

15-0: **Graphs**

- A graph consists of:
 - A set of **nodes** or **vertices** (terms are interchangeable)
 - A set of **edges** or **arcs** (terms are interchangeable)
- Edges in graph can be either directed or undirected

15-1: **Graphs & Edges**

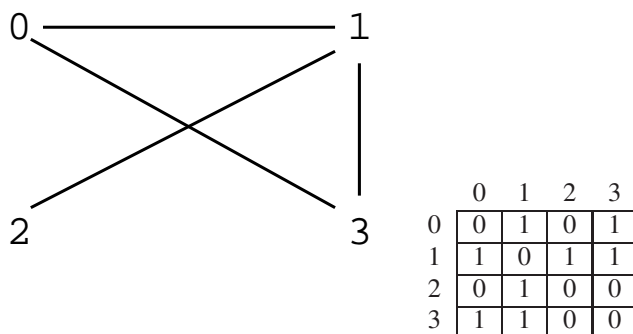
- Edges can be labeled or unlabeled
 - Edge labels are typically the *cost* associated with an edge
 - e.g., Nodes are cities, edges are roads between cities, edge label is the length of road

15-2: **Graph Representations**

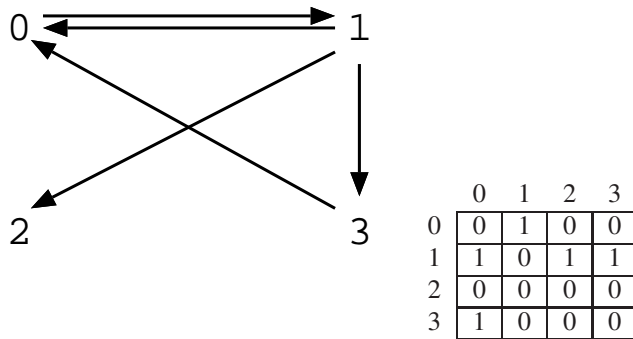
- Adjacency Matrix
- Represent a graph with a two-dimensional array G
 - $G[i][j] = 1$ if there is an edge from node i to node j
 - $G[i][j] = 0$ if there is no edge from node i to node j
- If graph is undirected, matrix is symmetric
- Can represent edges labeled with a cost as well:
 - $G[i][j] = \text{cost of link between } i \text{ and } j$
 - If there is no direct link, $G[i][j] = \infty$

15-3: **Adjacency Matrix**

- Examples:

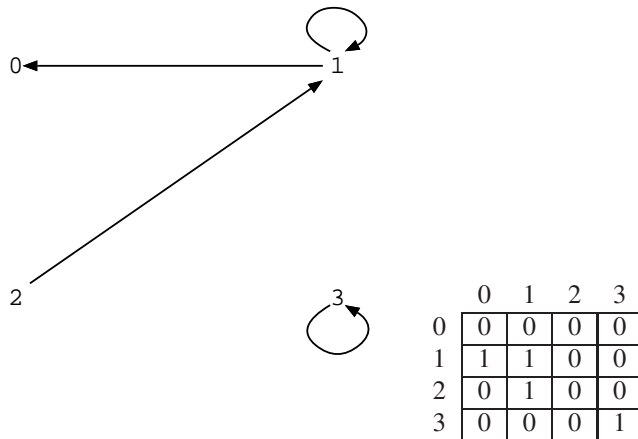
15-4: **Adjacency Matrix**

- Examples:



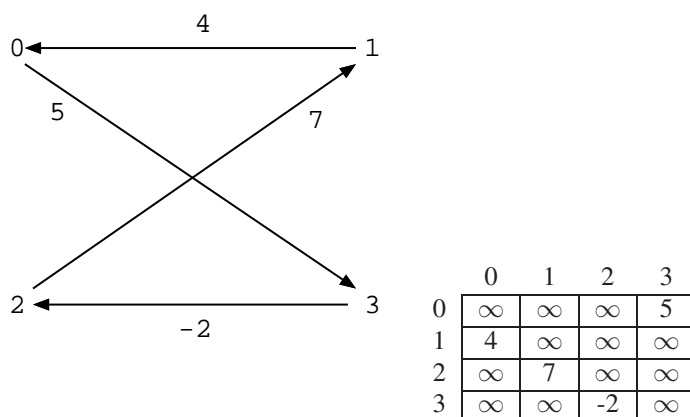
15-5: Adjacency Matrix

- Examples:



15-6: Adjacency Matrix

- Examples:



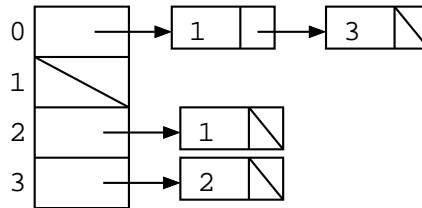
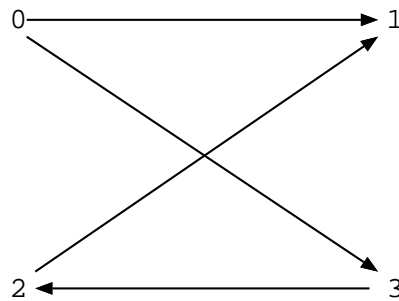
15-7: Graph Representations

- Adjacency List
- Maintain a linked-list of the neighbors of every vertex.
 - n vertices

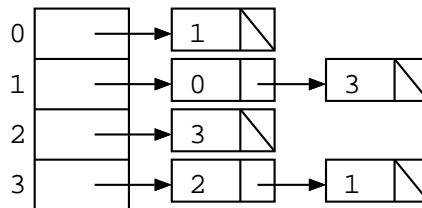
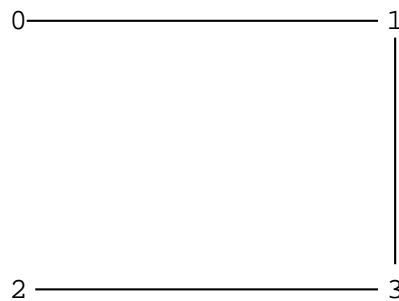
- Array of n lists, one per vertex
- Each list i contains a list of all vertices adjacent to i .

15-8: **Adjacency List**

- Examples:

15-9: **Adjacency List**

- Examples:



- Note – lists are not always sorted

15-10: **Sparse vs. Dense**

- Sparse graph – relatively few edges
- Dense graph – lots of edges
- Complete graph – contains all possible edges
 - These terms are fuzzy. “Sparse” in one context may or may not be “sparse” in a different context

15-11: **Breadth-First Search**

- Method for searching a graph
- Specify a source node in the graph
- Find all nodes reachable from that node
 - First find all nodes 1 unit away
 - Next find all nodes 2 units away

- ... etc

15-12: Breadth-First Search

- Auxiliary Data Structures
 - “color” for each vertex – white, black, grey
 - Used to make sure we don’t visit vertices more than once
 - Parent of each vertex (Path to source node)
 - Distance of each vertex from source

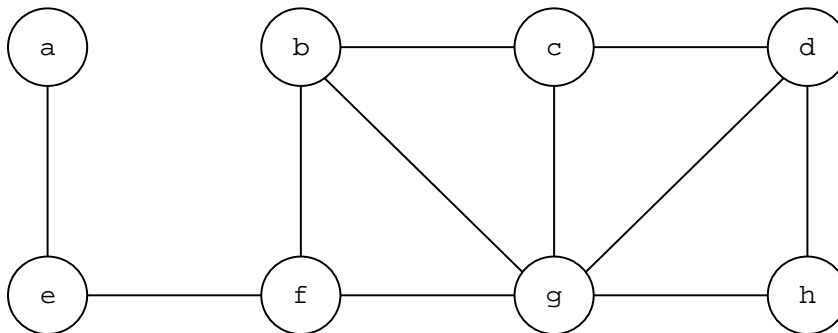
15-13: Breadth-First Search

```

BFS( $G, s$ )
  for each vertex  $u$  in  $V[G]$  do
     $\text{color}[u] \leftarrow \text{WHITE}$ 
     $d[u] \leftarrow \infty$ 
     $\pi[u] \leftarrow \text{nil}$ 
   $\text{color}[s] \leftarrow \text{GRAY}$ 
   $d[s] \leftarrow 0$ 
   $Q \leftarrow \{s\}$ 
  while  $Q$  not empty do
     $u \leftarrow Q.\text{dequeue}$ 
    for each  $v$  adj. to  $u$ 
      if  $\text{color}[v] = \text{WHITE}$ 
         $\text{color}[v] \leftarrow \text{GRAY}$ 
         $d[v] \leftarrow d[u] + 1$ 
         $\pi[v] \leftarrow u$ 
         $Q.\text{enqueue}(v)$ 
     $\text{color}[u] \leftarrow \text{BLACK}$ 

```

15-14: Breadth-First Search

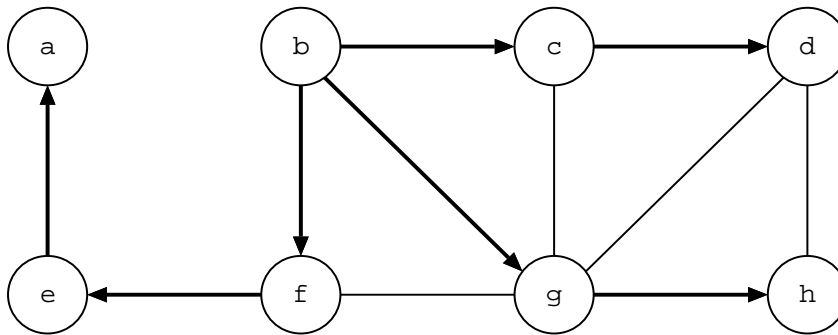


Q: b

15-15: Breadth-First Search

- BFS computes the shortest path from the start vertex to every other vertex
- We can run BFS on a directed or undirected tree
- Defines a “BFS Tree”
 - Parent pointers $p[v]$
 - BFS Tree is directed

15-16: Breadth-First Search



Q:

15-17: Breadth-First Search

- BFS Running time:
 - V vertices
 - E edges

15-18: Breadth-First Search

- BFS Running time:
 - V vertices
 - E edges
- Running time $\Theta(V + E)$
 - In terms of just V , $O(V^2)$ (why?)

15-19: Depth-First Search

DFS(G)

```

for each vertex  $v$  in  $G$  do
  color[ $v$ ] ← WHITE
   $\pi[v]$  = nil
time ← 0
for each vertex  $v$  in  $G$  do
  if color[ $v$ ] = WHITE
    DFS-VISIT( $v$ )
  
```

15-20: Depth-First Search

DFS-VISIT(v, G)

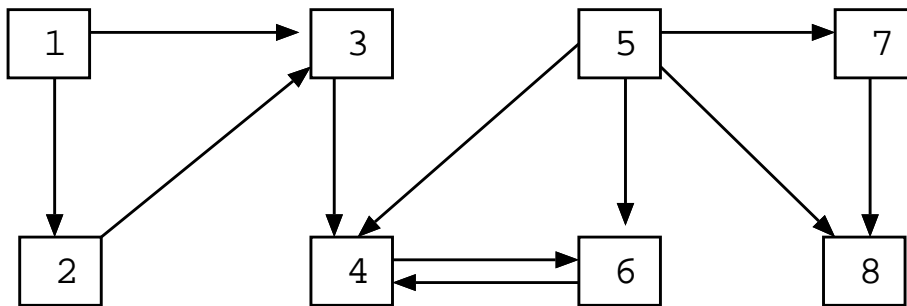
```

color[ $v$ ] ← GRAY
time ← time + 1
 $d[v]$  ← time
for each  $u$  adjacent to  $v$  in  $G$  do
  if color[ $u$ ] = WHITE then
     $\pi[u]$  ←  $v$ 
  
```

```

DFS-VISIT( $u, G$ )
color[ $v$ ]  $\leftarrow$  BLACK
time  $\leftarrow$  time + 1
 $f[v] \leftarrow$  time

```

15-21: **Depth-First Search**

(Do DFS, show discover/finish times & Depth First Forest) 15-22: **Depth-First Search**

- DFS creates a Depth First Forest
- We can use DFS to classify edges:
 - Tree edges
 - edges in the Depth First Forest
 - Back Edges
 - edge (u, v) that connects u to ancestor v in DFF
 - Forward edges
 - non-tree edge (u, v) that connects u to descendent v in DFF
 - Cross Edges
 - Everything Else

15-23: **Depth-First Search**

- Labeling edges
 - How could we label edges (tree/back/forward/cross) while we are doing DFS?

