

CS 457, Fall 2019

Drexel University, Department of Computer Science

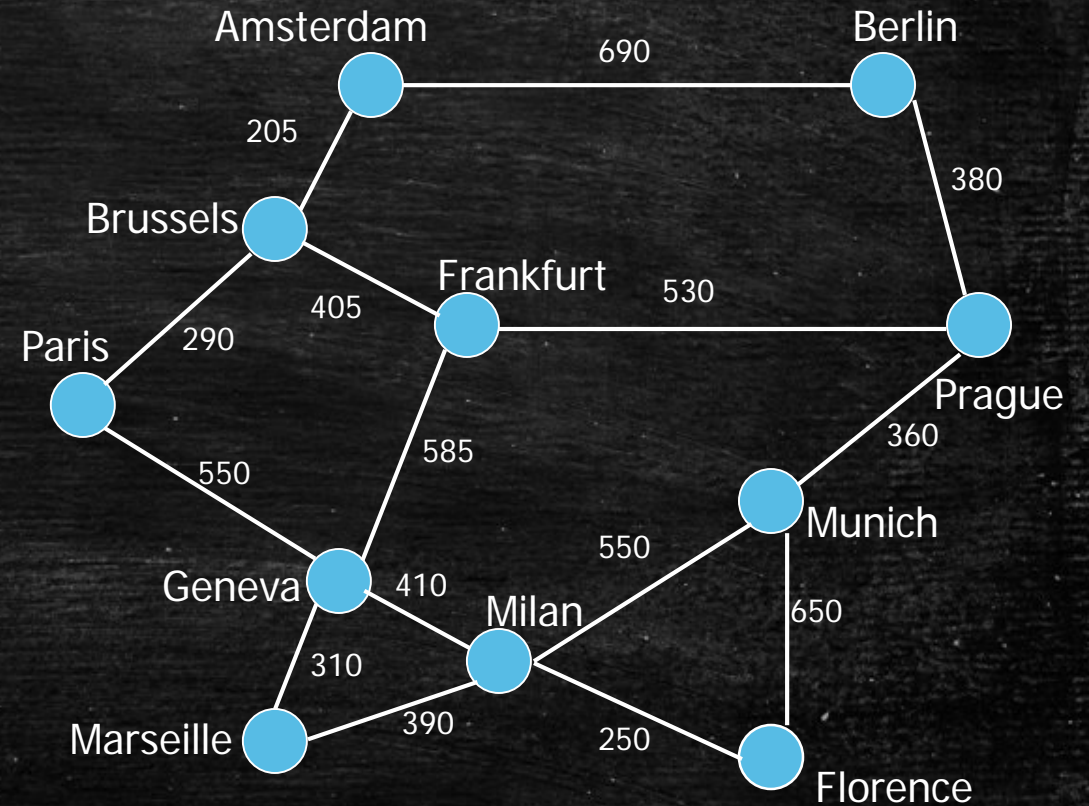
Lecture 16

Today's Lecture

- Graph algorithms
 - Minimum Spanning Tree
 - Breadth First Search
 - Depth First Search
 - Topological Sorting
 - Strongly Connected Components
- Greedy algorithms

Graphs

- Things to know:
 - Path
 - Cycle
 - Sub-graph
 - Degree of a vertex
 - Maximum and minimum degree
 - Maximum number of edges
 - Connected components
 - Shortest path (weighted & unweighted)
 - Distance of two vertices
 - Tree (rooted tree)
 - Spanning tree of a graph
 - Acyclic graph
 - Bipartite graph

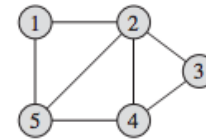


Definitions

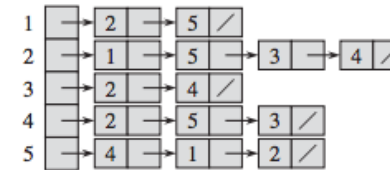
- Given A graph $G=(V,E)$, where
 - V is its vertex set, $|V|=n$,
 - E is its edge set, with $|E|=m=O(n^2)$
- If G is **connected** then for every pair of vertices u,v in G , there is path connecting them
- In an **undirected graph**, an edge $(u, v)=(v, u)$.
- In a **directed graph**, (u, v) is different from (v, u) .
- In a **weighted graph** there are weights associated with edges and/or vertices.
- **Running time** of graph algorithms are usually expressed in terms of n or m .

