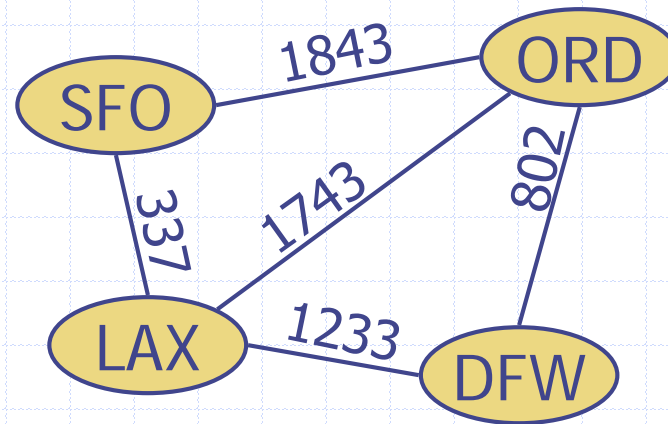


Chapter 6: Graphs

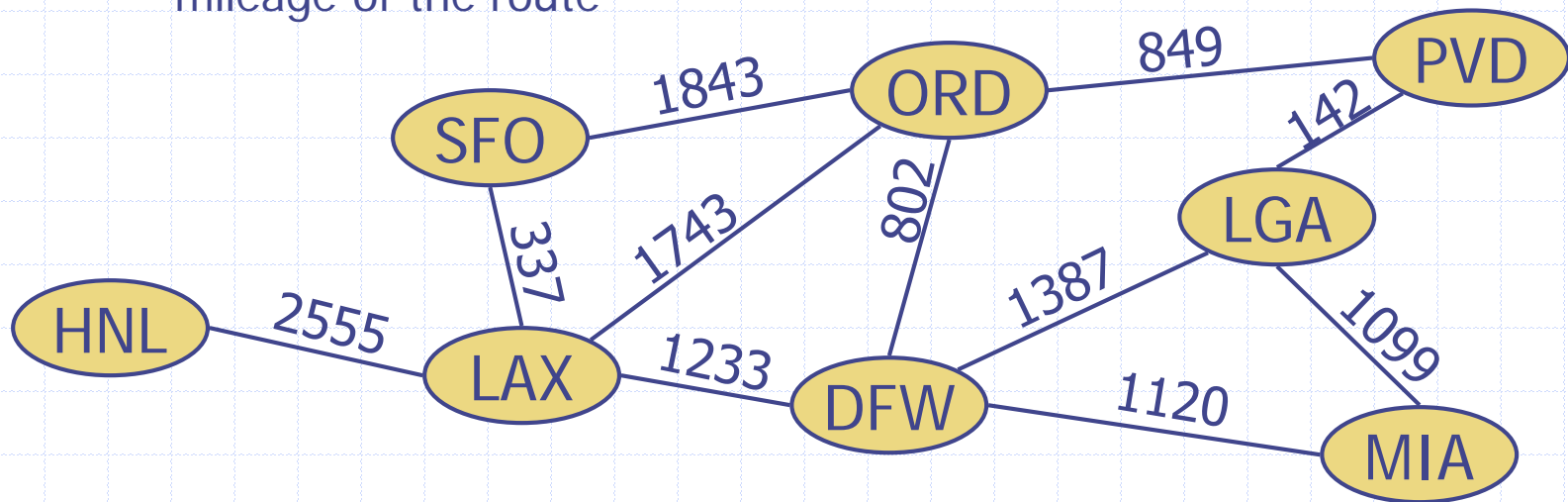


Outline and Reading

- ◆ Graphs (§6.1)
- ◆ Data Structures for Graphs (§6.2)
- ◆ Graph Traversal (§6.3)
- ◆ Directed Graph (§6.4)

6.1 Graph

- ◆ A graph is a pair (V, E) , where
 - V is a set of nodes, called **vertices**
 - E is a collection of pairs of vertices, called **edges**
 - Vertices and edges are positions and store elements
- ◆ Example:
 - A vertex represents an airport and stores the three-letter airport code
 - An edge represents a flight route between two airports and stores the mileage of the route



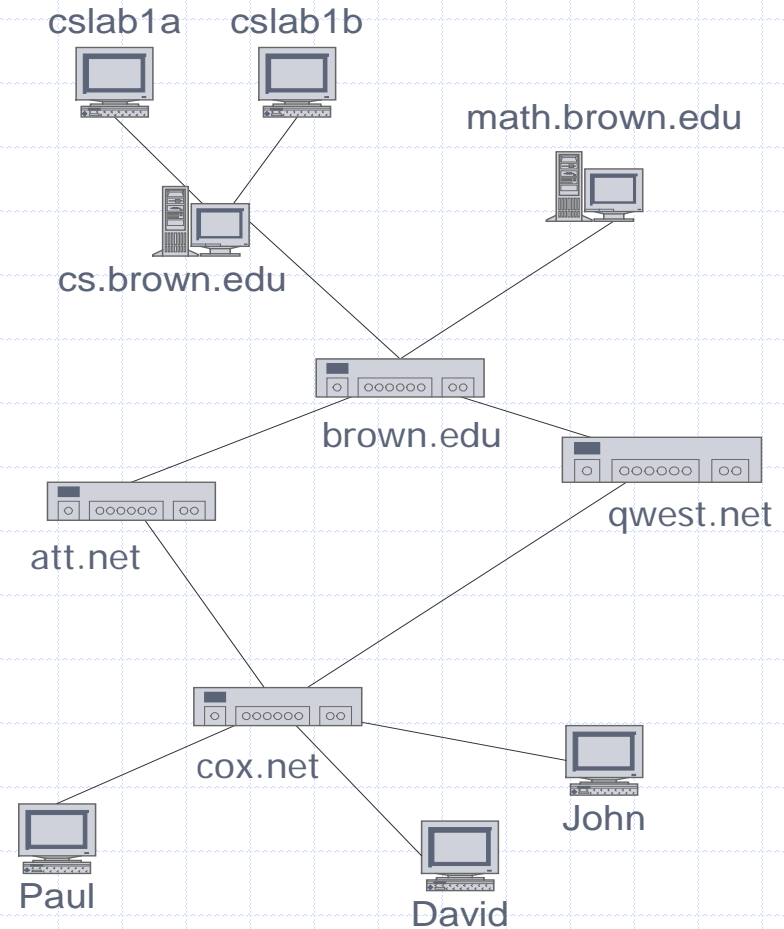
Edge Types

- ◆ Directed edge
 - ordered pair of vertices (u,v)
 - first vertex u is the origin
 - second vertex v is the destination
 - e.g., a flight
- ◆ Undirected edge
 - unordered pair of vertices (u,v)
 - e.g., a flight route
- ◆ Directed graph
 - all the edges are directed
 - e.g., flight network
- ◆ Undirected graph
 - all the edges are undirected
 - e.g., route network



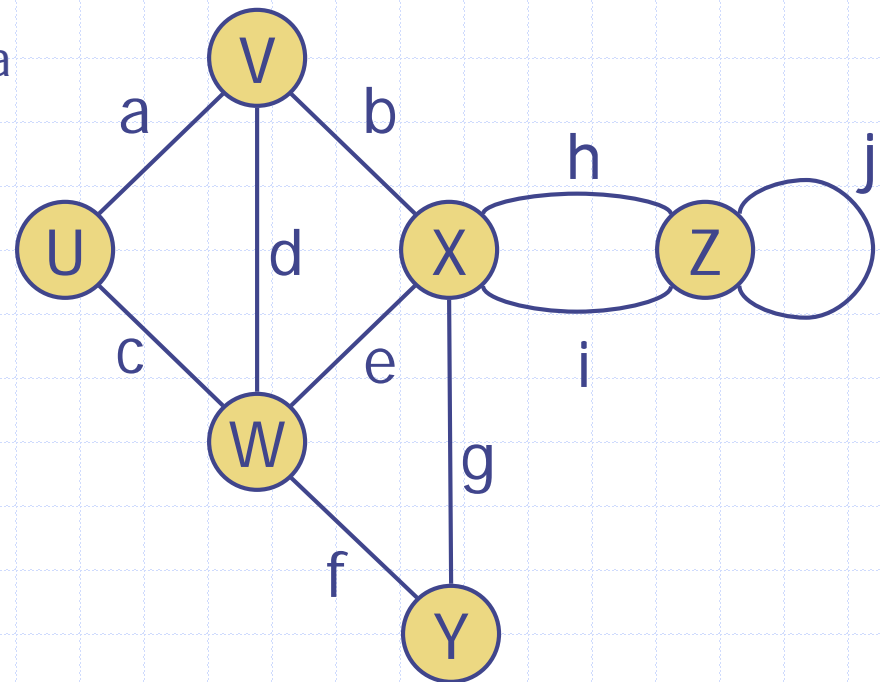
Applications

- ◆ Electronic circuits
 - Printed circuit board
 - Integrated circuit
- ◆ Transportation networks
 - Highway network
 - Flight network
- ◆ Computer networks
 - Local area network
 - Internet
 - Web
- ◆ Databases
 - Entity-relationship diagram



Terminology

- ◆ End vertices (or endpoints) of an edge
 - U and V are the endpoints of a
- ◆ Edges incident on a vertex
 - a, d, and b are incident on V
- ◆ Adjacent vertices
 - U and V are adjacent
- ◆ Degree of a vertex
 - X has degree 5
- ◆ Parallel edges
 - h and i are parallel edges
- ◆ Self-loop
 - j is a self-loop



Terminology (cont.)

◆ Path

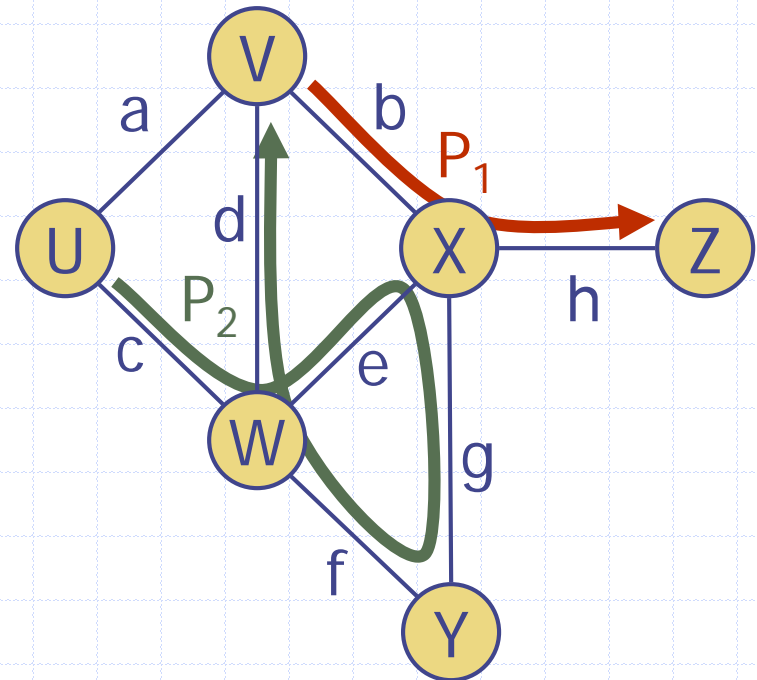
- sequence of alternating vertices and edges
- begins with a vertex
- ends with a vertex
- each edge is preceded and followed by its endpoints

◆ Simple path

- path such that all its vertices and edges are distinct

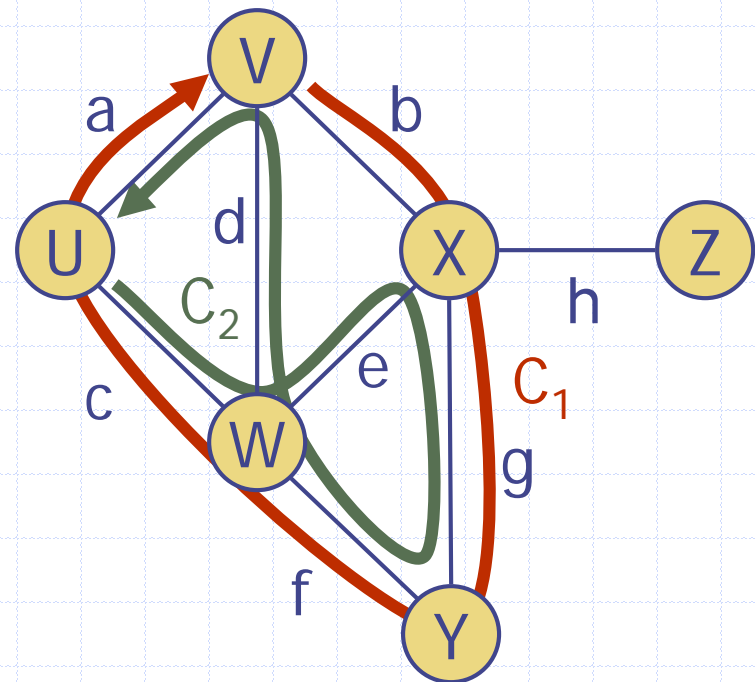
◆ Examples

- $P_1 = (V, b, X, h, Z)$ is a simple path
- $P_2 = (U, c, W, e, X, g, Y, f, W, d, V)$ is a path that is not simple



Terminology (cont.)

- ◆ Cycle
 - circular sequence of alternating vertices and edges
 - each edge is preceded and followed by its endpoints
- ◆ Simple cycle
 - cycle such that all its vertices and edges are distinct
- ◆ Examples
 - $C_1 = (V, b, X, g, Y, f, W, c, U, a, \rightarrow)$ is a simple cycle
 - $C_2 = (U, c, W, e, X, g, Y, f, W, d, V, a, \rightarrow)$ is a cycle that is not simple



Properties

Property 1

$$\sum_v \deg(v) = 2m$$

Proof: each edge is counted twice

Property 2

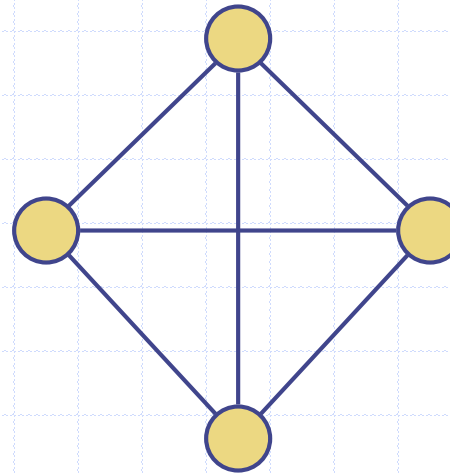
In an undirected graph with no self-loops and no multiple edges

$$m \leq n(n-1)/2$$

Proof: each vertex has degree at most $(n-1)$

Notation

n	number of vertices
m	number of edges
$\deg(v)$	degree of vertex v

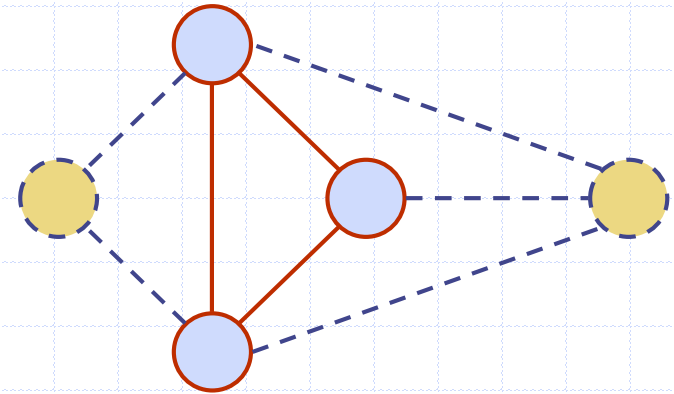


Example

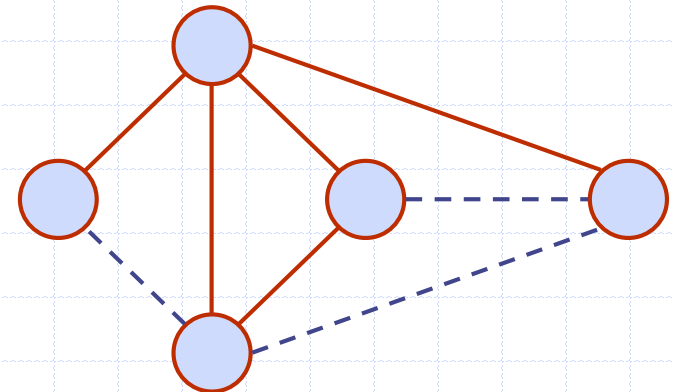
- $n = 4$
- $m = 6$
- $\deg(v) = 3$

Subgraphs

- ◆ A subgraph S of a graph G is a graph such that
 - The vertices of S are a subset of the vertices of G
 - The edges of S are a subset of the edges of G
- ◆ A spanning subgraph of G is a subgraph that contains all the vertices of G



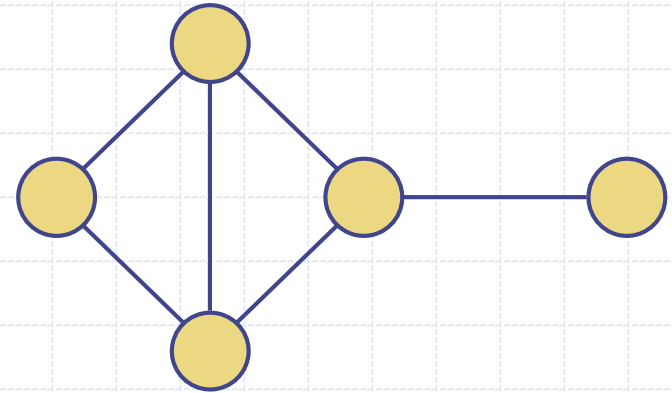
Subgraph



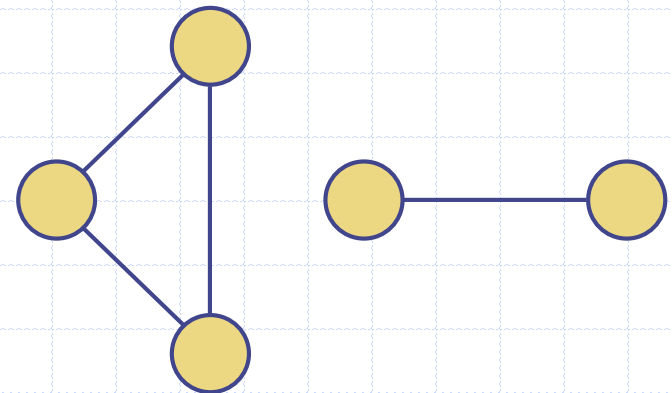
Spanning subgraph

Connectivity

- ◆ A graph is connected if there is a path between every pair of vertices
- ◆ A connected component of a graph G is a maximal connected subgraph of G



Connected graph



Non connected graph with two connected components

Trees and Forests

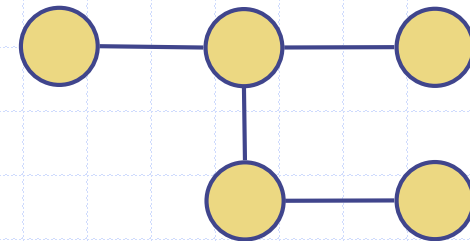
◆ A (free) tree is an undirected graph T such that

- T is connected
- T has no cycles

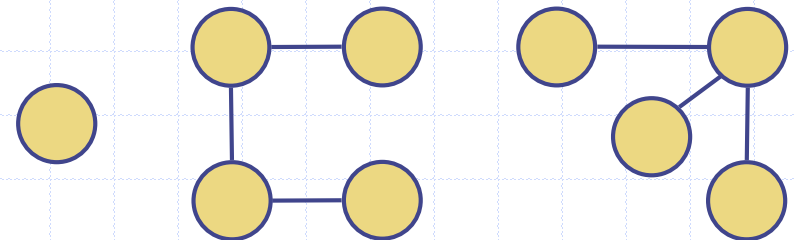
This definition of tree is different from the one of a rooted tree

◆ A forest is an undirected graph without cycles

◆ The connected components of a forest are trees



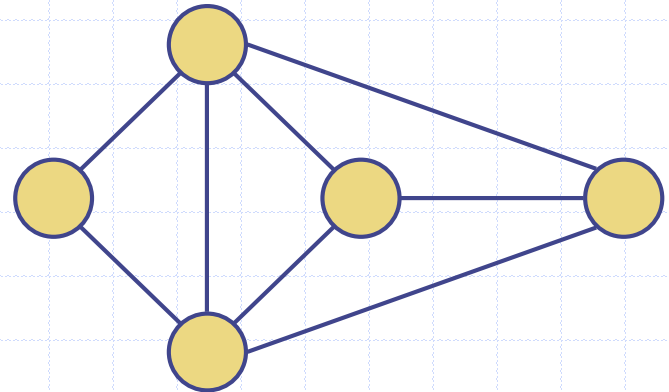
Tree



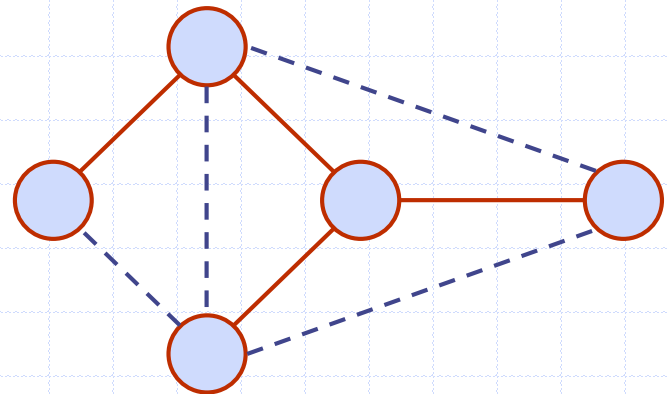
Forest

Spanning Trees and Forests

- ◆ A spanning tree of a connected graph is a spanning subgraph that is a tree
- ◆ A spanning tree is not unique unless the graph is a tree
- ◆ Spanning trees have applications to the design of communication networks
- ◆ A spanning forest of a graph is a spanning subgraph that is a forest



Graph



Spanning tree

Main Methods of the Graph ADT

◆ Vertices and edges

- are positions
- store elements

◆ Accessor methods

- `aVertex()`
- `incidentEdges(v)`
- `endVertices(e)`
- `isDirected(e)`
- `origin(e)`
- `destination(e)`
- `opposite(v, e)`
- `areAdjacent(v, w)`

◆ Update methods

- `insertVertex(o)`
- `insertEdge(v, w, o)`
- `insertDirectedEdge(v, w, o)`
- `removeVertex(v)`
- `removeEdge(e)`

