

# CS4102: Graph Traversals

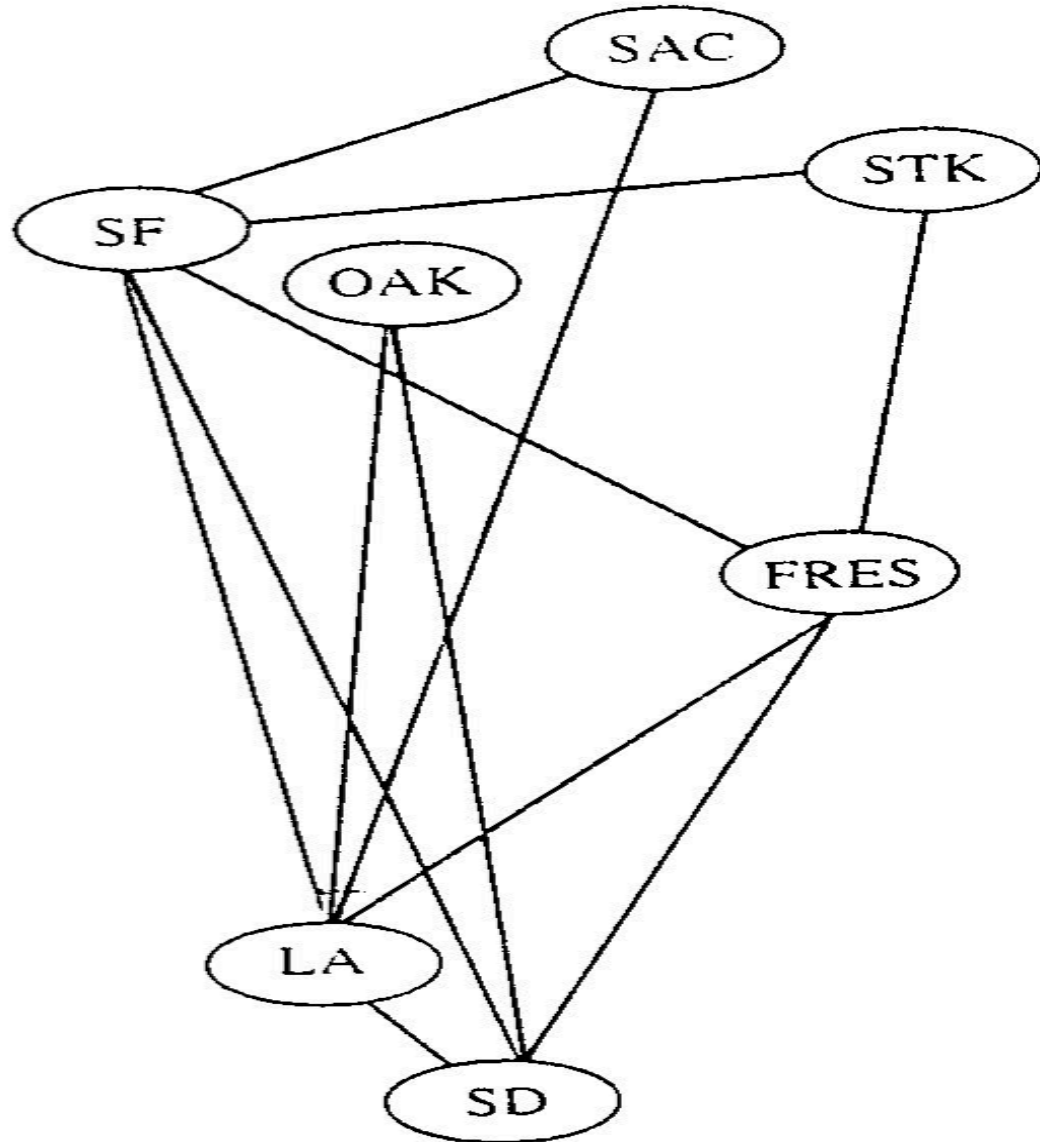
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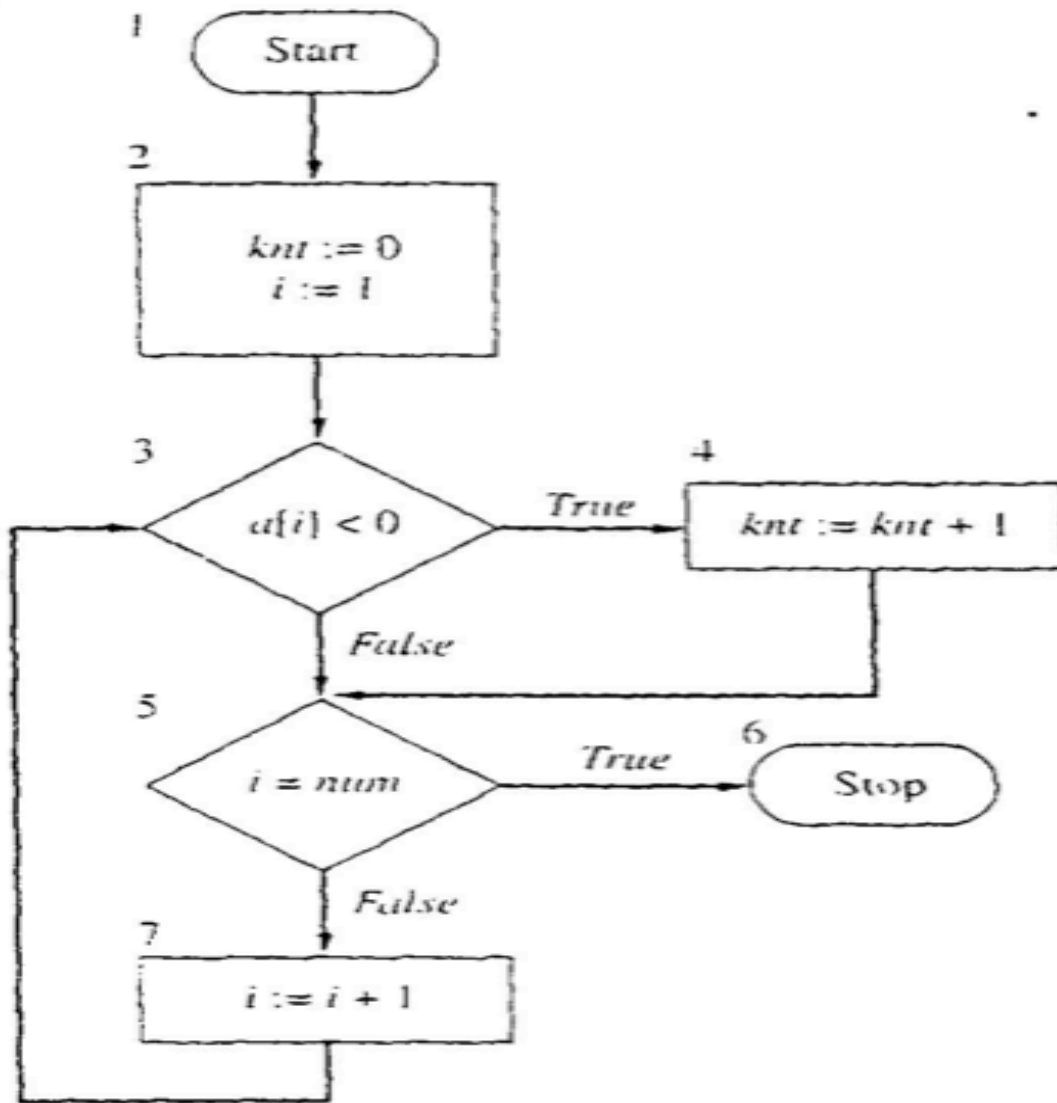
- Review: Section 2.5
  - Definitions, data structures
  - Note: review definitions, data structures, and BFS from CS216 slides from 4-16-03  
<http://www.cs.virginia.edu/~cs216/notes/slides/graphs2.pdf>
- Read: Chapter 4 (from 4.2 on)
  - Traversing Graphs
    - Depth-first Search (DFS)
    - Breadth-first Search (BFS)
  - Applications of DFS strategy (things not in text)
  - Backtracking, Exhaustive Search (handout)

# Problems: e.g. Airline Routes

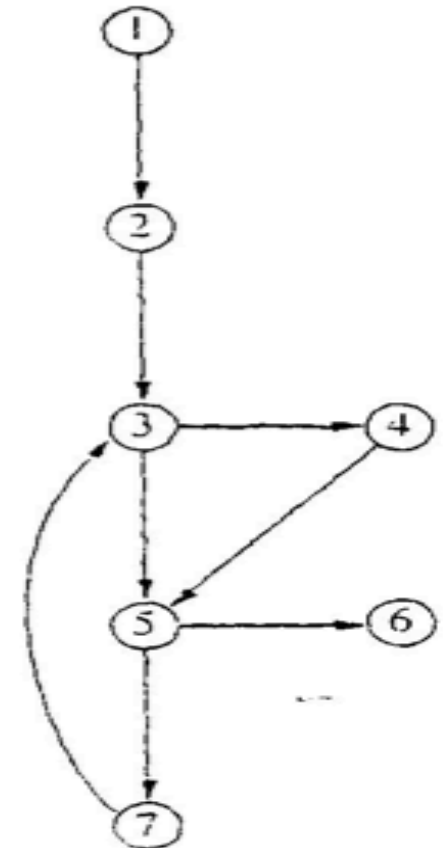
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# Problems: e.g. Flowcharts



