



# Graphs

Data Structures and  
Algorithms



# Warmup

Discuss with your neighbors:

Come up with as many kinds of relational data as you can (data that can be represented with a graph).

Scientific paper  
 $V$ : paper  
 $e$ : refs.

maps:  
 $V$ : dest.  
 $e$ : roads

family relationships  
 $V$ : people  
 $e$ : is parent of

food web:  
 $V$ : animals  
 $e$ :  $(u,v)$  mean  $u$  eats  $v$

# Announcements

Project 2 Due on Friday

Next individual HW assignment assigned tonight, due next Wednesday

- Two problems – merge sort and graph practice

Sign up for Project 3 partners by Thursday night!

I am gone tonight through Tuesday. Robbie will be lecturing again. No instructor office hours tomorrow or Tuesday next week.

# Graph: Formal Definition

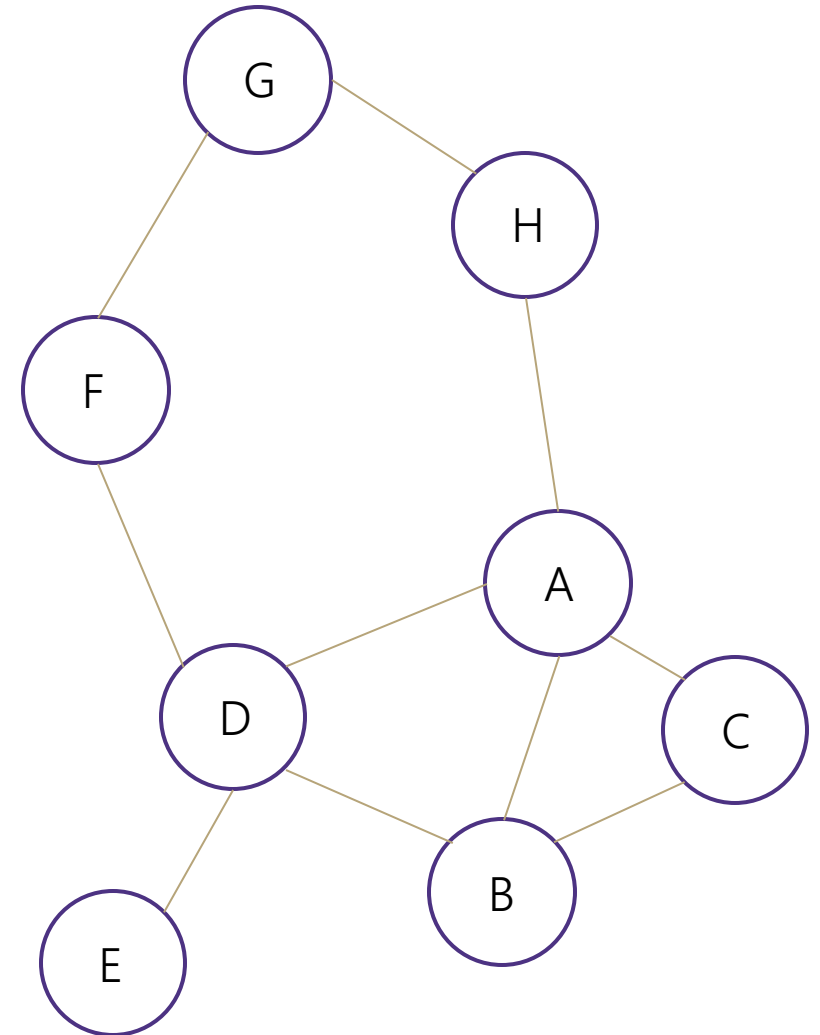
A **graph** is defined by a pair of sets  $G = (V, E)$  where...