Graph Representations

- Adjacency List
 - Good for sparse graphs
- Adjacency Matrix
 - Quick edge existence query
 - Simple



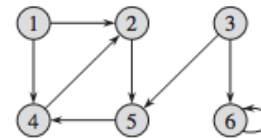
(a)



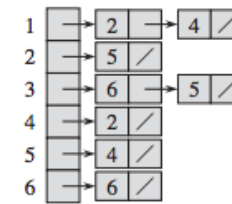
(b)

	1	2	3	4	5
1	0	1	0	0	1
2	1	0	1	1	1
3	0	1	0	1	0
4	0	1	1	0	1
5	1	1	0	1	0

(c)



(a)



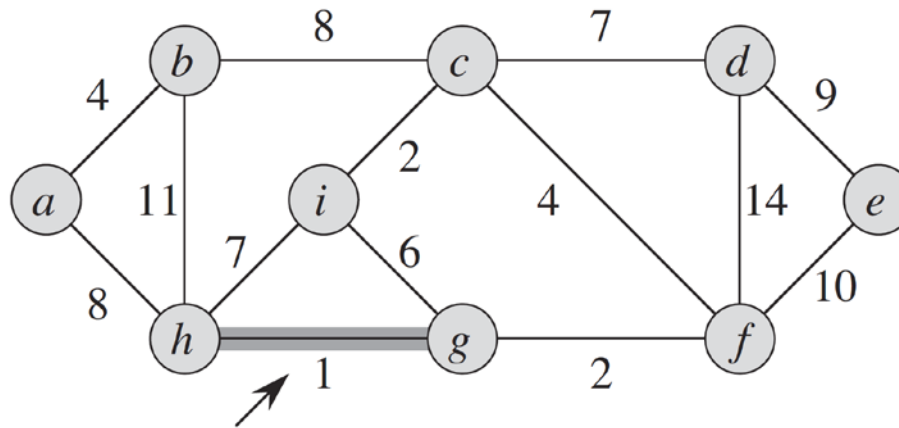
(b)

	1	2	3	4	5	6
1	0	1	0	1	0	0
2	0	0	0	0	1	0
3	0	0	0	0	1	1
4	0	1	0	0	0	0
5	0	0	0	1	0	0
6	0	0	0	0	0	1

(c)

Minimum Spanning Tree of a Weighted Graph

- Let $G=(V,E)$ be a graph on n vertices, m edges, and a **weight** w on edges in E .
- Sub-graph $T=(V,E')$ with $E' \subseteq E$ with no cycles is a **spanning tree**
- The **weight of T** is the sum of the weights of its edges: $w(T) = \sum_{(u,v) \in E'} w(u,v)$



