

# Lecture 9

## Introduction to Graphs

# Graphs

- A graph is a formalism for representing relationships among items. One way to write graphs:

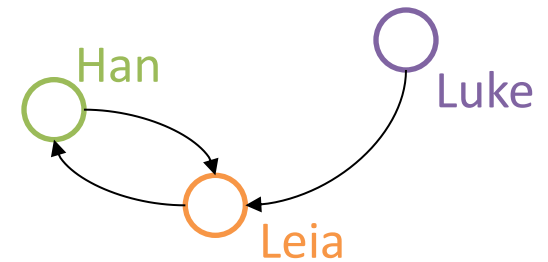
- A **graph**  $G = (V, E)$ 
  - A set of **vertices**, also known as **nodes**

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

- 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



$$V = \{\text{Han}, \text{Leia}, \text{Luke}\}$$

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

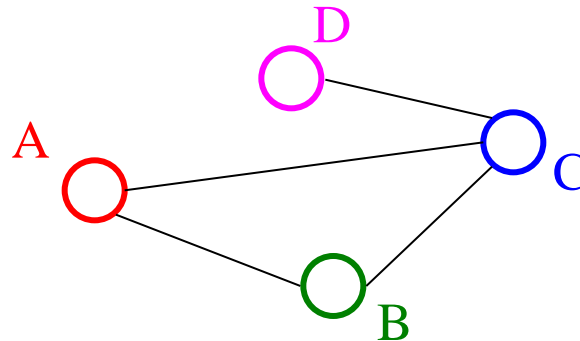
- Graphs can be **directed** or **undirected**

# Are Graphs An ADT?

- Can think of graphs as an ADT with operations like **isEdge** ( $(v_j, v_k)$ ) , **addVertex** ( $v_{\text{new}}$ ) , ...
- But it is unclear what the “standard operations” are
- Instead we tend to develop algorithms over graphs and then use data structures that are efficient for those algorithms
- Many important problems can be solved by:
  1. Formulating them in terms of graphs
  2. Applying a standard graph algorithm
- To make the formulation easy and standard, we have a lot of *standard terminology* about graphs

# Undirected Graphs

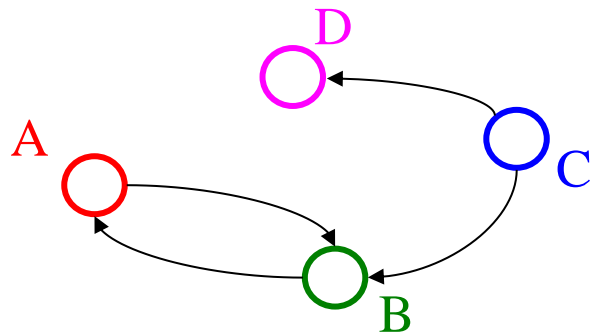
- In **undirected graphs**, edges have no specific direction
  - Edges are always “two-way”



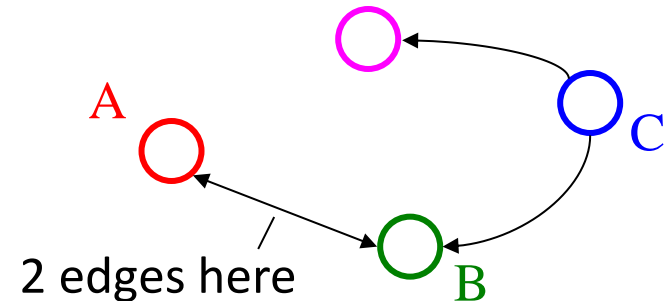
- Thus,  $(u, v) \in E$  implies  $(v, u) \in E$ 
  - Only one of these edges needs to be in the set
  - The other is implicit, so normalize how you check for it
- **Degree** of a vertex: number of edges containing that vertex
  - Put another way: the number of adjacent vertices

# Directed Graphs

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



or

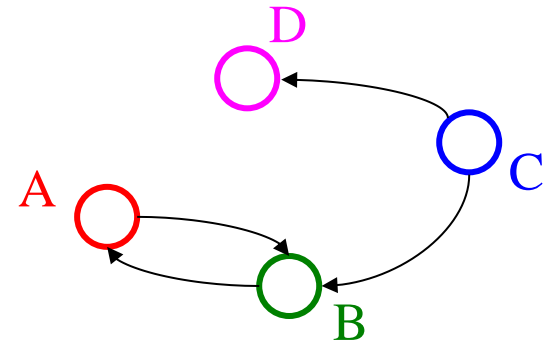


- Thus,  $(u, v) \in E$  does *not* imply  $(v, u) \in E$ .
  - Let  $(u, v) \in E$  mean  $u \rightarrow v$
  - Call  $u$  the **source** and  $v$  the **destination**
- In-degree** of a vertex: number of in-bound edges, i.e., edges where the vertex is the destination
- Out-degree** of a vertex: number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