- $V$  is a set of **vertices**
  - A vertex or "node" is a data entity

$V = \{A, B, C, D, E, F, G, H\}$

- $E$  is a set of **edges**
  - An edge is a connection between two vertices

$E = \{(A, B), (A, C), (A, D), (A, H),$   
 $(C, B), (B, D), (D, E), (D, F),$   
 $(F, G), (G, H)\}$



# Graph Vocabulary

## Graph Direction

- **Undirected graph** – edges have no direction and are two-way

$$V = \{ A, B, C \}$$

$$E = \{ (A, B), (B, C) \} \text{ inferred } (B, A) \text{ and } (C, B)$$

- **Directed graphs** – edges have direction and are thus one-way

$$V = \{ A, B, C \}$$

$$E = \{ (\underline{A}, \underline{B}), (B, C), (C, B) \}$$

## Degree of a Vertex

- **Degree** – the number of edges containing that vertex

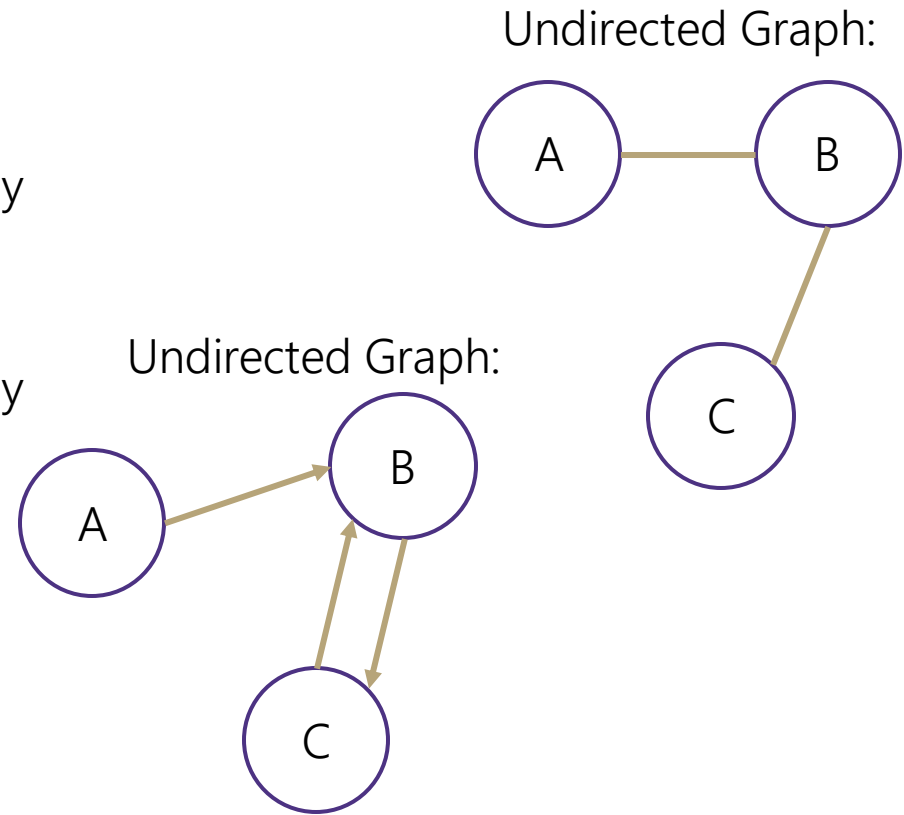
$$A : 1, B : 1, C : 1$$

- **In-degree** – the number of directed edges that point to a vertex

$$A : 0, B : 2, C : 1$$

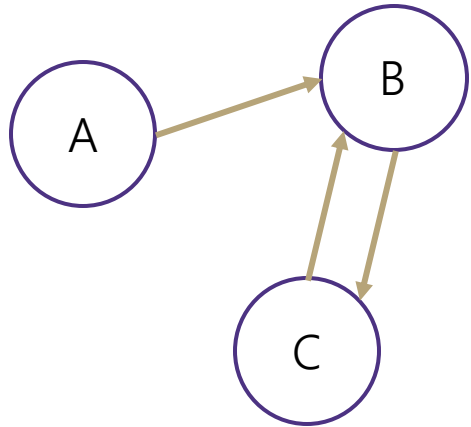
- **Out-degree** – the number of directed edges that start at a vertex

$$A : 1, B : 1, C : 1$$



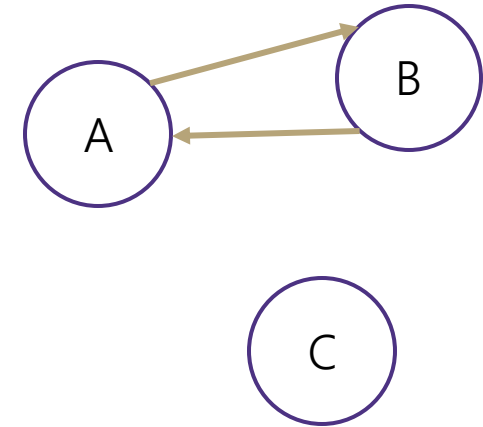
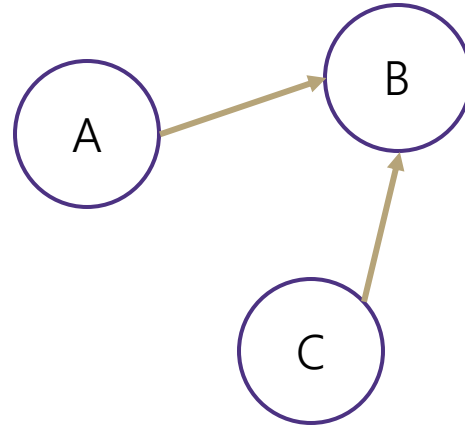
# Food for thought

Is a graph valid if there exists a vertex with a degree of 0? Yes



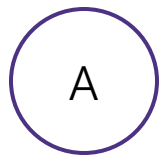
A has an "in degree" of 0

B has an "out degree" of 0

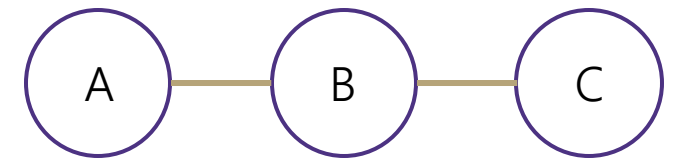
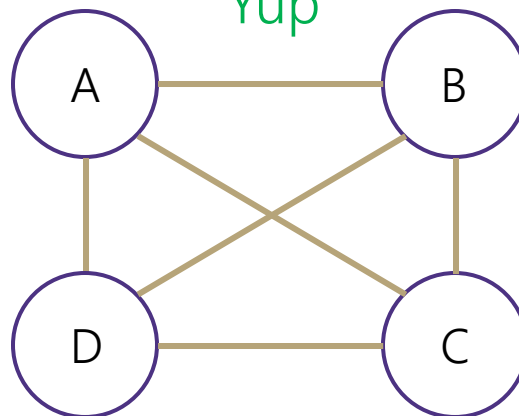


