



CSE373: Data Structures & Algorithms

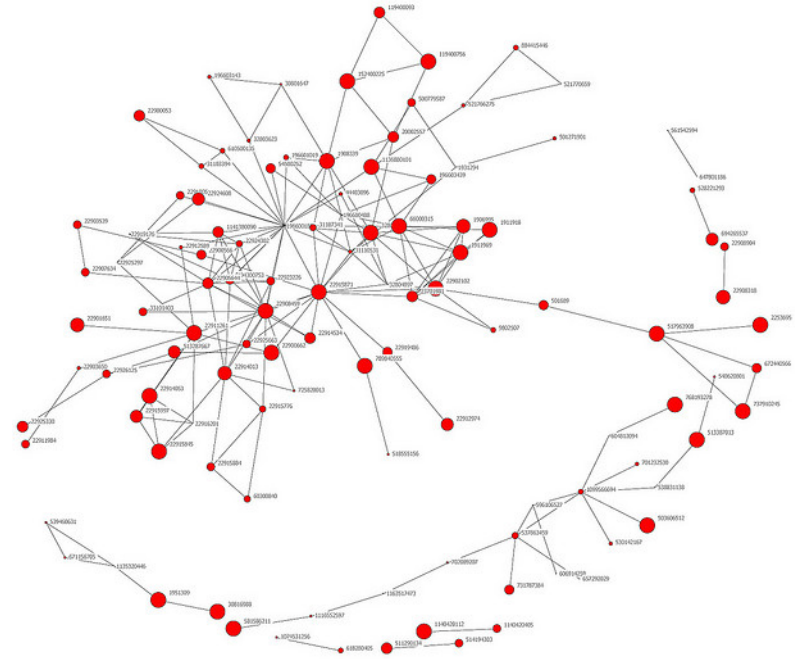
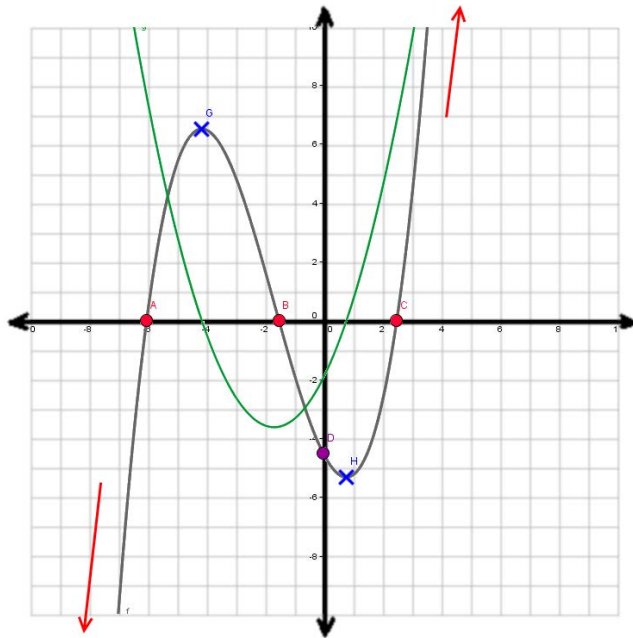
Lecture 16: Introduction to Graphs

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Announcements

What is a Graph?



Which kind of graph are we going to study?

Graphs: the mathematical definition

- A graph is a formalism for representing relationships among items
 - Very general definition because very general concept

- A graph is a pair

$$G = (V, E)$$

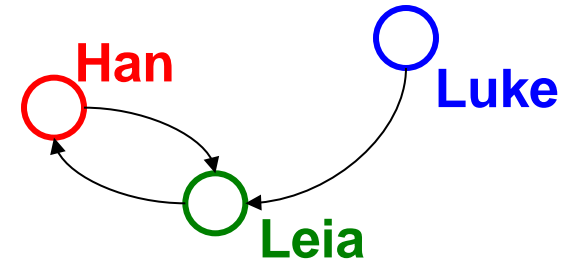
- A set of vertices, also known as nodes

$$V = \{v_1, v_2, \dots, v_n\}$$

- A set of edges

$$E = \{e_1, e_2, \dots, e_m\}$$

- Each edge e_i is a pair of vertices
(v_j, v_k)
- An edge “connects” the vertices

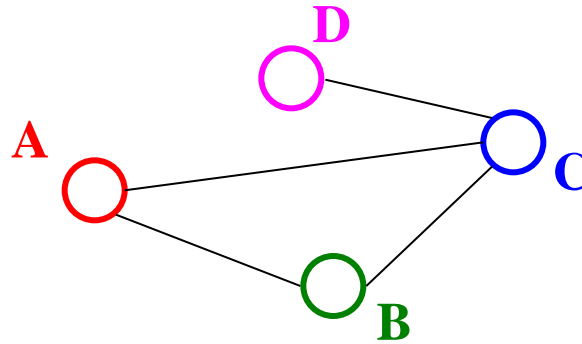


$$V = \{\text{Han}, \text{Leia}, \text{Luke}\}$$
$$E = \{(\text{Luke}, \text{Leia}), (\text{Han}, \text{Leia}), (\text{Leia}, \text{Han})\}$$

