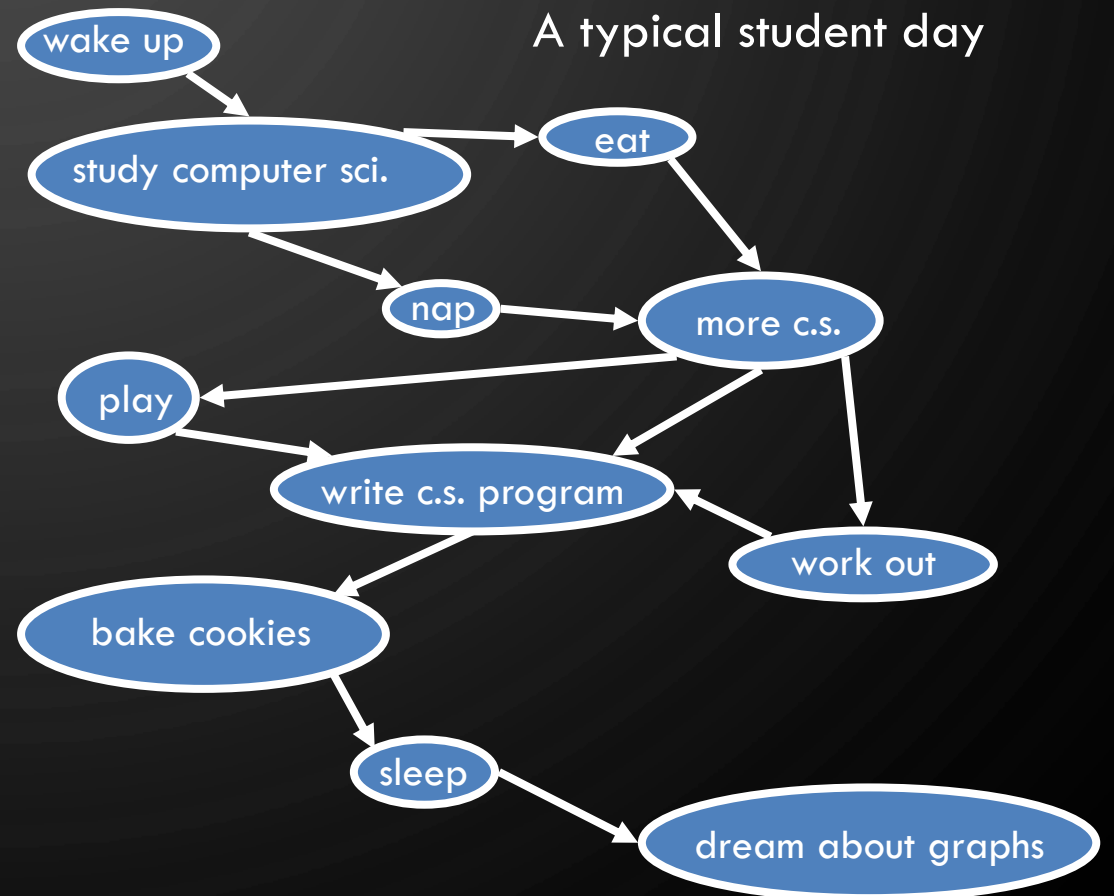


# EXERCISE

## TOPOLOGICAL SORTING

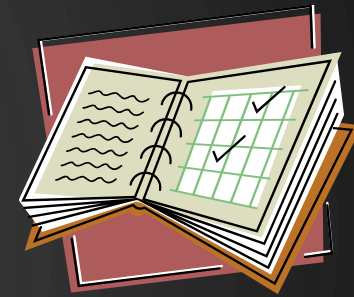


- Number vertices, so that  $(u, v)$  in  $E$  implies  $u < v$

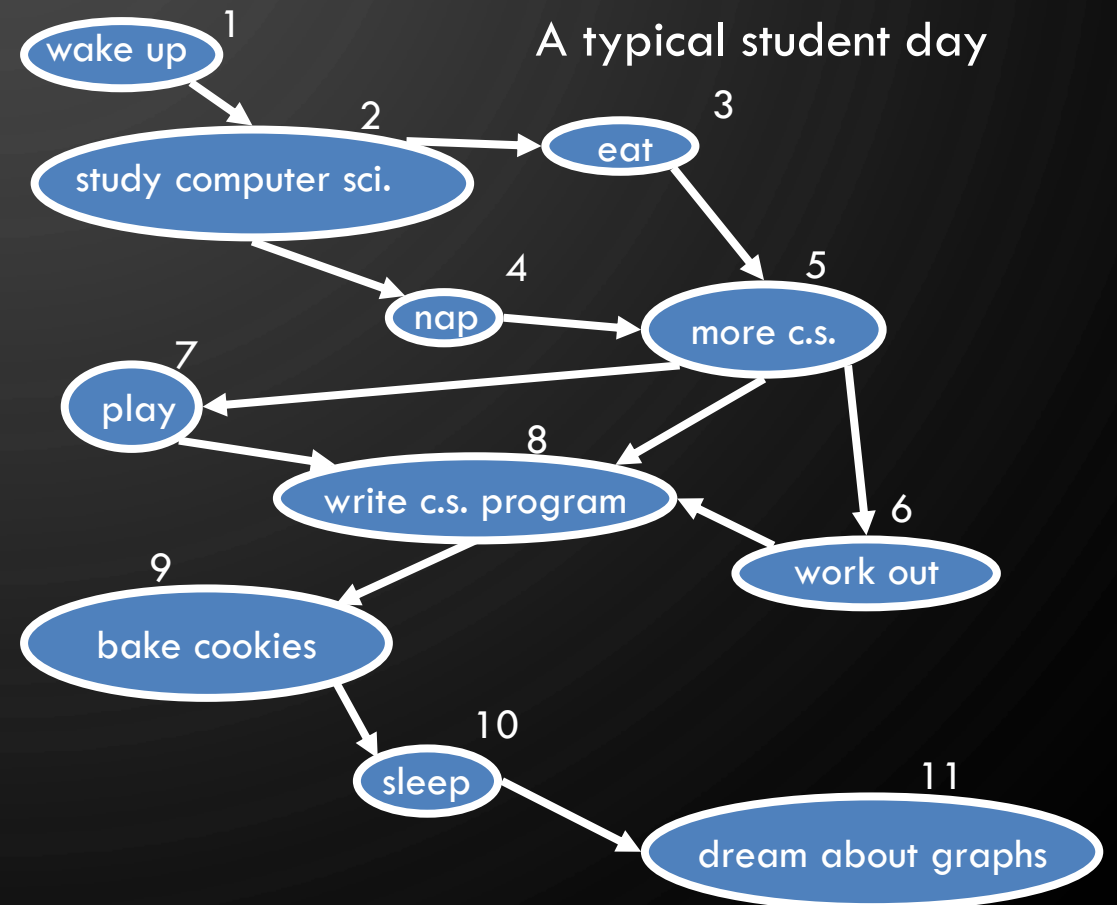


# EXERCISE

## TOPOLOGICAL SORTING



- Number vertices, so that  $(u, v)$  in  $E$  implies  $u < v$



# ALGORITHM FOR TOPOLOGICAL SORTING

**Algorithm** TopologicalSort( $G$ )

**Input:** Directed Acyclic Graph (DAG)  $G$

**Output:** Topological ordering of  $G$

1.  $H \leftarrow G$
2.  $n \leftarrow G.\text{numVertices}()$
3. **while**  $\neg H.\text{isEmpty}()$  **do**
4.     Let  $v$  be a vertex with no outgoing edges
5.     Label  $v \leftarrow n$
6.      $n \leftarrow n - 1$
7.      $H.\text{removeVertex}(v)$

# IMPLEMENTATION WITH DFS

- Simulate the algorithm by using depth-first search
- $O(n + m)$  time.