C has both an "in degree" and an "out degree" of 0

Is this a valid graph? Yup



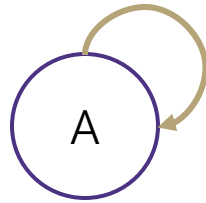
Yes!



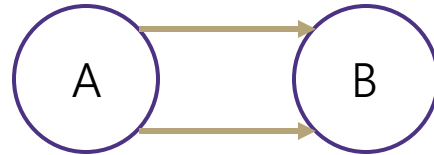
Sure

# Graph Vocabulary

**Self loop** – an edge that starts and ends at the same vertex



**Parallel edges** – two edges with the same start and end vertices



**Simple graph** – a graph with no self-loops and no parallel edges

**Complete graph** – a graph with edges between every pair of vertices

# For simple, undirected graphs...

What is the fewest number of edges a graph with  $n$  vertices can have?

0

What is the sum of the degrees of all vertices in a graph? (in terms of  $|V|$  and  $|E|$ )

$2|E|$

What is the maximum number of edges a graph with  $n$  vertices can have?

$n(n-1)/2 = O(n^2)$  - (the complete graph on  $n$  vertices  $K_n$ )





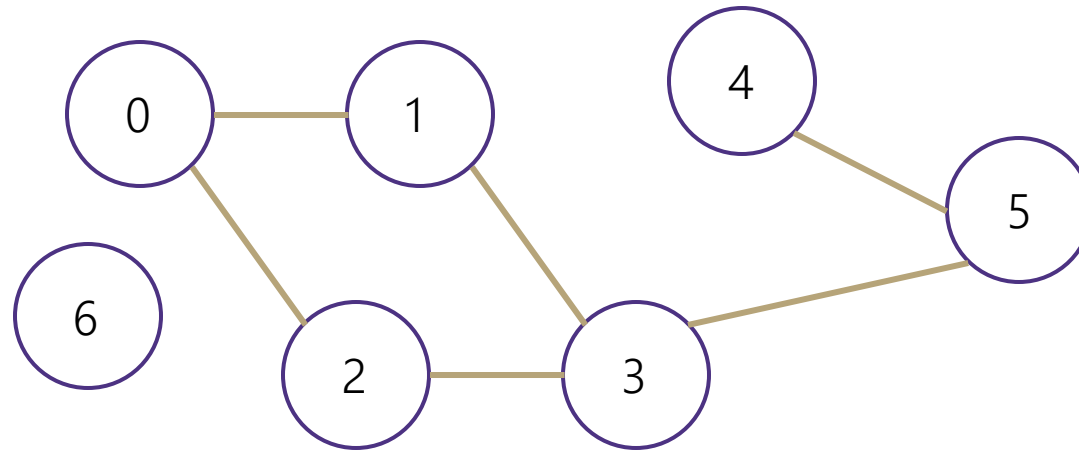
# Representing Graphs

*simple*

Discuss with your neighbor: How would you implement a non-weighted, undirected graph?

Assume there are  $n$  vertices, and those vertices are numbered  $0, 1, 2, \dots, n-1$ .

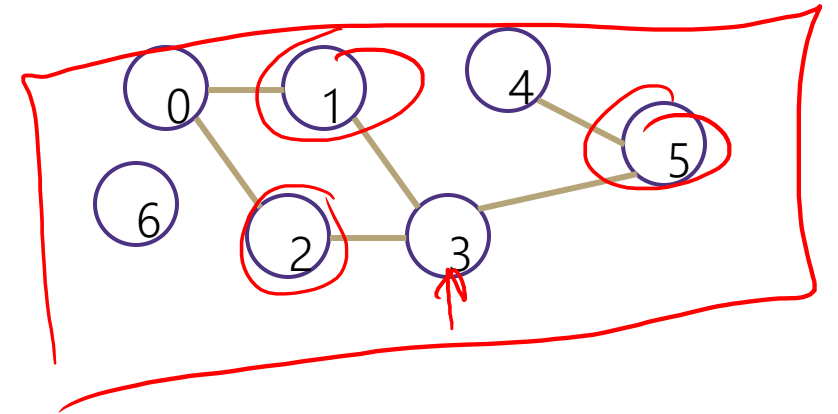
As an example:



$G = (V, E)$

$V = \{0,1,2,3,4,5,6\}$ ,  $E = \{ (0,1), (0,2), (1,3), (2,3), (3,5), (4,5) \}$

# Representing Graphs



$\text{Map}(\text{Vertex}, \text{Set}(\text{Vertex}))$

class Vertex:  
 DLL < vertices

adjacency list

+ Map, array of vertices

# Adjacency Matrix

A **matrix** is a table of numbers,  $a[u][v]$ .

