

---

# Graphs I

---

## AGENDA

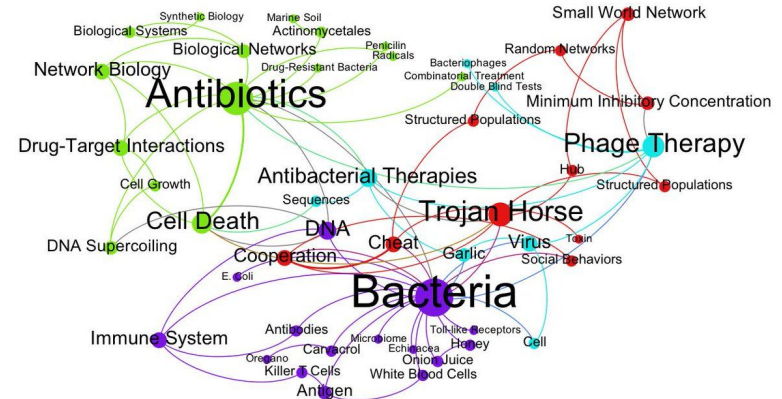
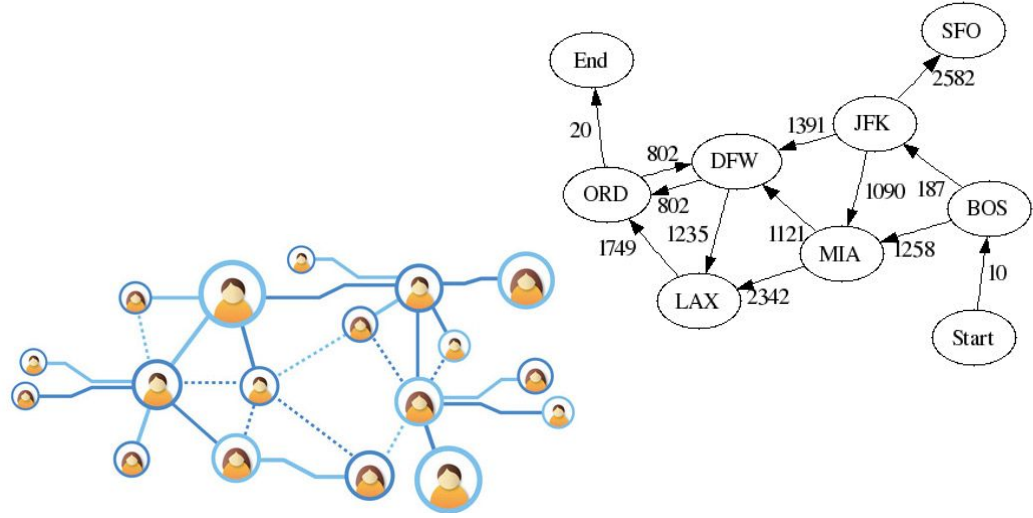
---

- Intro to Graphs
- Graph Properties
- Representing Graphs
- Graph Traversals

# Intro to Graphs

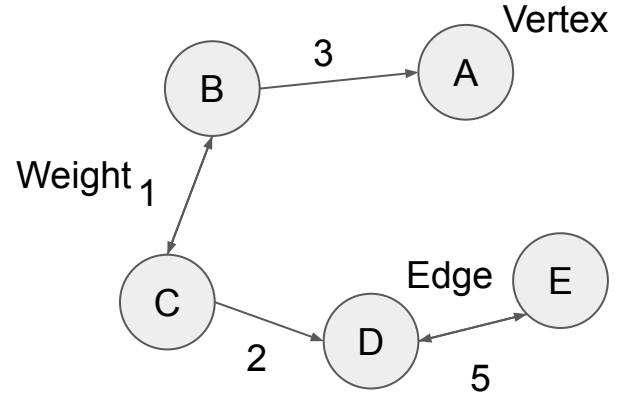
## INTRO TO GRAPHS

- A very versatile data structure that allows you to represent relationships between data
  - Social network, flight schedule, word relationships, etc.