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

- $|V|$  is the number of vertices
- $|E|$  is the number of edges (assuming no self loops)
  - Minimum? 0
  - Maximum for directed?  $|V| * (|V| - 1) \in O(|V|^2)$
  - Maximum for undirected?  $(|V| * (|V| - 1)) / 2 \in O(|V|^2)$

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

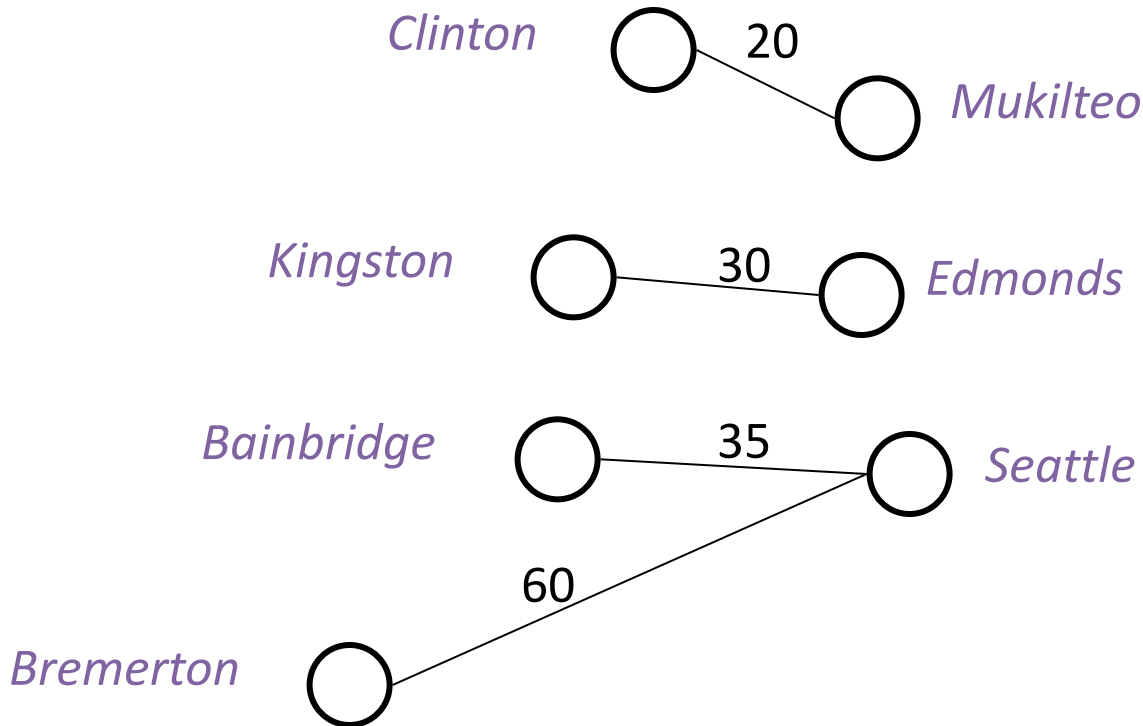
- Then  $v$  is a **neighbor** of  $u$ , i.e.,  $v$  is **adjacent** to  $u$
- Order matters for directed edges
  - $u$  is not **adjacent** to  $v$  unless  $(v, u) \in E$

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

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

# Weighted Graphs

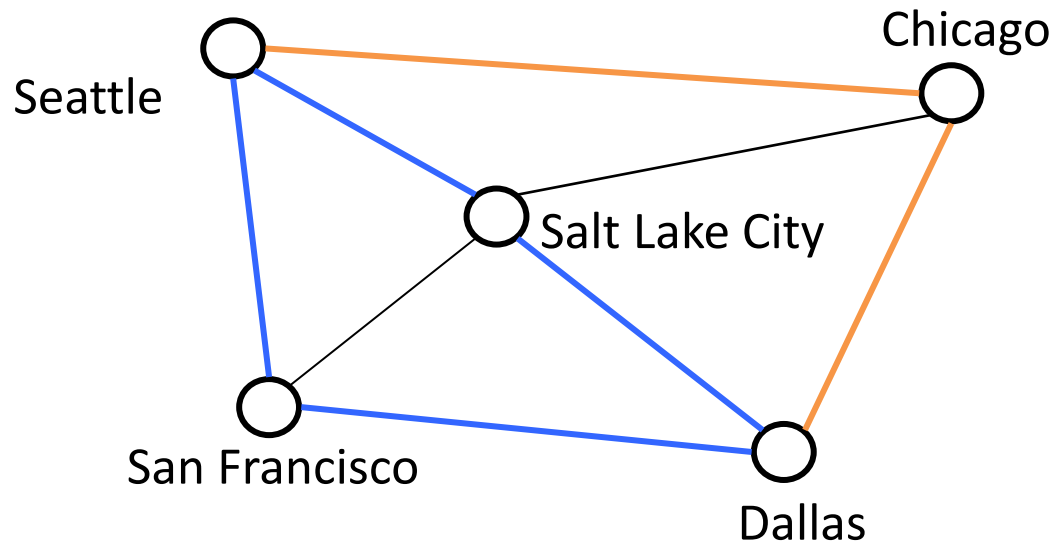
- In a weighed graph, each edge has a **weight** a.k.a. **cost**
  - Typically numeric (most examples use ints)
  - *Orthogonal* to whether graph is directed
  - Some graphs allow *negative weights*; many do not





