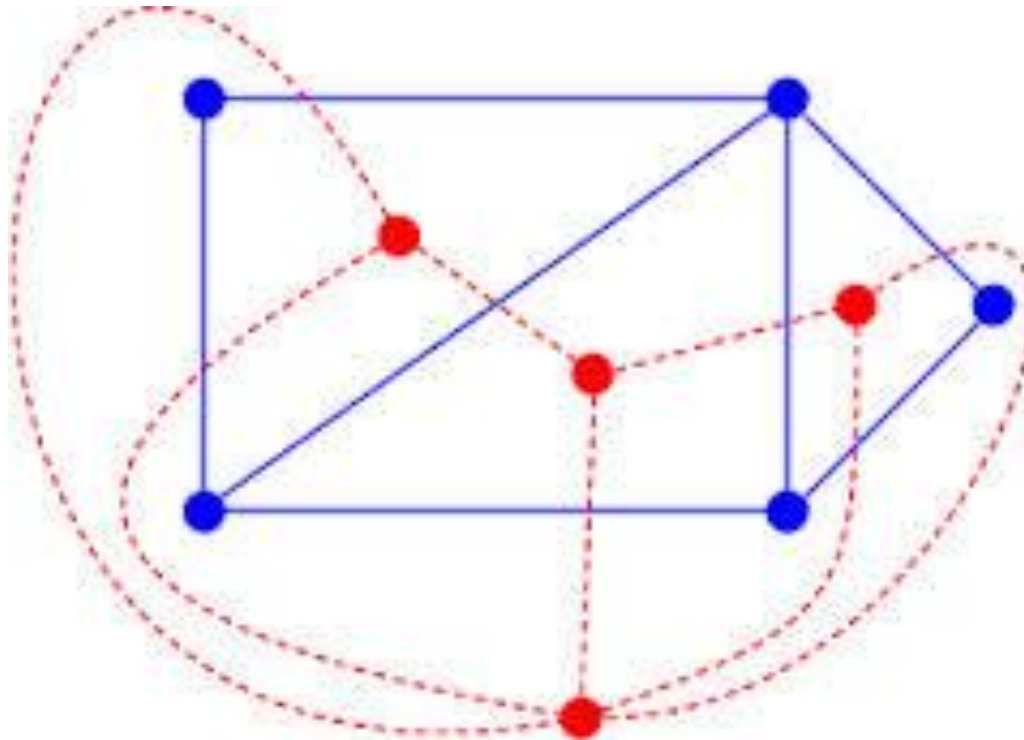


# Graphs

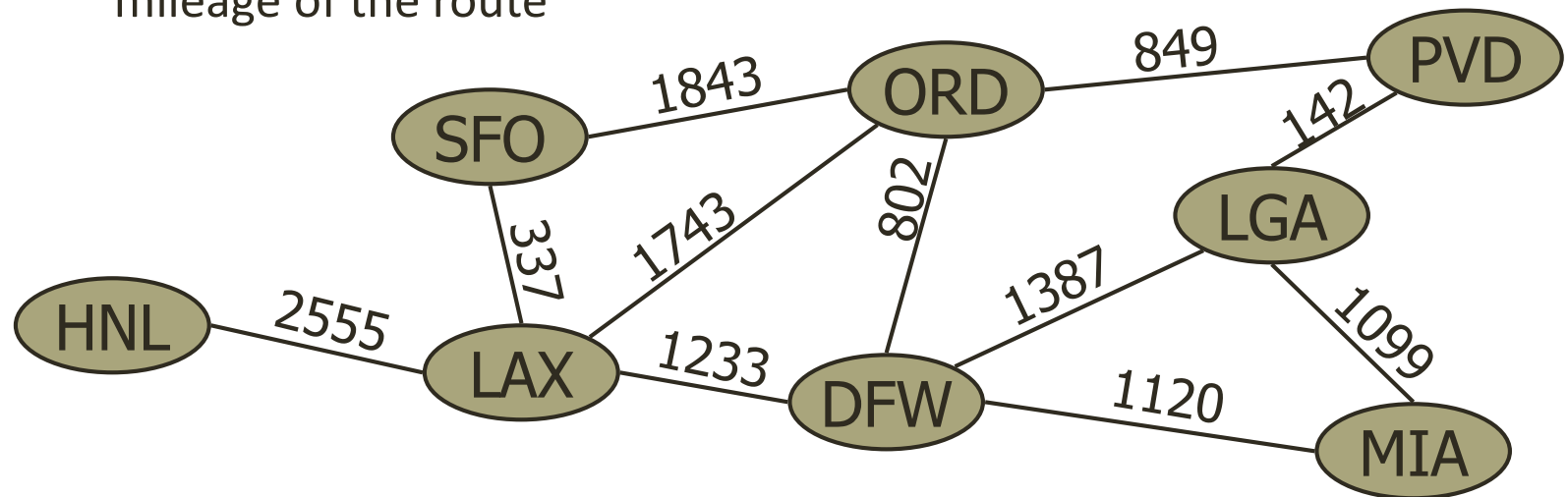


# Overview

- Graph terminology & representation data structures
- Traversals:
  - Depth First Search
  - Breadth First Search
- Topological Sorting of DAGs
- Transitive Closure: Floyd-Warshall
- Shortest Paths in Weighted graphs
  - Dijkstra, Bellman-Ford, DAGs
- Minimum Spanning Trees
  - Prim-Jarnik, Kruskal
- Union-Find Structures

# Graphs

- A graph is a pair  $(V, E)$ , where
  - $V$  is a set of nodes, called **vertices**
  - $E$  is a collection of pairs of vertices, called **edges**
  - Vertices and edges are positions and store elements
- **Example:**
  - A vertex represents an airport and stores the three-letter airport code
  - An edge represents a flight route between two airports and stores the mileage of the route



# Graph applications

graph	vertex	edge
communication	telephone, computer	fiber optic cable
circuit	gate, register, processor	wire
mechanical	joint	rod, beam, spring
financial	stock, currency	transactions
transportation	intersection	street
internet	class C network	connection
game	board position	legal move
social relationship	person	friendship
neural network	neuron	synapse
protein network	protein	protein-protein interaction
molecule	atom	bond

# Edge Types

- **Directed edge**

- ordered pair of vertices  $(u, v)$
- first vertex  $u$  is the origin
- second vertex  $v$  is the destination
- e.g., a flight



- **Undirected edge**

- unordered pair of vertices  $(u, v)$
- e.g., a flight route



- **Directed graph**

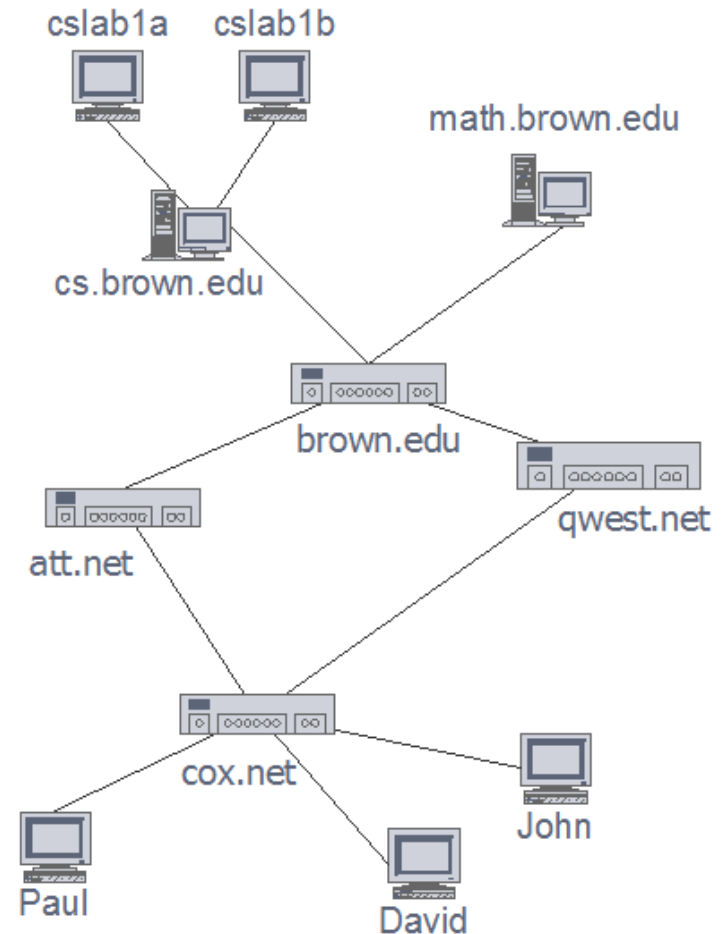
- all the edges are directed
- e.g., route network

- **Undirected graph**

- all the edges are undirected
- e.g., flight network

# Applications

- **Electronic circuits**
  - Printed circuit board
  - Integrated circuit
- **Transportation networks**
  - Highway network
  - Flight network
- **Computer networks**
  - Local area network
  - Internet
  - Web
- **Databases**
  - Entity-relationship diagram



