

# COMP2421 – DATA STRUCTURES AND ALGORITHMS

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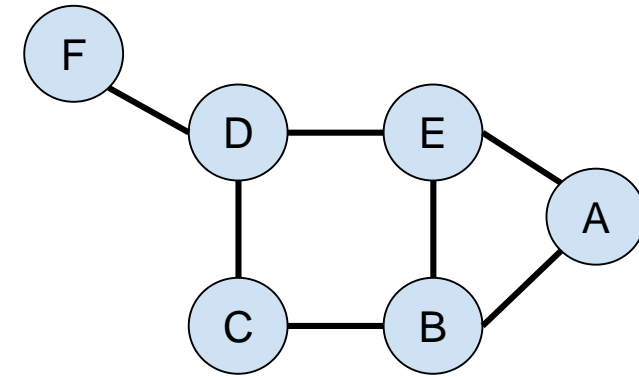
## Graphs

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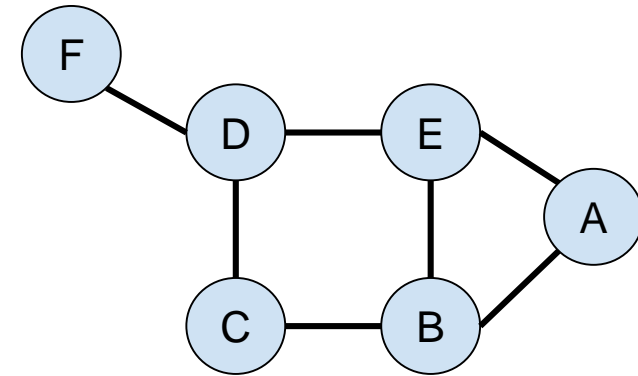
# Graphs

- Graphs are mathematical concepts that have many applications in computer science.
- They have many applications in real-life applications such as social networks, locations and routers in GPS, ...
- A graph consists of a finite set of vertices (i.e., nodes) and a set of edges connecting these vertices.
- Two vertices are called adjacent if they are connected to each other by the same edge.



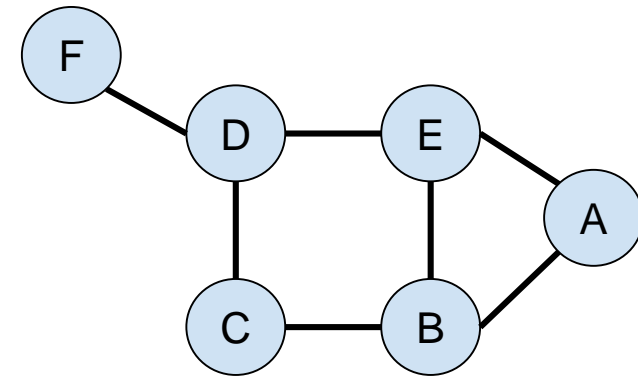
# Graphs

- A graph  $G=(V, E)$ , is a data structure that consists of a finite set of vertices (or nodes)  $V$ , and a set of edges,  $E$ .
- Each edge is a pair  $(v, w)$  where  $v$  and  $w$  are nodes from  $V$ .



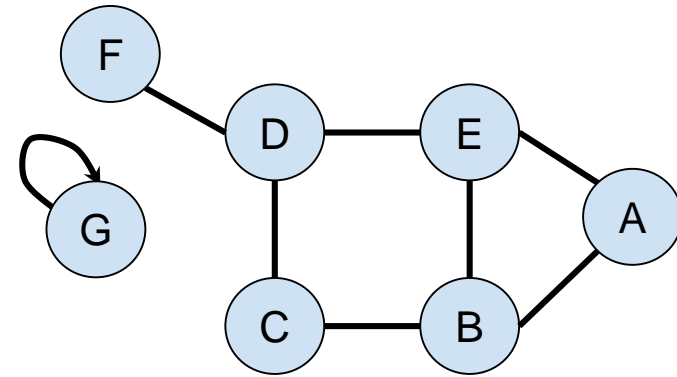
# Graphs

- If the pairs are ordered in the graph, then the graph is called **directed graph**(diagraphs).
- Vertex  $w$  is **adjacent** to  $v$  if and only if  $(v, w) \in E$ . In an undirected graph with edge  $(v, w)$ , and hence  $(w, v)$ ,  $w$  is adjacent to  $v$  and  $v$  is adjacent to  $w$ .



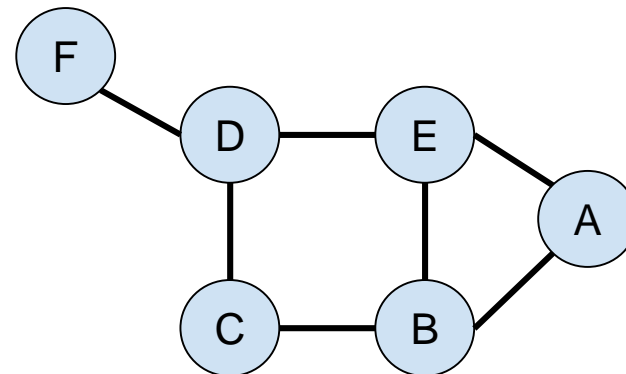
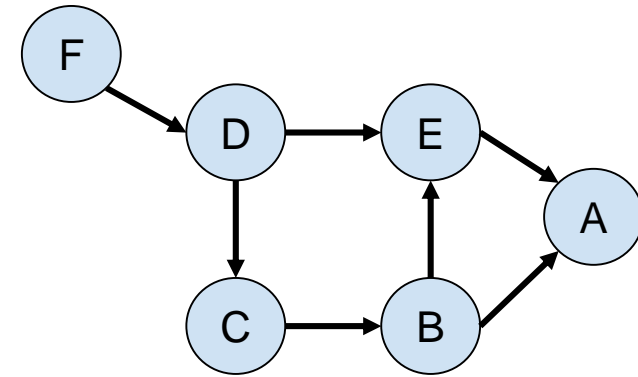
# Graphs - Definitions

- Order: is the number of vertices in a graph
- Size: is the number of edges in a graph
- Vertex degree: is the number of edges that are connected to a vertex
- Isolated vertex: is the vertex that is not connected to any other vertex in the graph
- Self-loop: an edge from a vertex to itself



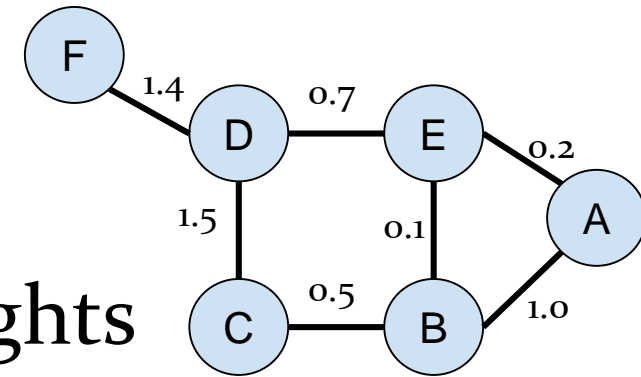
# Graphs - Definitions

- Directed graph: is a graph where all edges have directions indicating what is the start vertex and what is the end vertex
- Undirected graph: is a graph with edges that have no directions

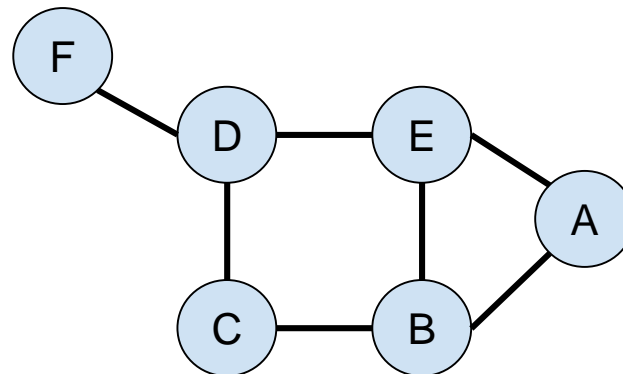


# Graphs - Definitions

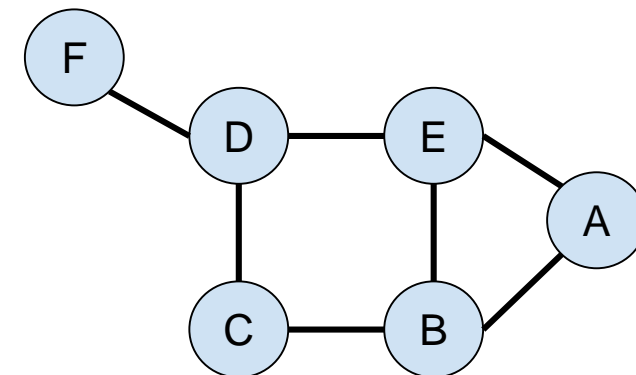
- Weighted graph: edges of a graph have weights



- Unweighted graph: edges of a graph have no weights



# Graphs - Definitions

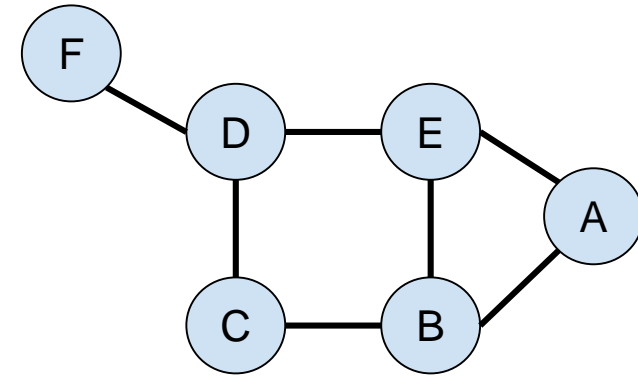


- A **path** in a graph is a sequence of vertices  $w_1, w_2, w_3, \dots, w_N$ , such that  $(w_i, w_{i+1}) \in E$  for  $1 \leq i < N$ . The **length** of such a path is the number of edges on the path, which is equal to  $N - 1$ .
- A path from a vertex to itself is allowed. If it does not contain edges, then the path length is 0. If edge  $(v, v)$ , then the path  $v$  (which is also referred to as a loop).
- Cycle: a path  $w_1, w_2, w_3, \dots, w_N$  for which  $N > 2$ , the first  $N - 1$  vertices are all different, and  $w_1 = w_N$ . For example, the sequence D, E, A, B, C, D is a cycle in the graph above.