Set Operations

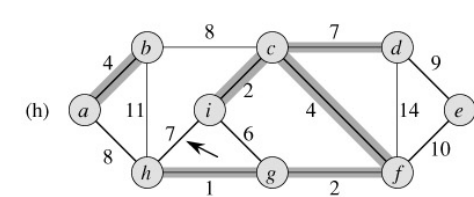
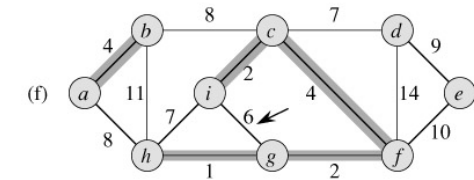
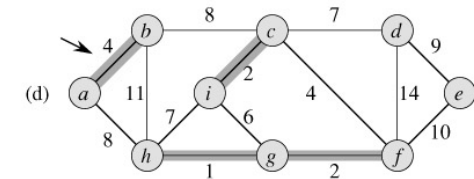
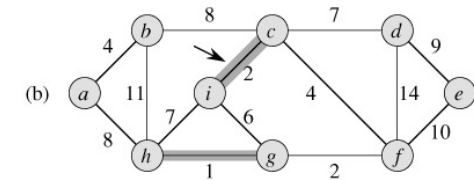
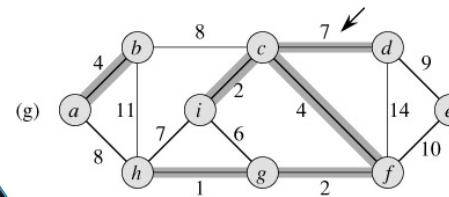
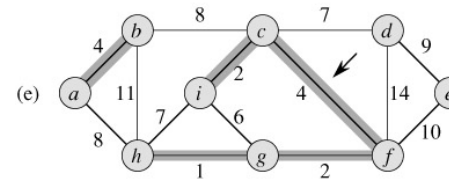
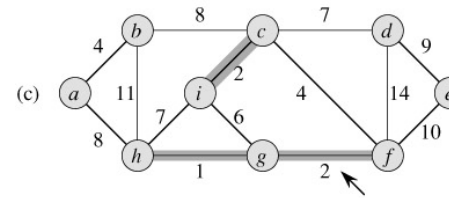
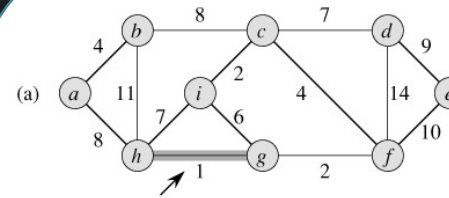
- We will use the following set operations:
 - **Make-Set(v)**: creates a set containing element v , i.e., $\{v\}$
 - **Find-Set(u)**: returns the set to which v belongs to
 - **Union(u, v)**: creates a set which is the union of two sets, the one containing v and the one containing u

Kruskal's Algorithm

MST-KRUSKAL(G, w)

```

1   $A = \emptyset$ 
2  for each vertex  $v \in G.V$ 
3      MAKE-SET( $v$ )
4  sort the edges of  $G.E$  into nondecreasing order by weight
5  for each edge  $(u, v) \in G.E$ , taken in nondecreasing order by weight
6      if FIND-SET( $u$ )  $\neq$  FIND-SET( $v$ )
7           $A = A \cup \{(u, v)\}$ 
8          UNION( $u, v$ )
9  return  $A$ 
    
```



Kruskal's Algorithm

$$\text{MST-KRUSKAL}(G, w)$$

1. $A = \emptyset$

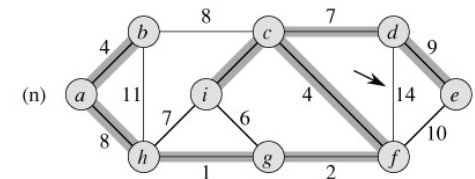
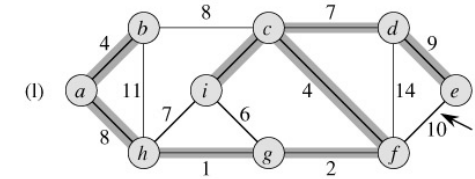
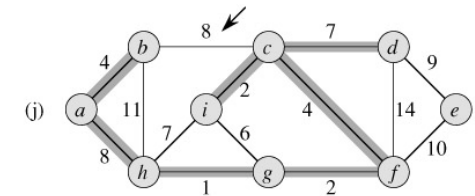
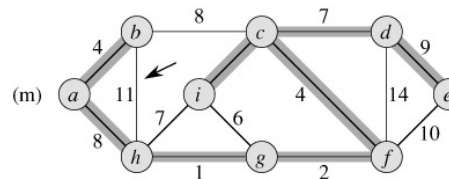
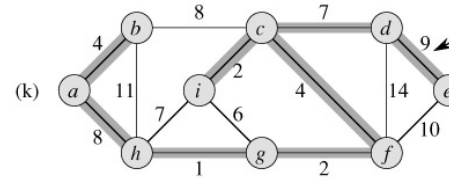
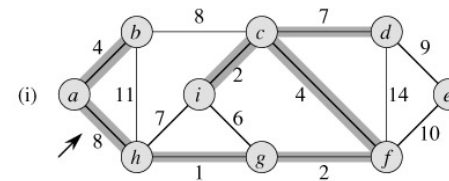
2 for each vertex $v \in G.V$