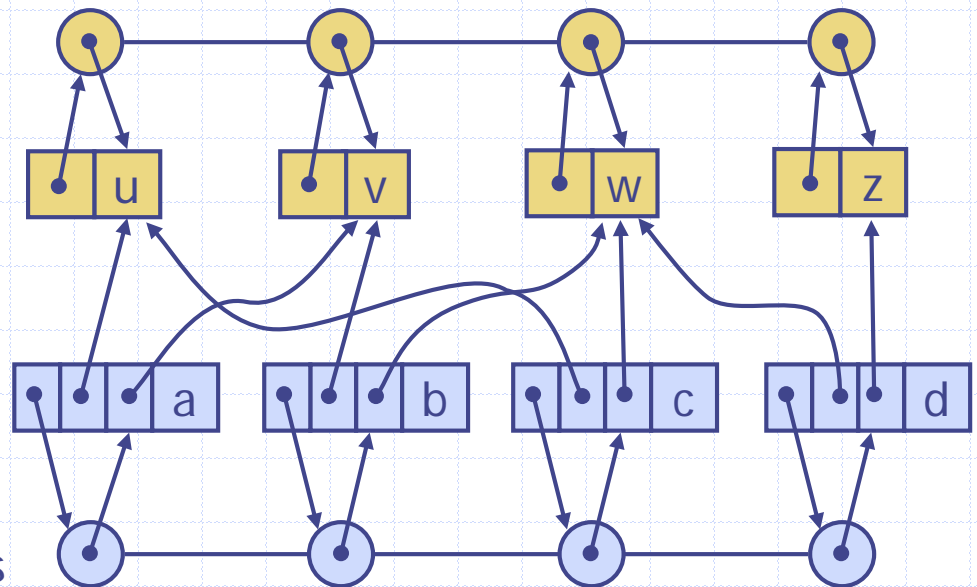
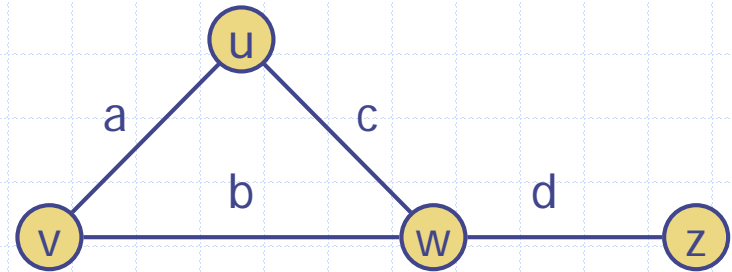
◆ Generic methods

- `numVertices()`
- `numEdges()`
- `vertices()`
- `edges()`

6.2 Data Structure for Graphs

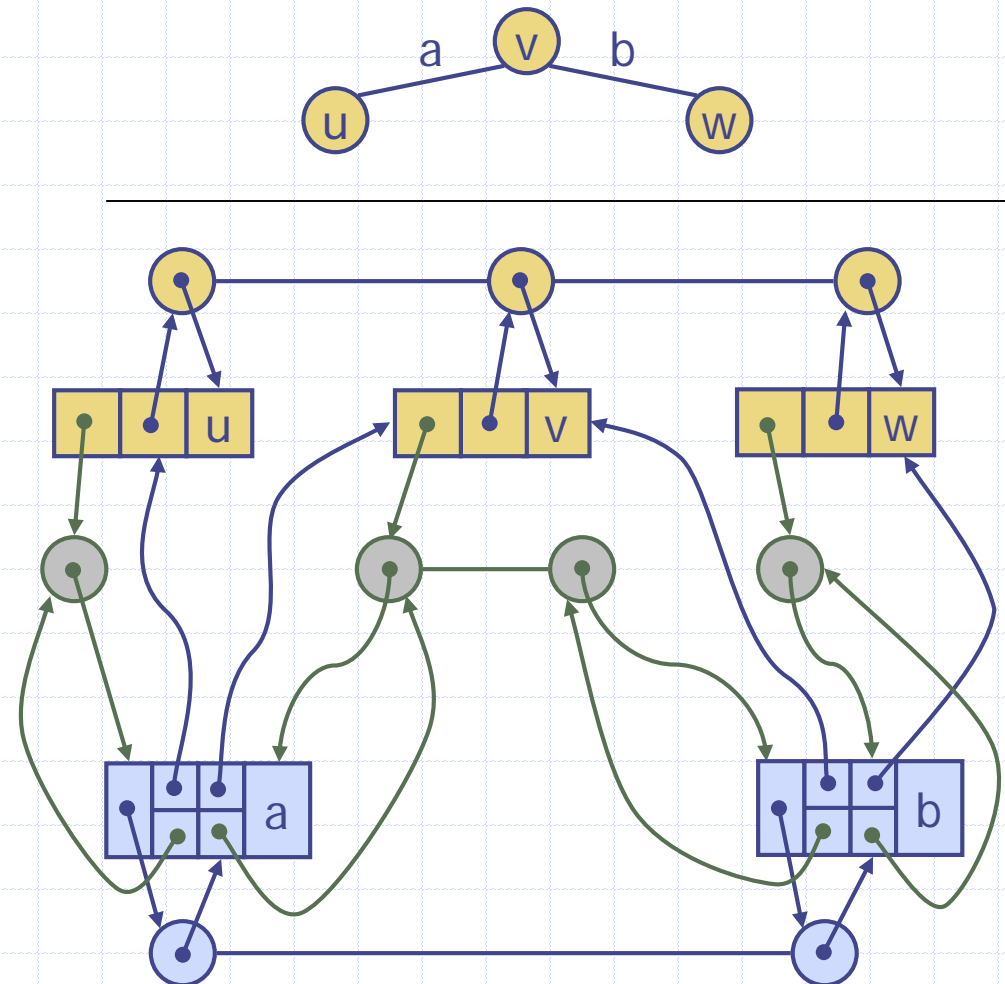
Edge List Structure

- ◆ Vertex object
 - element
 - reference to position in vertex sequence
- ◆ Edge object
 - element
 - origin vertex object
 - destination vertex object
 - reference to position in edge sequence
- ◆ Vertex sequence
 - sequence of vertex objects
- ◆ Edge sequence
 - sequence of edge objects



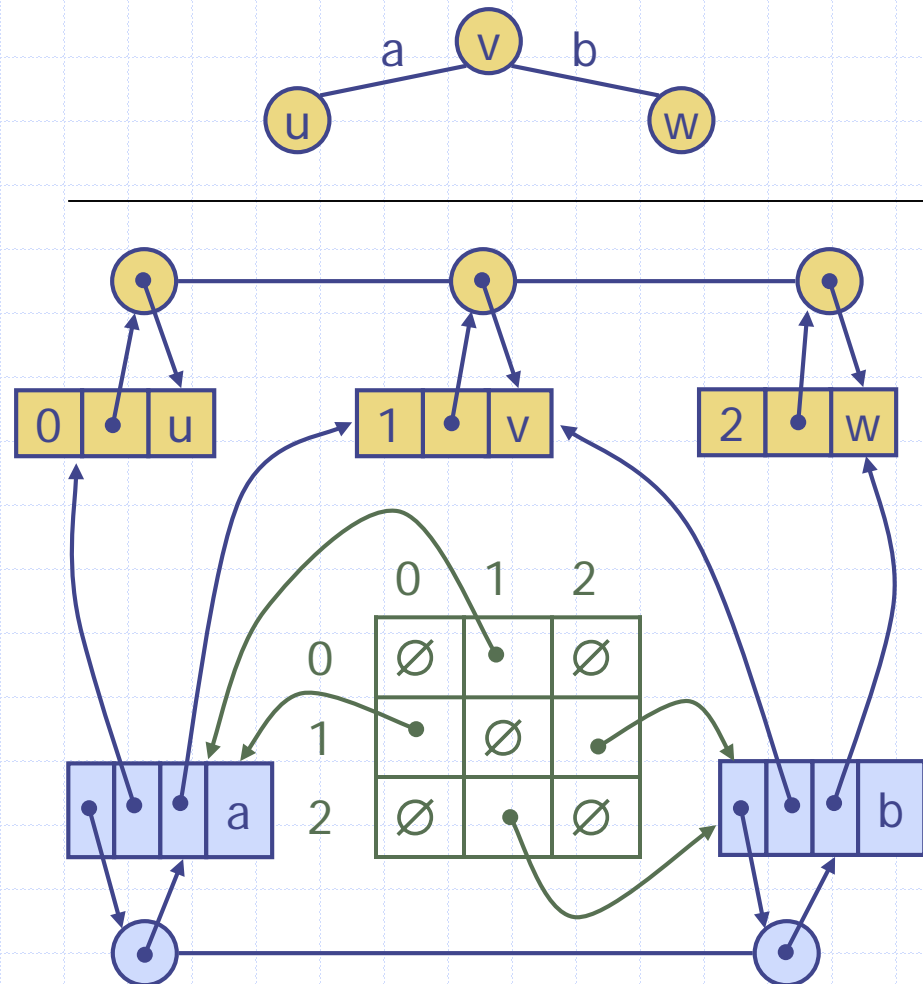
Adjacency List Structure

- ◆ Edge list structure
- ◆ Incidence sequence for each vertex
 - sequence of references to edge objects of incident edges
- ◆ Augmented edge objects
 - references to associated positions in incidence sequences of end vertices



Adjacency Matrix Structure

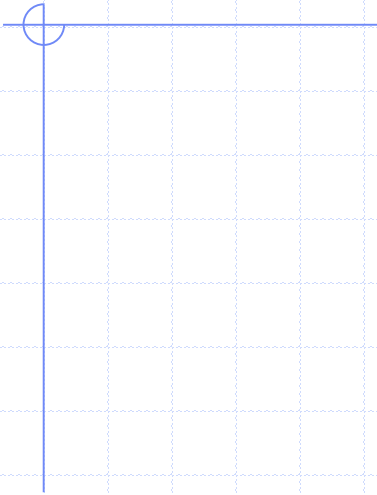
- ◆ Edge list structure
- ◆ Augmented vertex objects
 - Integer key (index) associated with vertex
- ◆ 2D adjacency array
 - Reference to edge object for adjacent vertices
 - Null for non adjacent vertices
- ◆ The “old fashioned” version just has 0 for no edge and 1 for edge

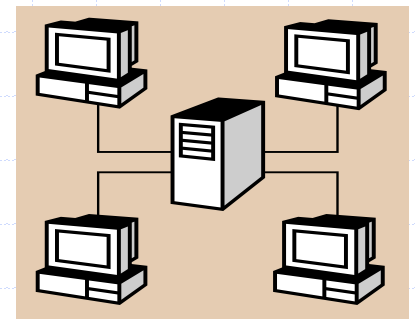


Asymptotic Performance

<ul style="list-style-type: none"> ◆ n vertices, m edges ◆ no parallel edges ◆ no self-loops ◆ Bounds are "big-Oh" 	Edge List	Adjacency List	Adjacency Matrix
Space	$n + m$	$n + m$	n^2
incidentEdges(v)	m	deg(v)	n
areAdjacent (v, w)	m	min(deg(v), deg(w))	1
insertVertex(o)	1	1	n^2
insertEdge(v, w, o)	1	1	1
removeVertex(v)	m	deg(v)	n^2
removeEdge(e)	1	1	1

6.3 Graph Traversal

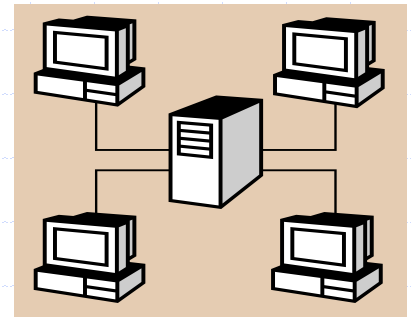




6.3.1 Depth-First Search

- ◆ Depth-first search (DFS) is a general technique for traversing a graph
- ◆ A DFS traversal of a graph G
 - Visits all the vertices and edges of G
 - Determines whether G is connected
 - Computes the connected components of G
 - Computes a spanning forest of G
- ◆ DFS on a graph with n vertices and m edges takes $O(n + m)$ time
- ◆ DFS can be further extended to solve other graph problems
 - Find and report a path between two given vertices
 - Find a cycle in the graph
- ◆ Depth-first search is to graphs what Euler tour is to binary trees

DFS Algorithm



- ◆ The algorithm uses a mechanism for setting and getting “labels” of vertices and edges

Algorithm *DFS(G)*

Input graph G

Output labeling of the edges of G
as discovery edges and
back edges

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

setLabel(u, UNEXPLORED)

for all $e \in G.edges()$

setLabel(e, UNEXPLORED)

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

if *getLabel(v) = UNEXPLORED*

DFS(G, v)

Algorithm *DFS(G, v)*

Input graph G and a start vertex v of G

Output labeling of the edges of G
in the connected component of v
as discovery edges and back edges

setLabel(v, VISITED)

for all $e \in G.incidentEdges(v)$

if *getLabel(e) = UNEXPLORED*

$w \leftarrow G.opposite(v, e)$

if *getLabel(w) = UNEXPLORED*

setLabel(e, DISCOVERY)

DFS(G, w)

else

setLabel(e, BACK)

Example



unexplored vertex



visited vertex



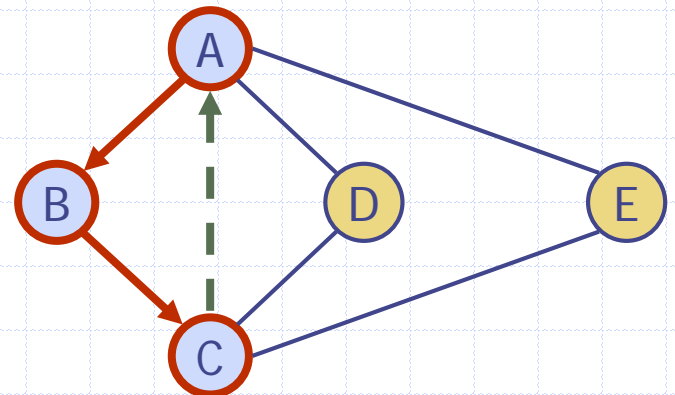
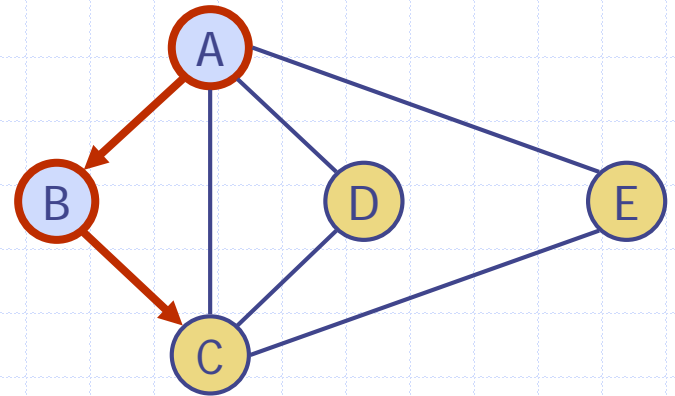
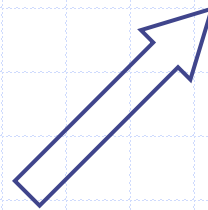
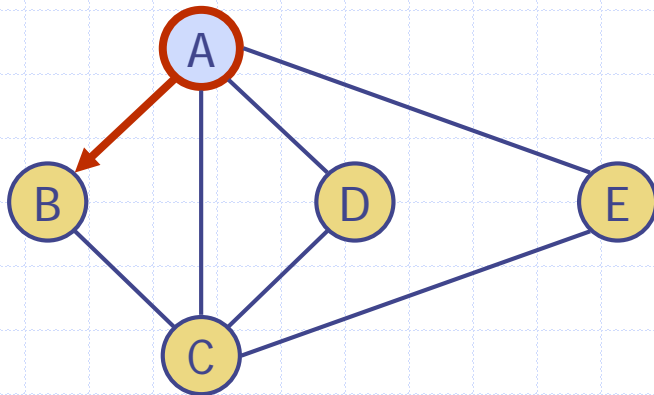
unexplored edge



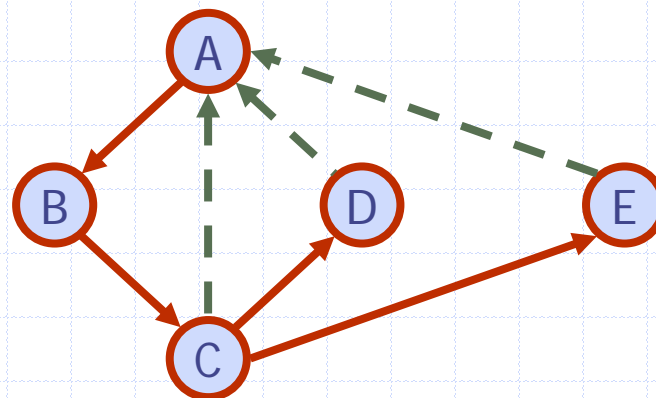
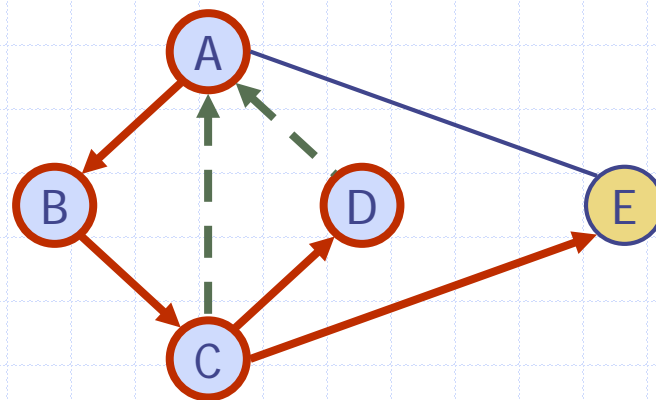
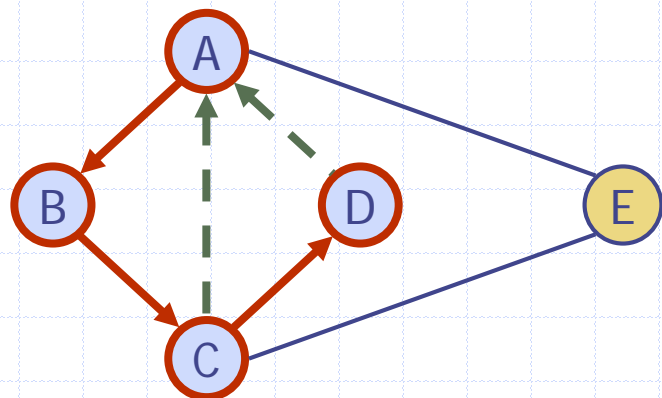
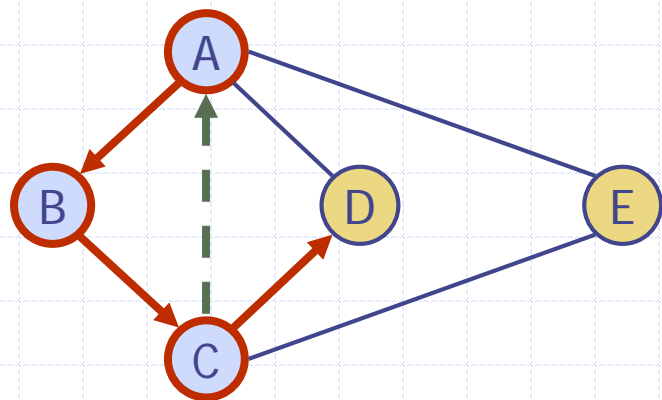
discovery edge



back edge



Example (cont.)



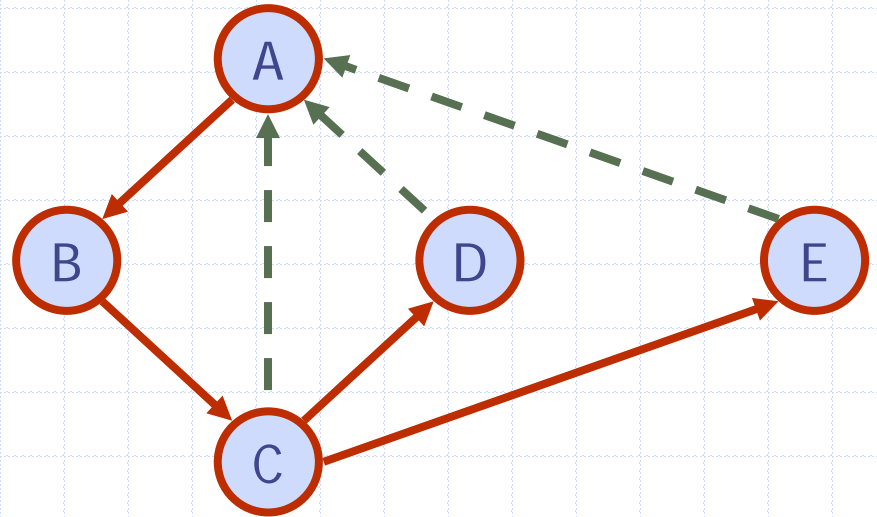
Properties of DFS

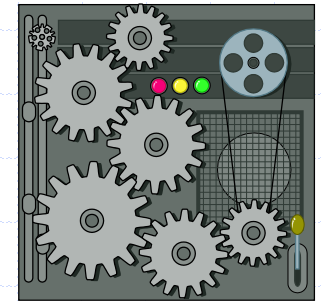
Property 1

$DFS(G, v)$ visits all the vertices and edges in the connected component of v

Property 2

The discovery edges labeled by $DFS(G, v)$ form a spanning tree of the connected component of v





Analysis of DFS

- ◆ Setting/getting a vertex/edge label takes $O(1)$ time
- ◆ Each vertex is labeled twice
 - once as UNEXPLORED
 - once as **VISITED**
- ◆ Each edge is labeled twice
 - once as UNEXPLORED
 - once as **DISCOVERY** or **BACK**
- ◆ Method incidentEdges is called once for each vertex
- ◆ DFS runs in $O(n + m)$ time provided the graph is represented by the adjacency list structure
 - Recall that $\sum_v \deg(v) = 2m$

Path Finding



- ◆ We can specialize the DFS algorithm to find a path between two given vertices u and z using the template method pattern
- ◆ We call $DFS(G, u)$ with u as the start vertex
- ◆ We use a stack S to keep track of the path between the start vertex and the current vertex
- ◆ As soon as destination vertex z is encountered, we return the path as the contents of the stack

```
Algorithm pathDFS( $G, v, z$ )
  setLabel( $v, VISITED$ )
  S.push( $v$ )
  if  $v = z$ 
    return S.elements()
  for all  $e \in G.incidentEdges(v)$ 
    if getLabel( $e$ ) = UNEXPLORED
       $w \leftarrow opposite(v, e)$ 
      if getLabel( $w$ ) = UNEXPLORED
        setLabel( $e, DISCOVERY$ )
        S.push( $e$ )
        pathDFS( $G, w, z$ )
        S.pop()      {  $e$  gets popped }
      else
        setLabel( $e, BACK$ )
  S.pop()          {  $v$  gets popped }
```

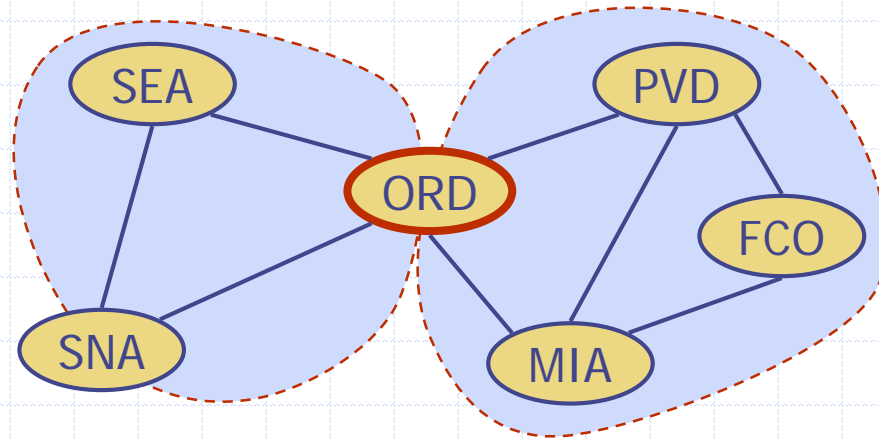
Cycle Finding



- ◆ We can specialize the DFS algorithm to find a simple cycle using the template method pattern
- ◆ We use a stack S to keep track of the path between the start vertex and the current vertex
- ◆ As soon as a back edge (v, w) is encountered, we return the cycle as the portion of the stack from the top to vertex w

```
Algorithm cycleDFS( $G, v, z$ )
  setLabel( $v, VISITED$ )
   $S.push(v)$ 
  for all  $e \in G.incidentEdges(v)$ 
    if getLabel( $e$ ) = UNEXPLORED
       $w \leftarrow opposite(v, e)$ 
       $S.push(e)$ 
      if getLabel( $w$ ) = UNEXPLORED
        setLabel( $e, DISCOVERY$ )
        pathDFS( $G, w, z$ )
         $S.pop()$ 
      else
         $C \leftarrow$  new empty stack
        repeat
           $o \leftarrow S.pop()$ 
           $C.push(o)$ 
        until  $o = w$ 
        return  $C.elements()$ 
   $S.pop()$ 
```