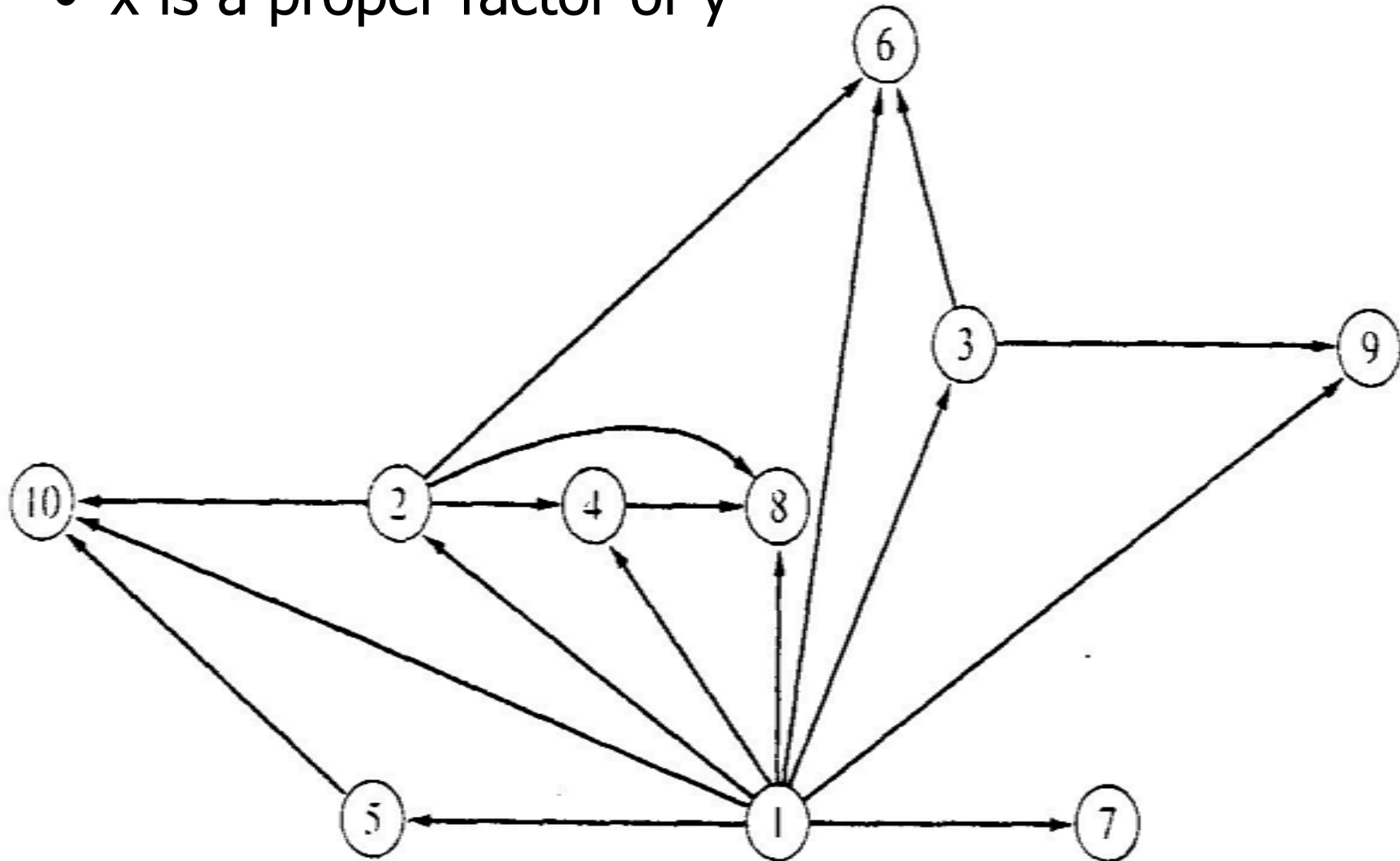
(a) Flowchart



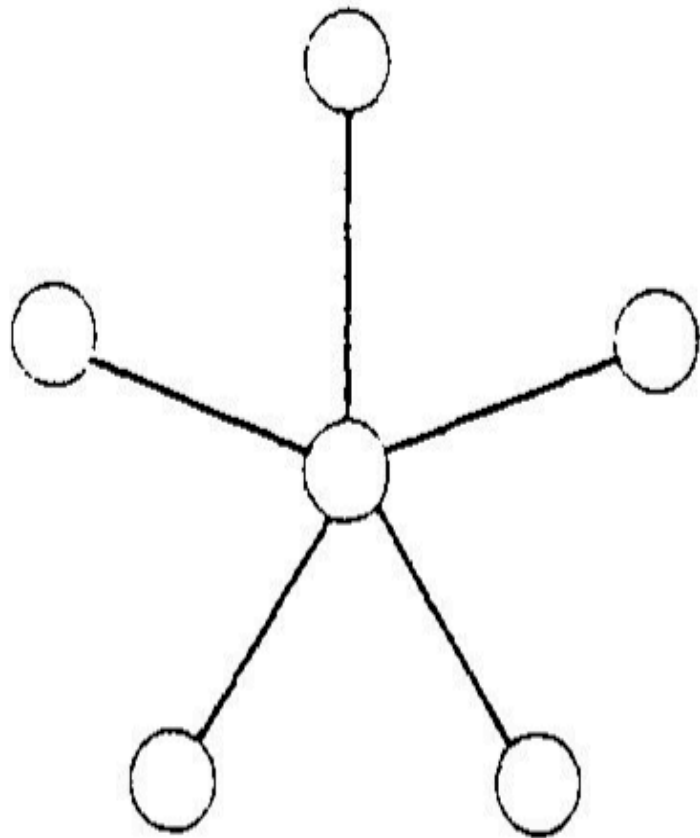
(b) Directed graph

# Problems: e.g. Binary relation

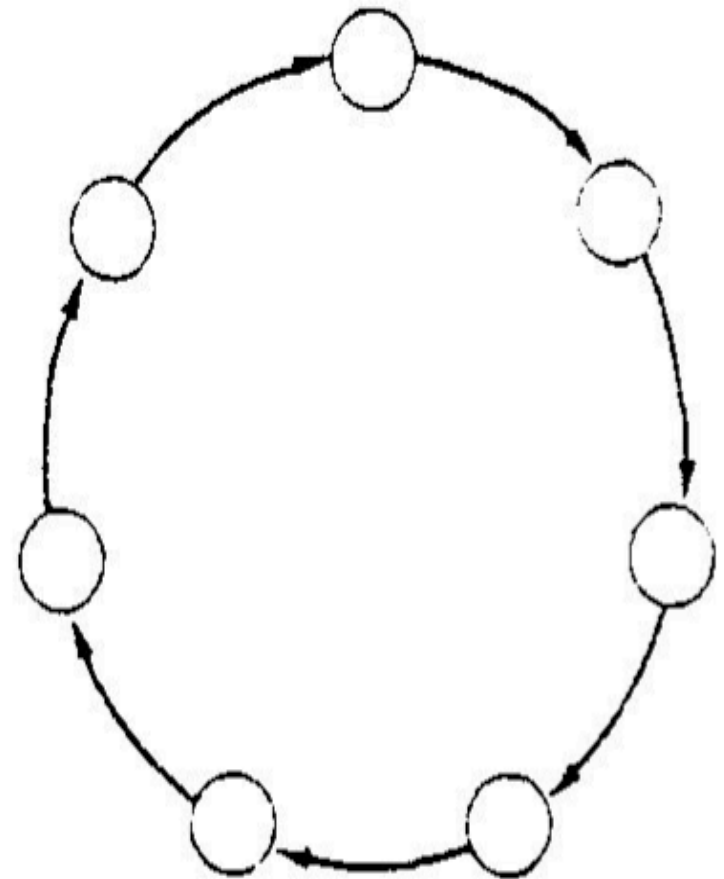
- $x$  is a proper factor of  $y$



# Problems: e.g. Computer Networks



(a) A star network



(b) A ring network

# Terms You Should Know or Learn Now

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- Vertex (plural *vertices*) or Node
- Edge (sometimes referred to as an *arc*)
  - Note the meaning of *incident*
- Degree of a vertex: how many adjacent vertices
  - Digraph: in-degree (num. of incoming edges) vs. out-degree
- Graphs can be:
  - Directed or undirected
  - Weighted or not weighted
    - weights can be reals, integers, etc.
    - weight also known as: cost, length, distance, capacity,...
- Undirected graphs:
  - Normally an edge can't connect a vertex to itself
- A directed graph (also known as a *digraph*)
  - "Originating" node is the *head*, the target the *tail*
  - An edge may connect a vertex to itself

# Terms You Should Know or Learn Now

- Size of graph? Two measures:
  - Number of nodes. Usually  $n$
  - Number of edges: usually  $m$
- Dense graph: many edges
  - Maximally dense?
  - Undirected: each node connects to all others, so  $m = n(n-1)/2$   
Called a *complete graph*
  - Directed:  $m = n(n-1)$       *why?*
- Sparse graph: fewer edges
  - Could be zero edges...

# **Terms You Should Know or Learn Now**

- Path vs. simple path
  - One vertex is *reachable* from another vertex
- *A connected graph*
  - undirected graph, where each vertex is reachable from all others
- *A strongly connected digraph:*
  - direction affects this!
  - node  $u$  may be reachable from  $v$ , but not  $v$  from  $u$
  - Strongly connected means both directions
- Connected components for undirected graphs



# Terms You Should Know or Learn Now

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- Cycle
  - Directed graph: non-empty path with same starting and ending node
  - An edge may appear more than once (but why?)
    - **Simple cycle**: no node repeated except start and end
  - Undirected graph: same idea
    - If an edge appears more than once (I.e. non-simple) then we traverse it in the same direction
- Acyclic: no-cycles
- A connected, acyclic undirected graph: *free tree*
  - If we specify a root, it's a *rooted tree*
  - Acyclic but not connected? a undirected *forest*
- Directed acyclic graph: a DAG

# **Self-test: Understand these Terms?**

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- Subgraph
- Symmetric digraph
- complete graph
- Adjacency relation
- Path, simple path, reachable
- Connected, Strongly Connected
- Cycle, simple cycle
- acyclic
- undirected forest
- free tree, undirected tree
- rooted tree
- Connected component



# Definition: Directed graph

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- Directed Graph
  - A directed graph, or digraph, is a pair
  - $G = (V, E)$
  - where  $V$  is a set whose elements are called vertices, and
  - $E$  is a set of ordered pairs of elements of  $V$ .
    - Vertices are often also called nodes.
    - Elements of  $E$  are called edges, or directed edges, or arcs.
    - For directed edge  $(v, w)$  in  $E$ ,  $v$  is its tail and  $w$  its head;
    - $(v, w)$  is represented in the diagrams as the arrow,  $v \rightarrow w$ .
    - In text we simple write  $vw$ .