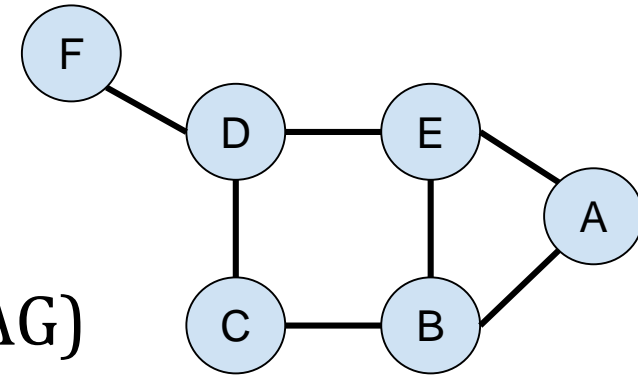
# Graphs - Definitions

- A simple path is a path such that all vertices are distinct (except that the first and last might be the same).
- A cycle in a directed graph is a path of length at least one such that  $w_1 = w_n$ .
- The path  $v, u, v$  is cyclic. However, it is not in undirected graph because  $(v,u)$  and  $(u,v)$  is the same path.



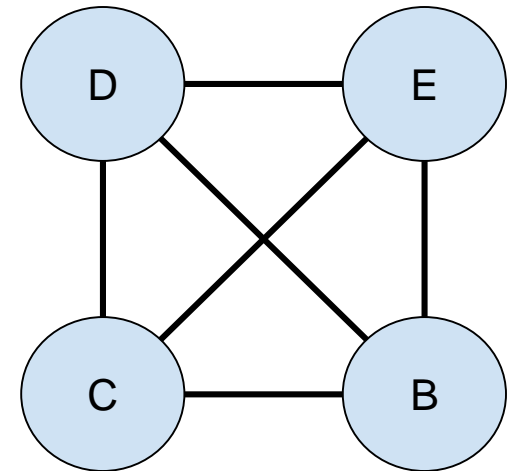
# Graphs - Definitions

- A directed graph is called acyclic if it has no cycles (DAG)
  - Acyclic directed graph.
- An undirected graph is called connected if there is a path from every node to every other node. A directed graph with this property is called strongly connected.



# Graphs - Definitions

- A complete graph is a graph in which there is an edge between every pair of vertices.



# Examples of using graphs

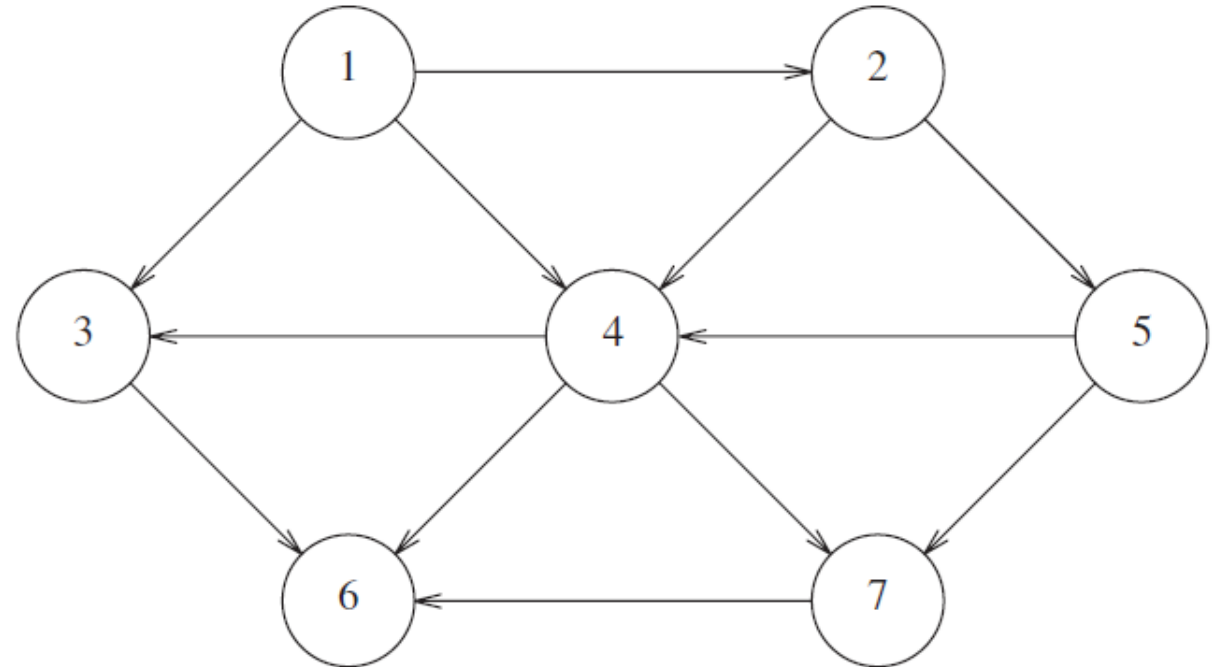
- Airport System
- Graphs are used to represent networks. The networks may include paths in a city or telephone network or circuit network.
- Graphs are also used in social networks like LinkedIn, Facebook. For example, in Facebook, each person is represented with a vertex(or node). Each node is a structure and contains information like person id, name, gender, and locale.

# REPRESENTATION OF GRAPHS

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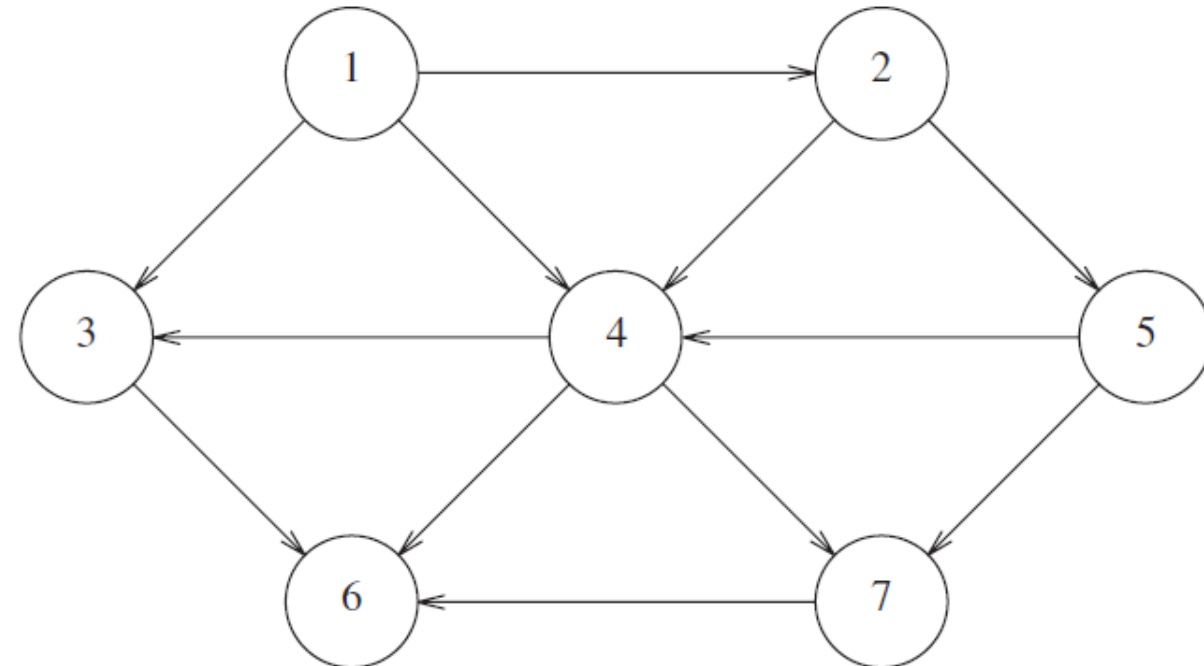
# Graph Representation

- A graph is a data structure that consists of two main components: a finite set of vertices (i.e., nodes); and a finite set of ordered pairs called edges
- Graphs are most commonly represented using
  - Adjacency matrix
  - Adjacency list



# Graph Representation

- Consider the following directed graph (the undirected graph is represented the same way)
- Suppose that we can number the vertices starting at 1. This graph has 7 vertices and 12 edges.
- One method is to represent a graph using a 2D array (adjacency matrix)