6.3.2 Biconnectivity



Separation Edges and Vertices

◆ Definitions

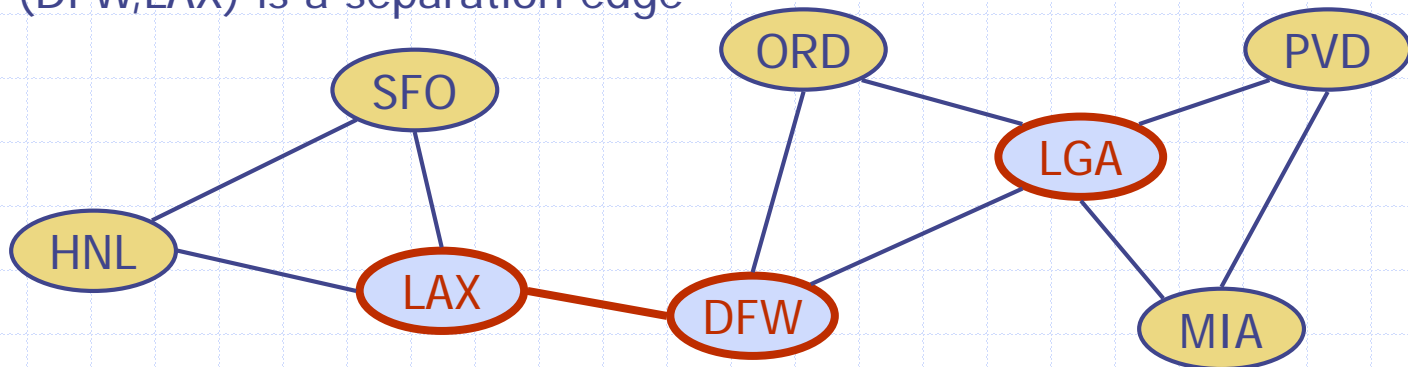
- Let G be a connected graph
- A separation edge of G is an edge whose removal disconnects G
- A separation vertex of G is a vertex whose removal disconnects G

◆ Applications

- Separation edges and vertices represent single points of failure in a network and are critical to the operation of the network

◆ Example

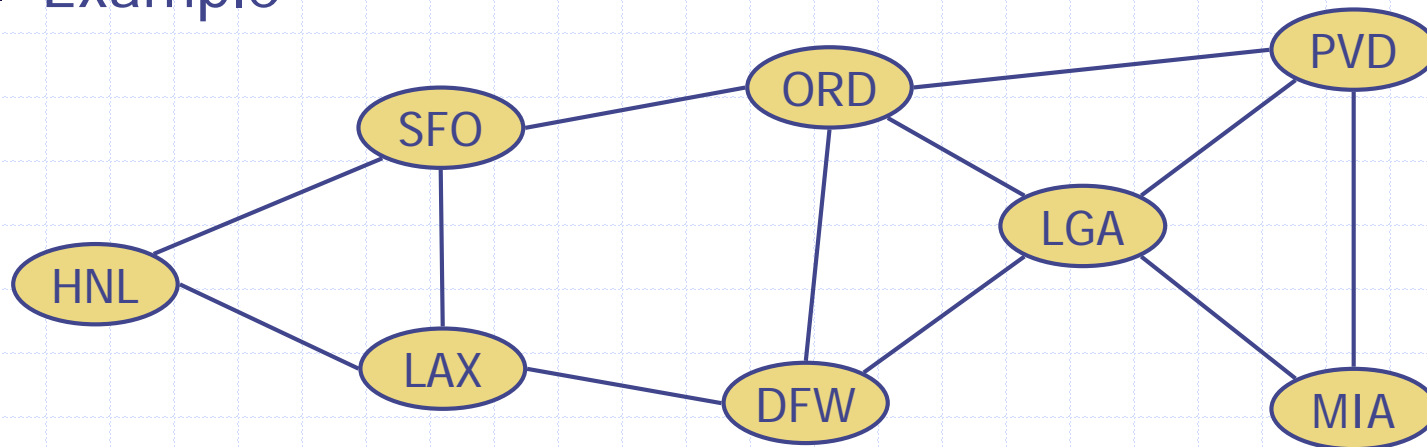
- DFW, LGA and LAX are separation vertices
- (DFW,LAX) is a separation edge



Biconnected Graph

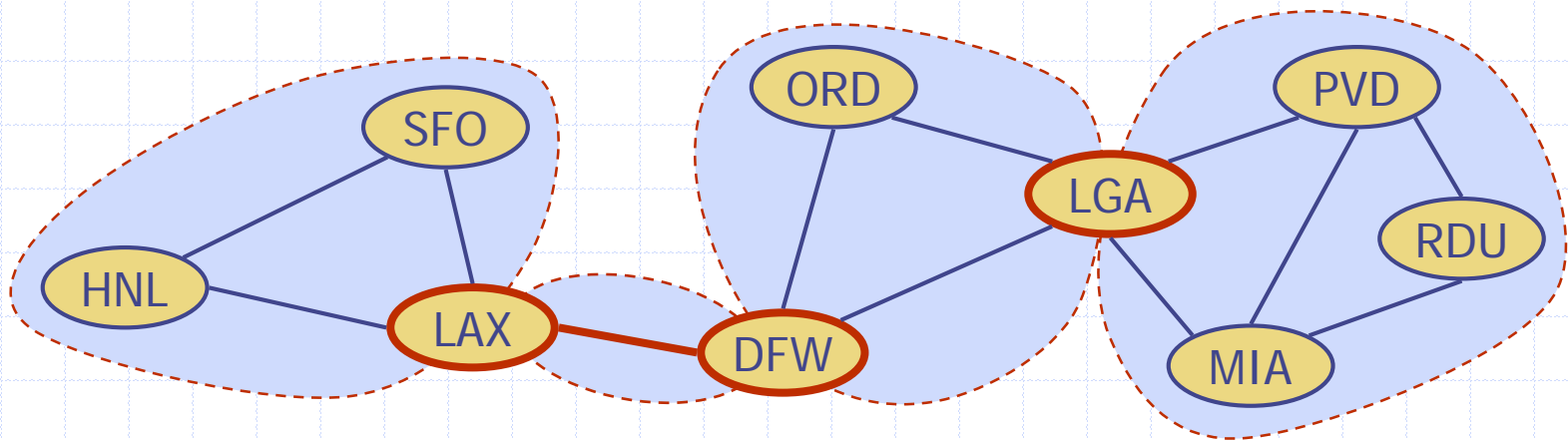
- ◆ Equivalent definitions of a biconnected graph G
 - Graph G has no separation edges and no separation vertices
 - For any two vertices u and v of G , there are two disjoint simple paths between u and v (i.e., two simple paths between u and v that share no other vertices or edges)
 - For any two vertices u and v of G , there is a simple cycle containing u and v

- ◆ Example



Biconnected Components

- ◆ Biconnected component of a graph G
 - A maximal biconnected subgraph of G , or
 - A subgraph consisting of a separation edge of G and its end vertices
- ◆ Interaction of biconnected components
 - An edge belongs to exactly one biconnected component
 - A nonseparation vertex belongs to exactly one biconnected component
 - A separation vertex belongs to two or more biconnected components
- ◆ Example of a graph with four biconnected components



Equivalence Classes

- ◆ Given a set S , a relation R on S is a set of ordered pairs of elements of S , i.e., R is a subset of $S \times S$
- ◆ An equivalence relation R on S satisfies the following properties
 - Reflexive:** $(x,x) \in R$
 - Symmetric:** $(x,y) \in R \Rightarrow (y,x) \in R$
 - Transitive:** $(x,y) \in R \wedge (y,z) \in R \Rightarrow (x,z) \in R$
- ◆ An equivalence relation R on S induces a partition of the elements of S into equivalence classes
- ◆ Example (connectivity relation among the vertices of a graph):
 - Let V be the set of vertices of a graph G
 - Define the relation
 - $C = \{(v,w) \in V \times V \text{ such that } G \text{ has a path from } v \text{ to } w\}$
 - Relation C is an equivalence relation
 - The equivalence classes of relation C are the vertices in each connected component of graph G

Link Relation

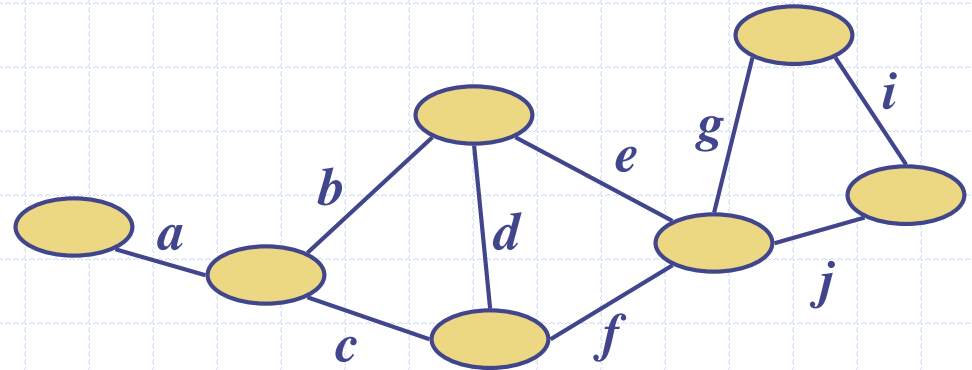
- ◆ Edges e and f of connected graph G are linked if
 - $e = f$, or
 - G has a simple cycle containing e and f

Theorem:

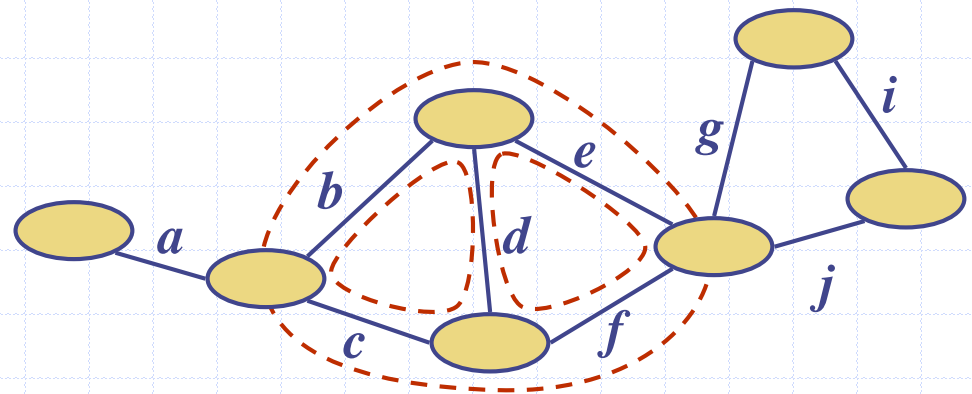
The link relation on the edges of a graph is an equivalence relation

Proof Sketch:

- The reflexive and symmetric properties follow from the definition
- For the transitive property, consider two simple cycles sharing an edge

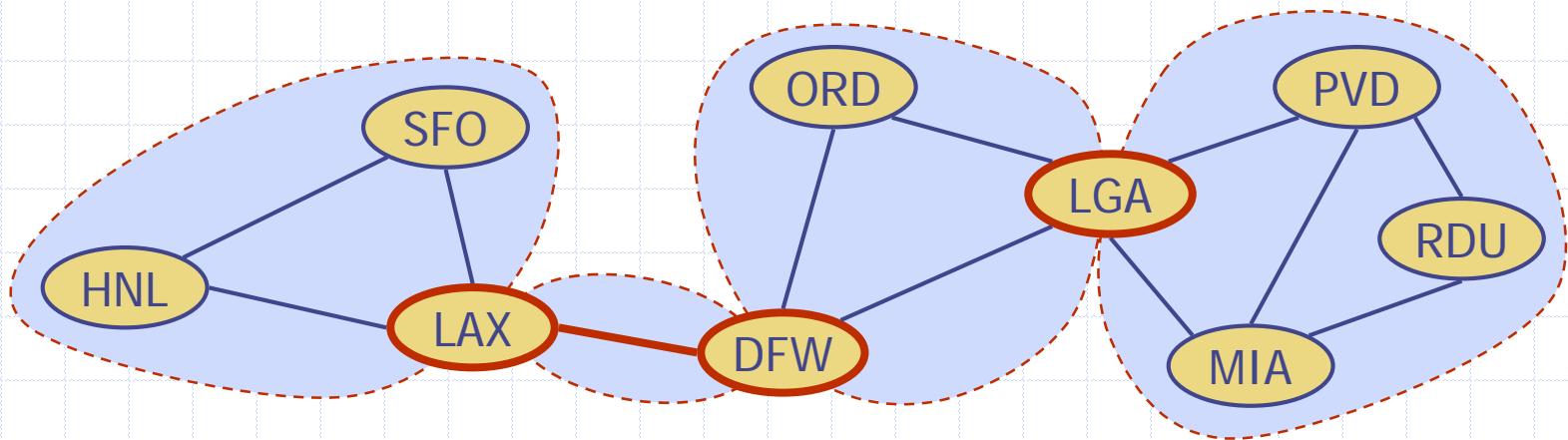


Equivalence classes of linked edges:
 $\{a\}$ $\{b, c, d, e, f\}$ $\{g, i, j\}$



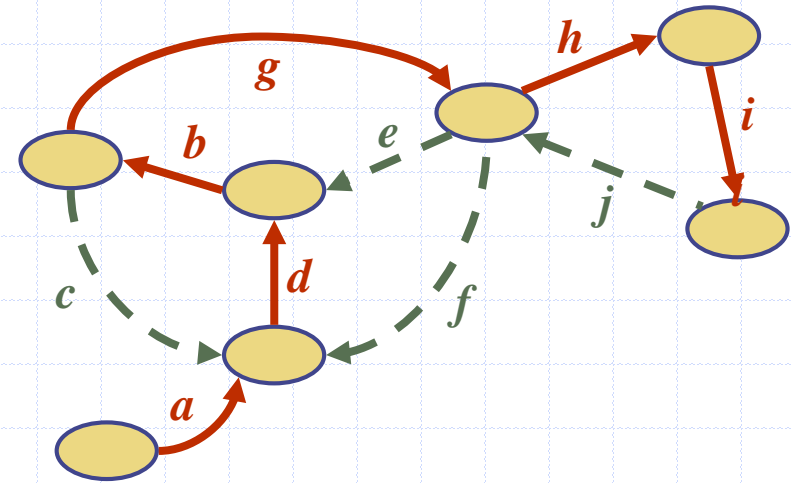
Link Components

- ◆ The link components of a connected graph G are the equivalence classes of edges with respect to the link relation
- ◆ A biconnected component of G is the subgraph of G induced by an equivalence class of linked edges
- ◆ A separation edge is a single-element equivalence class of linked edges
- ◆ A separation vertex has incident edges in at least two distinct equivalence classes of linked edge

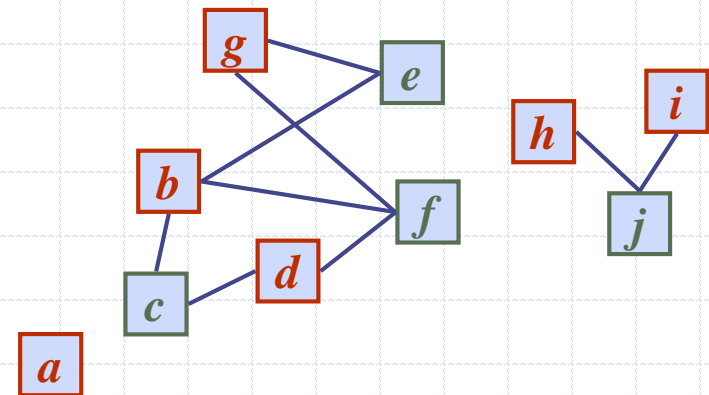


Auxiliary Graph

- ◆ Auxiliary graph B for a connected graph G
 - Associated with a DFS traversal of G
 - The vertices of B are the edges of G
 - For each back edge e of G , B has edges $(e, f_1), (e, f_2), \dots, (e, f_k)$, where f_1, f_2, \dots, f_k are the discovery edges of G that form a simple cycle with e
 - Its connected components correspond to the link components of G



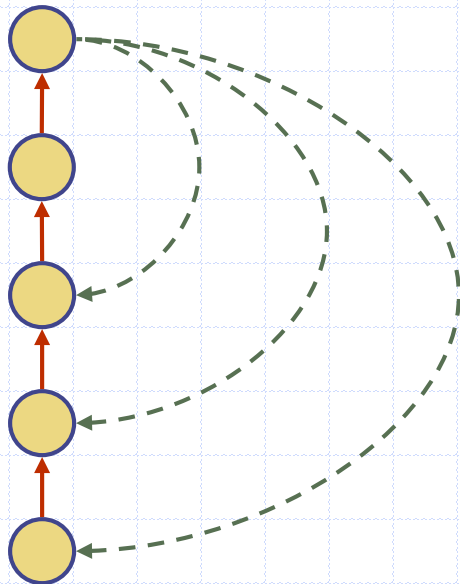
DFS on graph G



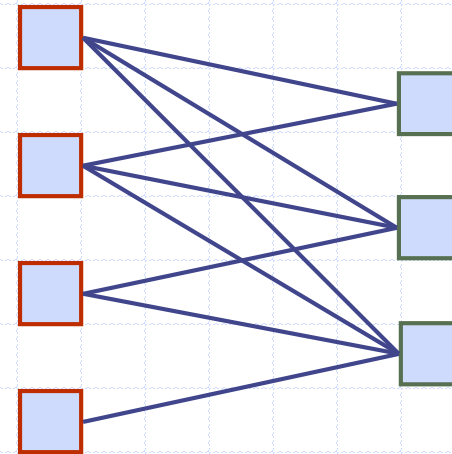
Auxiliary graph B

Auxiliary Graph (cont.)

- ◆ In the worst case, the number of edges of the auxiliary graph is proportional to nm



DFS on graph G



Auxiliary graph B

Proxy Graph

Algorithm *proxyGraph*(G)

Input connected graph G

Output proxy graph F for G

$F \leftarrow$ empty graph

DFS(G, s) { s is any vertex of G }

for all discovery edges e of G

$F.insertVertex(e)$

setLabel($e, UNLINKED$)

for all vertices v of G in DFS visit order

for all back edges $e = (u, v)$

$F.insertVertex(e)$

repeat

$f \leftarrow$ discovery edge with dest. u

$F.insertEdge(e, f, \emptyset)$

if $f.getLabel(f) = UNLINKED$

$setLabel(f, LINKED)$

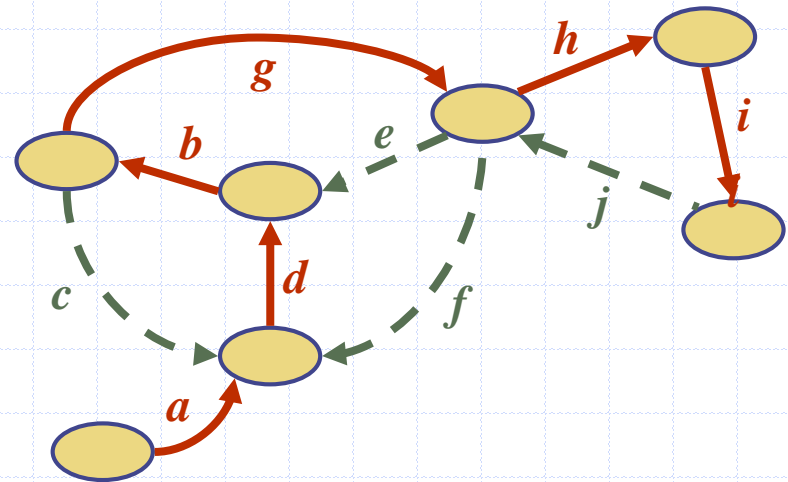
$u \leftarrow$ origin of edge f

else

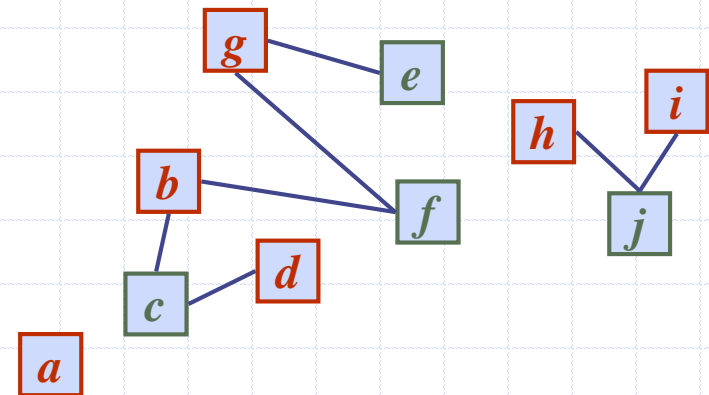
$u \leftarrow v$ { ends the loop }

until $u = v$

return F



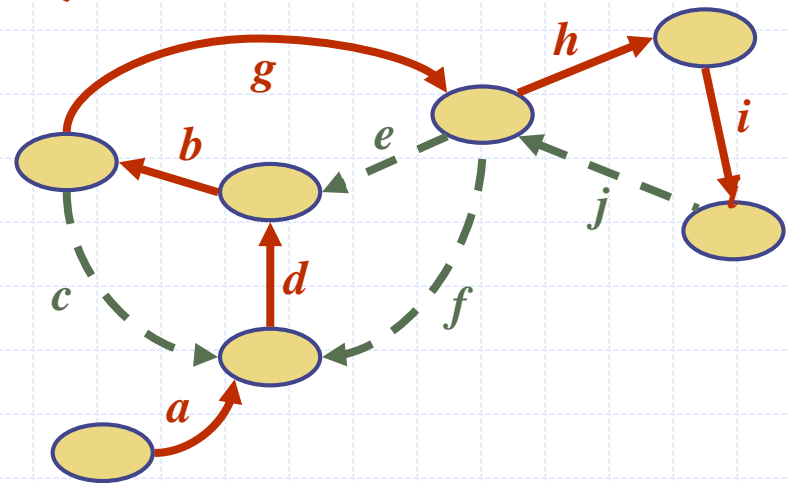
DFS on graph G



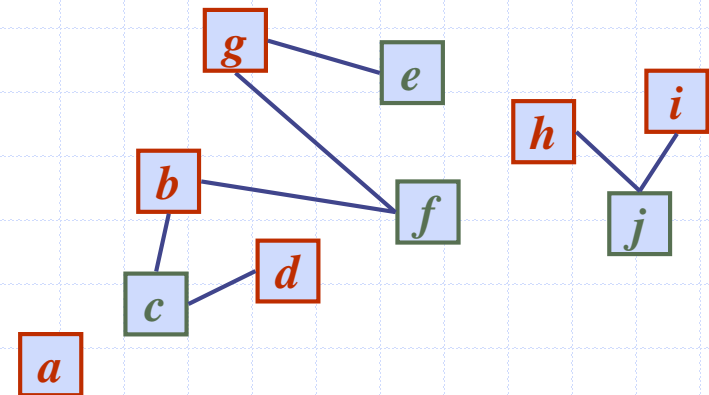
Proxy graph F

Proxy Graph (cont.)