# Definition: Undirected graph

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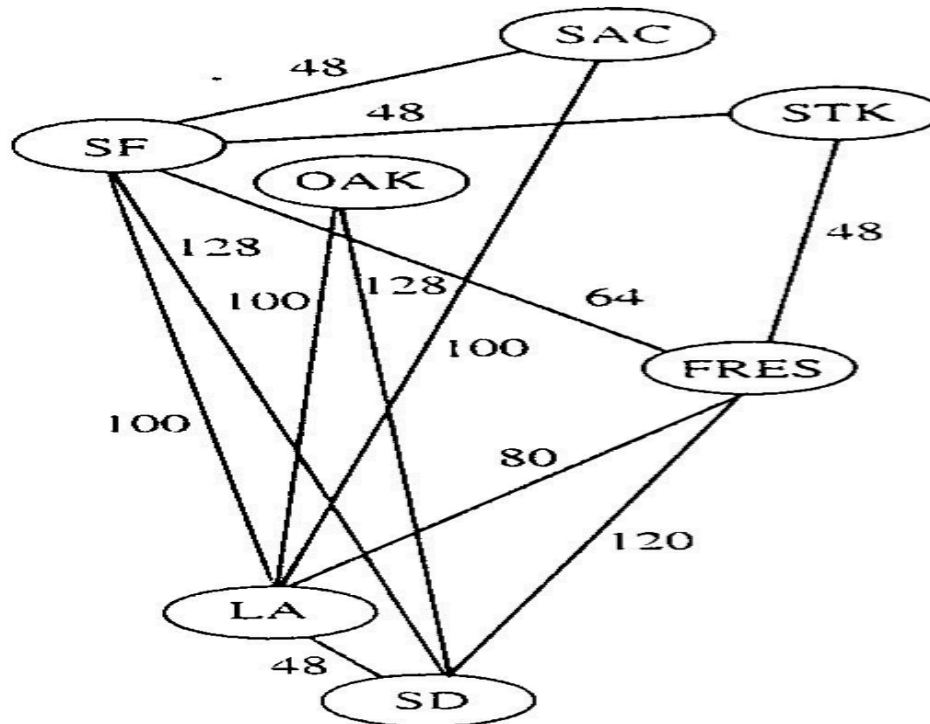
- Undirected Graph
  - A undirected graph is a pair
  - $G = (V, E)$
  - where  $V$  is a set whose elements are called vertices, and
  - $E$  is a set of *unordered* pairs of *distinct* elements of  $V$ .
    - Vertices are often also called nodes.
    - Elements of  $E$  are called edges, or undirected edges.
    - Each edge may be considered as a subset of  $V$  containing two elements,
    - $\{v, w\}$  denotes an undirected edge
    - In diagrams this edge is the line  $v\text{---}w$ .
    - In text we simple write  $vw$ , or  $wv$
    - $vw$  is said to be *incident* upon the vertices  $v$  and  $w$

# Definitions: Weighted Graph

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- A weighted graph is a triple  $(V, E, W)$ 
  - where  $(V, E)$  is a graph (directed or undirected) and
  - $W$  is a function from  $E$  into  $\mathbb{R}$ , the reals (integer or rationals).
  - For an edge  $e$ ,  $W(e)$  is called the weight of  $e$ .

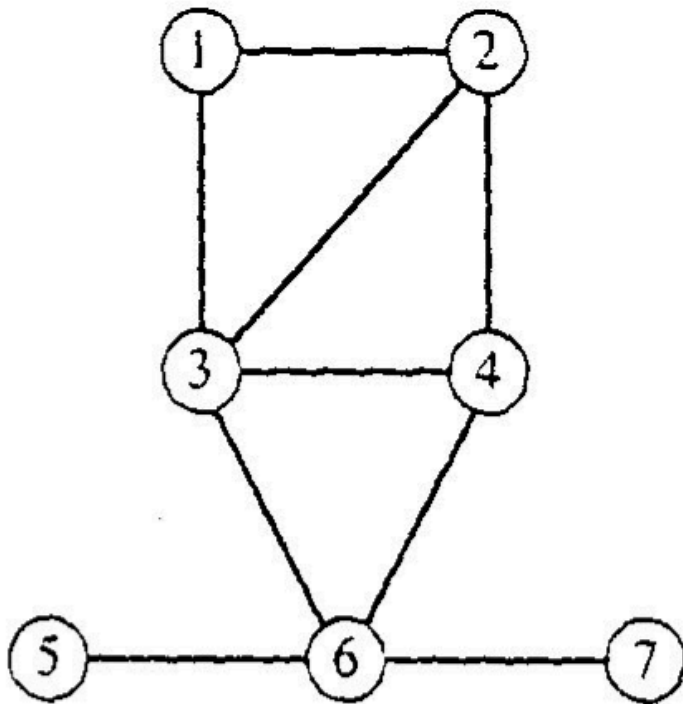


# Graph Representations using Data Structures

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- Adjacency Matrix Representation
  - Let  $G = (V, E)$ ,  $n = |V|$ ,  $m = |E|$ ,  $V = \{v_1, v_2, \dots, v_n\}$
  - $G$  can be represented by an  $n \times n$  matrix



(a) An undirected graph

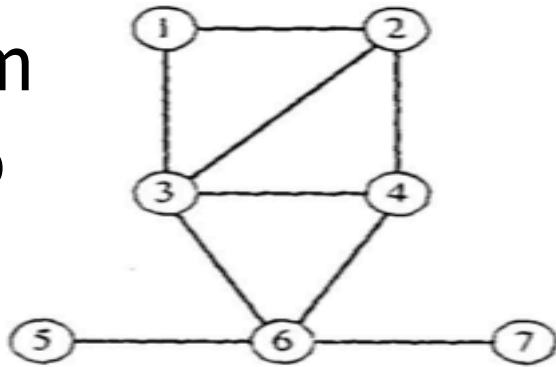
The adjacency matrix is a 7x7 matrix with a diagonal line from the top-left to the bottom-right. The entries are as follows:

0	1	1	0	0	0	0
1	0	1	1	0	0	0
1	1	0	1	0	1	0
0	1	1	0	0	1	0
0	0	0	0	0	1	0
0	0	1	1	1	0	1
0	0	0	0	0	1	0

(b) Its adjacency matrix

# Array of Adjacency Lists Representation

- From
- to

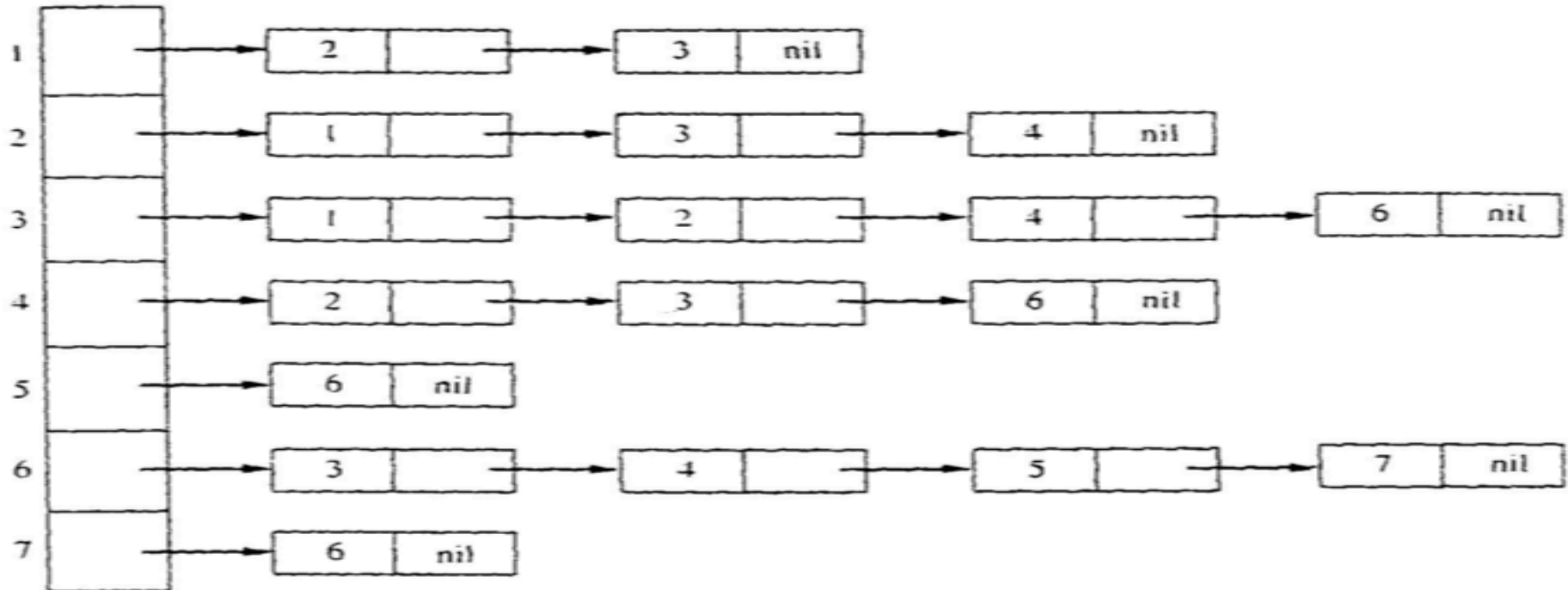


(a) An undirected graph

$$\begin{pmatrix} 0 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 \end{pmatrix}$$