**Algorithm** `topologicalDFS(G)`

**Input:** DAG  $G$

**Output:** Topological ordering of  $G$

1.  $n \leftarrow G.\text{numVertices}()$
2. Initialize all vertices as **UNEXPLORED**
3. **for each** vertex  $v \in G.\text{vertices}()$  **do**
4.     **if** `getLabel(v) = UNEXPLORED` **then**
5.         `topologicalDFS(G, v)`

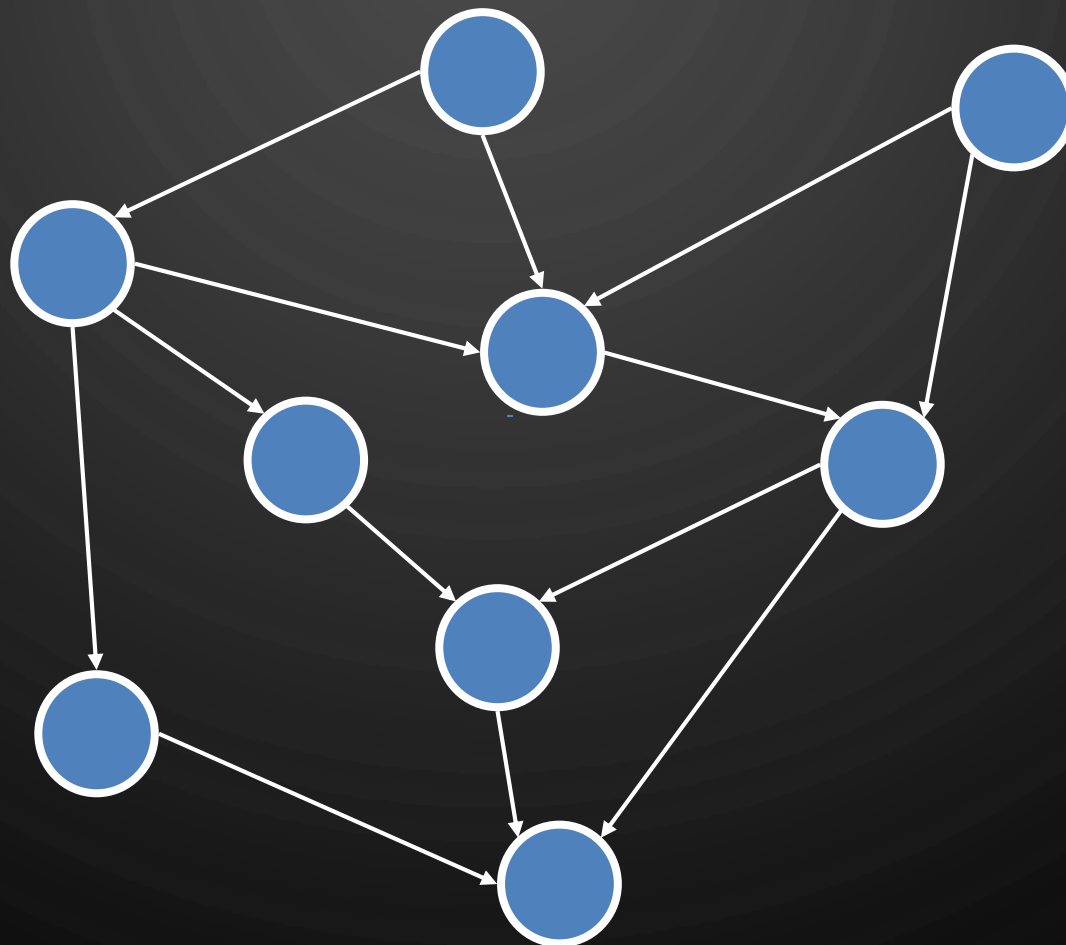
**Algorithm** `topologicalDFS(G, v)`

**Input:** DAG  $G$ , start vertex  $v$

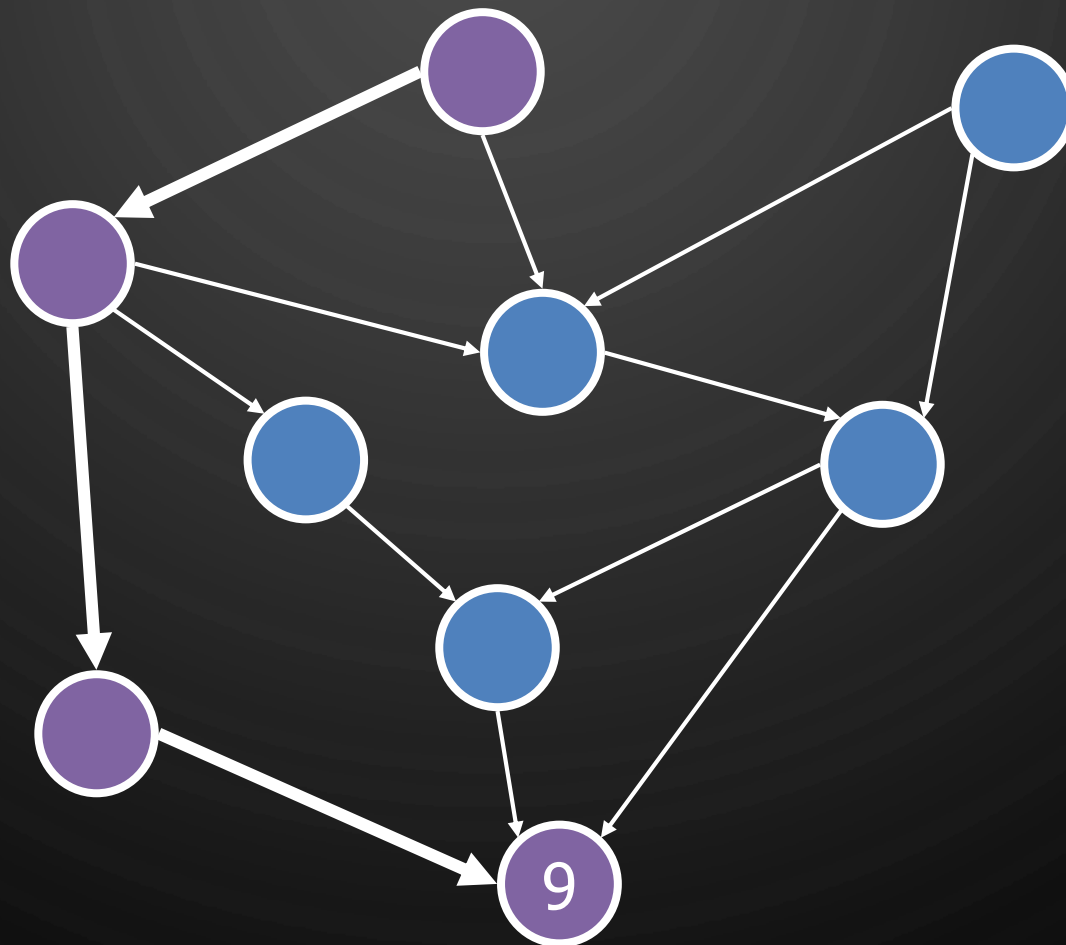
**Output:** Labeling of the vertices of  $G$   
in the connected component of  $v$

1. `setLabel(v, VISITED)`
2. **for each**  $e \in G.\text{outgoingEdges}(v)$  **do**
3.      $w \leftarrow G.\text{opposite}(v, e)$
4.     **if** `getLabel(w) = UNEXPLORED` **then**
5.         //  $e$  is a discovery edge
6.         `topologicalDFS(G, w)`
7.     **else**
8.         //  $e$  is a forward, cross, or back edge
9.     Label  $v$  with topological number  $n$
10.  $n \leftarrow n - 1$