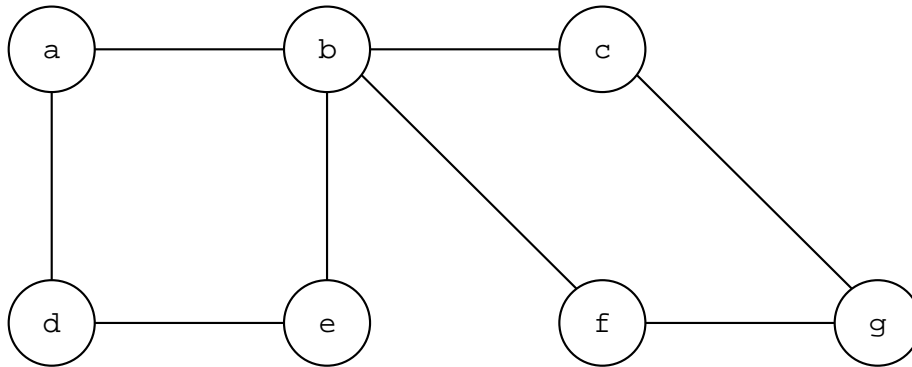
15-24: **Depth-First Search**

- Labeling edges
 - How could we label edges (tree/back/forward/cross) while we are doing DFS?
 - When examining edge (u, v) , if v is:
 - WHITE – tree edge
 - GRAY – back edge
 - BLACK – forward edge or cross edge

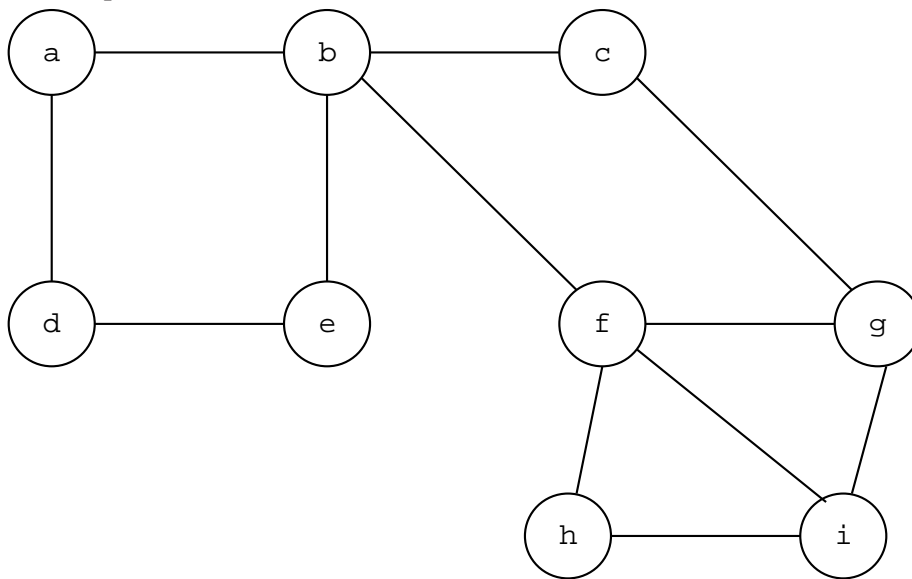
15-25: **Depth-First Search**

- Labeling edges

- Can we have cross edges in a DFS of an undirected graph?
- Can we have forward edges in a DFS of an undirected graph?

15-26: **Depth-First Search**

(Do DFS, show discover/finish times & Depth First Forest)

15-27: **Depth-First Search**

(Do DFS, show discover/finish times & Depth First Forest)

15-28: **Topological Sort**

- Directed Acyclic Graph, Vertices $v_1 \dots v_n$
- Create an ordering of the vertices
 - If there a path from v_i to v_j , then v_i appears before v_j in the ordering
- Example: Prerequisite chains

15-29: **Topological Sort**

- How could we use DFS to do a Topological Sort?
 - (Hint – Use discover and/or finish times)

15-30: **Topological Sort**

- How could we use DFS to do a Topological Sort?
 - (Hint – Use discover and/or finish times)
 - (What does it mean if node x finished before node y ?)

15-31: **Topological Sort**

- How could we use DFS to do a Topological Sort?
 - Do DFS, computing finishing times for each vertex
 - As each vertex is finished, add to front of a linked list
 - This list is a valid topological sort

15-32: **Topological Sort**

- Second method for doing topological sort:
- Which node(s) could be first in the topological ordering?
 - Node(s) with no incident (incoming) edges

15-33: **Topological Sort**

- Pick a node v_k with no incident edges
- Add v_k to the ordering
- Remove v_k and all edges from v_k from the graph
- Repeat until all nodes are picked.

15-34: **Topological Sort**

- How can we find a node with no incident edges?
- Count the incident edges of all nodes

15-35: **Topological Sort**

```
for (i=0; i < NumberOfVertices; i++)
    NumIncident[i] = 0;

for(i=0; i < NumberOfVertices; i++)
    each node k adjacent to i
        NumIncident[k]++
```

15-36: **Topological Sort**

```
for(i=0; i < NumberOfVertices; i++)
    NumIncident[i] = 0;

for(i=0; i < NumberOfVertices; i++)
    for(tmp=G[i]; tmp != null; tmp=tmp.next())
        NumIncident[tmp.neighbor()]++
```


15-37: **Topological Sort**

- Create NumIncident array
- Repeat
 - Search through NumIncident to find a vertex v with $\text{NumIncident}[v] == 0$
 - Add v to the ordering
 - Decrement NumIncident of all neighbors of v
 - Set $\text{NumIncident}[v] = -1$
- Until all vertices have been picked

15-38: **Topological Sort**

- In a graph with V vertices and E edges, how long does this version of topological sort take?

15-39: **Topological Sort**

- In a graph with V vertices and E edges, how long does this version of topological sort take?
 - $\Theta(V^2 + E) = \Theta(V^2)$
 - Since $E \in O(V^2)$

15-40: **Topological Sort**

- Where are we spending “extra” time

15-41: **Topological Sort**

- Where are we spending “extra” time
 - Searching through NumIncident each time looking for a vertex with no incident edges
 - Keep around a set of all nodes with no incident edges
 - Remove an element v from this set, and add it to the ordering
 - Decrement NumIncident for all neighbors of v
 - If $\text{NumIncident}[k]$ is decremented to 0, add k to the set.
 - How do we implement the set of nodes with no incident edges?

15-42: **Topological Sort**

- Where are we spending “extra” time
 - Searching through NumIncident each time looking for a vertex with no incident edges
 - Keep around a set of all nodes with no incident edges
 - Remove an element v from this set, and add it to the ordering
 - Decrement NumIncident for all neighbors of v
 - If $\text{NumIncident}[k]$ is decremented to 0, add k to the set.
 - How do we implement the set of nodes with no incident edges?
 - Use a stack

15-43: **Topological Sort**

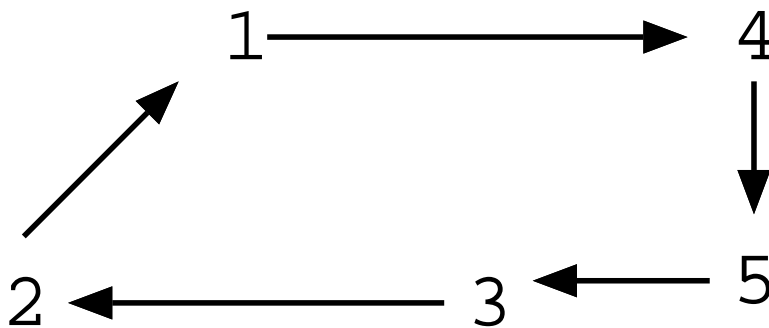
- Examples!!
 - Graph
 - Adjacency List
 - NumIncident
 - Stack

15-44: **More DFS Applications**

- Depth First Search can be used to calculate the connected components of a directed graph
- First, some definitions and examples:

15-45: **Strongly Connected Graph**

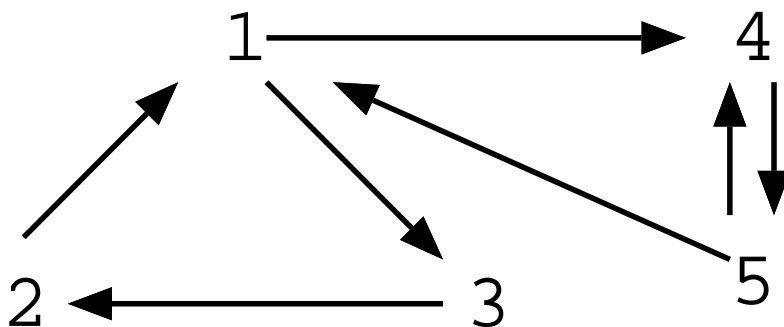
- Directed Path from every node to every other node



- Strongly Connected

15-46: **Strongly Connected Graph**

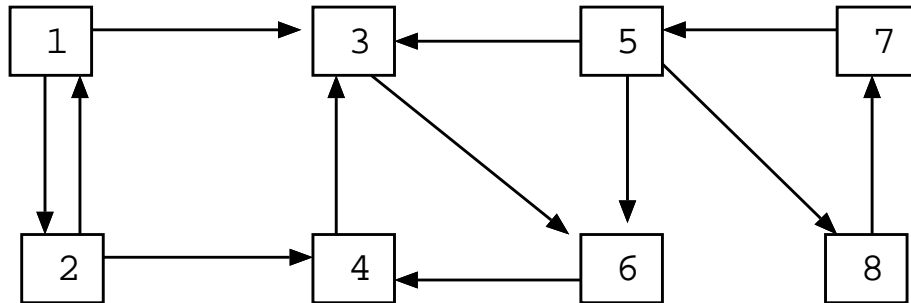
- Directed Path from every node to every other node



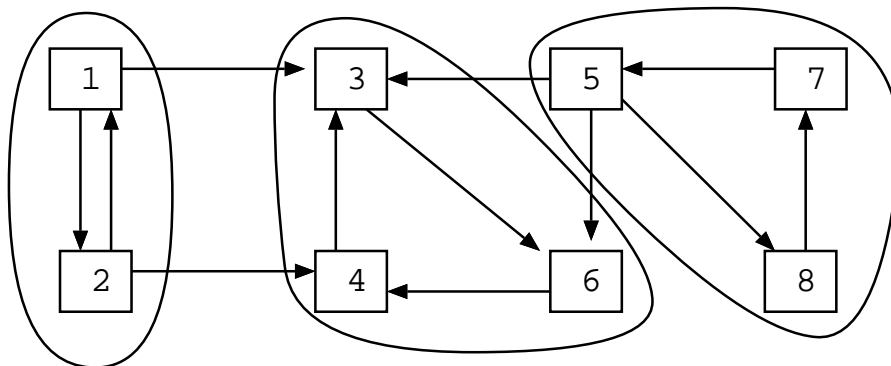
- Strongly Connected

15-47: **Connected Components**

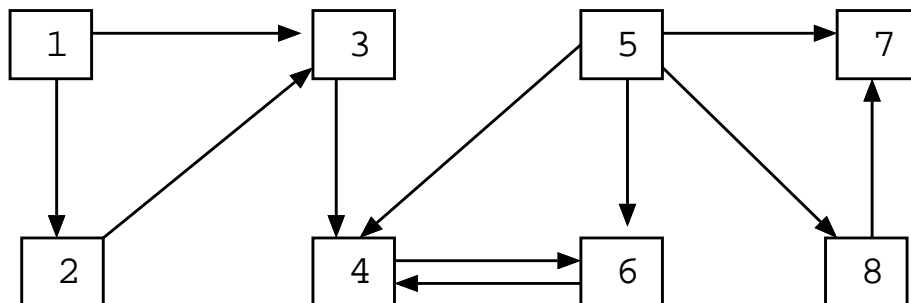
- Subgraph (subset of the vertices) that is strongly connected.