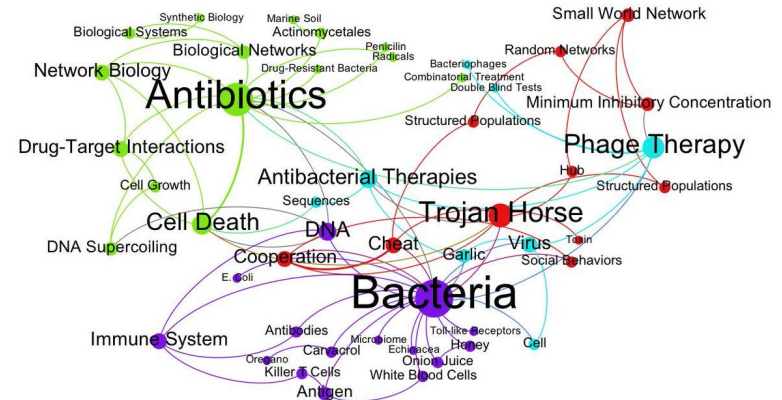
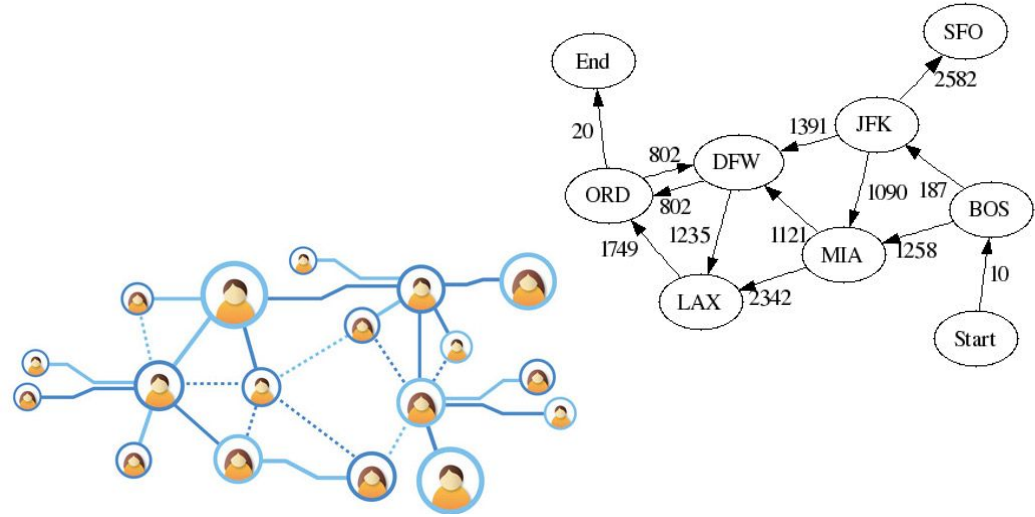
## COMPONENTS OF A GRAPH

- *Vertex* - also called nodes
- *Edge* - connects a pair of nodes
  - *Unidirectional* - Path from A to B doesn't mean there's a path from B to A
  - *Bidirectional* - A is friends with B, then B is friends with A
- *Weight* - used to represent a value associated with the edge (usually a cost)



## EXAMPLES OF GRAPHS

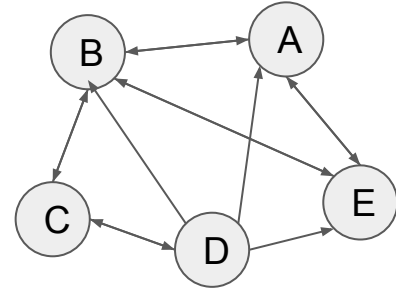
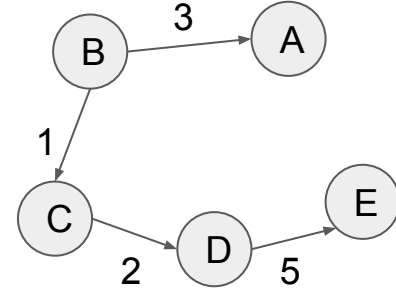
- Social Networks
  - Each node is a user, edges are friendships between nodes
  - Edges can also be to which groups you are a part of
- Transportation Systems (BART, Maps, etc.)
  - Each node is a location, each edge is a route to another one. A weight can represent time to get there
- The Internet!
  - Each page can be represented as a node, an directed edge is a link to another web page