# TOPOLOGICAL SORTING EXAMPLE

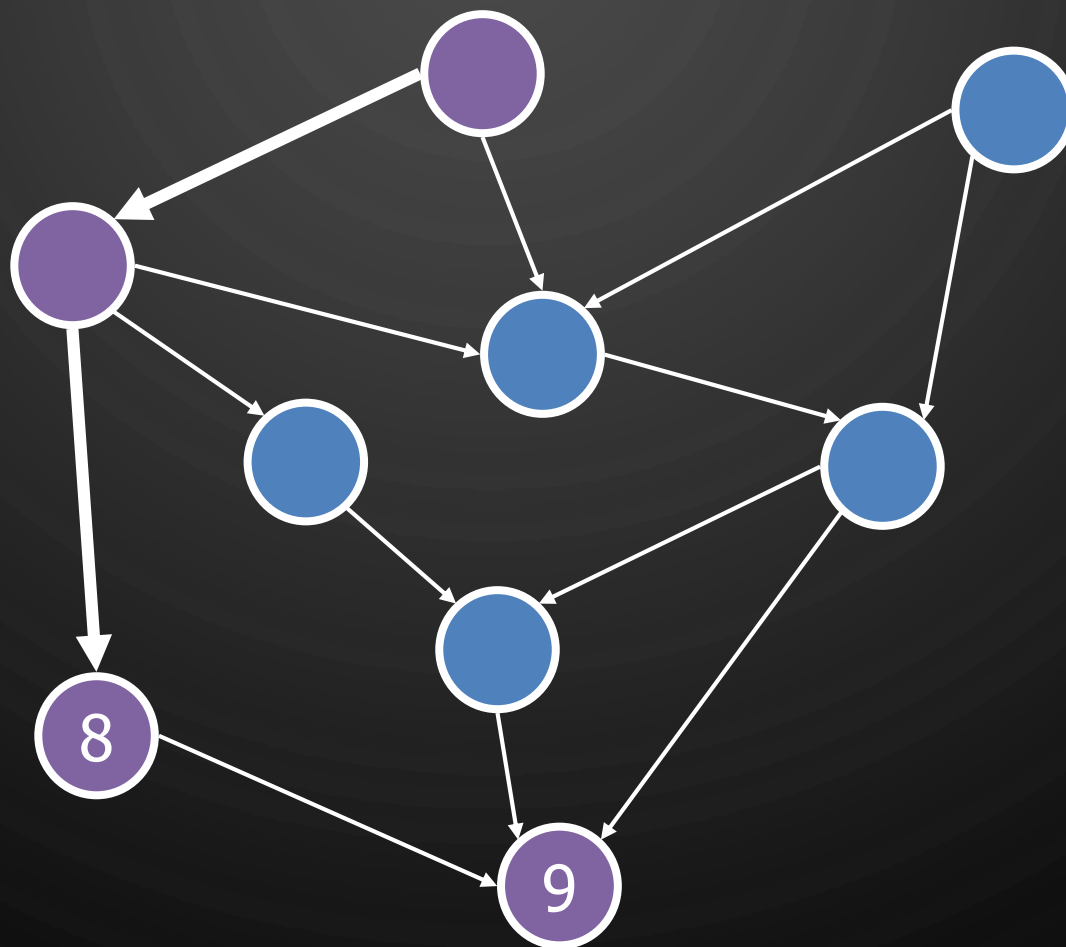


# TOPOLOGICAL SORTING EXAMPLE

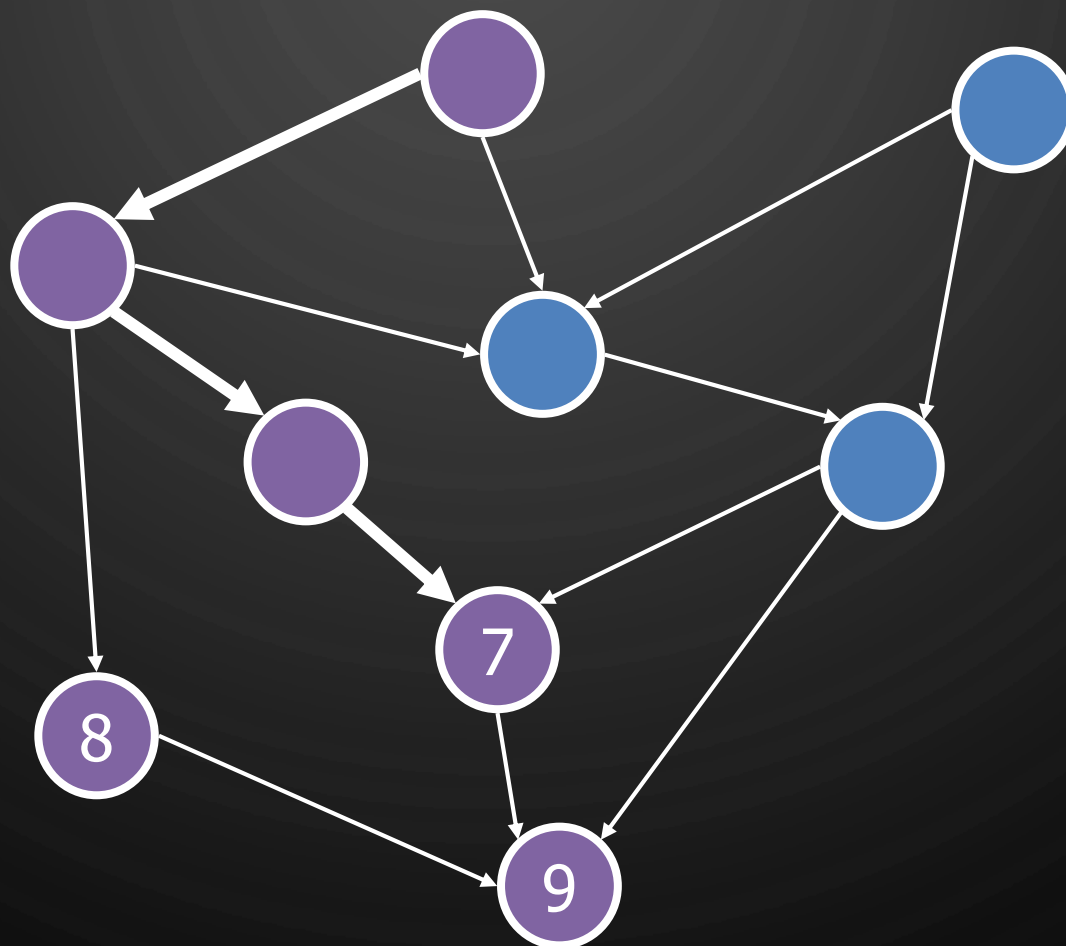




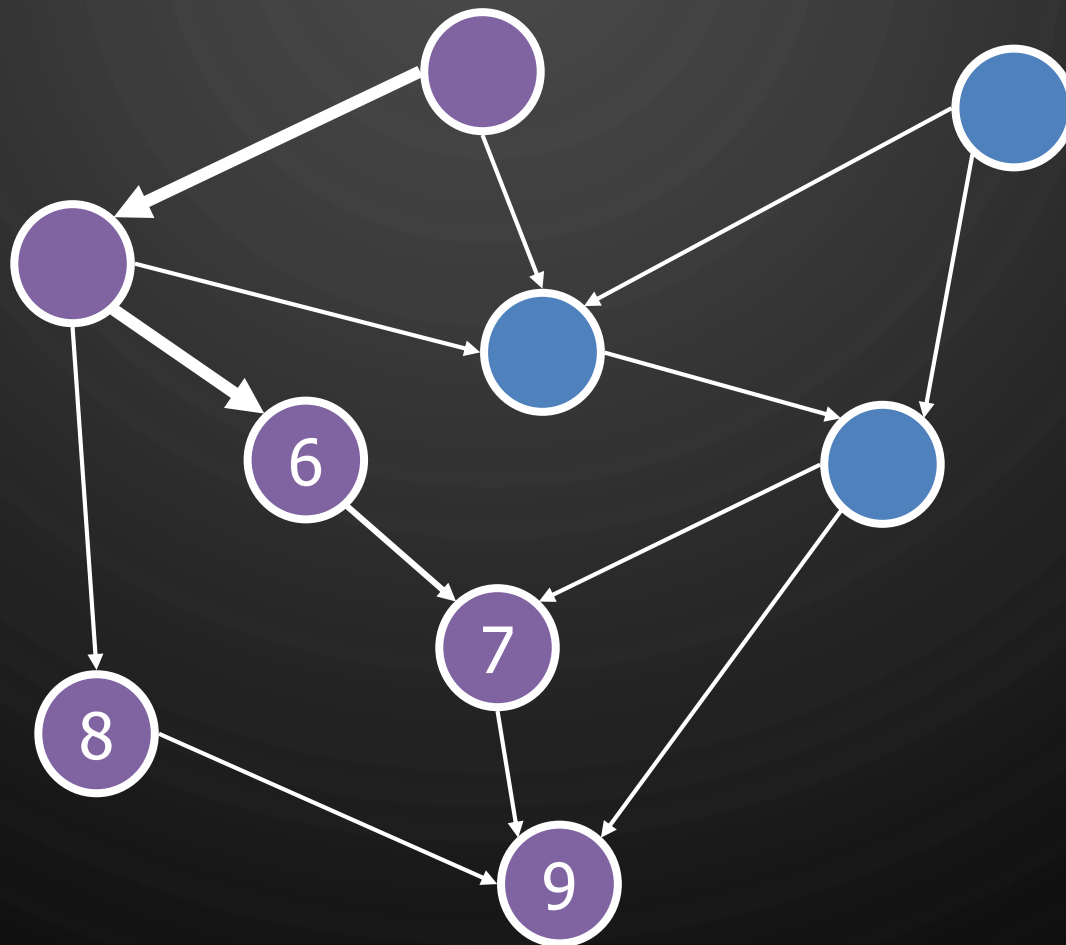
