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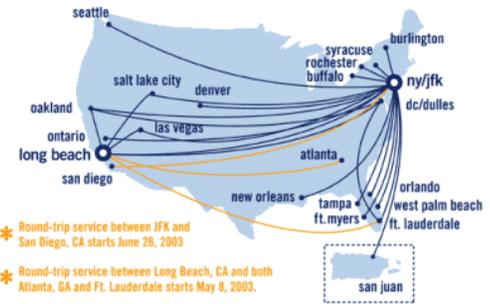
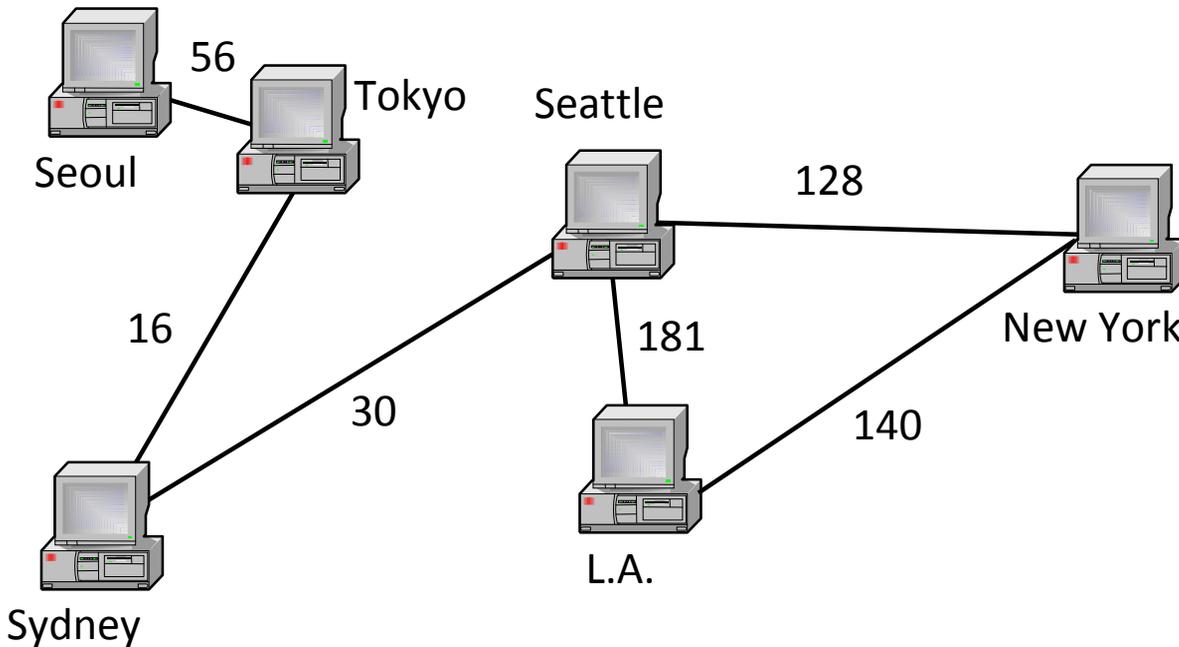
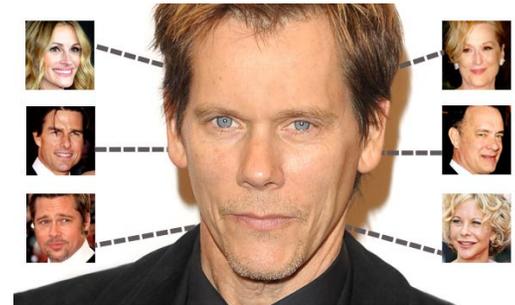
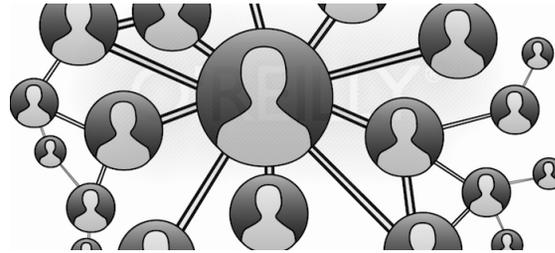
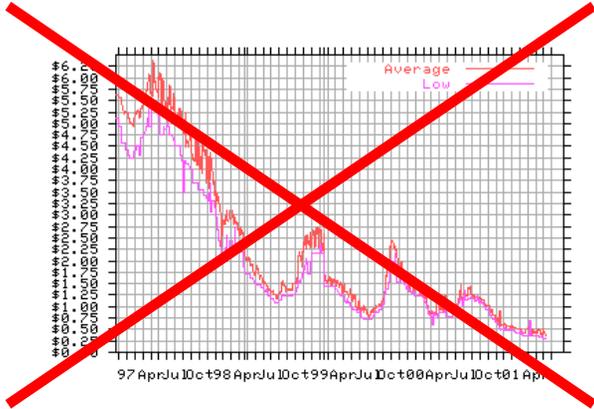
# CSE 373