# Graph Properties

## GRAPH PROPERTIES

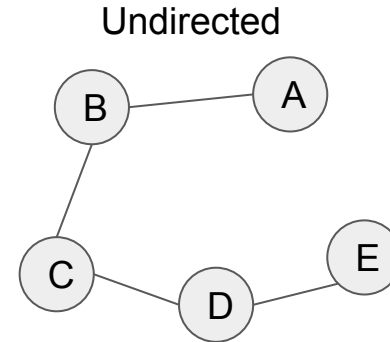
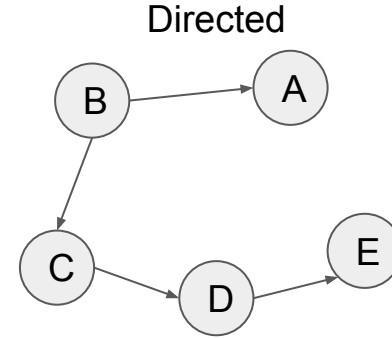
- A graph can have multiple properties
- Knowing these different properties are important so you can build/solve graph problems!





## DIRECTED VS. UNDIRECTED

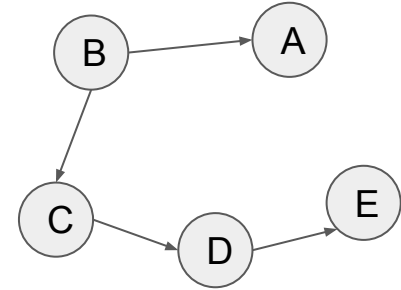
- A graph can be either *directed* or *undirected*
- *Directed* - An edge from A to B doesn't mean there's an edge from B to A
- *Undirected* - An edge from A to B means there's also an edge from B to A



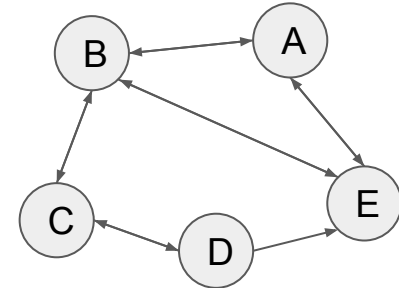
## CYCLIC VS. ACYCLIC

- Applies to directed graphs
- *cyclic* - there's at least one path from a node back to itself
- *acyclic* - there are no paths such that no node can be traversed back to itself

Acyclic Graph



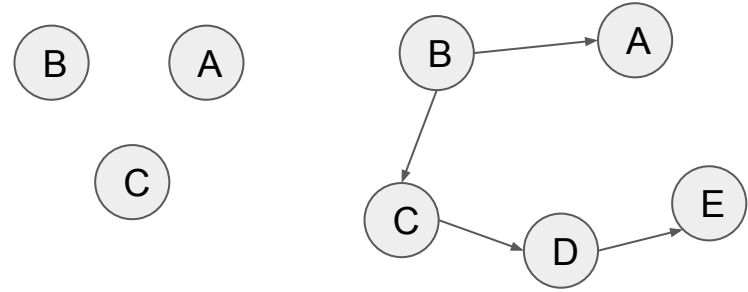
Cyclic Graph



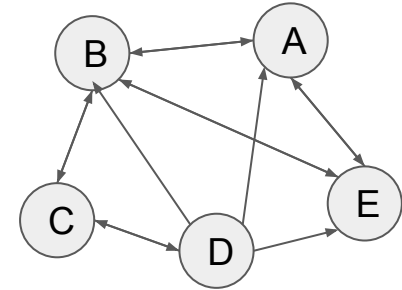
## DENSE VS. SPARSE

- A graph can be sparse/dense or anything in between
- *Dense* - contains close to the maximum edges possible
- *Sparse* - contains close to the minimum edges possible