# TOPOLOGICAL SORTING EXAMPLE



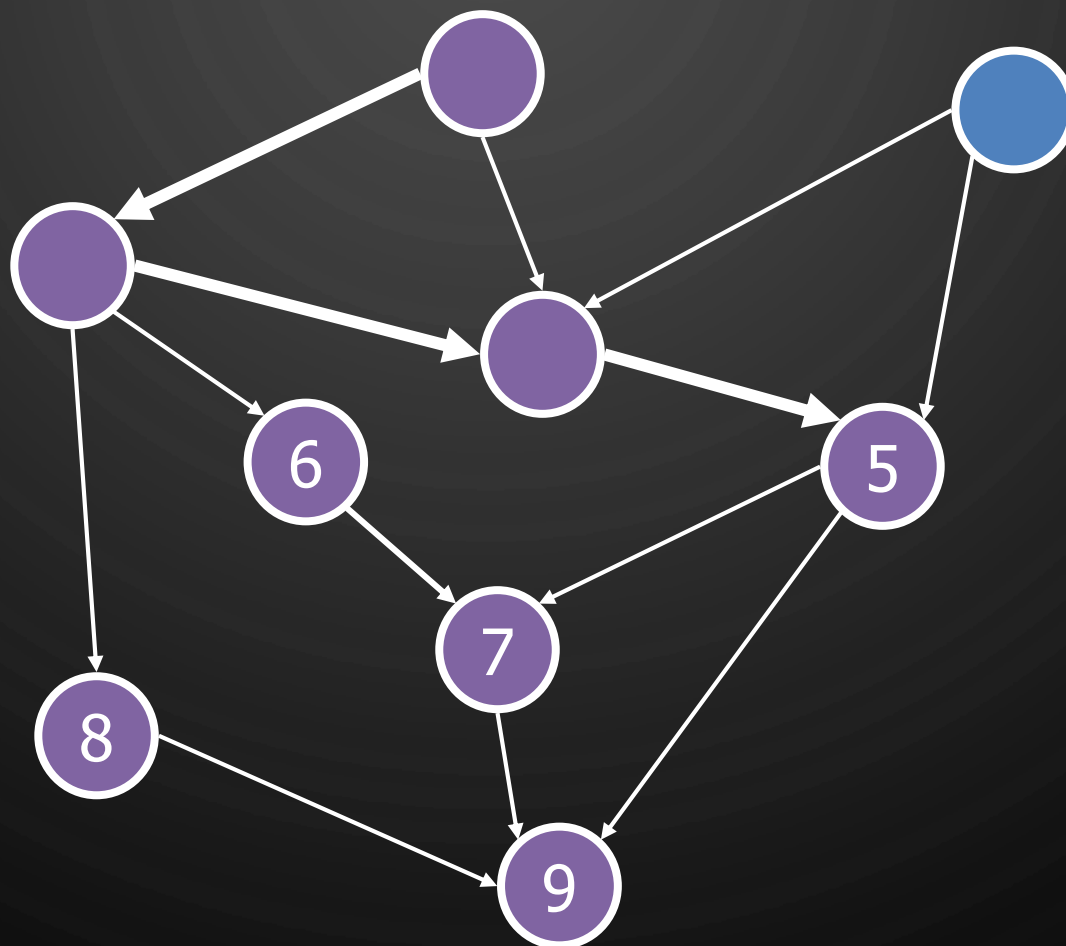
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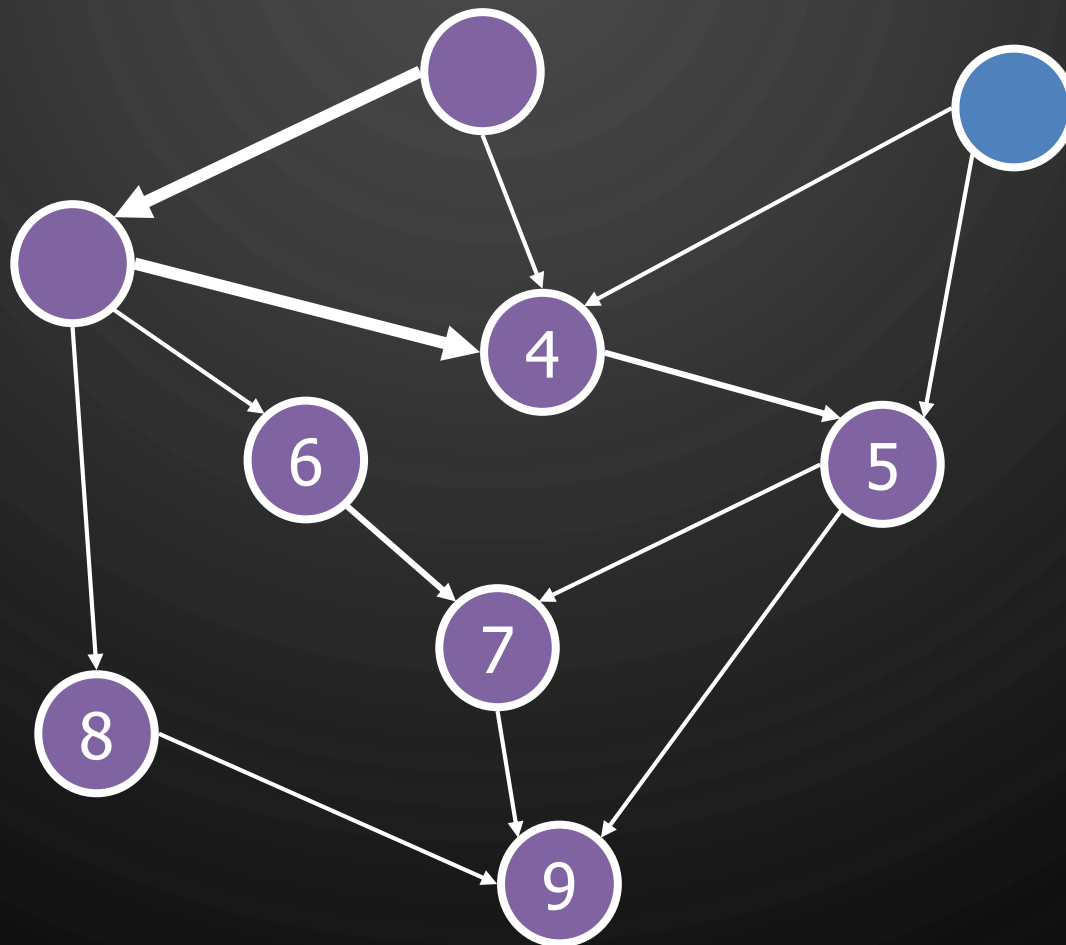
# TOPOLOGICAL SORTING EXAMPLE