# Terminology

- **End vertices (or endpoints) of an edge**

- U and V are the endpoints of a

- **Edges incident on a vertex**

- a, d, and b are incident on V

- **Adjacent vertices**

- U and V are adjacent

- **Degree of a vertex**

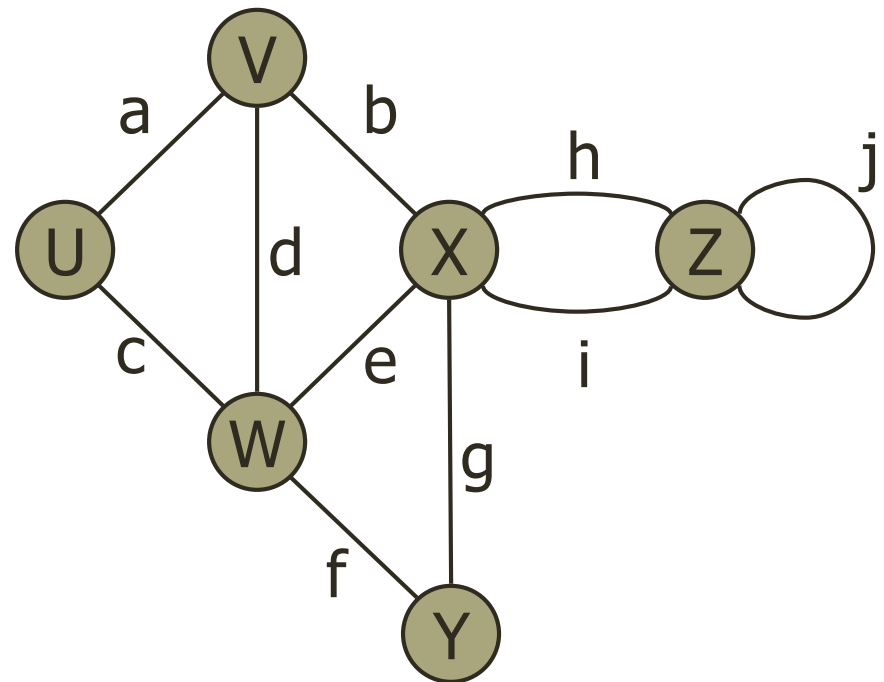
- X has degree 5

- **Parallel edges**

- h and i are parallel edges

- **Self-loop**

- j is a self-loop



# Terminology (cont.)

- **Path**

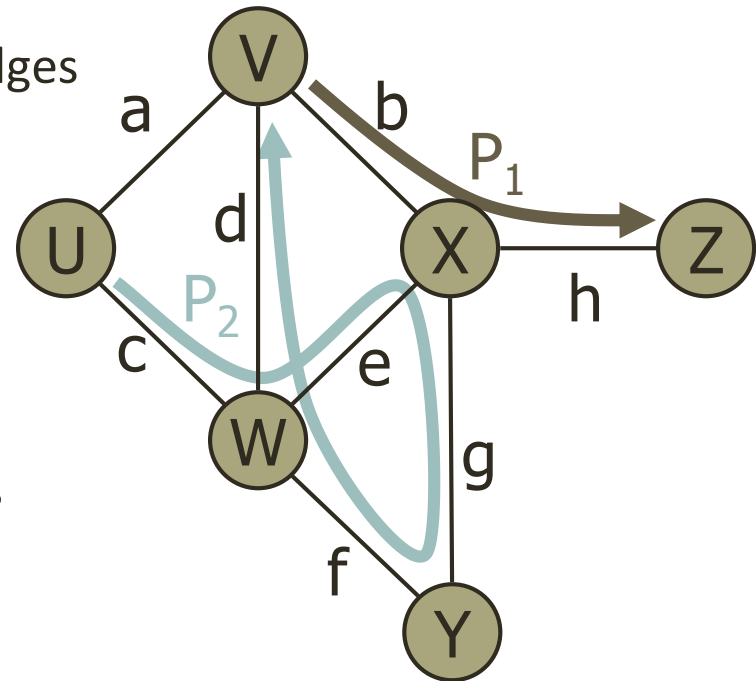
- sequence of alternating vertices and edges
- begins with a vertex
- ends with a vertex
- each edge is preceded and followed by its endpoints

- **Simple path**

- path such that all its vertices and edges are distinct

- **Examples:**

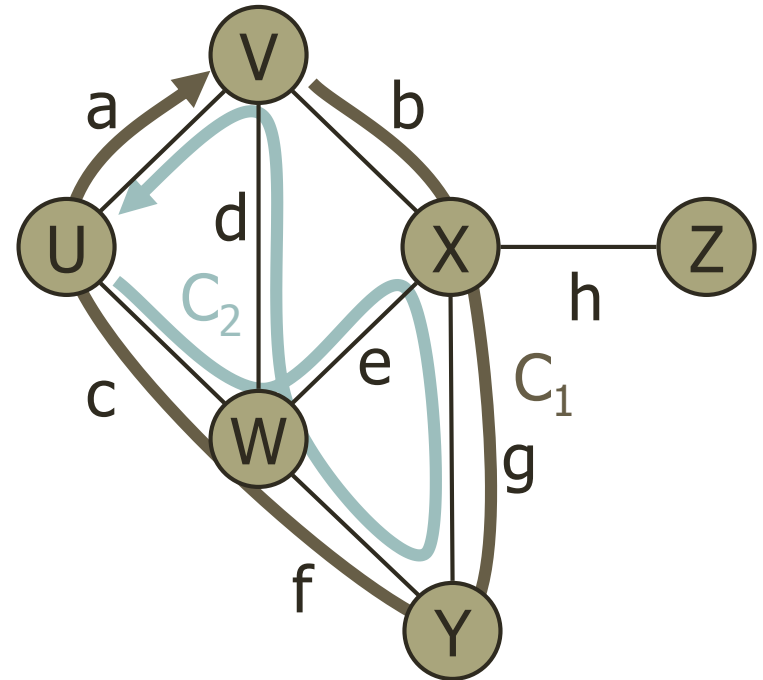
- $P_1 = (V, b, X, h, Z)$  is a simple path
- $P_2 = (U, c, W, e, X, g, Y, f, W, d, V)$  is a path that is not simple





# Terminology (cont.)

- **Cycle**
  - circular sequence of alternating vertices and edges
  - each edge is preceded and followed by its endpoints
- **Simple cycle**
  - cycle such that all its vertices and edges are distinct
- **Examples:**
  - $C_1 = (V, b, X, g, Y, f, W, c, U, a, \hookrightarrow)$  is a simple cycle
  - $C_2 = (U, c, W, e, X, g, Y, f, W, d, V, a, \hookrightarrow)$  is a cycle that is not simple



# Properties

## Property 1

$$\sum_v \deg(v) = 2m$$

**Proof:** each edge is counted twice

## Notation

$n$  number of vertices

$m$  number of edges

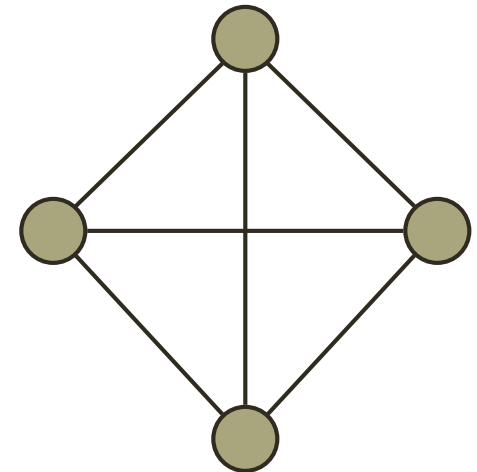
$\deg(v)$  degree of vertex  $v$

## Property 2

In an undirected graph with no self-loops and no multiple edges

$$m \leq n(n-1)/2$$

**Proof:** each vertex has degree at most  $(n-1)$



# Vertices and Edges

- A **graph** is a collection of **vertices** and **edges**.
- We model the abstraction as a combination of three data types: Vertex, Edge, and Graph.
- A **Vertex** is a lightweight object that stores an arbitrary element provided by the user (e.g., an airport code)
  - We assume it supports a method, `element()`, to retrieve the stored element.
- An **Edge** stores an associated object (e.g., a flight number, travel distance, cost), retrieved with the `element()` method.

# Graph ADT: part 1

- `numVertices()`: Returns the number of vertices of the graph.
- `vertices()`: Returns an iteration of all the vertices of the graph.
- `numEdges()`: Returns the number of edges of the graph.
- `edges()`: Returns an iteration of all the edges of the graph.
- `getEdge( $u, v$ )`: Returns the edge from vertex  $u$  to vertex  $v$ , if one exists; otherwise return null. For an undirected graph, there is no difference between `getEdge( $u, v$ )` and `getEdge( $v, u$ )`.
- `endVertices( $e$ )`: Returns an array containing the two endpoint vertices of edge  $e$ . If the graph is directed, the first vertex is the origin and the second is the destination.
- `opposite( $v, e$ )`: For edge  $e$  incident to vertex  $v$ , returns the other vertex of the edge; an error occurs if  $e$  is not incident to  $v$ .

# Graph ADT: part 2

`outDegree( $v$ )`: Returns the number of outgoing edges from vertex  $v$ .

`inDegree( $v$ )`: Returns the number of incoming edges to vertex  $v$ . For an undirected graph, this returns the same value as does `outDegree( $v$ )`.

`outgoingEdges( $v$ )`: Returns an iteration of all outgoing edges from vertex  $v$ .

`incomingEdges( $v$ )`: Returns an iteration of all incoming edges to vertex  $v$ . For an undirected graph, this returns the same collection as does `outgoingEdges( $v$ )`.

`insertVertex( $x$ )`: Creates and returns a new Vertex storing element  $x$ .

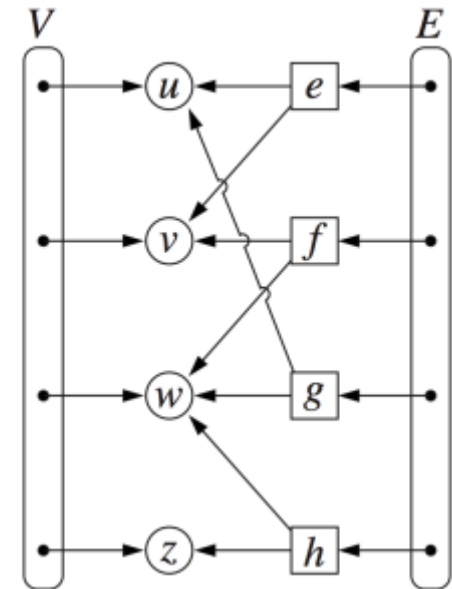
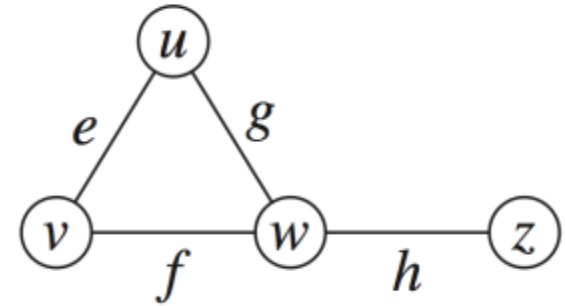
`insertEdge( $u$ ,  $v$ ,  $x$ )`: Creates and returns a new Edge from vertex  $u$  to vertex  $v$ , storing element  $x$ ; an error occurs if there already exists an edge from  $u$  to  $v$ .