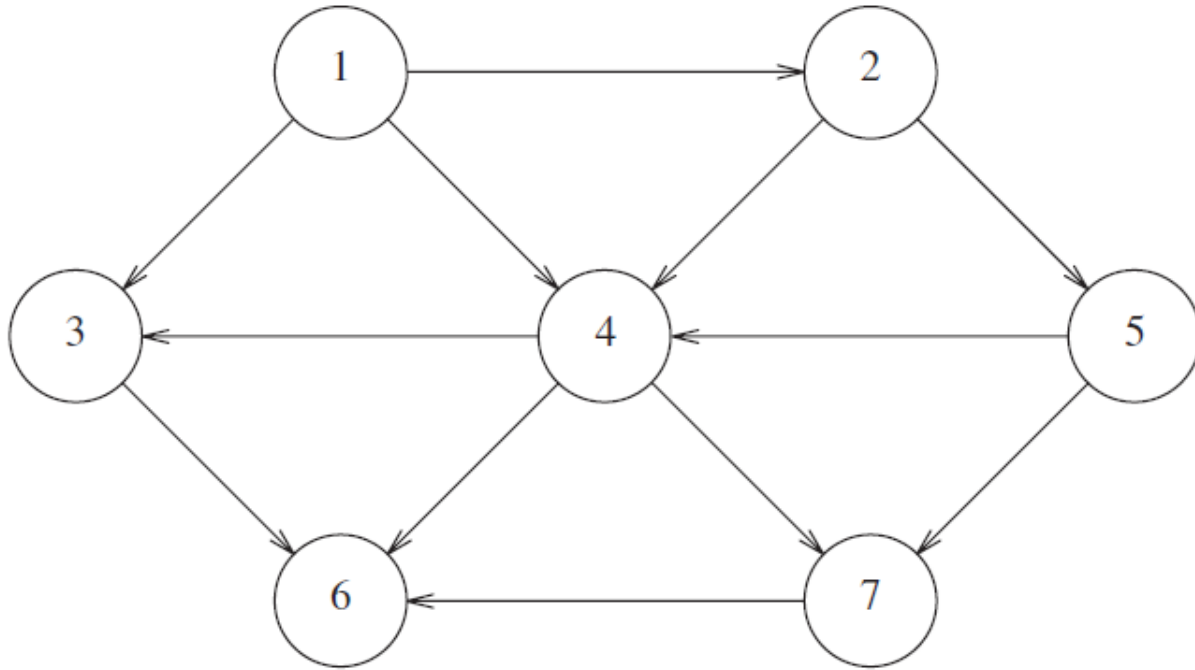


# Adjacency Matrix

- Adjacency Matrix: maintain a 2D-Boolean array of size  $v * v$  where  $v$  is the number of vertices in the graph.
- Let the adjacency matrix  $adj$ , each edge is represented with the value true:  $adj[v][w] = true$  for the edge  $(v, w)$
- The boolean value can be replaced with a weight to represent a weighted graph
- For undirected graph, the adjacency matrix is symmetric



# Adjacency Matrix



	1	2	3	4	5	6	7
1		1	1	1			
2				1	1		
3						1	
4			1			1	1
5				1			1
6							
7						1	

# Adjacency Matrix



# Adjacency Matrix

## Advantages:

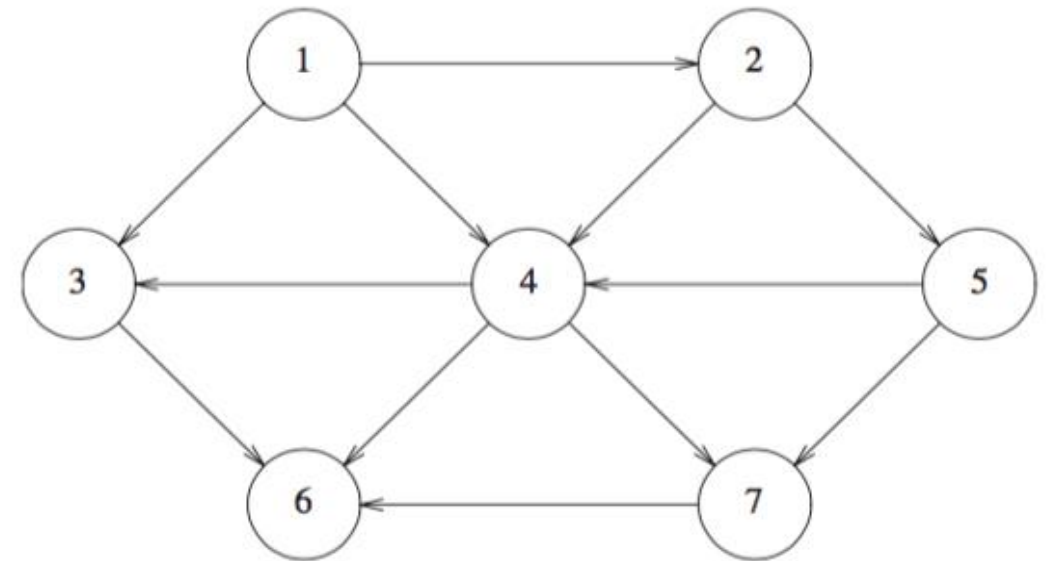
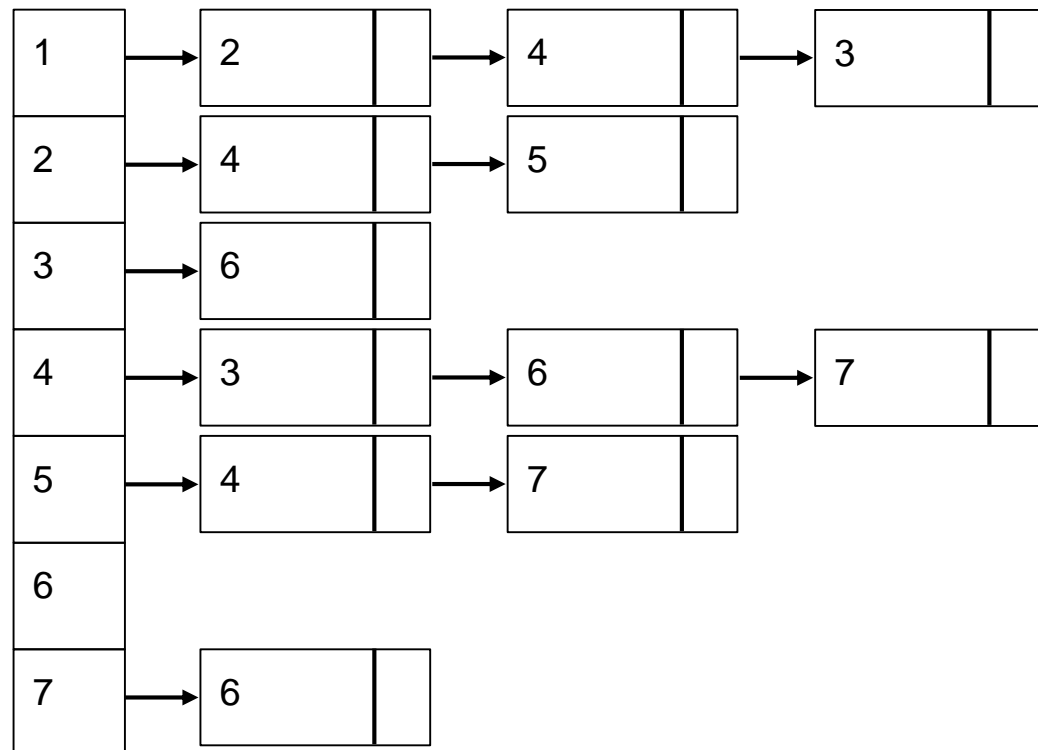
- Easy to implement and follow
- Removing and edge/checking if an edge exists in the graph takes  $O(1)$

## Disadvantages:

- Requires more space  $O(n^2)$  if the graph has a few number of edges between vertices
- Adding a vertex will consume  $O(n^2)$
- Very slow to iterate over all edges

# Adjacency List

- Is a better solution if the graph is sparse (not dense)
- For each vertex, we keep a list of all adjacent vertices
- The space requirement is then  $O(|E| + |V|)$ , which is linear in the size of the graph



**Figure 9.1** A directed graph

# Adjacency List

- Adjacency lists are the standard way to represent graphs
- Undirected graphs can be similarly represented; each edge  $(u, v)$  appears in two lists, so the space usage essentially are doubled
- A common requirement in graph algorithms is to find all vertices adjacent to some given vertex  $v$ , and this can be done in time proportional to the number of such vertices found, by a simple scan down the appropriate adjacency list

# Adjacency List

## Advantages:

- Fast to iterate over all edges
- Fast to add/delete a node (vertex)
- Fast to add a new edge  $O(1)$
- Memory depends more on the number of edges (and less on the number of nodes), which saves more memory if the adjacency matrix is sparse