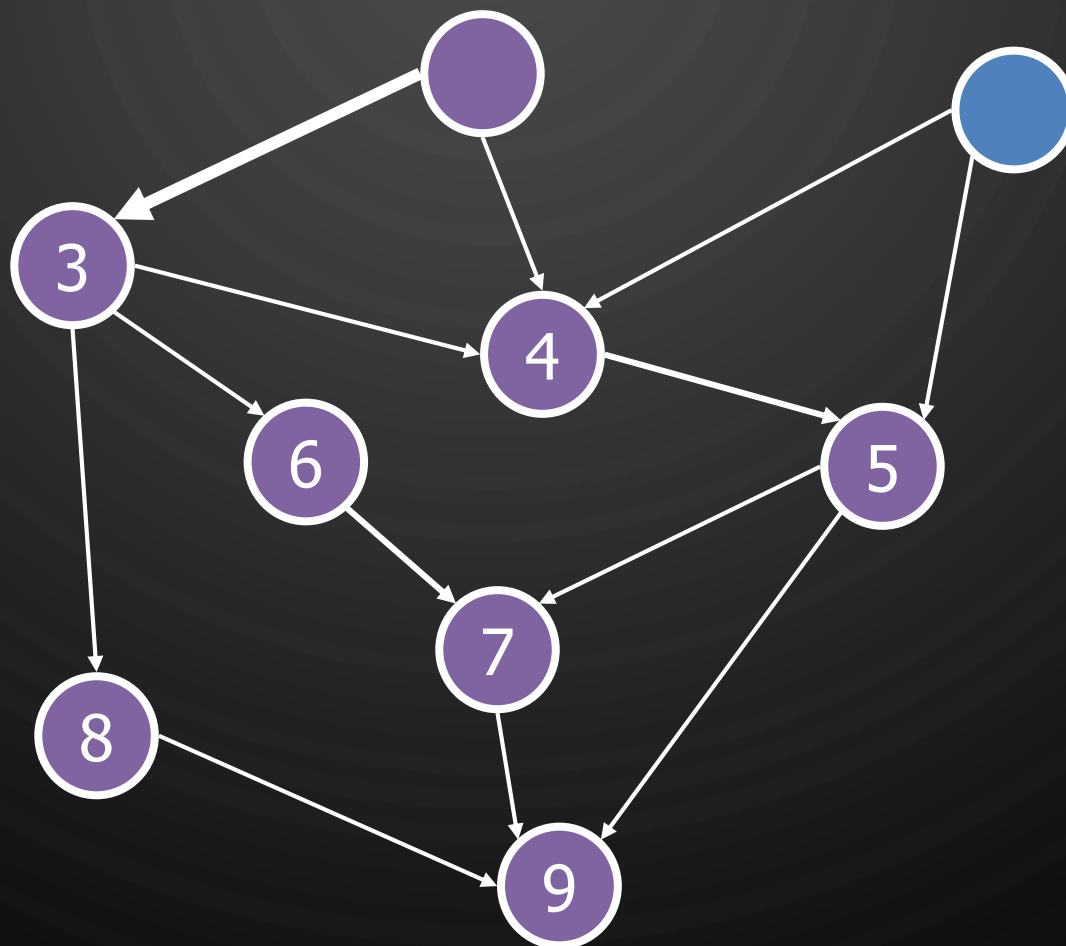
# TOPOLOGICAL SORTING EXAMPLE



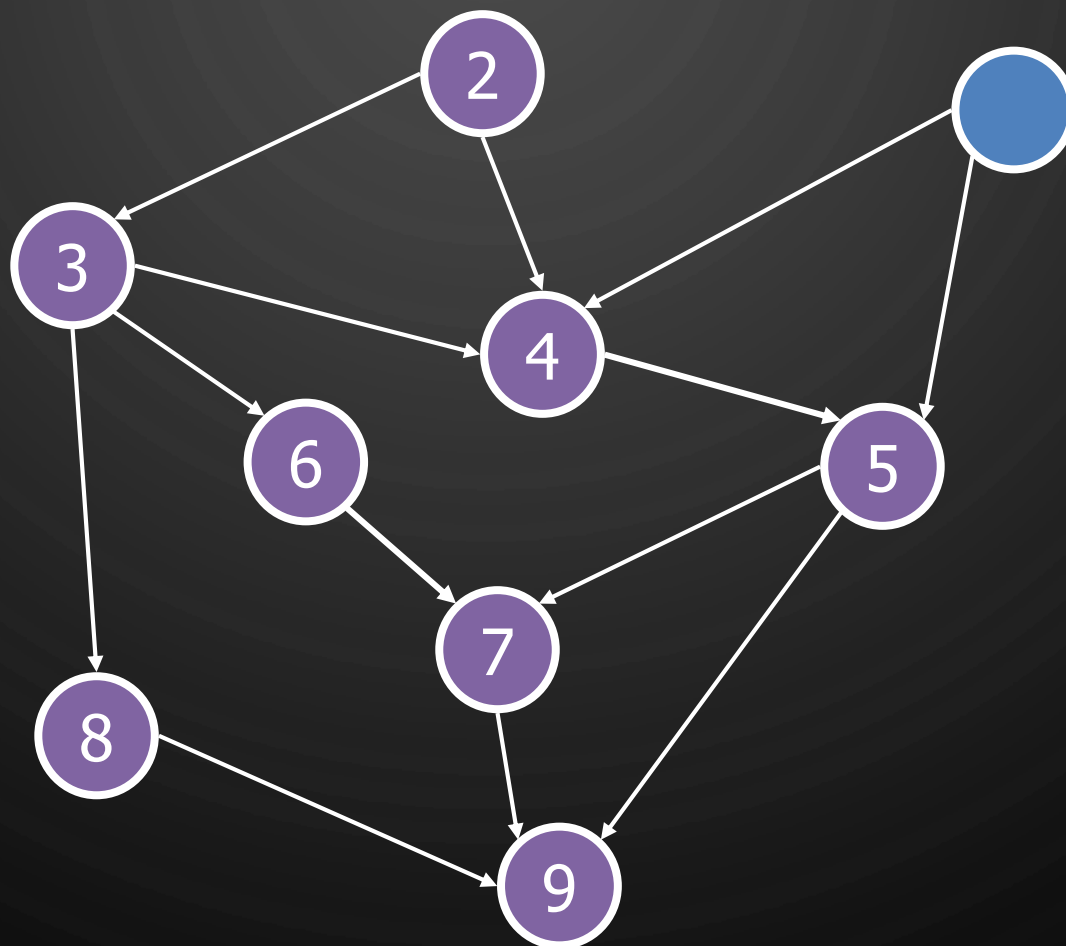
# TOPOLOGICAL SORTING EXAMPLE



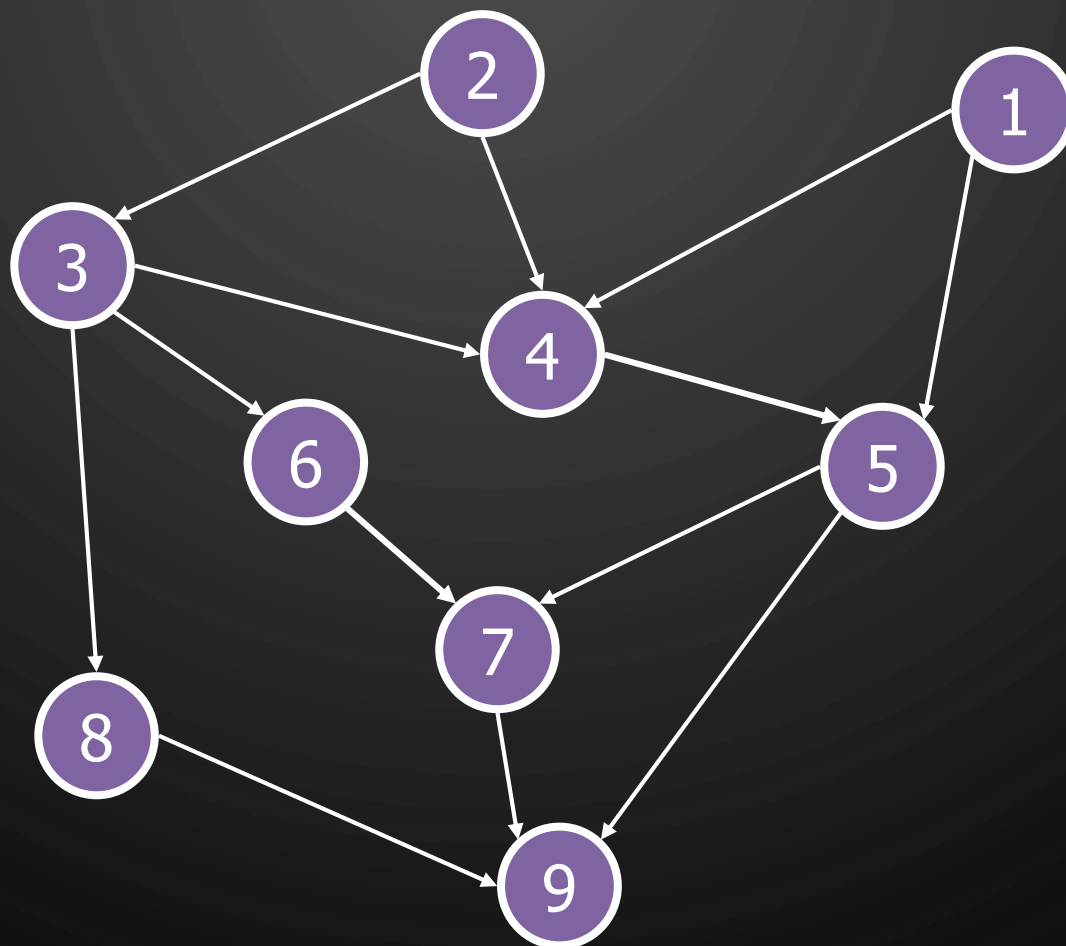
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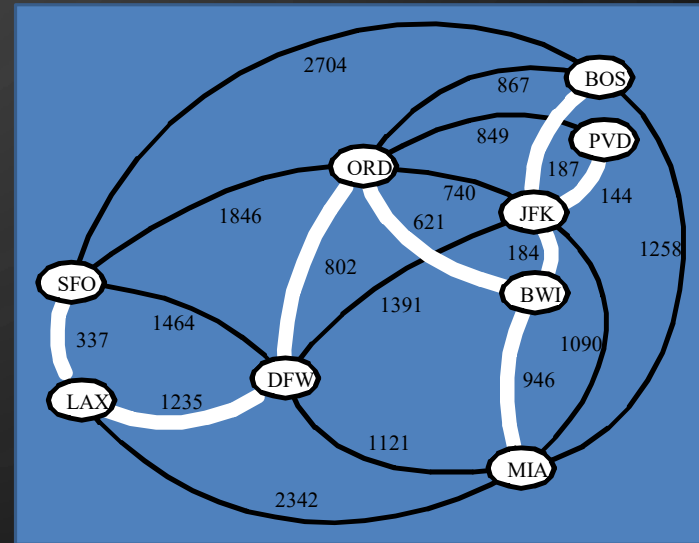