Sparse

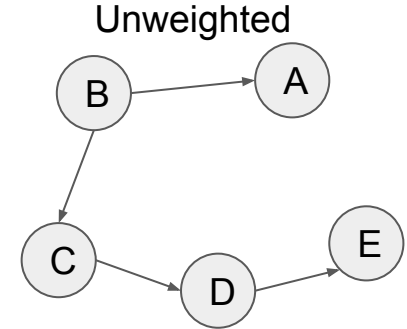
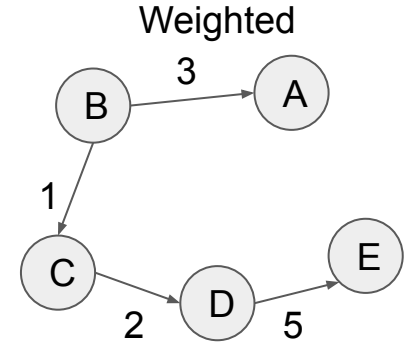


Dense



## WEIGHTED VS. UNWEIGHTED

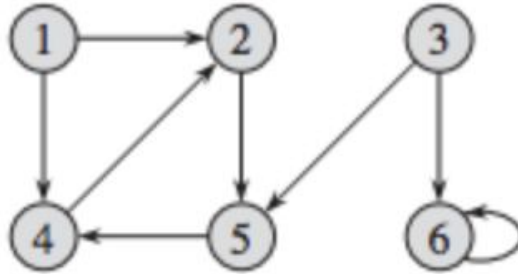
- A graph can either be weighted or unweighted
- Weight determines a value associated with an edge (usually a cost)
- *Weighted* - Each edge has an associated value
- *Unweighted* - Each edge has no associated value



# Representing Graphs

## ADJACENCY LIST

- Use a dictionary with sets to represent the edges of a particular vertex to other neighboring vertices
- $adjacencyList[i]$  is a set of all the edges to its neighbors for vertex  $i$



```
{  
  1: {2, 4},  
  2: {5},  
  3: {6, 5},  
  4: {2},  
  5: {4},  
  6: {6}  
}
```

# Adjacency List Demo

## ADJACENCY LIST RUNTIME/SPACE COMPLEXITIES

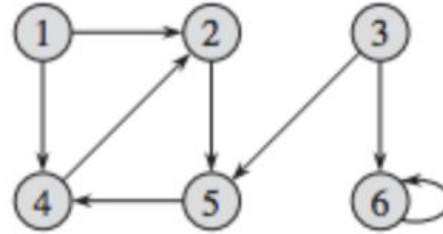
- Space:  $O(\text{vertices}^2)$ 
  - Imagine a dense graph
- Add vertex:  $O(1)$
- Remove vertex:  $O(\text{vertices})$
- Add edge:  $O(1)$
- Remove edge:  $O(1)$
- Find edge:  $O(1)$
- Get all edges:  $O(1)$

```
{  
  1: {2, 4},  
  2: {5},  
  3: {6, 5},  
  4: {2},  
  5: {4},  
  6: {6}  
}
```



## ADJACENCY MATRIX

- Use a matrix to represent whether or not there exists an edge between two vertices
- $matrix[i][j]$  is True if there exists an edge from vertex  $i$  to vertex  $j$

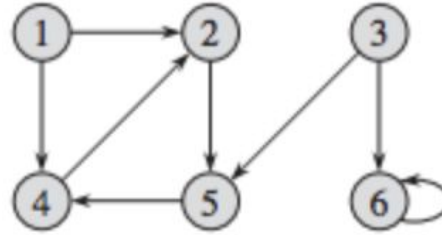


	1	2	3	4	5	6
1	0	1	0	1	0	0
2	0	0	0	0	1	0
3	0	0	0	0	1	1
4	0	1	0	0	0	0
5	0	0	0	1	0	0
6	0	0	0	0	0	1

