

Data Structures & Algorithms 2

Directed Graphs

Lecturer: Dr. Hadi Tabatabaee

Materials: Dr. Hadi Tabatabaee

Maynooth University

Online at <http://moodle.maynoothuniversity.ie>

Overview

Aims

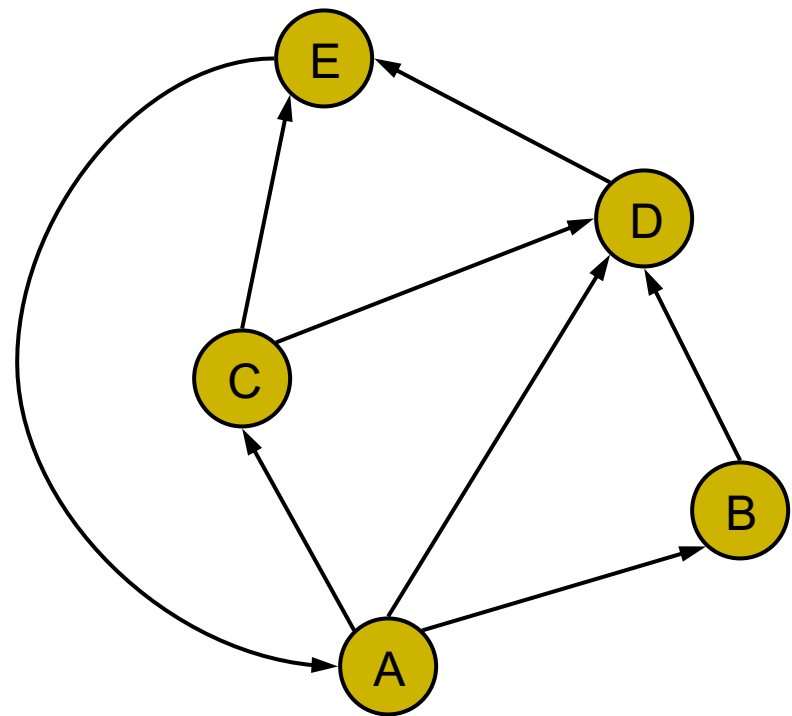
- Introduce Transitive Closure in a directed graph.
- Introduce Directed Acyclic Graphs.

Learning outcomes: You should be able to...

- Use Floyd-Warshall Algorithm to investigate the reachability of vertices in a directed graph.
- Use topological ordering in a directed acyclic graph.

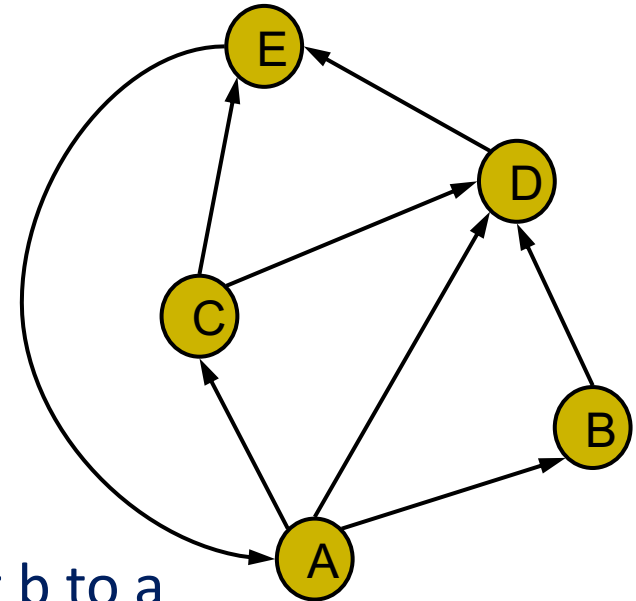
Digraphs

- A **digraph** is a graph whose edges are all directed
 - Short for “directed graph”
- Applications
 - one-way streets
 - flights
 - task scheduling



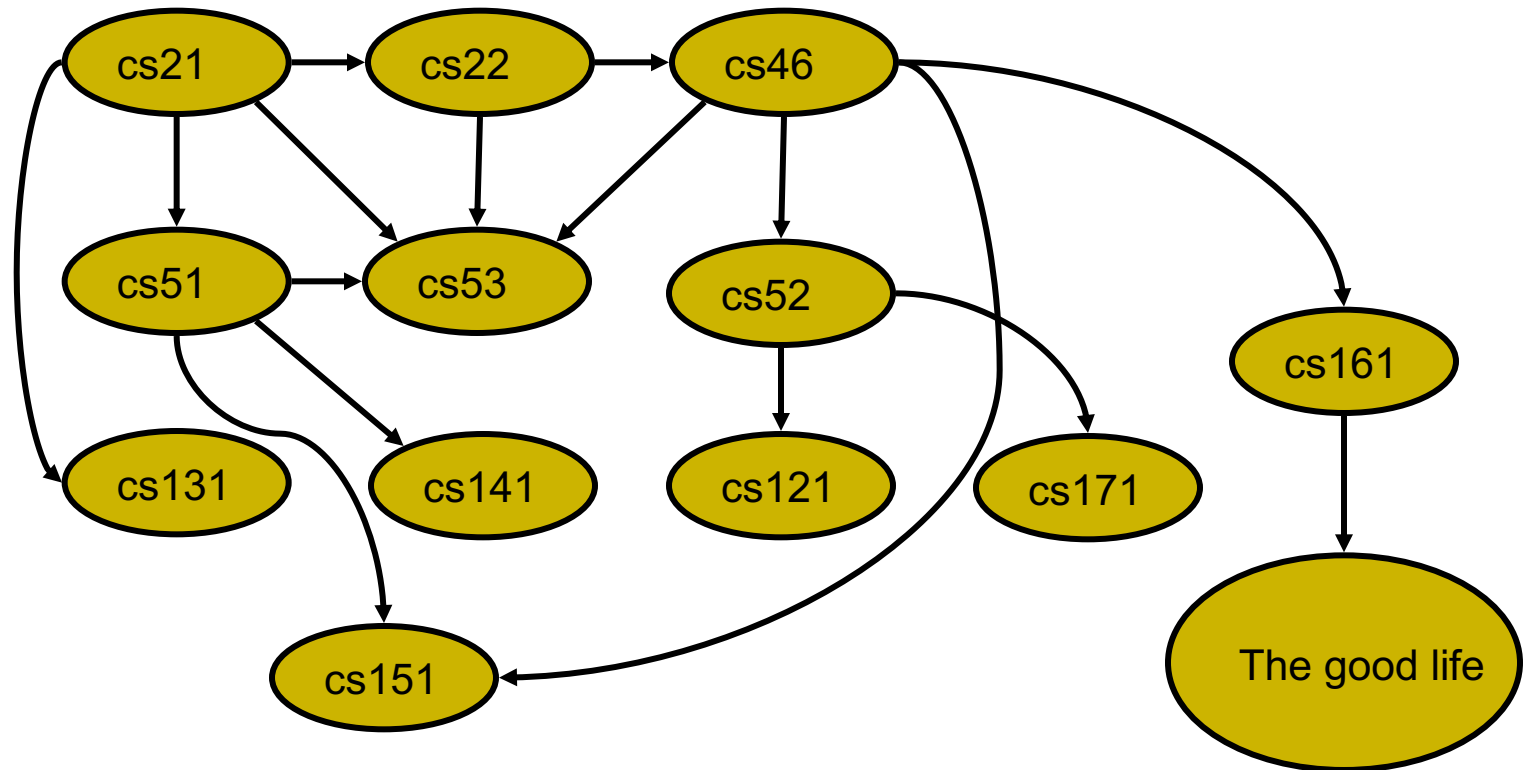
Digraph Properties

- A graph $G=(V,E)$ such that
 - Each edge goes in **one direction**:
Edge (a,b) goes from a to b , but not b to a
- If G is simple, $m \leq n \cdot (n - 1)$
- If we keep in-edges and out-edges in separate adjacency lists, we can perform a listing of incoming edges and outgoing edges in time proportional to their size



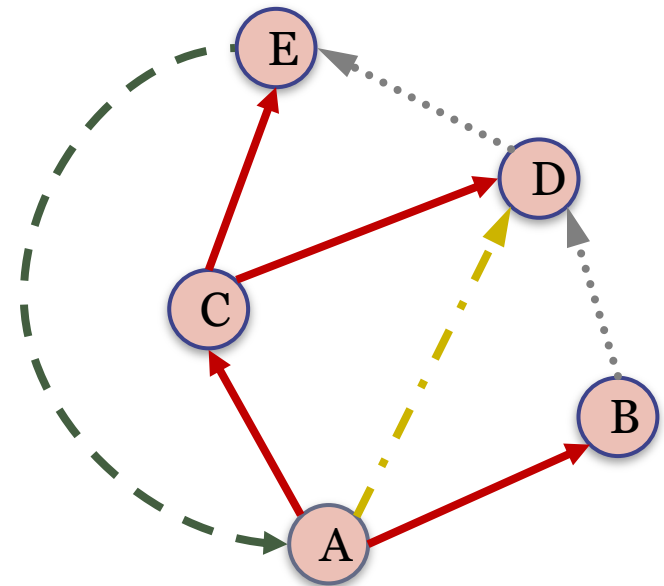
Digraph Application

Scheduling: edge (a,b) means task **a** must be completed before **b** can be started.



Directed DFS

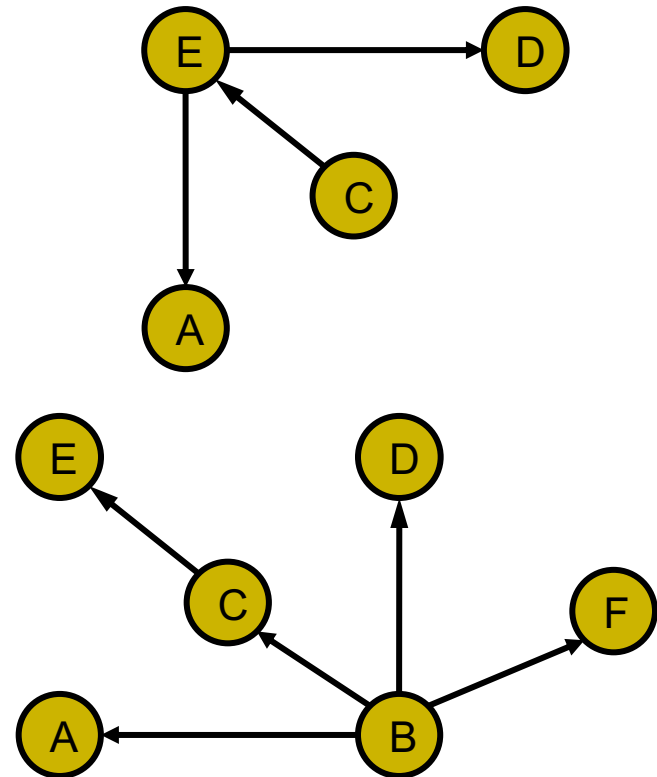
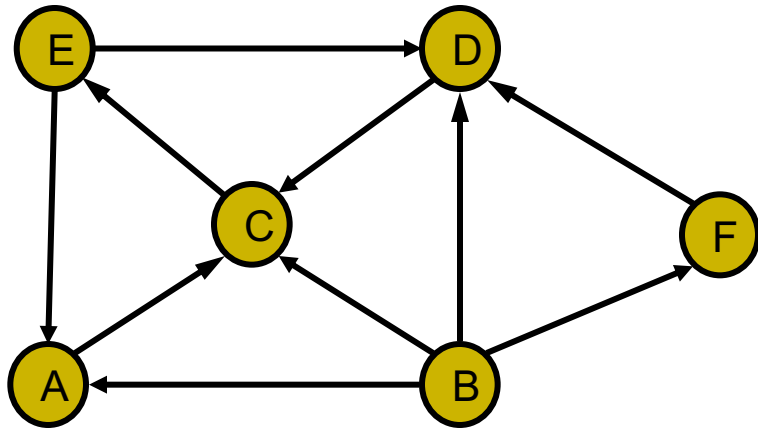
- We can specialize the traversal algorithms (DFS and BFS) to digraphs by traversing edges only along their direction.
- In the directed DFS algorithm, we have four types of edges
 - **discovery edges**: discovers a new vertex (unvisited)
 - **back edges**: connect a vertex to an ancestor in the DFS tree
 - **forward edges**: which connect a vertex to a descendant in the DFS tree
 - **cross edges**: connect a vertex to a vertex that is neither its ancestor nor its descendant
- A directed DFS starting at a vertex s determines the vertices **reachable** from s .



Reachability



DFS tree rooted at **v**: vertices reachable from **v** via directed paths.