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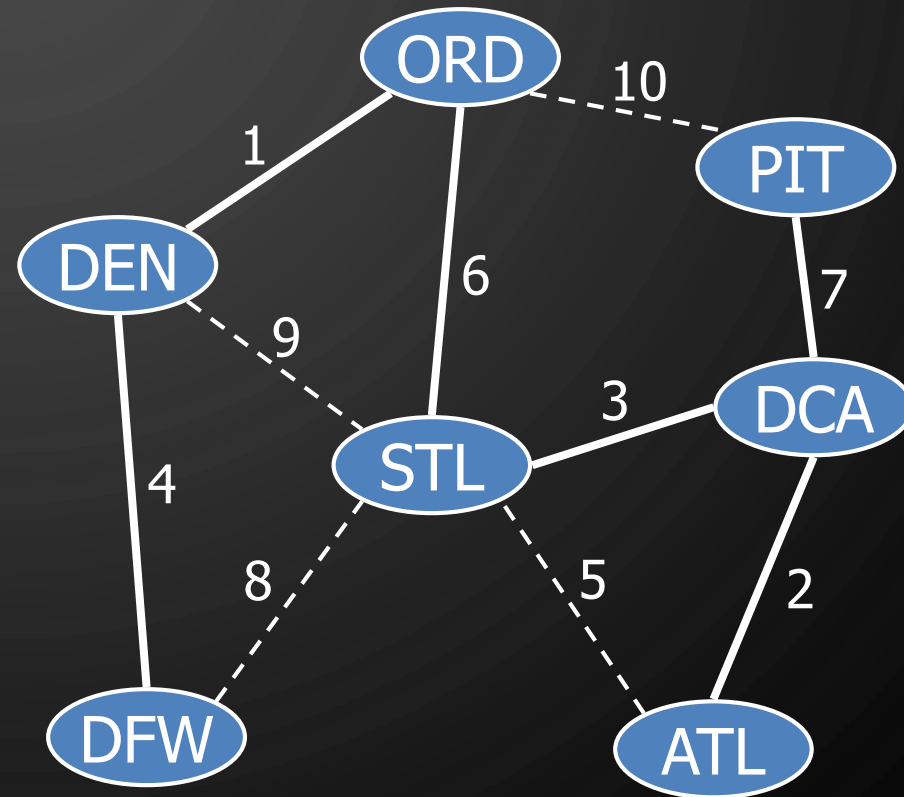


# MINIMUM SPANNING TREES



# MINIMUM SPANNING TREE

- Minimum spanning tree (MST)
  - Spanning tree of a weighted graph with minimum total edge weight
- Applications
  - Communications networks
  - Transportation networks

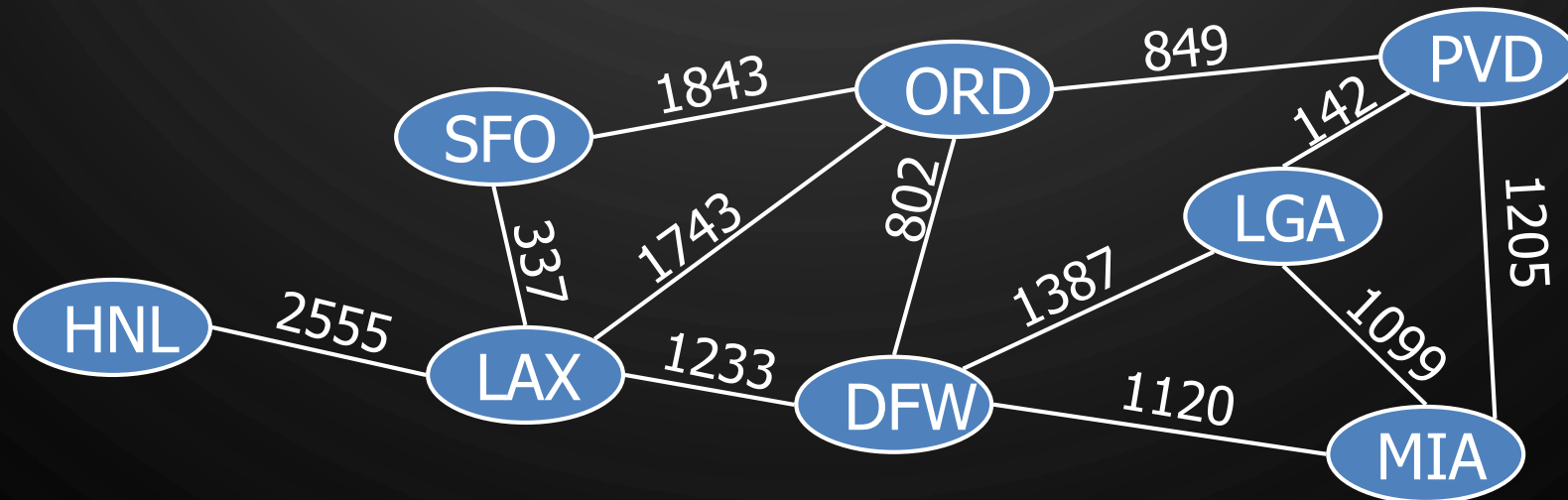


# EXERCISE

## MST



- Show an MST of the following graph.



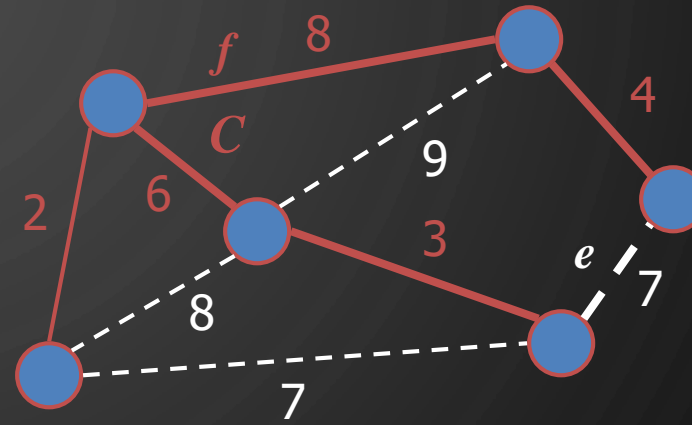
# CYCLE PROPERTY

- **Cycle Property:**

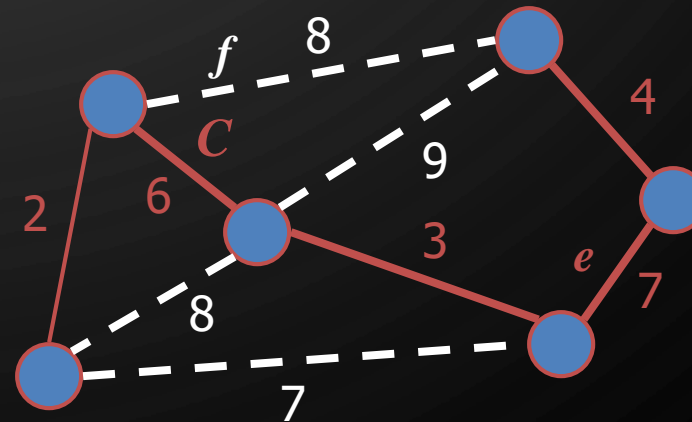
- Let  $T$  be a minimum spanning tree of a weighted graph  $G$
- Let  $e$  be an edge of  $G$  that is not in  $T$  and  $C$  let be the cycle formed by  $e$  with  $T$
- For every edge  $f$  of  $C$ ,  
 $weight(f) \leq weight(e)$

- **Proof by contradiction:**

- If  $weight(f) > weight(e)$  we can get a spanning tree of smaller weight by replacing  $e$  with  $f$