Graphs 1: Concepts,  
Depth/Breadth-First Search  
reading: Weiss Ch. 9

slides created by Marty Stepp  
<http://www.cs.washington.edu/373/>

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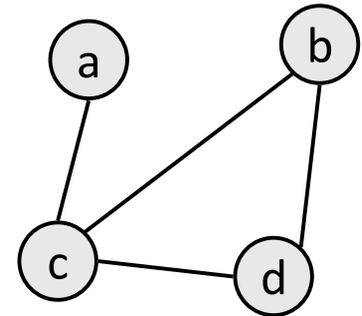
# What is a graph?



# Graphs

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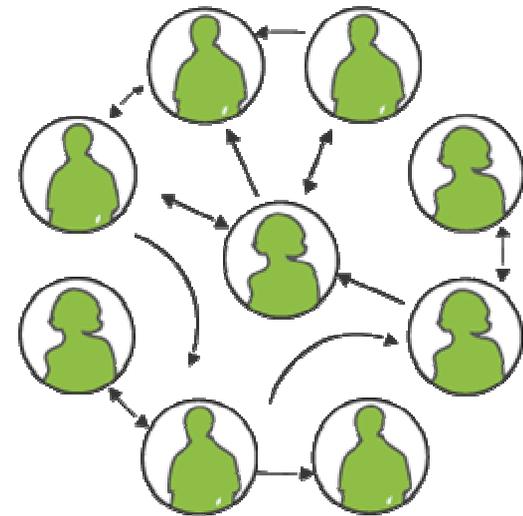
- **graph**: A data structure containing:
  - a set of **vertices**  $V$ , (*sometimes called nodes*)
  - a set of **edges**  $E$ , where an edge represents a connection between 2 vertices.
    - Graph  $G = (V, E)$
    - an edge is a pair  $(v, w)$  where  $v, w$  are in  $V$
- the graph at right:
  - $V = \{a, b, c, d\}$
  - $E = \{(a, c), (b, c), (b, d), (c, d)\}$
- **degree**: number of edges touching a given vertex.
  - at right:  $a=1, b=2, c=3, d=2$



# Graph examples

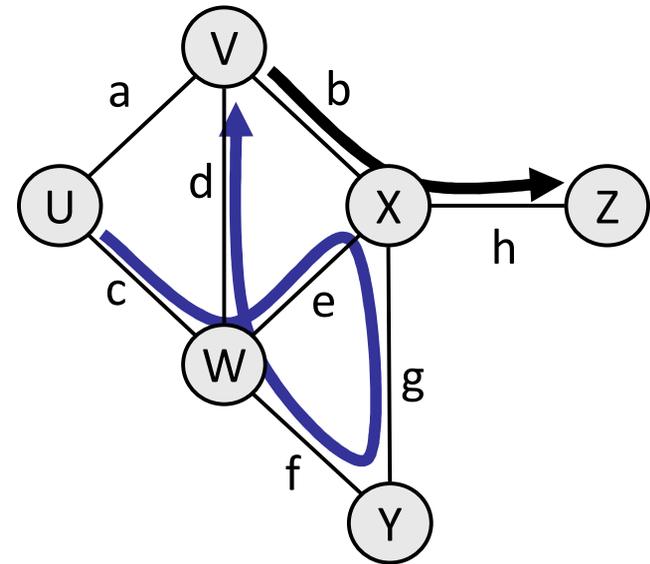
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- For each, what are the vertices and what are the edges?
  - Web pages with links
  - Methods in a program that call each other
  - Road maps (e.g., Google maps)
  - Airline routes
  - Facebook friends
  - Course pre-requisites
  - Family trees
  - Paths through a maze