- Graphs can be directed or undirected

Undirected Graphs

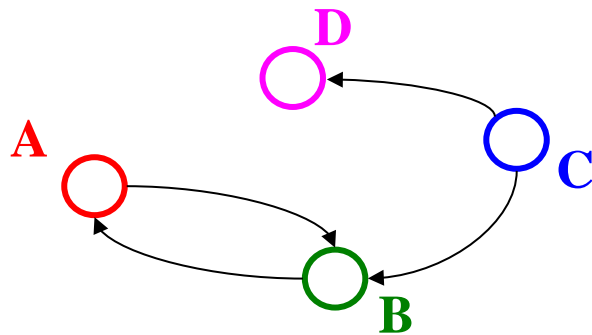
- In **undirected graphs**, edges have no specific direction
 - Edges are always “two-way”



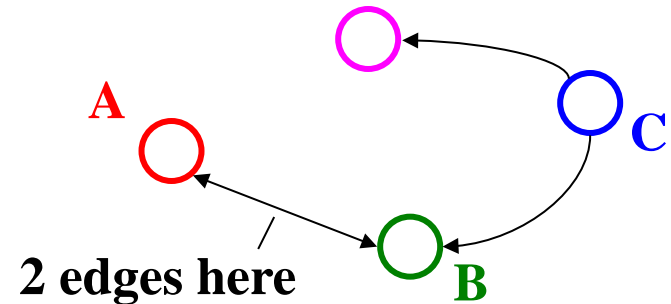
- Thus, $(u, v) \in \mathbf{E}$ implies $(v, u) \in \mathbf{E}$ (What do we call this property?)
 - Only one of these edges needs to be in the set
 - The other is implicit, so normalize how you check for it
- **Degree** of a vertex: number of edges containing that vertex
 - Put another way: the number of adjacent vertices

Directed Graphs

- In **directed graphs** (sometimes called **digraphs**), edges have a direction

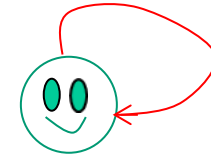


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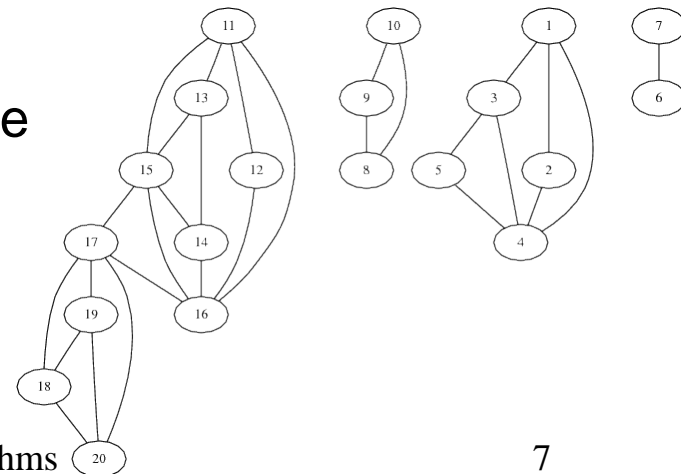


- Thus, $(u, v) \in \mathbf{E}$ does *not* imply $(v, u) \in \mathbf{E}$.
 - Let $(u, v) \in \mathbf{E}$ mean $u \rightarrow v$
 - Call u the **source** and v the **destination**
- In-degree** of a vertex: number of in-bound edges, i.e., edges where the vertex is the destination
- Out-degree** of a vertex: number of out-bound edges i.e., edges where the vertex is the source

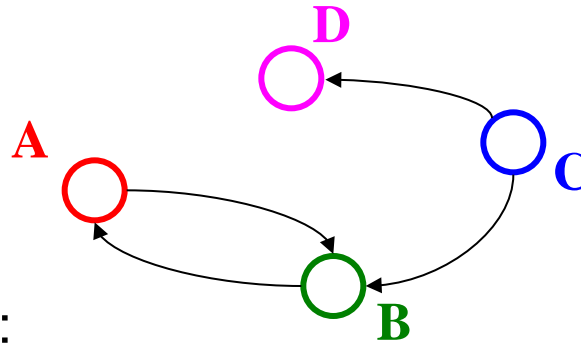
Self-Edges, Connectedness



- A **self-edge** a.k.a. a **loop** is an edge of the form (u, u)
 - Depending on the use/algorithm, a graph may have:
 - No self edges
 - Some self edges
 - All self edges (often therefore implicit, but we will be explicit)
- A node can have a degree / in-degree / out-degree of **zero**
- A graph does not have to be **connected**
 - Even if every node has non-zero degree



More notation

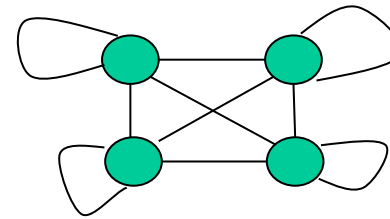


$$V = \{A, B, C, D\}$$

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

For a graph $G = (V, E)$:

- $|V|$ is the number of vertices
- $|E|$ is the number of edges



- Minimum?