- ◆ Proxy graph F for a connected graph G
 - Spanning forest of the auxiliary graph B
 - Has m vertices and $O(m)$ edges
 - Can be constructed in $O(n + m)$ time
 - Its connected components (trees) correspond to the link components of G
- ◆ Given a graph G with n vertices and m edges, we can compute the following in $O(n + m)$ time
 - The biconnected components of G
 - The separation vertices of G
 - The separation edges of G

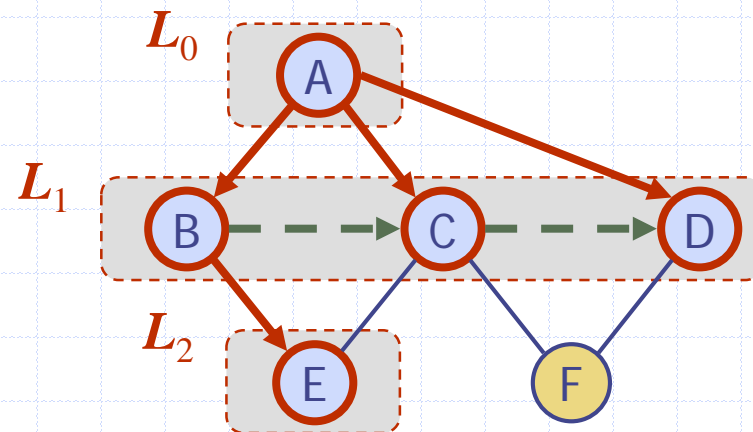


DFS on graph G



Proxy graph F

6.3.3 Breadth-First Search



Breadth-First Search

- ◆ Breadth-first search (BFS) is a general technique for traversing a graph
- ◆ A BFS traversal of a graph G
 - Visits all the vertices and edges of G
 - Determines whether G is connected
 - Computes the connected components of G
 - Computes a spanning forest of G
- ◆ BFS on a graph with n vertices and m edges takes $O(n + m)$ time
- ◆ BFS can be further extended to solve other graph problems
 - Find and report a path with the minimum number of edges between two given vertices
 - Find a simple cycle, if there is one

BFS Algorithm

- ◆ The algorithm uses a mechanism for setting and getting “labels” of vertices and edges

Algorithm *BFS(G)*

Input graph G

Output labeling of the edges and partition of the vertices of G

```
for all  $u \in G.vertices()$ 
     $setLabel(u, UNEXPLORED)$ 
for all  $e \in G.edges()$ 
     $setLabel(e, UNEXPLORED)$ 
for all  $v \in G.vertices()$ 
    if  $getLabel(v) = UNEXPLORED$ 
         $BFS(G, v)$ 
```

Algorithm *BFS(G, s)*

```
 $L_0 \leftarrow$  new empty sequence
 $L_0.insertLast(s)$ 
 $setLabel(s, VISITED)$ 
 $i \leftarrow 0$ 
while  $\neg L_i.isEmpty()$ 
     $L_{i+1} \leftarrow$  new empty sequence
    for all  $v \in L_i.elements()$ 
        for all  $e \in G.incidentEdges(v)$ 
            if  $getLabel(e) = UNEXPLORED$ 
                 $w \leftarrow opposite(v, e)$ 
                if  $getLabel(w) = UNEXPLORED$ 
                     $setLabel(e, DISCOVERY)$ 
                     $setLabel(w, VISITED)$ 
                     $L_{i+1}.insertLast(w)$ 
                else
                     $setLabel(e, CROSS)$ 
     $i \leftarrow i + 1$ 
```

Example

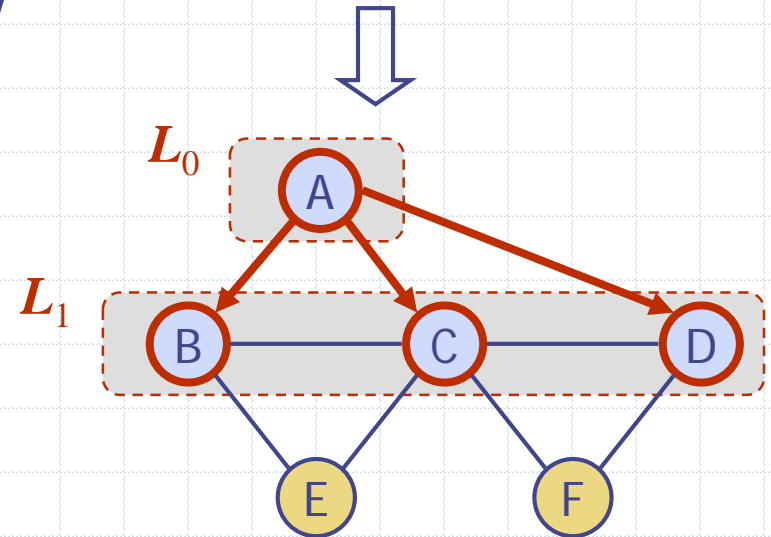
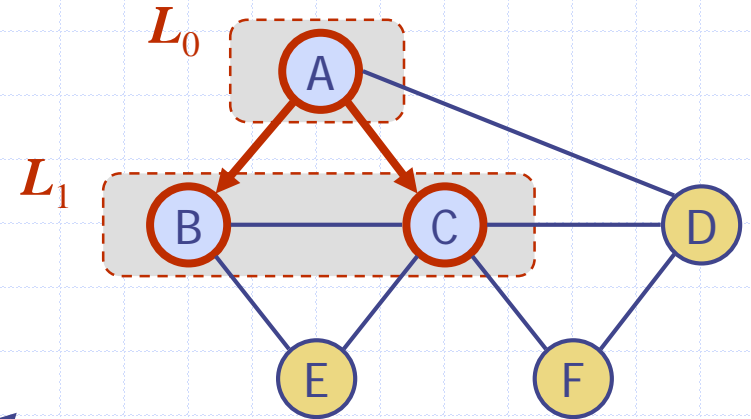
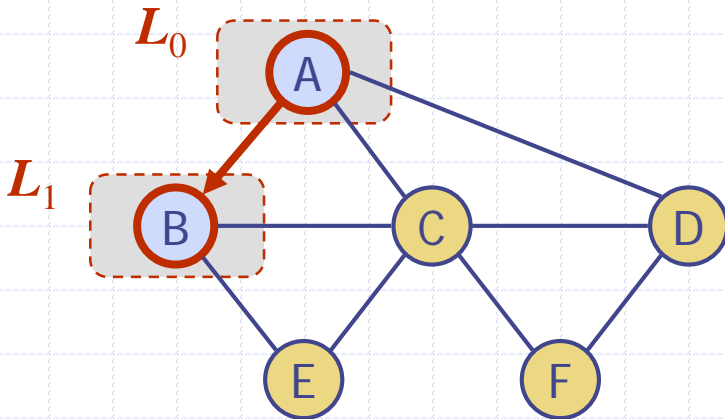
 unexplored vertex

 visited vertex

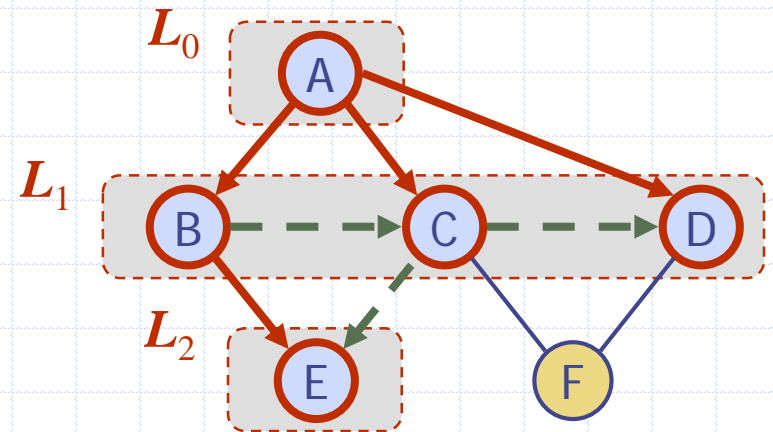
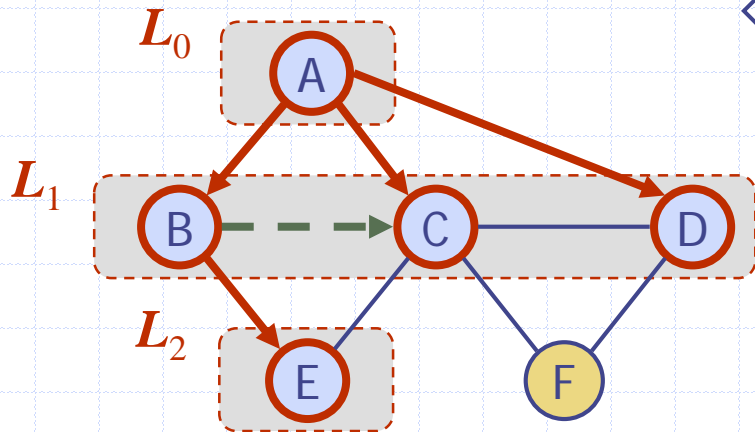
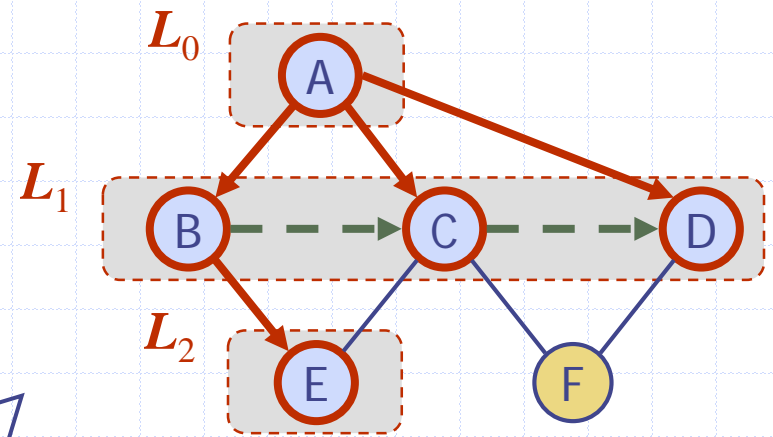
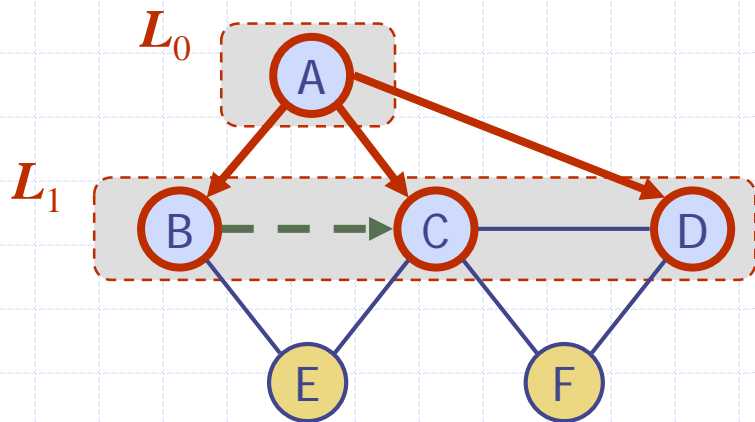
 unexplored edge

 discovery edge

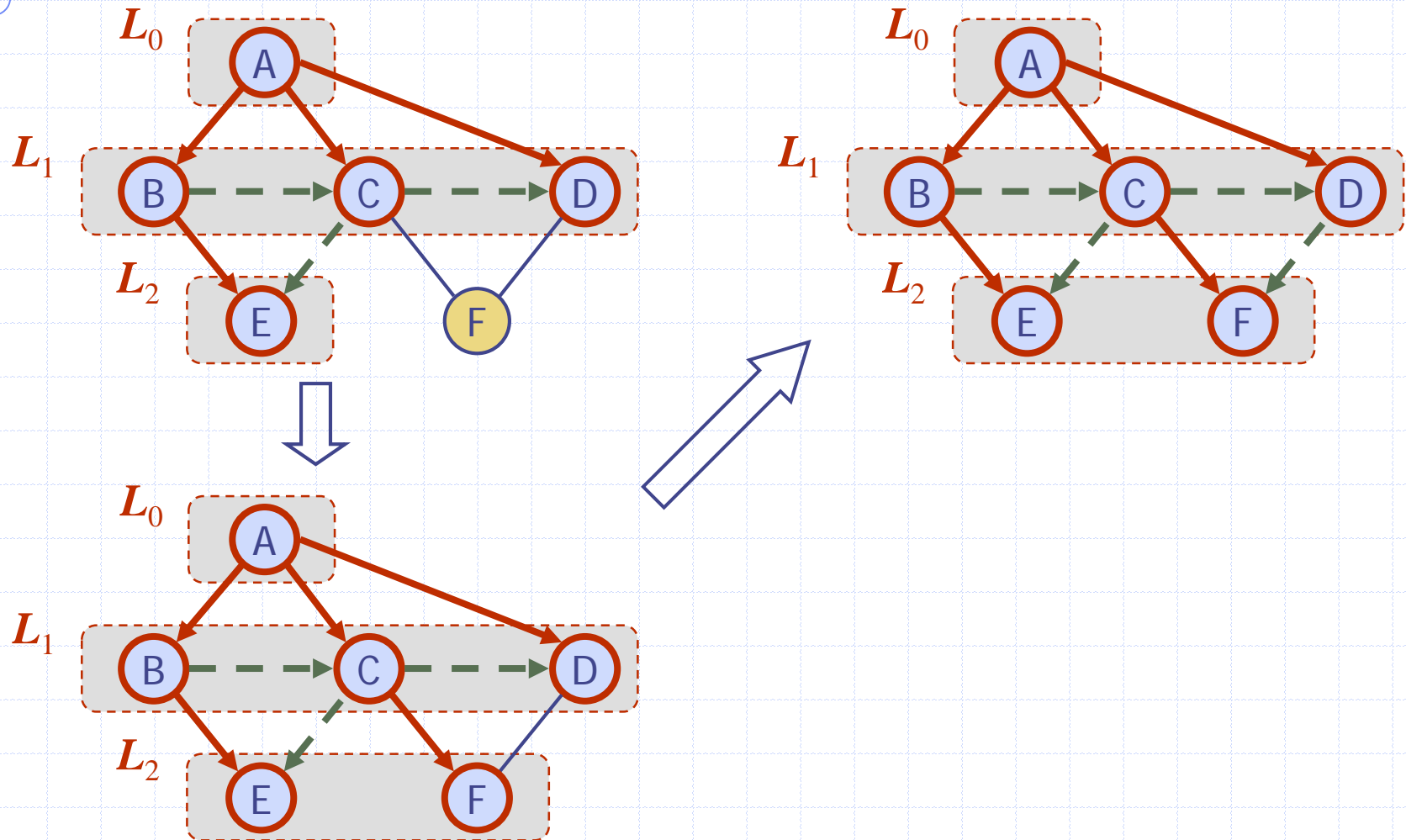
 cross edge



Example (cont.)



Example (cont.)



Properties

Notation

G_s : connected component of s

Property 1

$BFS(G, s)$ visits all the vertices and edges of G_s

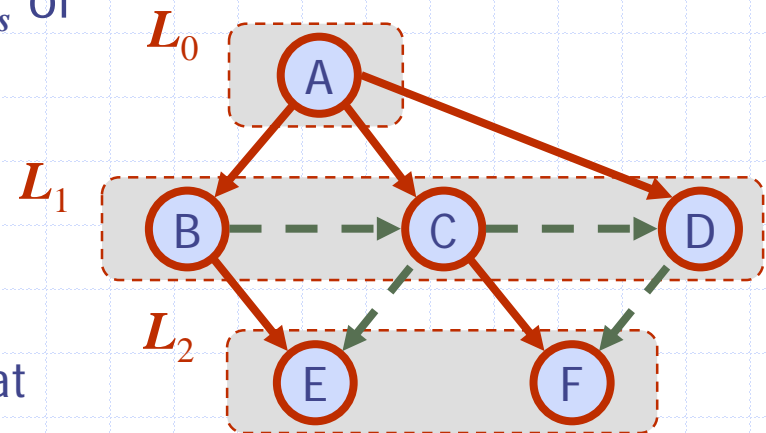
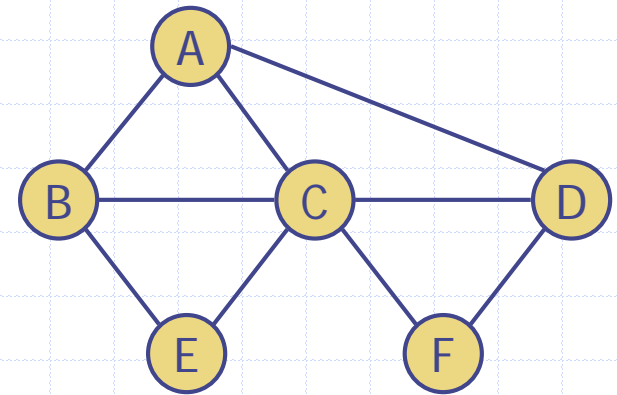
Property 2

The discovery edges labeled by $BFS(G, s)$ form a spanning tree T_s of G_s

Property 3

For each vertex v in L_i

- The path of T_s from s to v has i edges
- Every path from s to v in G_s has at least i edges



Analysis

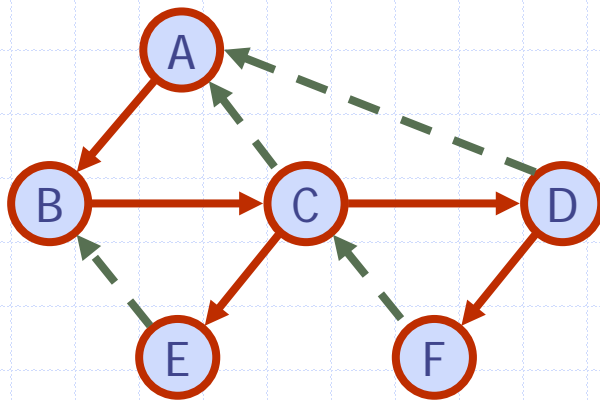
- ◆ Setting/getting a vertex/edge label takes $O(1)$ time
- ◆ Each vertex is labeled twice
 - once as UNEXPLORED
 - once as VISITED
- ◆ Each edge is labeled twice
 - once as UNEXPLORED
 - once as DISCOVERY or CROSS
- ◆ Each vertex is inserted once into a sequence L_i
- ◆ Method incidentEdges is called once for each vertex
- ◆ BFS runs in $O(n + m)$ time provided the graph is represented by the adjacency list structure
 - Recall that $\sum_v \deg(v) = 2m$

Applications

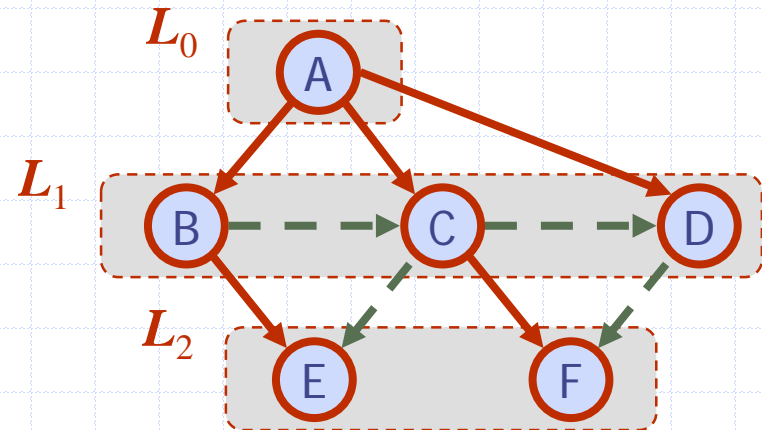
- ◆ Using the template method pattern, we can specialize the BFS traversal of a graph G to solve the following problems in $O(n + m)$ time
 - Compute the connected components of G
 - Compute a spanning forest of G
 - Find a simple cycle in G , or report that G is a forest
 - Given two vertices of G , find a path in G between them with the minimum number of edges, or report that no such path exists

DFS vs. BFS

Applications	DFS	BFS
Spanning forest, connected components, paths, cycles	✓	✓
Shortest paths		✓
Biconnected components	✓	



DFS

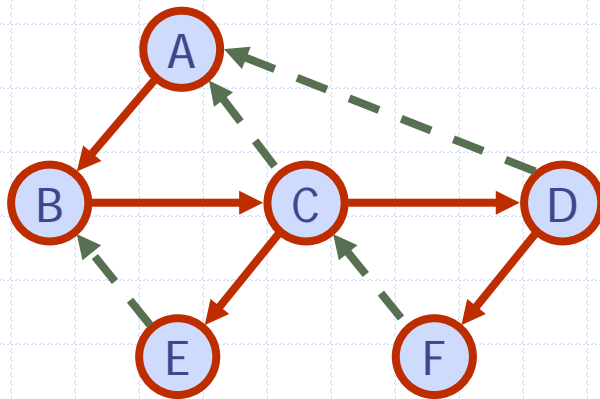


BFS

DFS vs. BFS (cont.)