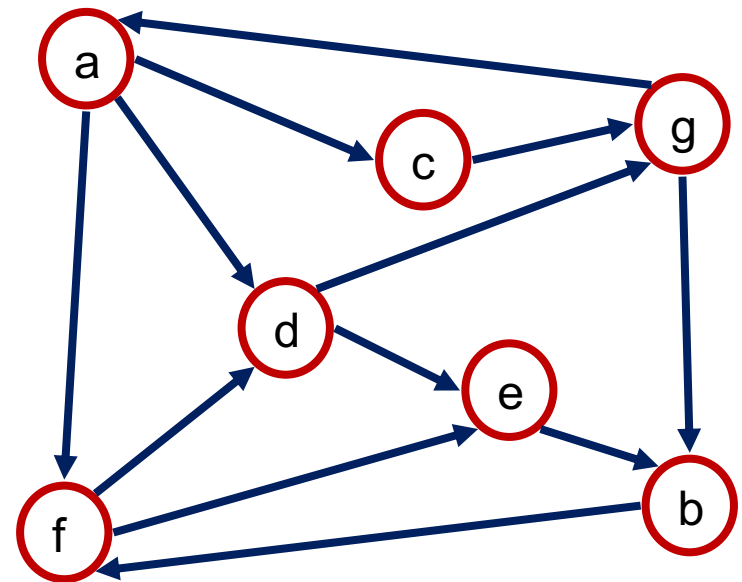


Strong Connectivity

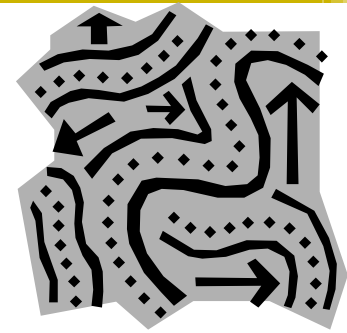
Each vertex can reach all other vertices



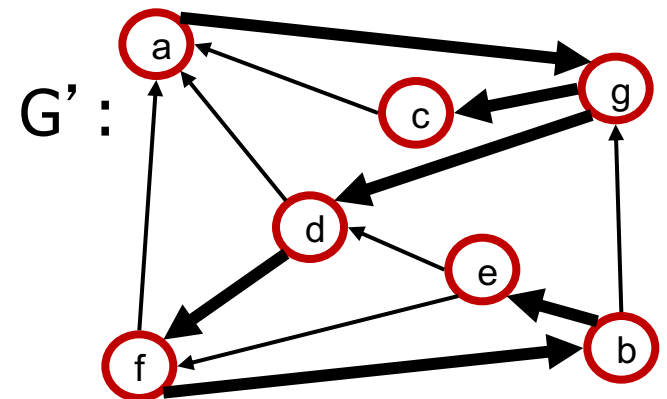
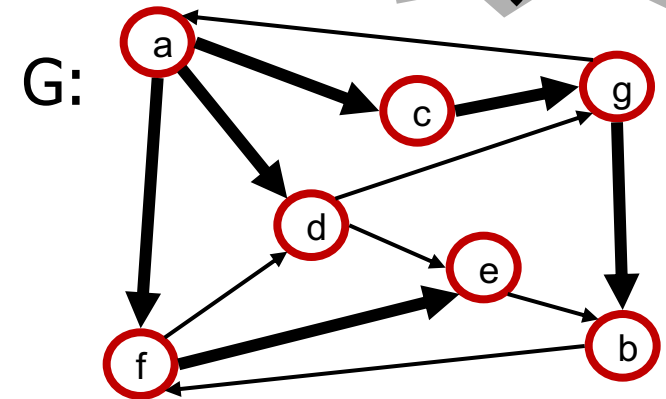
If we start an independent call to DFS from each vertex, we could determine whether this was the case, but those n calls when combined would run in $O(n(n+m))$. However, we can determine if G is strongly connected much faster than this, requiring only two depth-first searches.



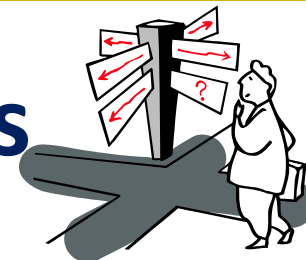
Strong Connectivity Algorithm



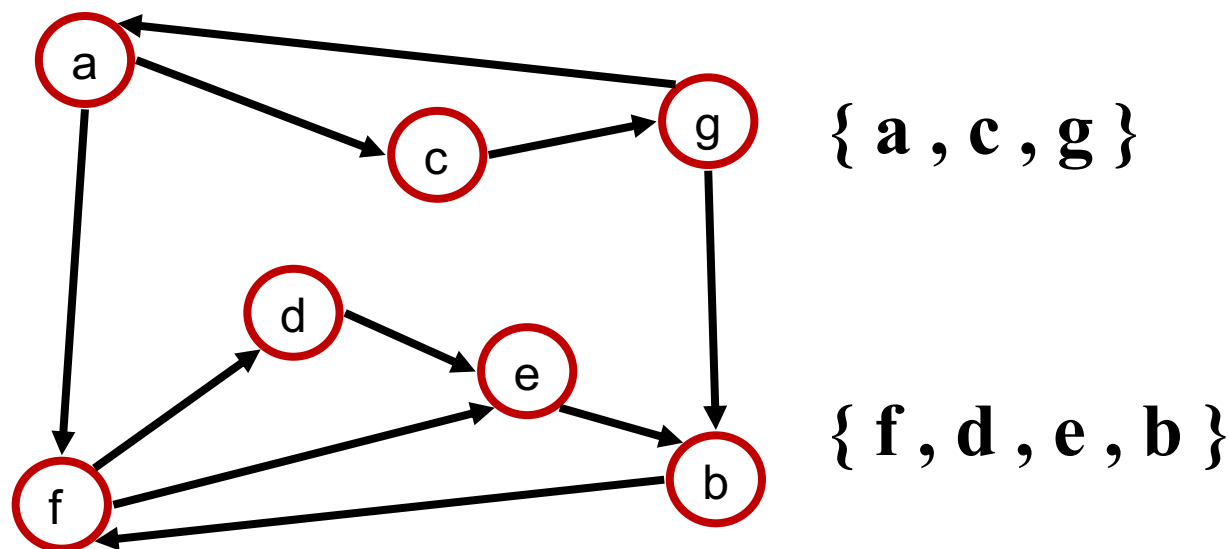
- Pick a vertex v in G
- Perform a DFS from v in G
 - If there's a w not visited, print "no"
- Let G' be G with edges reversed
- Perform a DFS from v in G'
 - If there's a w not visited, print "no"
 - Else, print "yes"
- Running time: $O(n+m)$



Strongly Connected Components



- Maximal subgraphs such that each vertex can reach all other vertices in the subgraph.
- It can also be done in $O(n+m)$ time using DFS, but it is more complicated (similar to biconnectivity).



Transitive Closure

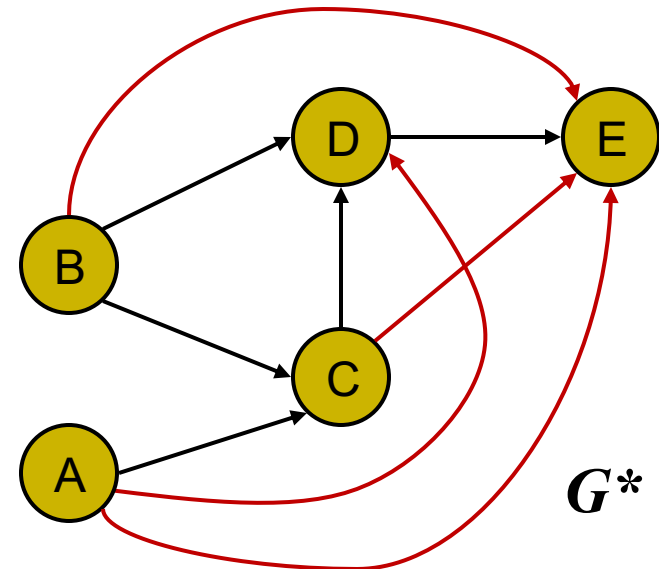
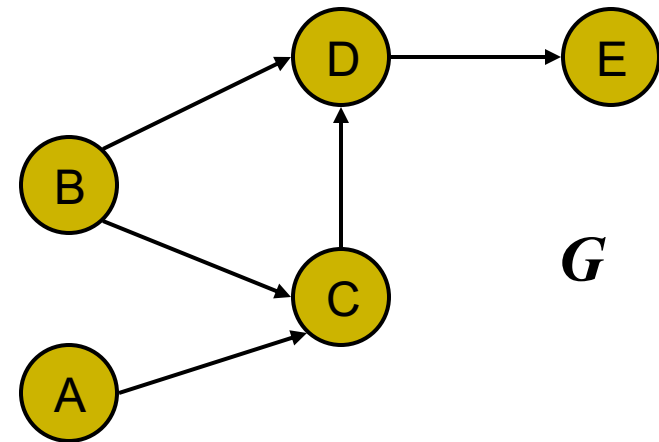


Transitive Closure

Given a digraph G , the transitive closure of G is the digraph G^* such that

- G^* has the same vertices as G
- if G has a directed path from u to v ($u \neq v$), G^* has a directed edge from u to v

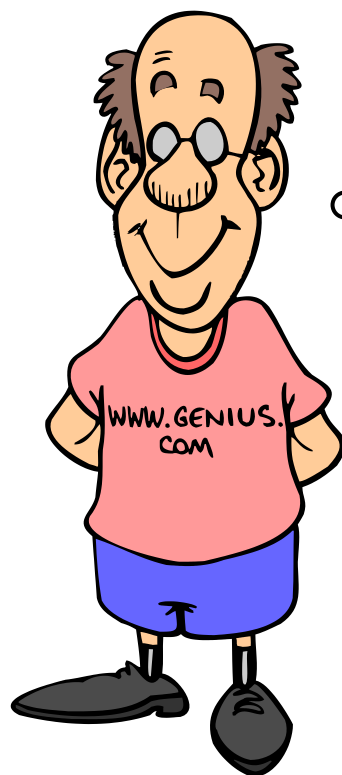
The transitive closure provides reachability information about a digraph



Computing the Transitive Closure

- We can perform DFS starting at each vertex
 - $O(n(n+m))$

If there's a way to get from **A** to **B** and from **B** to **C**, then there's a way to get from **A** to **C**.

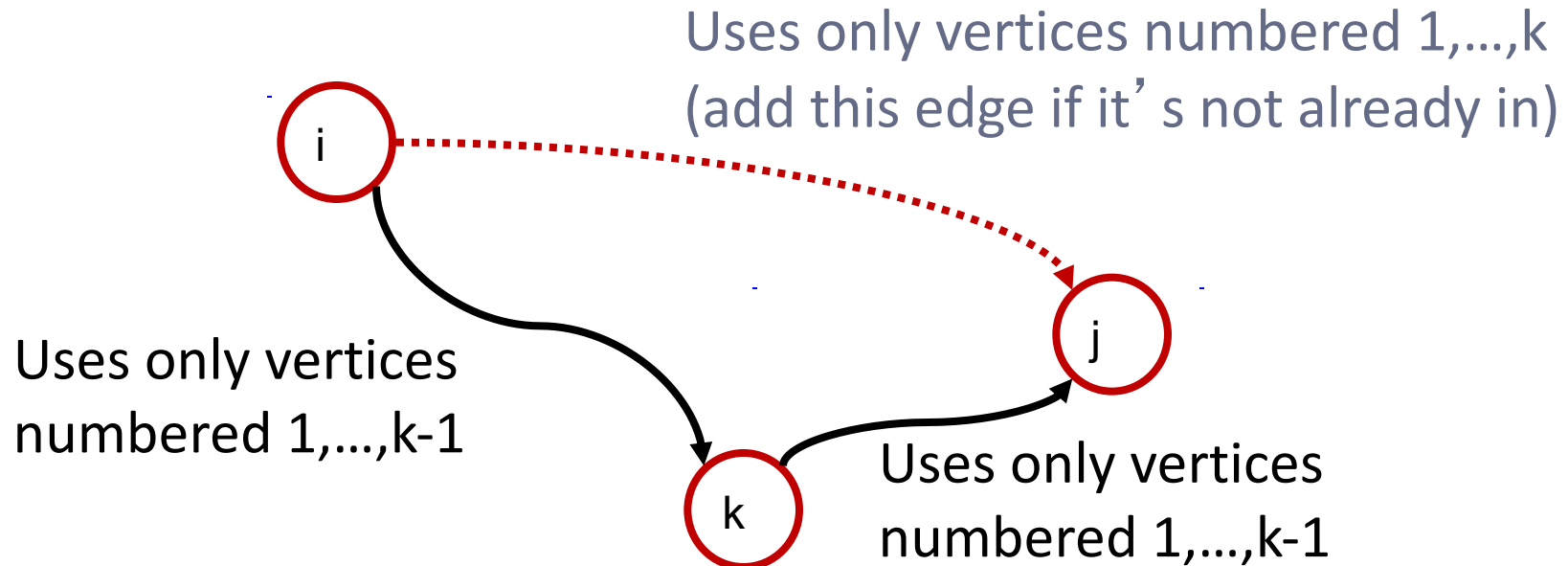
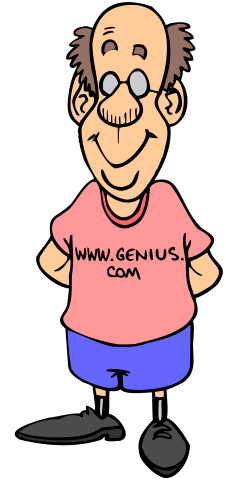