## Disadvantages:

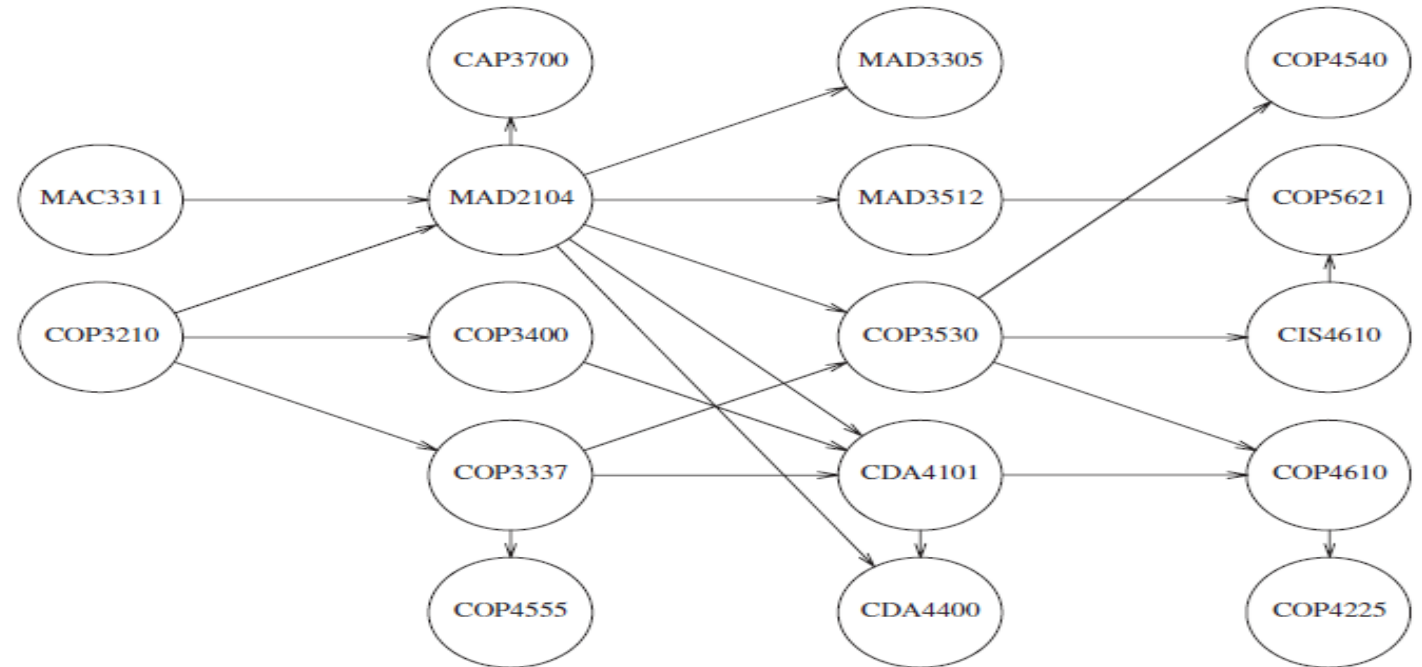
- Finding a specific edge between any two nodes is slightly slower than the matrix  $O(k)$ ; where  $k$  is the number of neighbors nodes

# SORTING GRAPHS

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# Topological Sort

- A linear order of the vertices in a directed graph
- A topological sort is an ordering of vertices in a directed acyclic graph, such that if there is a path from  $v_i$  to  $v_j$ , then  $v_j$  appears after  $v_i$  in the ordering
- An example is the a directed graph that represents the prerequisite of courses in the figure below

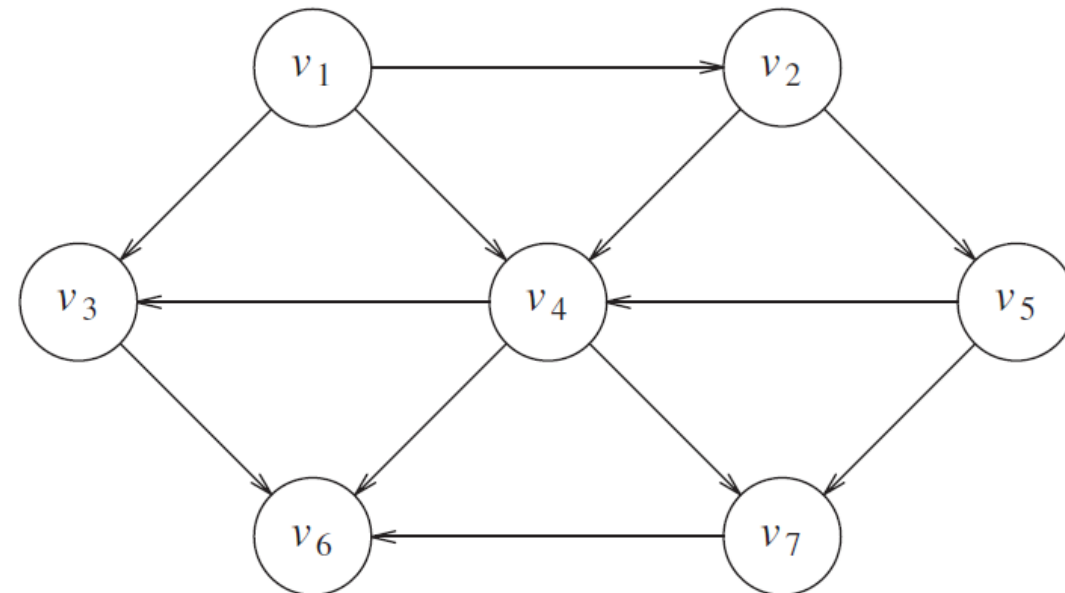


**Figure 9.3** An acyclic graph representing course prerequisite structure



# Topological Sort

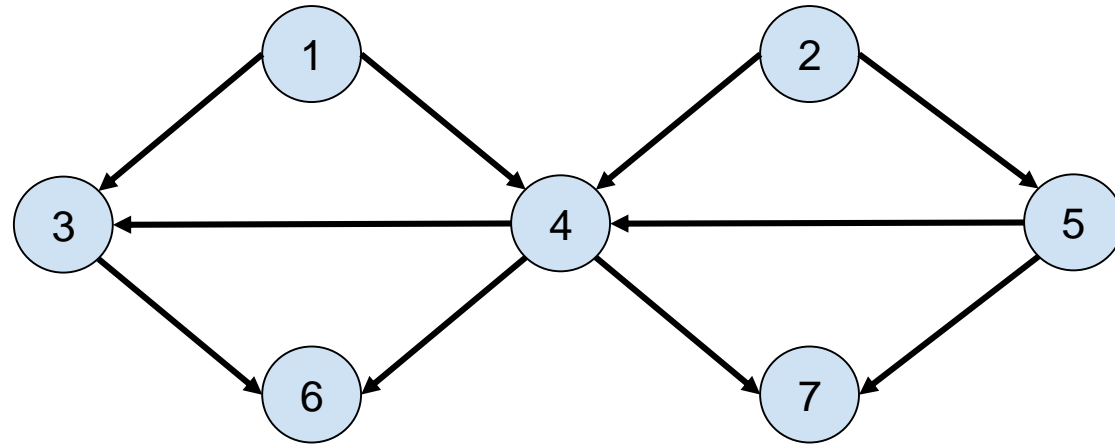
- A directed edge  $(v, w)$  indicates that course  $v$  must be completed before course  $w$  may be attempted
- A topological ordering of these courses is any course sequence that does not violate the prerequisite requirement
- Topological ordering is not possible if the graph has a cycle, since for two vertices  $v$  and  $w$  on the cycle,  $v$  precedes  $w$  and  $w$  precedes  $v$ .
- The ordering is not necessarily unique; any legal ordering will work.
- In this graph,  $v_1, v_2, v_5, v_4, v_3, v_7, v_6$  and  $v_1, v_2, v_5, v_4, v_7, v_3, v_6$  are both topological orderings.



# Topological Sort

- Main idea: find a vertex with nothing going into it (i.e., Starting point). Write it down. Remove it and go through the other vertices and check for anyone with nothing coming into it. Repeat.
- scan all vertices to find the starting point
- \* if edge (A, B) exists, A must precede B in the final order.
- Algorithm:
- Assume indegree is sorted with each node
- Repeat until no nodes remain
  - Choose a node of zero indegree and output it
  - Remove the node and all its edges and update indegree

# Topological Sort - Example



- Indegree:

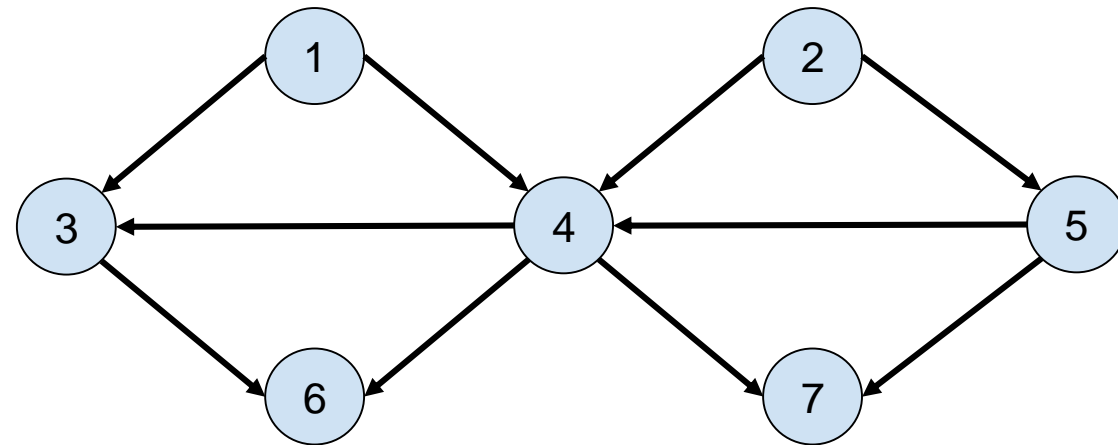
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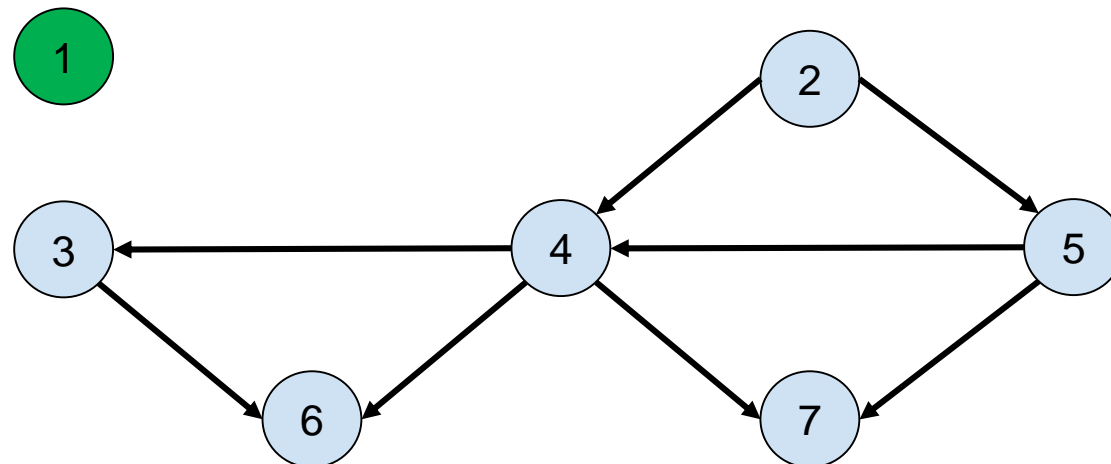
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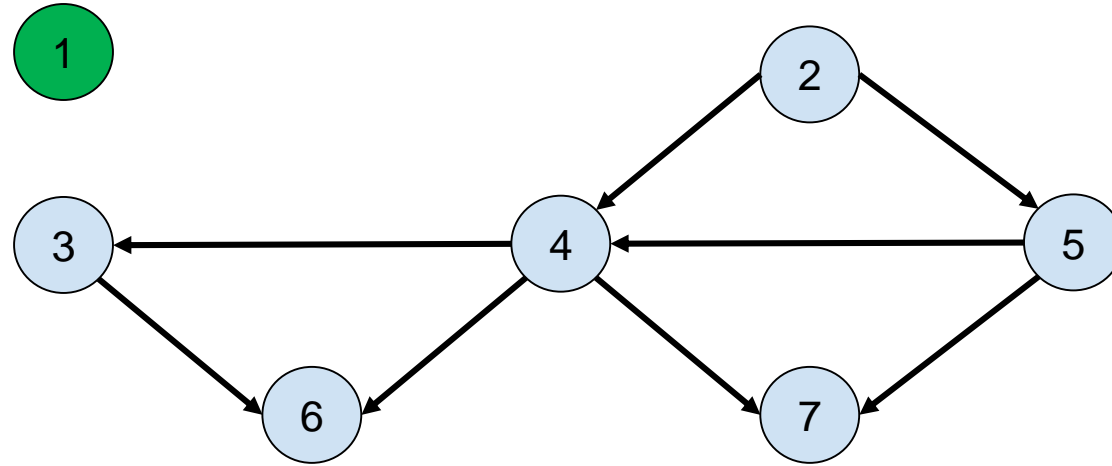
# Topological Sort - Example



- Pick 1 and then update:



# Topological Sort - Example



- Indegree:

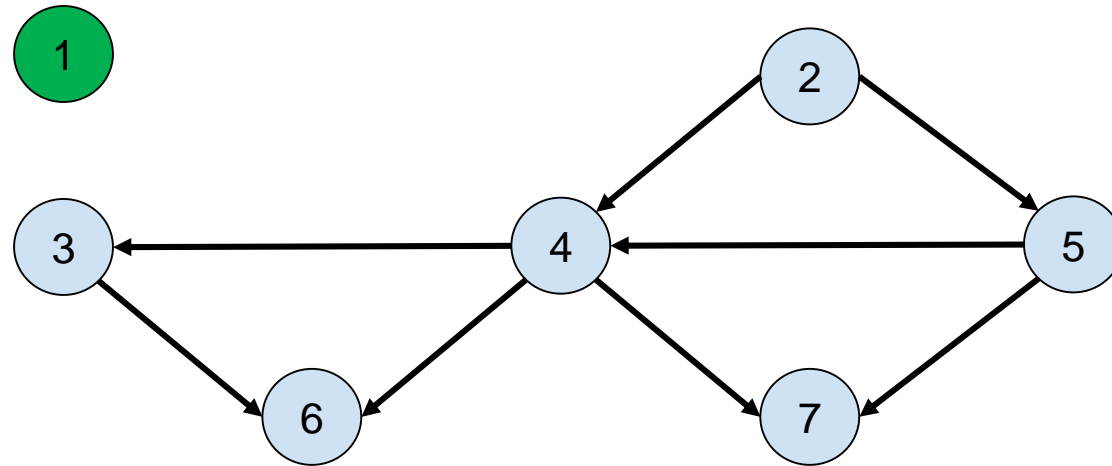
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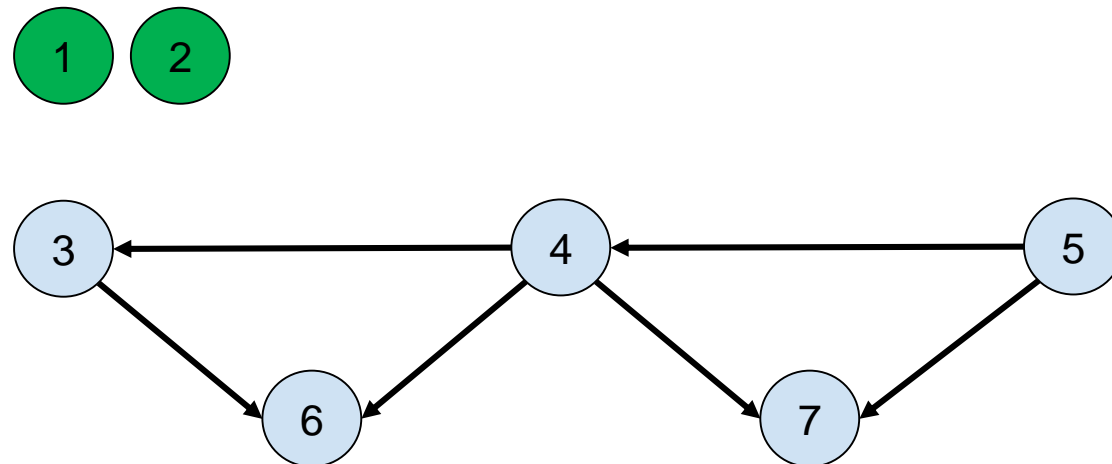
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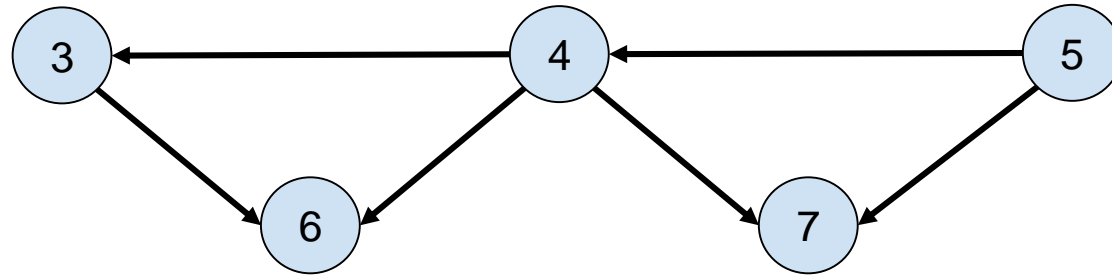
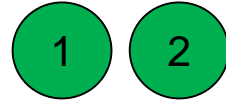
# Topological Sort - Example



- Pick 2 and then update:



# Topological Sort - Example



- Indegree:

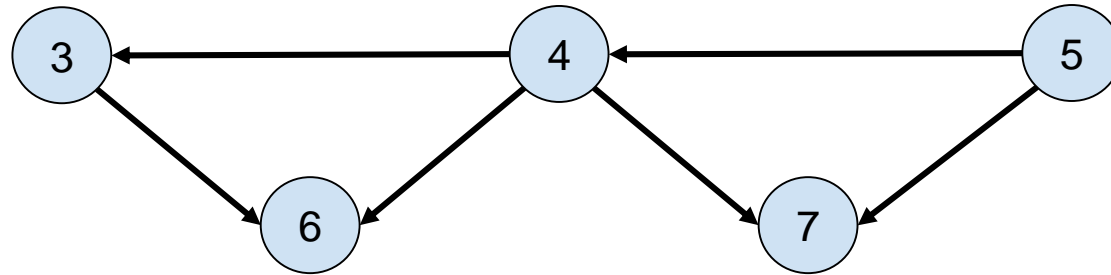
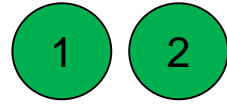
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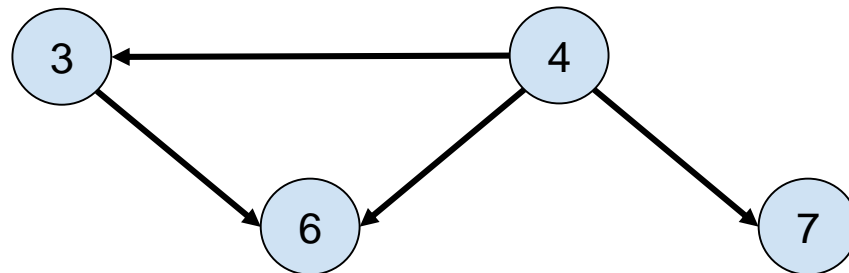
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# Topological Sort - Example