`removeVertex( $v$ )`: Removes vertex  $v$  and all its incident edges from the graph.

`removeEdge( $e$ )`: Removes edge  $e$  from the graph.

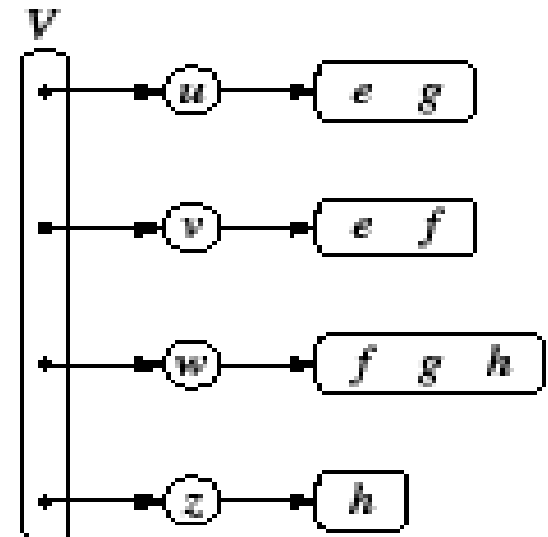
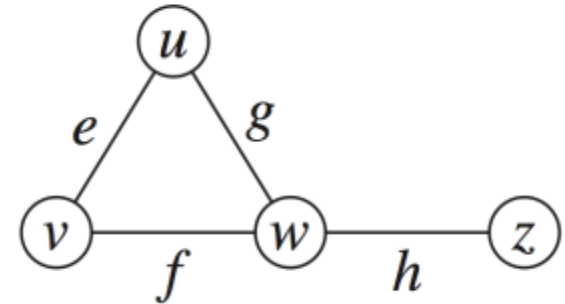
# Edge List Structure

- **Vertex object**
  - element
  - reference to position in vertex sequence
- **Edge object**
  - element
  - origin vertex object
  - destination vertex object
  - reference to position in edge sequence
- **Vertex sequence**
  - sequence of vertex objects
- **Edge sequence**
  - sequence of edge objects



# Adjacency List Structure

- **Incidence sequence for each vertex**
  - sequence of references to edge objects of incident edges
- **Augmented edge objects**
  - references to associated positions in incidence sequences of end vertices

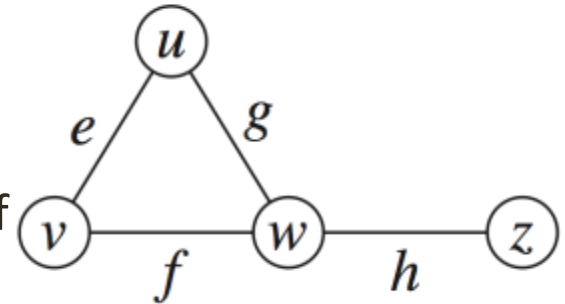




# Adjacency Map Structure

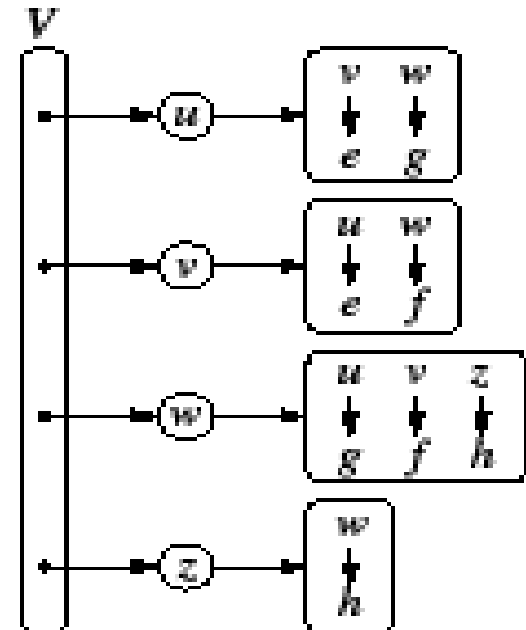
- Incidence sequence for each vertex**

- sequence of references to adjacent vertices, each mapped to edge object of the incident edge



- Augmented edge objects**

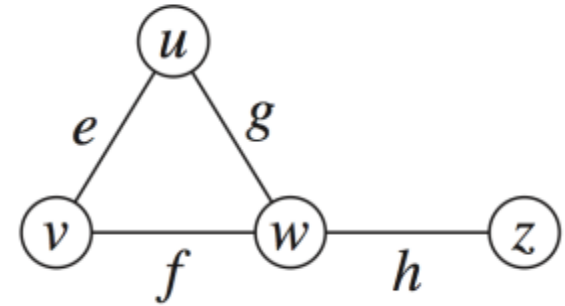
- references to associated positions in incidence sequences of end vertices





# Adjacency Matrix Structure

- **Edge list structure**
- **Augmented vertex objects**
  - Integer key (index) associated with vertex
- **2D-array adjacency array**
  - Reference to edge object for adjacent vertices
  - Null for non-adjacent vertices



- The “old fashioned” version just has 0 for no edge and 1 for edge

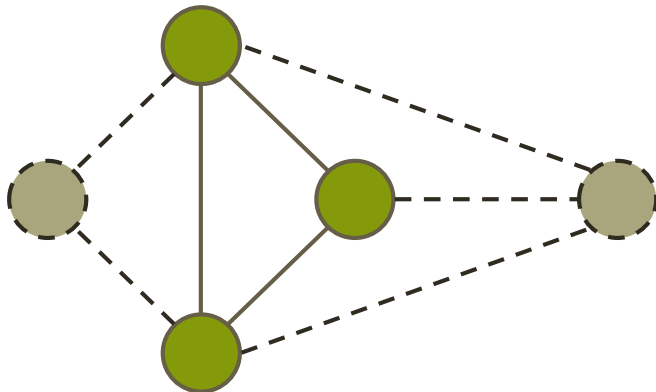
		0	1	2	3
$u \longrightarrow$	0		$e$	$g$	
$v \longrightarrow$	1	$e$		$f$	
$w \longrightarrow$	2	$g$	$f$		$h$
$z \longrightarrow$	3			$h$	

# Performance

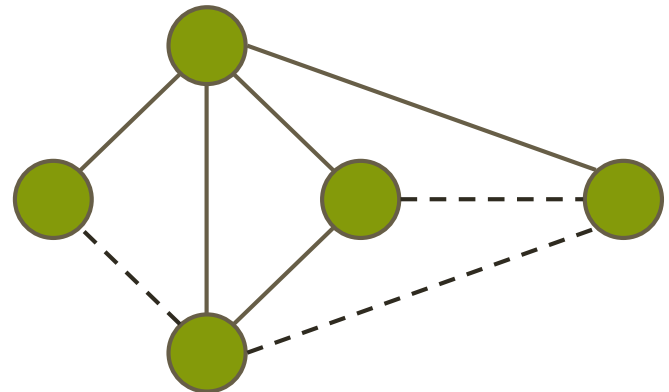
<ul style="list-style-type: none"> <li>▪ <math>n</math> vertices, <math>m</math> edges</li> <li>▪ no parallel edges</li> <li>▪ no self-loops</li> </ul>	Edge List	Adjacency List	Adjacency Matrix
Space	$n + m$	$n + m$	$n^2$
incidentEdges( $v$ )	$m$	deg( $v$ )	$n$
areAdjacent ( $v, w$ )	$m$	min(deg( $v$ ), deg( $w$ ))	1
insertVertex( $o$ )	1	1	$n^2$
insertEdge( $v, w, o$ )	1	1	1
removeVertex( $v$ )	$m$	deg( $v$ )	$n^2$
removeEdge( $e$ )	1	max(deg( $v$ ), deg( $w$ ))	1

# Subgraphs

- A subgraph  $S$  of a graph  $G$  is a graph such that
  - The vertices of  $S$  are a subset of the vertices of  $G$
  - The edges of  $S$  are a subset of the edges of  $G$
- A **spanning** subgraph of  $G$  is a subgraph that contains all the vertices of  $G$



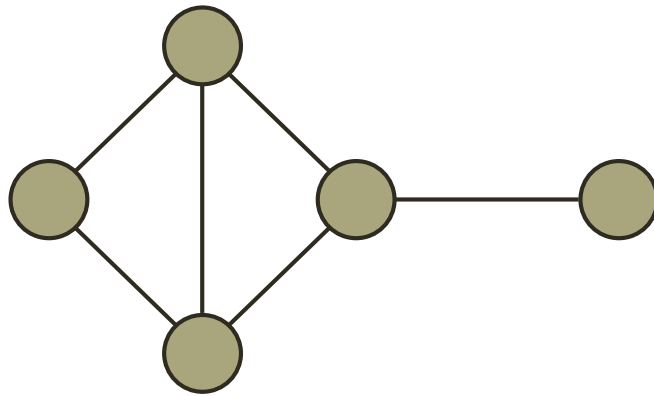
Subgraph



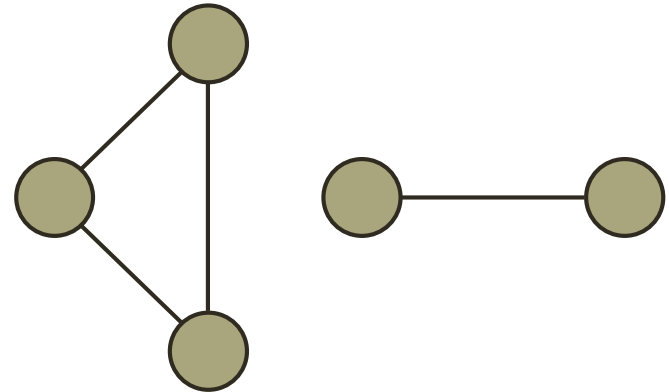
Spanning subgraph

# Connectivity

- A graph is **connected** if there is a path between every pair of vertices
- A **connected component** of a graph  $G$  is a maximal connected subgraph of  $G$