In an adjacency matrix  $a[u][v]$  is 1 if there is an edge  $(u,v)$ , and 0 otherwise.

Can represent both undirected and directed graphs

Can represent self-loops and parallel edges (interpret as the # of edges between two vertices)

Time Complexity ( $|V| = n$ ,  $|E| = m$ ):

Add Edge:  $O(1)$

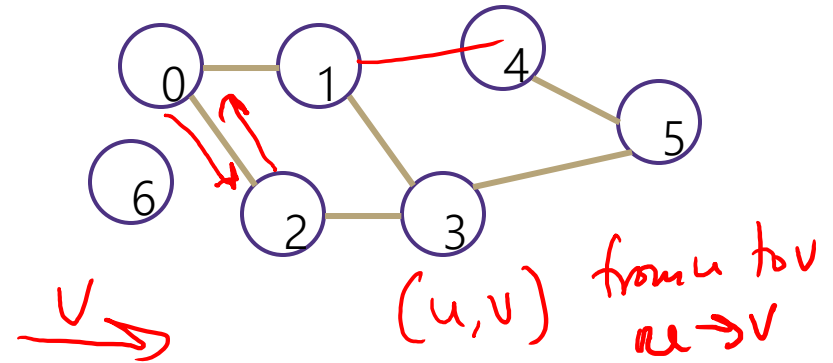
Remove Edge:  $O(1)$

Check edge exists from  $(u,v)$ :  $O(1)$

Get neighbors of  $u$  (out):  $O(n)$

Get neighbors of  $u$  (in):  $O(n)$

Space Complexity:  $O(n^2)$



|   | 0 | 1 | 2 | 3 | 4 | 5 | 6 |
|---|---|---|---|---|---|---|---|
| 0 | 0 | 1 | 1 | 0 | 0 | 0 | 0 |
| 1 | 1 | 0 | 0 | 1 | 1 | 0 | 0 |
| 2 | 1 | 0 | 0 | 1 | 0 | 0 | 0 |
| 3 | 0 | 1 | 1 | 0 | 0 | 1 | 0 |
| 4 | 0 | 1 | 0 | 0 | 0 | 1 | 0 |
| 5 | 0 | 0 | 0 | 1 | 1 | 0 | 0 |
| 6 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |

# Sparsity

A **Dense Graph** is a graph with many edges  $|E| \approx |V|^2$

A **Sparse Graph** is a graph with few edges  $|E| \approx |V|$

Adjacency Matrices are very wasteful of space for sparse graphs – they are almost all 0s!

How could we save space?

# Adjacency List

An array where the  $u$ 'th element contains a list of neighbors of  $u$ .

Can represent both undirected and directed graphs

In the directed case, put the out neighbors ( $a[u]$  has  $v$  for all  $(u,v)$  in  $E$ )

Can represent self-loops and parallel edges (repeat neighbor)

Time Complexity ( $|V| = n$ ,  $|E| = m$ ):

Add Edge:  $O(1)$

Remove Edge:  $O(\min(n, m))$

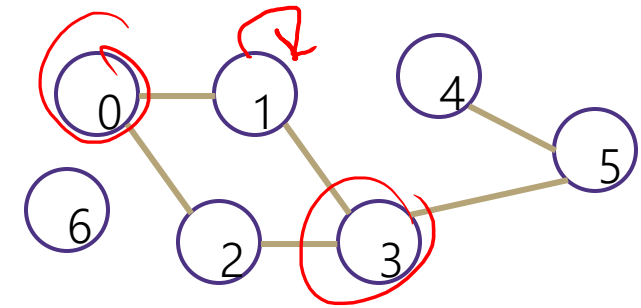
Check edge exists from  $(u,v)$ :  $O(\min(n, m))$

Get neighbors of  $u$  (out):  $O(n)$

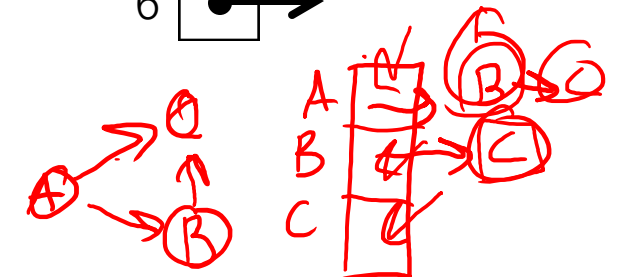
Get neighbors of  $u$  (in):  $O(n + m)$