15-48: **Connected Components**

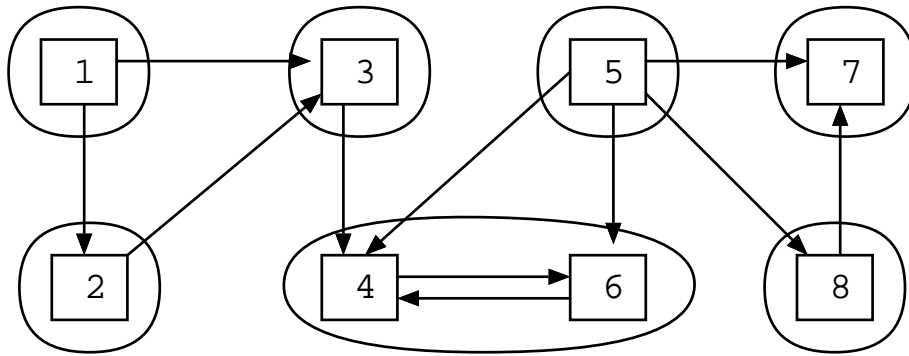
- Subgraph (subset of the vertices) that is strongly connected.

15-49: **Connected Components**

- Subgraph (subset of the vertices) that is strongly connected.

15-50: **Connected Components**

- Subgraph (subset of the vertices) that is strongly connected.

15-51: **Connected Components**

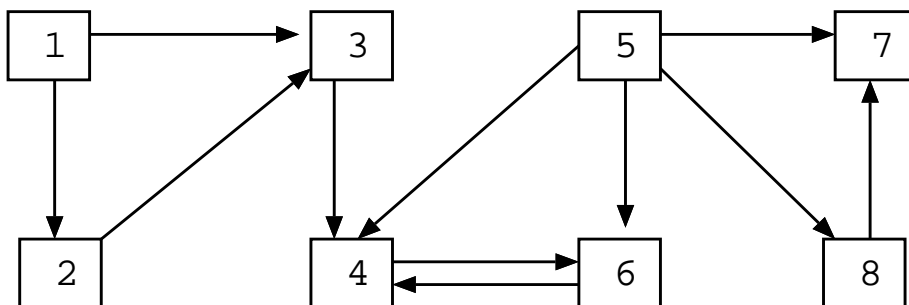
- Connected components of the graph are the *largest possible* strongly connected subgraphs
- If we put each vertex in its own component – each component would be (trivially) strongly connected
 - Those would not be the connected components of the graph – unless there were no larger connected subgraphs

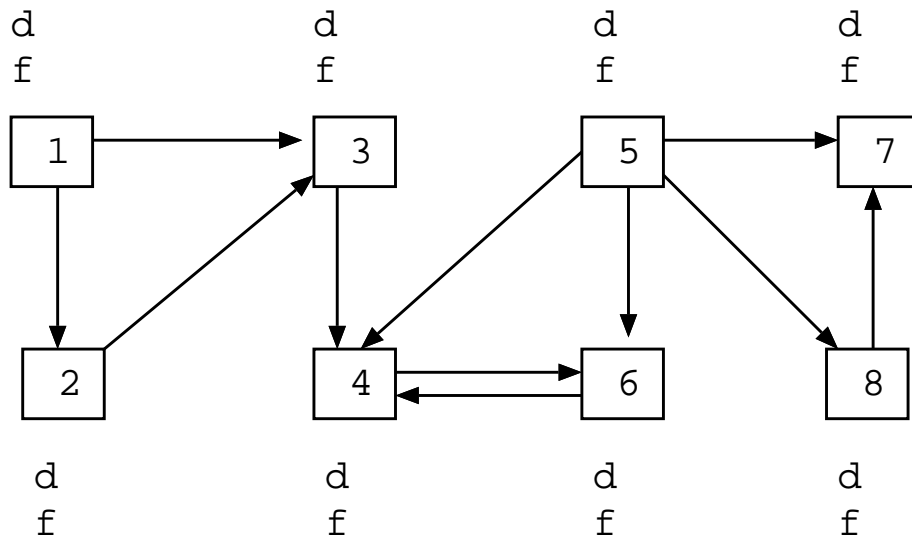
15-52: **Connected Components**

- Calculating Connected Components
 - Two vertices v_1 and v_2 are in the same connected component if and only if:
 - Directed path from v_1 to v_2
 - Directed path from v_2 to v_1
 - To find connected components – find directed paths
 - Use DFS: $d[v]$ and $f[v]$

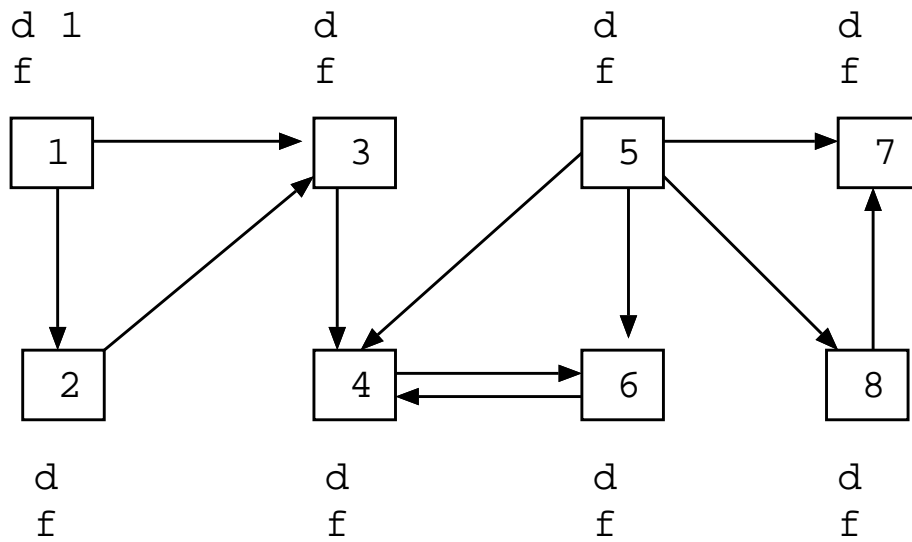
15-53: **DFS Revisited**

- Recall that we calculate the order in which we visit the elements in a Depth-First Search
- For any vertex v in a DFS:
 - $d[v]$ = *Discovery* time – when the vertex is first visited
 - $f[v]$ = *Finishing* time – when we have finished with a vertex (and all of its children)

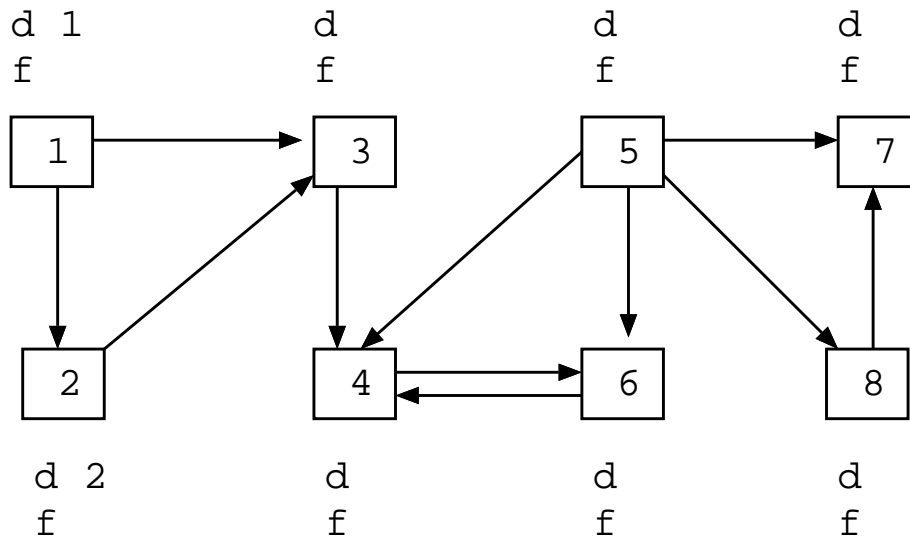
15-54: **DFS Example**15-55: **DFS Example**