Connected graph



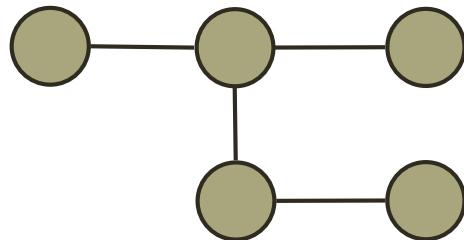
Non connected graph with two connected components

# Trees and Forests

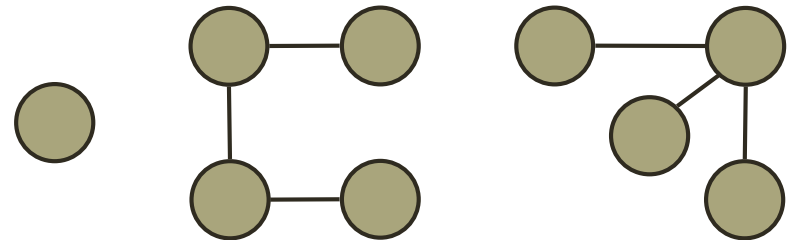
- A **(free) tree** is an undirected graph  $T$  such that
  - $T$  is **connected**
  - $T$  has **no cycles**

This definition of tree is different from the one of a **rooted tree**

- A **forest** is an undirected graph without cycles
- The **connected components** of a forest are trees



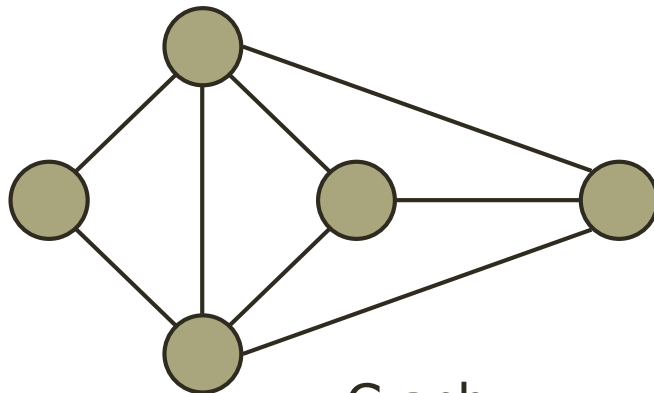
Tree



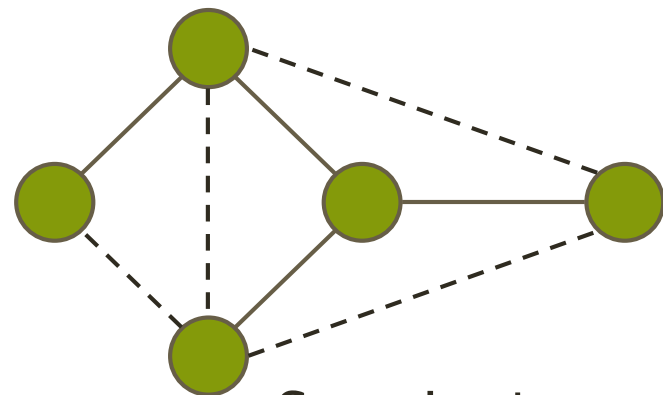
Forest

# Spanning Trees and Forests

- A **spanning tree** of a connected graph is a spanning **subgraph** that is a **tree**
- A spanning tree is not unique unless the graph is a tree
- Spanning trees have applications to the design of communication networks
- A spanning forest of a graph is a spanning subgraph that is a forest



Graph



Spanning tree

- 1. Depth First Search (DFS)**
- 2. Breadth First Search (BFS)**

# Depth-First Search



# Depth-First Search

- DFS is a general graph traversal technique
- DFS(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 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

Mark vertex  $u$  as visited.

**for** each of  $u$ 's outgoing edges,  $e = (u, v)$  **do**

**if** vertex  $v$  has not been visited **then**

        Record edge  $e$  as the discovery edge for vertex  $v$ .

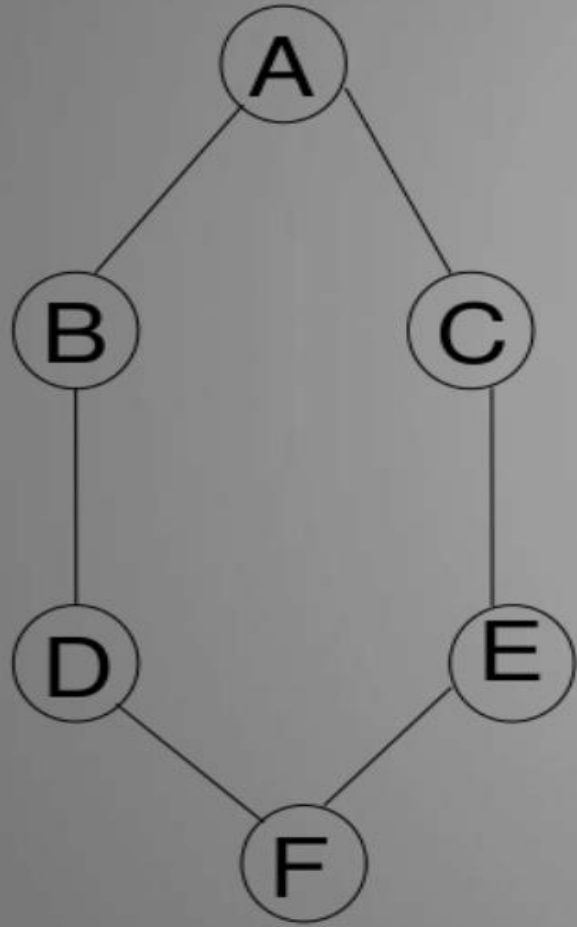
        Recursively call DFS( $G, v$ ).

- DFS stands for Depth First Search.
- DFS is an algorithm for traversing or searching a tree or graph data structures.
- It uses a stack data structure for implementation.
- In DFS one starts at the root and explores as far as possible along each branch before backtracking.

# Algorithm for DFS:

- [1]-- Initialize all nodes with status=1.(ready state)
- [2]– Put starting node in the stack and change status to status=2(waiting state).
- [3]– Loop:-
  - Repeat step- 4 and step- 5 until stack Get empty.
- [4]– Remove top node N from stack process them and change the status of N processed state (status=3).
- [5]– Add all the neighbours of N to the top of stack and change status to waiting status-2.