# 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 ( $v_0 == v_n$ )



**Path:** [Seattle, Chicago, Dallas]

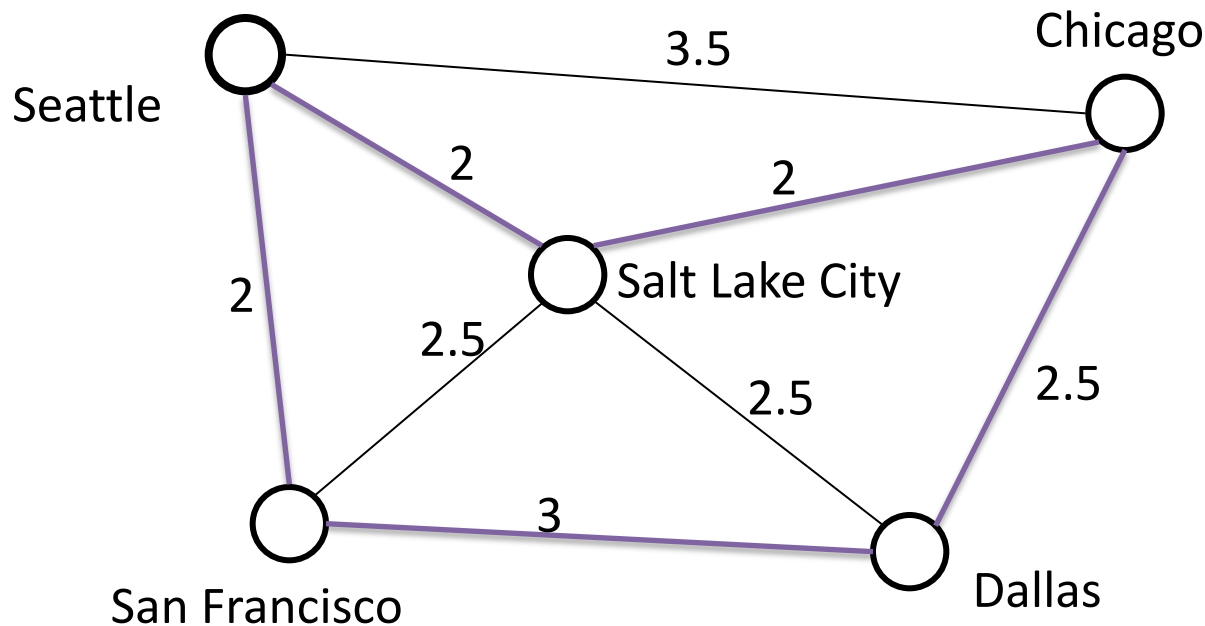
**Cycle:** [Seattle, Salt Lake City, Dallas, San Francisco, Seattle]

# Path Length and Cost

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

Example:

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



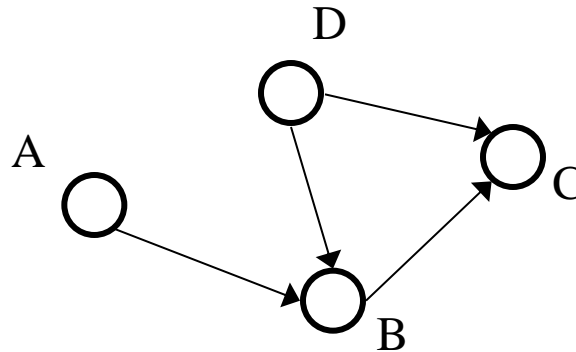
$\text{length}(P) = 5$   
 $\text{cost}(P) = 11.5$

# 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 a cycle and a simple path  
[Seattle, Salt Lake City, San Francisco, Dallas, Seattle]

# Paths and Cycles in Directed Graphs

Example:

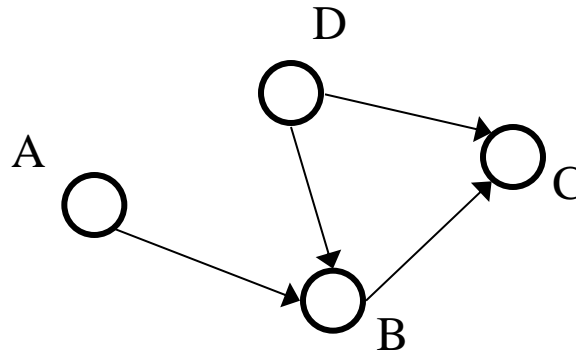


Is there a path from A to D?