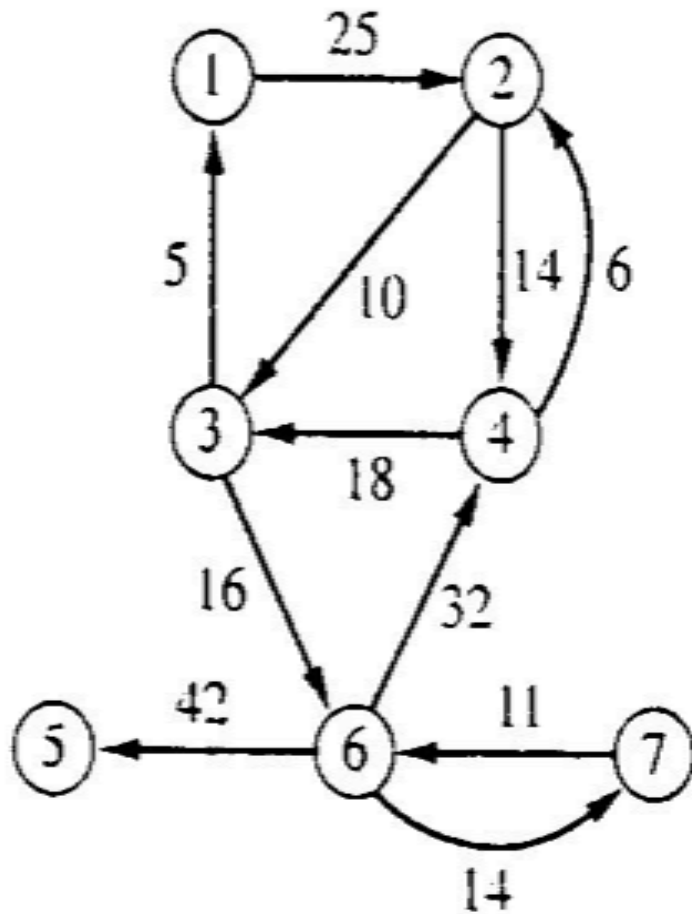
(b) Its adjacency matrix

adjVertices





# Adjacency Matrix for weight digraph

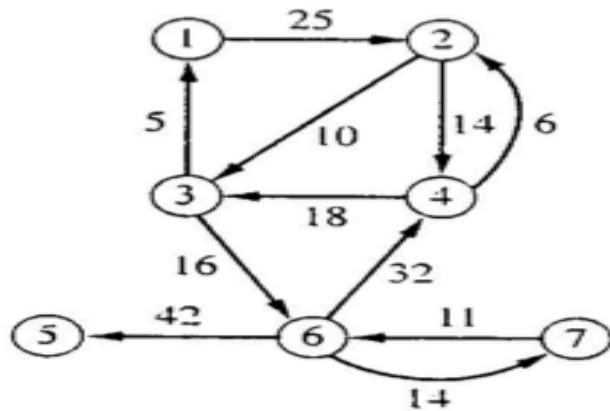


(a) A weighted digraph

$$\begin{pmatrix} 0 & 25.0 & \infty & \infty & \infty & \infty & \infty \\ \infty & 0 & 10.0 & 14.0 & \infty & \infty & \infty \\ 5.0 & \infty & 0 & \infty & \infty & 16.0 & \infty \\ \infty & 6.0 & 18.0 & 0 & \infty & \infty & \infty \\ \infty & \infty & \infty & \infty & 0 & \infty & \infty \\ \infty & \infty & \infty & 32.0 & 42.0 & 0 & 14.0 \\ \infty & \infty & \infty & \infty & \infty & 11.0 & 0 \end{pmatrix}$$

(b) Its adjacency matrix

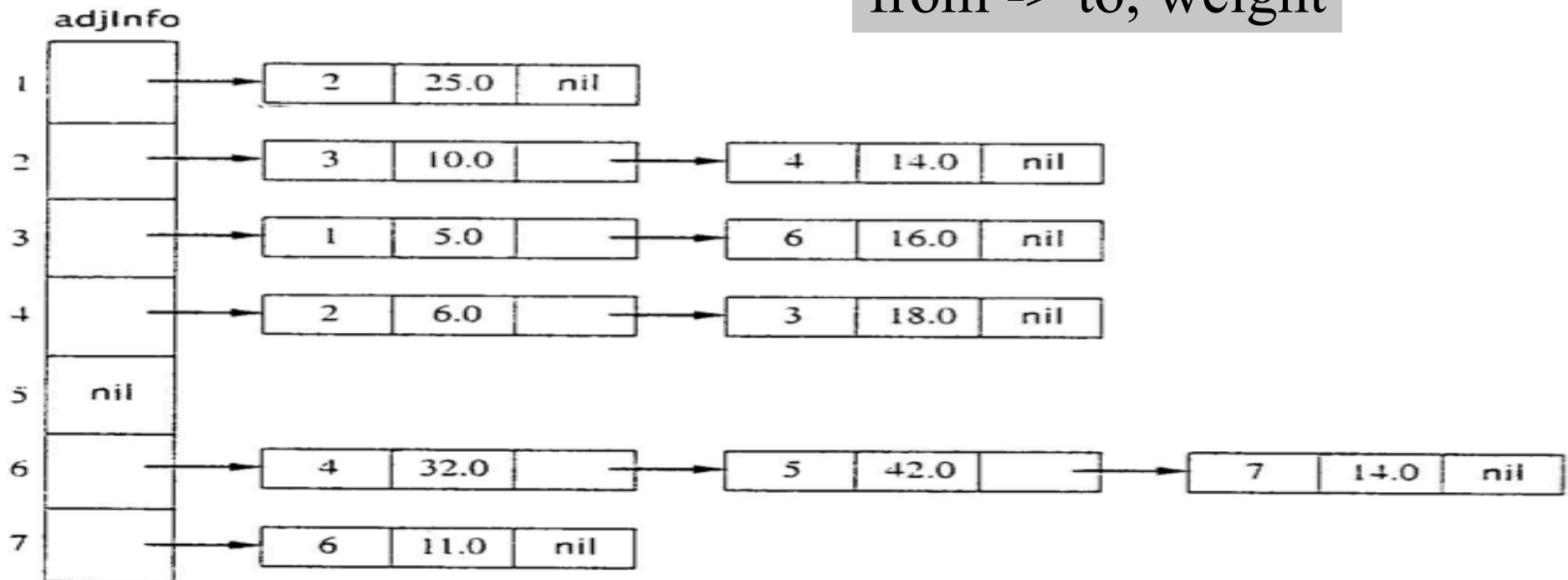
# Array of Adjacency Lists Representation



(a) A weighted digraph

$$\begin{pmatrix}
 0 & 25.0 & \infty & \infty & \infty & \infty & \infty \\
 \infty & 0 & 10.0 & 14.0 & \infty & \infty & \infty \\
 5.0 & \infty & 0 & \infty & \infty & 16.0 & \infty \\
 \infty & 6.0 & 18.0 & 0 & \infty & \infty & \infty \\
 \infty & \infty & \infty & \infty & 0 & \infty & \infty \\
 \infty & \infty & \infty & 32.0 & 42.0 & 0 & 14.0 \\
 \infty & \infty & \infty & \infty & \infty & 11.0 & 0
 \end{pmatrix}$$

from -> to, weight



# Traversing Graphs

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- “Traversing” means processing each vertex edge in some organized fashion by following edges between vertices
  - We speak of *visiting* a vertex. Might do something while there.
- Recall traversal of binary trees:
  - Several strategies: In-order, pre-order, post-order
  - Traversal strategy implies an order of visits
  - We used recursion to describe and implement these
- Graphs can be used to model interesting, complex relationships
  - Often traversal used just to process the set of vertices or edges
  - Sometimes traversal can identify interesting properties of the graph
  - Sometimes traversal (perhaps modified, enhanced) can answer interesting questions about the problem-instance that the graph models

# Traversal Strategies

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- Note: traversal algorithms start at some vertex
  - Which? Trees have a root, but graphs don't.
  - Might matter, might not.
- Breadth-first search and depth-first search
  - efficient way to "visit" each vertex and edge exactly once.
- Later we'll see exhaustive search
  - Can visit vertices and edges more than once
  - Exhaustively finds... (wait and see!)
- We'll see that BFS will tell us something about distances between a vertex and other vertices
- We'll see that DFS will be a generally useful approach for solving many graph problems.