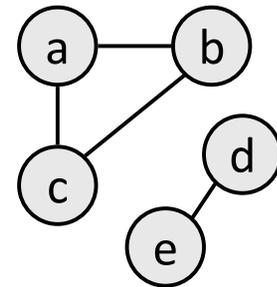
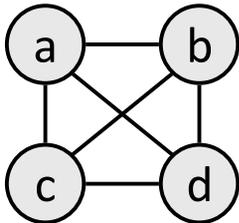
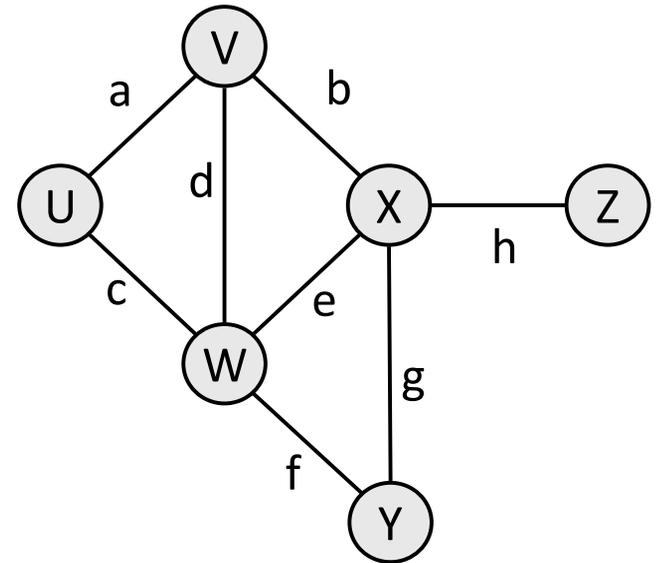
# Paths

- **path:** A path from vertex  $a$  to  $b$  is a sequence of edges that can be followed starting from  $a$  to reach  $b$ .
  - can be represented as vertices visited, or edges taken
  - example, one path from  $V$  to  $Z$ :  $\{b, h\}$  or  $\{V, X, Z\}$
  - What are two paths from  $U$  to  $Y$ ?
- **path length:** Number of vertices or edges contained in the path.
- **neighbor or adjacent:** Two vertices connected directly by an edge.
  - example:  $V$  and  $X$



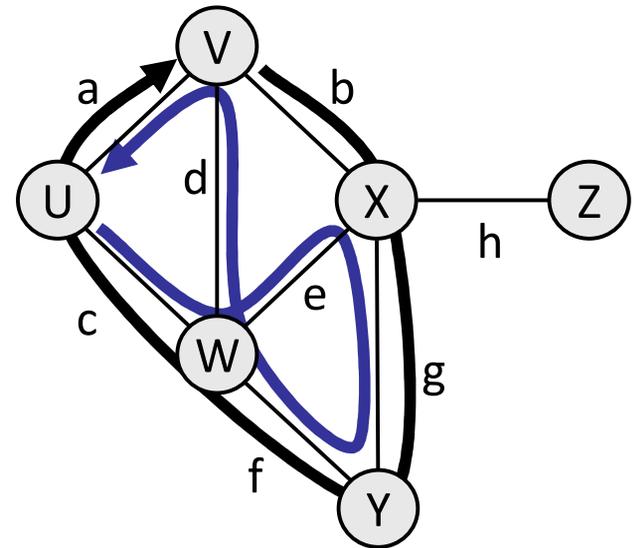
# Reachability, connectedness

- **reachable:** Vertex  $a$  is *reachable* from  $b$  if a path exists from  $a$  to  $b$ .
- **connected:** A graph is *connected* if every vertex is reachable from any other.
  - Is the graph at top right connected?
- **strongly connected:** When every vertex has an edge to every other vertex.