Does the graph contain any cycles?

# Paths and Cycles in Directed Graphs

Example:

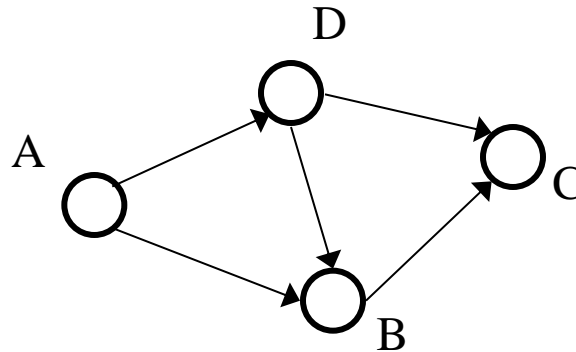


Is there a path from A to D? No

Does the graph contain any cycles? No

# Paths and Cycles in Directed Graphs

Example:

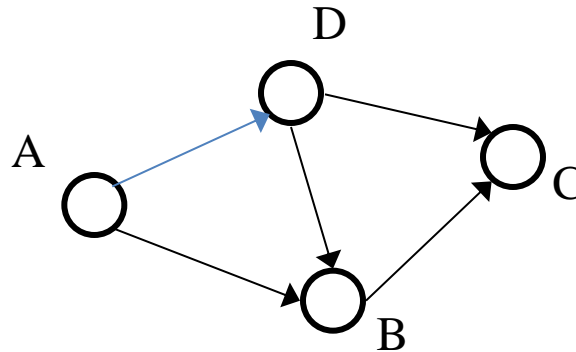


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# Paths and Cycles in Directed Graphs

Example:

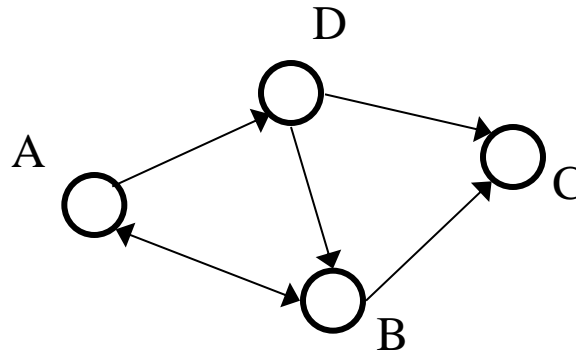


Is there a path from A to D? Yes

Does the graph contain any cycles? No

# Paths and Cycles in Directed Graphs

Example:



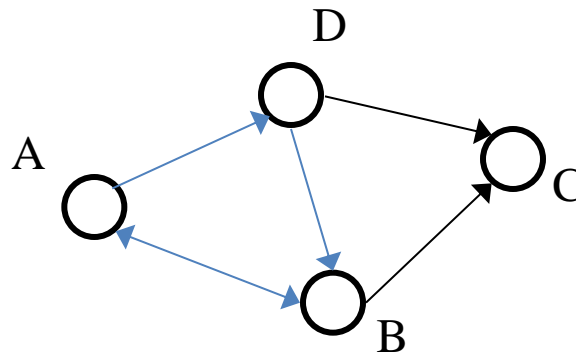
Is there a path from A to D?

Does the graph contain any cycles?



# Paths and Cycles in Directed Graphs

Example:

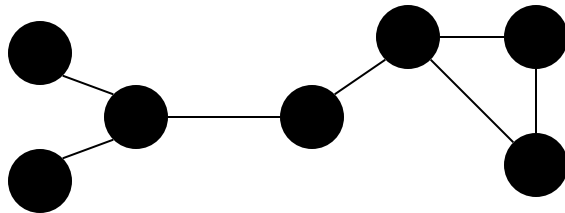


Is there a path from A to D? **Yes**

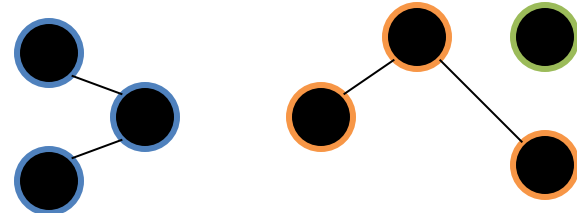
Does the graph contain any cycles? **Yes**