3 **MAKE-SET**(v)

4 sort the edges of $G.E$ into nondecreasing order by weight w

5 for each edge $(u, v) \in G.E$, taken in nondecreasing order by weight

6 if FIND-SET(u) \neq FIND-SET(v)

$$7 \quad A = A \cup \{(u, v)\}$$
8 UNION(u, v)9 return A 

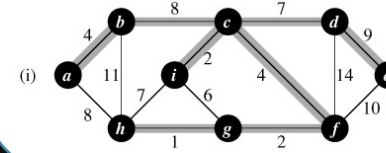
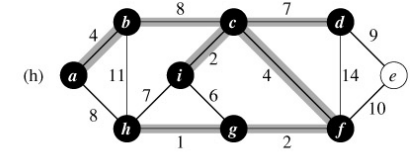
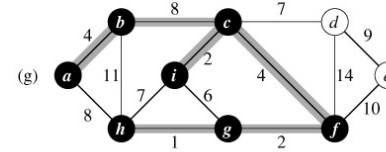
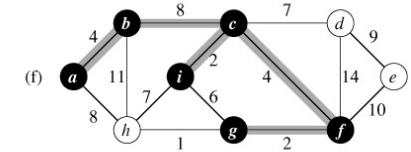
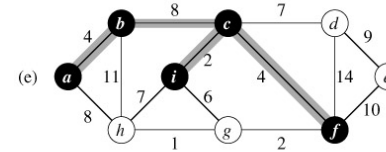
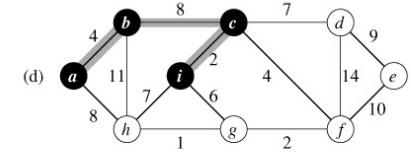
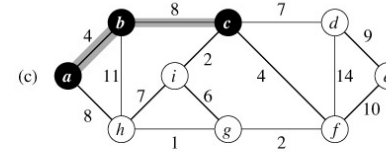
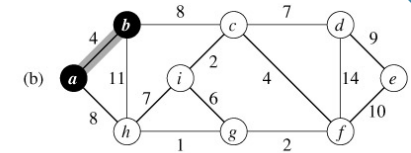
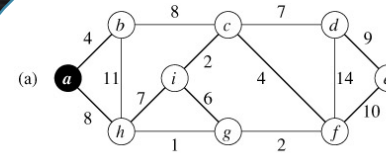
Prim's Algorithm

MST-PRIM(G, w, r)

```

1  for each  $u \in G.V$ 
2     $u.key = \infty$ 
3     $u.\pi = \text{NIL}$ 
4   $r.key = 0$ 
5   $Q = G.V$ 
6  while  $Q \neq \emptyset$ 
7     $u = \text{EXTRACT-MIN}(Q)$ 
8    for each  $v \in G.Adj[u]$ 
9      if  $v \in Q$  and  $w(u, v) < v.key$ 
10        $v.\pi = u$ 
11        $v.key = w(u, v)$ 

```



Practice Problem

- Is the path between two vertices in an MST necessarily a shortest path between the two vertices in the full graph? Give a proof or a counterexample
 - Answer: No it is not. For example, consider a graph that forms a single n -vertex cycle. The minimum spanning tree will remove just one edge (u, v) , significantly increasing the distance between u and v

Breadth First Search (BFS)

- Given a graph $G = (V, E)$, BFS starts at some **source vertex s** and discovers which vertices are **reachable from s**
- The **distance** between a vertex v and s is the minimum number of edges on a path from s to v (the shortest path length)
- BFS discovers vertices in increasing order of distance, and hence can be used as an algorithm for computing **shortest paths** from s
- At any given time there is a **frontier** of vertices that have been discovered, but not yet processed.
- BFS first visits all vertices across the **breadth** of this frontier (hence the name)