Space Complexity:  $O(n + m)$

*m possible edges  
371 need to check all  
n array positions*



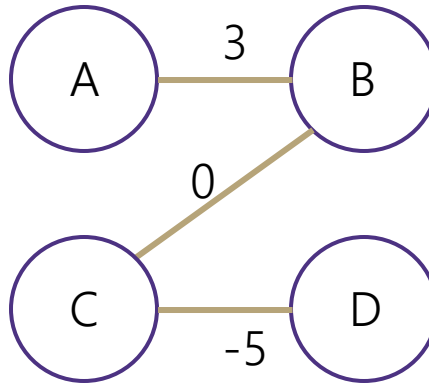
|   |     |           |
|---|-----|-----------|
| 0 | ● → | 1 → 2     |
| 1 | ● → | 0 → 3     |
| 2 | ● → | 0 → 3     |
| 3 | ● → | 1 → 2 → 5 |
| 4 | ● → | 5         |
| 5 | ● → | 3 → 4     |
| 6 | ● → |           |





# Graph Vocabulary

**Weighted Graph** – a graph with numeric weights associated with each edge



- can be directed or undirected
- weights can be negative, positive or 0
- common to specify “only non-negative edge weights”, or “only positive edge weights”
- denoted  $e = (u, v, w)$  or sometimes  $e = ((u,v), w)$
- Weights often carry meaning such as “distance”, (e.g. driving time between cities)

# Representing Weighted Graphs

Adjacency Matrix – You can make the value of at each element the **weight** of the edge

- In an int array (`int[]`) you can't distinguish between 0 weight edge and no edge
  - Solution 1 – You know ahead of time that 0 weight (or negative weight) edges do not exist and use that to represent no edge
  - Solution 2 (Java) – Use an Integer array (`Integer[]`) – have null represent no edge

Adjacency list – Store pairs (neighbor, edge weight)

# Storing Data In Graphs

We often have data associated with vertices and edges

## Vertex Data Examples:

- Facebook: Name, Age, Birthday, Hometown, Likes
- Google Maps: City name, elevation, hours of operation
- Internet: Page title, page contents, date last modified

## Edge Data Examples:

- Facebook: Date friendship was made, friend vs. acquaintance, etc.
- Google Maps: Length of road, speed limit

# Storing Data in Graphs

We could have a  $\text{Graph}\langle V, E \rangle$  where  $V$  is a data type for Vertices and  $E$  is one for Edges

Adjacency Matrix:  $E[][]$  – now each entry has a pointer to edge data, or null if that edge is not in the graph

Adjacency List:  $E[]$  – the neighbor lists are now a list of edges

Both: Maintain a list  $V[]$  of vertices

Alternative Adjacency List:  $V[]$  just a list of vertices, where each vertex contains within it a list of (outgoing) edges.

# Arrays → HashTables

When we analyze and describe graph algorithms, for simplicity we assume that each vertex has a unique identity 0, 1, 2, ...  $n-1$ .

Accesses in Adjacency Matrices and getting an adjacency list are both **worst case**  $O(1)$  for arrays.

In reality, we often don't have unique sequential integers for each vertex. In a real graph implementation, we often use **HashMaps** or **HashSets**.

This would get us **average case**  $O(1)$ , but **worst case**  $O(n)$ . This is bad in analysis, but fine in practice. An adjacency list that stores references to the actual vertex objects can avoid repeated table lookups.

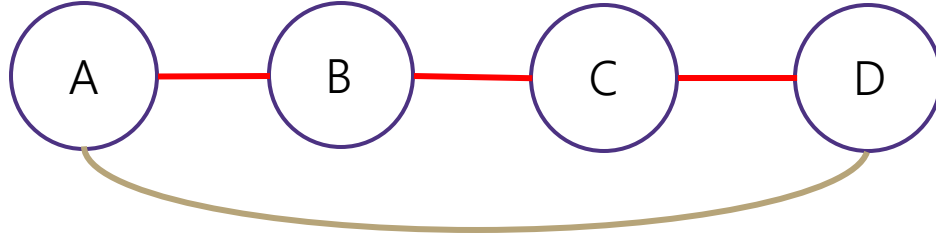


We *could* always use a hash table to give unique integer IDs to every element, incurring an  $O(n)$  **worst case**, but  $O(1)$  **average case** overhead before each call (which can save our worst case analysis for any  $O(n)$  or slower algorithm), but we don't usually bother to.

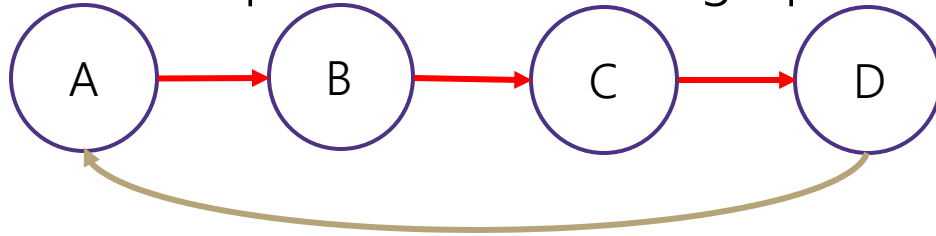


# Graph Vocabulary

**Path** – A sequence of connected vertices



**(Directed) Path** – A path in a directed graph must follow the direction of the edges



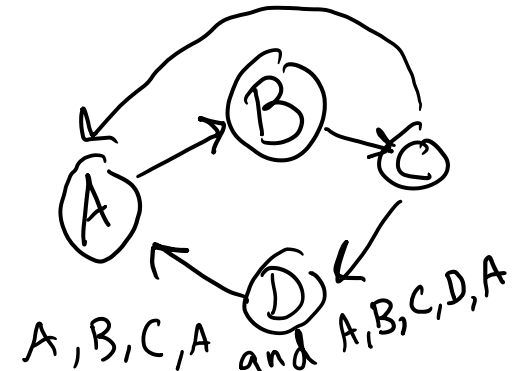
**Path Length (unweighted)** – The number of edges in a path

- (A,B,C,D) has length 3.