Back edge (v, w)

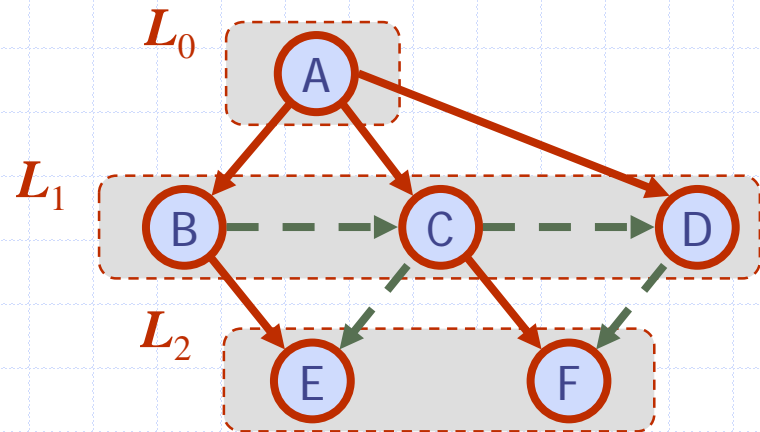
- w is an ancestor of v in the tree of discovery edges



DFS

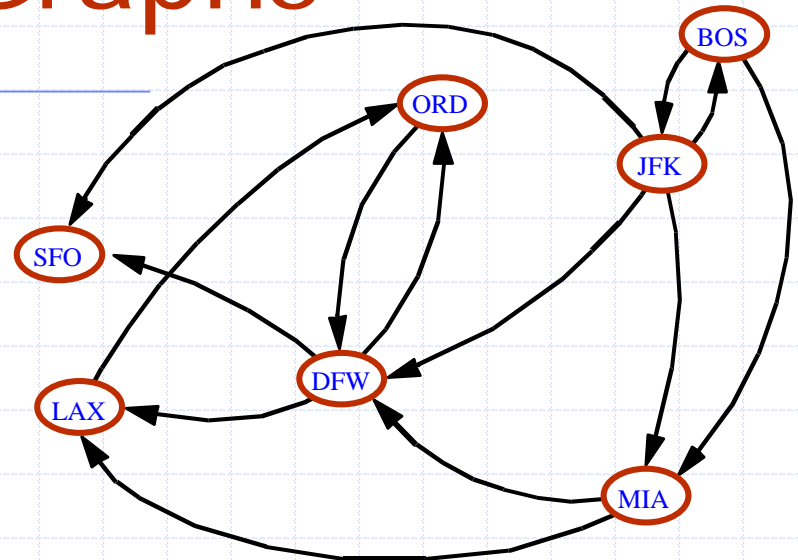
Cross edge (v, w)

- w is in the same level as v or in the next level in the tree of discovery edges



BFS

6.4 Directed Graphs



Outline and Reading (§6.4)

◆ Reachability (§6.4.1)

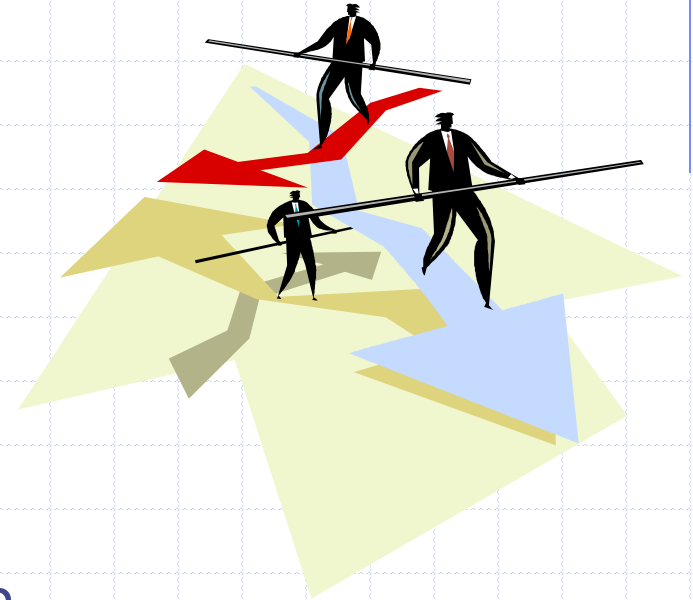
- Directed DFS
- Strong connectivity

◆ Transitive closure (§6.4.2)

- The Floyd-Warshall Algorithm

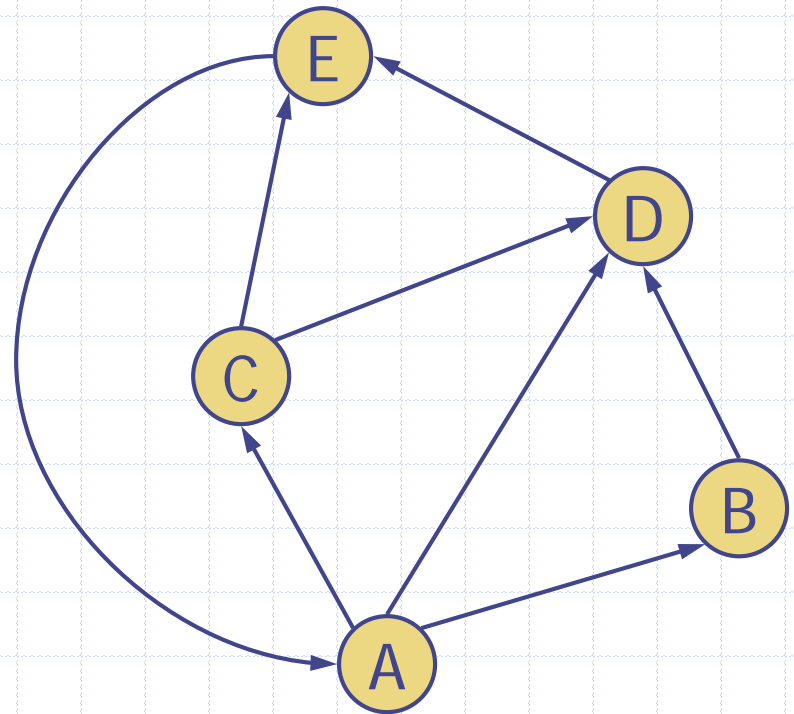
◆ Directed Acyclic Graphs (DAG's) (§6.4.4)

- Topological Sorting

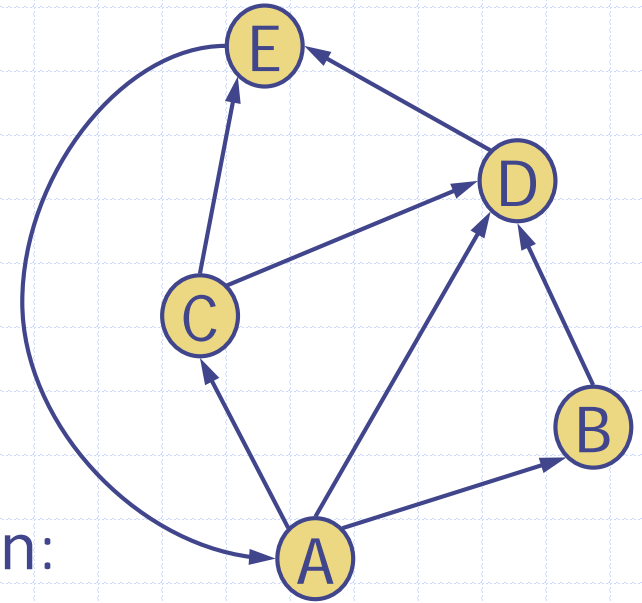


Digraphs

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



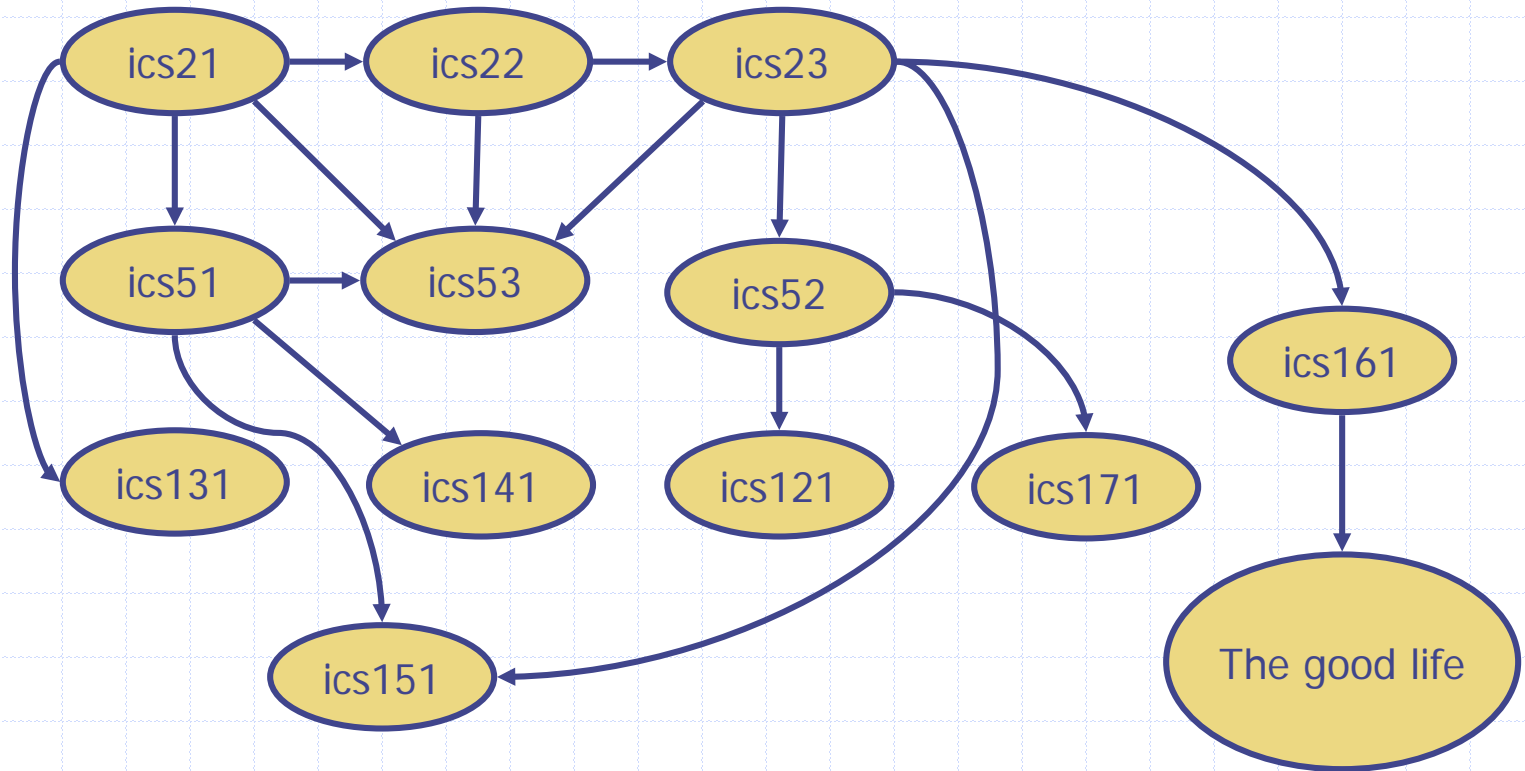
Digraph Properties



- ◆ 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, $m \leq n(n-1)$.
- ◆ If we keep in-edges and out-edges in separate adjacency lists, we can perform listing of the sets of in-edges and out-edges in time proportional to their size.

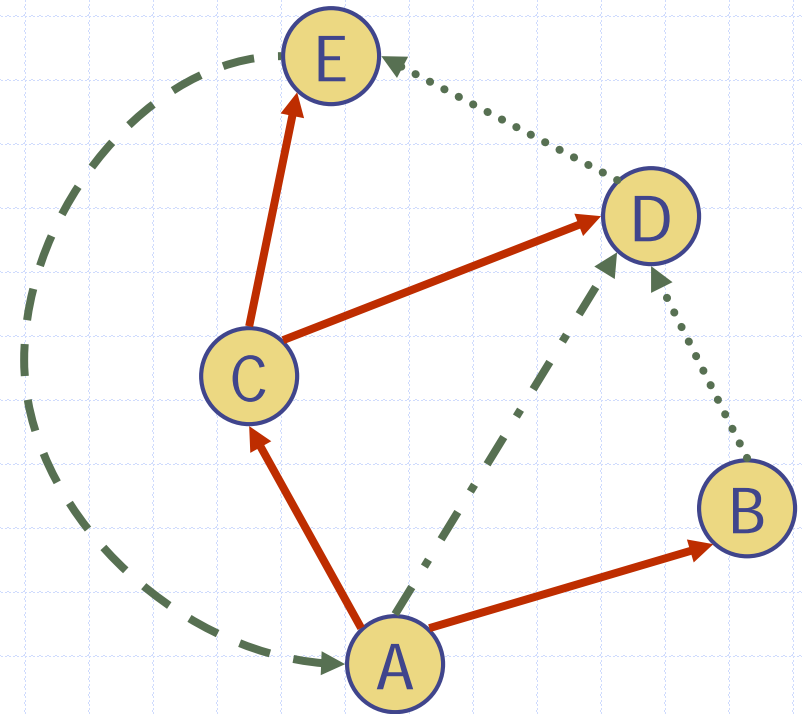
Digraph Application

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



Directed DFS

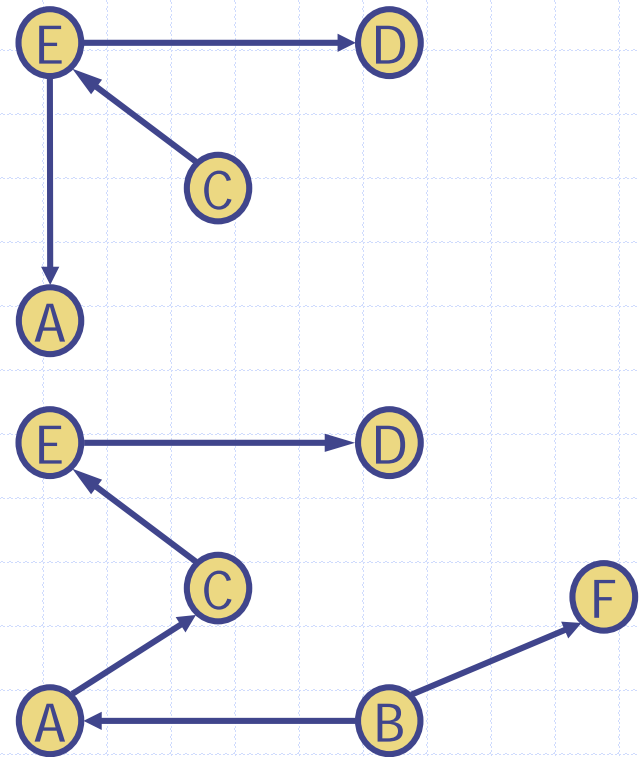
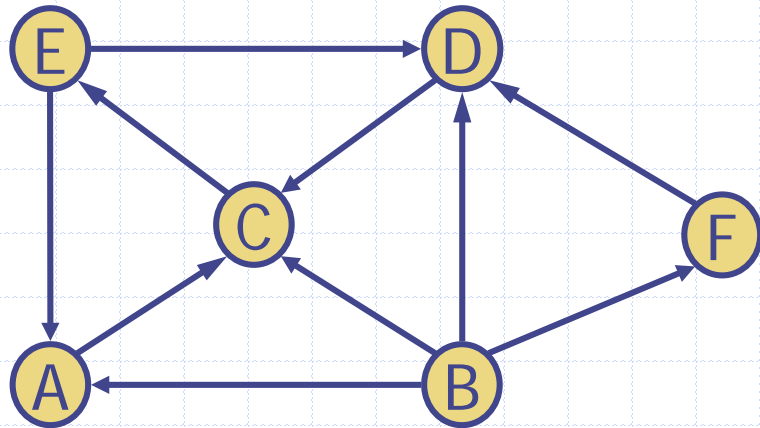
- ◆ We can specialize the traversal algorithms (DFS and BFS) to digraphs by traversing edges only along their direction
- ◆ In the directed DFS algorithm, we have four types of edges
 - discovery edges
 - back edges
 - forward edges
 - cross edges
- ◆ A directed DFS starting at a vertex s determines the vertices reachable from s

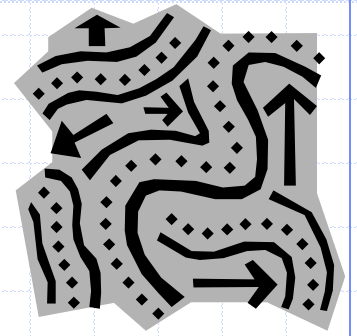




Reachability

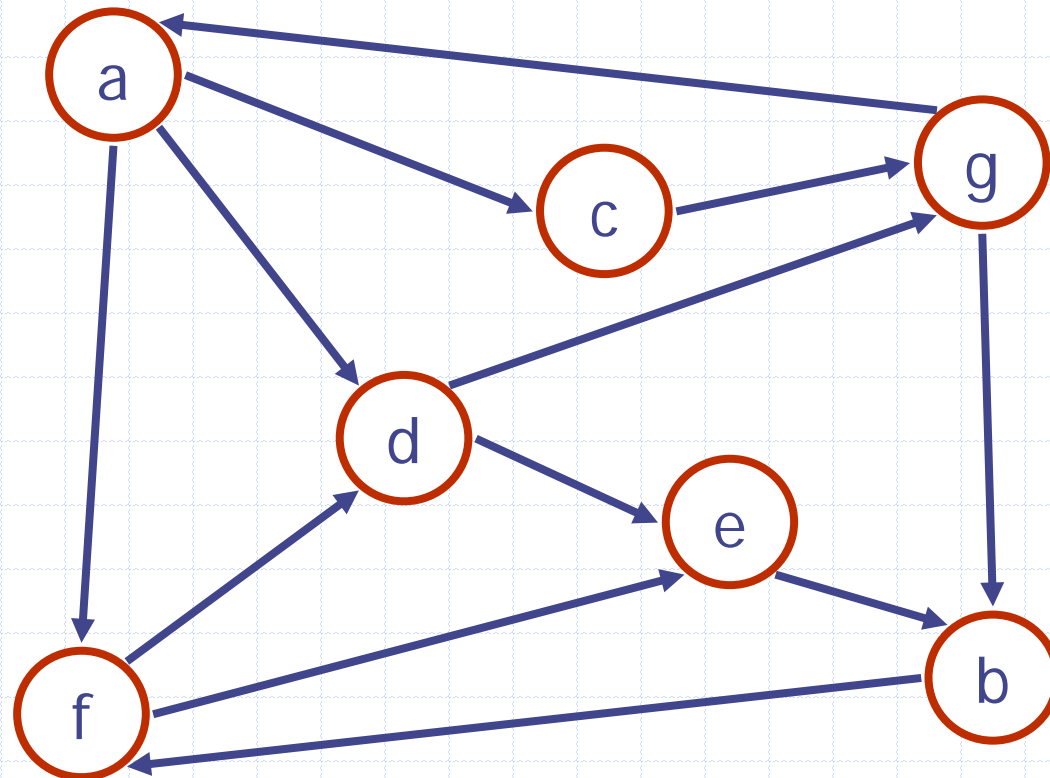
- ◆ DFS tree rooted at v : vertices reachable from v via directed paths



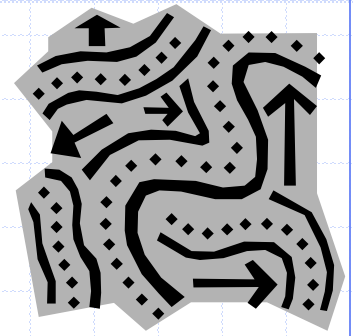


Strong Connectivity

- ◆ Each vertex can reach all other vertices

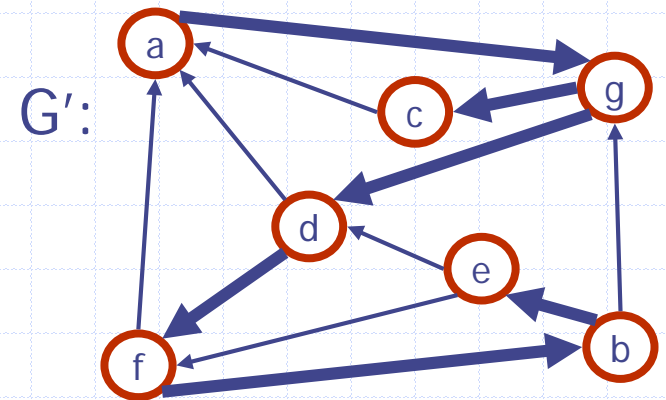
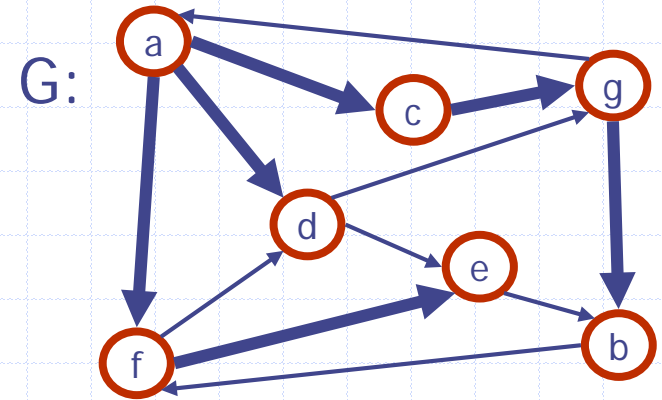


Strong Connectivity Algorithm

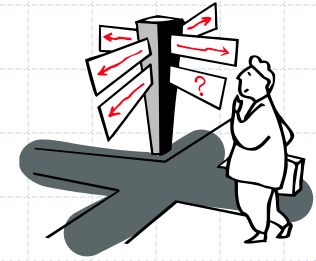


- ◆ Pick a vertex v in G .
- ◆ Perform a DFS from v in G .
 - If there's a w not visited, print "no".
- ◆ Let G' be G with edges reversed.
- ◆ Perform a DFS from v in G' .
 - If there's a w not visited, print "no".
 - Else, print "yes".

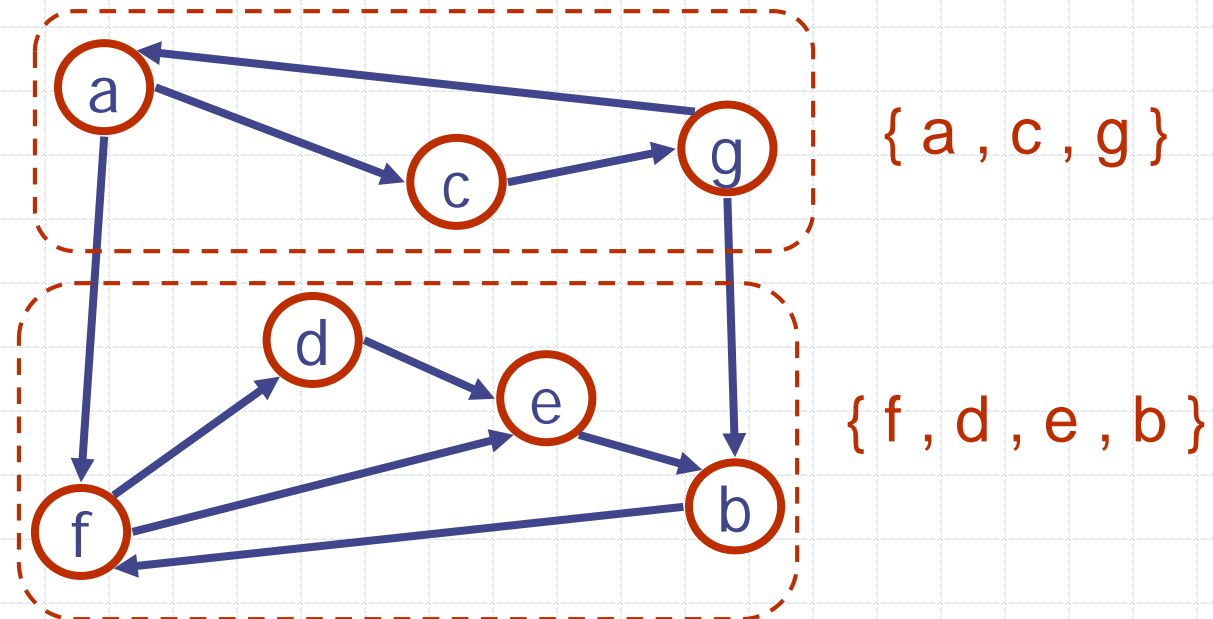
- ◆ Running time: $O(n+m)$.