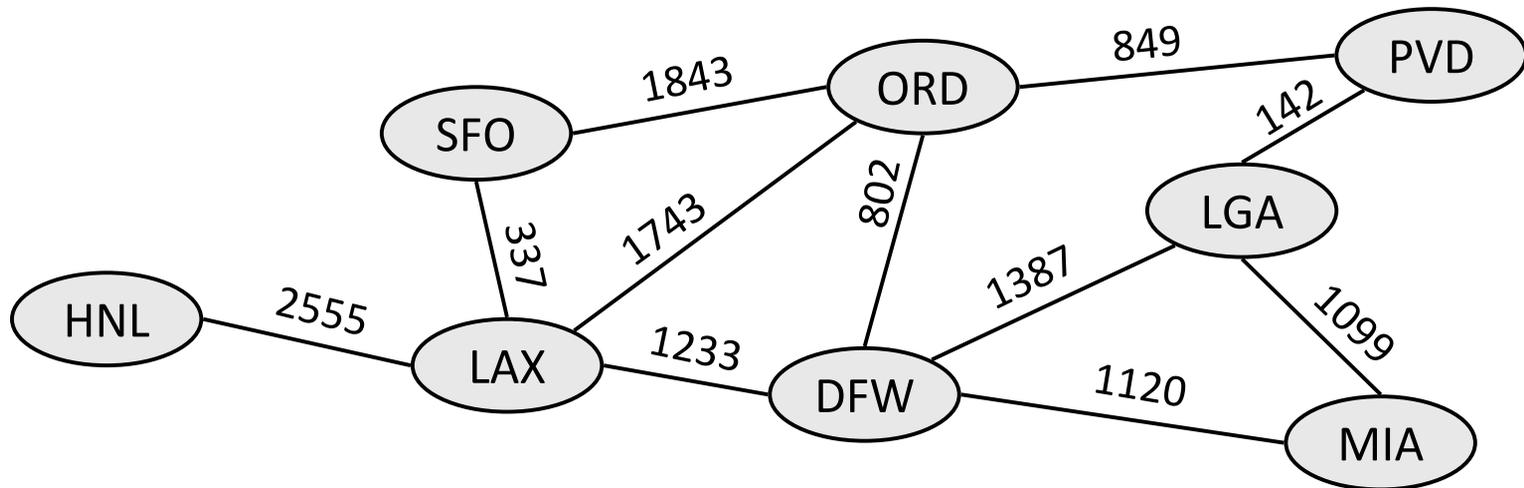
# Loops and cycles

- **cycle:** A path that begins and ends at the same node.
  - example: {b, g, f, c, a} or {V, X, Y, W, U, V}.
  - example: {c, d, a} or {U, W, V, U}.
  - **acyclic graph:** One that does not contain any cycles.
- **loop:** An edge directly from a node to itself.
  - Many graphs don't allow loops.



# Weighted graphs

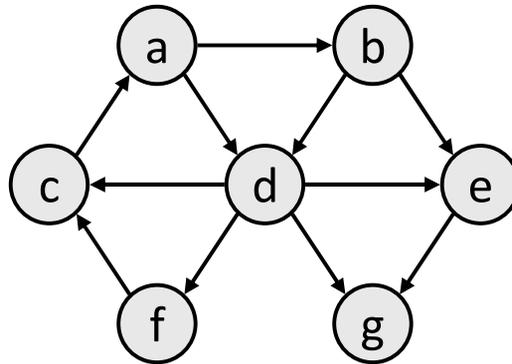
- **weight:** Cost associated with a given edge.
  - Some graphs have weighted edges, and some are unweighted.
  - Edges in an unweighted graph can be thought of as having equal weight (e.g. all 0, or all 1, etc.)
  - Most graphs do not allow negative weights.
- *example:* graph of airline flights, weighted by miles between cities:



# Directed graphs

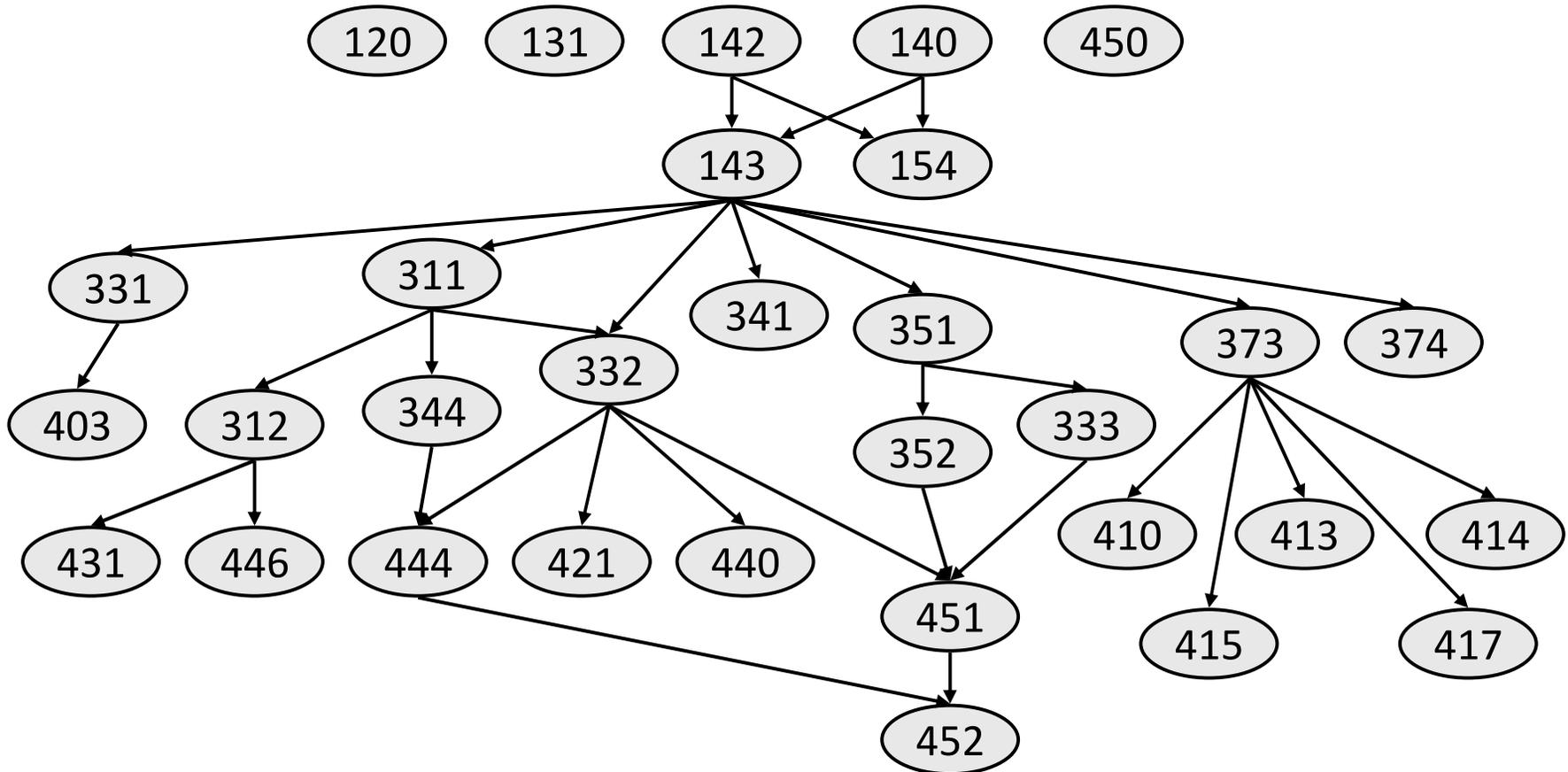
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- **directed graph** ("digraph"): One where edges are *one-way* connections between vertices.
  - If graph is directed, a vertex has a separate in/out degree.
  - A digraph can be weighted or unweighted.
  - Is the graph below connected? Why or why not?