# BFS Strategy

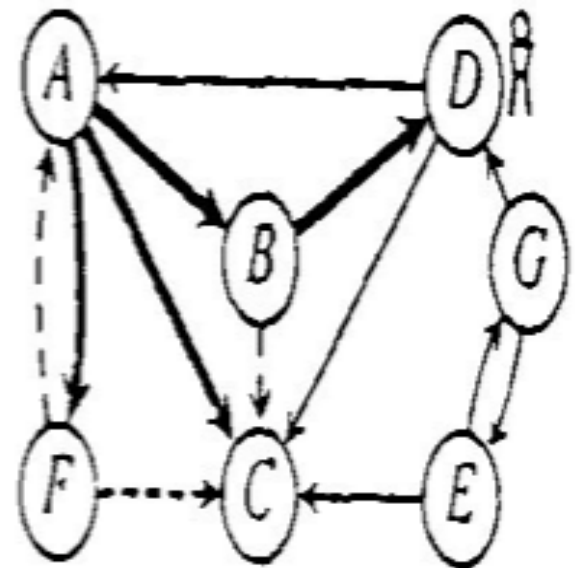
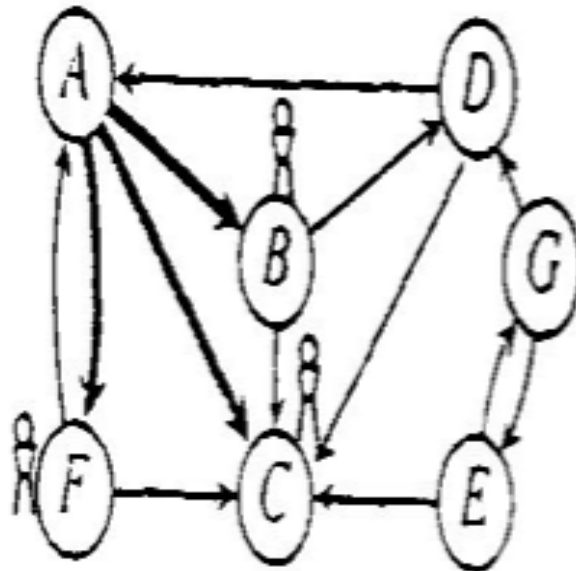
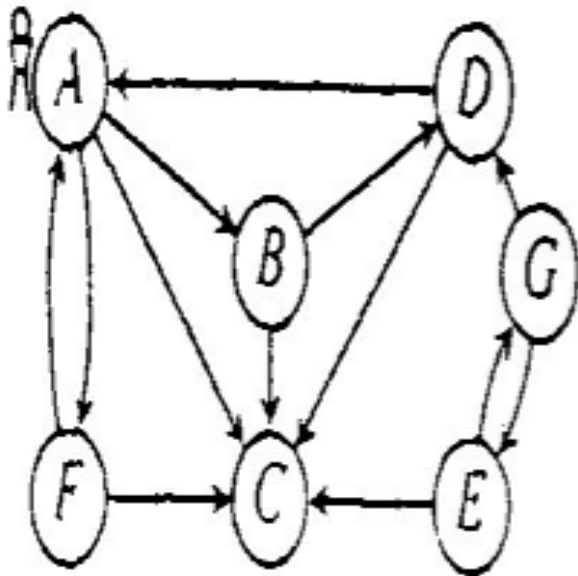
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- Breadth-first search: Strategy (for digraph)
  - choose a starting vertex, distance  $d = 0$
  - vertices are visited in order of increasing distance from the starting vertex,
  - examine all edges leading from vertices (at distance  $d$ ) to adjacent vertices (at distance  $d+1$ )
  - then, examine all edges leading from vertices at distance  $d+1$  to distance  $d+2$ , and so on,
  - until no new vertex is discovered

# Breath-first search, e.g.

- e.g. Start from vertex A, at  $d = 0$ 
  - visit B, C, F; at  $d = 1$
  - visit D; at  $d = 2$
- e.g. Start from vertex E, at  $d = 0$ 
  - visit G: at  $d = 1$



# Breadth-first search: I/O Data Structures

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*Input:*  $G = (V, E)$ , a graph represented by an adjacency list structure, `adjVertices`, as described in Section 7.2.3, where  $V = \{1, \dots, n\}$ ;  $s \in V$ , the vertex from which the search begins.

*Output:* A breadth-first spanning tree, stored in the parent array. The parent array is passed in and the algorithm fills it.

*Remarks:* For a queue  $Q$ , we assume operations of the Queue abstract data type (Section 2.4.2) are used. The array `color[1], ..., color[n]` denotes the current search status of all vertices. Undiscovered vertices are white; those that are discovered but not yet processed (in the queue) are gray; those that are processed are black.

# Breadth-first search: Algorithm

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```
void breadthFirstSearch(IntList[] adjVertices, int n, int s, int[] parent)
    int[] color = new int[n+1];
    Queue pending = create(n);
    Initialize color[1], . . . , color[n] to white.