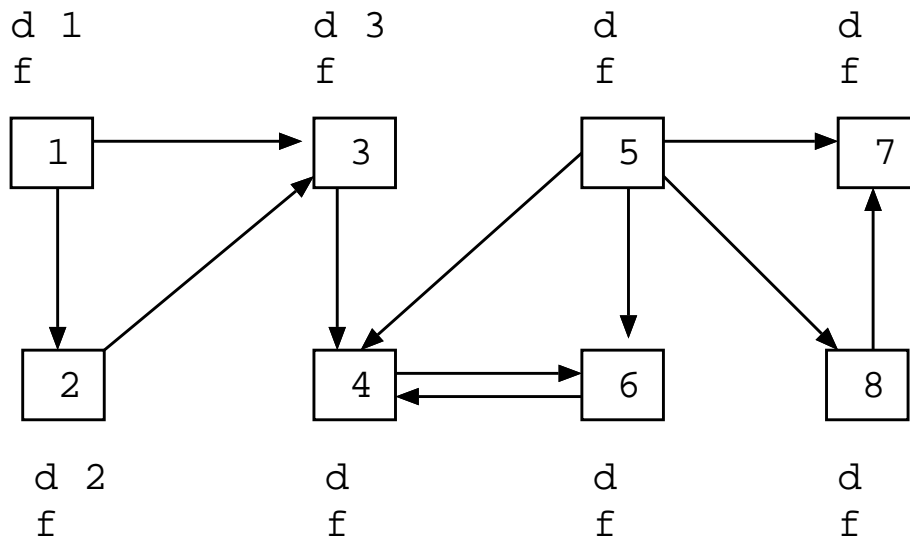
15-56: DFS Example



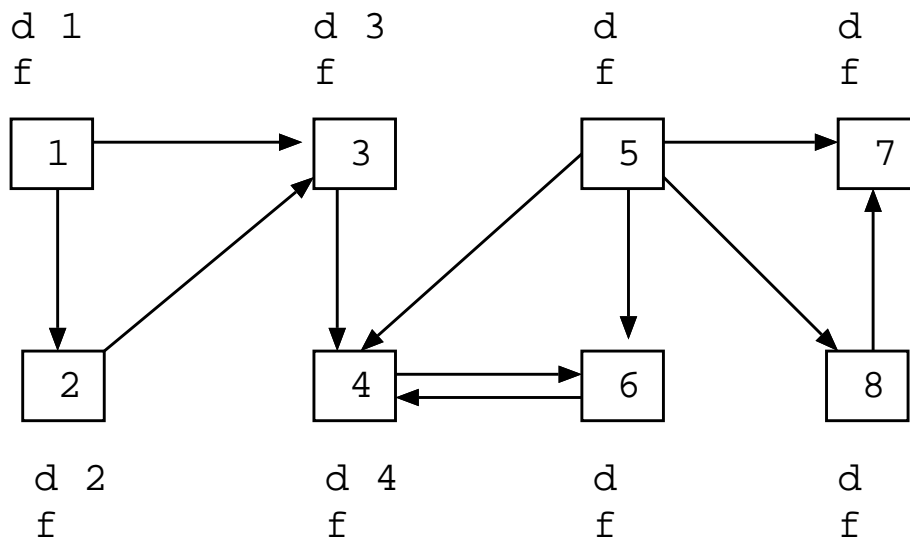
15-57: DFS Example



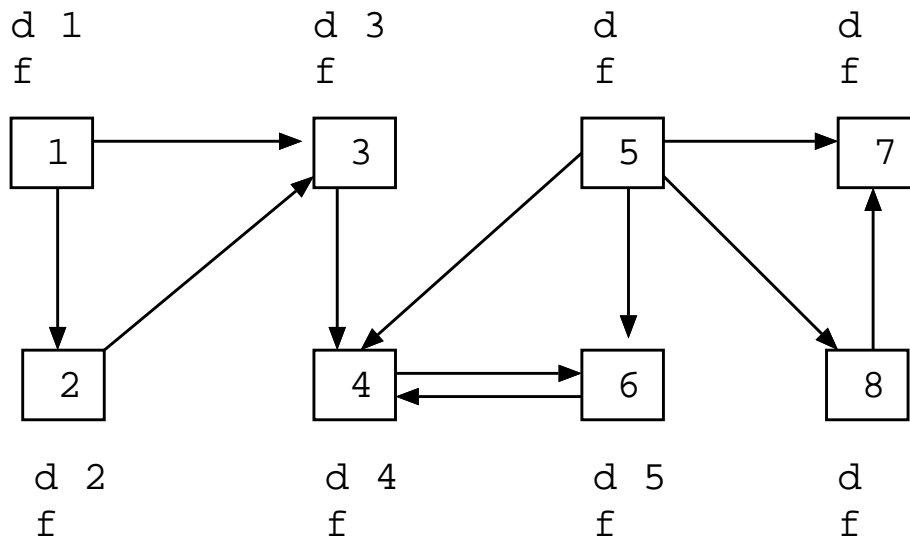
15-58: DFS Example



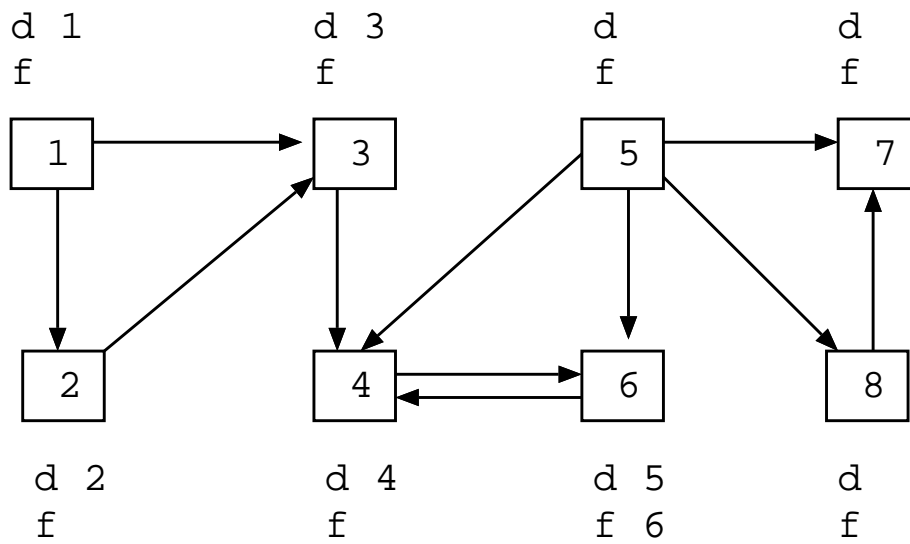
15-59: DFS Example



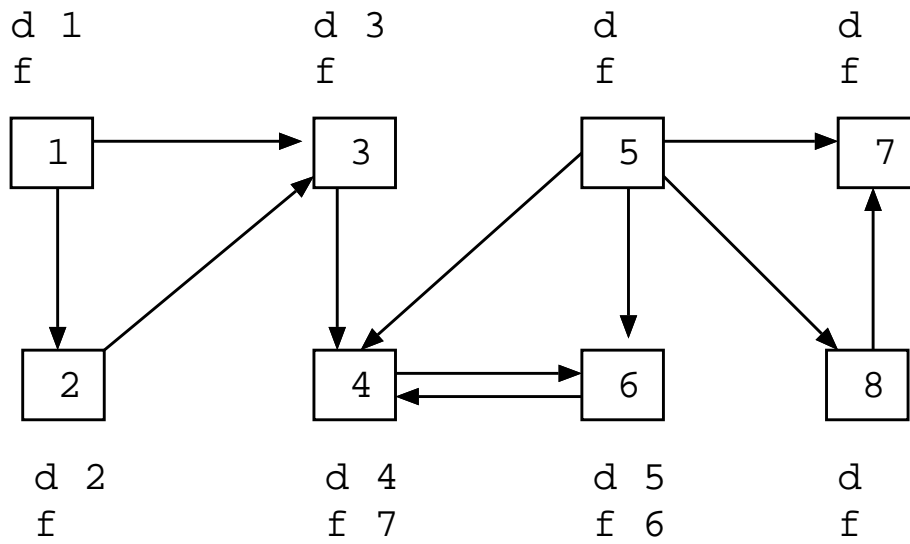
15-60: DFS Example



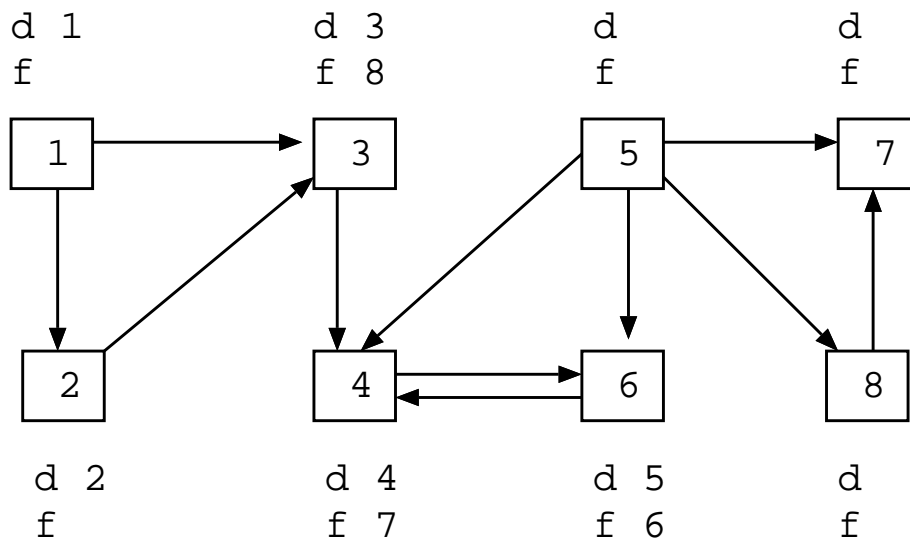
15-61: DFS Example



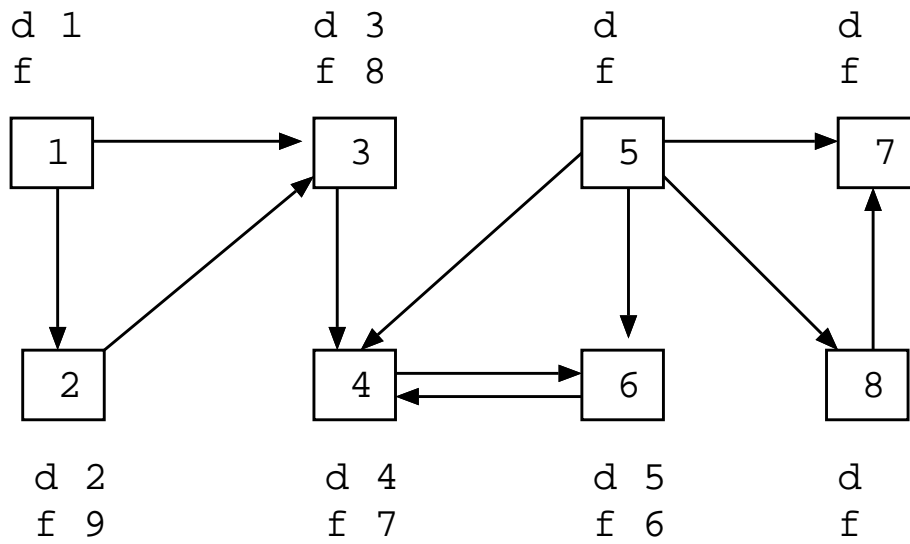
15-62: DFS Example



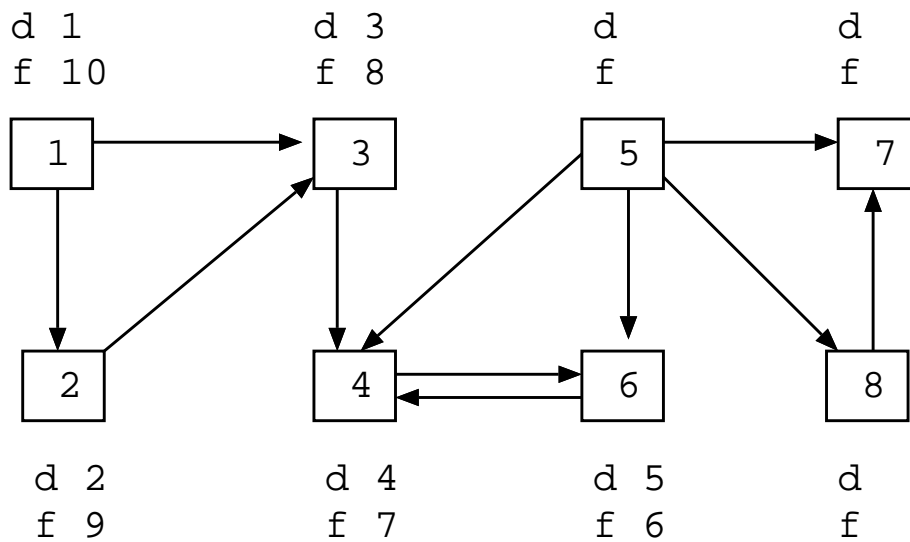
15-63: DFS Example



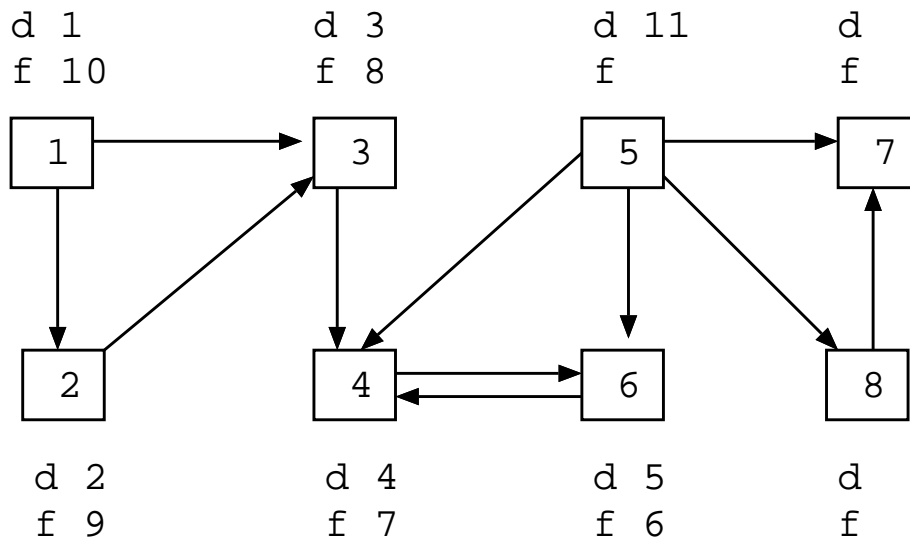
15-64: DFS Example



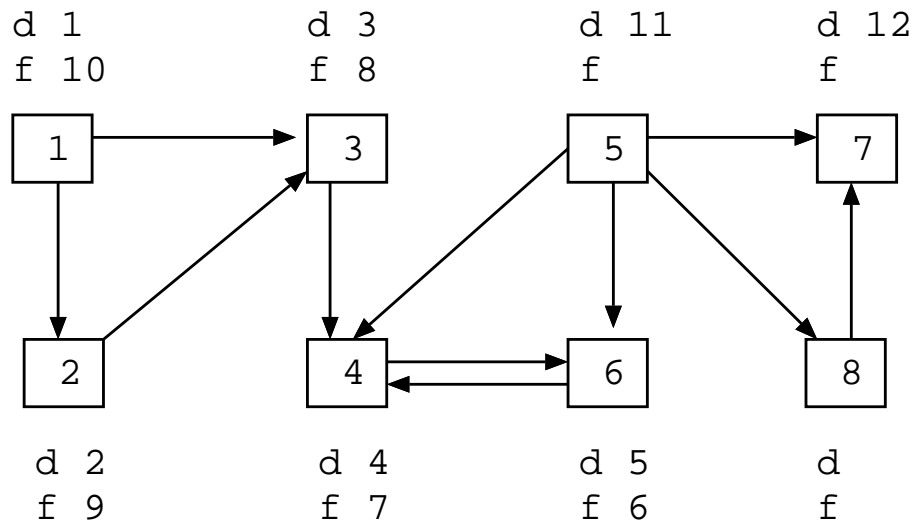
15-65: DFS Example



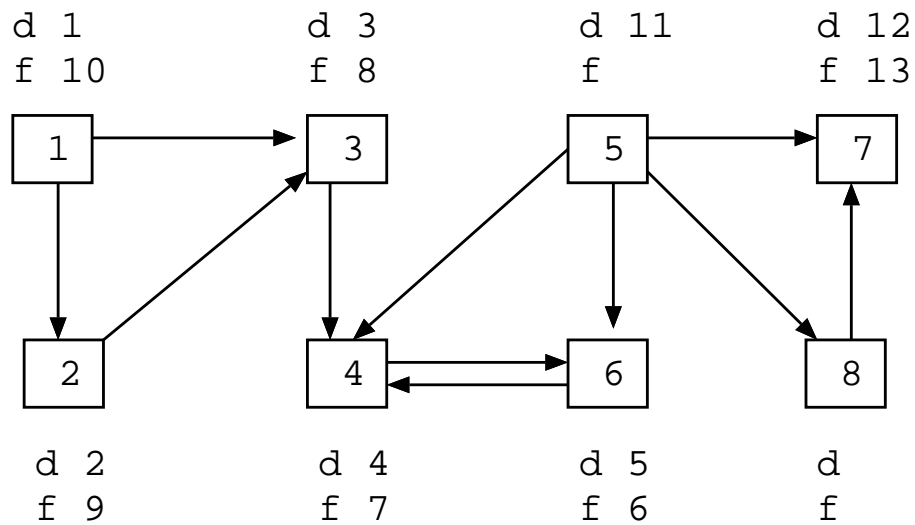
15-66: DFS Example



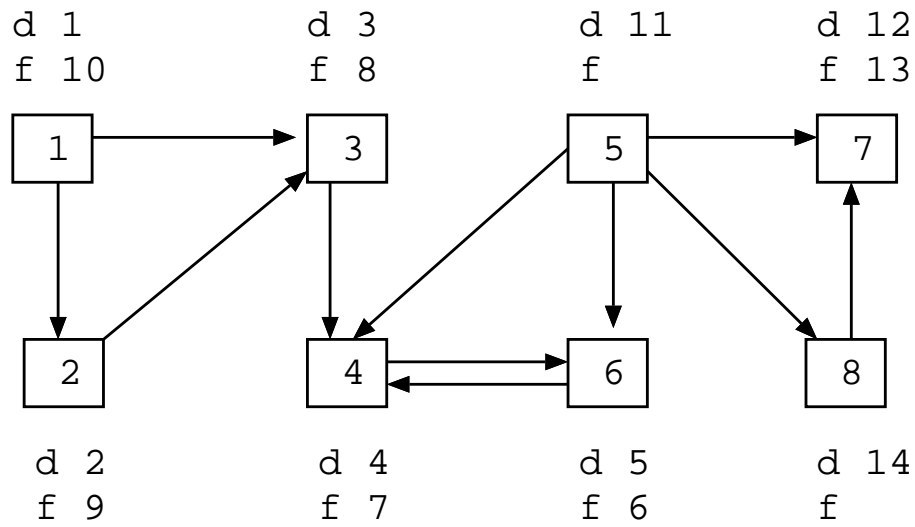
15-67: DFS Example



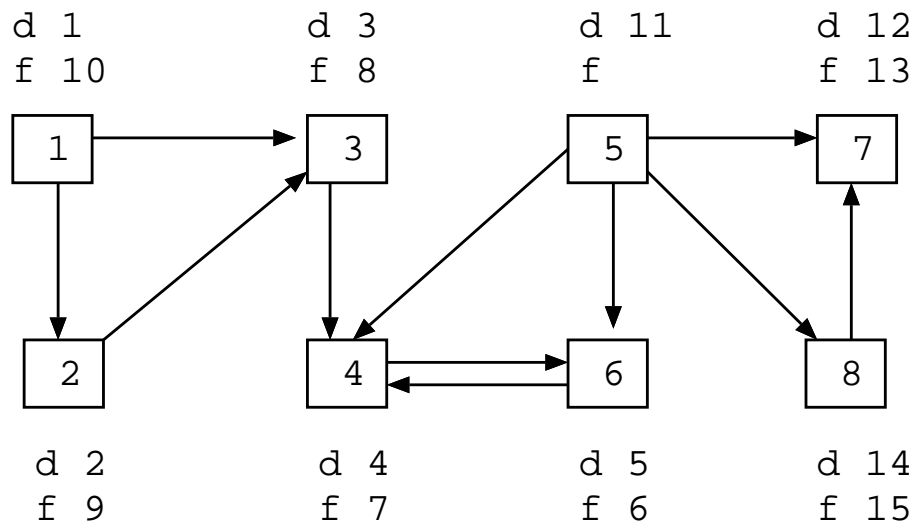
15-68: DFS Example