# Adjacency Matrix Demo

## ADJACENCY MATRIX RUNTIME/SPACE COMPLEXITIES

- Space:  $O(\text{vertices}^2)$ 
  - Even in a sparse graph, but good for dense graphs b/c lists are space efficient
- Add vertex:  $O(\text{vertices}^2)$
- Remove vertex:  $O(\text{vertices}^2)$
- Add edge:  $O(1)$
- Remove edge:  $O(1)$
- Find edge:  $O(1)$
- Get all edges:  $O(\text{vertices})$



	1	2	3	4	5	6
1	0	1	0	1	0	0
2	0	0	0	0	1	0
3	0	0	0	0	1	1
4	0	1	0	0	0	0
5	0	0	0	1	0	0
6	0	0	0	0	0	1

## ADJACENCY MATRIX VS. ADJACENCY LISTS

	Space	Add Vertices	Remove Vertices	Add Edge	Remove Edge	Find Edge	Get All Edges
<b>Adj. Matrix</b>	$O(V^2)$	$O(V^2)$	$O(V^2)$	$O(1)$	$O(1)$	$O(1)$	$O(V)$
<b>Adj. List</b>	$O(V^2)$	$O(1)$	$O(1)$	$O(1)$	$O(1)$	$O(1)$	$O(1)$

	1	2	3	4	5	6
1	0	1	0	1	0	0
2	0	0	0	0	1	0
3	0	0	0	0	1	1
4	0	1	0	0	0	0
5	0	0	0	1	0	0
6	0	0	0	0	0	1

```
{  
  1: {2, 4},  
  2: {5},  
  3: {6, 5},  
  4: {2},  
  5: {4},  
  6: {6}  
}
```

## ADJACENCY MATRIX VS. ADJACENCY LISTS

- The best representation mainly depends on whether or not the graph is sparse/dense and what you're optimizing for (space/runtime)
- Representing dense graphs are probably better with adjacency matrix because lists are very space efficient in comparison to dictionaries/sets
- You'll probably deal with more adjacency lists

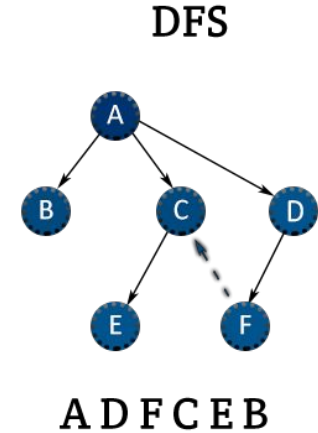
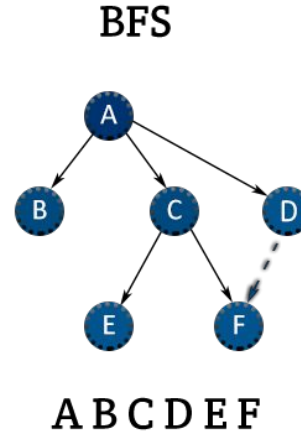
	1	2	3	4	5	6
1	0	1	0	1	0	0
2	0	0	0	0	1	0
3	0	0	0	0	1	1
4	0	1	0	0	0	0
5	0	0	0	1	0	0
6	0	0	0	0	0	1

```
{  
  1: {2, 4},  
  2: {5},  
  3: {6, 5},  
  4: {2},  
  5: {4},  
  6: {6}  
}
```