    parent[s] = -1;
    color[s] = gray;
    enqueue(pending, s);
    while (pending is nonempty)
        v = front(pending);
        dequeue(pending);
        For each vertex w in the list adjVertices[v]:
            if (color[w] == white)
                color[w] = gray;
                enqueue(pending, w);
                parent[w] = v; // Process tree edge vw.
            // Continue through list.
        // Process vertex v here.
        color[v] = black;
    return;
```



# Breadth-first search: Analysis

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- For a digraph having  $n$  vertices and  $m$  edges
  - Each edge is processed once in the while loop for a cost of  $\theta(m)$
  - Each vertex is put into the queue once and removed from the queue and processed once, for a cost  $\theta(n)$
  - Extra space is used for color array and queue, there are  $\theta(n)$
- From a *tree* (breadth-first spanning tree)
  - the path in the tree from start vertex to any vertex contains the *minimum* possible number of edges
- Not all vertices are necessarily reachable from a selected starting vertex

# DFS: the Strategy in Words

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- Depth-first search: Strategy
  - Go as deep as can visiting un-visited nodes
    - Choose any un-visited vertex when you have a choice
  - When stuck at a dead-end, backtrack as little as possible
    - Back up to where you could go to another unvisited vertex
  - Then continue to go on from that point
  - Eventually you'll return to where you started
    - Reach all vertices? Maybe, maybe not
- Things are a bit different for directed vs. undirected graphs
  - It's not really that different, until you get interested in using DFS to find cycles

# Observations about the DFS Strategy

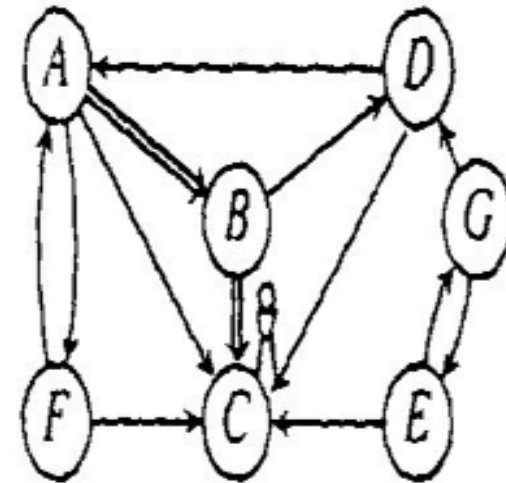
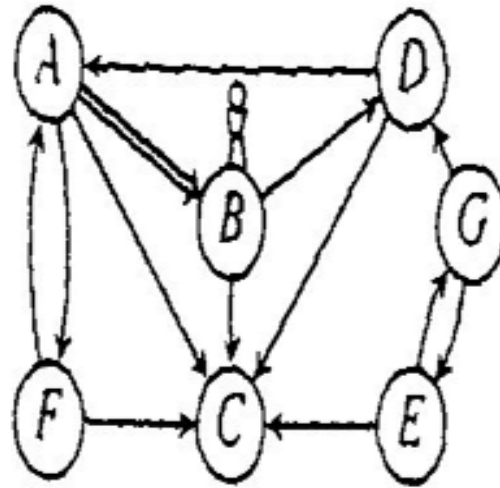
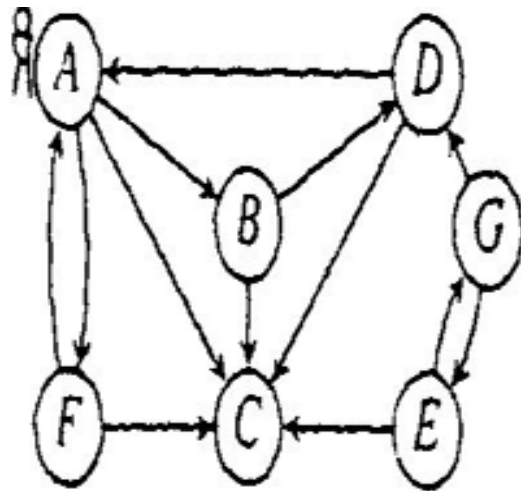
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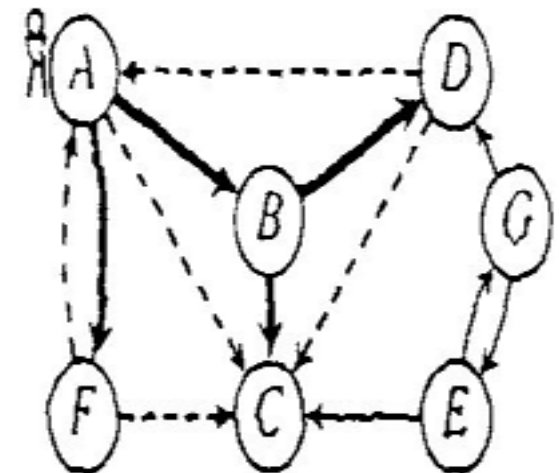
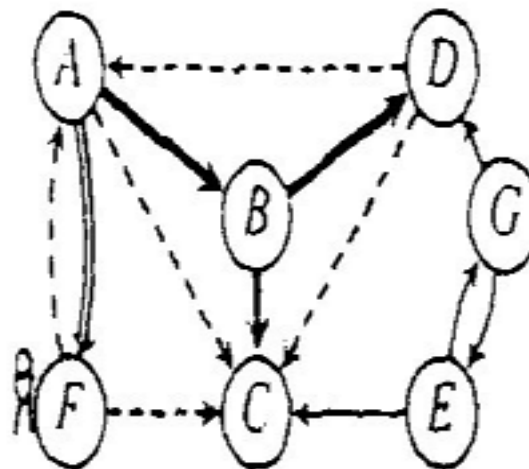
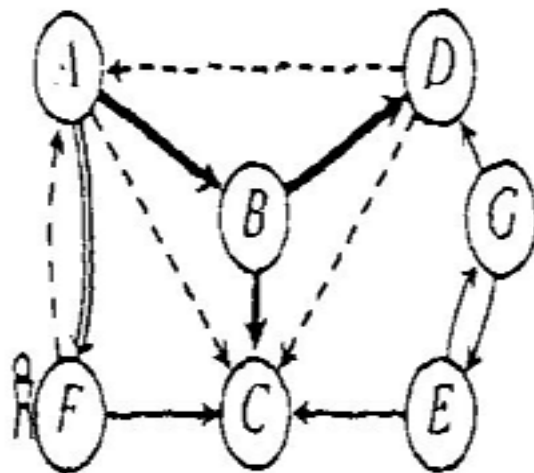
- Note: we must keep track of what nodes we've visited
- DFS traverses a subset of  $E$  (the set of edges)
  - Creates a tree, rooted at the starting point: the Depth-first Search Tree (DFS tree)
  - Each node in the DFS tree has a distance from the start. (We often don't care about this, but we could.)
- At any point, all nodes are either:
  - Un-discovered
  - Finished (you backed up from it), or
  - Discovered (I.e. visited) but not finished
    - On the path from the current node back to the root
    - We might back up to it
  - (Later we'll call these states: white, black and gray)

# An Example of DFS

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and so on...

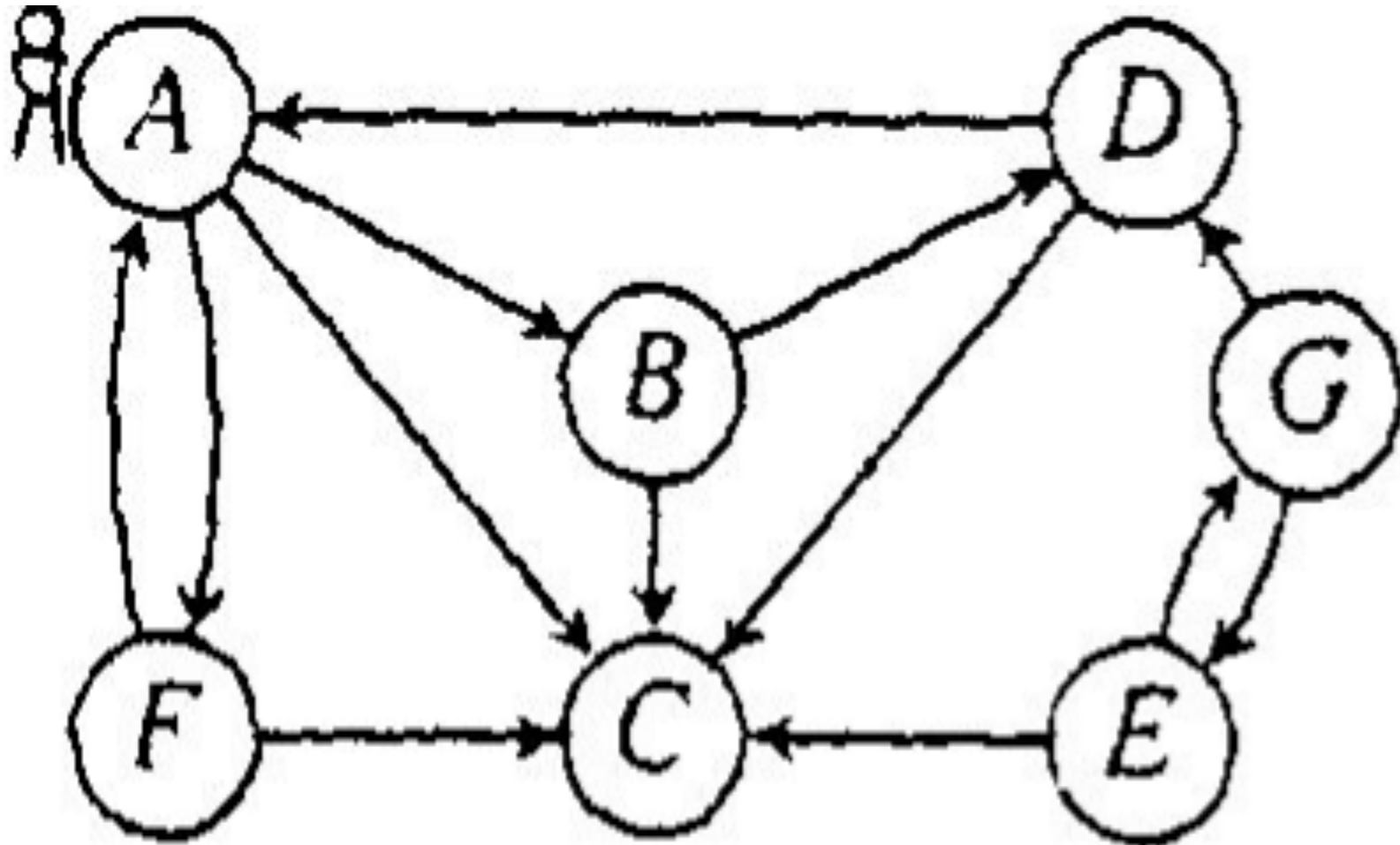


# Depth-first Search, e.g. trace it, in order

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- Vertex status: undiscovered, discovered, finished
- Edge status: part of DFS tree or not?



# Recursive DFS visit function

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```
dfs_rekurs(adj, start) {
    // reached node "start"; do something?
    visit[start] = true
    trav = adj[start]
    while (trav != null) {
        v = trav.ver
        if (!visit[v])
            dfs_rekurs(adj, v)
        trav = trav.next
    }
    // about to leave "start"; do something?
}
```

- Sometimes called dfs\_visit().

# Calling Function for DFS

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- Purpose: do all required initializations, then call `dfs_rekurs()` at a given node (just one call)

Input Parameters: `adj, start`

Output Parameters: None

```
dfs(adj, start) {  
    // do any initializations  
    n = adj.last  
    for i = 1 to n  
        visit[i] = false  
  
    // one call to recursive function at start  
    dfs_rekurs(adj, start)  
}
```

# **DFS to Process all Vertices in a Graph**

- Purpose: do all required initializations, then call `dfs_rekurs()` as many times as needed to visit all nodes. May create a DFS forest.

```
dfs_sweep(adj) {
    n = adj.last
    // do any initializations
    for i = 1 to n
        visit[i] = false

    // loop called on any unvisited node
    for i = 1 to n
        if (!visit[i]) dfs_rekurs(adj, i)
}
```



## **Notes on dfs\_rekurs() function**

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- Often called "dfsVisit" (or something like that)
- Creates one DFS tree from a given start node
  - Must be called by some caller function
  - May not visit all nodes in a the graph G
- Assumes that all nodes have been initialized as "undiscovered"
- Sometimes an "else" clause that does something to nodes not visited (or edges to those)

# General Skeleton Similar to DFS\_rekurs (Cormen)

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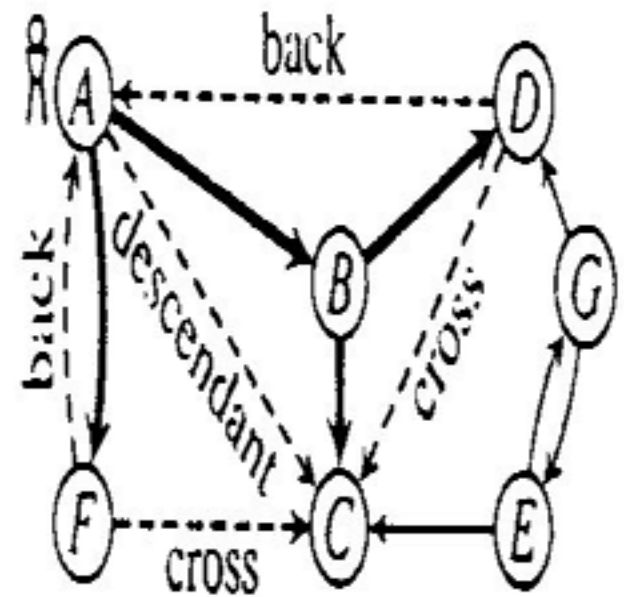
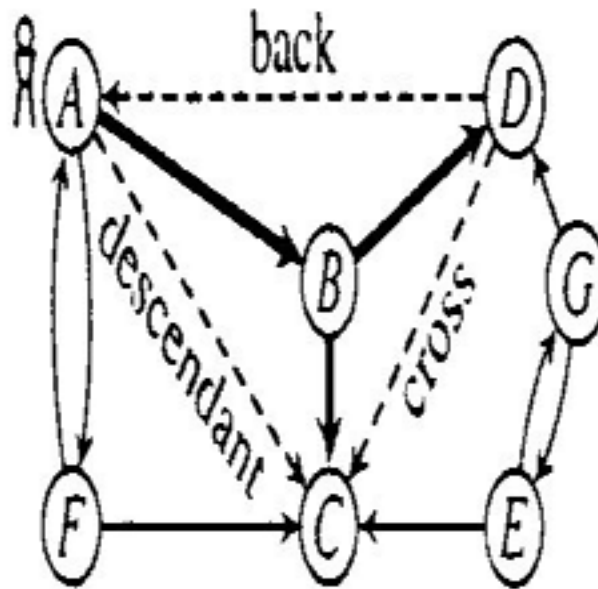
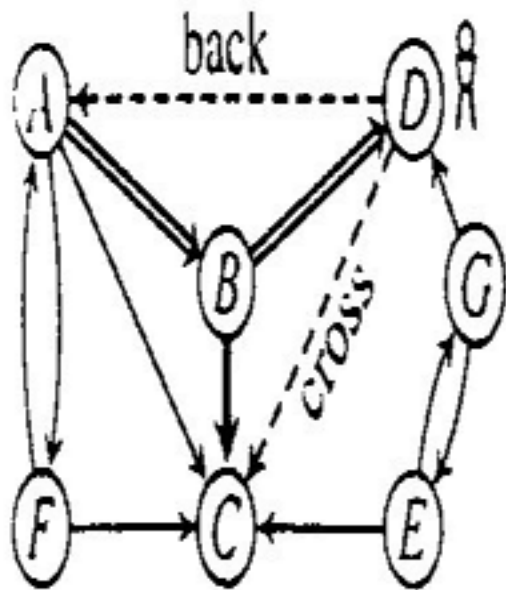
```
int dfs(IntList[] adjVertices, int[] color, int v, ...)  
    int w;  
    IntList remAdj;  
    int ans;  
1. color[v] = gray;  
2. Preorder processing of vertex  $v$   
3. remAdj = adjVertices[v];  
4. while (remAdj  $\neq$  nil)  
5.     w = first(remAdj);  
6.     if (color[w] == white)  
7.         Exploratory processing for tree edge  $vw$   
8.         int wAns = dfs(adjVertices, color, w, ...);  
9.         Backtrack processing for tree edge  $vw$ . using wAns (like inorder)  
10.    else  
11.        Checking (i.e., processing) for nontree edge  $vw$   
12.    remAdj = rest(remAdj)  
13. Postorder processing of vertex  $v$ . including final computation of ans  
14. color[v] = black;  
15. return ans;
```