# How DFS works:



output



A



A  
B



A  
B  
D



A  
B  
D  
F

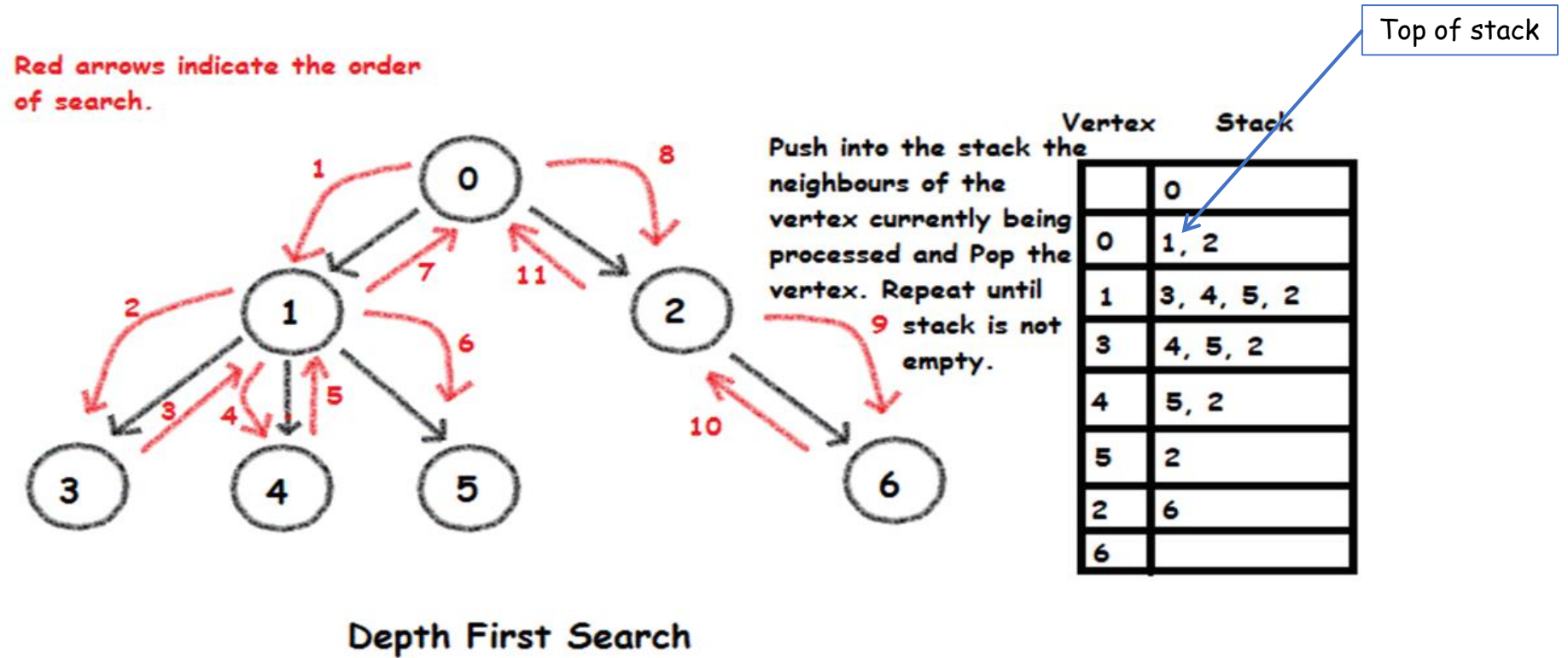


A  
B  
D  
F  
E



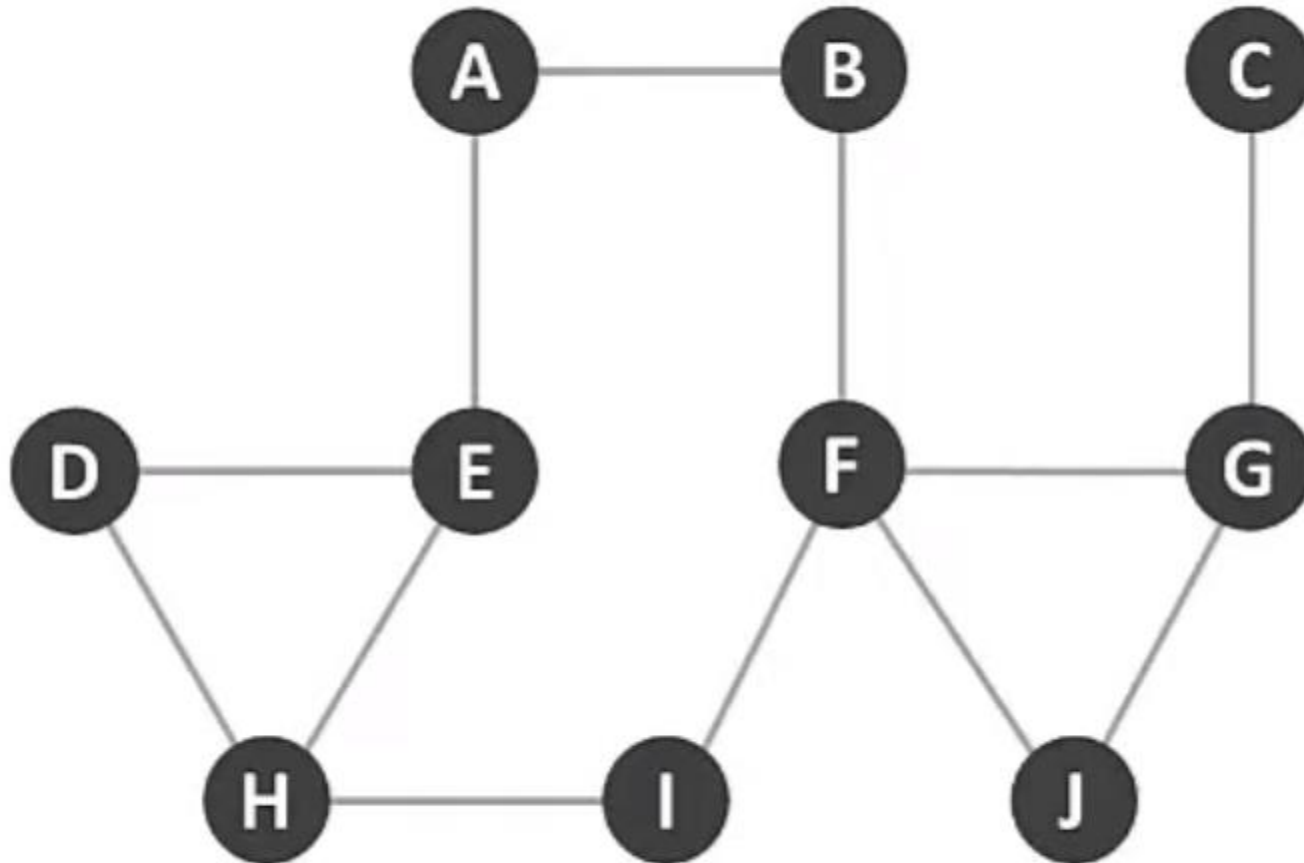
A  
B  
D  
F  
E  
C

# DFS: Example



# Exercise

Apply DFS algorithm for this graph:  
starting from Vertex A and going alphabetical order



# Breadth-First Search

- BFS is a general graph traversal technique
- A BFS traversal of a graph  $G$ 
  - Visits all vertices and edges of  $G$
  - Determines whether  $G$  is connected
  - Computes connected components of  $G$
  - Computes a spanning forest of  $G$
- BFS takes  $O(n + m)$  time
- BFS can be extended to solve other graph problems, e.g.,
  - Find a path with minimum number of edges between two given vertices
  - Minimum number of hops in a computer network (routing)
  - Find a simple cycle, if there is one



- BFS stands for Breadth First Search.
- BFS is an algorithm for traversing or searching a tree or graph data structures.
- It uses a queue data structure for implementation.
- In BFS traversal we visit all the nodes level by level and the traversal completed when all the nodes are visited.

# Algorithm for BFS:

**Step 1:** Initialize all nodes with status=1.(ready state)

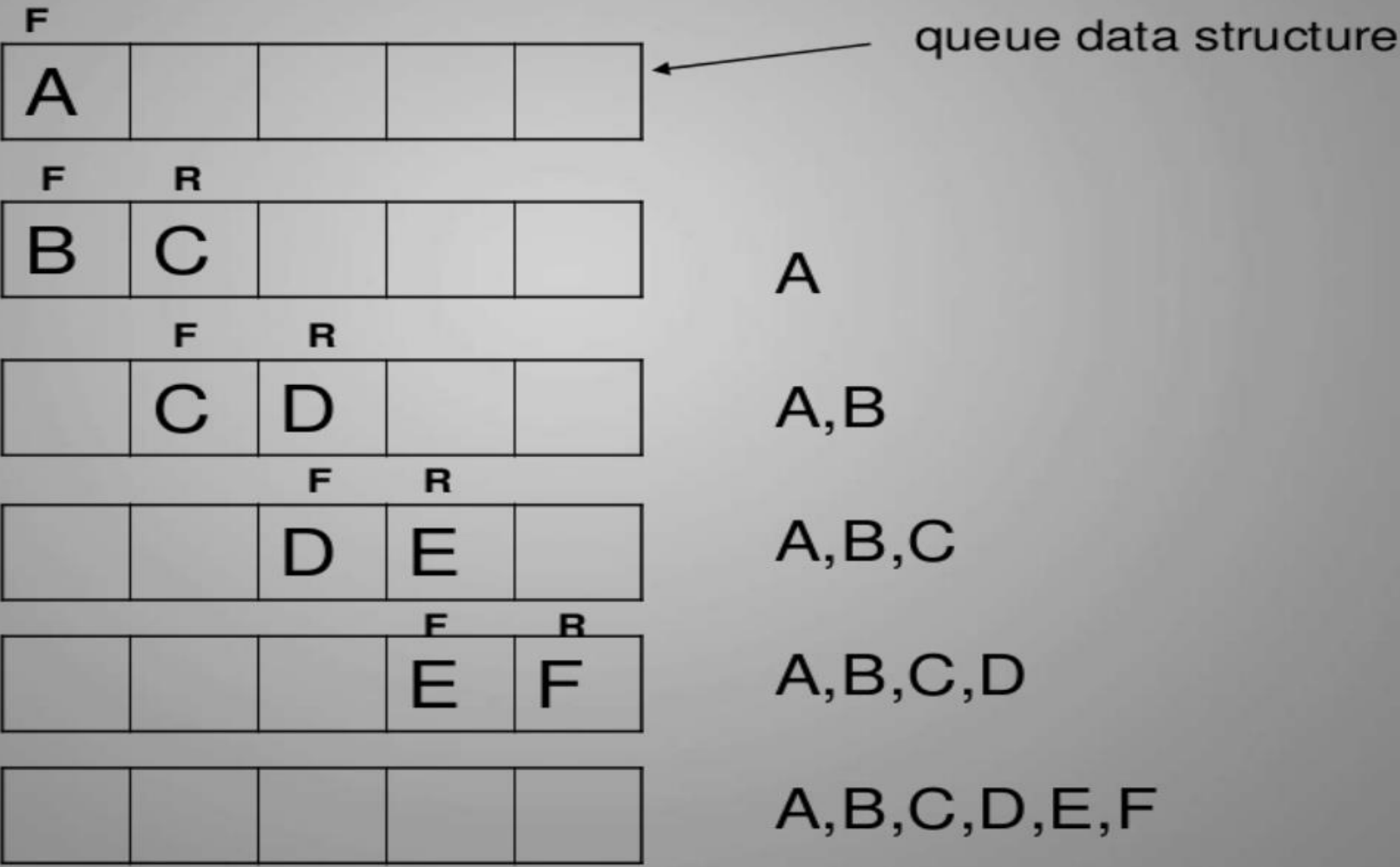
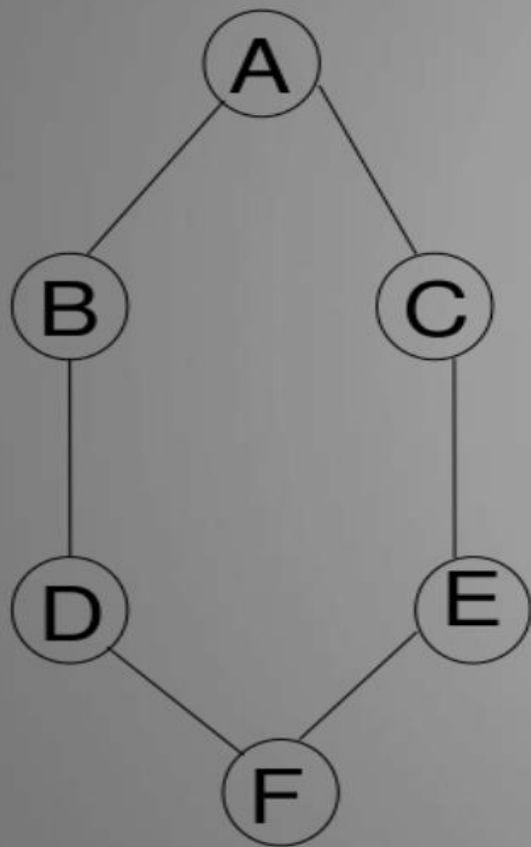
**Step 2:** Put starting node in a queue and change status to status=2.(waiting state)