# Digraph example

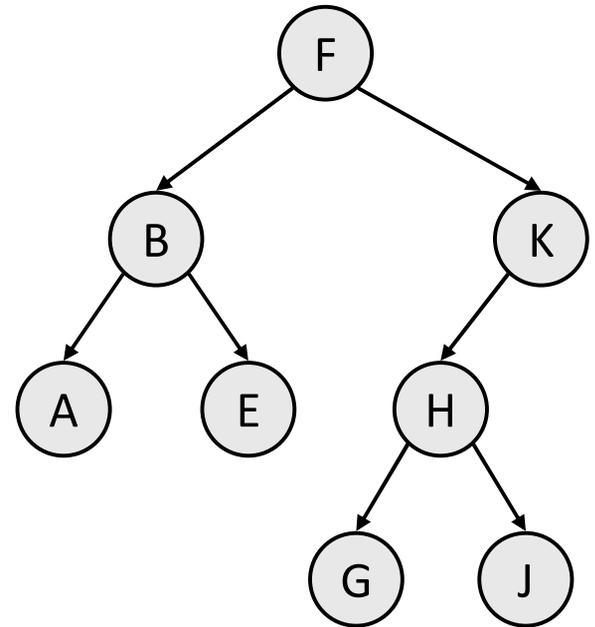
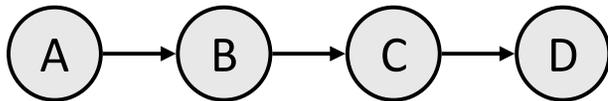
- Vertices = UW CSE courses (incomplete list)
- Edge (a, b) = a is a prerequisite for b



# Linked Lists, Trees, Graphs

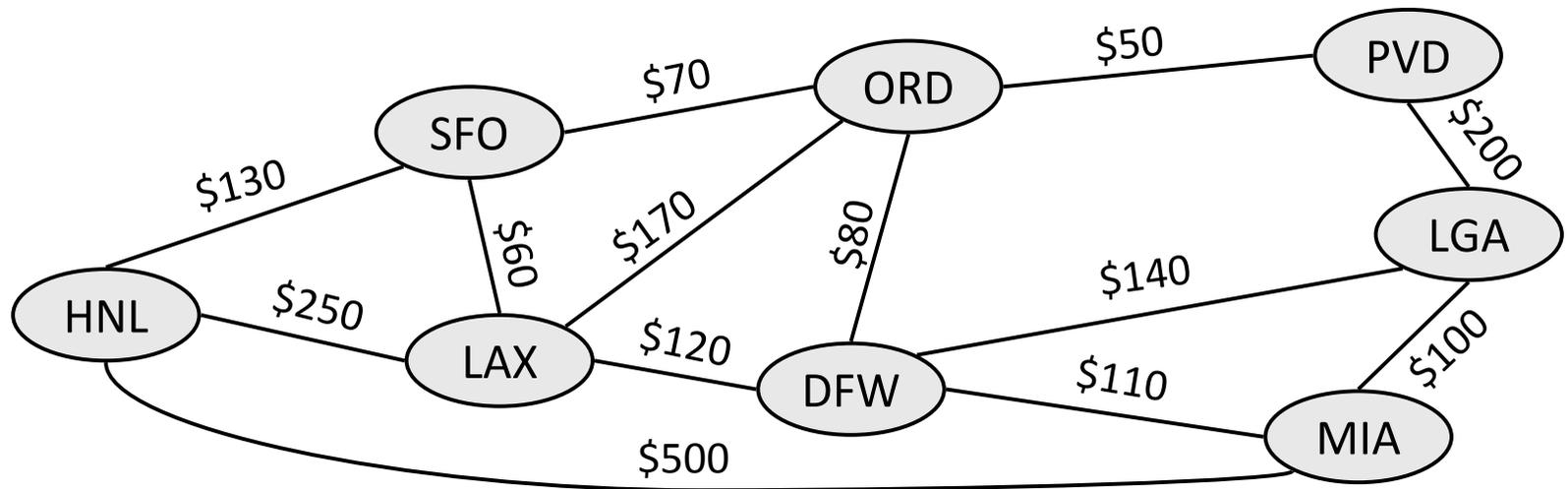
- A *binary tree* is a graph with some restrictions:
  - The tree is an unweighted, directed, acyclic graph (DAG).
  - Each node's in-degree is at most 1, and out-degree is at most 2.
  - There is exactly one path from the root to every node.