# Undirected-Graph Connectivity

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

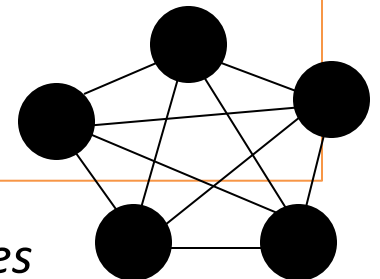


Connected graph



Disconnected graph

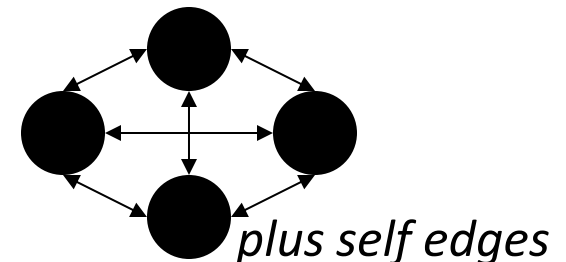
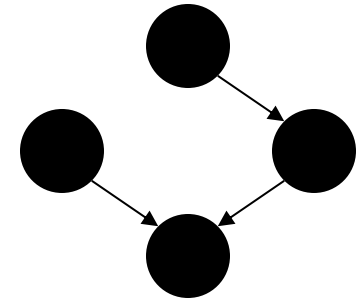
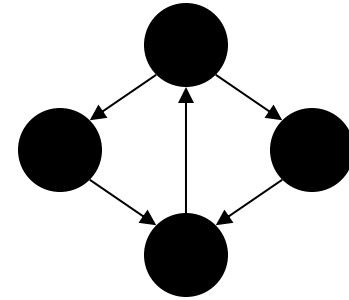
- An undirected graph is **complete**, a.k.a. **fully connected** if for *all* pairs of vertices  $u, v$ , there exists an *edge* from  $u$  to  $v$



*plus self edges*

# 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



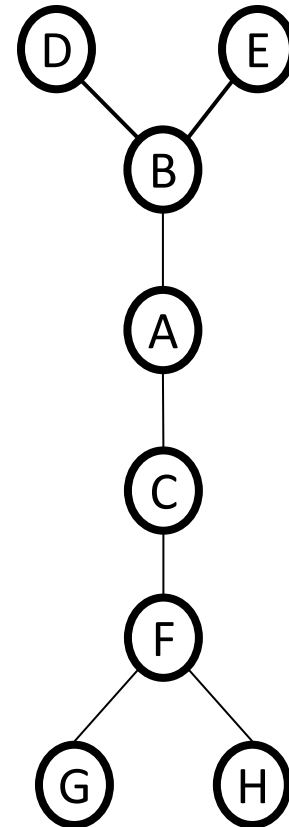
# Trees as Graphs

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

- Acyclic (no cycles)
- Connected

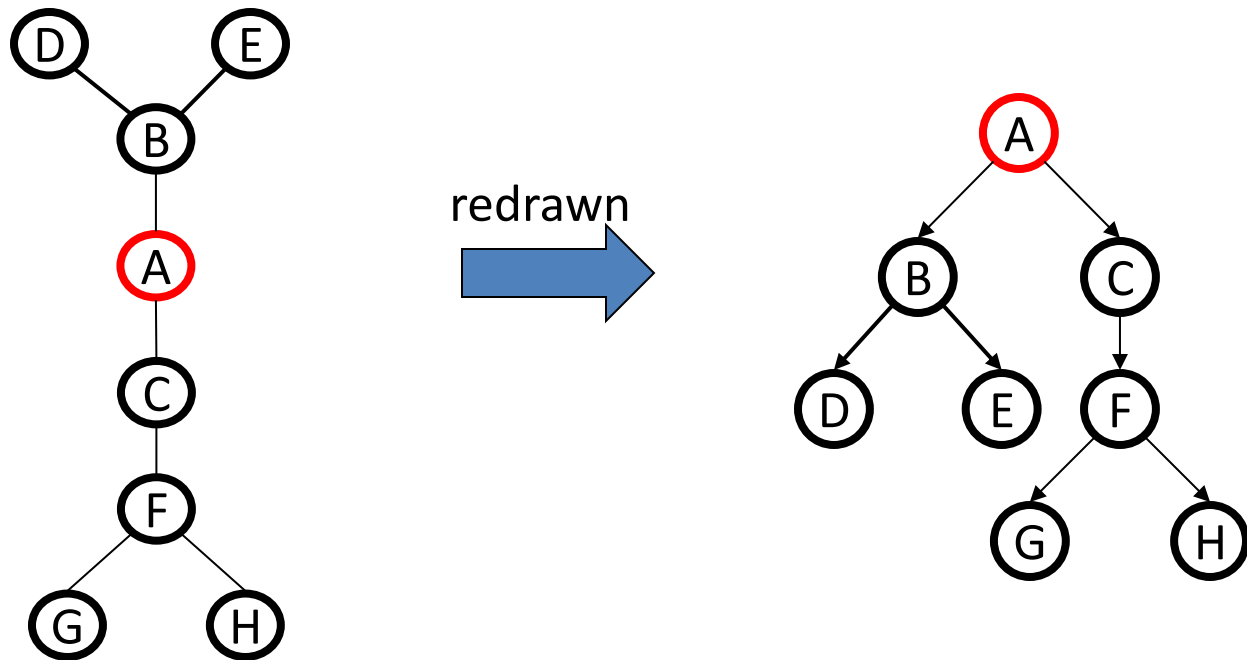
So all trees are graphs,  
but not all graphs are  
trees