*weighted  $\rightarrow \sum \text{edge weights}$*

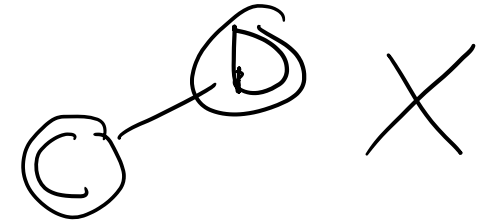
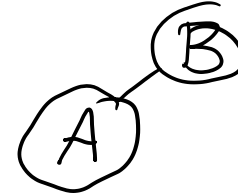
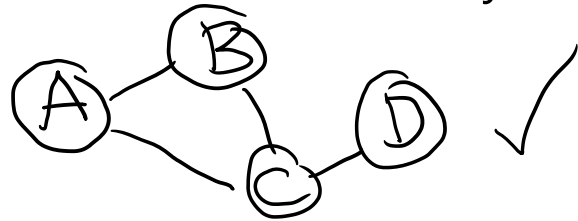
**Simple Path** – A path that doesn't repeat vertices (except maybe first=last)

**Cycle** – A path that starts and ends at the same vertex (of length at least 1)

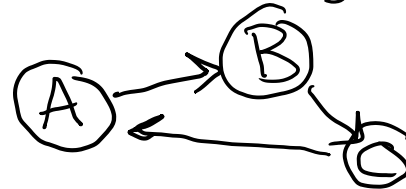


# Graph Vocabulary

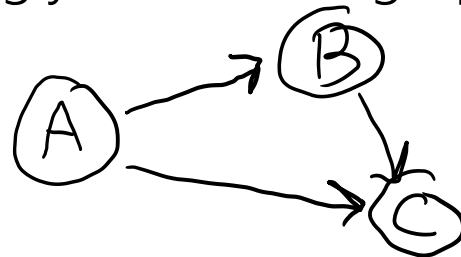
**Connected Graph** – A graph that has a path from every vertex to every other vertex (i.e. every vertex is **reachable** from every other vertex).



**Strongly Connected Graph** – A directed graph that is connected (note the direction of the edges!)



**Weakly Connected Graph** – A directed graph that is connected when interpreted as undirected. (Note all strongly connected graphs are also weakly connected)



# Paths and Reachability

Very common questions:

- Is there a path between two vertices? (Can I drive from Seattle to LA?)
- What is the length of the shortest path between two vertices? (How long will it take?)

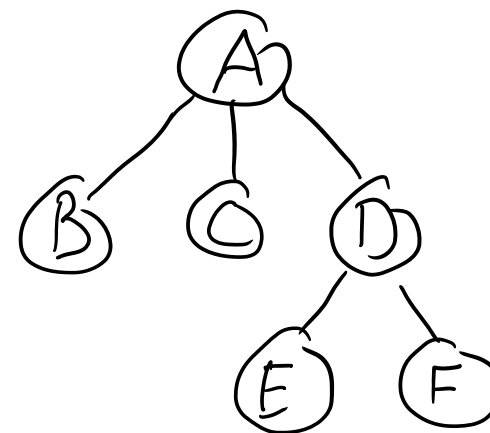
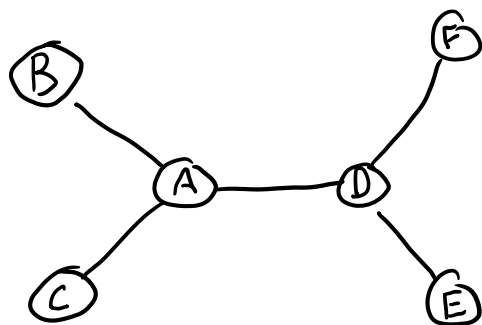
Less common, but still important:

What is the longest path in a graph?

- In general a “hard”™ problem
- 7 degrees of Kevin Bacon
- Length of this path is called the “diameter” of a graph

# Trees

A **tree** is a connected, acyclic graph.



In a tree there exist **exactly one** path between every pair of vertices.

A graph consisting of several disconnected trees is called a **forest**.

The trees we have seen so far have been **rooted trees** – interpret one vertex as the **root**, and its neighbors are now **children**, and the root of their own **subtrees**.