# Graph Traversals

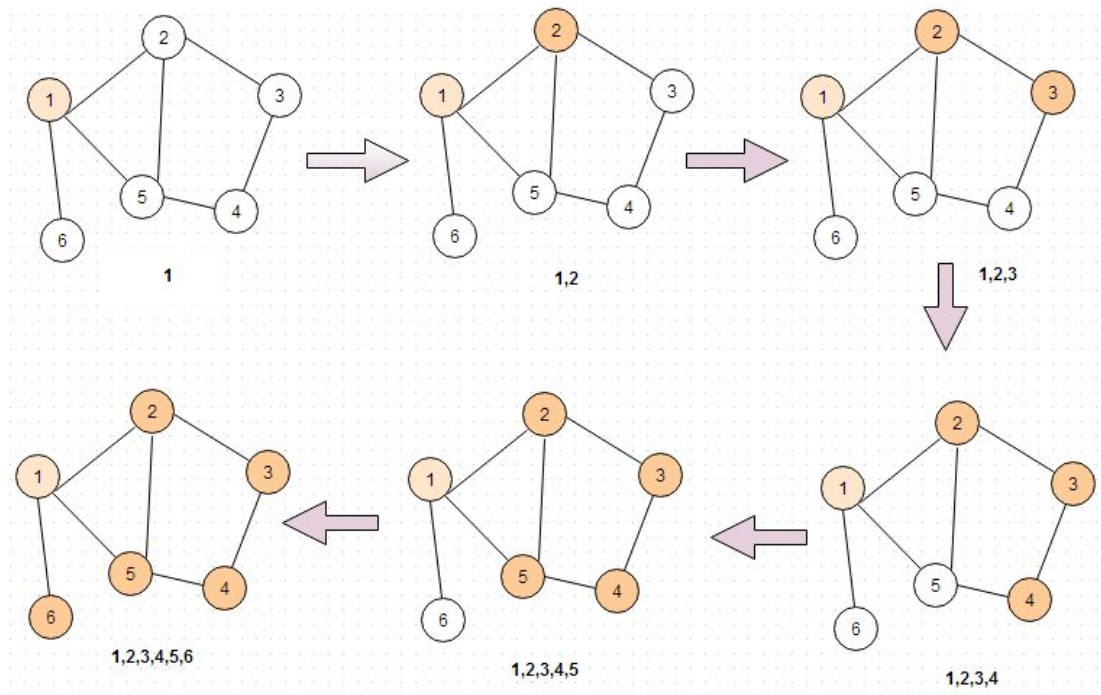
## GRAPH TRAVERSALS

- There are two primary ways to traverse a graph:  
*Depth-first* and *Breadth-first*
- Traversal vs. Search
  - In a search, you stop once you find the node you're searching for
  - In a traversal, you traverse the entire graph



## DEPTH-FIRST TRAVERSAL

- Traverse the graph in a depth-ward motion using a stack/recursion





## DEPTH-FIRST TRAVERSAL ITERATIVE PSEUDOCODE

```
procedure DFS_iterative( $G, v$ ) is  
    let  $S$  be a stack  
     $S.push(v)$   
    while  $S$  is not empty do  
         $v = S.pop()$   
        if  $v$  is not labeled as discovered then  
            label  $v$  as discovered  
            for all edges from  $v$  to  $w$  in  $G.adjacentEdges(v)$  do  
                 $S.push(w)$ 
```

*Note: The function name should be DFT\_iterative*

## DEPTH-FIRST TRAVERSAL RECURSIVE PSEUDOCODE

```
procedure DFS(G, v) is  
    label v as discovered  
    for all directed edges from v to w that are in G.adjacentEdges(v) do  
        if vertex w is not labeled as discovered then  
            recursively call DFS(G, w)
```

*Note: The function name should be DFT\_recursive*

# Depth-First Demo

## BREADTH-FIRST

---

- Traverse the graph in a breadth-ward motion using a queue
- Very useful for finding **shortest path** from node to node

## BREADTH-FIRST SEARCH PSEUDOCODE

```
procedure BFS(G, root) is  
    let Q be a queue  
    label root as discovered  
    Q.enqueue(root)  
    while Q is not empty do  
        v := Q.dequeue()  
        if v is the goal then  
            return v  
        for all edges from v to w in G.adjacentEdges(v) do  
            if w is not labeled as discovered then  
                label w as discovered  
                w.parent := v  
                Q.enqueue(w)
```

# Breadth-First Demo