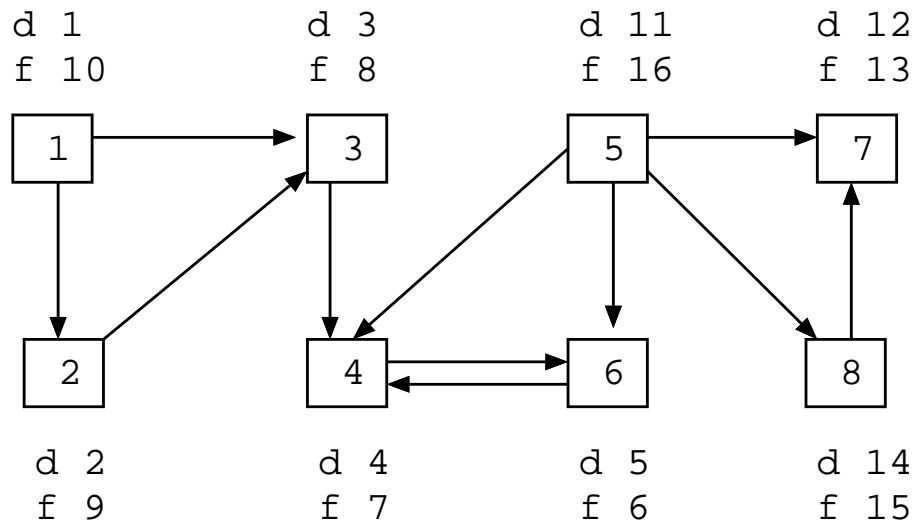
15-69: DFS Example



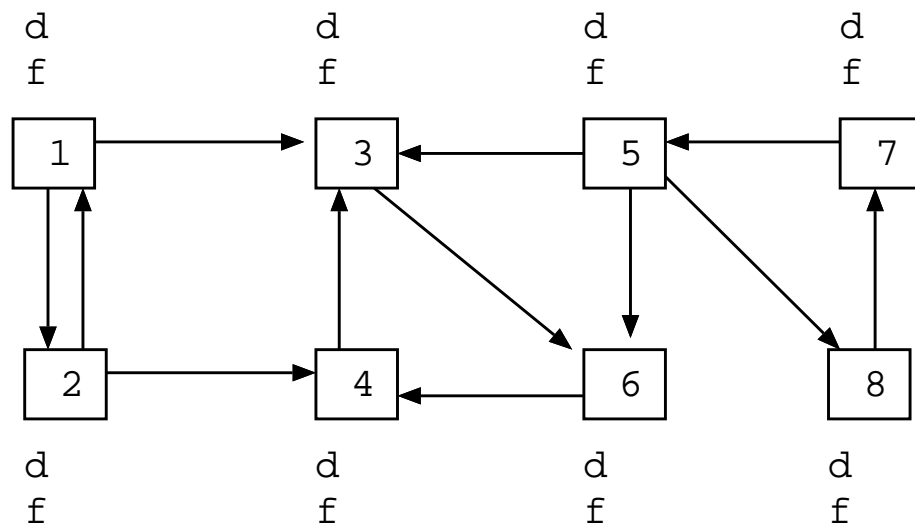
15-70: DFS Example



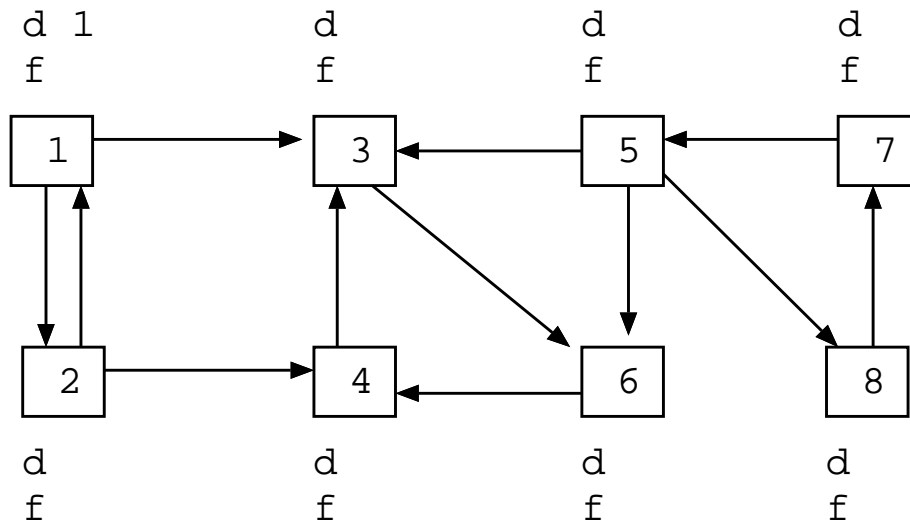
15-71: DFS Example



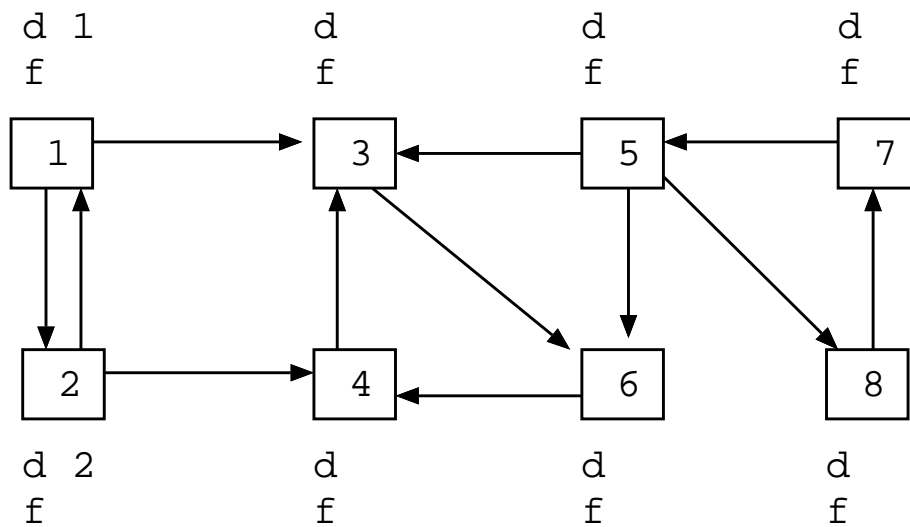
15-72: DFS Example



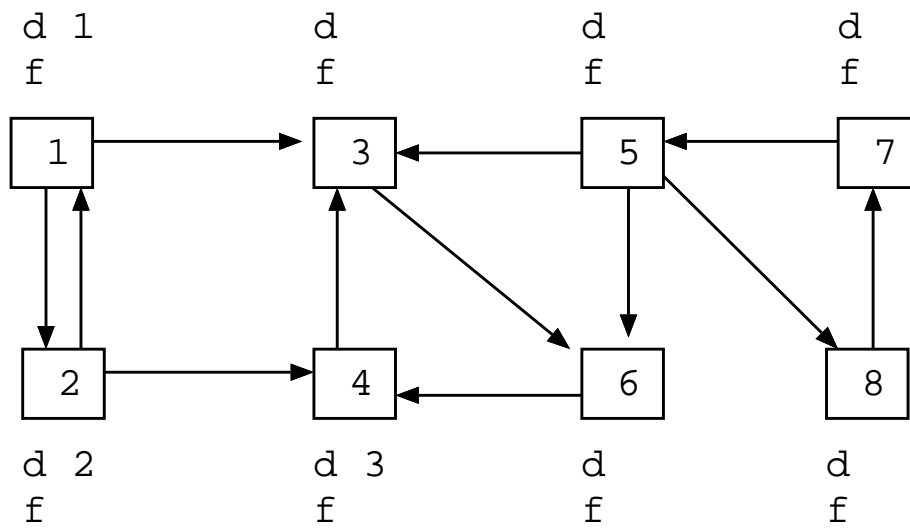
15-73: DFS Example



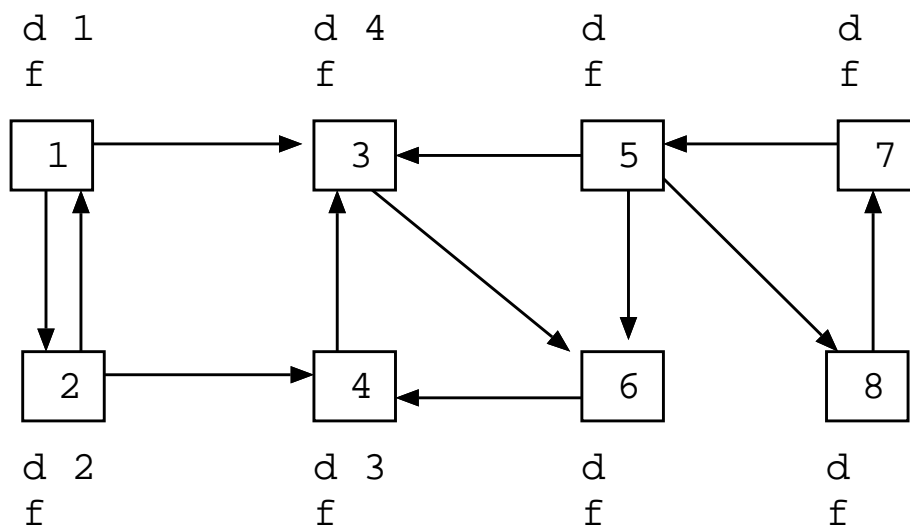
15-74: DFS Example



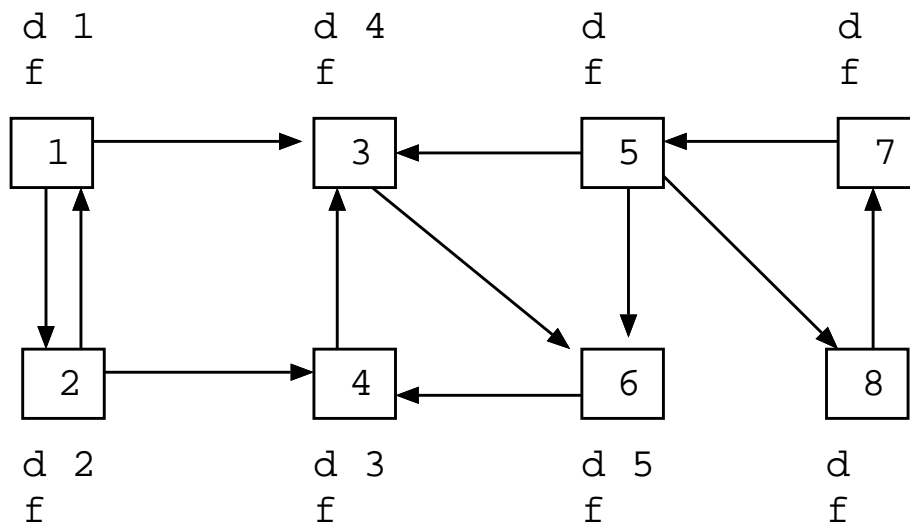
15-75: DFS Example



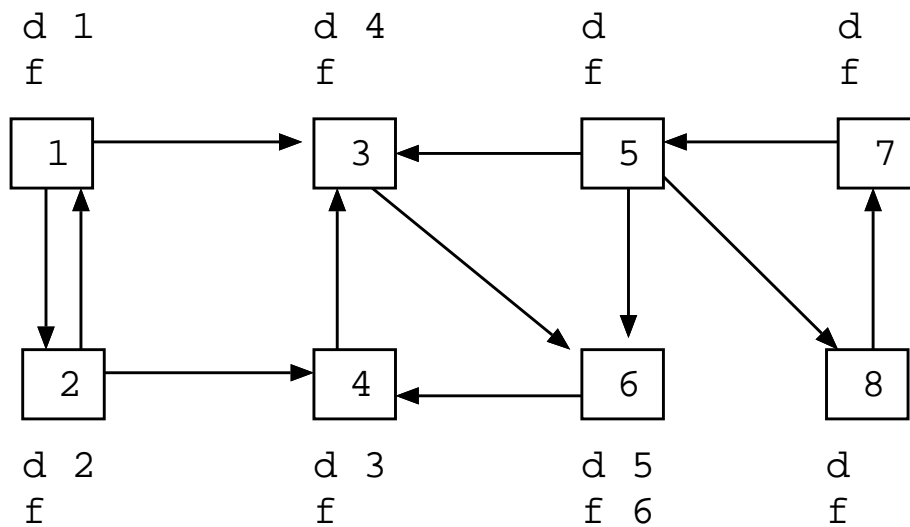
15-76: DFS Example



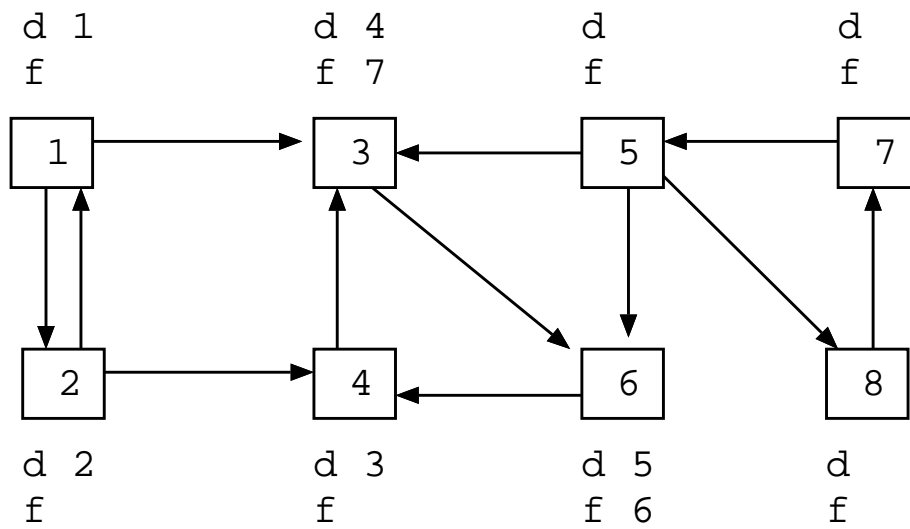
15-77: DFS Example



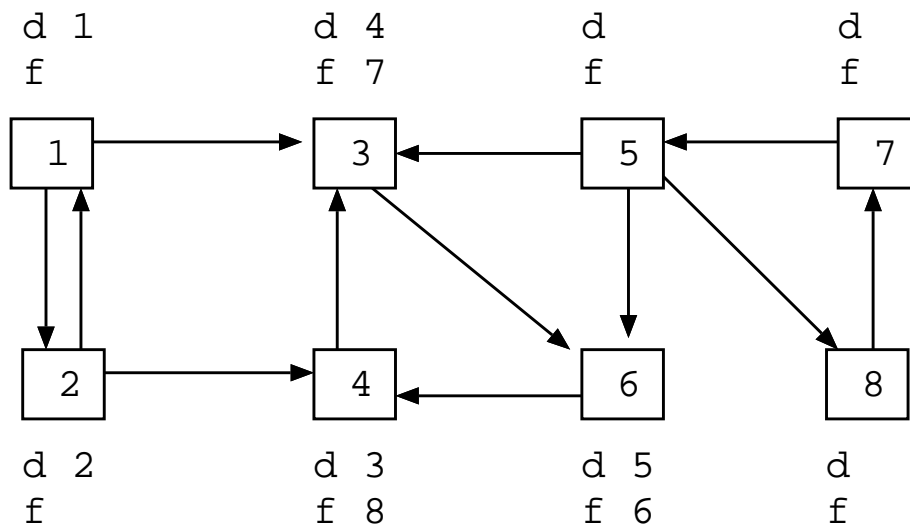
15-78: DFS Example



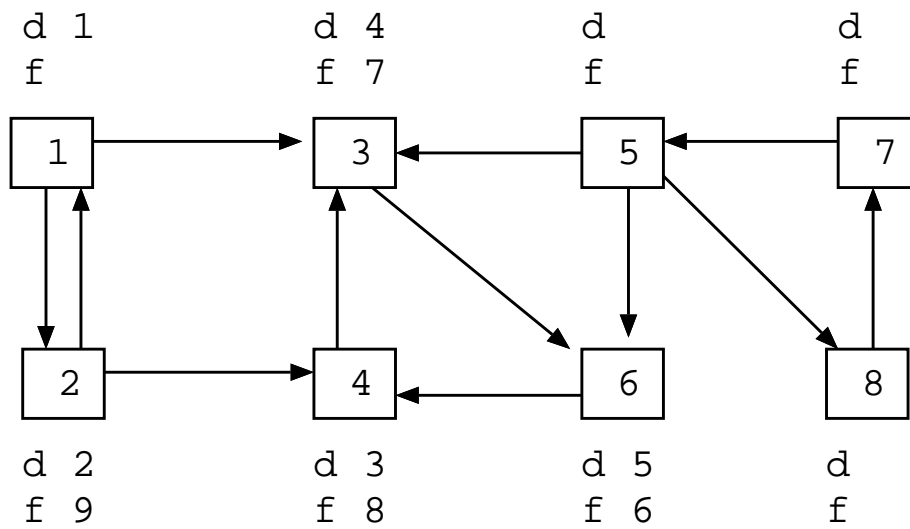
15-79: DFS Example



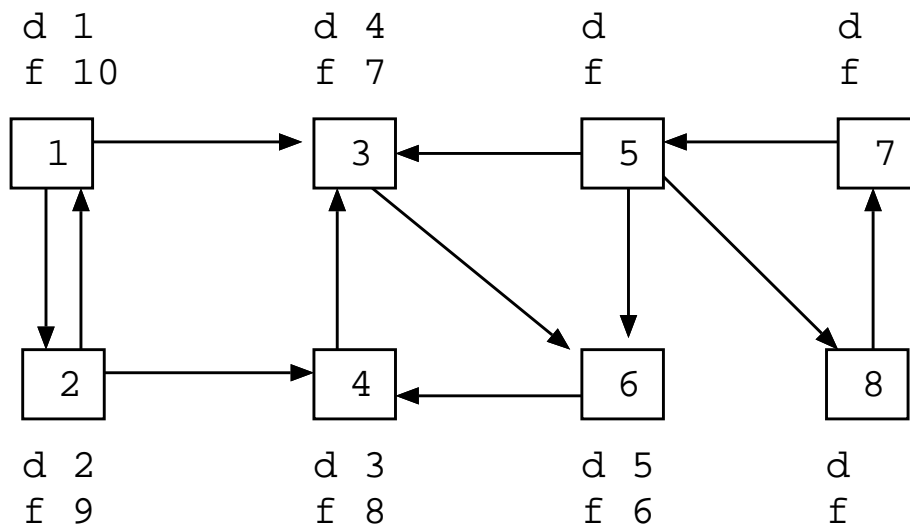
15-80: DFS Example



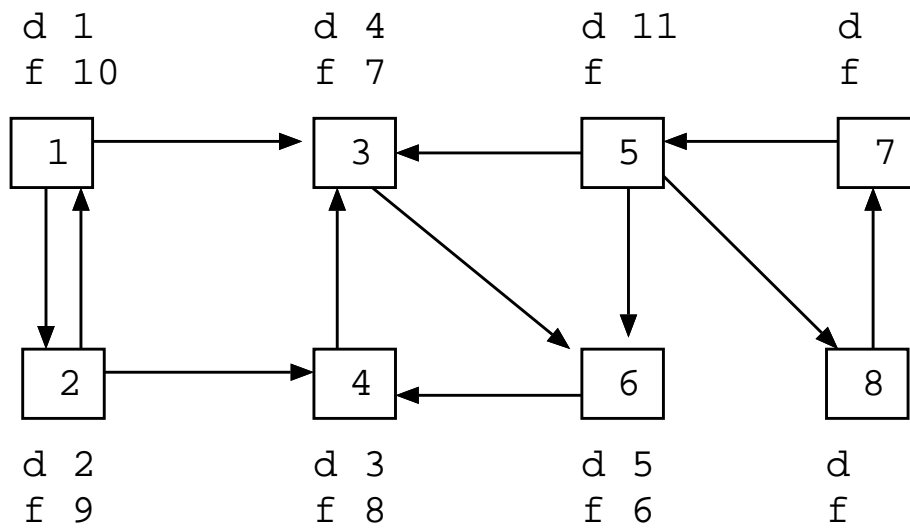
15-81: DFS Example



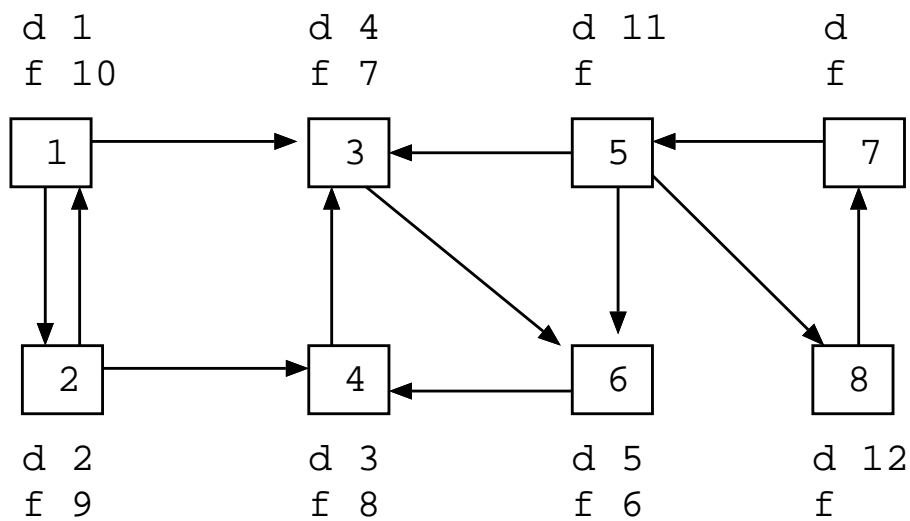