Strongly Connected Components

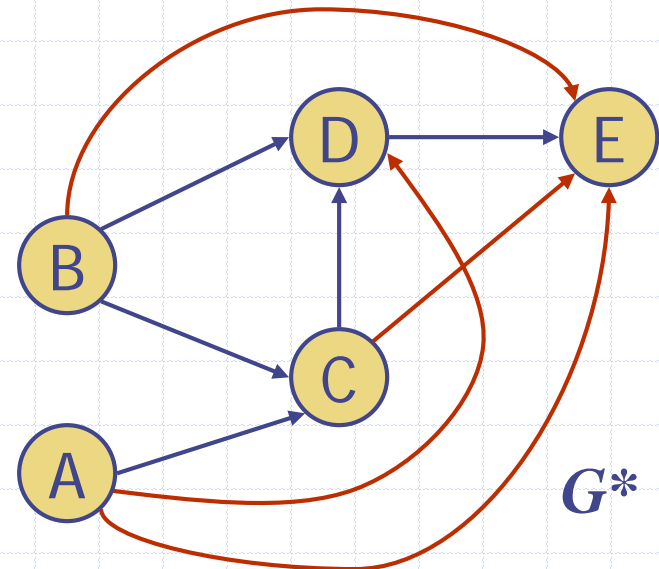
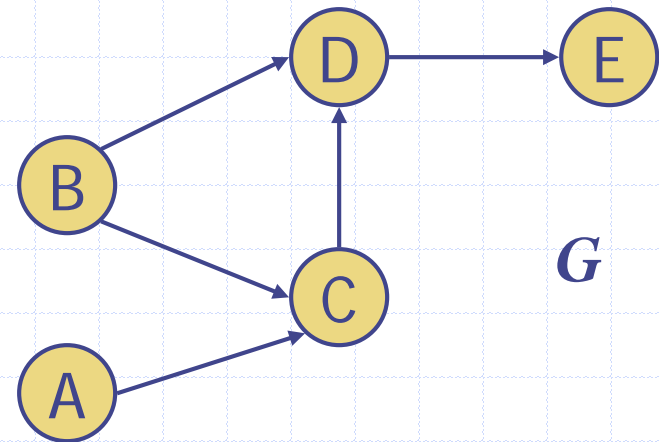


- ◆ Maximal subgraphs such that each vertex can reach all other vertices in the subgraph
- ◆ Can also be done in $O(n+m)$ time using DFS, but is more complicated (similar to biconnectivity).



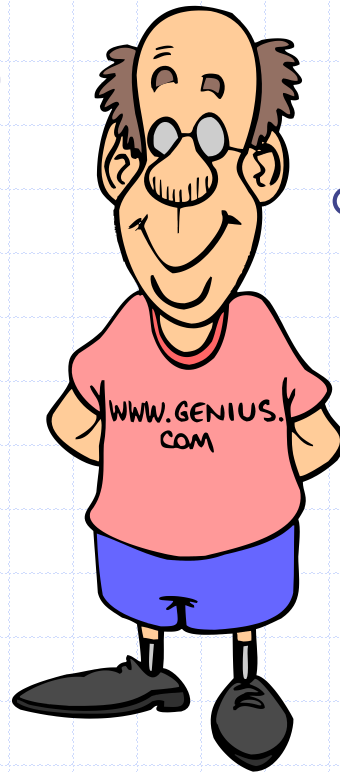
Transitive Closure

- ◆ 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

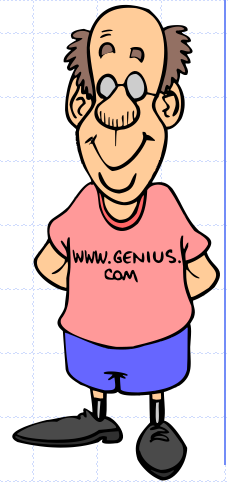
- ◆ We can perform DFS starting at each vertex
 - $O(n(n+m))$



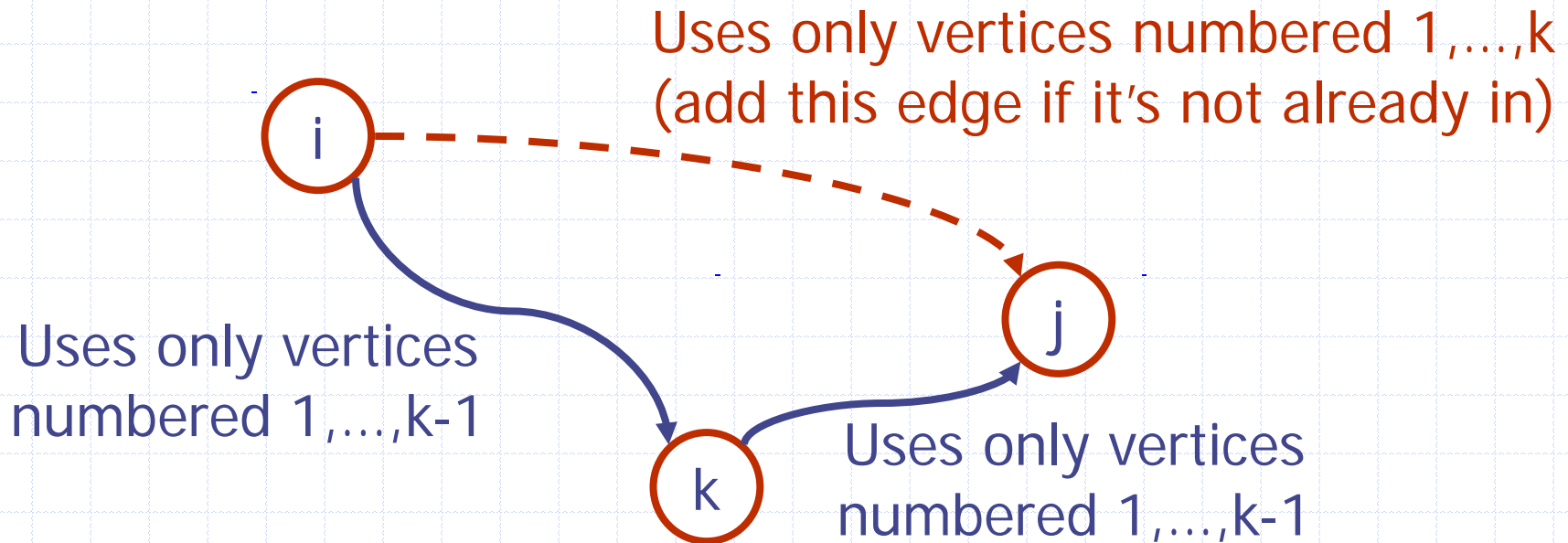
If there's a way to get from **A** to **B** and from **B** to **C**, then there's a way to get from **A** to **C**.

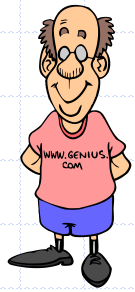
- ◆ Alternatively ... Use dynamic programming: the Floyd-Warshall Algorithm

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

- ◆ Floyd-Warshall's algorithm numbers the vertices of G as v_1, \dots, v_n and computes a series of digraphs G_0, \dots, G_n
 - $G_0 = G$
 - G_k has a directed edge (v_i, v_j) if G has a directed path from v_i to v_j with intermediate vertices in the set $\{v_1, \dots, v_k\}$
- ◆ 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 $v \in G.vertices()$