Alternatively ... Use dynamic programming:

The Floyd-Warshall Algorithm

Floyd-Warshall Transitive Closure

- Idea #1: Number the vertices $1, 2, \dots, n$.
- Idea #2: Consider paths that use only vertices numbered $1, 2, \dots, k$, as intermediate vertices:



Floyd-Warshall's Algorithm



- Number vertices v_1, \dots, v_n
- Compute digraphs G_0, \dots, G_n
 - $G_0 = G$
 - G_k has directed edge (v_i, v_j) if G has a directed path from v_i to v_j with intermediate vertices in $\{v_1, \dots, v_k\}$
- We have that $G_n = G^*$
- In phase k , digraph G_k is computed from G_{k-1}
- Running time: $O(n^3)$, assuming areAdjacent is $O(1)$ (e.g., adjacency matrix)

Algorithm *FloydWarshall*(G)

Input digraph G

Output transitive closure G^* of G

$i \leftarrow 1$

for all $v \in G.\text{vertices}()$

denote v as v_i

$i \leftarrow i + 1$

$G_0 \leftarrow G$

for $k \leftarrow 1$ **to** n **do**

$G_k \leftarrow G_{k-1}$

for $i \leftarrow 1$ **to** n ($i \neq k$) **do**

for $j \leftarrow 1$ **to** n ($j \neq i, k$) **do**

if $G_{k-1}.\text{areAdjacent}(v_i, v_k) \wedge$
 $G_{k-1}.\text{areAdjacent}(v_k, v_j)$

if $\neg G_k.\text{areAdjacent}(v_i, v_j)$

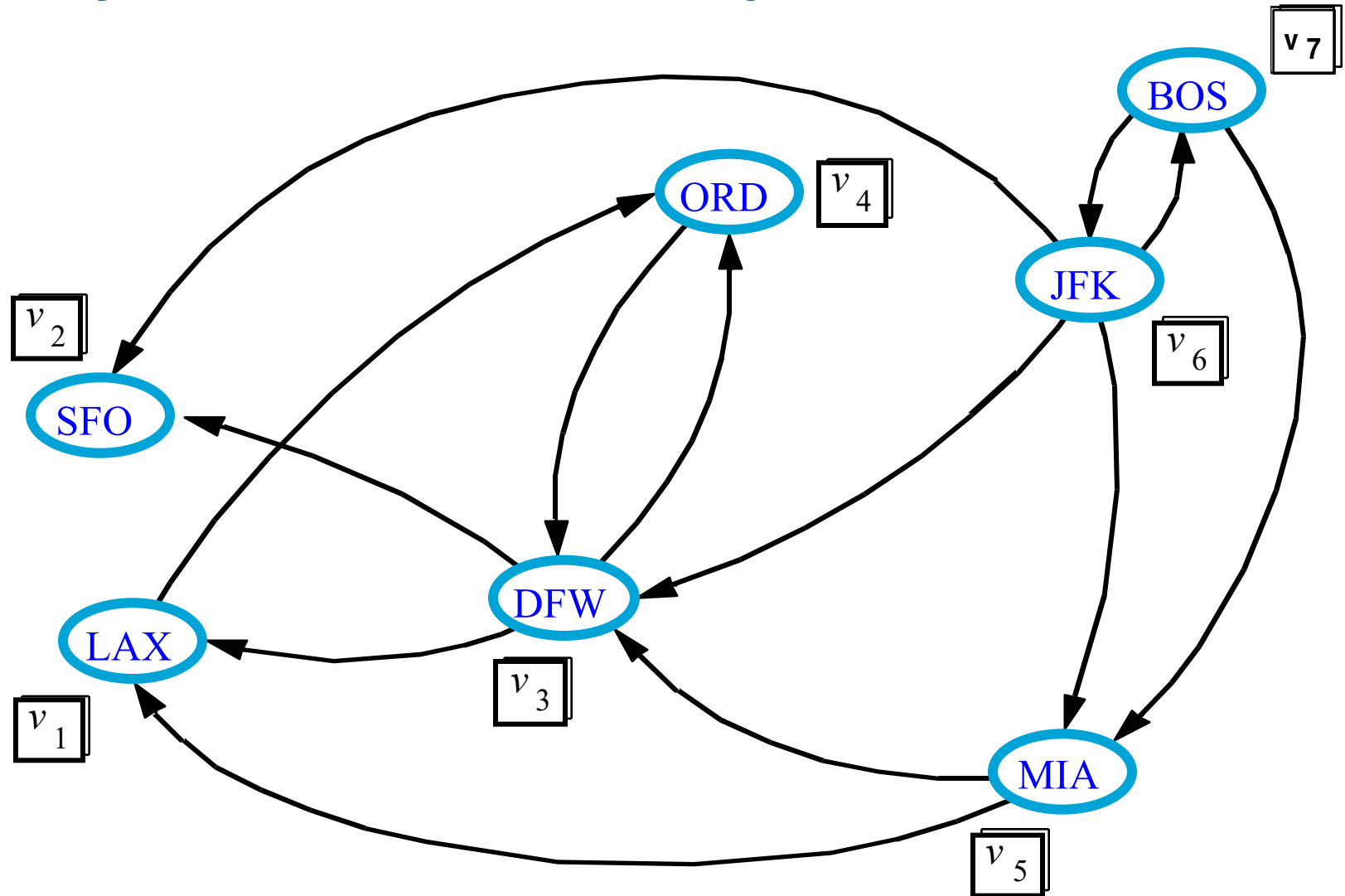
$G_k.\text{insertDirectedEdge}(v_i, v_j, k)$

return G_n

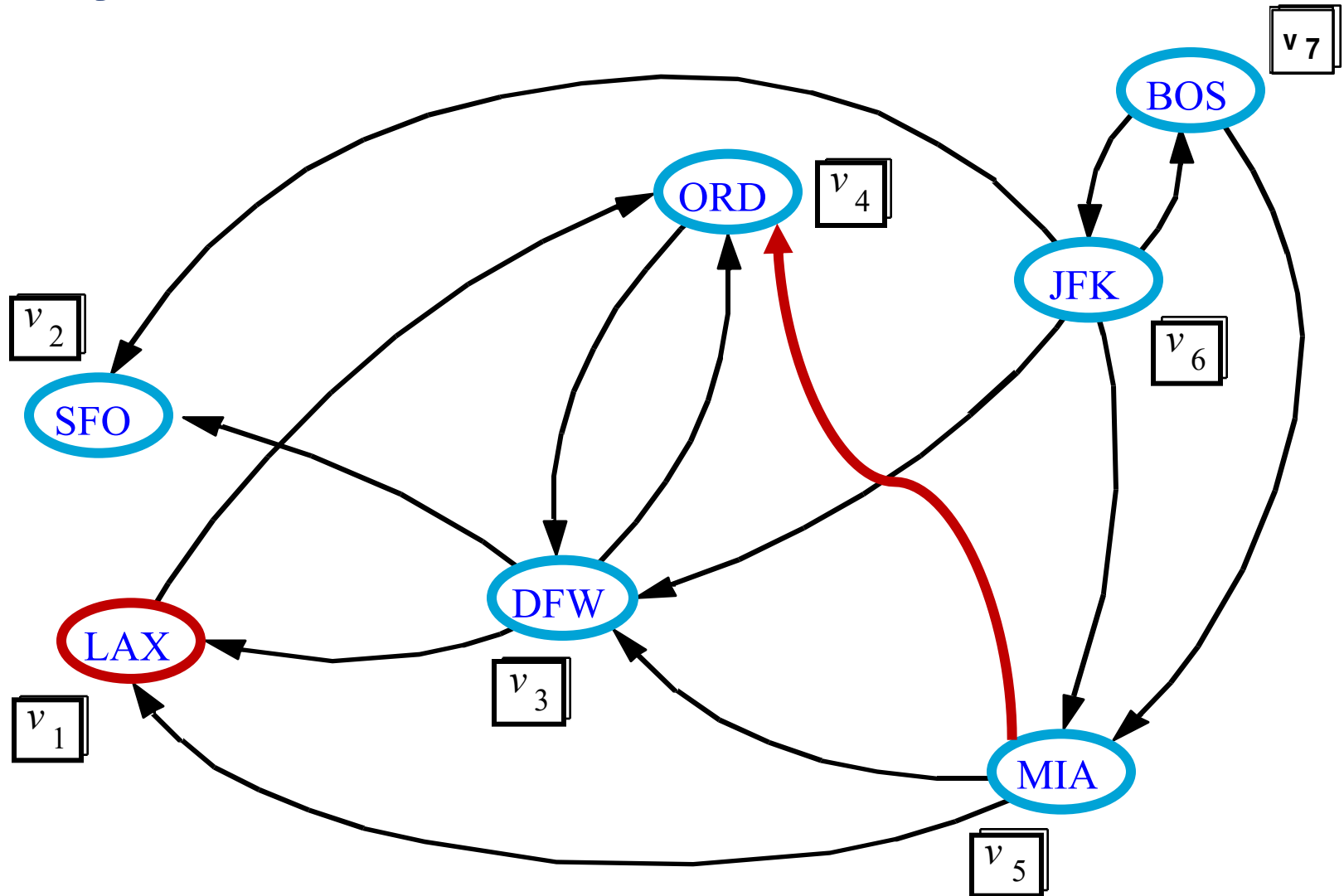
Java Implementation

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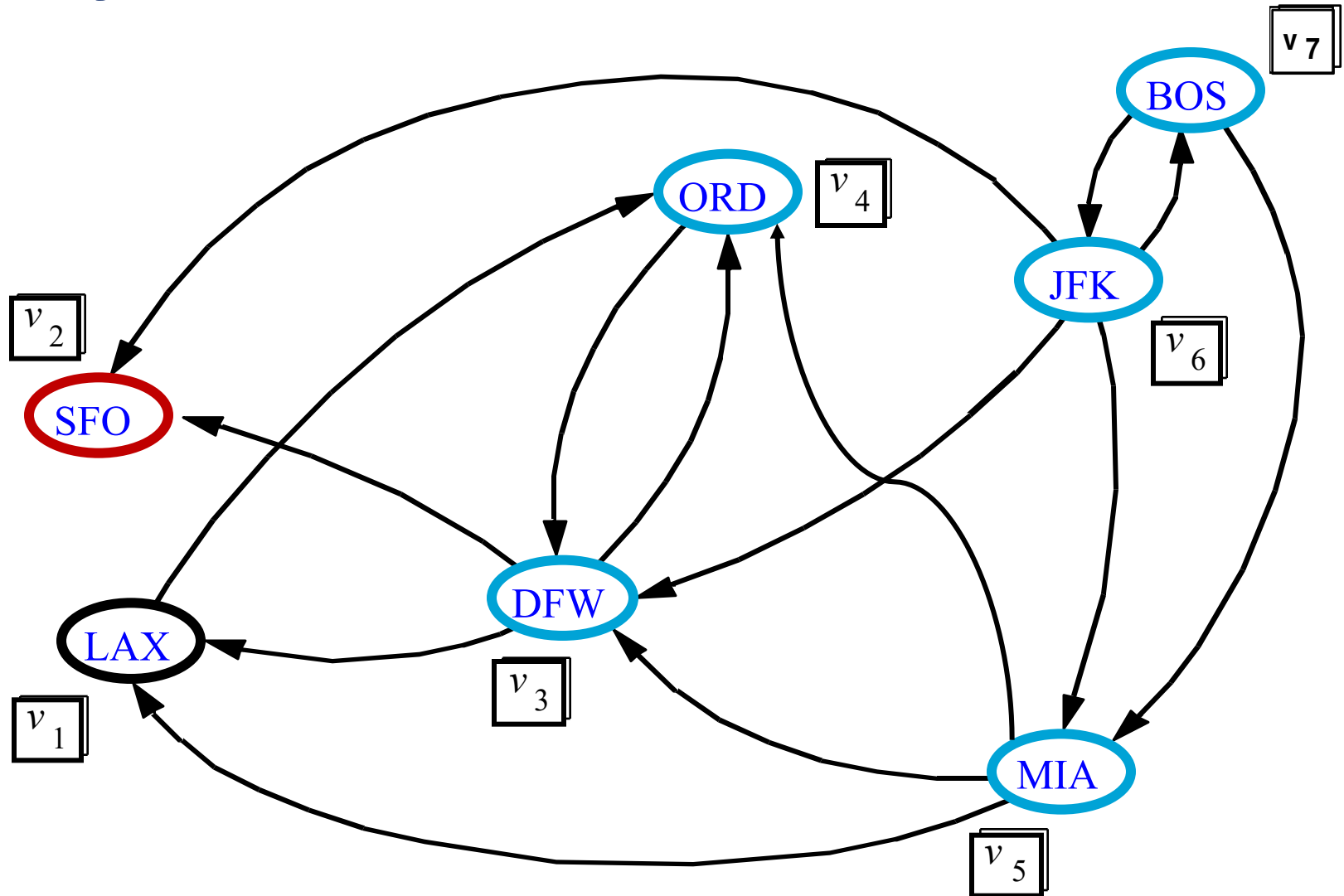
Floyd-Warshall Example