How many edges does a tree with  $n$  vertices have?  $n - 1$

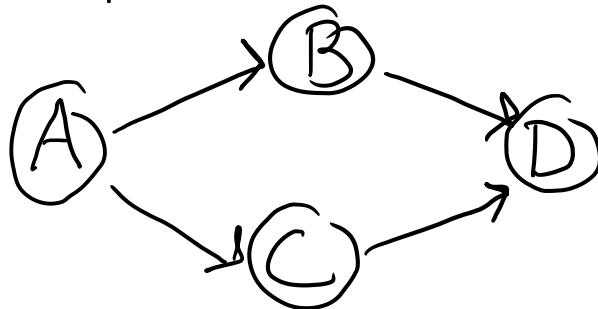
# DAGs

DAG stands for **D**irected, **A**cyclic, **G**raph

This is the directed graph analog of a forest.

The trees we have made so far in this class have been implemented as weakly connected DAGs.

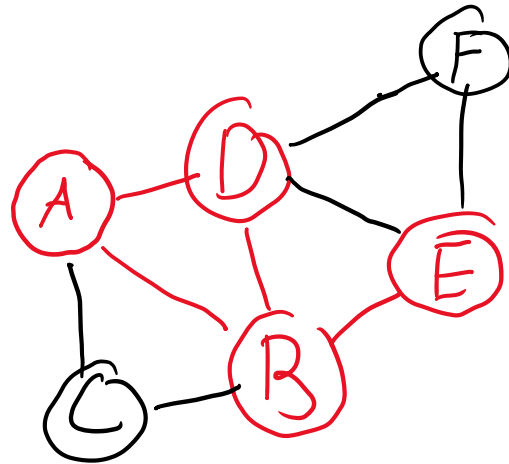
Can be used to represent **dependencies**: i.e. A must be completed before either B or C, and both B and C must be completed before D. Scheduling these tasks is called **topological sort**.





# Subgraphs

Take a graph, and delete vertices and edges so you still have a graph.



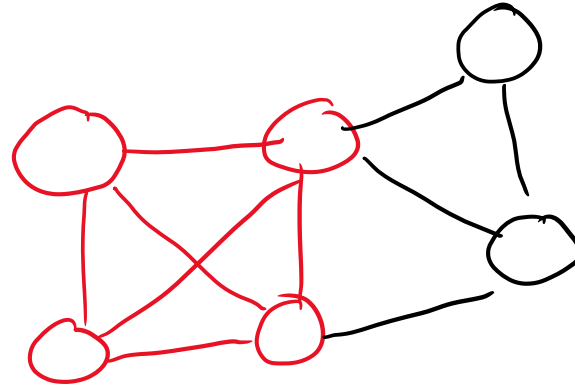
The remaining **red** vertices and edges form a subgraph.

Formally:  $G' = (V', E') \subseteq G = (V, E) \Leftrightarrow V' \subseteq V$  and  $E' \subseteq E$  and  $G'$  is a graph

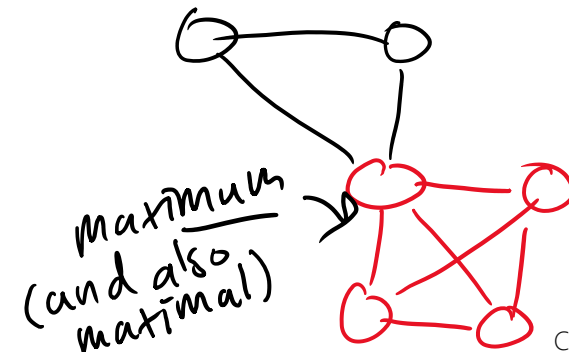
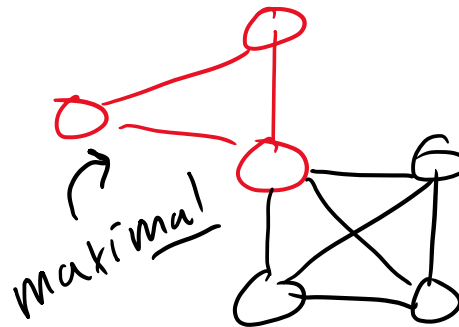
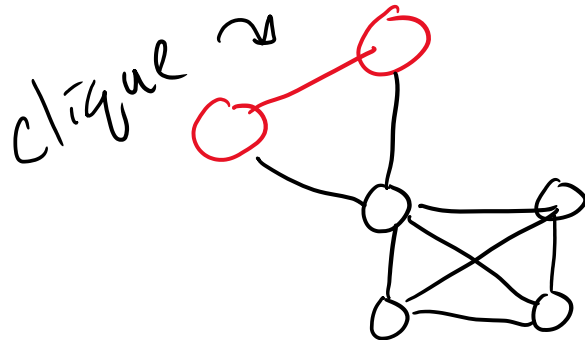
# Interesting Subgraphs