- Maximum for undirected? $\frac{|V|(|V+1)|}{2} \in O(|V|^2)$

- Maximum for directed? $|V|^2 \in O(|V|^2)$

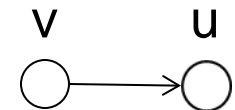
(assuming self-edges allowed, else subtract $|V|$)

- If $(u, v) \in E$

- Then v is a **neighbor** of u , i.e., v is **adjacent** to u

- **Order matters for directed edges**

- u is not **adjacent** to v unless $(v, u) \in E$



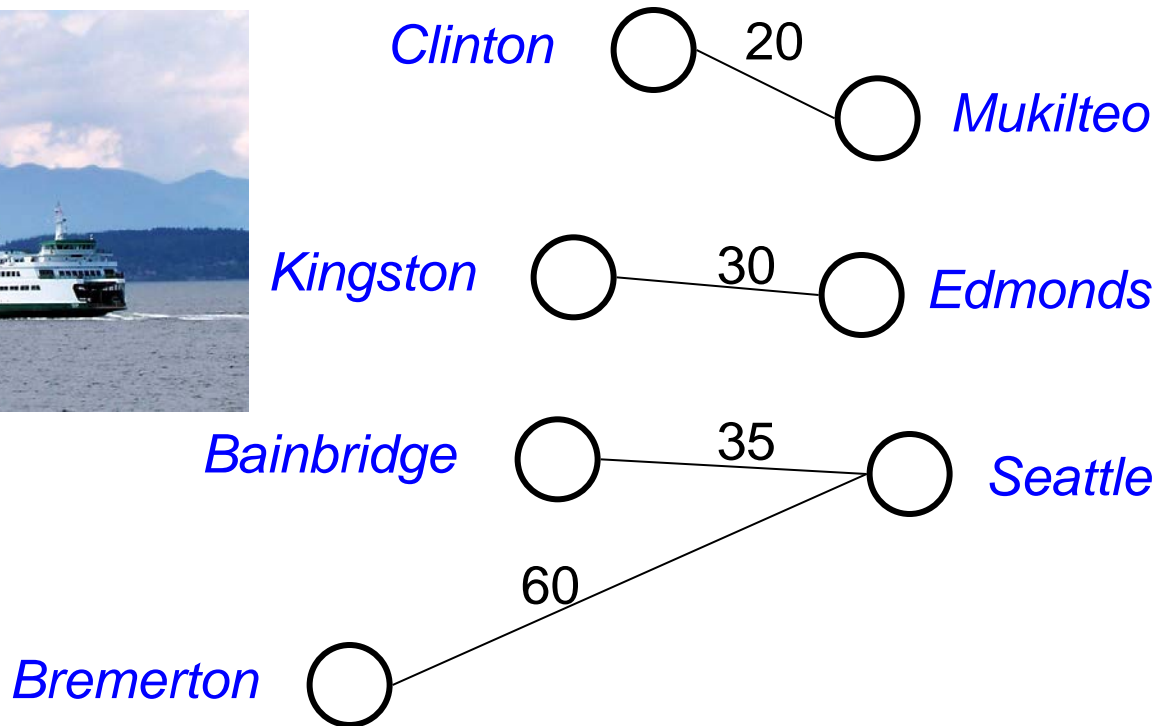
Examples

Which would use **directed edges**? Which would have **self-edges**?
Which would be **connected**? Which could have **0-degree nodes**?

1. Web pages with links
2. Facebook friends
3. Methods in a program that call each other
4. Road maps (e.g., Google maps)
5. Airline routes
6. Family trees
7. Course pre-requisites

Weighted Graphs

- In a weighed graph, each edge has a **weight** a.k.a. **cost**
 - Typically numeric (most examples use ints)
 - *Orthogonal* to whether graph is directed
 - Some graphs allow *negative weights*; many do not