15-82: DFS Example



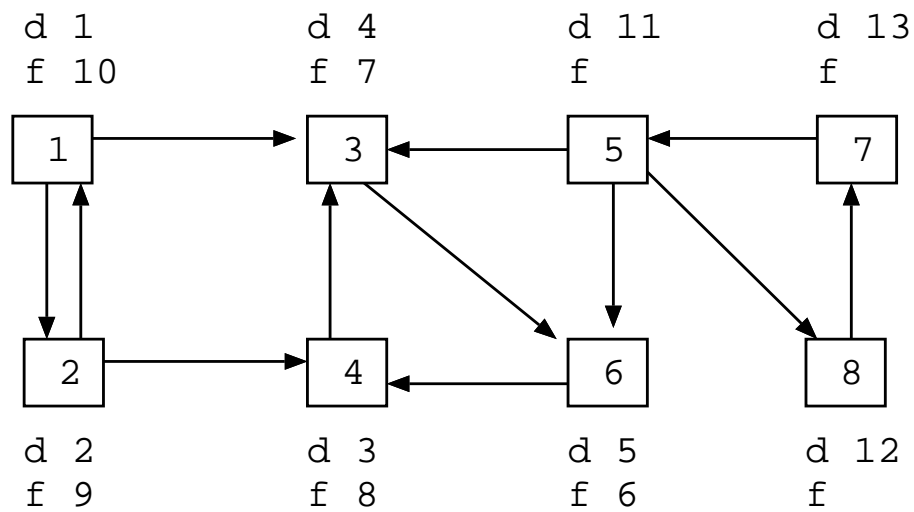
15-83: DFS Example



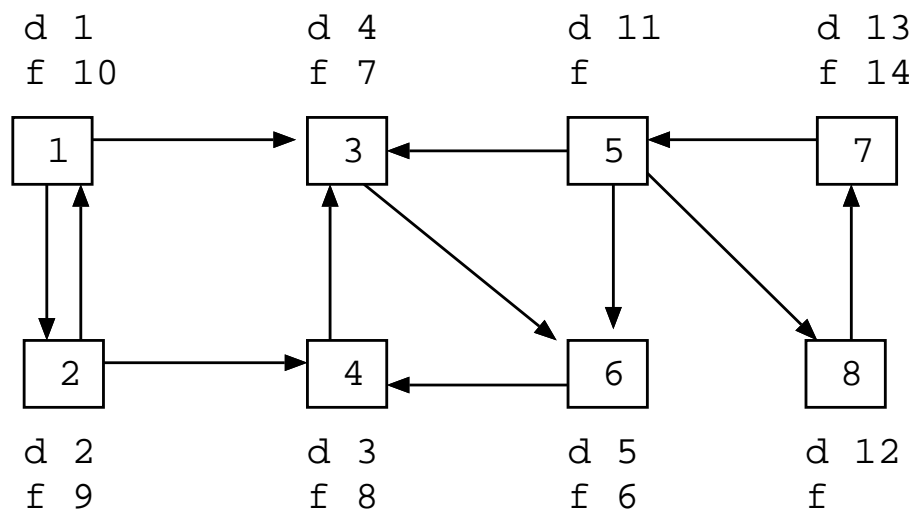
15-84: DFS Example



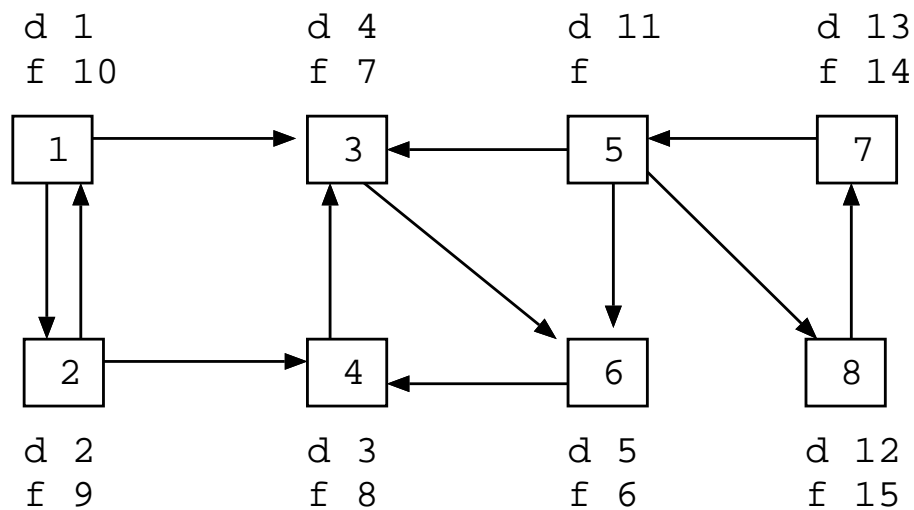
15-85: DFS Example



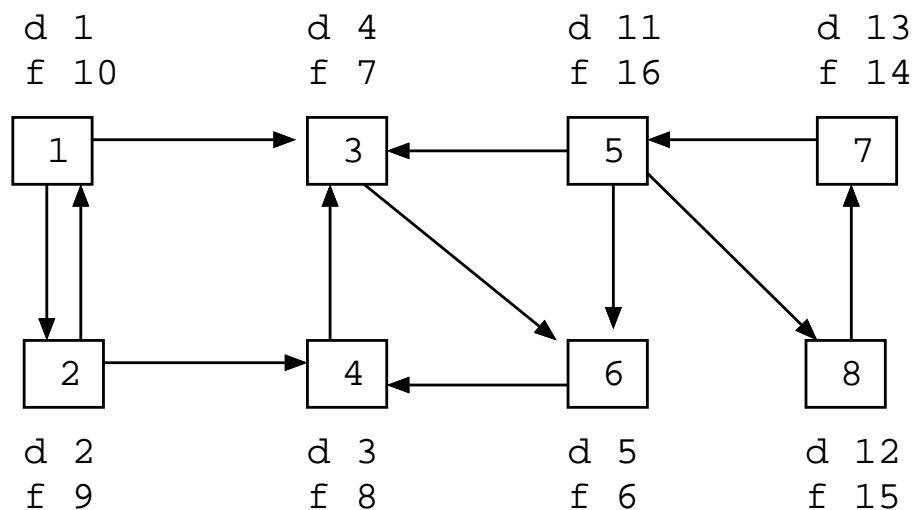
15-86: DFS Example



15-87: DFS Example



15-88: DFS Example

15-89: Using $d[]$ & $f[]$

- Given two vertices v_1 and v_2 , what do we know if $f[v_2] < f[v_1]$?

15-90: Using $d[]$ & $f[]$

- Given two vertices v_1 and v_2 , what do we know if $f[v_2] < f[v_1]$?
 - Either:
 - Path from v_1 to v_2
 - Start from v_1
 - Eventually visit v_2
 - Finish v_2
 - Finish v_1

15-91: Using $d[]$ & $f[]$

- Given two vertices v_1 and v_2 , what do we know if $f[v_2] < f[v_1]$?
 - Either:
 - Path from v_1 to v_2
 - No path from v_2 to v_1
 - Start from v_2