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

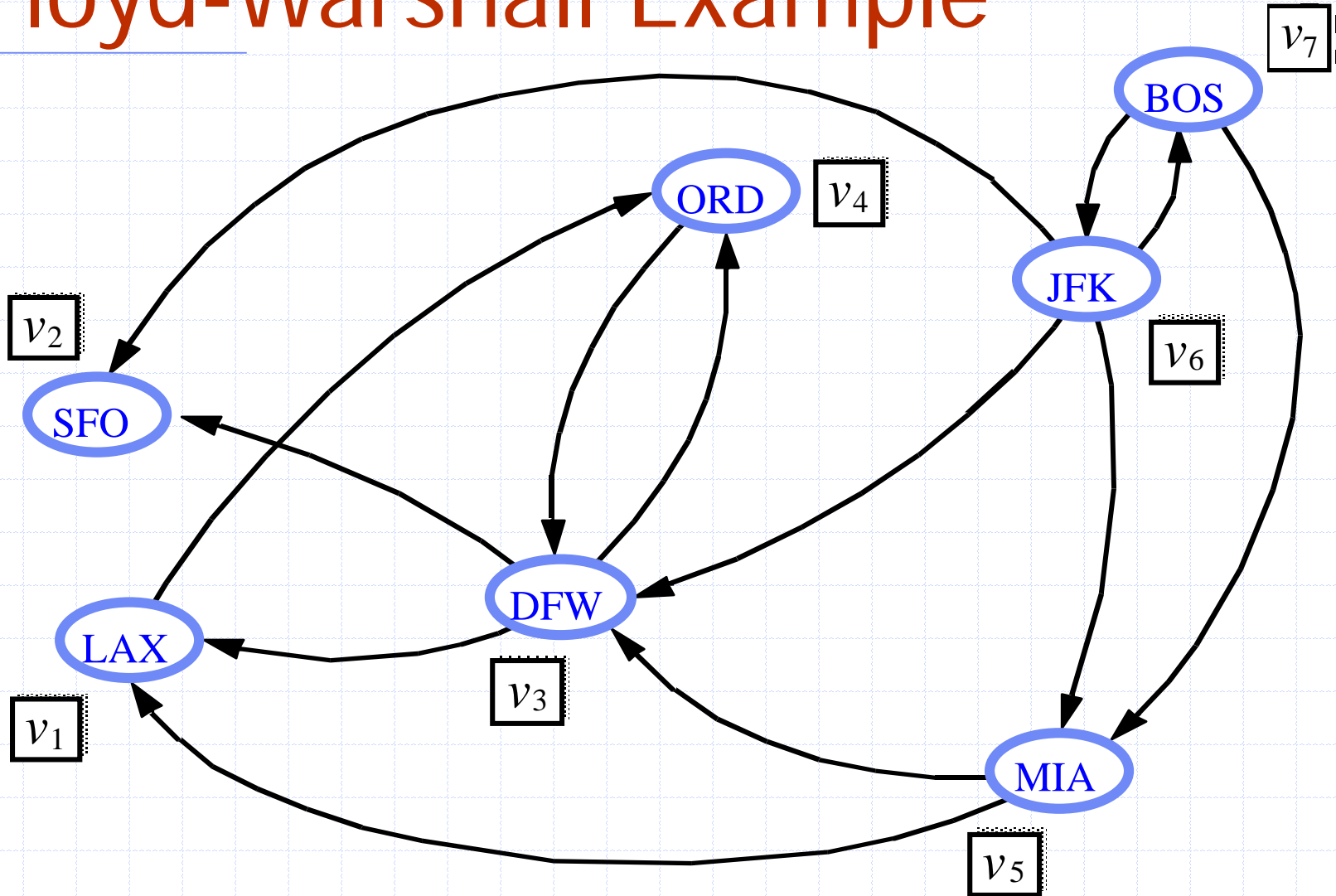
$G_{k-1}.areAdjacent(v_k, v_j)$

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

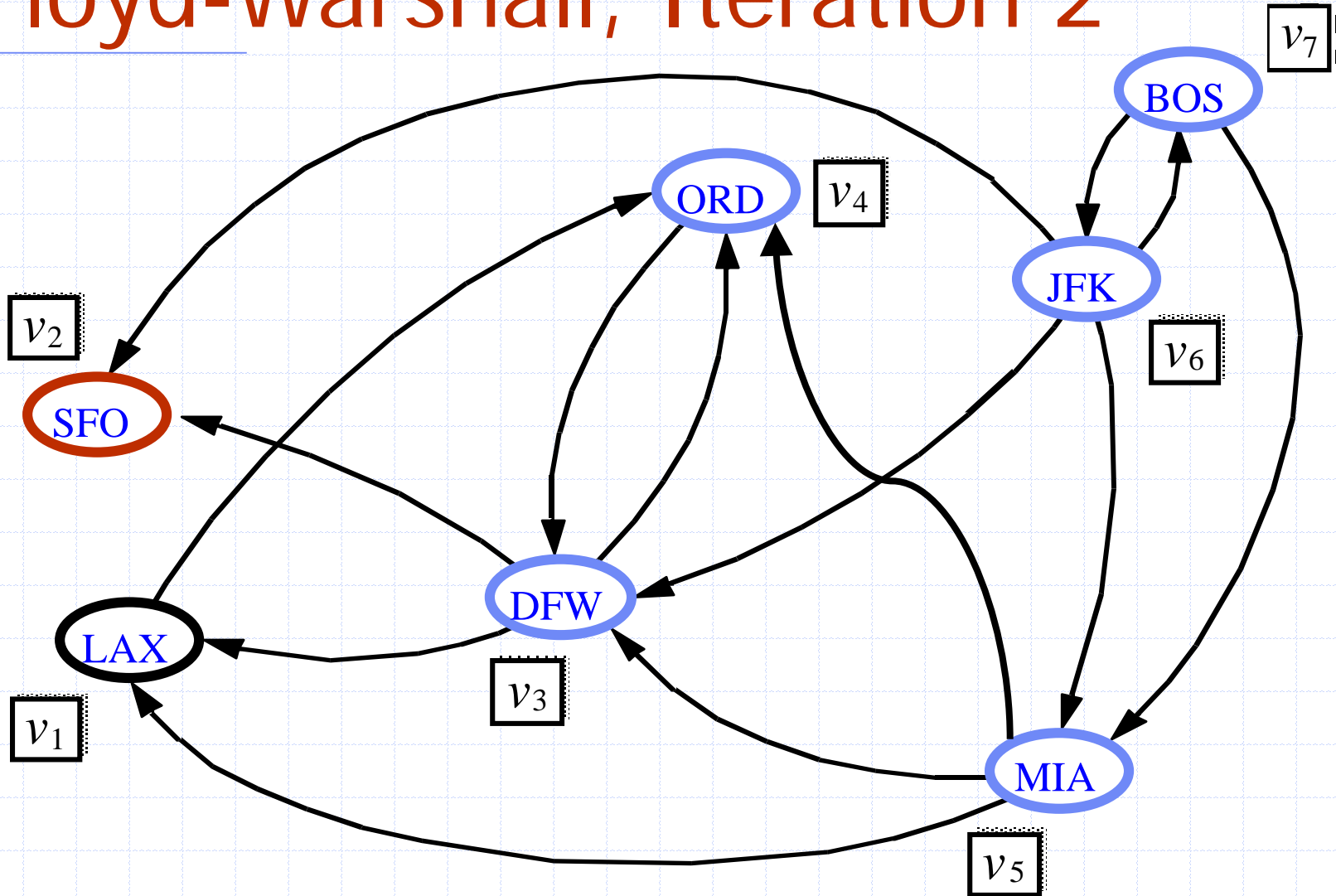
$G_k.insertDirectedEdge(v_i, v_j, k)$

return G_n

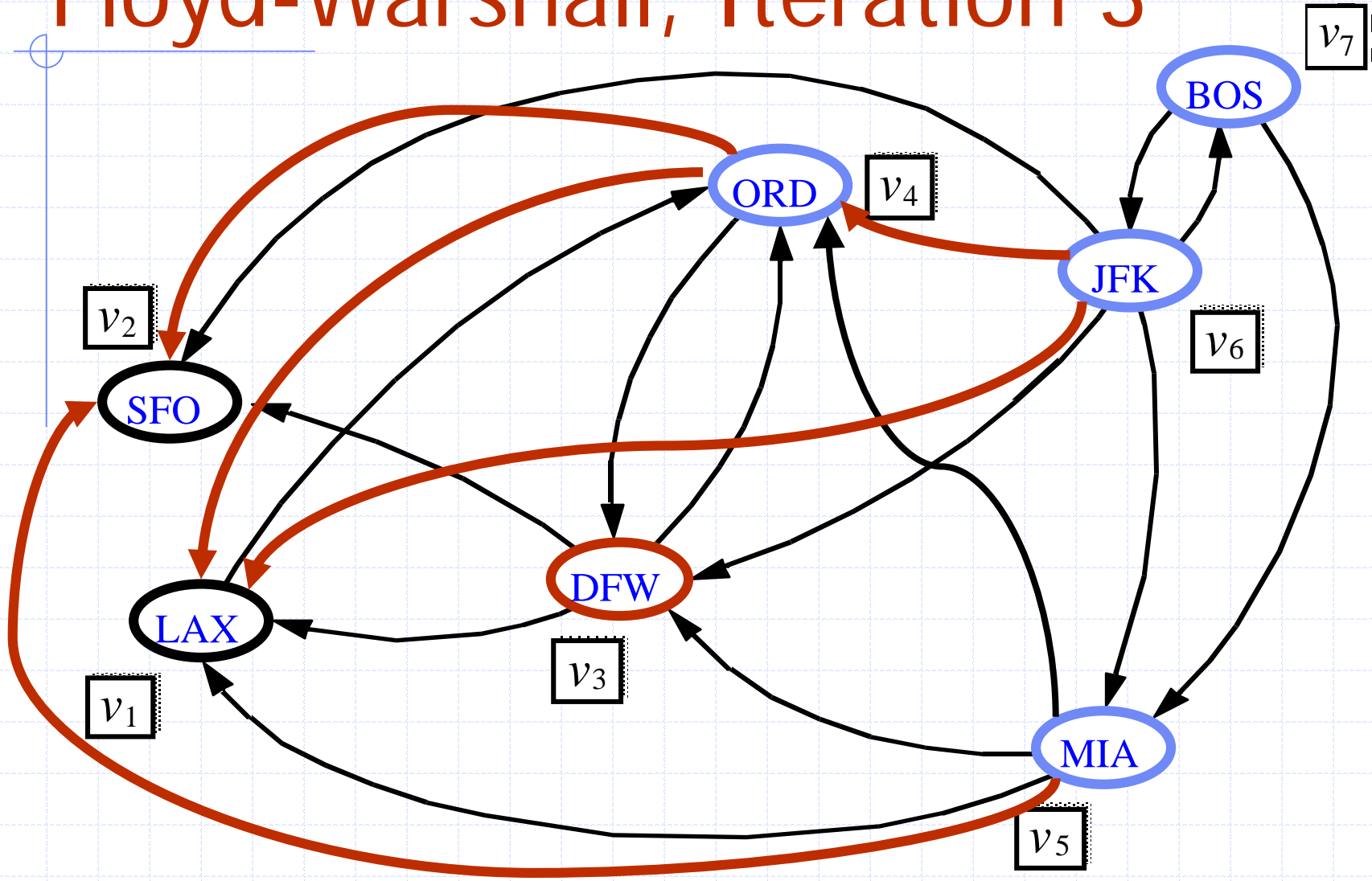
Floyd-Warshall Example



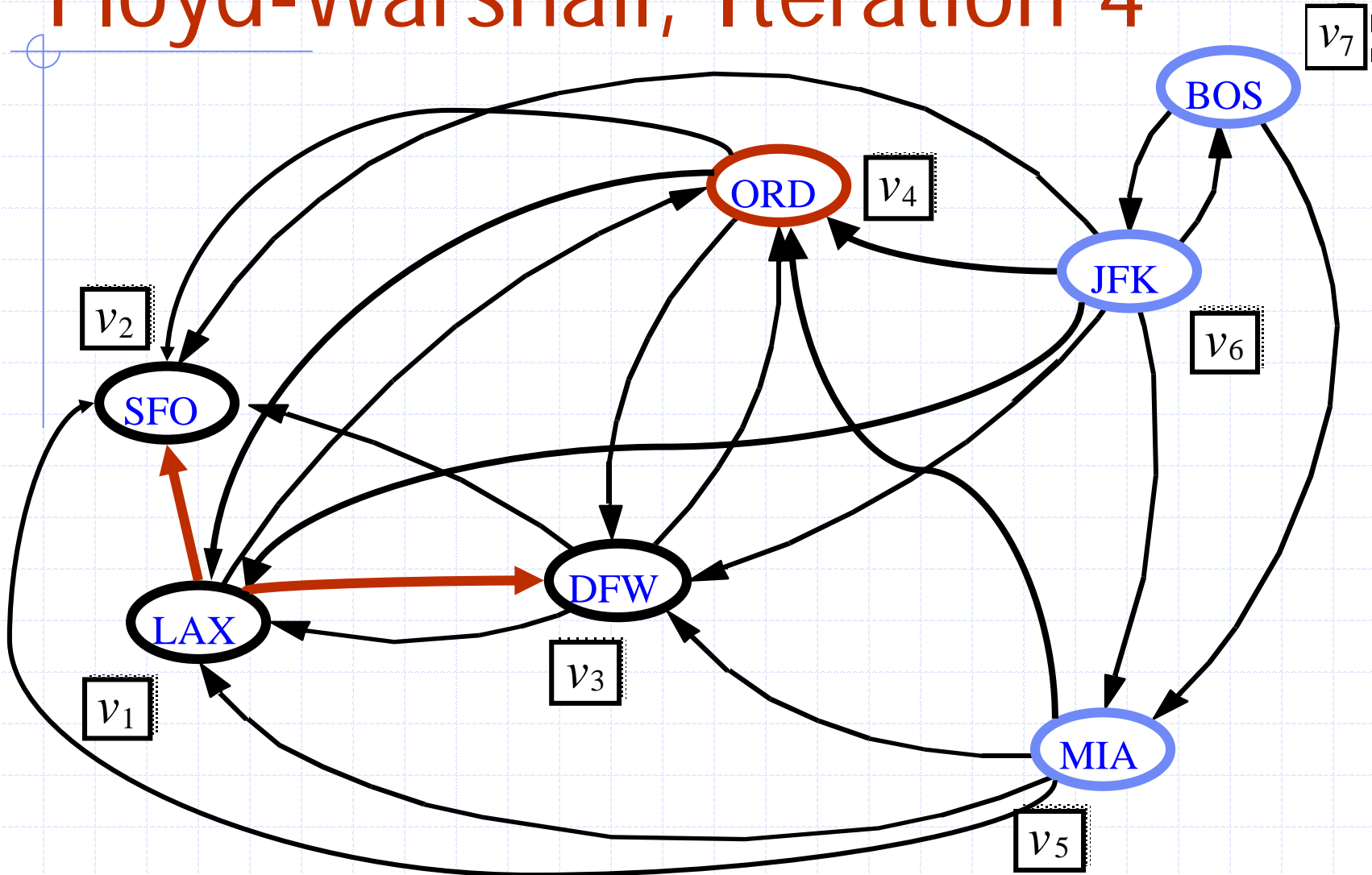
Floyd-Warshall, Iteration 2



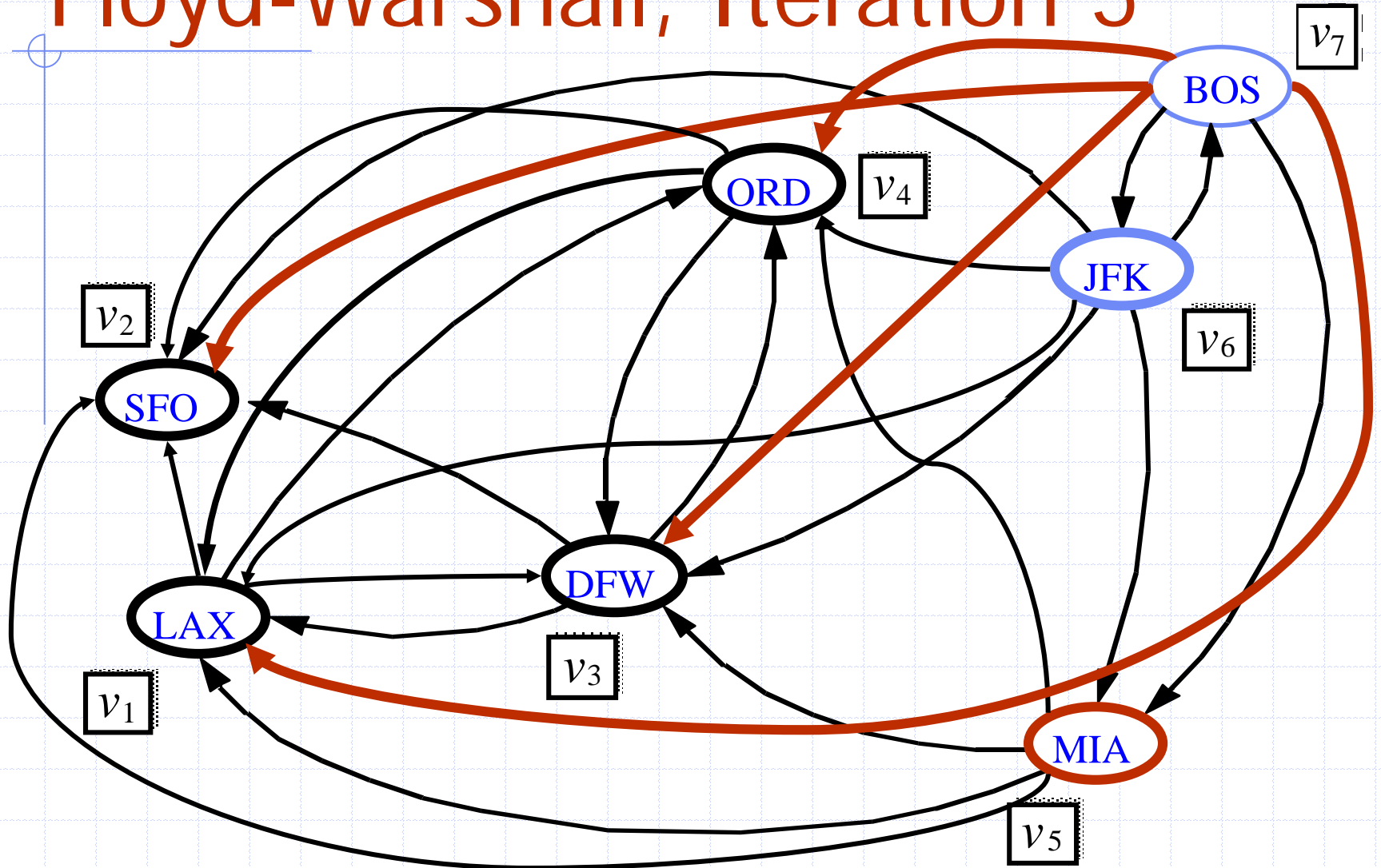
Floyd-Warshall, Iteration 3



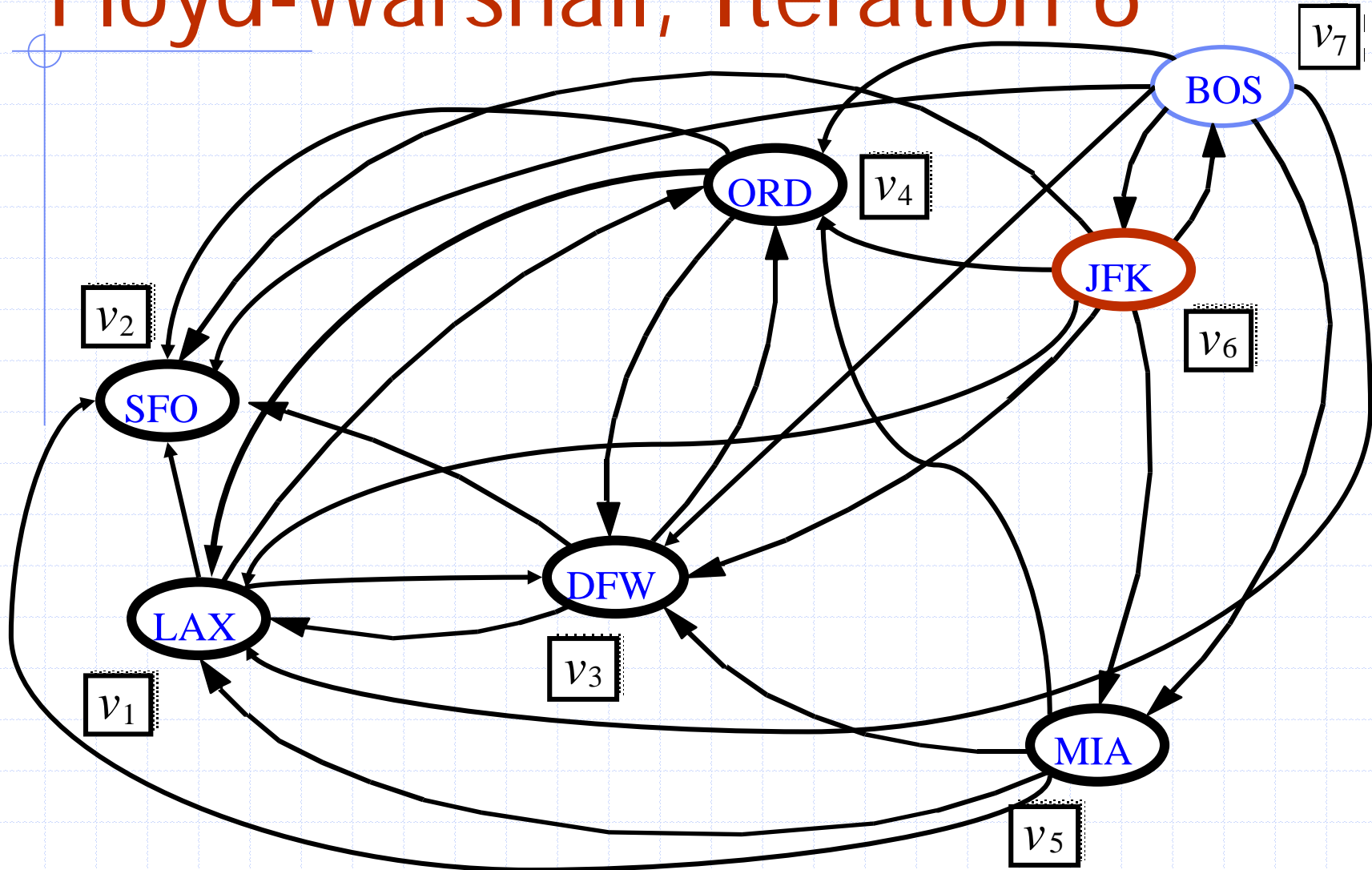
Floyd-Warshall, Iteration 4



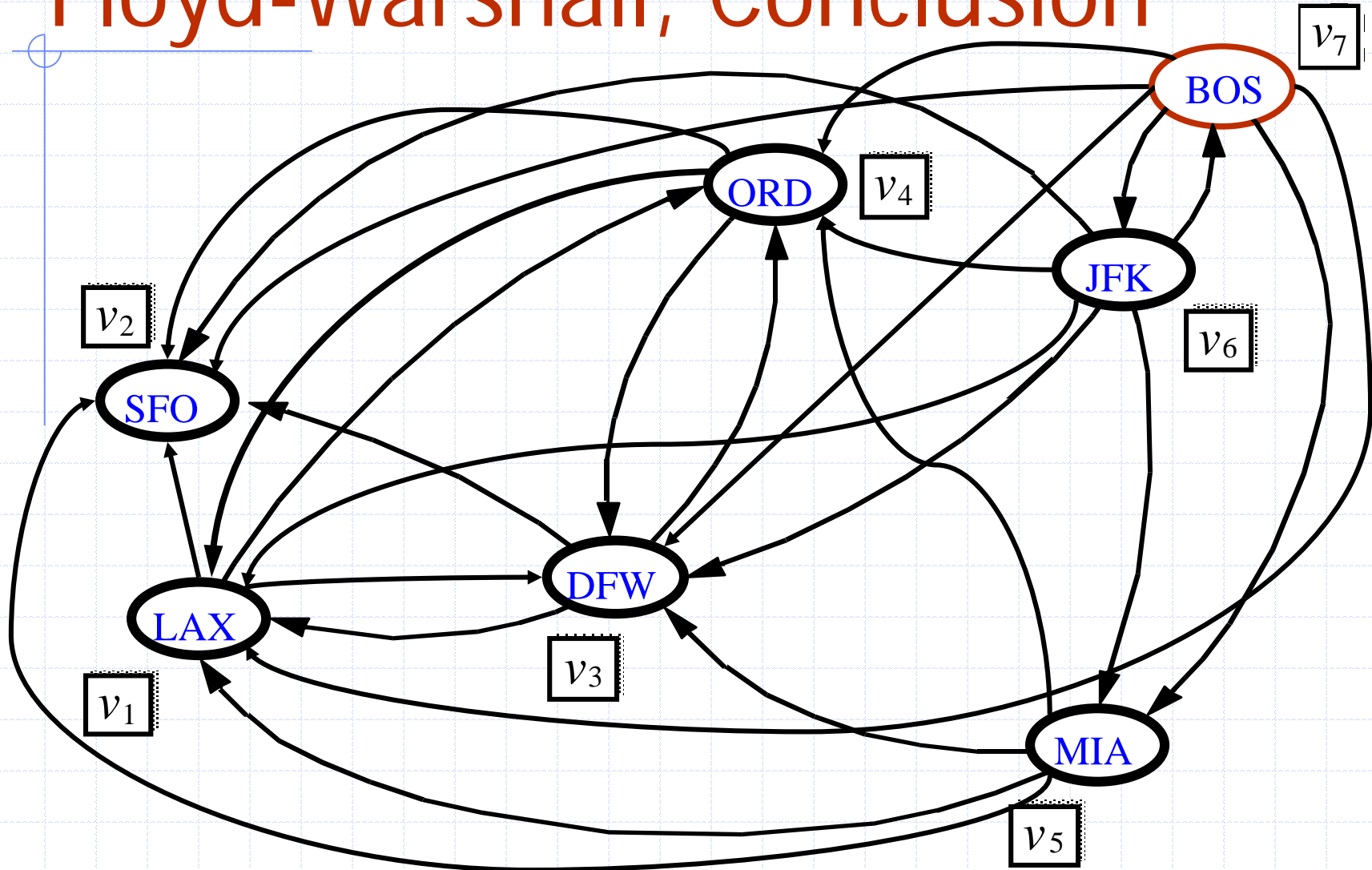
Floyd-Warshall, Iteration 5



Floyd-Warshall, Iteration 6



Floyd-Warshall, Conclusion



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

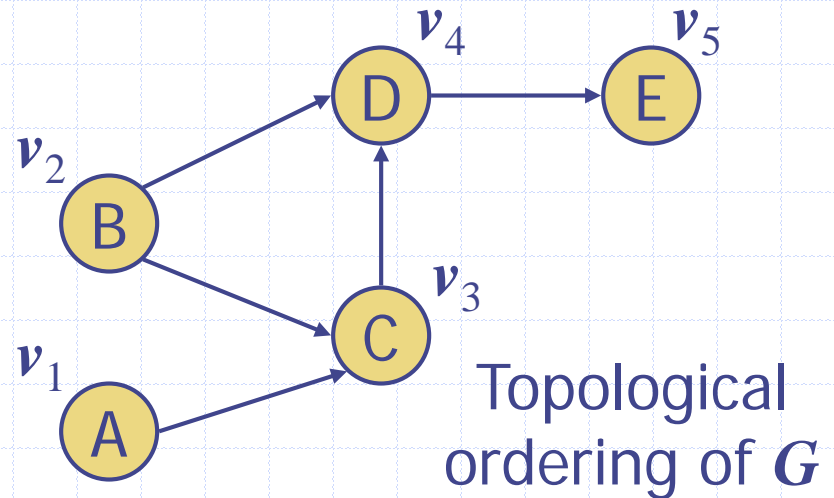
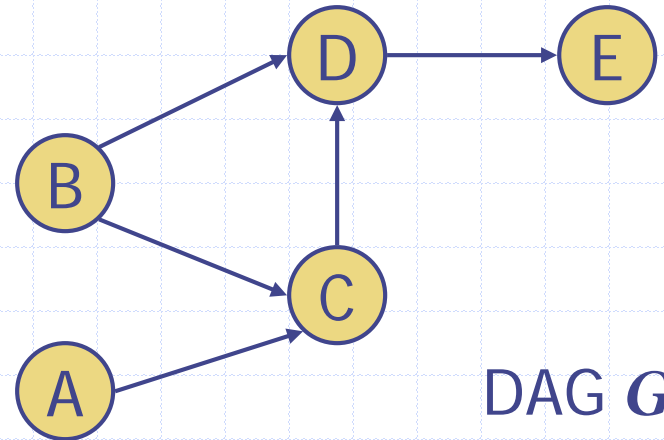
$$v_1, \dots, v_n$$

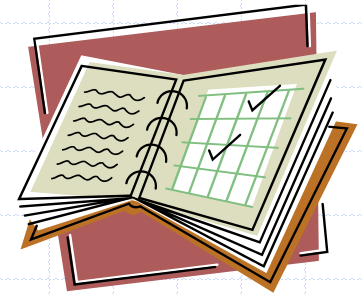
of the vertices such that for every edge (v_i, v_j) , we have $i < j$

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

Theorem

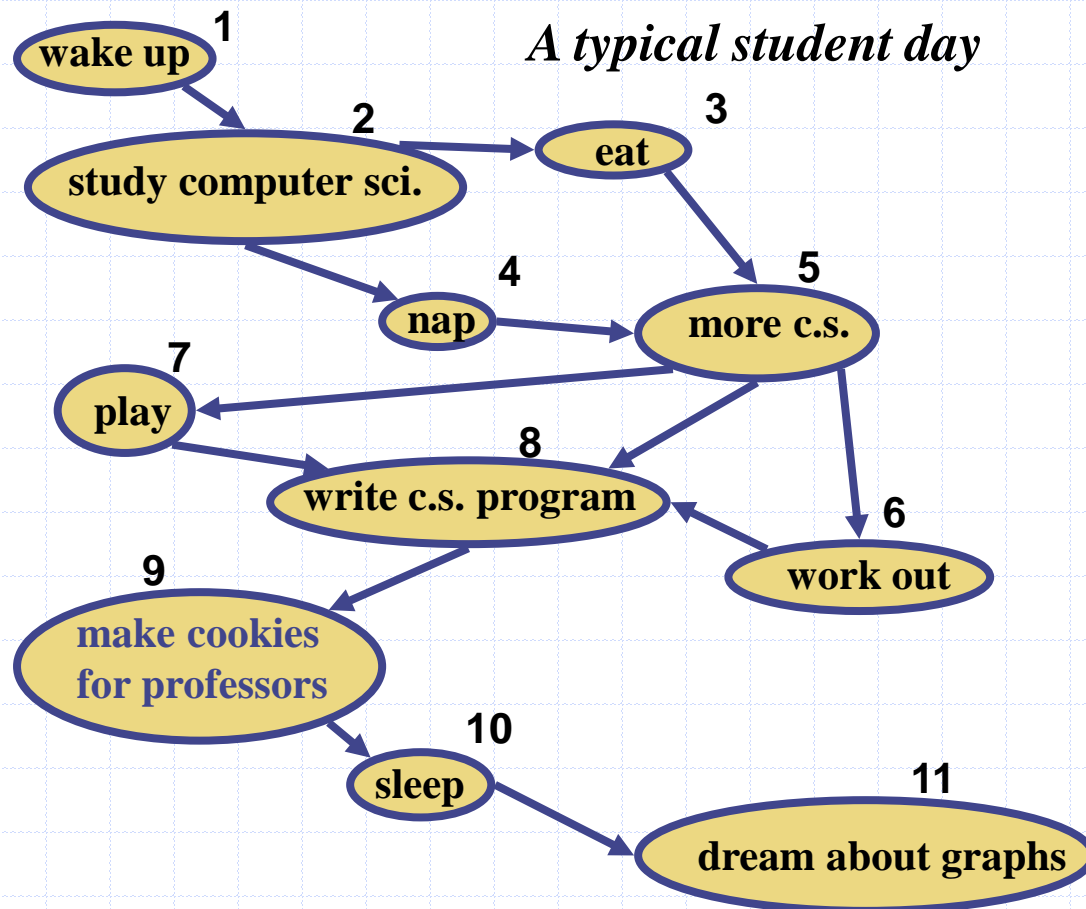
A digraph admits a topological ordering if and only if it is a DAG





Topological Sorting

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



Algorithm for Topological Sorting

- ◆ Note: This algorithm is different than the one in Goodrich-Tamassia

```
Method 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: $O(n + m)$. How...?

Topological Sorting Algorithm using DFS

- ◆ Simulate the algorithm by using depth-first search

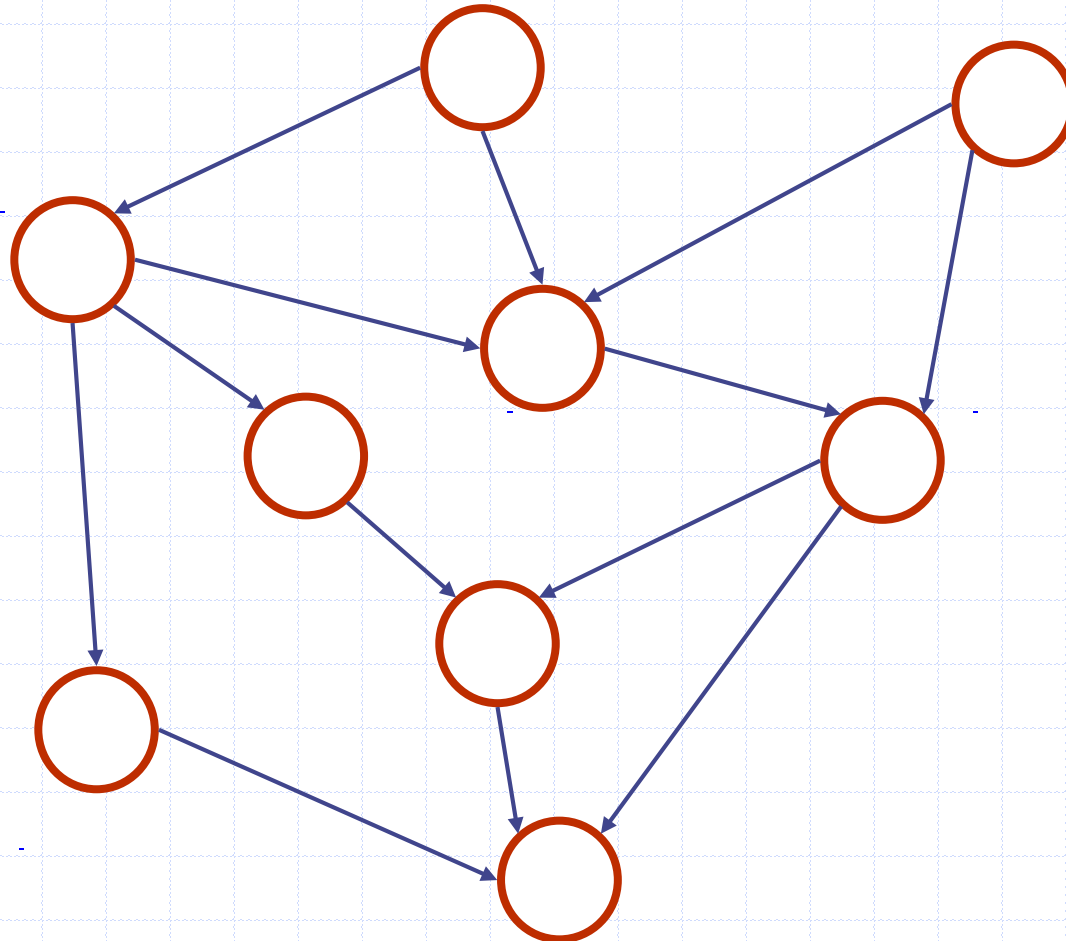
```
Algorithm topologicalDFS(G)  
  Input dag  $G$   
  Output topological ordering of  $G$   
   $n \leftarrow G.numVertices()$   
  for all  $u \in G.vertices()$   
    setLabel(u, UNEXPLORED)  
  for all  $e \in G.edges()$   
    setLabel(e, UNEXPLORED)  
  for all  $v \in G.vertices()$   
    if getLabel(v) = UNEXPLORED  
      topologicalDFS(G, v)
```

- ◆ $O(n+m)$ time.

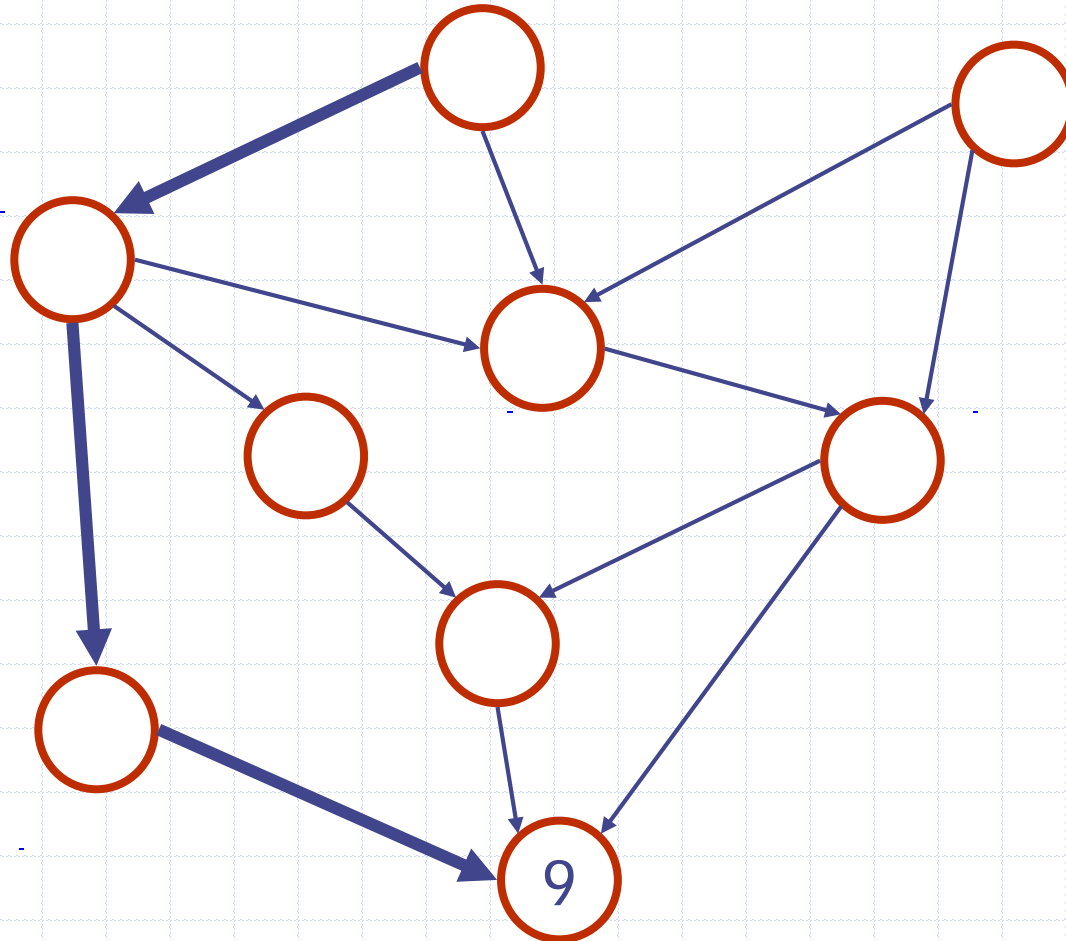
```
Algorithm topologicalDFS(G, v)
```