Examples

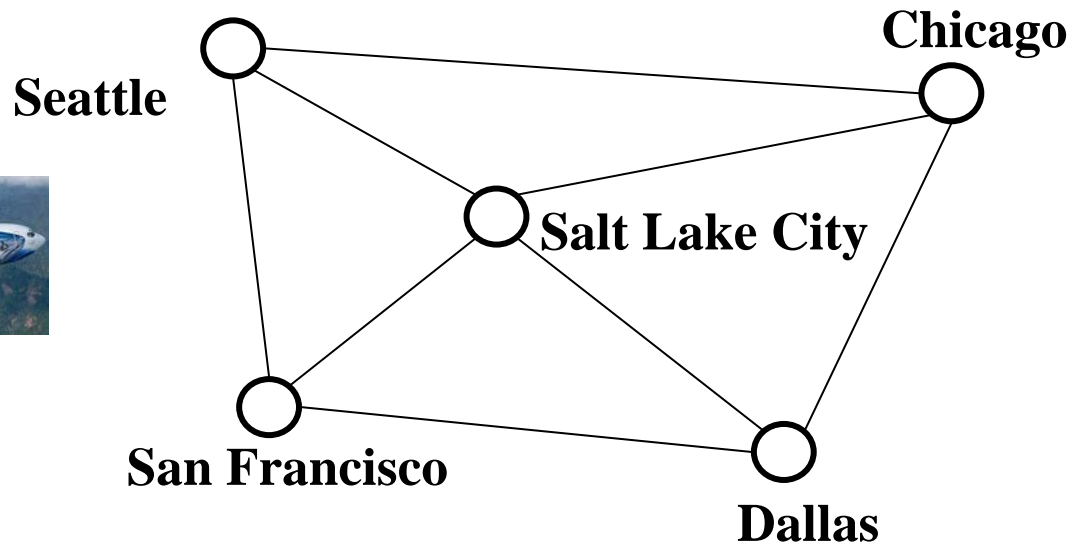
What, if anything, might weights represent for each of these?

Do negative weights make sense?

- Web pages with links
- Facebook friends
- Methods in a program that call each other
- Road maps (e.g., Google maps)
- Airline routes
- Family trees
- Course pre-requisites

Paths and Cycles

- A **path** is a list of vertices $[v_0, v_1, \dots, v_n]$ such that $(v_i, v_{i+1}) \in E$ for all $0 \leq i < n$. Say “a path from v_0 to v_n ”
- A **cycle** is a path that begins and ends at the same node ($v_0 = v_n$)



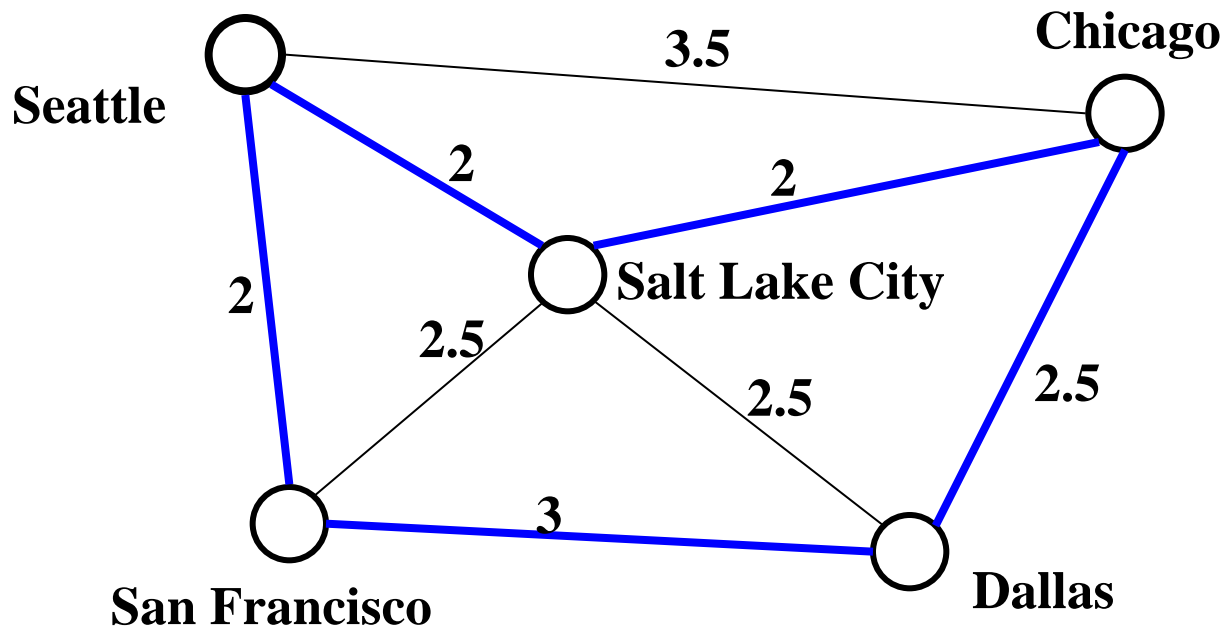
Example: [Seattle, Salt Lake City, Chicago, Dallas, San Francisco, Seattle]

Path Length and Cost

- Path length: Number of *edges* in a path
- Path cost: Sum of *weights* of edges in a path

Example where

$P = [\text{Seattle}, \text{Salt Lake City}, \text{Chicago}, \text{Dallas}, \text{San Francisco}, \text{Seattle}]$