- A *linked list* is also a graph:
  - Unweighted DAG.
  - In/out degree of at most 1 for all nodes.



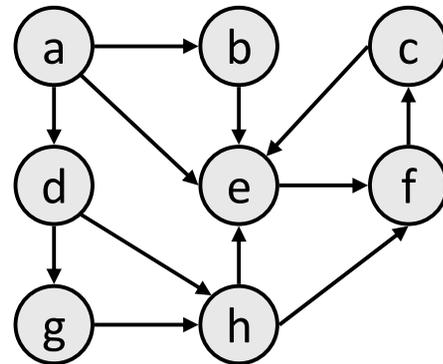
# Searching for paths

- Searching for a path from one vertex to another:
  - Sometimes, we just want *any* path (or want to know there *is* a path).
  - Sometimes, we want to minimize path *length* (# of edges).
  - Sometimes, we want to minimize path *cost* (sum of edge weights).
- What is the shortest path from MIA to SFO?  
Which path has the minimum cost?



# Depth-first search

- **depth-first search (DFS)**: Finds a path between two vertices by exploring each possible path as far as possible before backtracking.
  - Often implemented recursively.
  - Many graph algorithms involve *visiting* or *marking* vertices.
- Depth-first paths from *a* to all vertices (assuming ABC edge order):
  - to b: {a, b}
  - to c: {a, b, e, f, c}
  - to d: {a, d}
  - to e: {a, b, e}
  - to f: {a, b, e, f}
  - to g: {a, d, g}
  - to h: {a, d, g, h}



# DFS pseudocode

```
function dfs( $v_1, v_2$ ):  
  dfs( $v_1, v_2, \{ \}$ ).
```

```
function dfs( $v_1, v_2, path$ ):
```

```
   $path += v_1$ .
```

```
  mark  $v_1$  as visited.
```

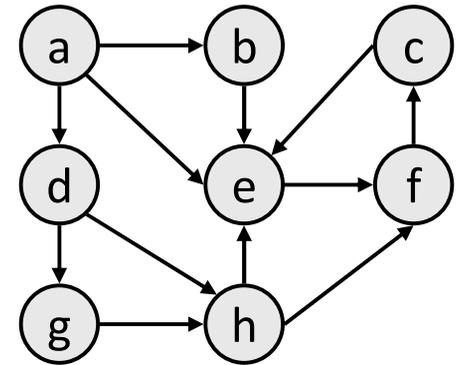
```
  if  $v_1$  is  $v_2$ :
```

```
    a path is found!
```

```
  for each unvisited neighbor  $n$  of  $v_1$ :
```

```
    if dfs( $n, v_2, path$ ) finds a path: a path is found!
```

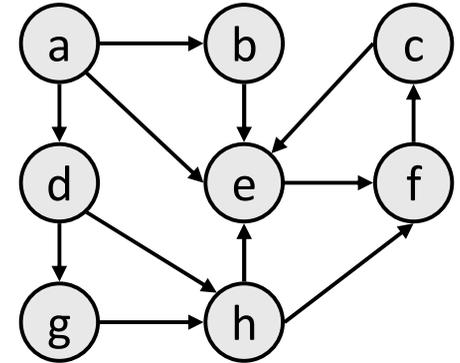
```
   $path -= v_1$ . // path is not found.
```



- The *path* param above is used if you want to have the path available as a list once you are done.
  - Trace dfs( $a, f$ ) in the above graph.