**Step 3:** loop:  
repeat step 4 and step 5 until queue gets empty.

**Step 4:** Remove front node N from queue, process them and change the status of N to status=3.(processed state)

**Step 5:** Add all the neighbours of N to the rear of queue and change status to status=2.(waiting status)

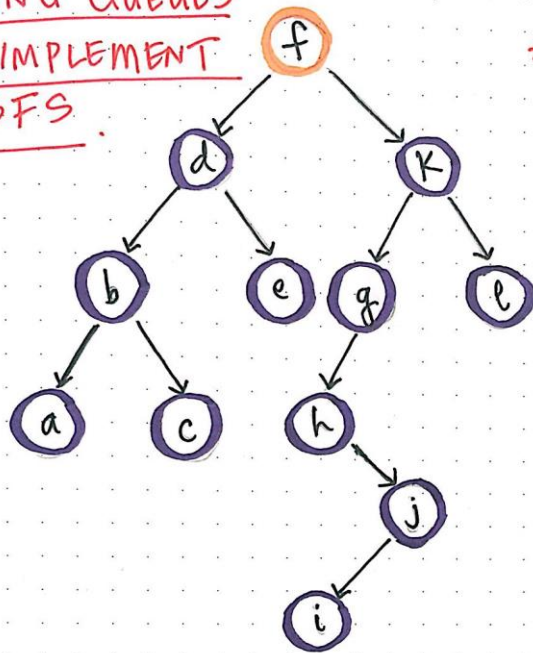
# How BFS works:



Queue gets empty and the algorithm ends

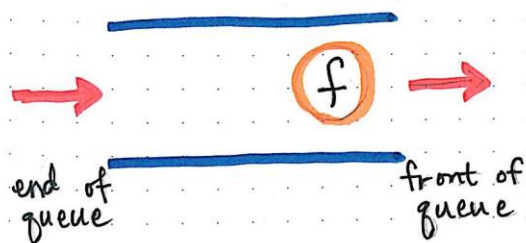
# BFS: Example

## USING QUEUES TO IMPLEMENT BFS.



\* **Discovered node:**  
a node we add to our queue, whose location we know, but we haven't actually visited yet.

→ To start, the root node will be the only discovered node.

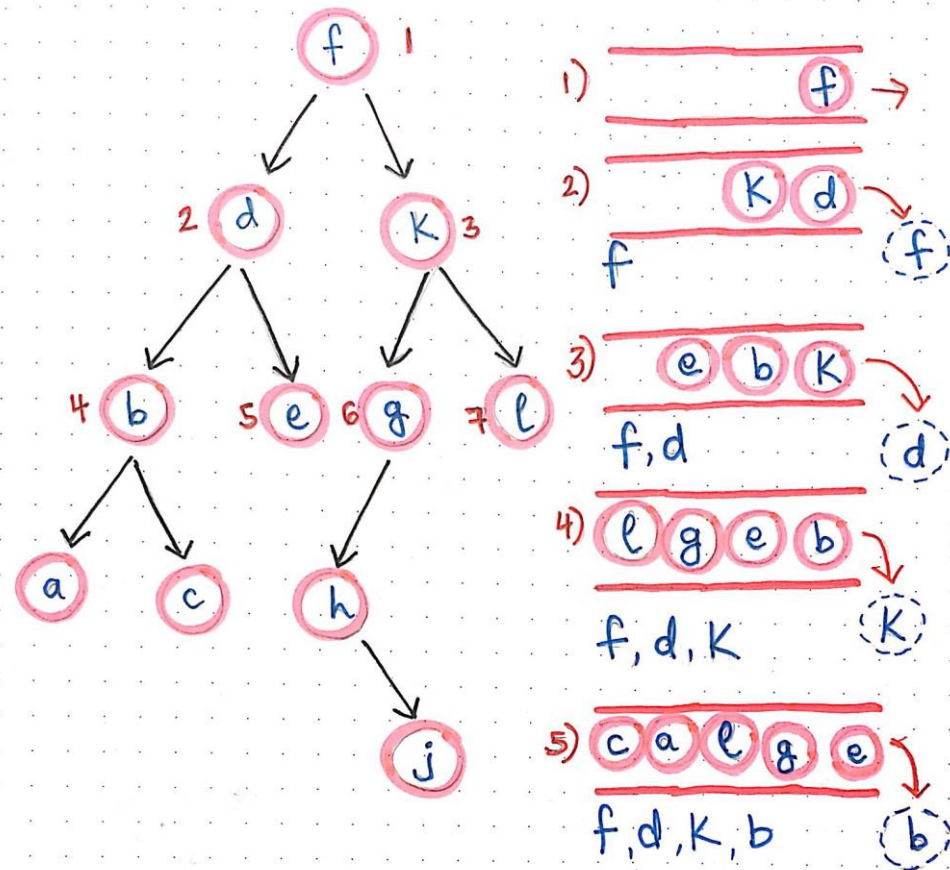


Steps to BFS:

- 1) Visit node f (print its value).
- 2) Enqueue left child.
- 3) Enqueue right child.

\* Once we have at least one node enqueued (and our queue isn't empty) we can start visiting the nodes and enqueueing their children.

} Continue until our queue is totally empty!

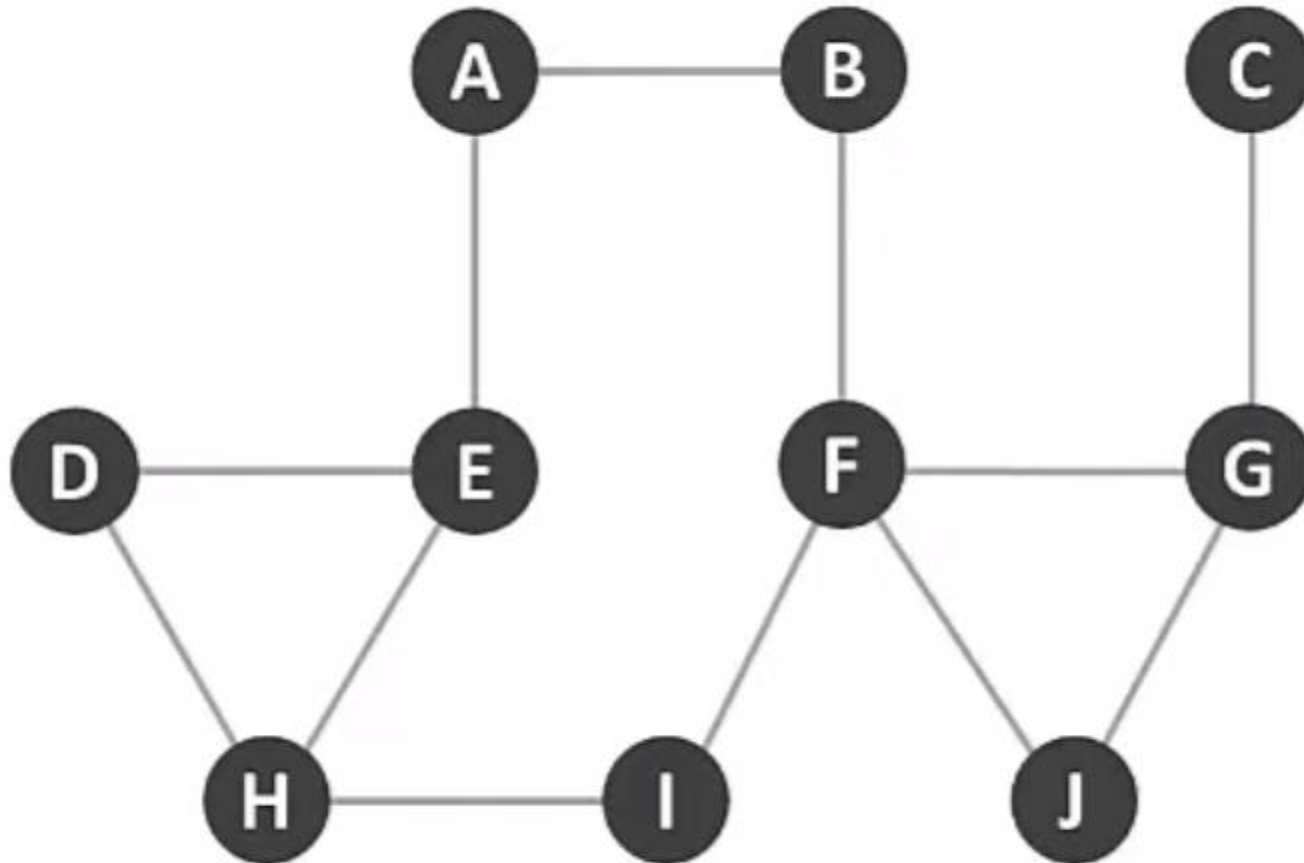


But what about space-time complexity?

- Visiting a node (reading its data and enqueueing its children) takes constant time. Since we are only visiting each node once, the time it will take us to use a BFS is  $O(n)$ , where  $n$  is the number of nodes.
- The space complexity depends on the size of the queue at its worst, which could be up to  $O(n)$  also.