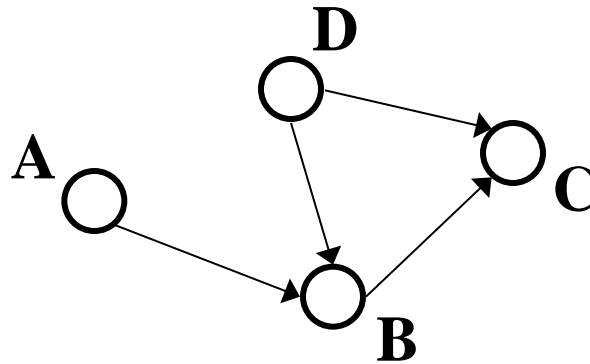
$$\text{length}(P) = 5$$
$$\text{cost}(P) = 11.5$$

Simple Paths and Cycles

- A **simple path** repeats no vertices, except the first might be the last
[Seattle, Salt Lake City, San Francisco, Dallas]
[Seattle, Salt Lake City, San Francisco, Dallas, Seattle]
- Recall, a **cycle** is a path that ends where it begins
[Seattle, Salt Lake City, San Francisco, Dallas, Seattle]
[Seattle, Salt Lake City, Seattle, Dallas, Seattle]
- A **simple cycle** is both a cycle and a simple path
[Seattle, Salt Lake City, San Francisco, Dallas, Seattle]

Paths and Cycles in Directed Graphs

Example:

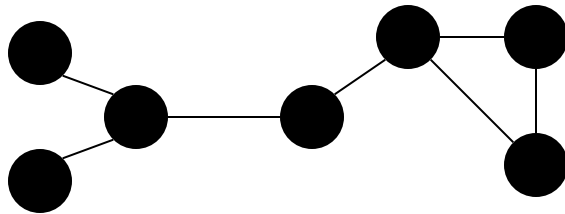


Is there a path from A to D? **No**

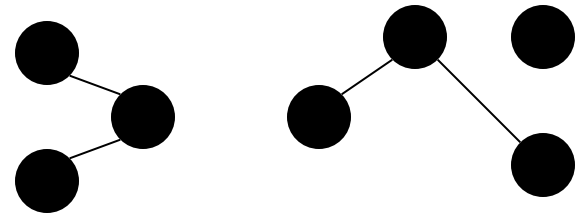
Does the graph contain any cycles? **No**

Undirected-Graph Connectivity

- An undirected graph is **connected** if for all pairs of vertices u, v , there exists a *path* from u to v

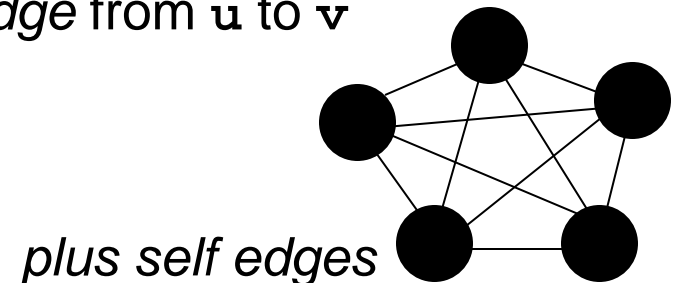


Connected graph



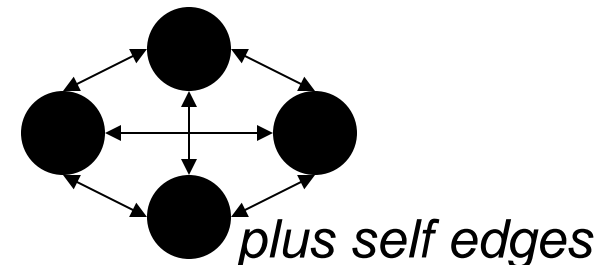
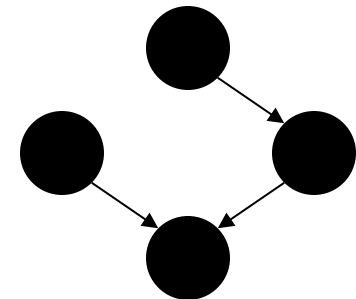
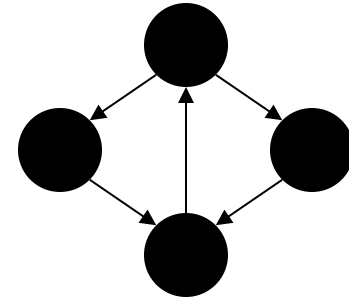
Disconnected graph

- An undirected graph is **complete**, a.k.a. **fully connected** if for *all* pairs of vertices u, v , there exists an *edge* from u to v



Directed-Graph Connectivity

- A directed graph is **strongly connected** if there is a path from every vertex to every other vertex
- A directed graph is **weakly connected** if there is a path from every vertex to every other vertex *ignoring direction of edges*
- A **complete** a.k.a. **fully connected** directed graph has an edge from every vertex to every other vertex