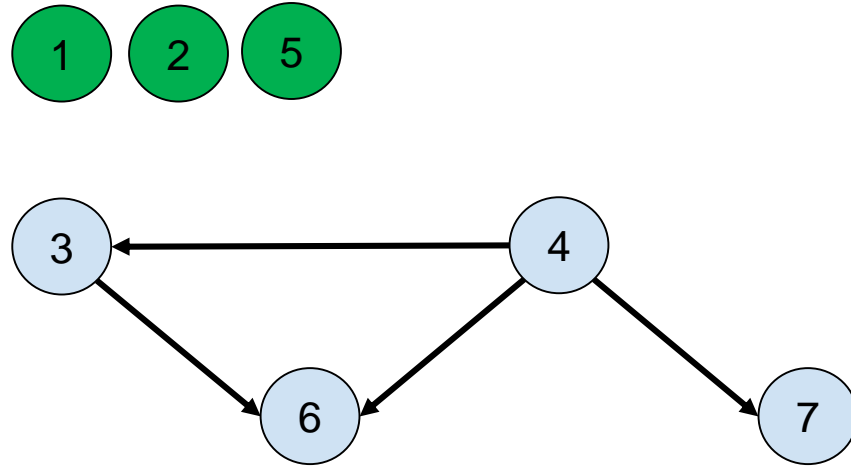


- Pick 5 and then update:





# Topological Sort - Example



- Indegree:

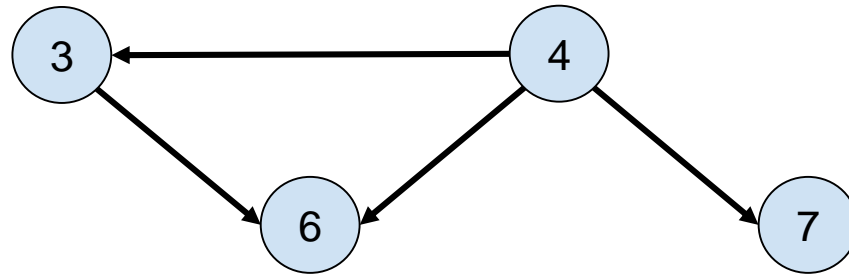
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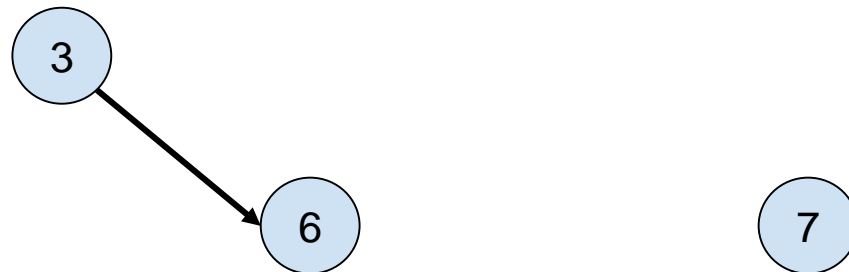
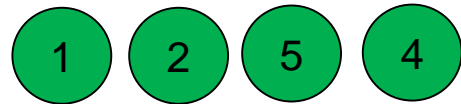
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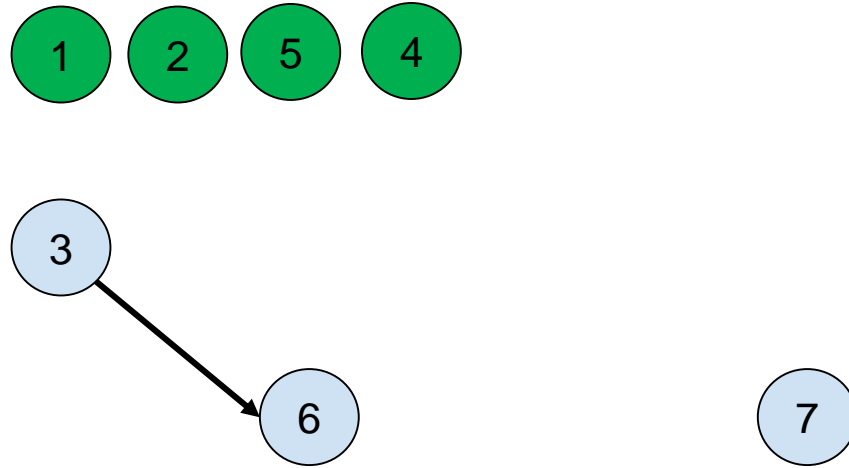
# Topological Sort - Example



- Pick 4 and then update:



# Topological Sort - Example



- Indegree:

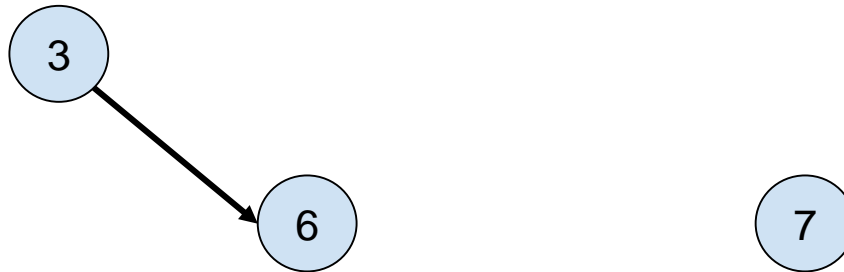
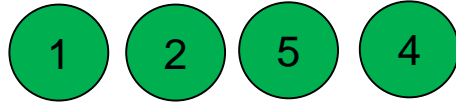
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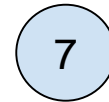
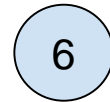
# Topological Sort - Example



- Pick 3 and then update:



# Topological Sort - Example



- Indegree:

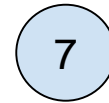
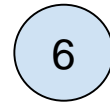
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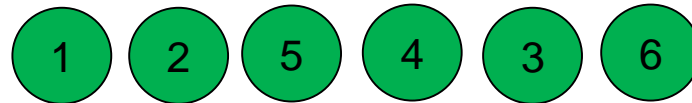
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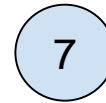
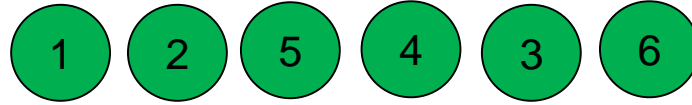
# Topological Sort - Example



- Pick 6 and then update:



# Topological Sort - Example



- Indegree:

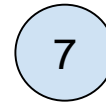
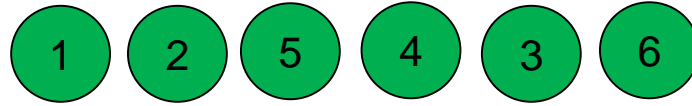
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# Topological Sort - Example



- Pick 7 and then update:





# Topological Sort

- First we find the nodes with no predecessors.
- Then, using a queue, we can keep the nodes with no predecessors and on each dequeue we can remove the edges from the node to all other nodes.

# Topological Sort

- **Pseudocode:**

1. Represent the graph with two lists on each vertex (incoming edges and outgoing edges)
2. Make an empty queue  $Q$ ;
3. Make an empty topologically sorted list  $T$ ;
4. Push all items with no predecessors in  $Q$ ;
5. While  $Q$  is not empty
  - Dequeue from  $Q$  into  $u$ ;
  - Push  $u$  in  $T$ ;
  - Remove all outgoing edges from  $u$ ;
6. Return  $T$ ;

# Topological Sort

- This approach will give us a running time complexity is  $O(|V| + |E|)$ .
- The problem is that we need additional space and an operational queue.