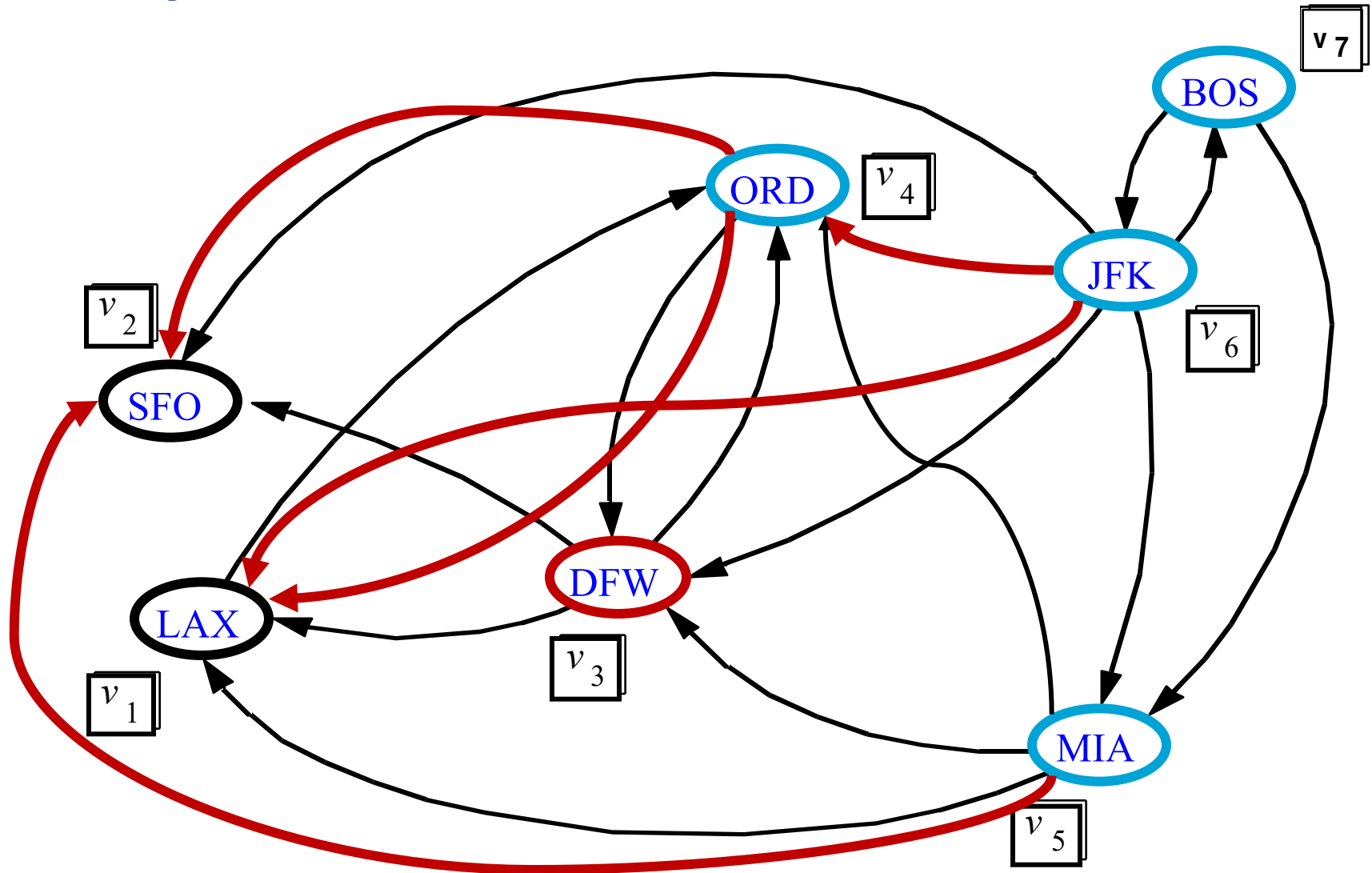
Floyd-Warshall, Iteration 1



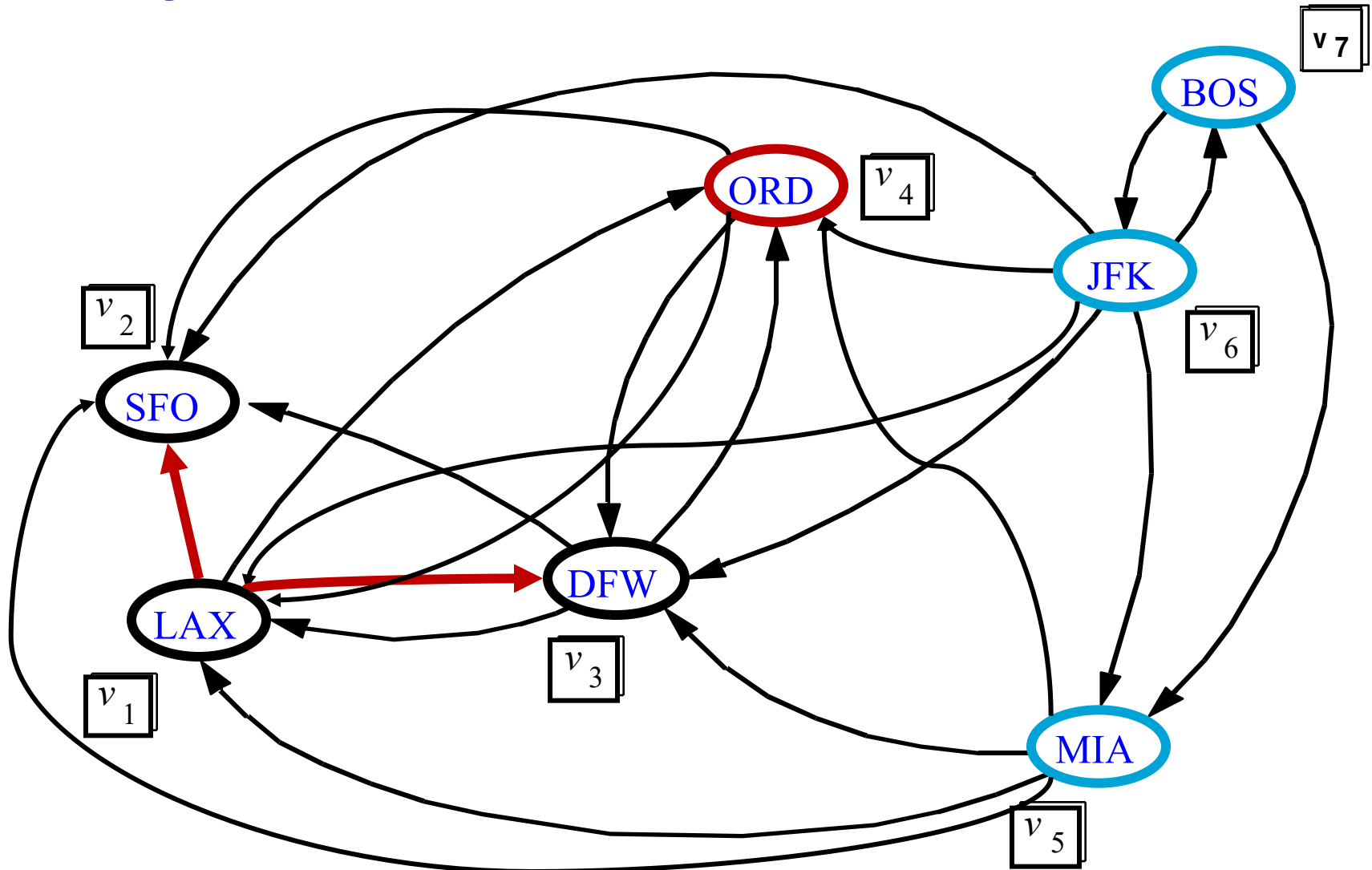
Floyd-Warshall, Iteration 2



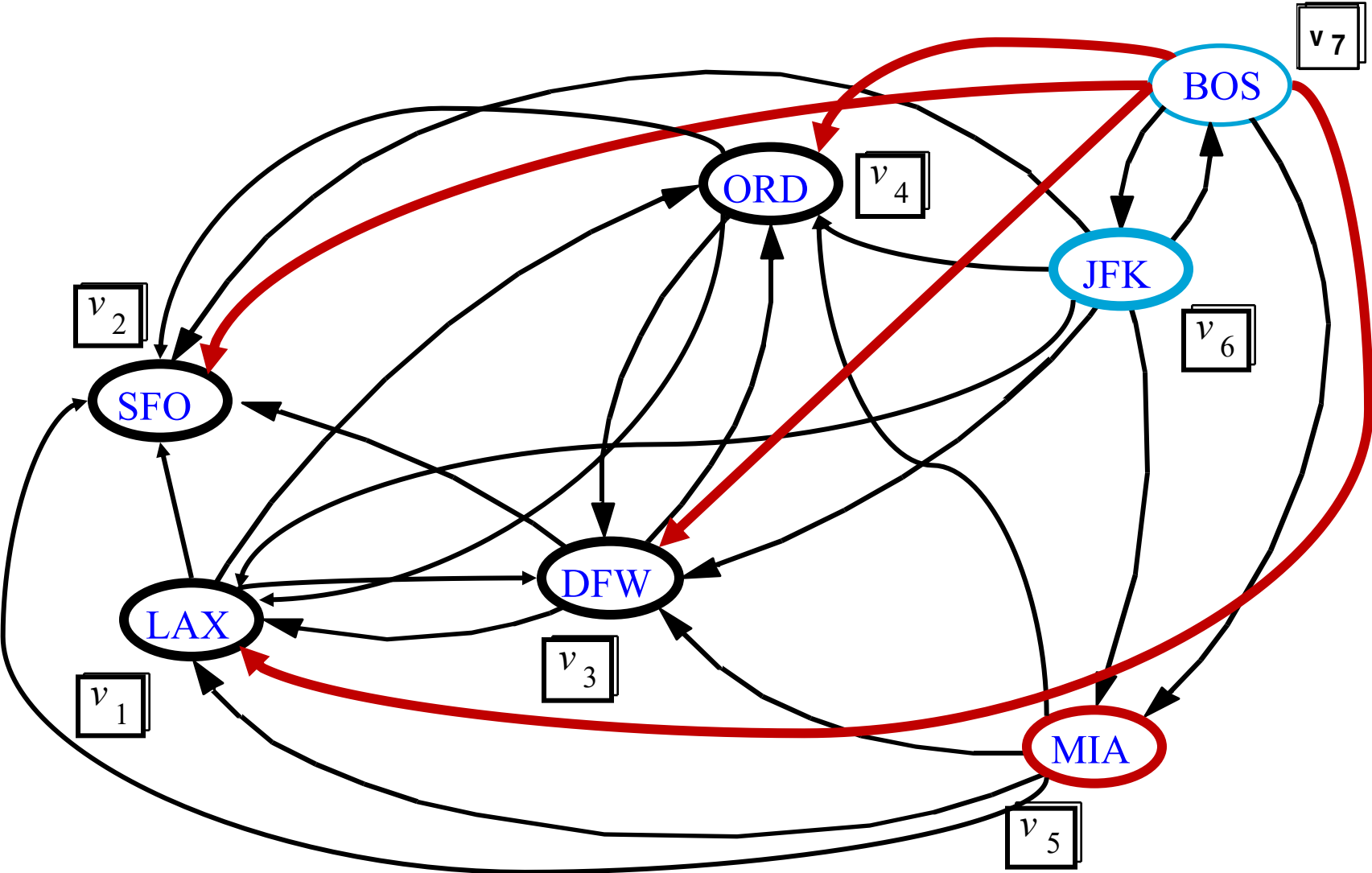
Floyd-Warshall, Iteration 3



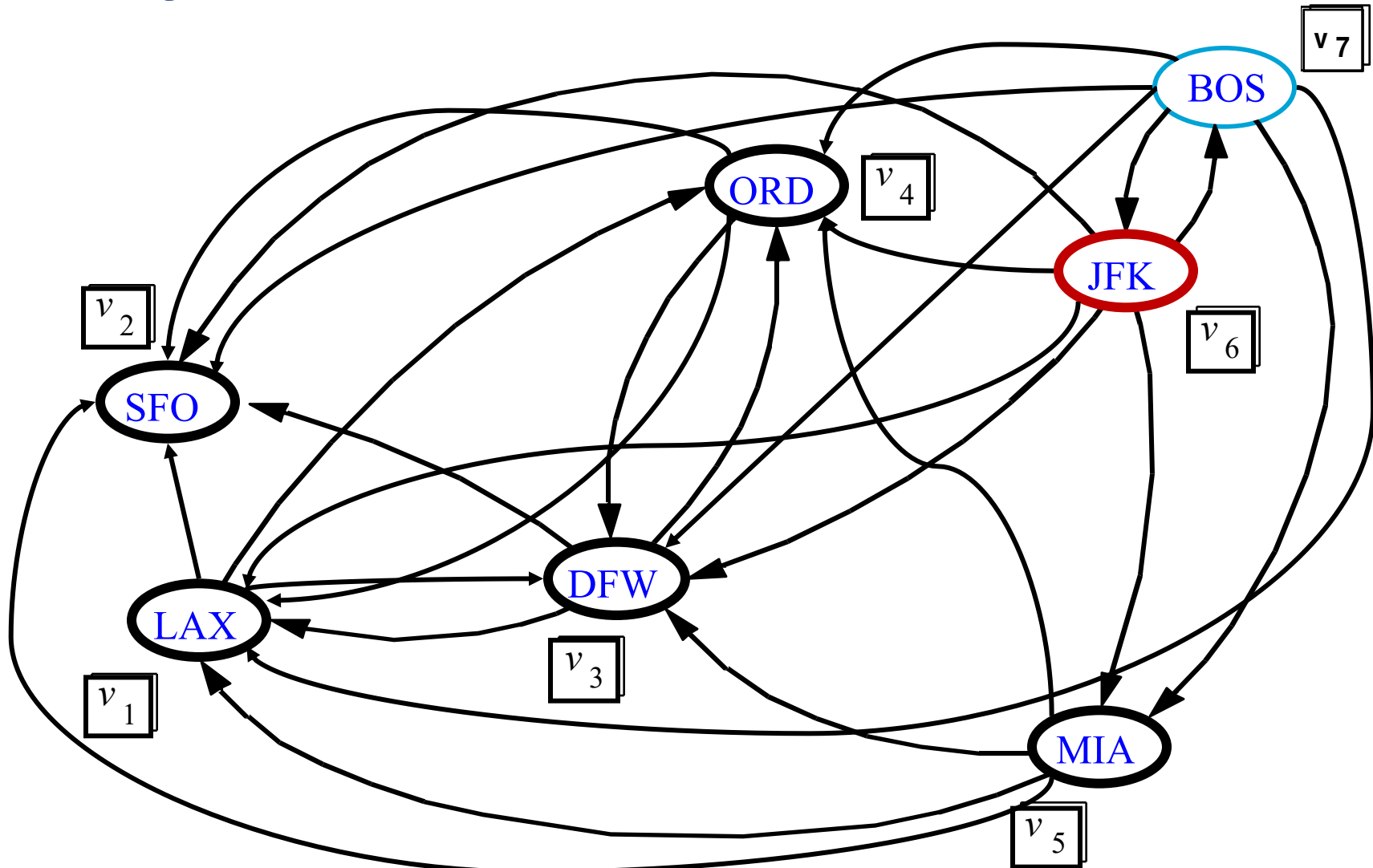
Floyd-Warshall, Iteration 4



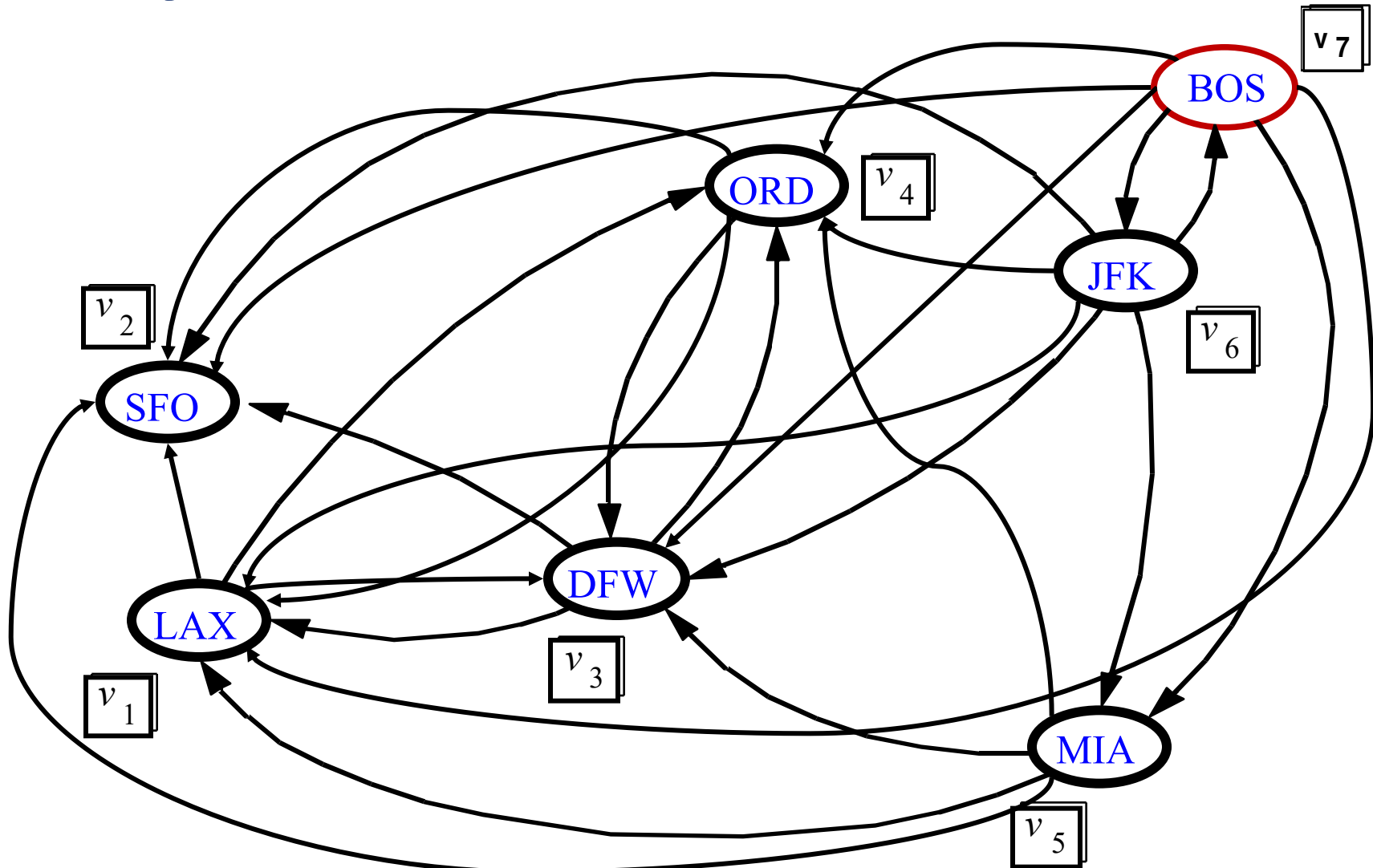
Floyd-Warshall, Iteration 5



Floyd-Warshall, Iteration 6



Floyd-Warshall, Conclusion



Directed Acyclic Graphs (DAG)

A decorative graphic consisting of several horizontal bars. A thick yellow bar spans the width of the slide. Below it, on the right side, are