Replacing  $f$  with  $e$  yields a better spanning tree



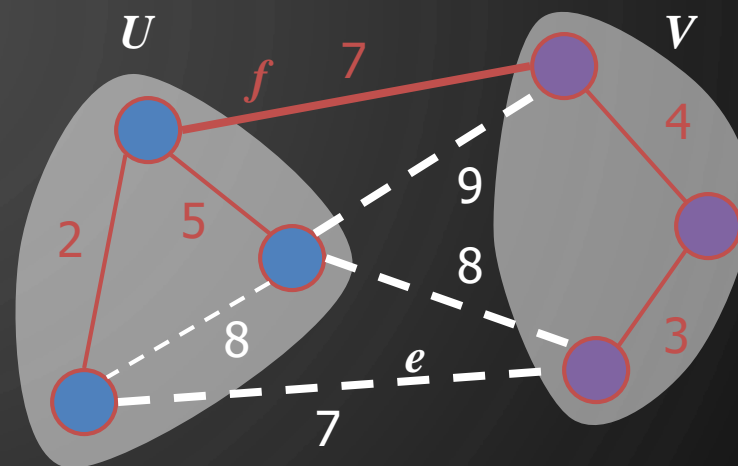
# PARTITION PROPERTY

- **Partition Property:**

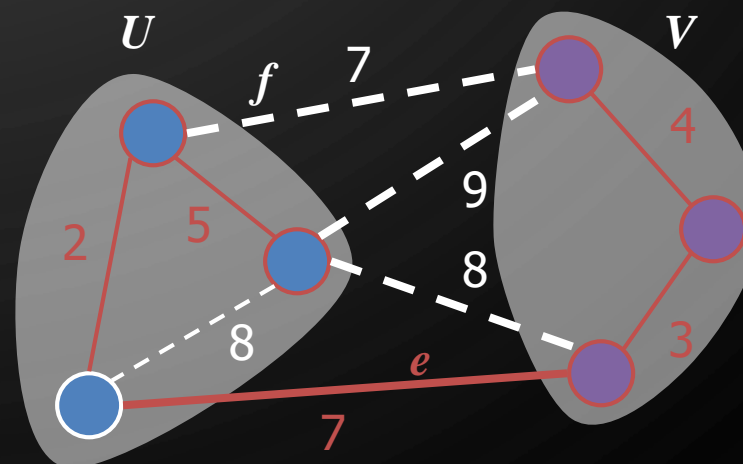
- Consider a partition of the vertices of  $G$  into subsets  $U$  and  $V$
- Let  $e$  be an edge of minimum weight across the partition
- There is a minimum spanning tree of  $G$  containing edge  $e$

- **Proof by contradiction:**

- Let  $T$  be an MST of  $G$
- If  $T$  does not contain  $e$ , consider the cycle  $C$  formed by  $e$  with  $T$  and let  $f$  be an edge of  $C$  across the partition
- By the cycle property,  
$$\text{weight}(f) \leq \text{weight}(e)$$
- Thus,  $\text{weight}(f) = \text{weight}(e)$
- We obtain another MST by replacing  $f$  with  $e$

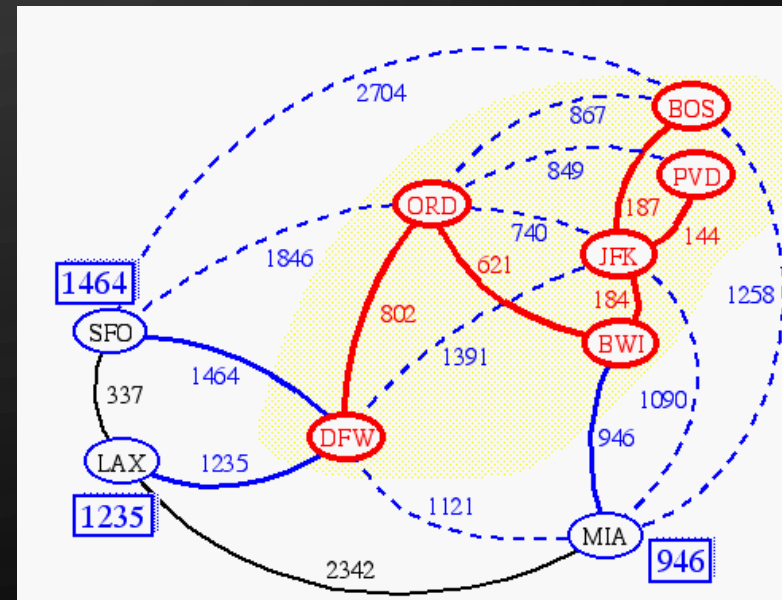


Replacing  $f$  with  $e$  yields another MST



# PRIM-JARNIK'S ALGORITHM

- We pick an arbitrary vertex  $s$  and we grow the MST as a cloud of vertices, starting from  $s$
- We store with each vertex  $v$  a label  $d(v)$  representing the smallest weight of an edge connecting  $v$  to a vertex in the cloud
- At each step:
  - We add to the cloud the vertex  $u$  outside the cloud with the smallest distance label
  - We update the labels of the vertices adjacent to  $u$



# PRIM-JARNIK'S ALGORITHM

- An adaptable priority queue stores the vertices outside the cloud
  - Key: distance,  $D[v]$
  - Element: vertex  $v$
  - $Q.replace(i, k)$  changes the key of an item
- We store three labels with each vertex  $v$ :
  - Distance  $D[v]$
  - Parent edge in MST  $P[v]$
  - Locator in priority queue

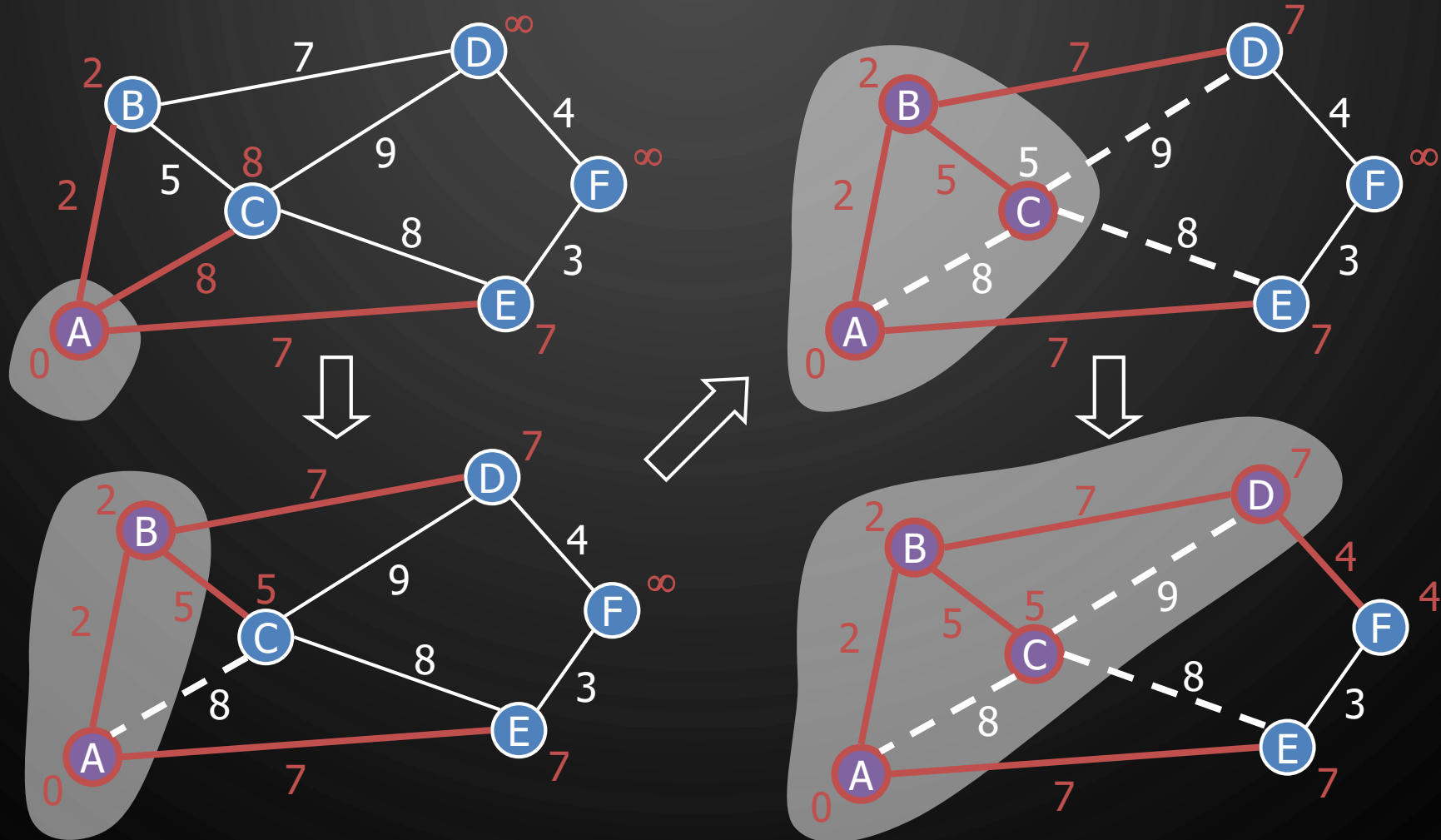
**Algorithm** PrimJarnikMST( $G$ )