# Topological Sort - Example

# Topological Sort - Example

# Topological Sort - Example

# SEARCH ALGORITHMS

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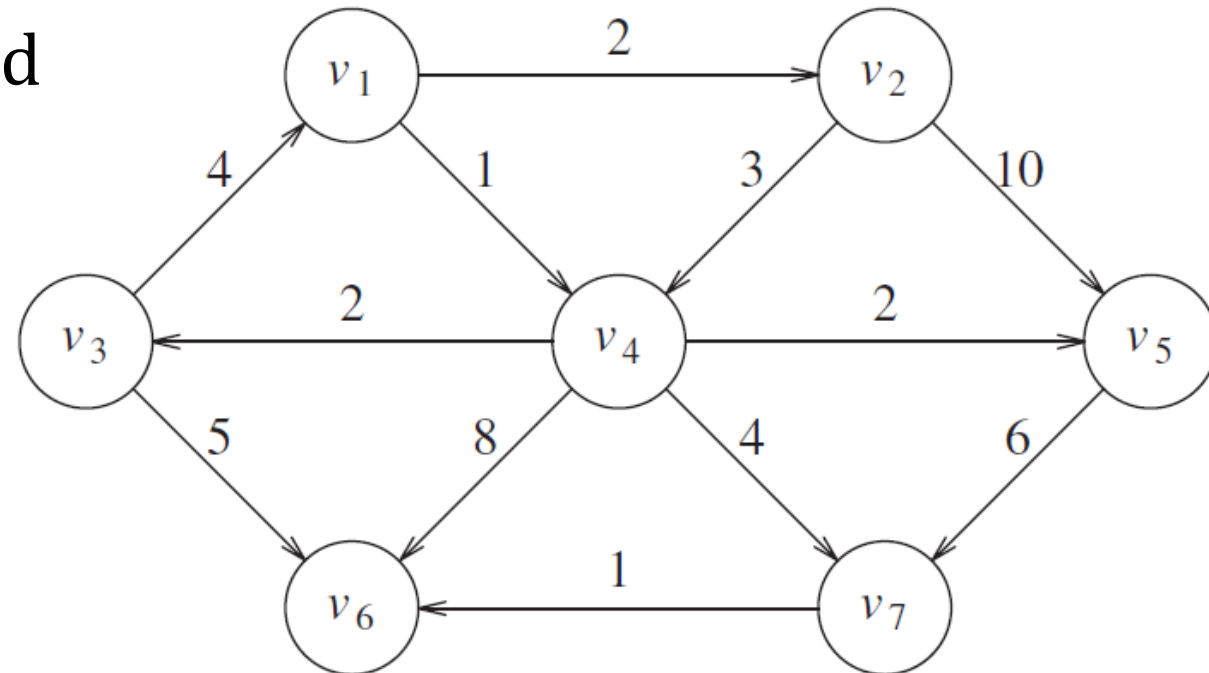
# Shortest-Path Algorithms

- Shortest-path algorithms aim at finding the shortest path between nodes in a graph
- The input is a weighted graph: associated with each edge  $(v_i, v_j)$  is a cost  $c_{i,j}$  to traverse the edge
- The cost of a path  $v_1 v_2 \dots v_N$  is  $\sum_{i=1}^{N-1} c_{i, i+1}$
- This is referred to as the **weighted path length**
- The unweighted path length is the number of edges on the path, namely,  $N - 1$



# Single-Source Shortest-Path Algorithms

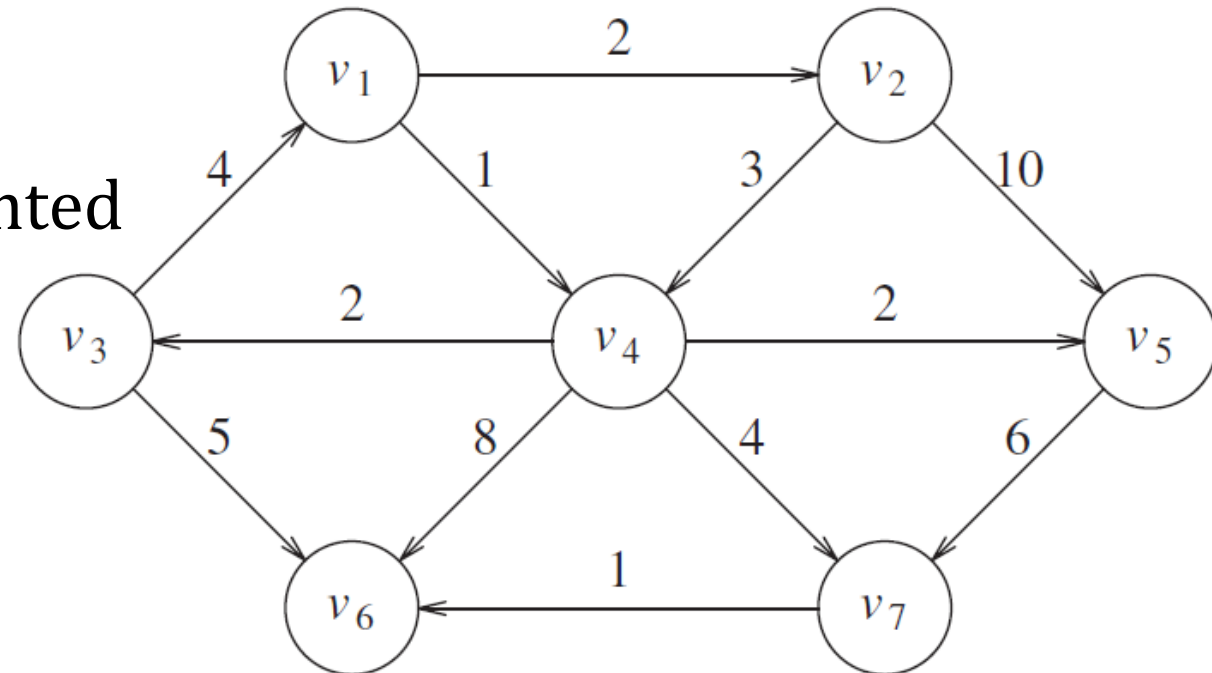
- Given as input a weighted graph,  $G = (V, E)$ , and a distinguished vertex,  $s$ , find the shortest weighted path from  $s$  to every other vertex in  $G$ .
- For example, the shortest weighted path from  $v_1$  to  $v_6$  has a cost of 6 and goes from  $v_1$  to  $v_4$  to  $v_7$  to  $v_6$ .
- The shortest unweighted path between these vertices is 2



**Figure 9.8** A directed graph  $G$

# Single-Source Shortest-Path Algorithms

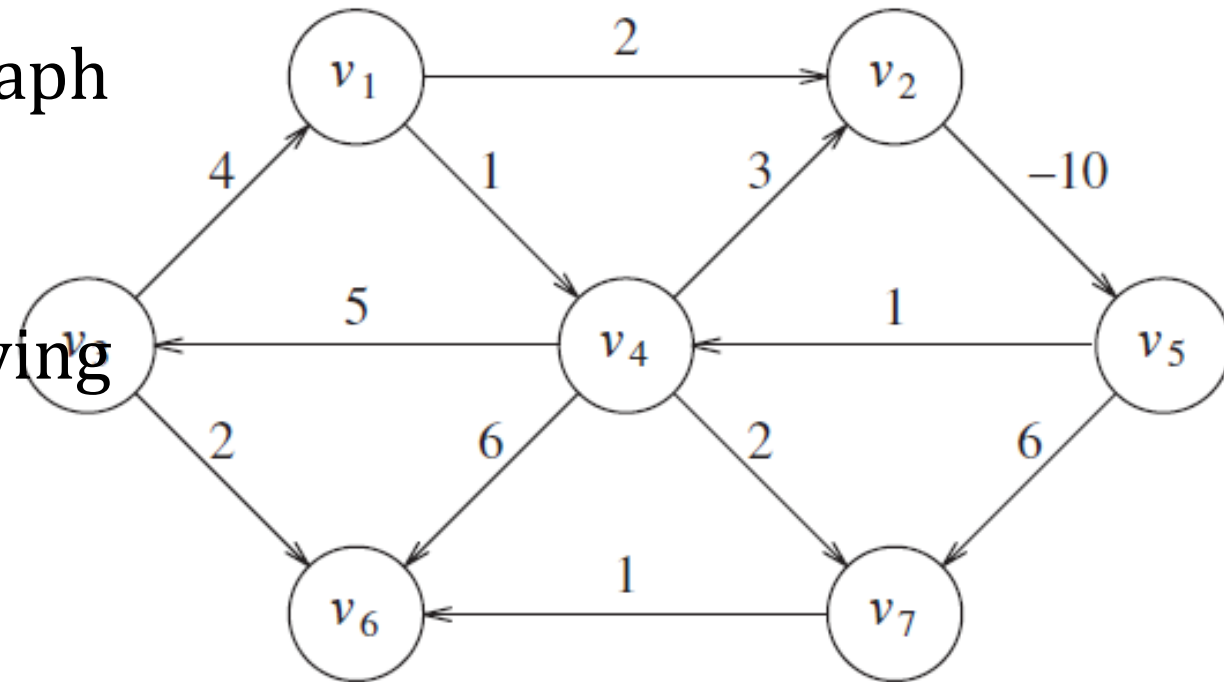
- The shortest unweighted path between these vertices is 2
- Generally, when it is not specified whether we are referring to a weighted or an unweighted path, the path is weighted if the graph is.