# Using DFS to Find if a Graph is Acyclic

- Does a graph have a cycle?
  - DFS is great for this
  - But, slightly harder if graph is undirected
- Use DFS tree: classify edges and nodes as you process them
  - Nodes:
    - White: unvisited
    - Black: done with it, backed up from it (never to return)
    - Gray: Have reached it; exploring it's adjacent nodes; but not done with it
  - Also, have a "time counter", say, `ctr`
    - Set  $d[v] = ctr++$  as discovery time
    - Set  $f[v] = ctr++$  as finish time

# Depth-first search tree

- edges classified:
  - tree edge, back edge, descendant edge, and cross edge



# **Using Non-Tree Edges to Identify Cycles**

- From the previous graph, note that:
- Back edges (indicates a cycle)
  - `dfs_rekurs()` sees a vertex that is gray
  - This back edge goes back up the DFS tree to a vertex that is on the path from the current node to the root
- Cross Edges and Descendant Edges (not cycles)
  - `dfs_rekurs()` sees a vertex that is black
  - Descendant edge: connects current node to a descendant in the DFS tree
  - Cross edge: connects current node to a node in another subtree – not a descendant of current node

# Non-tree Edges in DFS

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- Question 1: Finding back edges for an undirected tree is not **quite** this simple:
  - The parent node of the current node is gray
  - Not a cycle, is it? It's the same edge you just traversed
  - Question: how would you modify our code to recognize this?
- Question 2:
  - How could you modify the code to distinguish cross edges from descendant edges?
  - Hint: use discovery and finish times

# Time Complexity of DFS

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- For a digraph having  $n$  vertices and  $m$  edges
  - Each edge is processed once in the while loop of `dfs_rekurs()` for a cost of  $\theta(m)$ 
    - Think about adjacency list data structure.
    - Traverse each list exactly once. (Never back up)
    - There are a total of  $2m$  nodes in all the lists
  - The `dfs_sweep()` algorithm will do  $\theta(n)$  work even if there are no edges in the graph
  - Thus over all time-complexity is  $\theta(n+m)$ 
    - Remember: this means the larger of the two values
    - Note: This is considered “linear” for graphs since there are two size parameters for graphs.
- Extra space is used for color array.  
Space complexity is  $\theta(n)$

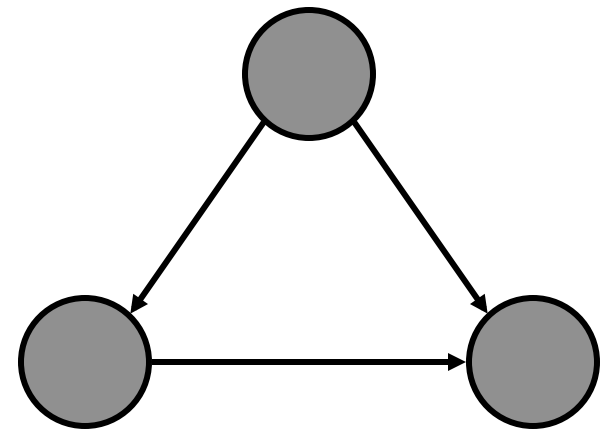
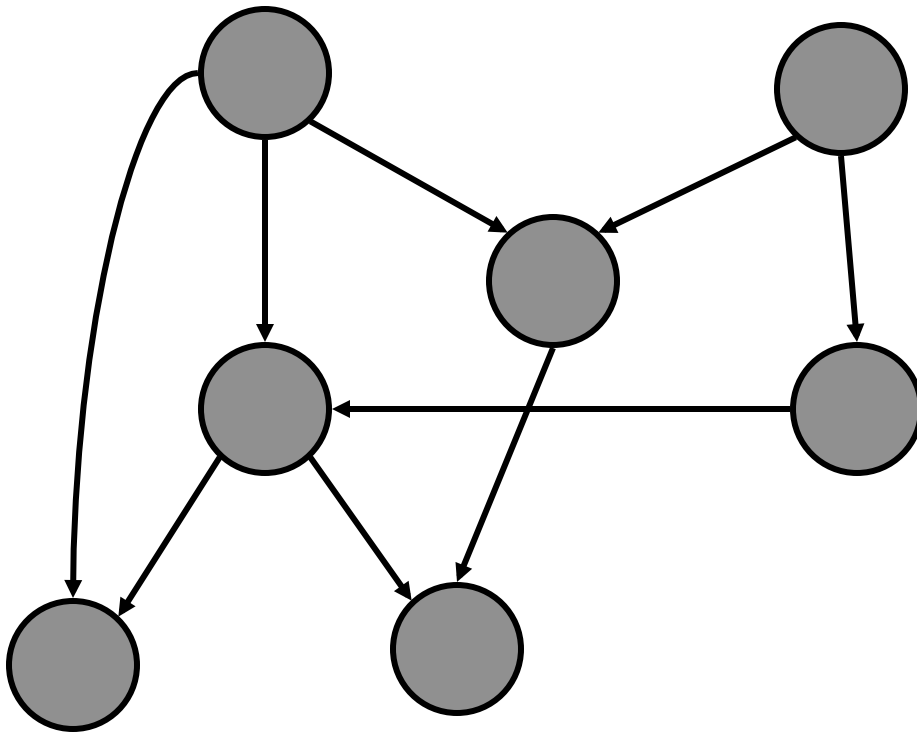




# Directed Acyclic Graphs

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- A *directed acyclic graph* or *DAG* is a directed graph with no directed cycles:



# Topological Sort

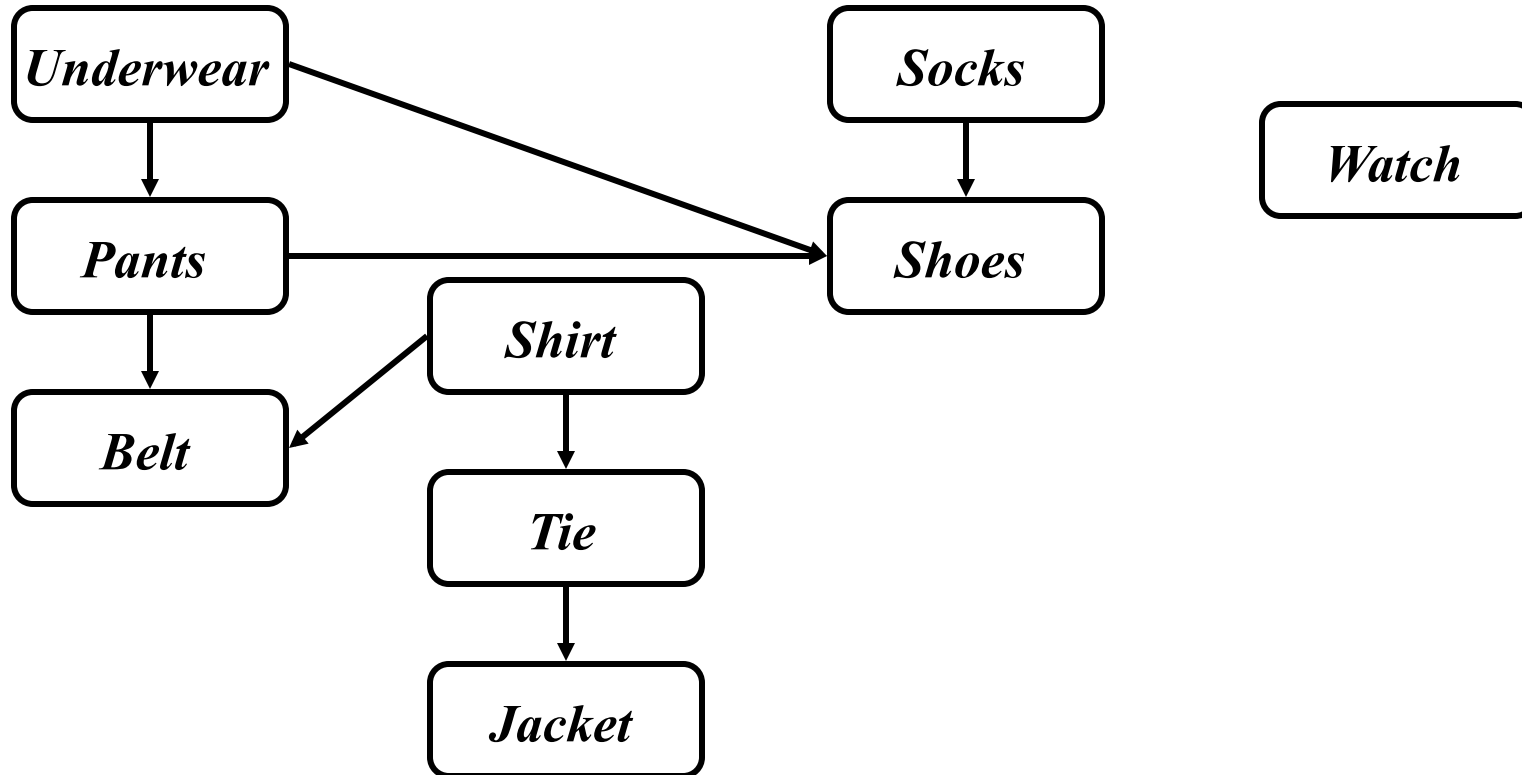
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- *Topological sort* of a DAG:
  - Linear ordering of all vertices in graph  $G$  such that vertex  $u$  comes before vertex  $v$  if edge  $(u, v) \in G$
- Real-world example: getting dressed