# Exercise

Apply BFS algorithm for this graph



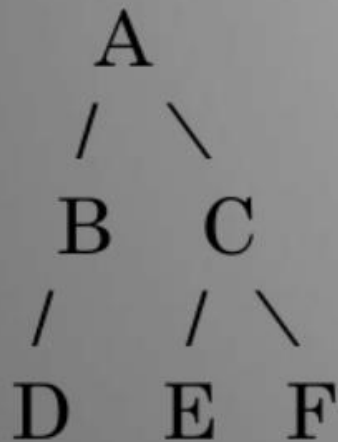


# DFS vs. BFS

- DFS stands for Depth First Search.
- DFS can be done with the help of STACK i.e., LIFO.
- In DFS has higher time and space complexity, because at a time it needs to back tracing in graph for traversal.
- BFS stands for Breadth First Search.
- BFS can be done with the help of QUEUE i.e., FIFO.
- In BFS the space & time complexity is lesser as there is no need to do back tracing

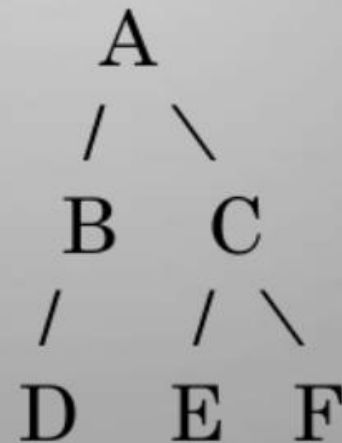
# DFS vs. BFS

- DFS is more faster then BFS.
- DFS requires less memory compare to BFS.
- DFS is not so useful in finding shortest path.
- Example :



Ans : A,B,D,C,E,F

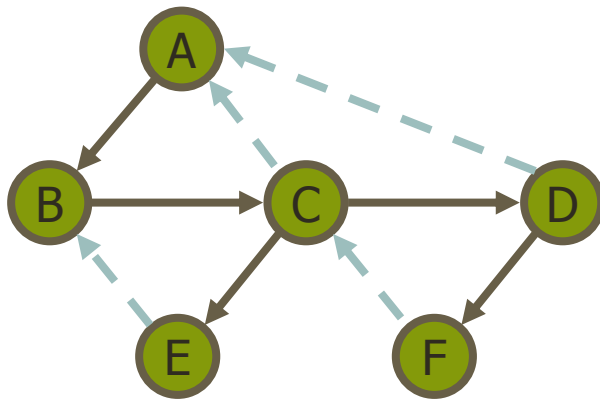
- BFS is slower than DFS.
- BFS requires more memory compare to DFS.
- BFS is useful in finding shortest path.
- Example :



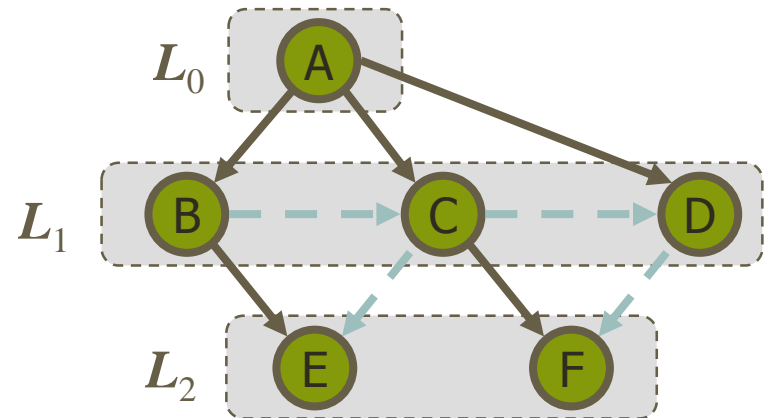
Ans : A,B,C,D,E,F

# DFS vs. BFS

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



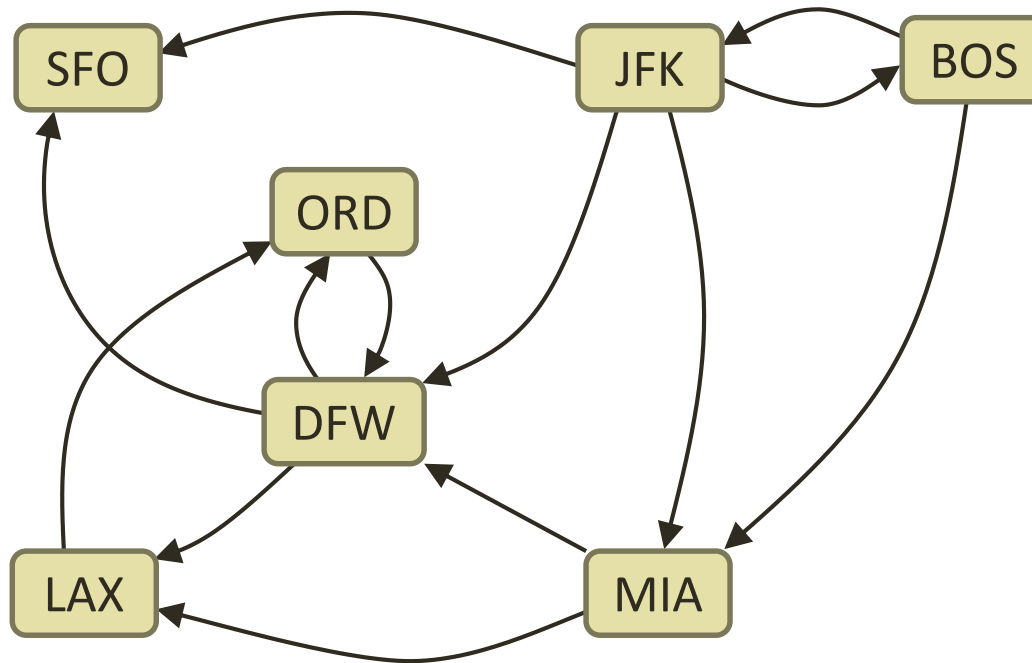
DFS



BFS

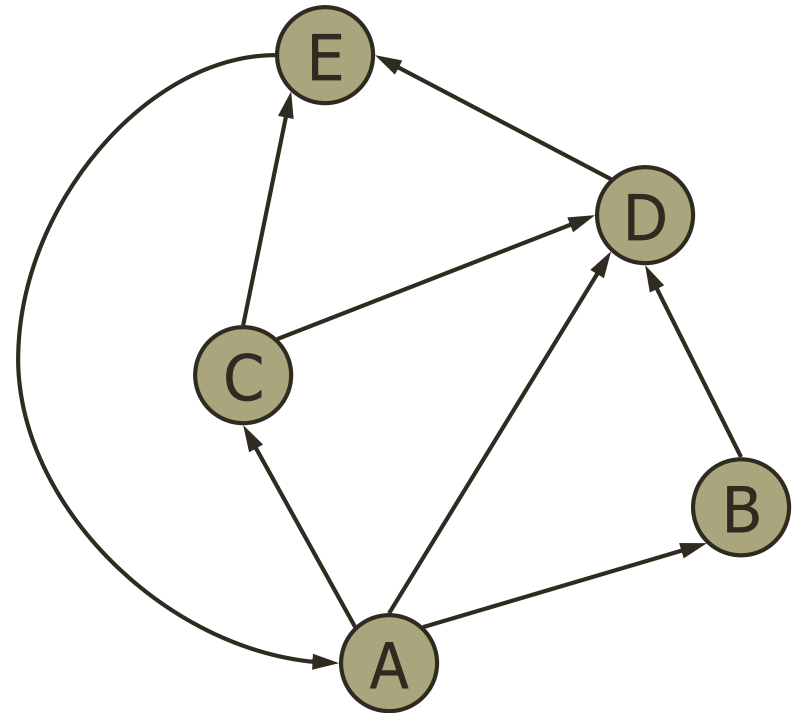


# Directed Graphs

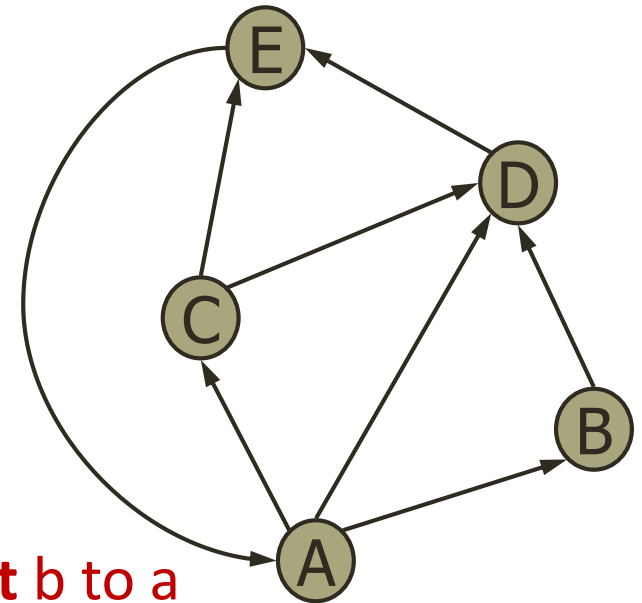


# Digraphs

- A **digraph** is a graph whose edges are all directed
  - Short for “directed graph”
- **Applications**
  - one-way streets
  - flights
  - task scheduling