 - Eventually finish v_2
 - Start from v_1
 - Eventually finish v_1

15-92: Using $d[]$ & $f[]$

- If $f[v_2] < f[v_1]$:
 - Either a path from v_1 to v_2 , or no path from v_2 to v_1
 - If there is a path from v_2 to v_1 , then there must be a path from v_1 to v_2
- $f[v_2] < f[v_1]$ and a path from v_2 to $v_1 \Rightarrow v_1$ and v_2 are in the same connected component

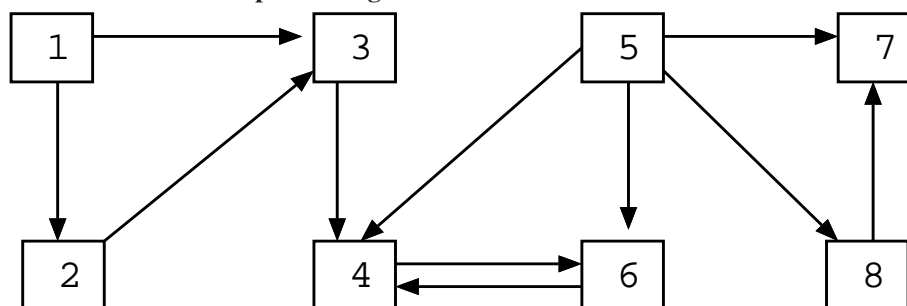
15-93: Calculating paths

- Path from v_2 to v_1 in G if and only if there is a path from v_1 to v_2 in G^T
 - G^T is the transpose of G – G with all edges reversed
- If after DFS, $f[v_2] < f[v_1]$
- Run second DFS on G^T , starting from v_1 , and v_1 and v_2 are in the same DFS spanning tree
- v_1 and v_2 must be in the same connected component

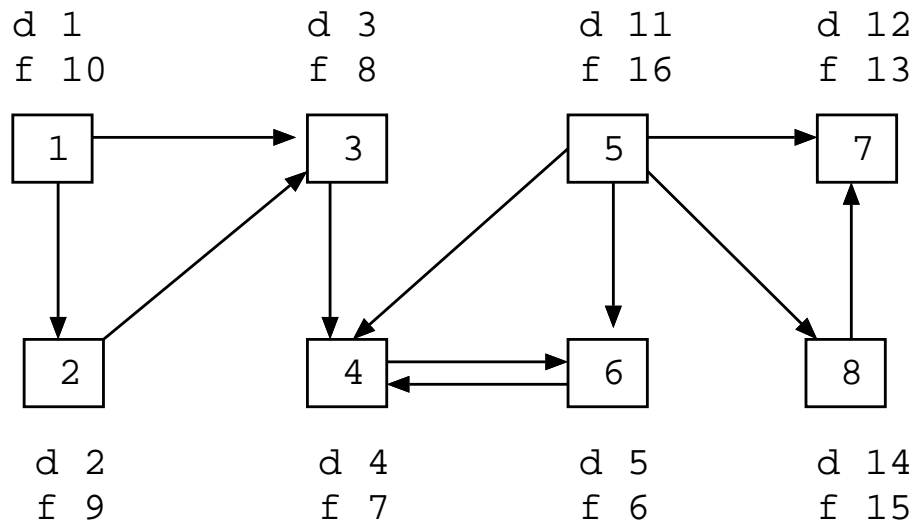
15-94: Connected Components