several thinner white and yellow bars of varying lengths, creating a stepped, modern look.

DAGs and Topological Ordering

- A directed acyclic graph (DAG) is a digraph that has no directed cycles
- A topological ordering of a digraph is a numbering

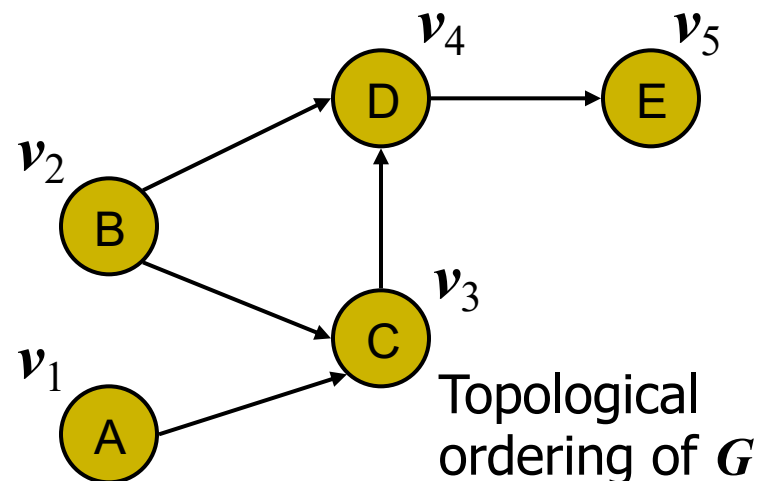
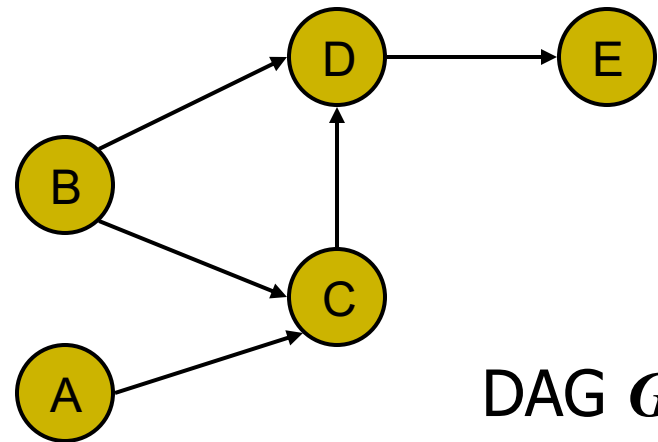
$$v_1, \dots, v_n$$

of the vertices such that for every edge (v_i, v_j) , we have $i < j$

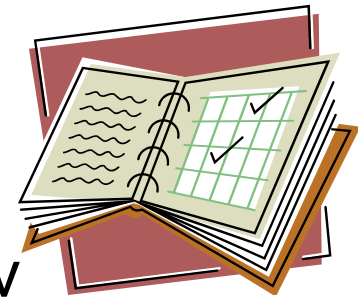
- For example: in a task scheduling digraph, a topological ordering a task sequence that satisfies the precedence constraints