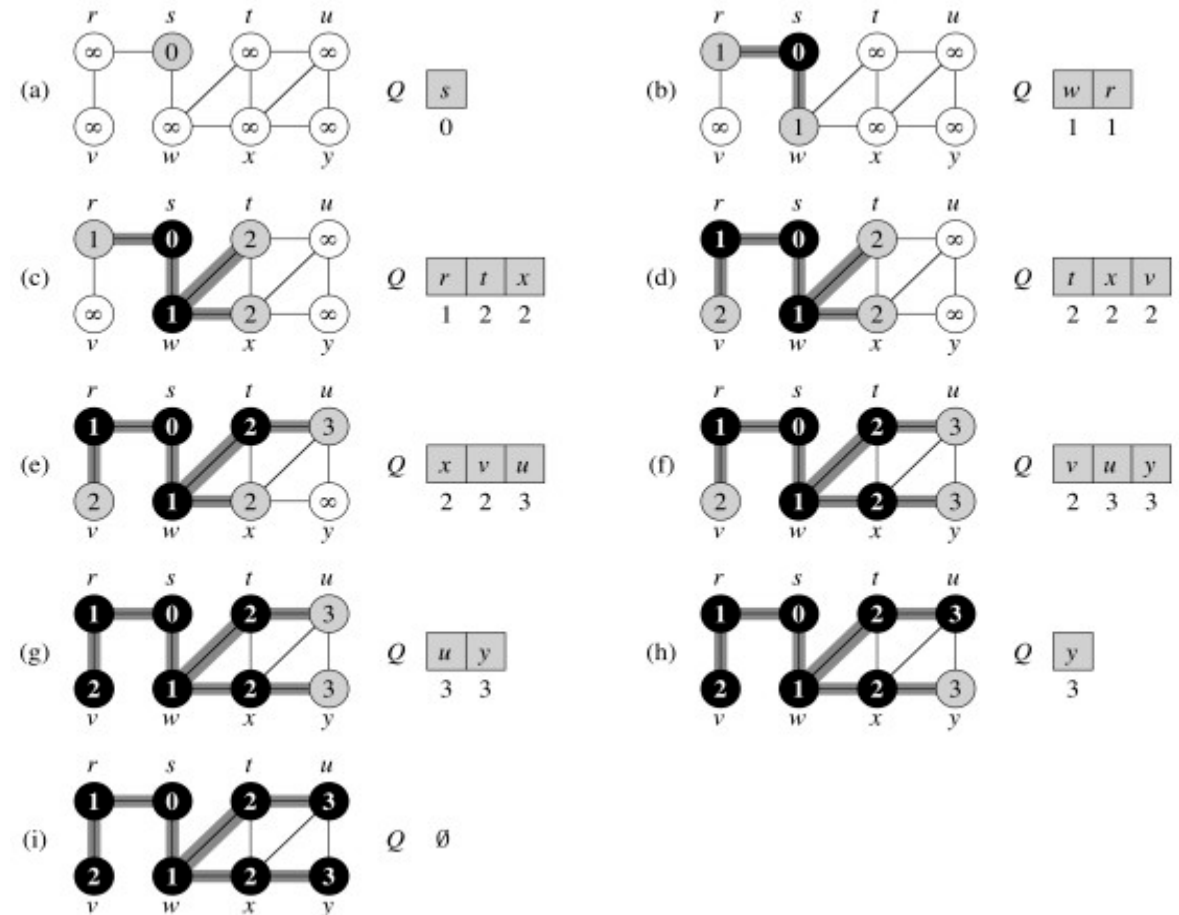
BFS: coloring

- We will use the following coloring procedure to show the status of BFS at each instance of time:
 - Initially all vertices (except the source) are colored **white**, meaning that they are **undiscovered**
 - When a vertex has first been **discovered**, it is colored **gray** (and is part of the frontier)
 - When a gray vertex is **processed**, it becomes **black**

BFS Algorithm

BFS (G, s)