Example:



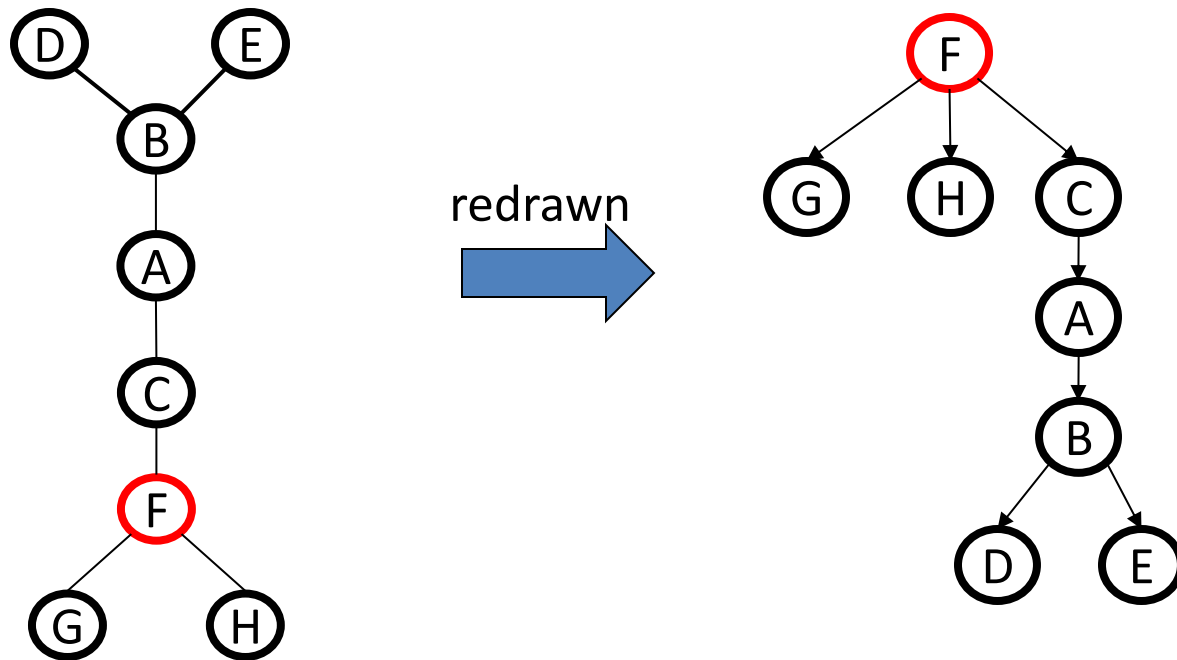
# Rooted Trees

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



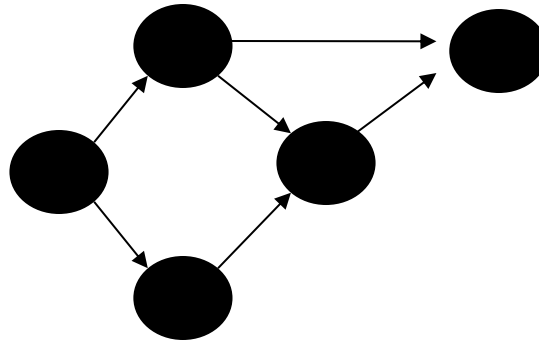
# Rooted Trees

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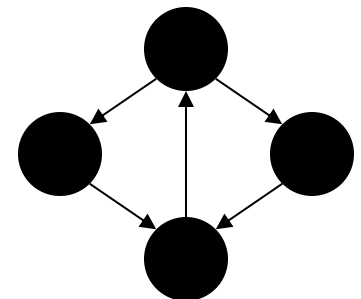