Theorem

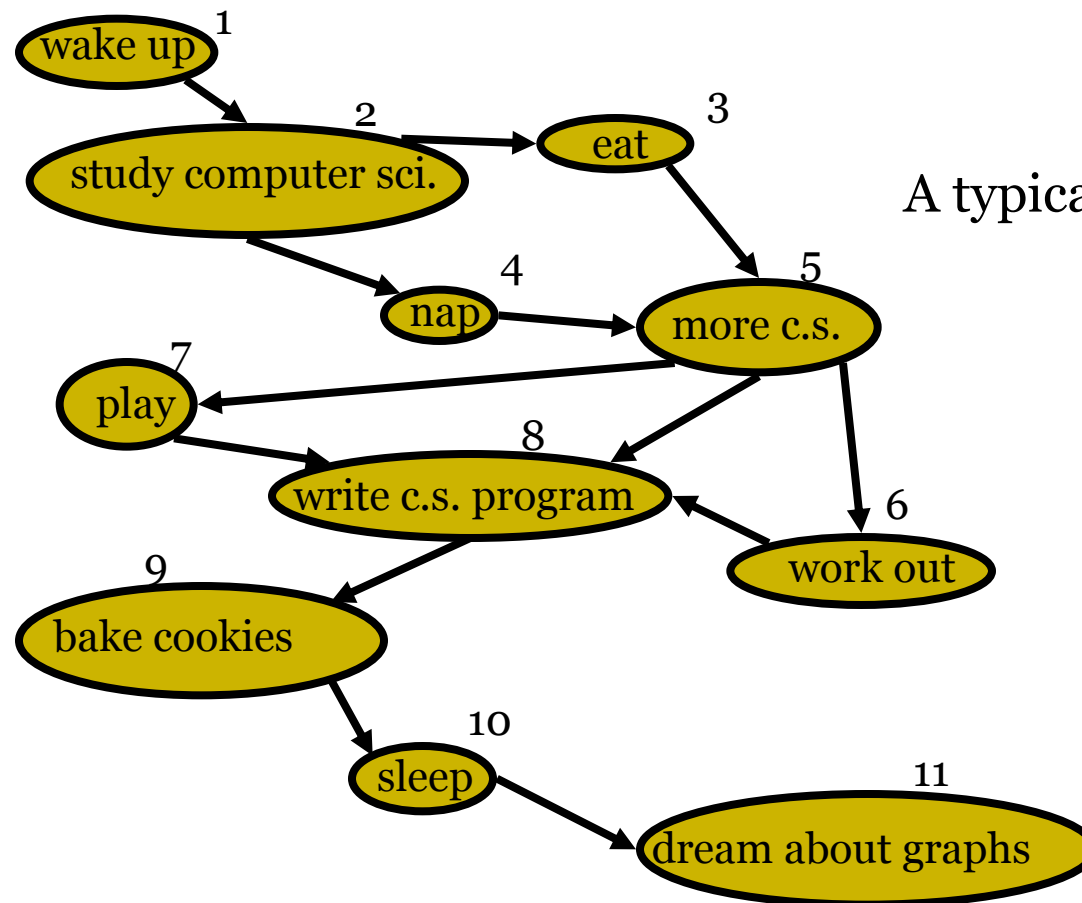
A digraph admits a topological ordering if and only if it is a DAG



Topological Sorting



Number vertices, so that (u,v) in E implies $u < v$



A typical student day

Algorithm for Topological Sorting

- Note: This algorithm is different than the one in the book

```
Algorithm TopologicalSort(G)  
  H ← G           // Temporary copy of G  
  n ← G.numVertices()  
  while H is not empty do  
    Let v be a vertex with no outgoing edges  
    Label v ← n  
    n ← n − 1  
    Remove v from H
```

- Running time: $O(n + m)$

Implementation with DFS

- Simulate the algorithm by using depth-first search
- $O(n+m)$ time.