**Input:** A weighted connected graph  $G$

**Output:** A minimum spanning tree  $T$  of  $G$

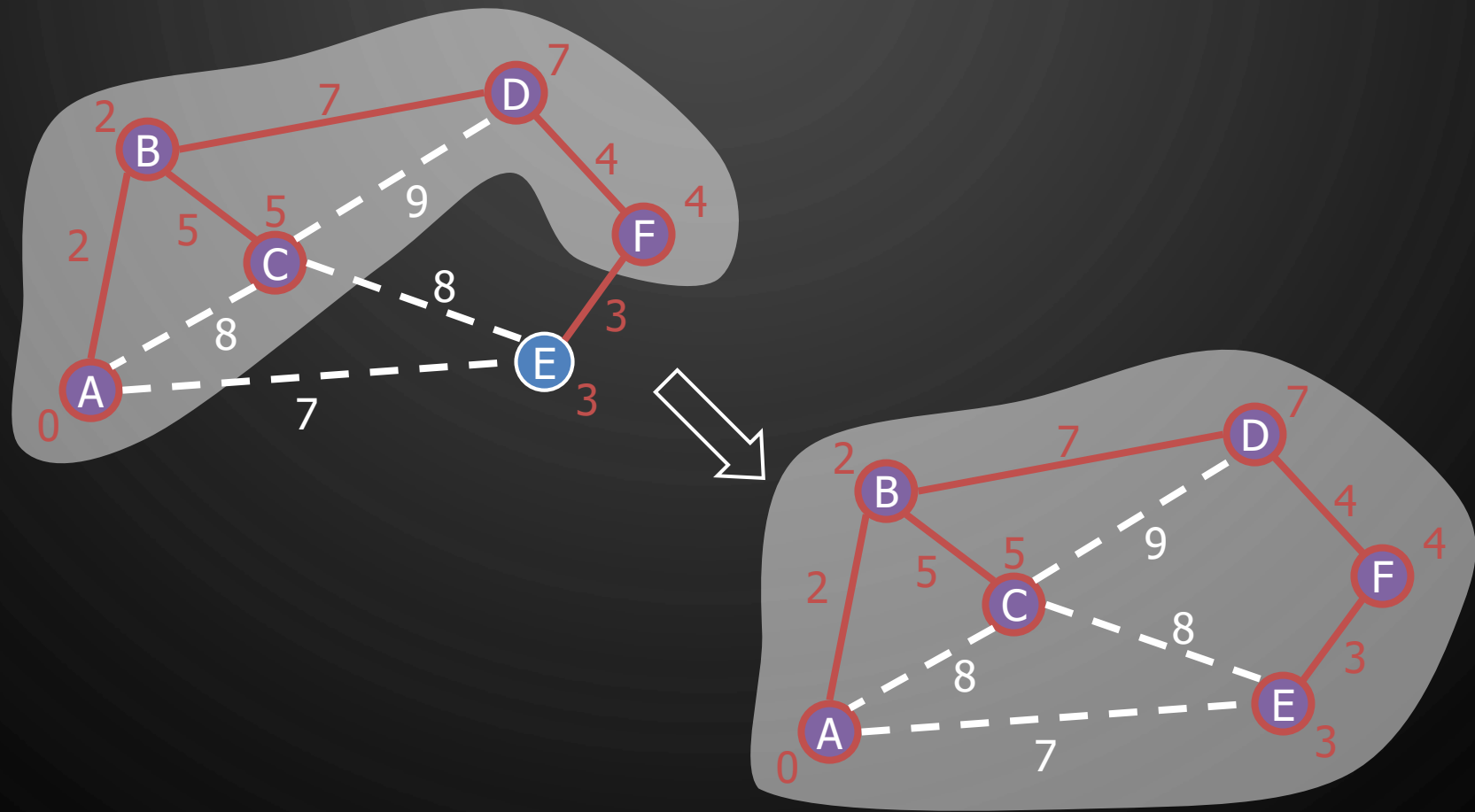
1. Pick any vertex  $s$  of  $G$
2.  $D[s] \leftarrow 0$ ;  $P[s] \leftarrow \emptyset$
3. **for each** vertex  $v \neq s$  **do**
4.      $D[v] \leftarrow \infty$ ;  $P[v] \leftarrow \emptyset$
5.  $T \leftarrow \emptyset$
6. Priority queue  $Q$  of vertices with  $D[v]$  as the key
7. **while**  $\neg Q.isEmpty()$  **do**
8.      $u \leftarrow Q.removeMin()$
9.     Add vertex  $u$  and edge  $P[u]$  to  $T$
10.    **for each**  $e \in u.outgoingEdges()$  **do**
11.      $v \leftarrow G.opposite(u, e)$
12.     **if**  $e.weight() < D[v]$  **then**
13.          $D[v] \leftarrow e.weight()$ ;  $P[v] \leftarrow e$
14.      $Q.replace(v, D[v])$
15. **return**  $T$

# EXAMPLE





# EXAMPLE

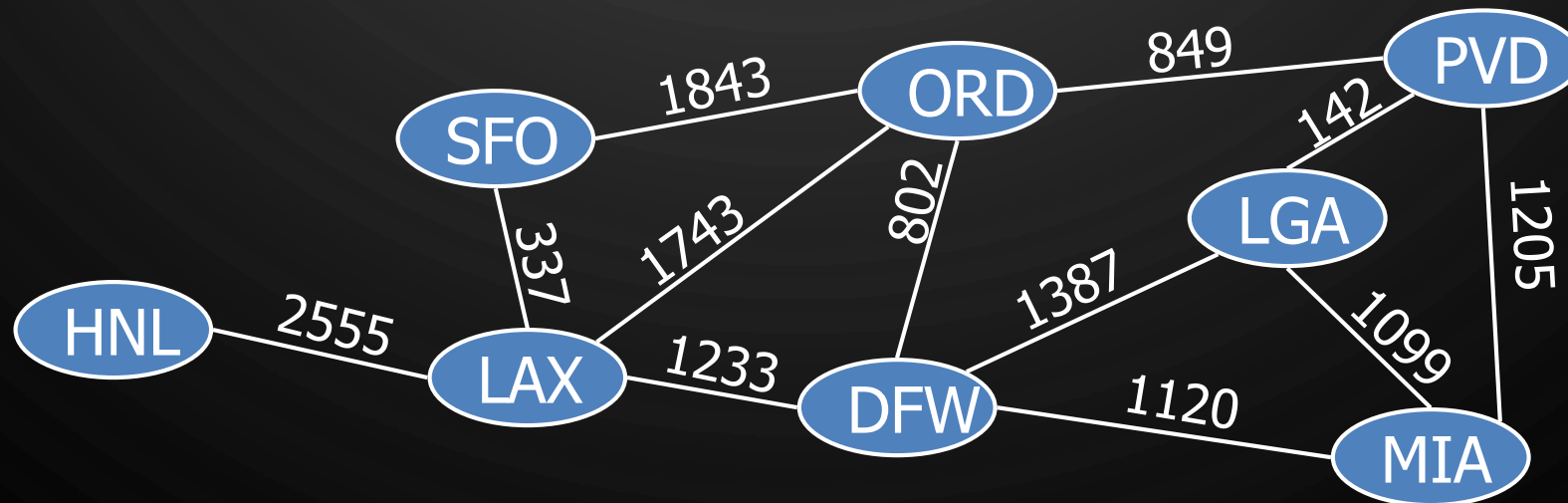


# EXERCISE

## PRIM'S MST ALGORITHM



- Show how Prim's MST algorithm works on the following graph, assuming you start with SFO
  - Show how the MST evolves in each iteration.



# ANALYSIS

- Graph operations
  - Method `incidentEdges` is called once for each vertex
- Label operations
  - We set/get the distance, parent and locator labels of vertex  $z$   $O(\deg(z))$  times
  - Setting/getting a label takes  $O(1)$  time
- Priority queue operations
  - Each vertex is inserted once into and removed once from the priority queue, where each insertion or removal takes  $O(\log n)$  time
  - The key of a vertex  $w$  in the priority queue is modified at most  $\deg(w)$  times, where each key change takes  $O(\log n)$  time
- Prim-Jarnik's algorithm runs in  $O((n + m) \log n)$  time provided the graph is represented by the adjacency list structure
  - Recall that  $\sum_v \deg(v) = 2m$
- If the graph is connected the running time is  $O(m \log n)$