# Directed Acyclic Graphs (DAGs)

- 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



- Not every directed graph is acyclic



# Density / Sparsity

- Recall: In an undirected graph,  $0 \leq |E| < |V|^2$
- Recall: In a directed graph:  $0 \leq |E| \leq |V|^2$
- So for any graph,  $O(|E| + |V|^2)$  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 upper bound, it just is often not tight
  - If it is tight, i.e.,  $|E|$  is  $\Theta(|V|^2)$  we say the graph is **dense**
  - If  $|E|$  is  $O(|V|)$  we say the graph is **sparse**

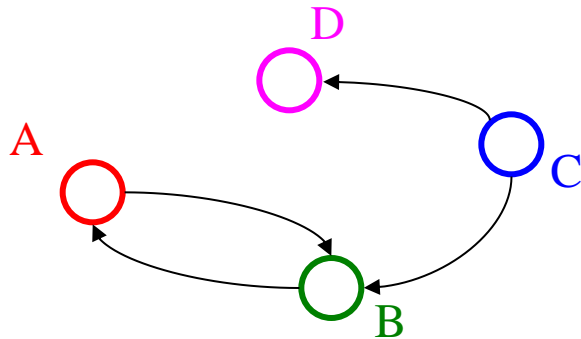


# How do we implement this?

- The “best” implementation can depend on:
  - Properties of the graph (e.g., dense vs sparse)
  - The common queries (e.g., “is  $(\mathbf{u}, \mathbf{v})$  an edge?” vs “what are the neighbors of node  $\mathbf{u}$ ?”)
- We’ll discuss the two standard graph representations
  - Adjacency Matrix and Adjacency List
  - Different trade-offs, particularly time versus space

# Adjacency Matrix

- Assign each vertex/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  $\mathbf{M}$  is the matrix, then  $\mathbf{M}[\mathbf{u}][\mathbf{v}]$  being **true** means there is an edge from  $\mathbf{u}$  to  $\mathbf{v}$



	A	B	C	D
A	F	T	F	F
B	T	F	F	F
C	F	T	F	T
D	F	F	F	F

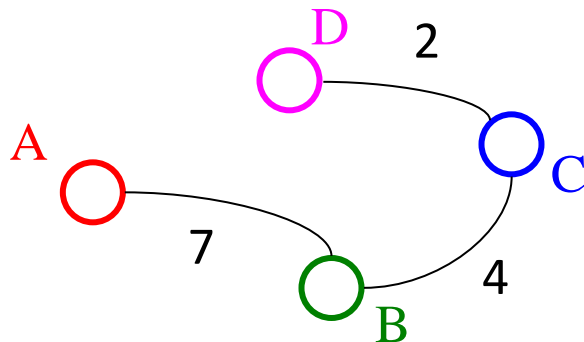
# Adjacency Matrix Properties

- 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
- Better for sparse or dense graphs?
  - Better for dense graphs

	A	B	C	D
A	F	T	F	F
B	T	F	F	F
C	F	T	F	T
D	F	F	F	F

# Adjacency Matrix Properties

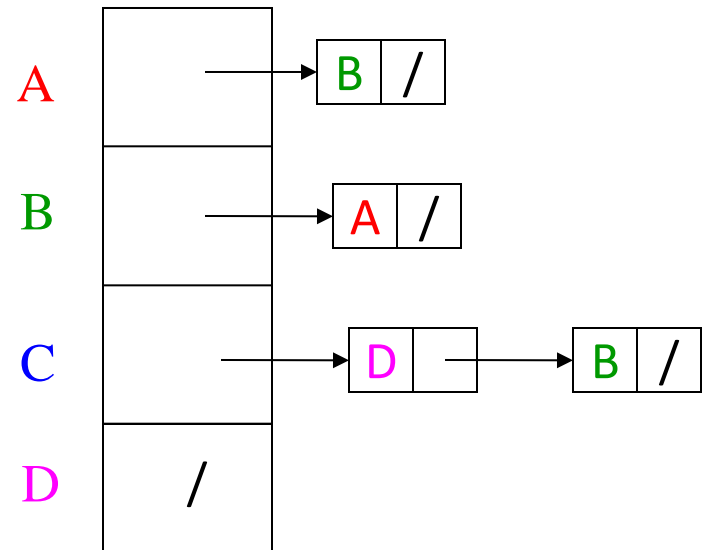
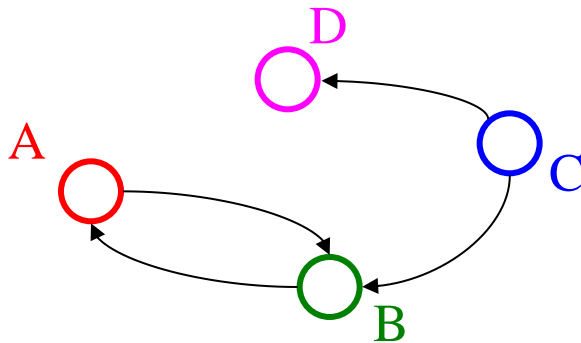
- How will the adjacency matrix vary for an *undirected graph*?
  - Undirected will be symmetric around the diagonal
- How can we adapt the representation for *weighted graphs*?
  - Instead of a Boolean, store a number in each cell
  - Need some value to represent 'not an edge'
    - In *some* situations,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