Algorithm *topologicalDFS(G)*

Input dag G

Output topological ordering of G
 $n \leftarrow G.numVertices()$

for all $u \in G.vertices()$

$setLabel(u, UNEXPLORED)$

for all $v \in G.vertices()$

if $getLabel(v) = UNEXPLORED$
 $topologicalDFS(G, v)$

Algorithm *topologicalDFS(G, v)*

Input graph G and a start vertex v of G

Output labeling of the vertices of G
in the connected component of v

$setLabel(v, VISITED)$

for all $e \in G.outEdges(v)$
 { outgoing edges }

$w \leftarrow opposite(v, e)$

if $getLabel(w) = UNEXPLORED$
 { e is a discovery edge }

$topologicalDFS(G, w)$

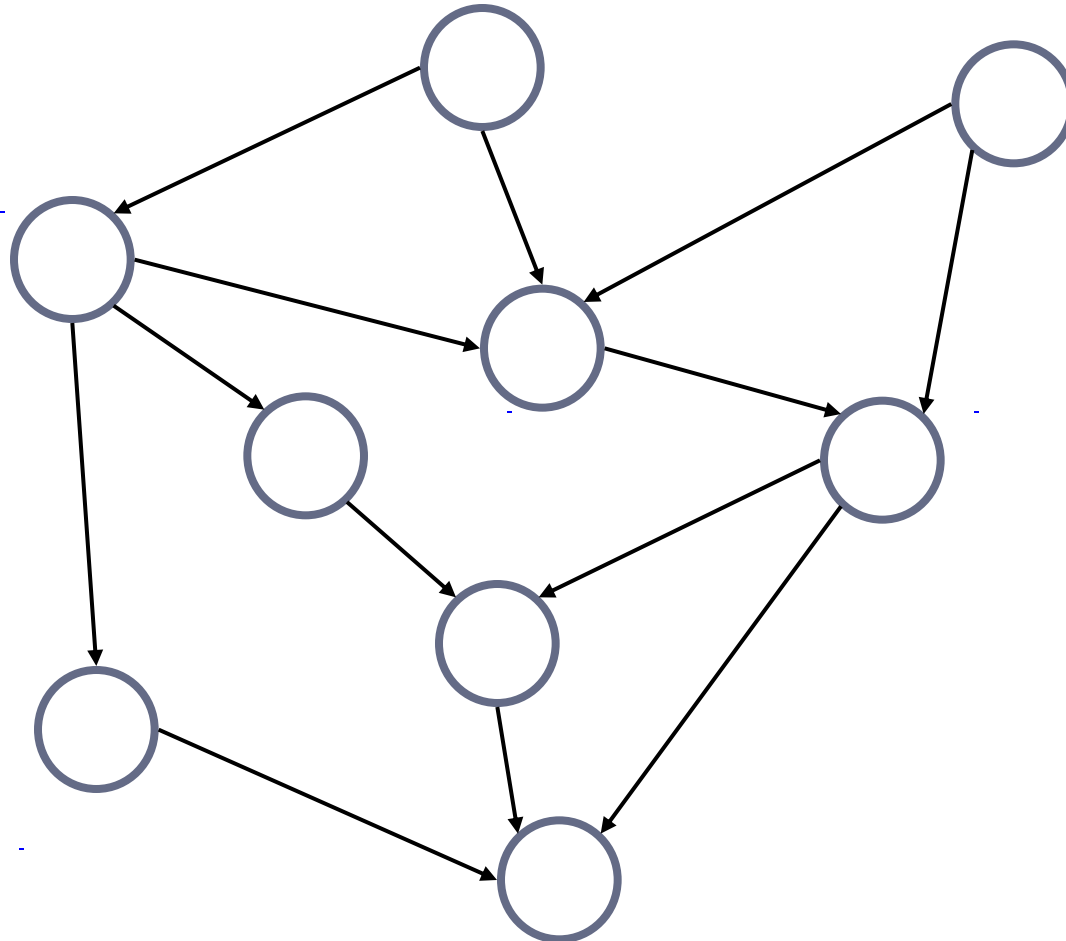
else

 { e is a forward or cross edge }

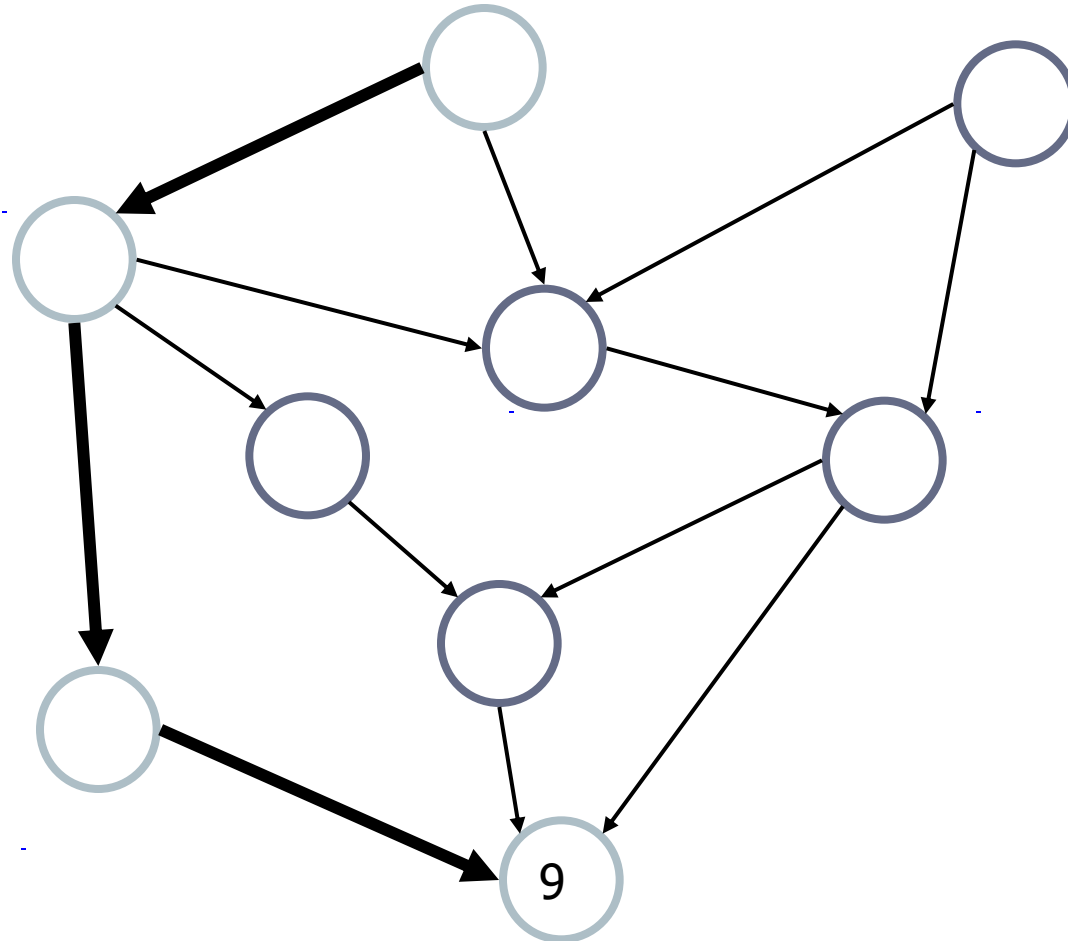
Label v with topological number n

$n \leftarrow n - 1$

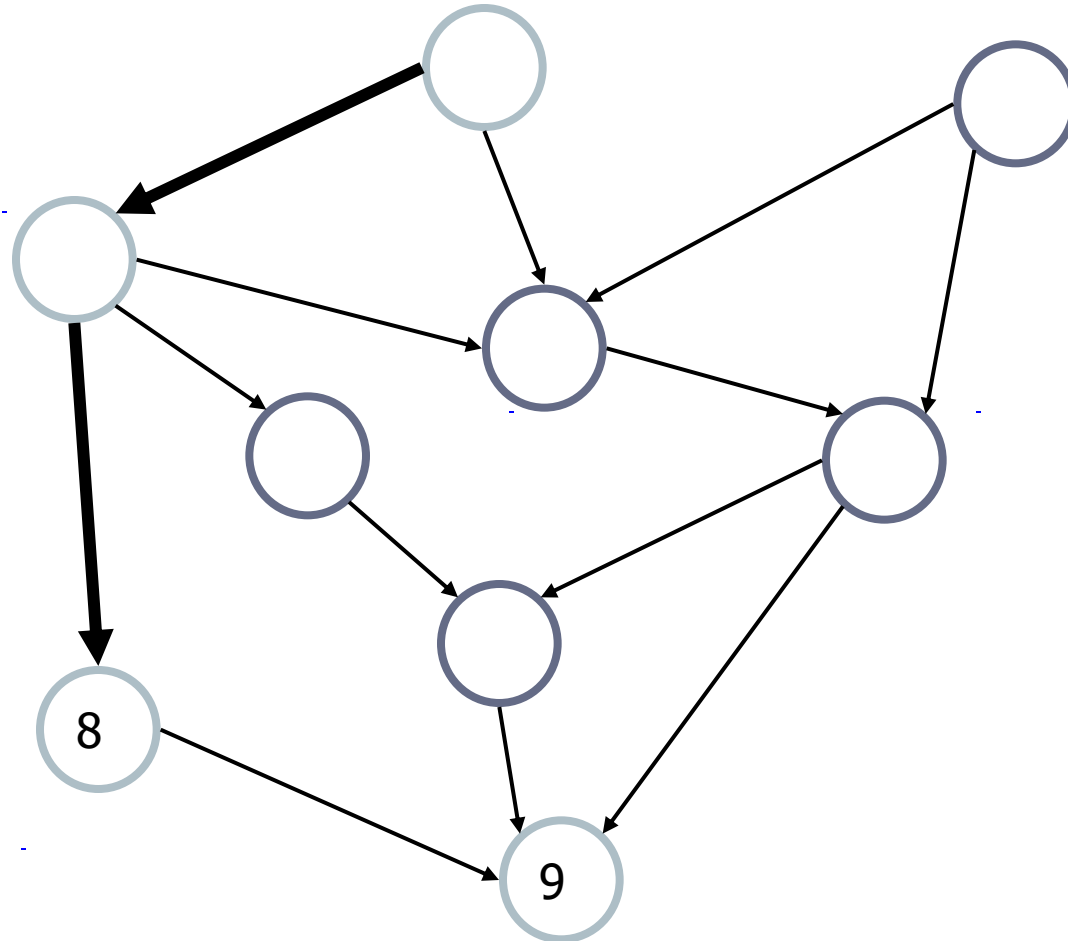
Topological Sorting Example



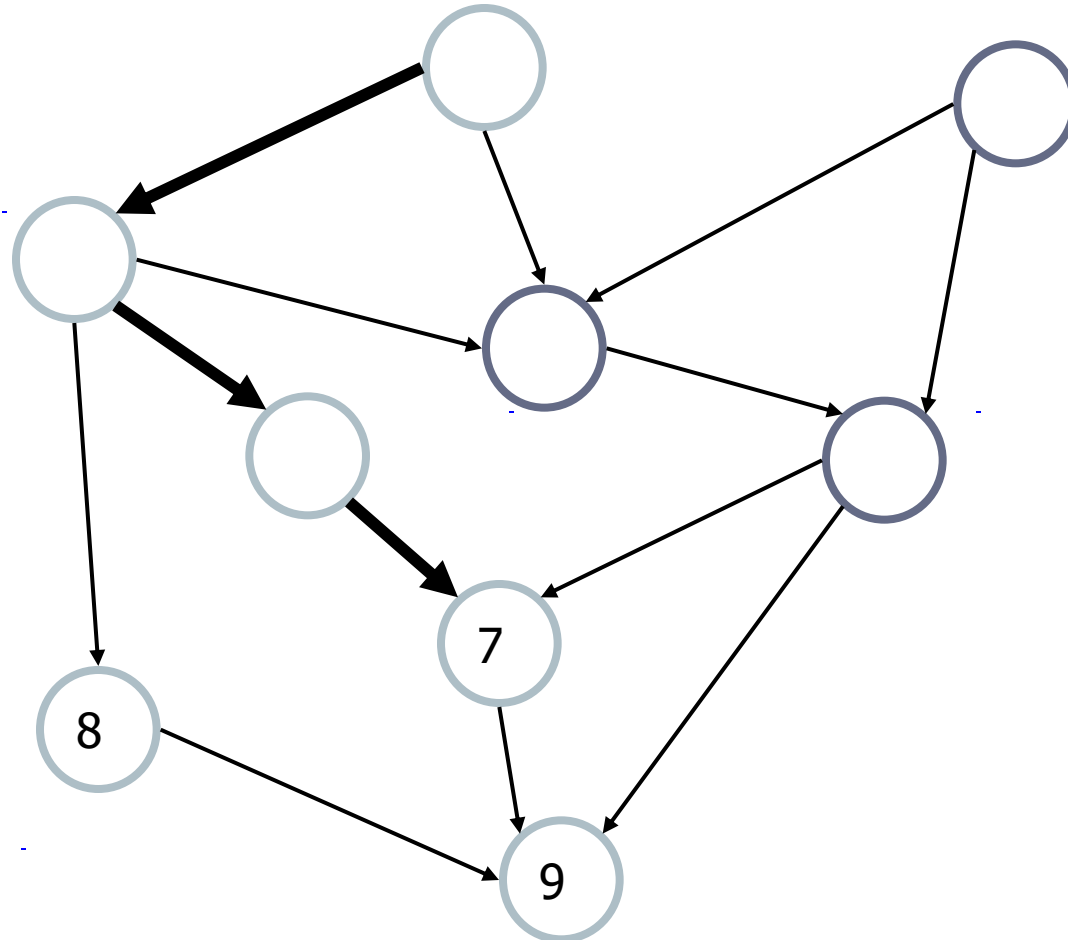
Topological Sorting Example



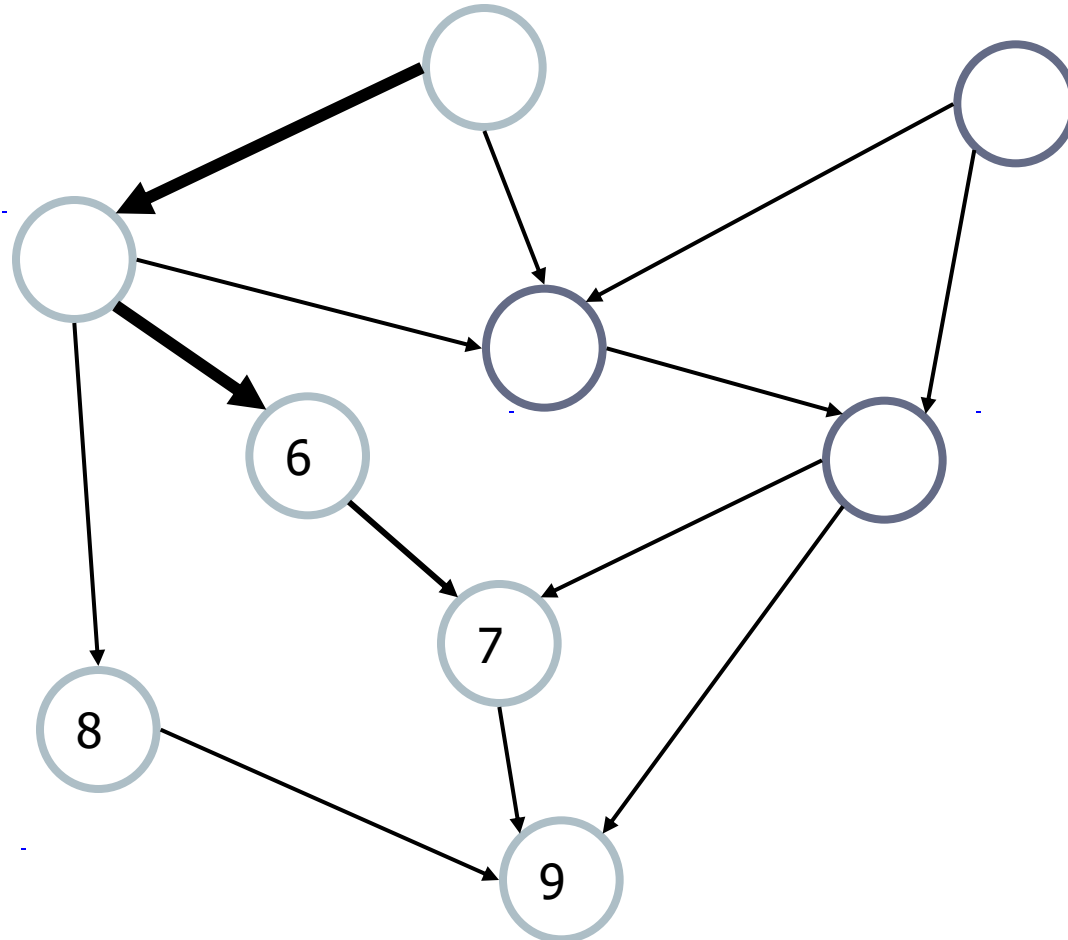
Topological Sorting Example



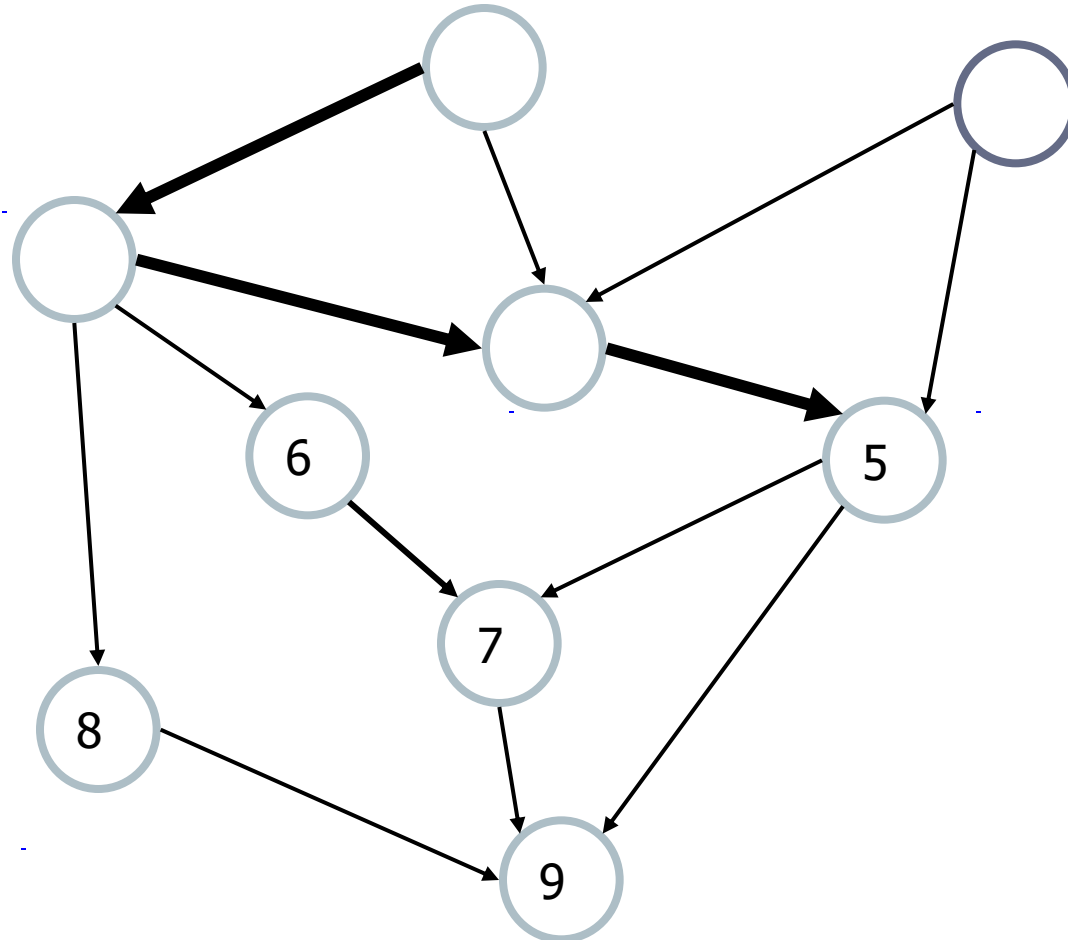
Topological Sorting Example



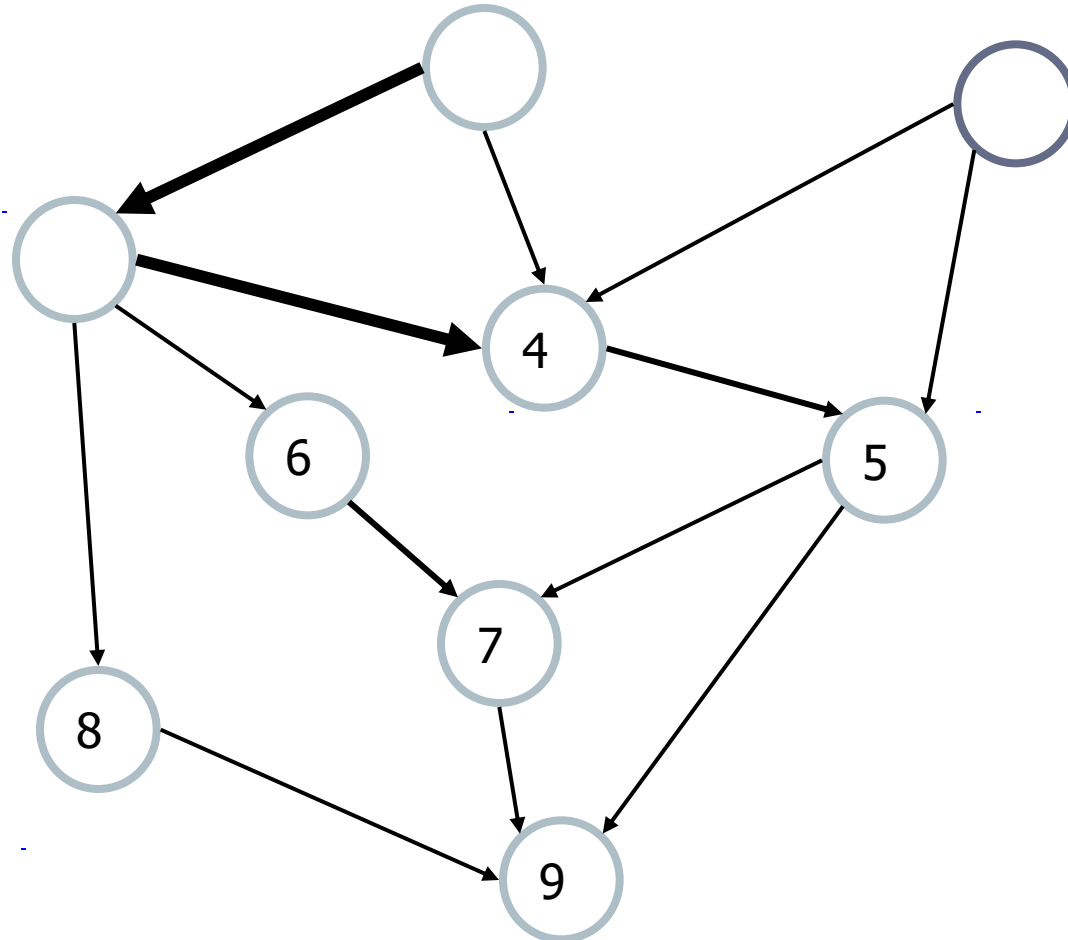
Topological Sorting Example



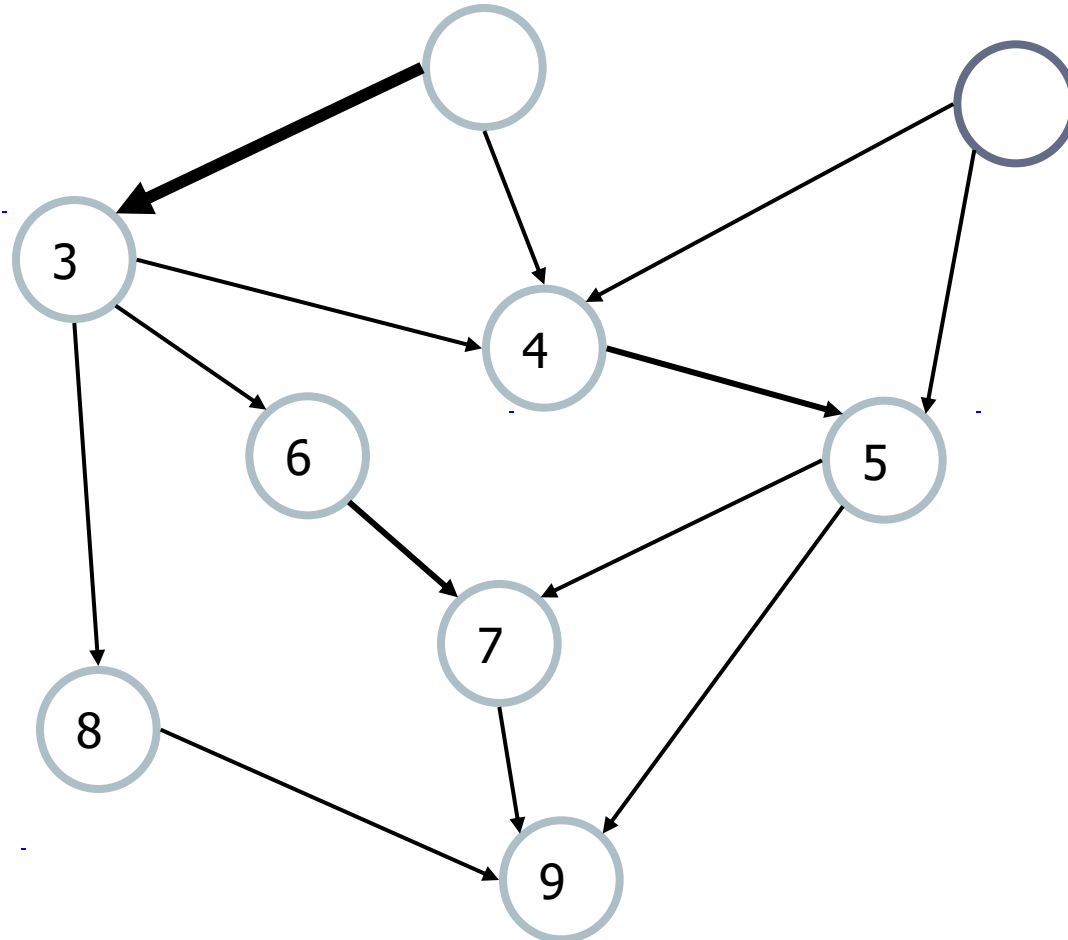
Topological Sorting Example



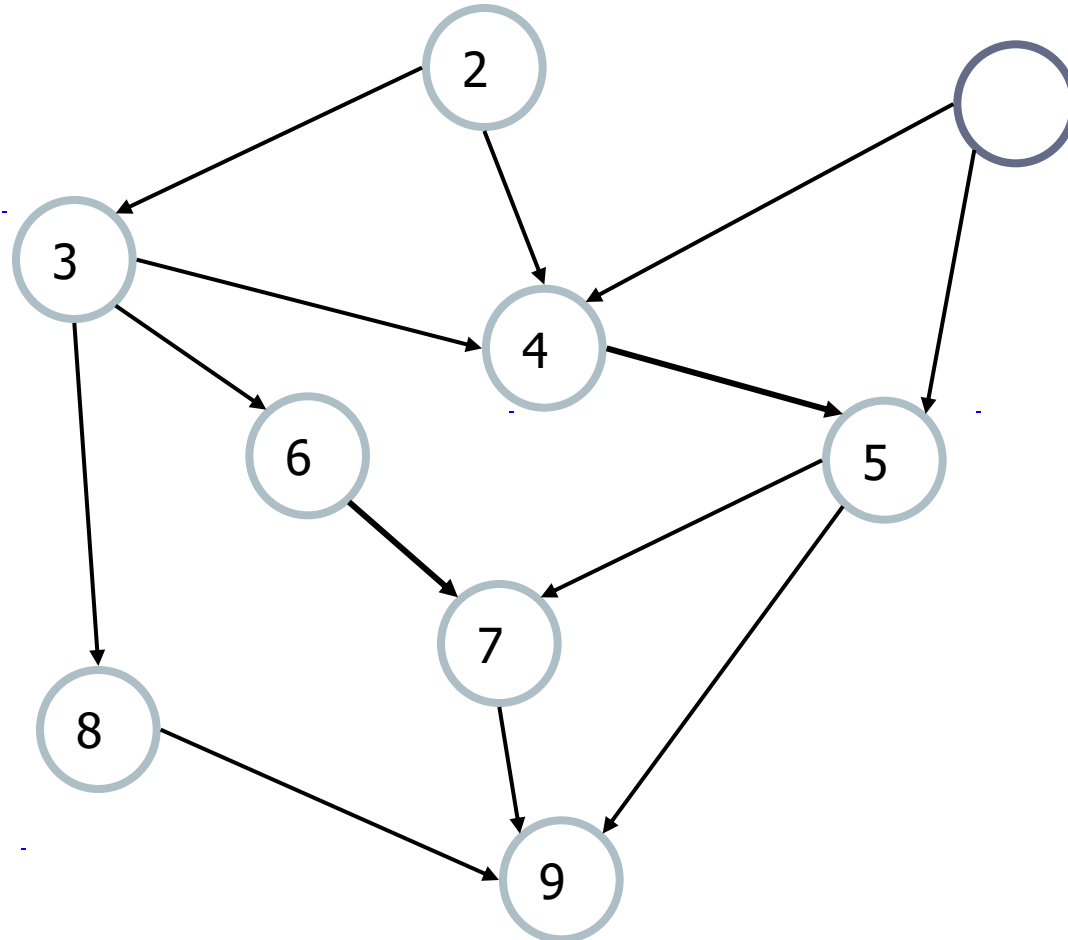
Topological Sorting Example



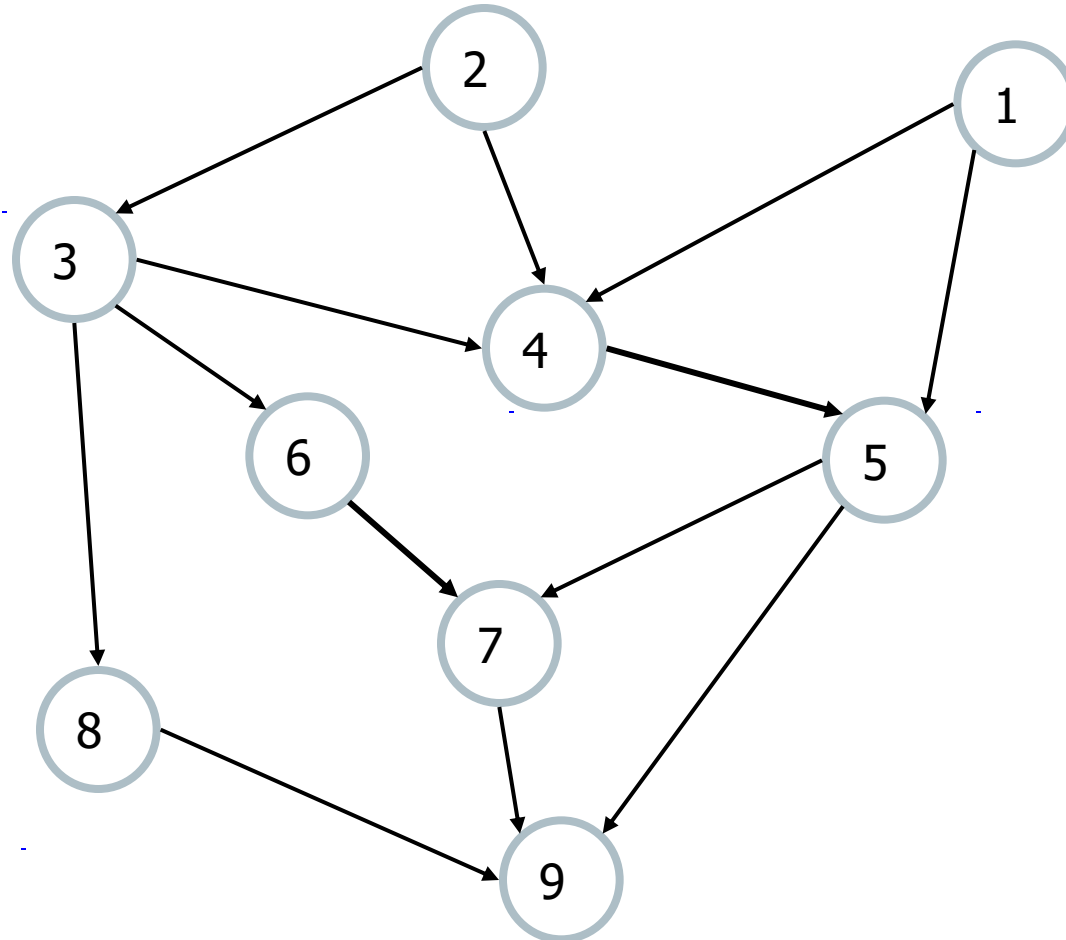
Topological Sorting Example



Topological Sorting Example