# Getting Dressed

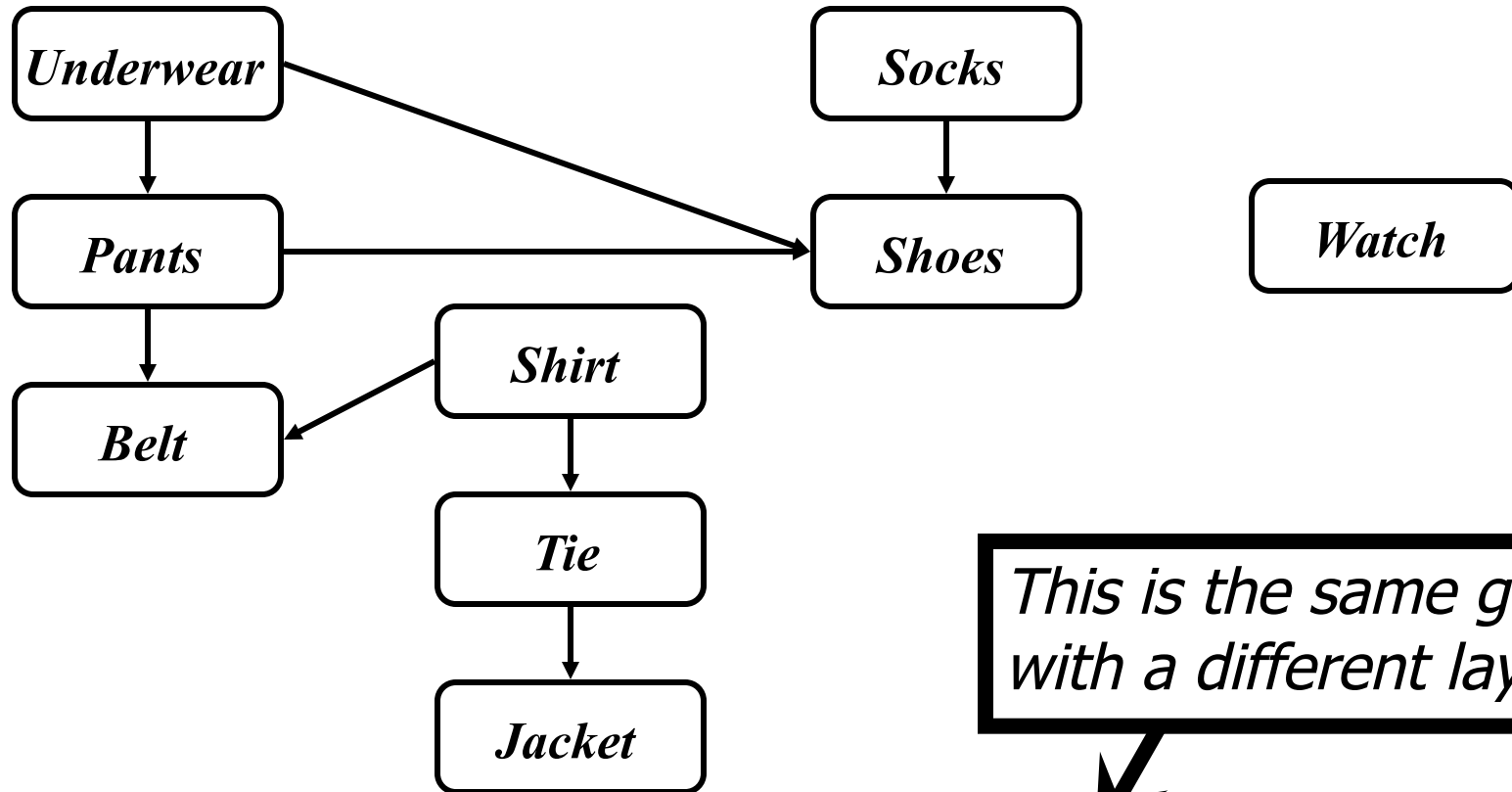
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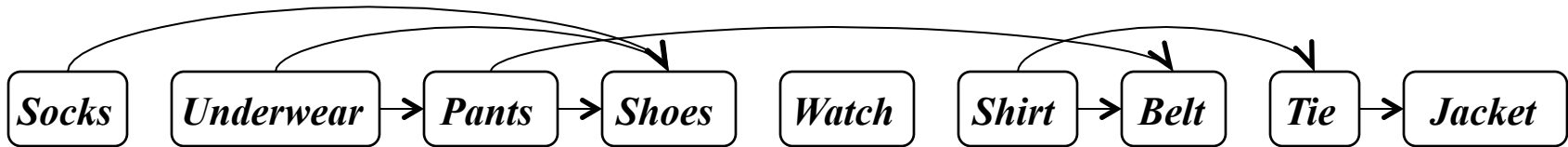
# Getting Dressed

---

---



*This is the same graph with a different layout.*



# Topological Sort Algorithm

---

```
Topological-Sort()
```

```
{
```

```
    Run DFS_rekurs()
```

```
    When a vertex is finished, output it
```

```
    Vertices are output in reverse  
    topological order
```

```
        (or add to stack/list)
```

```
}
```

- Can stack/store vertices as found to store them in topologically sorted order
- Time:  $O(V+E)$

# Topological Sort, Recursive Function

```
top_sort_rekurs(adj, start, ts) {  
    visit[start] = true  
    trav = adj[start]  
    while (trav != null) {  
        v = trav.ver  
        if (!visit[v])  
            top_sort_rekurs(adj,v,ts)  
        trav = trav.next  
    }  
    ts[k] = start  
    k = k - 1  
}
```

# Topological Sort: Driver

---

---

```
top_sort(adj, ts) {
    n = adj.last
    // k is the index in ts where the next vertex is to be
    // stored in topological sort. k is assumed global.
    k = n
    for i = 1 to n
        visit[i] = false
    for i = 1 to n
        if (!visit[v])
            top_sort_rekurs(adj, i, ts)
    }
}
```

# Forward vs. Reverse

---

---

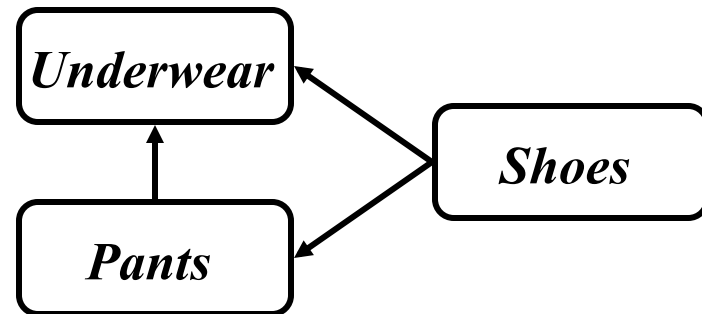
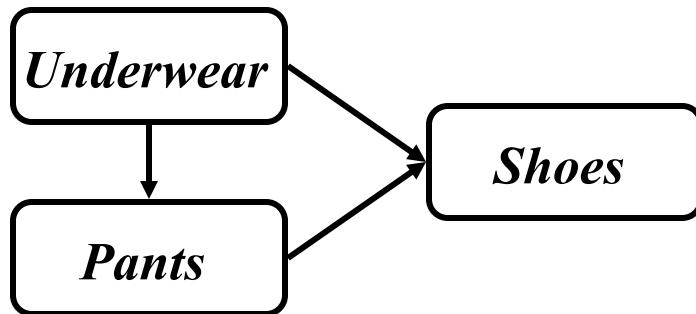
- Topological sort is a type of sort
  - Implies an ordering
  - Can sort backwards, of course
- Forward topological order
  - If edge  **$vw$**  in graph, then  $\text{topo}[v] < \text{topo}[w]$
- Reverse topological order
  - If edge  **$vw$**  in graph, then  $\text{topo}[v] > \text{topo}[w]$
- And, every directed graph has a transpose, which means... (see next slide)



# What's an Edge Mean?

---

- What's our graph model?
  - Edge  $uv$  means do  $u$  first, then  $v$ . Or, ...
  - Edge  $uv$  means task  $u$  depends on  $v$  (I.e.  $v$  must be done first)



- The latter called a dependency graph
- “forward in time” vs. “depend on this one”
- Big deal? No, we can order vertices in reverse topological order if needed

# Sort this!

---

---