- Run DFS on G , calculating $f[]$ times
- Compute G^T
- Run DFS on G^T – examining nodes in *inverse order of finishing times* from first DFS
- Any nodes that are in the same DFS search tree in G^T must be in the same connected component

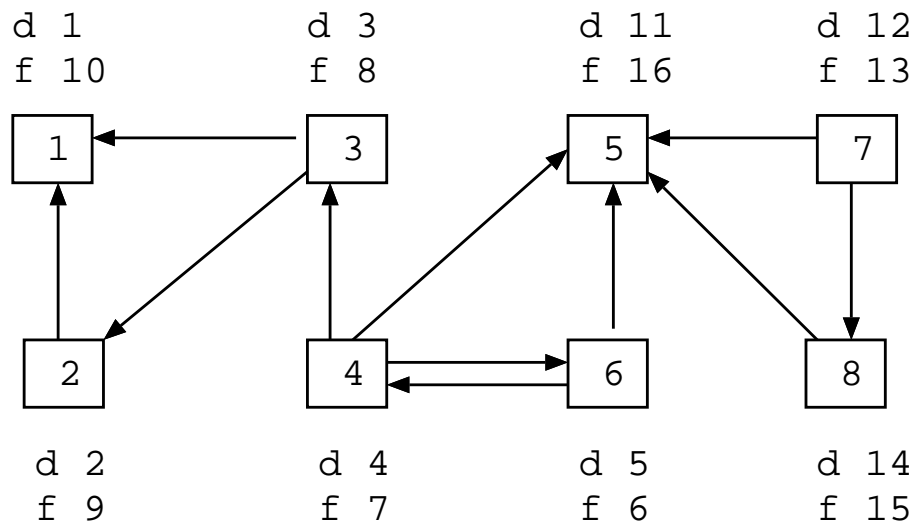
15-95: Connected Components Eg.



15-96: Connected Components Eg.



15-97: Connected Components Eg.



15-98: Connected Components Eg.

d 1
f 10



d 3
f 8



d 11
f 16



d 12
f 13



d 2
f 9



d 4
f 7

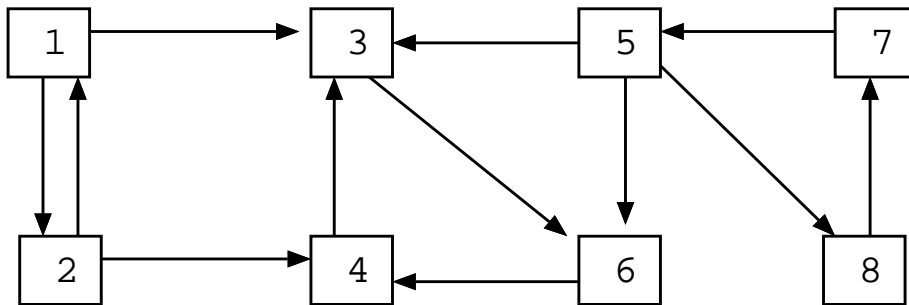


d 5
f 6



d 14
f 15

15-99: Connected Components Eg.



15-100: Connected Components Eg.

d 1
f 10



d 4
f 7



d 11
f 16



d 13
f 14



d 2
f 9



d 3
f 8

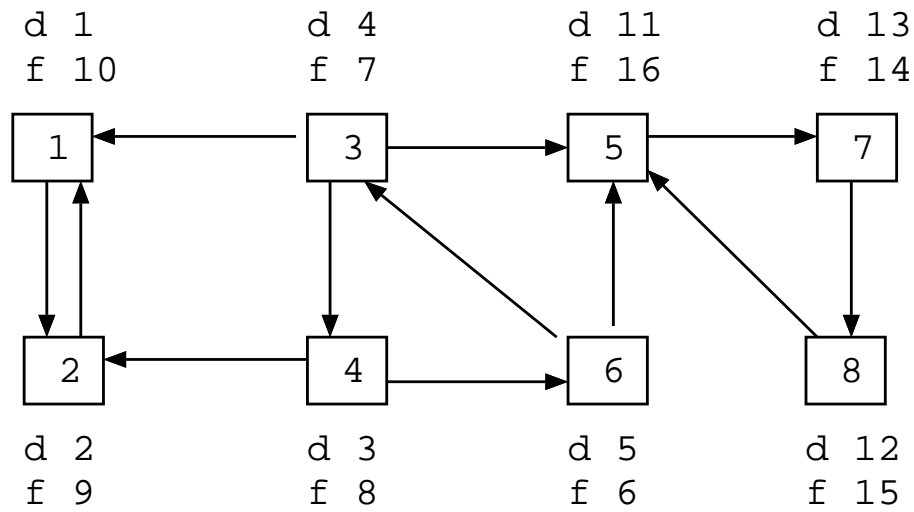


d 5
f 6



d 12
f 15

15-101: Connected Components Eg.



15-102: Connected Components Eg.