Topological Sorting Example



Java Implementation

```

1  /** Returns a list of vertices of directed acyclic graph g in topological order. */
2  public static <V,E> PositionalList<Vertex<V>> topologicalSort(Graph<V,E> g) {
3      // list of vertices placed in topological order
4      PositionalList<Vertex<V>> topo = new LinkedPositionalList<>();
5      // container of vertices that have no remaining constraints
6      Stack<Vertex<V>> ready = new LinkedStack<>();
7      // map keeping track of remaining in-degree for each vertex
8      Map<Vertex<V>, Integer> inCount = new ProbeHashMap<>();
9      for (Vertex<V> u : g.vertices()) {
10         inCount.put(u, g.inDegree(u));           // initialize with actual in-degree
11         if (inCount.get(u) == 0)                 // if u has no incoming edges,
12             ready.push(u);                       // it is free of constraints
13     }
14     while (!ready.isEmpty()) {
15         Vertex<V> u = ready.pop();
16         topo.addLast(u);
17         for (Edge<E> e : g.outgoingEdges(u)) { // consider all outgoing neighbors of u
18             Vertex<V> v = g.opposite(u, e);
19             inCount.put(v, inCount.get(v) - 1); // v has one less constraint without u
20             if (inCount.get(v) == 0)
21                 ready.push(v);
22         }
23     }
24     return topo;
25 }

```


Questions