Trees as Graphs

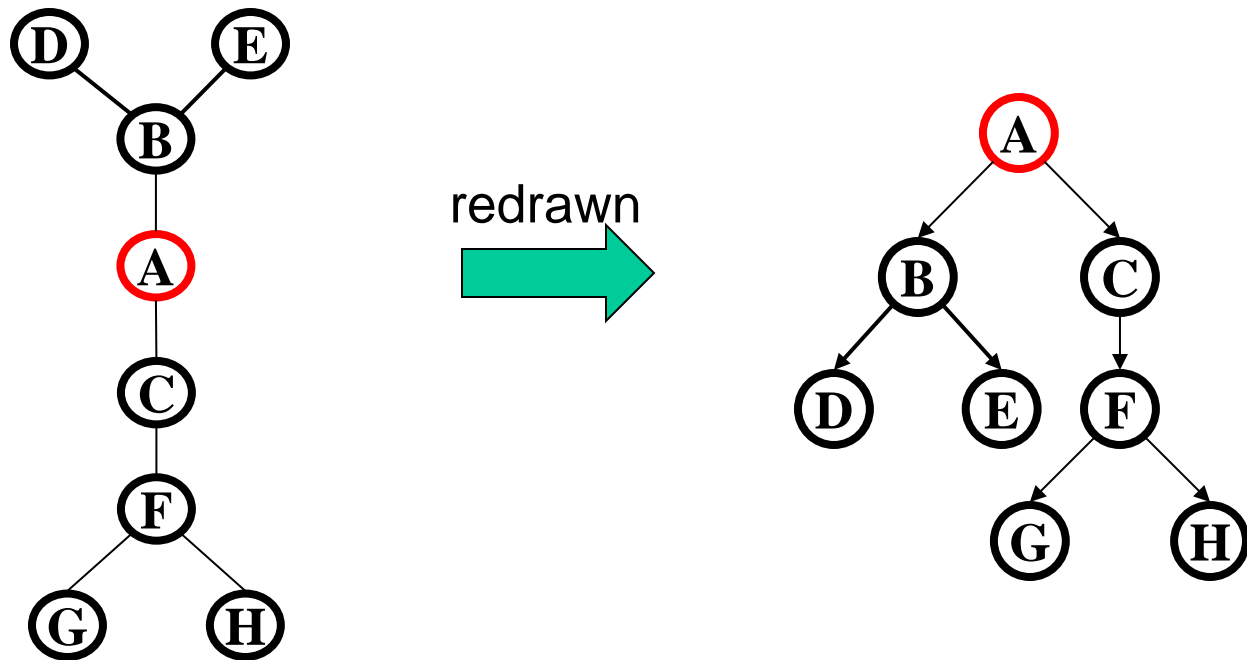
When talking about graphs,
we say **a tree is a graph** that is:

- **Undirected**
- **Acyclic**
- **Connected**

So all trees are graphs, but not
all graphs are trees

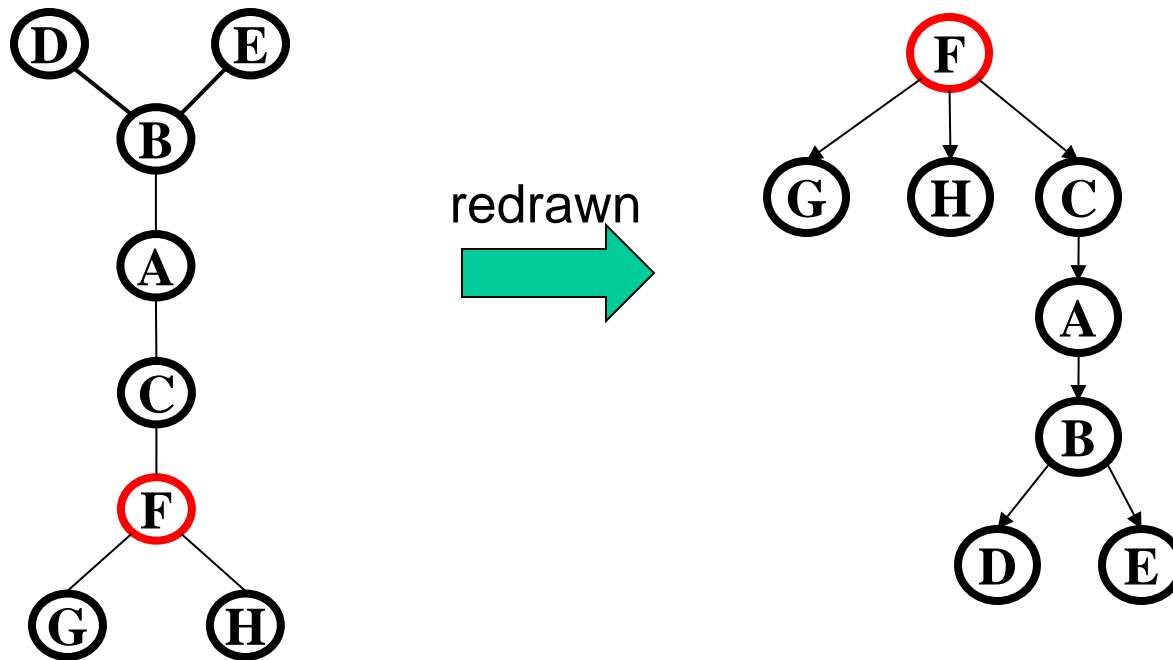
Rooted Trees

- We are more accustomed to **rooted trees** where:
 - We identify a unique root
 - We think of edges as directed: parent to children
- Given a tree, picking a root gives a unique rooted tree
 - The tree is just drawn differently