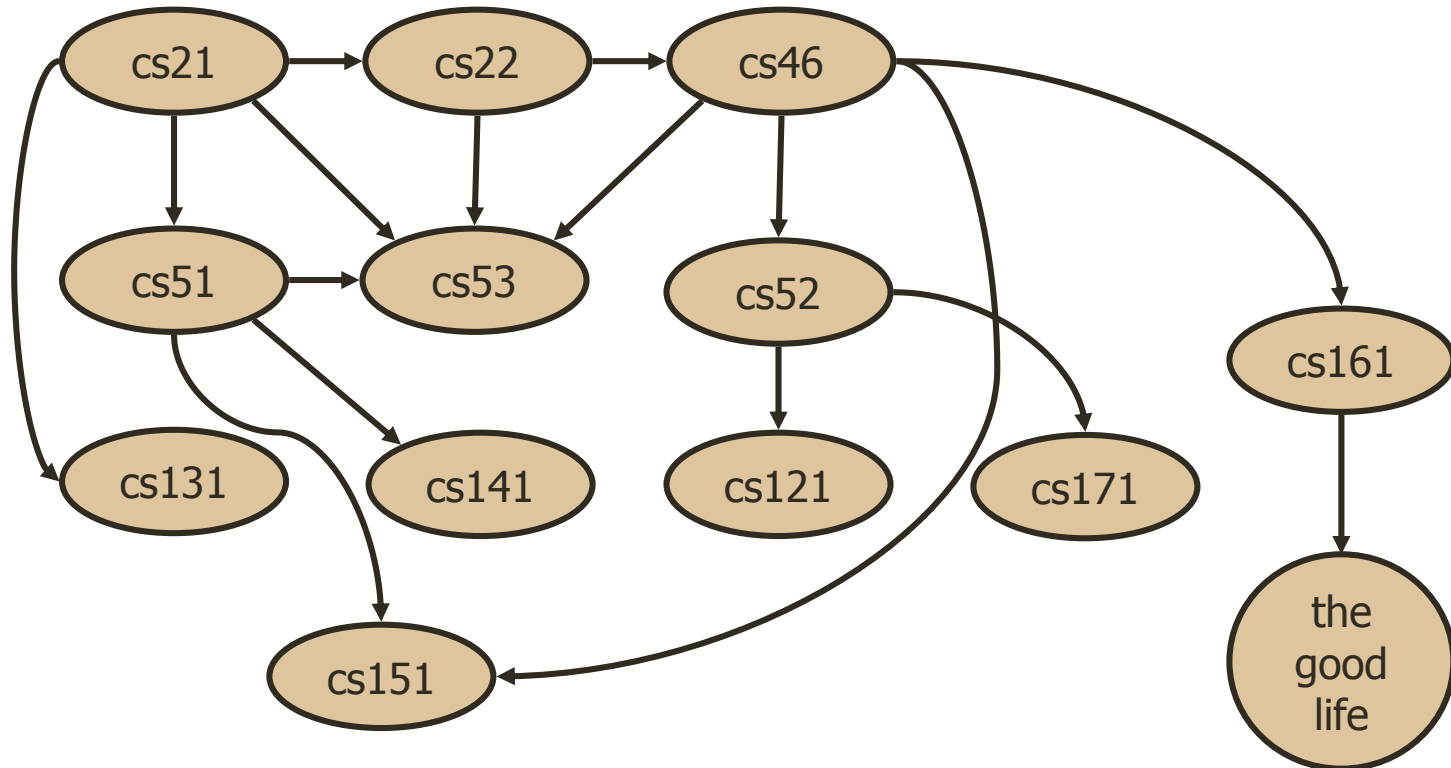
# Digraph Properties



- A graph  $G=(V,E)$  such that
  - Each edge goes in one direction:
  - Edge  $(a,b)$  goes from  $a$  to  $b$ , but **not**  $b$  to  $a$
- If  $G$  is simple,  $m \leq n \cdot (n - 1)$
- If we keep in-edges and out-edges in **separate** adjacency lists, we can perform listing of incoming edges and outgoing edges in time proportional to their size

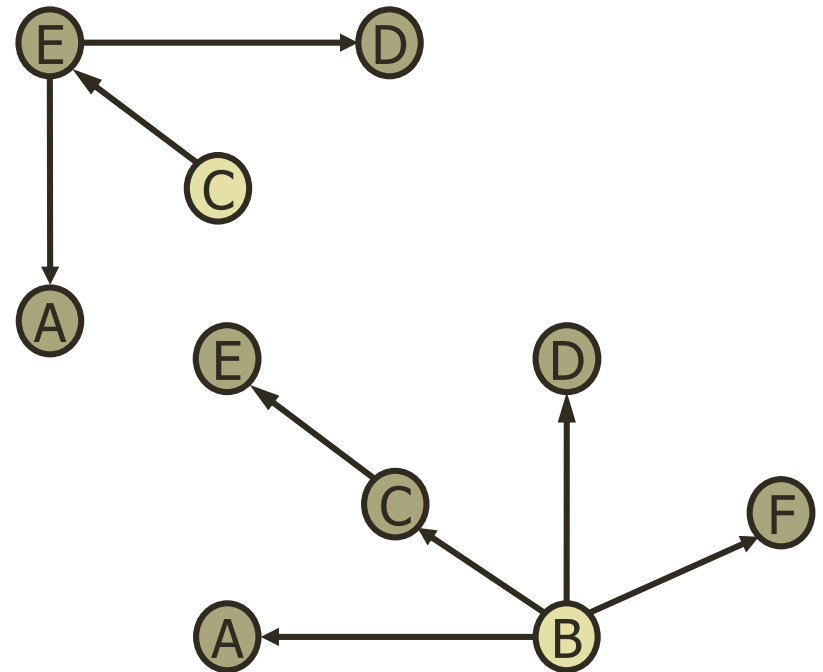
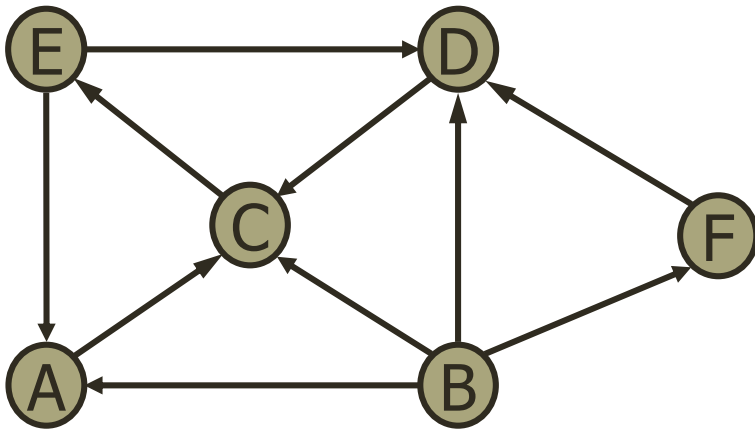
# Digraph Application

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



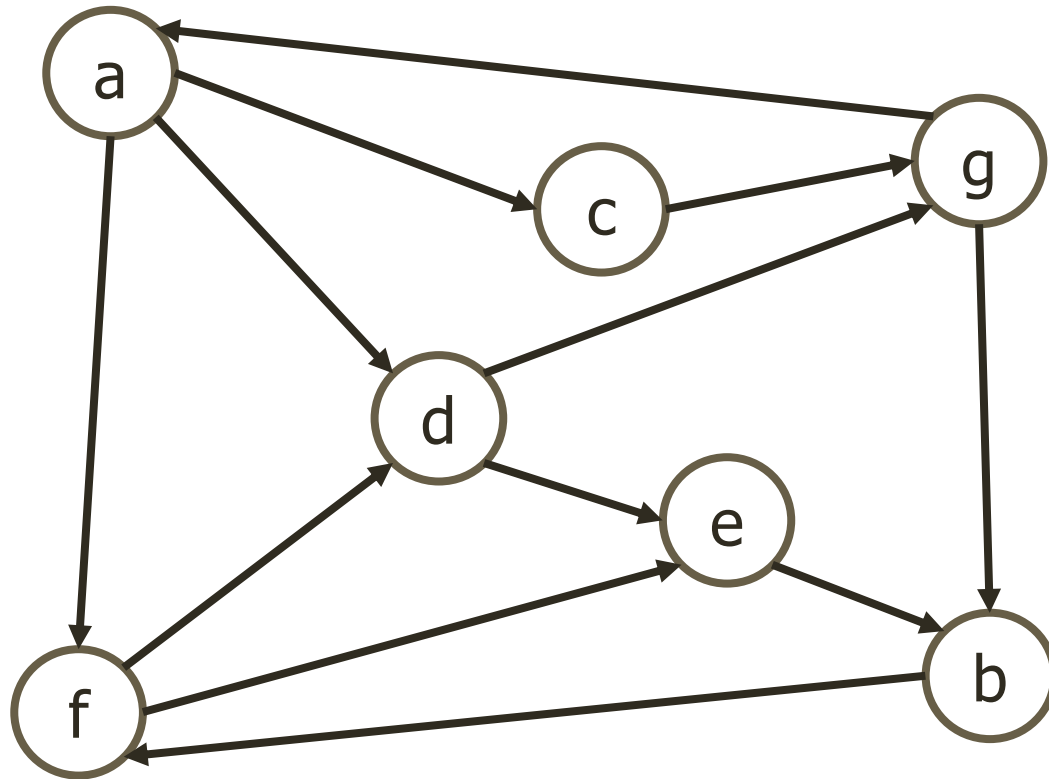
# Reachability

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

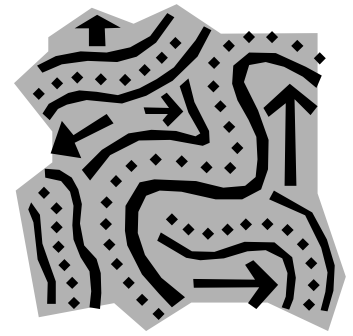


# Strong Connectivity

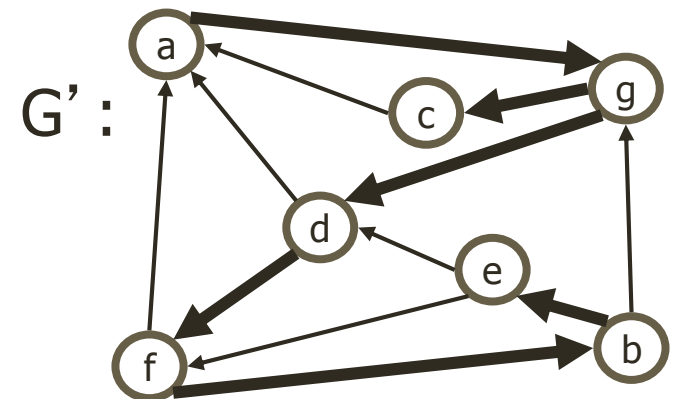
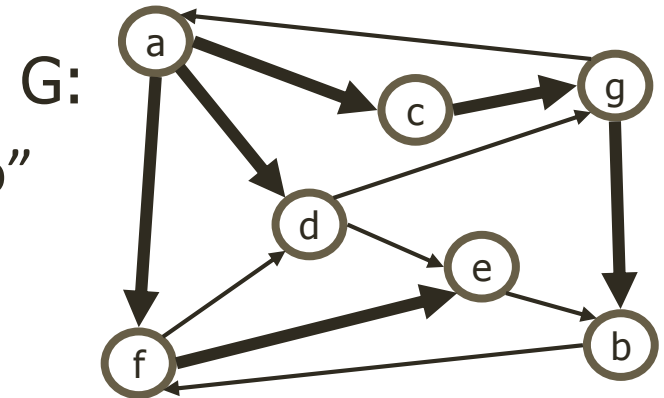
**Each** vertex can reach **all** other vertices



# Strong Connectivity Algorithm



- For each vertex  $v$  in  $G$  do:
  - Perform a DFS from  $v$  in  $G$ 
    - If there's a  $w$  not visited, return “no”
  - Let  $G'$  be  $G$  with edges reversed
  - Perform a DFS from  $v$  in  $G'$ 
    - If there's a  $w$  not visited, return “no”
- Else, return “yes”
- Running time:  $O(n(n+m))$



# **3. Topological Sorting of DAGs**