```
  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  $e \in G.incidentEdges(v)$   
    if getLabel(e) = UNEXPLORED  
       $w \leftarrow opposite(v, e)$   
      if getLabel(w) = UNEXPLORED  
        setLabel(e, DISCOVERY)  
        topologicalDFS(G, w)  
      else  
        { $e$  is a forward or cross edge}  
  Label  $v$  with topological number  $n$   
   $n \leftarrow n - 1$ 
```

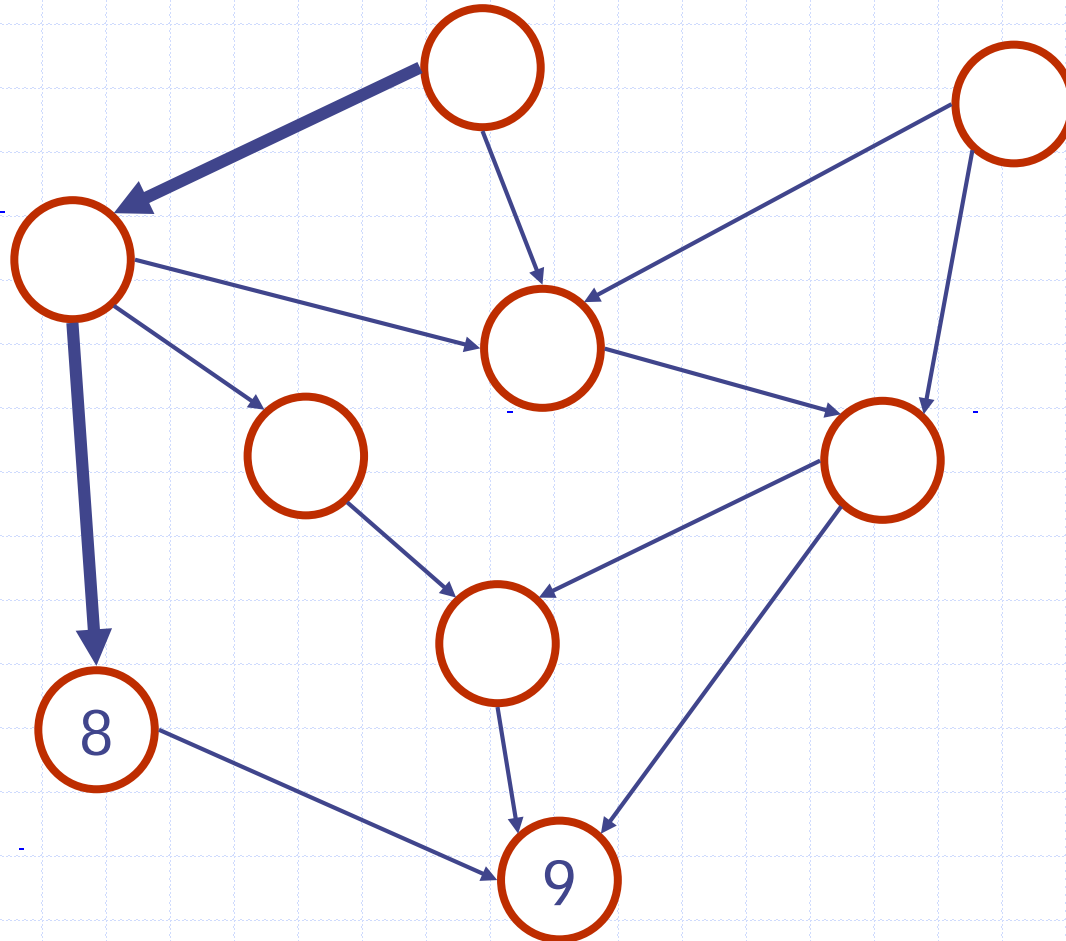
Topological Sorting Example



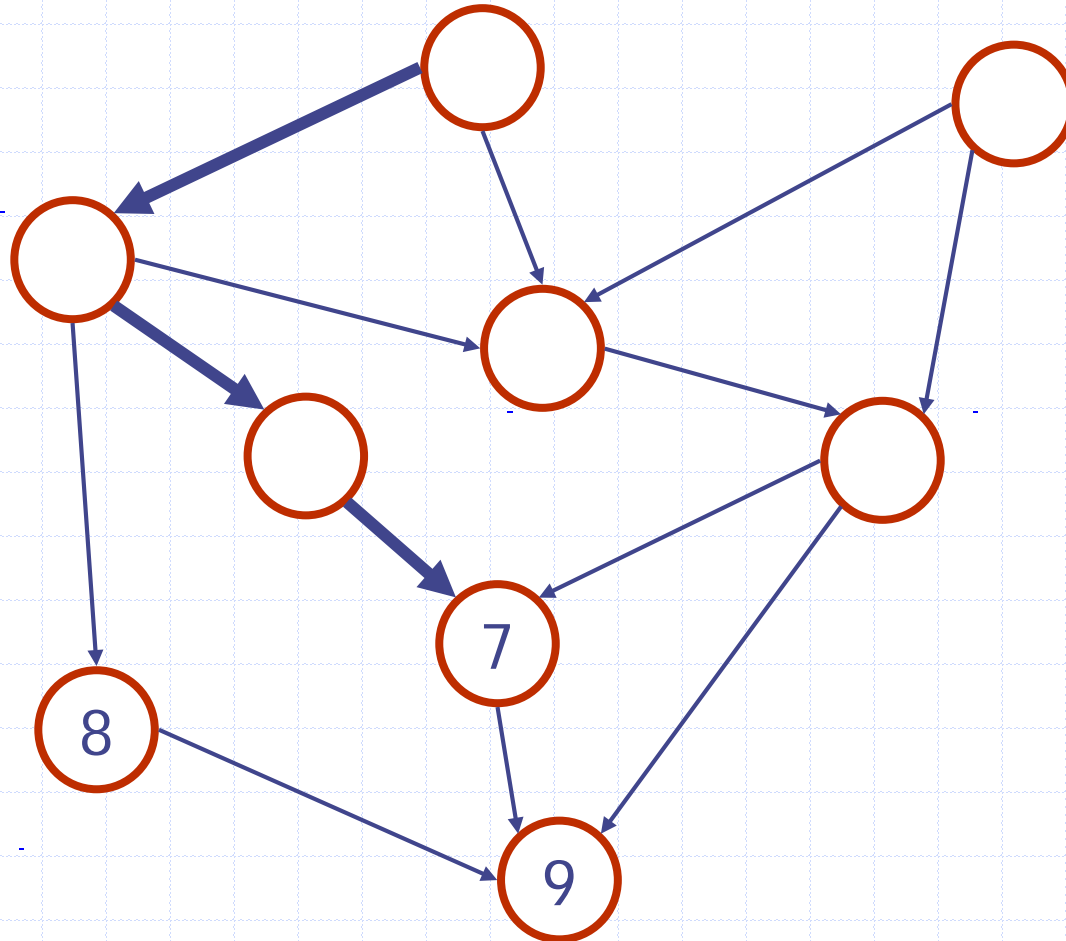
Topological Sorting Example



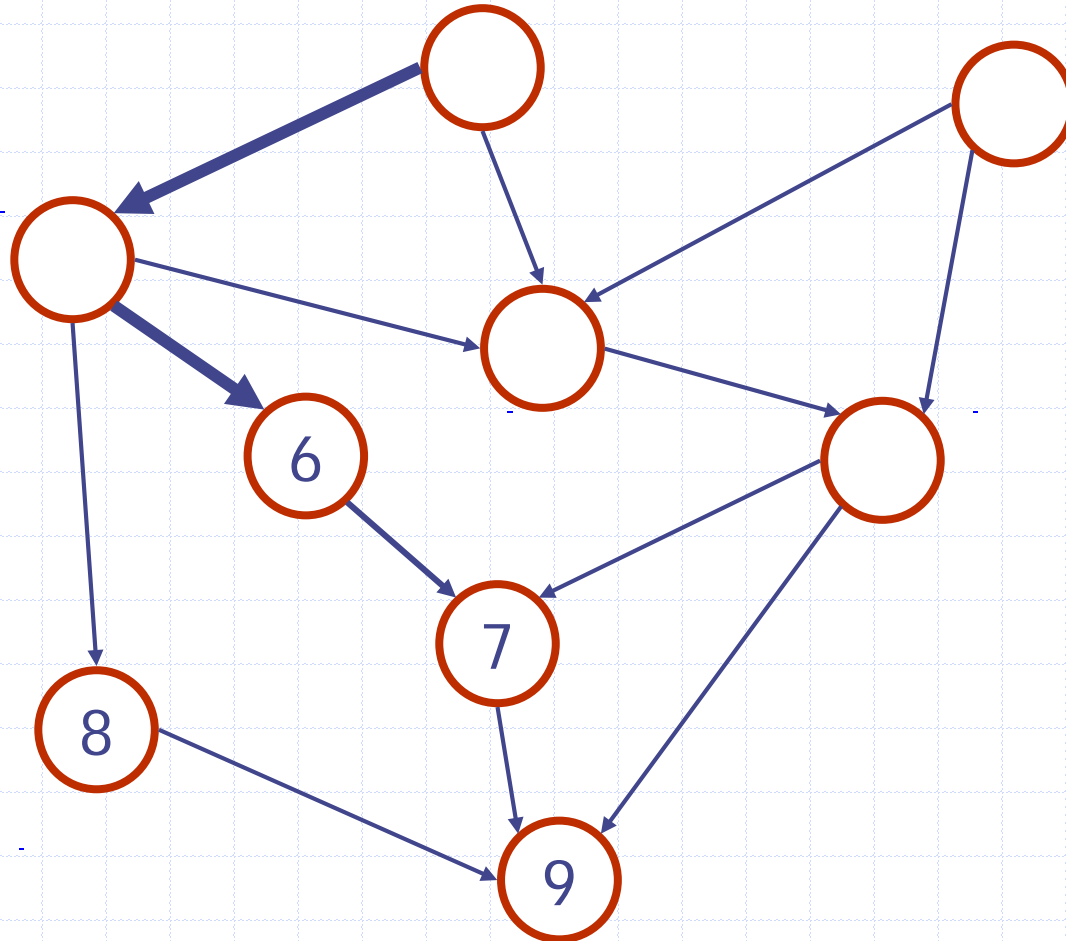
Topological Sorting Example



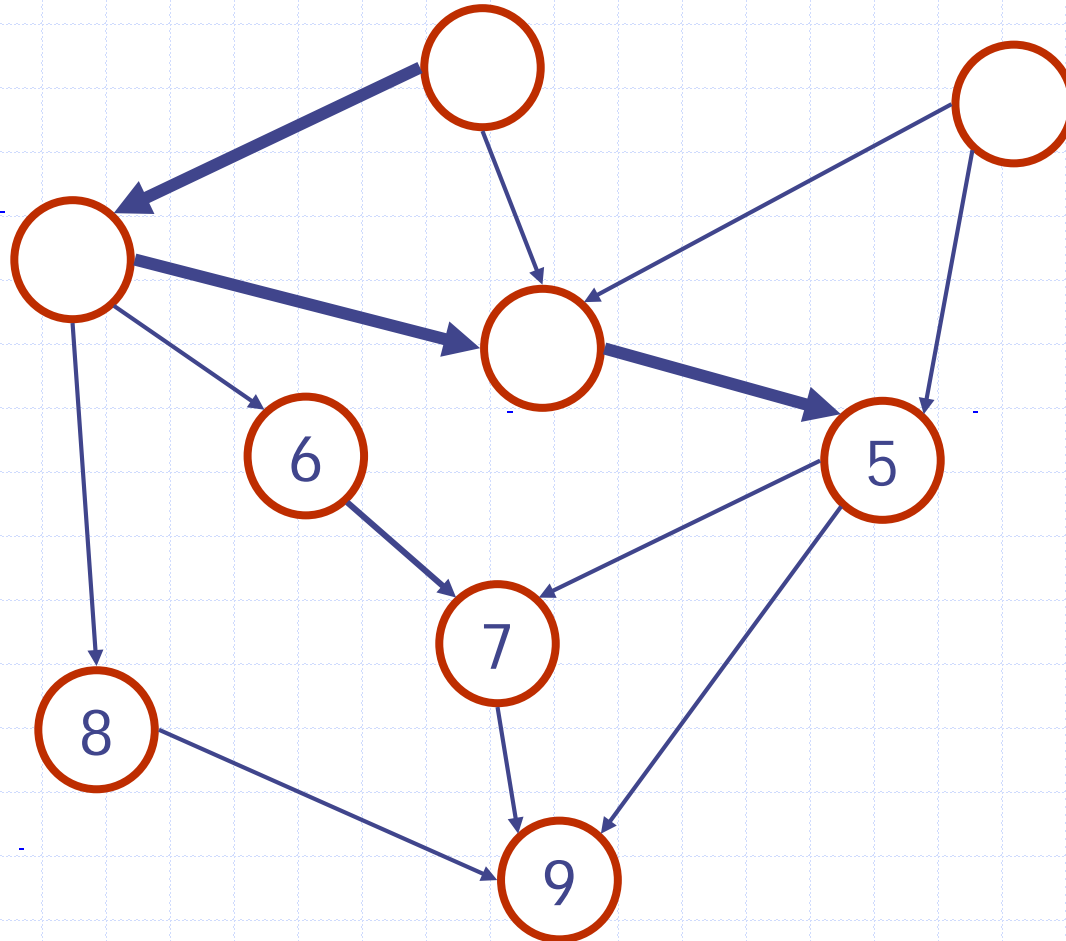
Topological Sorting Example



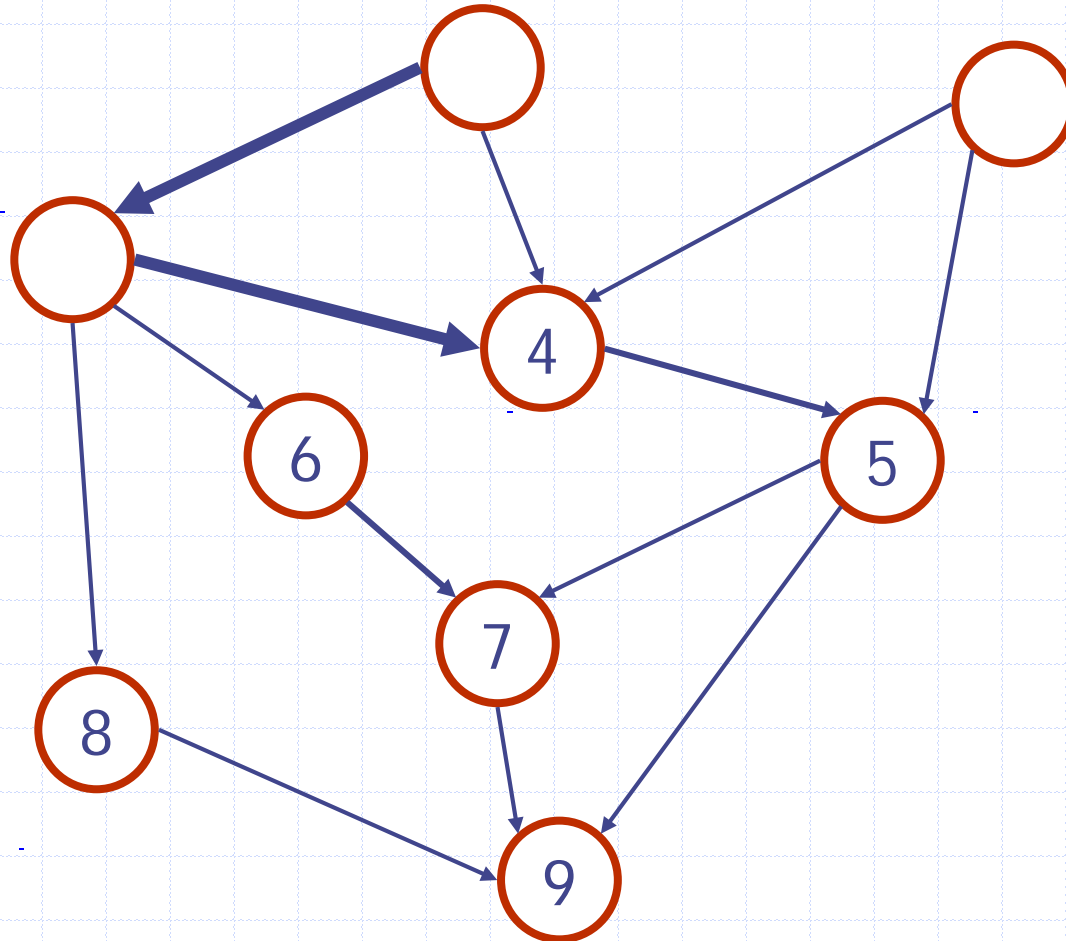
Topological Sorting Example



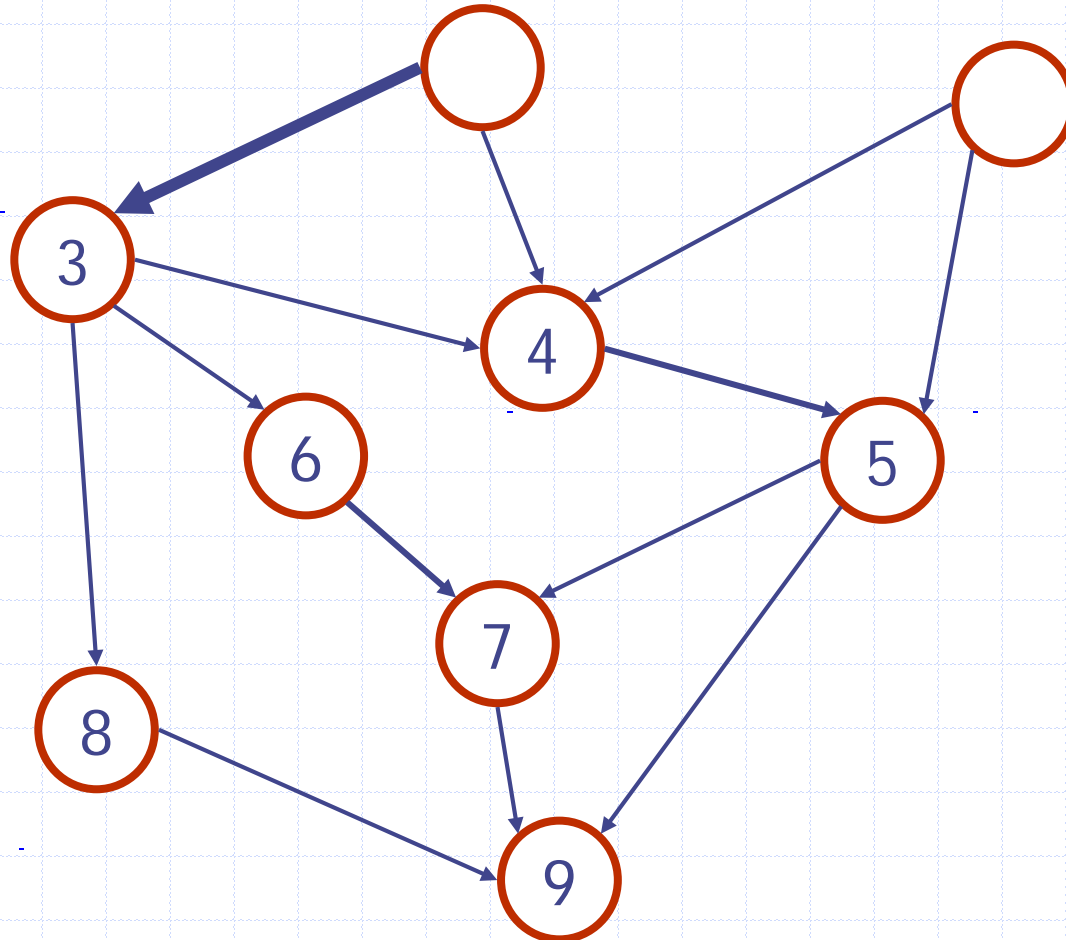
Topological Sorting Example



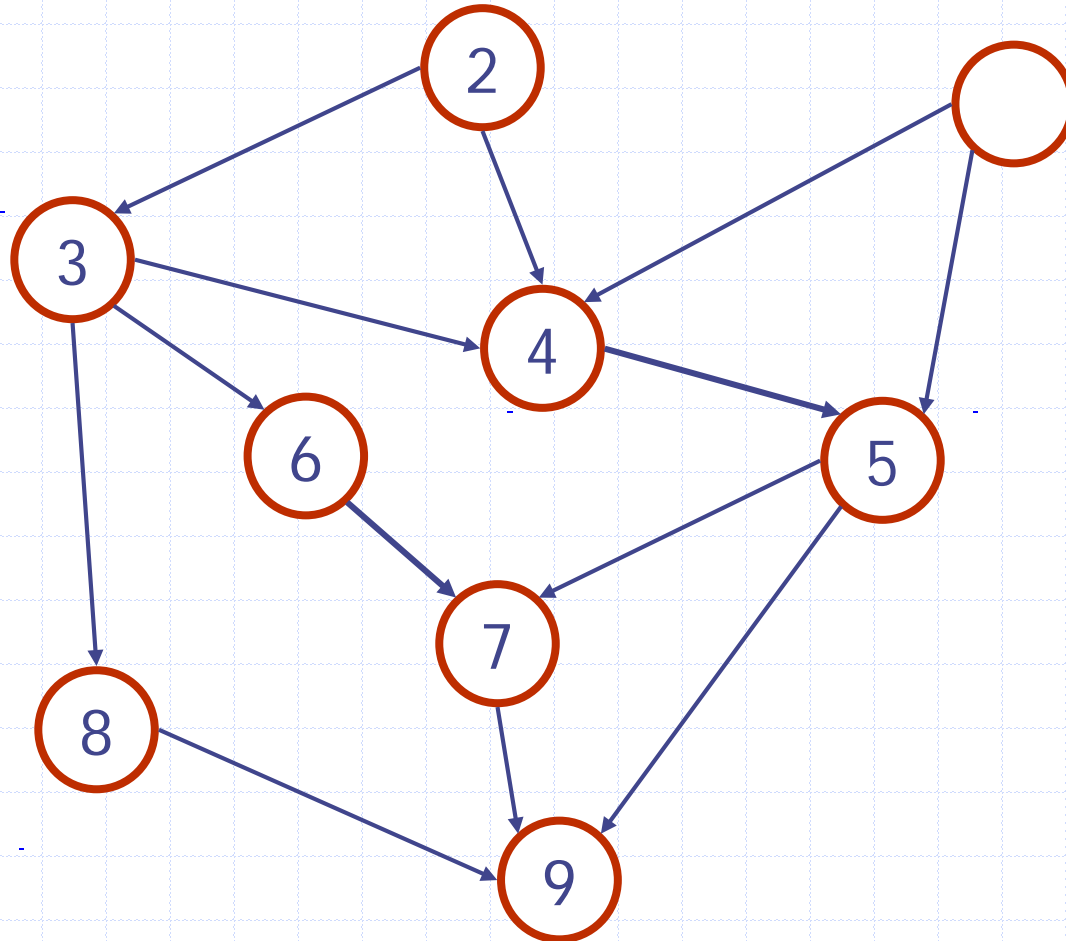
Topological Sorting Example



Topological Sorting Example



Topological Sorting Example



Topological Sorting Example