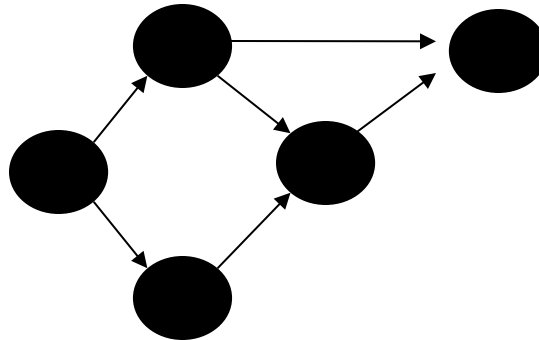
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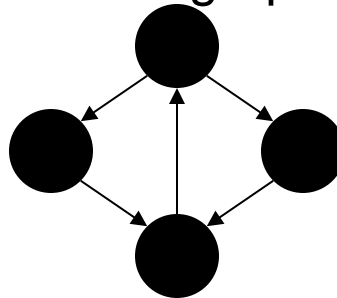


Directed Acyclic Graphs (DAGs)

- A **DAG** is a directed graph with no (directed) cycles
 - Every rooted directed tree is a DAG
 - **But not every DAG is a rooted directed tree**



- Every DAG is a directed graph
- But not every directed graph is a DAG



Examples

Which of our directed-graph examples do you expect to be a DAG?

- Web pages with links
- Methods in a program that call each other
- Airline routes
- Family trees
- Course pre-requisites

Density / Sparsity

- Let E be the set of **edges** and V the set of **vertices**.
- Then $0 \leq |E| \leq |V|^2$
- And $|E|$ is $O(|V|^2)$

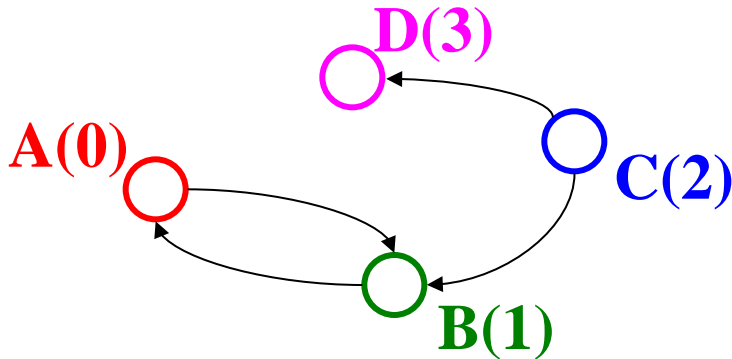
- Because $|E|$ is often much smaller than its maximum size, we do not always approximate $|E|$ as $O(|V|^2)$
 - This is a correct bound, it just is often not tight
 - If it is tight, i.e., $|E|$ is $\Theta(|V|^2)$ we say the graph is **dense**
 - More sloppily, dense means “lots of edges”
 - If $|E|$ is $O(|V|)$ we say the graph is **sparse**
 - More sloppily, sparse means “most possible edges missing”

What is the Data Structure?

- So graphs are really useful for lots of data and questions
 - For example, “what’s the lowest-cost path from x to y ”
- But we need a data structure that represents graphs
- The “best one” can depend on:
 - Properties of the graph (e.g., dense versus sparse)
 - The common queries (e.g., “is (u, v) an edge?” versus “what are the neighbors of node u ?”)
- So we’ll discuss the **two standard graph representations**
 - **Adjacency Matrix** and **Adjacency List**
 - Different trade-offs, particularly time versus space

Adjacency Matrix

- Assign each node a number from 0 to $|V| - 1$
- A $|V| \times |V|$ **matrix** (i.e., 2-D array) of **Booleans** (or **1 vs. 0**)
 - If M is the matrix, then $M[u][v]$ being **true** means there is an edge from u to v



	0	1	2	3
0	F	T	F	F
1	T	F	F	F
2	F	T	F	T
3	F	F	F	F

Adjacency Matrix Properties

- Running time to:
 - Get a vertex's out-edges: $O(|V|)$
 - Get a vertex's in-edges: $O(|V|)$
 - Decide if some edge exists: $O(1)$
 - Insert an edge: $O(1)$
 - Delete an edge: $O(1)$
- Space requirements:
 - $|V|^2$ bits
- Best for sparse or dense graphs?
 - Best for dense graphs