**Figure 9.8** A directed graph  $G$

# Single-Source Shortest-Path Algorithms

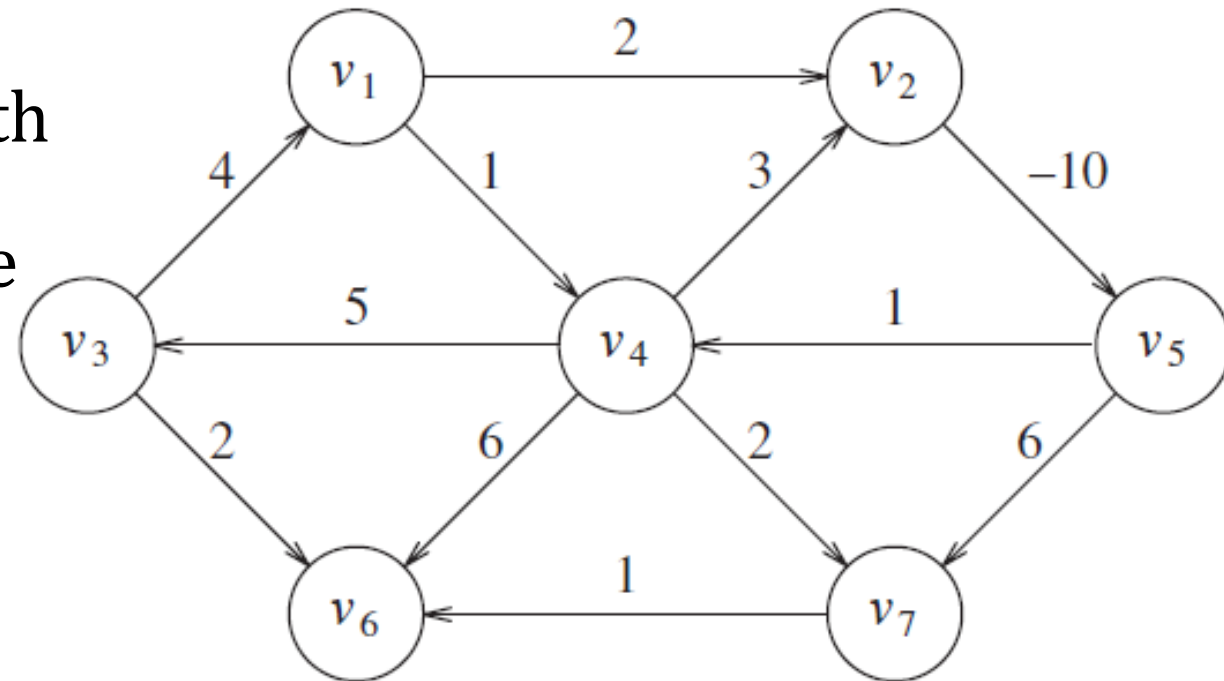
- Having negative weights in the graph may cause some problems.
- The path from  $v_5$  to  $v_4$  has cost 1, but a shorter path exists by following the loop  $v_5, v_4, v_2, v_5, v_4$ , which has cost  $-5$
- This path is still not the shortest, because we could stay in the loop arbitrarily long.
- Thus, the shortest path between these two points is **undefined**.



**Figure 9.9** A graph with a negative-cost cycle

# Single-Source Shortest-Path Algorithms

- Another example, the shortest path from  $v_1$  to  $v_6$  is undefined, because we can get into the same loop.
- This loop is known as a
- **negative-cost cycle**; when one is present in the graph, the shortest paths are not defined.



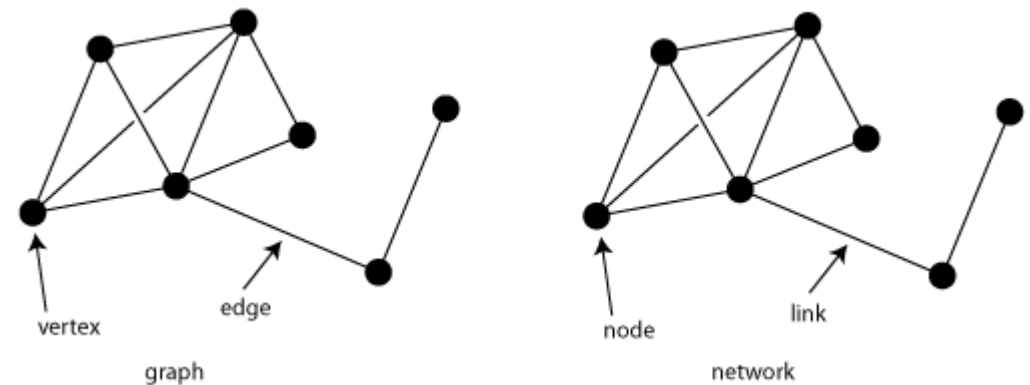
**Figure 9.9** A graph with a negative-cost cycle

# Single-Source Shortest-Path Algorithms

- Negative-cost edges are not necessarily bad, as the cycles are, but their presence seems to make the problem harder.
- For convenience, in the absence of a negative-cost cycle, the shortest path from  $s$  to  $s$  is zero.

# Single-Source Shortest-Path Algorithms

- There are many examples where we might want to solve the shortest-path problem.
- If the vertices represent computers; the edges represent a link between computers; and the costs represent communication costs (phone bill per megabyte of data), delay costs (number of seconds required to transmit a megabyte), or a combination of these and other factors, then we can use the shortest-path algorithm to find the cheapest way to send electronic news from one computer to a set of other computers.

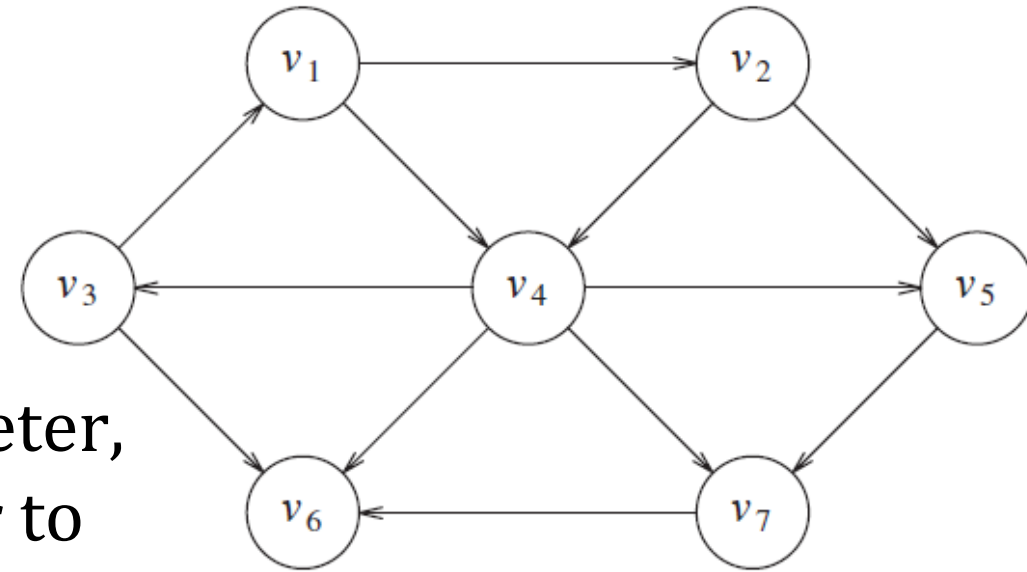


# Single-Source Shortest-Path Algorithms

- Another example is to model an airplane (or transportation routes) by graphs and use a shortest path algorithm to compute the best route between two points.
- In this and many practical applications, we might want to find the shortest path from one vertex,  $s$ , to only one other vertex,  $t$ .
- Currently there are no algorithms in which finding the path from  $s$  to one vertex is any faster (by more than a constant factor) than finding the path from  $s$  to all vertices.
- We will solve 4 variations of this problem

# Unweighted Shortest Paths

- Given an unweighted graph,  $G$ . Using some vertex,  $s$ , which is an input parameter, we want to find the shortest path from  $s$  to all other vertices.
- We are only interested in the number of edges contained on the path (because there are no weights).
- This is clearly a special case of the weighted shortest-path problem, since we could assign all edges a weight of 1.

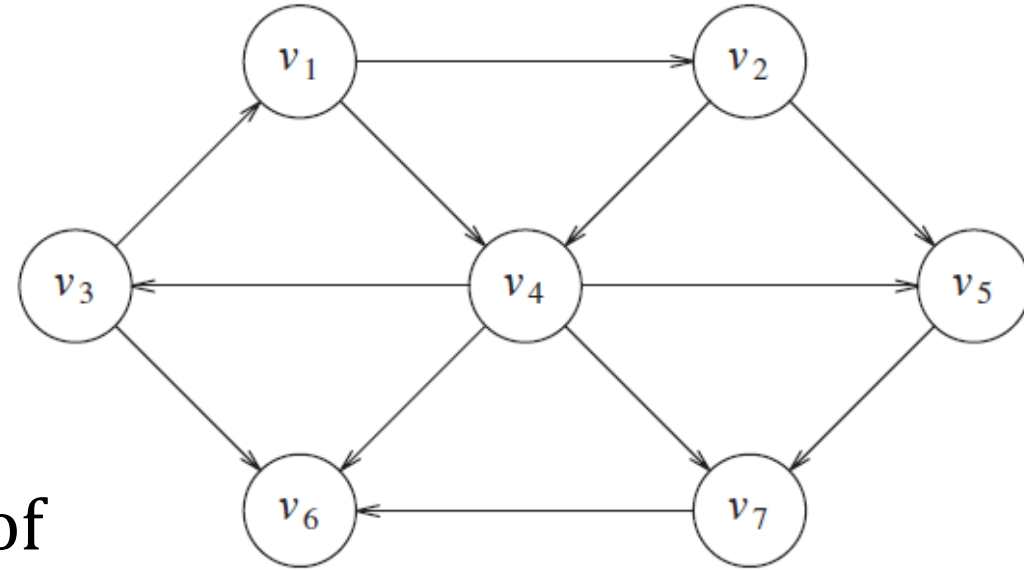


**Figure 9.10** An unweighted directed graph  $G$



# Unweighted Shortest Paths

- Suppose we are interested in the length of the shortest path not in the
- actual paths themselves. Keeping track of the actual paths will turn out to be a matter of simple bookkeeping.



**Figure 9.10** An unweighted directed graph  $G$

# Weighted Shortest-Path

# Breadth-First Search (BFS)

# Dijkstra's Algorithm

# Minimum Spanning Tree