	A	B	C	D
A	-1	7	-1	-1
B	7	-1	4	-1
C	-1	4	-1	2
D	-1	-1	2	-1

# 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)



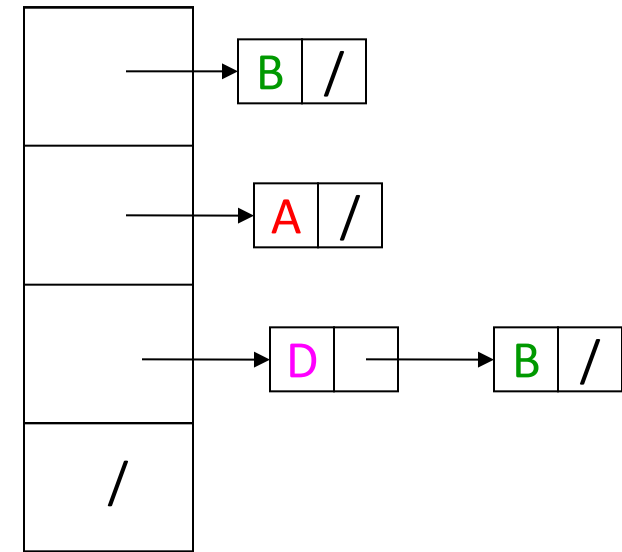
# Adjacency List Properties

A

B

C

D



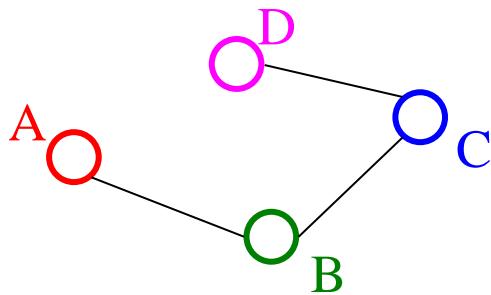
- Running time to:
  - Get all of a vertex's out-edges:  
 $O(d)$  where  $d$  is out-degree of vertex
  - Get all of a vertex's in-edges:  
 $O(|E| + |V|)$  (but could keep a second adjacency list for this!)
  - Decide if some edge exists:  
 $O(d)$  where  $d$  is out-degree of source
  - Insert an 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:
  - $O(|V| + |E|)$
- Better for dense or sparse graphs?
  - Better for sparse graphs

# Undirected Graphs

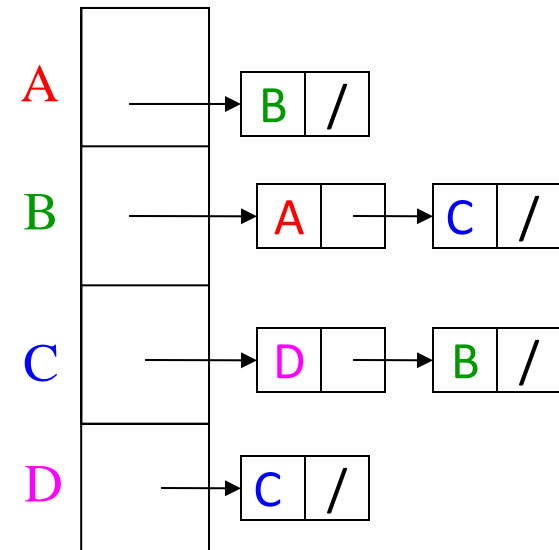
Adjacency matrices & adjacency lists both do fine for undirected graphs

- Matrix: Can save roughly 2x space
  - But may slow down operations in languages with “proper” 2D arrays (not Java, which has only arrays of arrays)
  - How would you “get all neighbors”?
- Lists: Each edge in two lists to support efficient “get all neighbors”

Example:



	A	B	C	D
A	F			
B	T	F		
C	F	T	F	
D	F	F	T	F



# Some Applications as Graphs

For each of the following examples:

- what are the **vertices** and what are the **edges**?
- would you use **directed edges**? Would they have **self-edges**?
- Are there **0-degree nodes**? Is it **strongly** or **weakly** connected?
- Does it have weights? Do negative weights make sense?
- Does it have cycles? Is it a DAG?

- 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
- Political donations to candidates