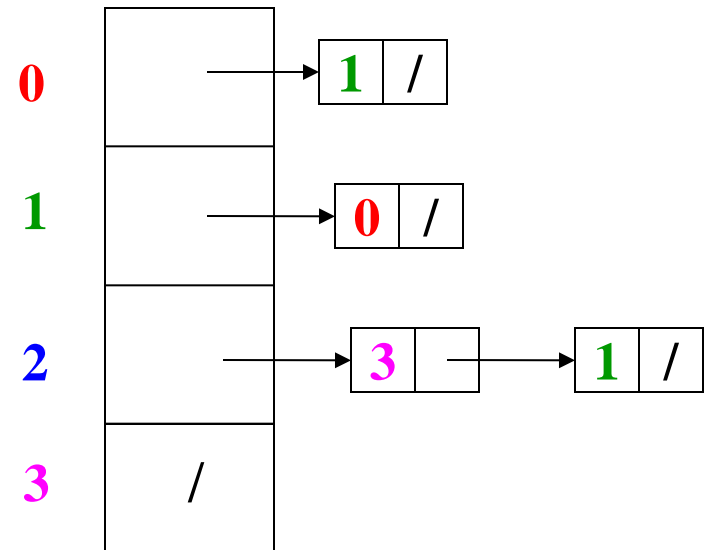
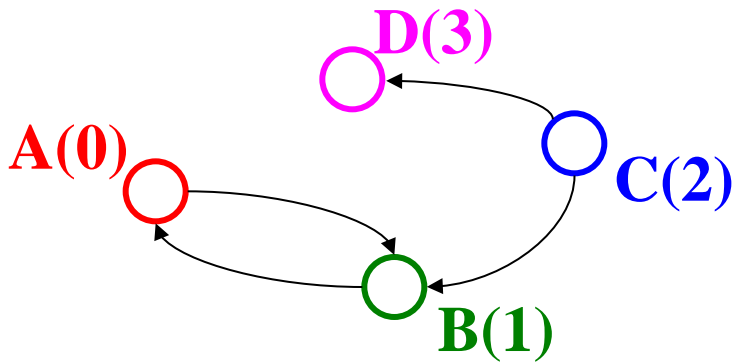
	0	1	2	3
0	F	T	F	F
1	T	F	F	F
2	F	T	F	T
3	F	F	F	F

Adjacency Matrix Properties

- How will the adjacency matrix vary for an *undirected graph*?
 - Undirected will be symmetric around the diagonal
- How can we adapt the representation for *weighted graphs*?
 - Instead of a Boolean, store a number in each cell
 - Need some value to represent 'not an edge'
 - In *some* situations, 0 or -1 works

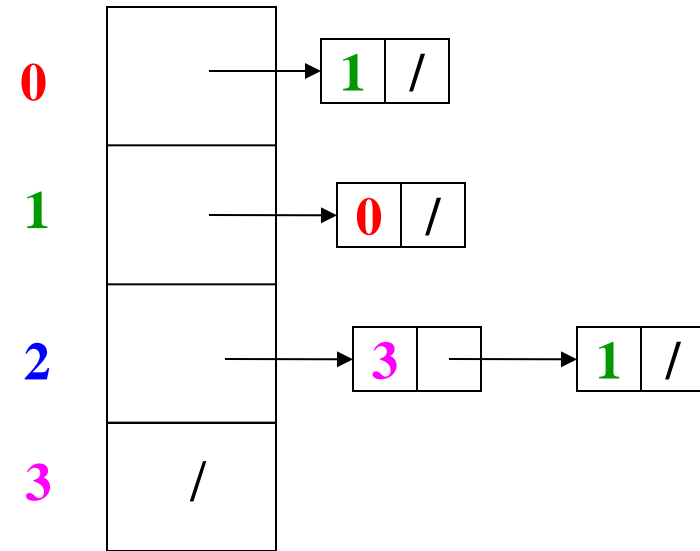
Adjacency List

- Assign each node a number from 0 to $|V| - 1$
- An array of length $|V|$ in which each entry stores a list of all adjacent vertices (e.g., linked list)



Adjacency List Properties

- Running time to:
 - Get all of a vertex's out-edges:
 $O(d)$ where d is out-degree of vertex
 - Get all of a vertex's in-edges:
 $O(|E|)$ (but could keep a second adjacency list for this!)
 - Decide if some edge exists:
 $O(d)$ where d is out-degree of source
 - Insert an edge:
 $O(1)$ (unless you need to check if it's there)
 - Delete an edge:
 $O(d)$ where d is out-degree of source
- Space requirements:
 - $O(|V|+|E|)$
- Good for sparse graphs



Next...

Okay, we can represent graphs

Next lecture we'll implement some useful and non-trivial algorithms

- **Topological sort:** Given a DAG, order all the vertices so that every vertex comes before all of its neighbors
- **Shortest paths:** Find the shortest or lowest-cost path from x to y
 - Related: Determine if there even is such a path