## Cliques

- A **clique** is a complete subgraph



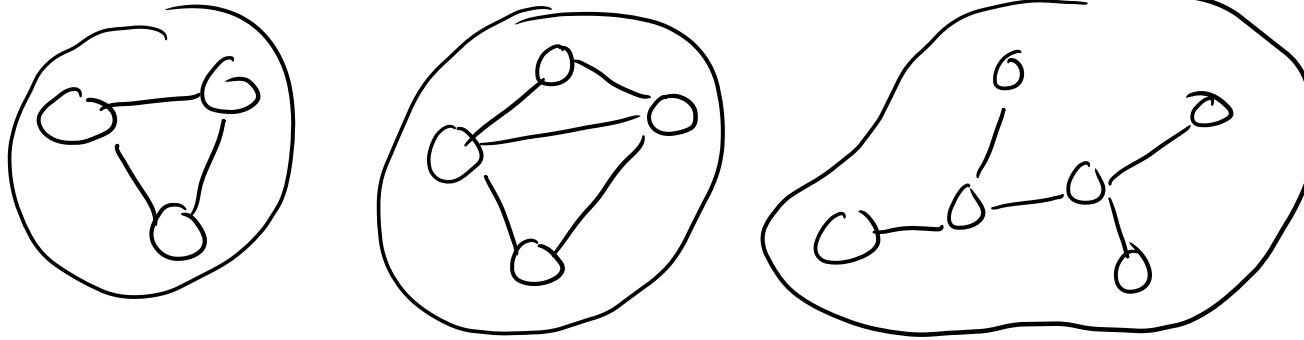
- A **maximal** clique is a clique that you could not add any more vertices to and still have a clique. This is distinct from the **maximum** clique, which is the largest clique in a graph.



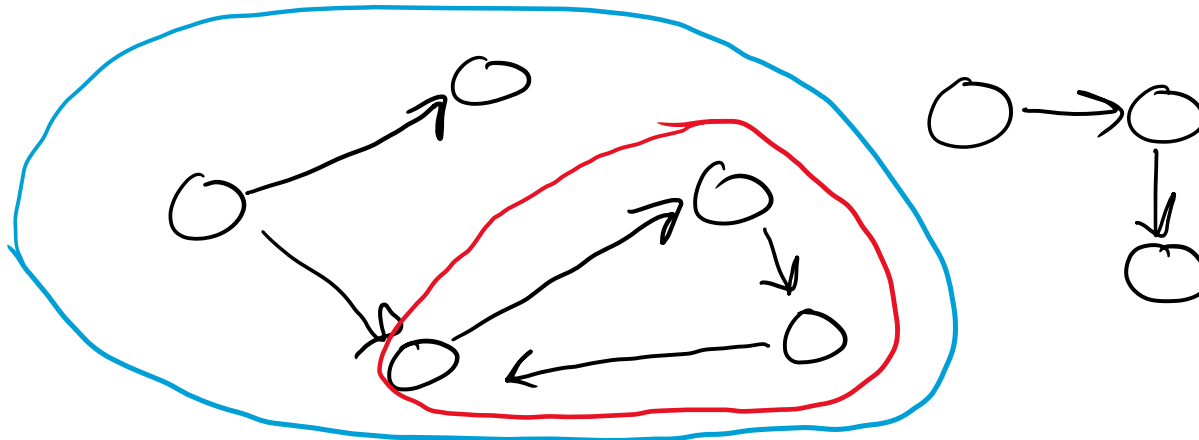
# Interesting Subgraphs

## Connected Components

- A connected component is a maximal, connected subgraph

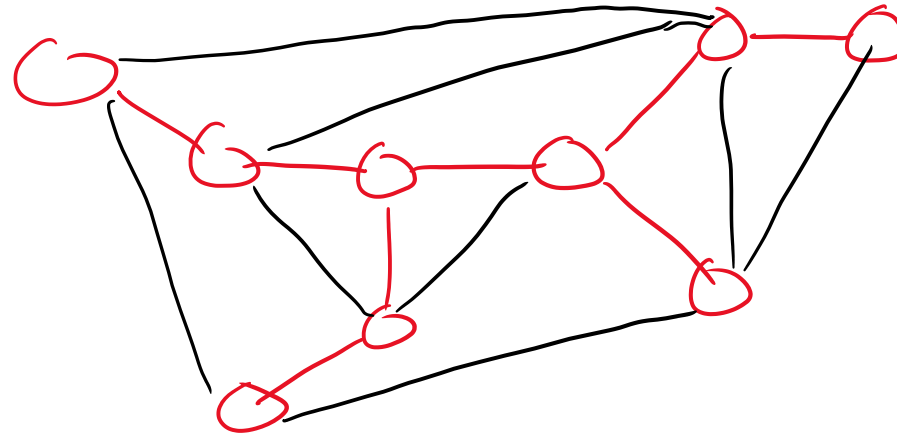


- In **directed** graphs, you have two kinds: **strongly connected**, and **weakly connected**:



# Interesting Subgraphs

A **Spanning Tree** is a subgraph that is both a **tree** and includes **every vertex** (it **spans** the graph).



Every **connected** graph has at least one spanning tree. They are like skeletons of the graph.

An important problem is finding the **minimum spanning tree**. We will learn 2 algorithms for this. It is useful for optimization tasks (e.g. minimum cost to build a road network).

# Other interesting Graph Problems

- Circuits – paths or cycles that touch every vertex
- Reductions – Everything we've seen so far in this class can be represented as a graph – a lot of other problems can too! Graphs can solve many problems.