# DFS observations

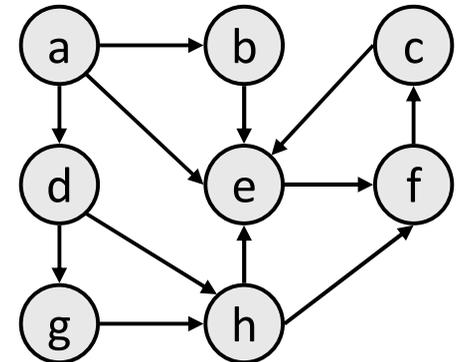
- *discovery*: DFS is guaranteed to find a path if one exists.
- *retrieval*: It is easy to retrieve exactly what the path is (the sequence of edges taken) if we find it
- *optimality*: not optimal. DFS is guaranteed to find a path, not necessarily the best/shortest path
  - Example:  $\text{dfs}(a, f)$  returns  $\{a, d, c, f\}$  rather than  $\{a, d, f\}$ .



# Breadth-first search

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- **breadth-first search (BFS)**: Finds a path between two nodes by taking one step down all paths and then immediately backtracking.
  - Often implemented by maintaining a queue of vertices to visit.
- BFS always returns the shortest path (the one with the fewest edges) between the start and the end vertices.
  - to b: {a, b}
  - to c: **{a, e, f, c}**
  - to d: {a, d}
  - to e: **{a, e}**
  - to f: **{a, e, f}**
  - to g: {a, d, g}
  - to h: **{a, d, h}**



# BFS pseudocode

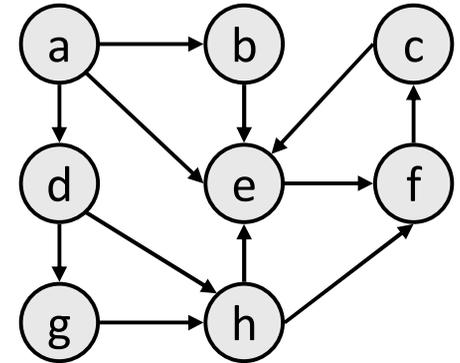
```
function bfs( $v_1, v_2$ ):  
   $queue := \{v_1\}$ .  
  mark  $v_1$  as visited.
```

```
  while  $queue$  is not empty:  
     $v := queue.removeFirst()$ .  
    if  $v$  is  $v_2$ :  
      a path is found!
```

```
    for each unvisited neighbor  $n$  of  $v$ :  
      mark  $n$  as visited.  
       $queue.addLast(n)$ .
```

```
  // path is not found.
```

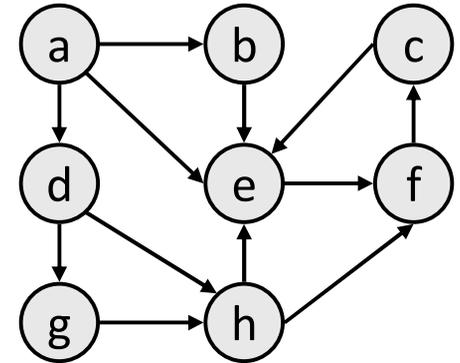
- Trace  $bfs(a, f)$  in the above graph.



# BFS observations

- *optimality*:

- always finds the shortest path (fewest edges).
- in unweighted graphs, finds optimal cost path.
- In weighted graphs, *not* always optimal cost.



- *retrieval*: harder to reconstruct the actual sequence of vertices or edges in the path once you find it

- conceptually, BFS is exploring many possible paths in parallel, so it's not easy to store a path array/list in progress
- solution: We can keep track of the path by storing predecessors for each vertex (each vertex can store a reference to a *previous* vertex).

- DFS uses less memory than BFS, easier to reconstruct the path once found; but DFS does not always find shortest path. BFS does.

# DFS, BFS runtime

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- What is the expected runtime of DFS and BFS, in terms of the number of vertices  $V$  and the number of edges  $E$  ?
- Answer:  $O(|V| + |E|)$ 
  - where  $|V|$  = number of vertices,  $|E|$  = number of edges
  - Must potentially visit every node and/or examine every edge once.
  - why not  $O(|V| * |E|)$  ?
- What is the space complexity of each algorithm?
  - (How much memory does each algorithm require?)