1. for each $u \in G.V - \{s\}$
2. $u.color = WHITE$
3. $u.d = \infty$
4. $u.\pi = NIL$
5. $s.color = GRAY$
6. $s.d = 0$
7. $s.\pi = NIL$
8. $Q = \emptyset$
9. ENQUEUE(Q, s)
10. while $Q \neq \emptyset$
11. $u = DEQUEUE(Q)$
12. for each $v \in G.Adj[u]$
13. if $v.color == WHITE$
14. $v.color = GRAY$
15. $v.d = u.d + 1$
16. $v.\pi = u$
17. ENQUEUE(Q, v)
18. $u.color = BLACK$



BFS predecessor subgraph of G

For a graph $G = (V, E)$ with source s the predecessor subgraph of G is:

- $G_\pi = (V_\pi, E_\pi)$, where
- $V_\pi = \{v \in V : v.\pi \neq \text{NIL}\} \cup \{s\}$
- $E_\pi = \{(v.\pi, v) : v \in V_\pi - \{s\}\}$

DFS Algorithm

DFS (G)

1. for each vertex $u \in G.V$
2. $u.color = WHITE$
3. $u.\pi = NIL$
4. time = 0
5. for each vertex $u \in G.V$
6. if $u.color = WHITE$
7. DFS-Visit(G, u)

DFS-Visit(G, u)

1. time = time + 1
2. $u.d = time$
3. $u.color = GRAY$
4. for each vertex $v \in G.Adj[u]$
5. if $v.color == WHITE$
6. $v.\pi = u$
7. DFS-Visit(G, v)
8. $u.color = BLACK$
9. time = time + 1
10. $u.f = time$

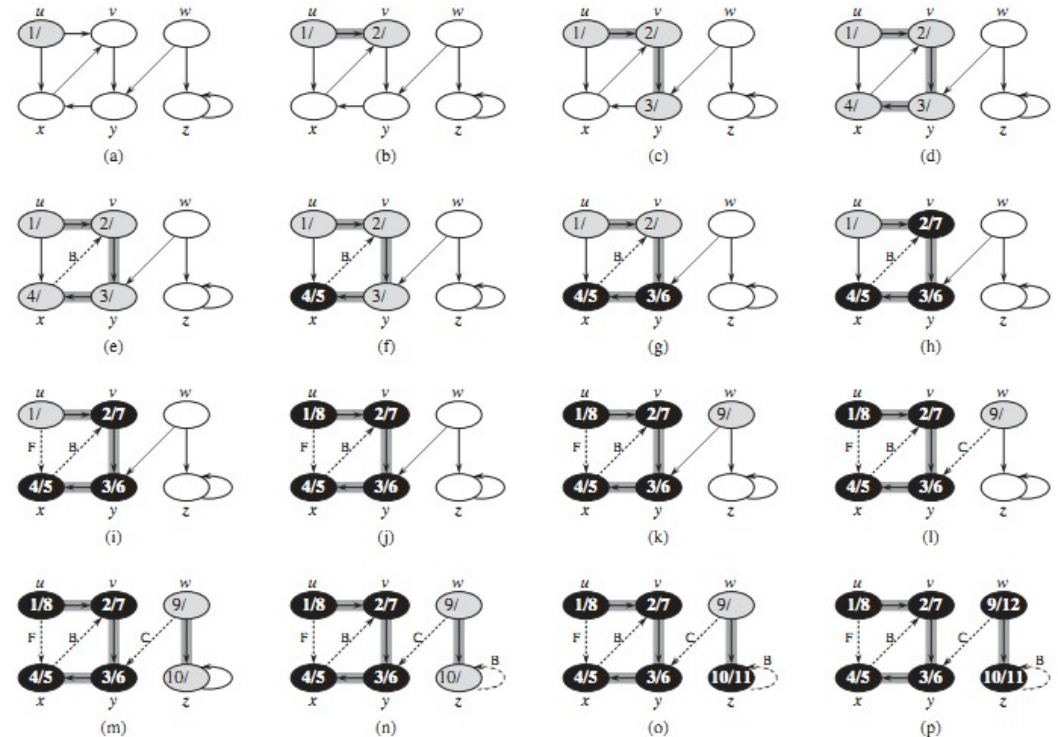
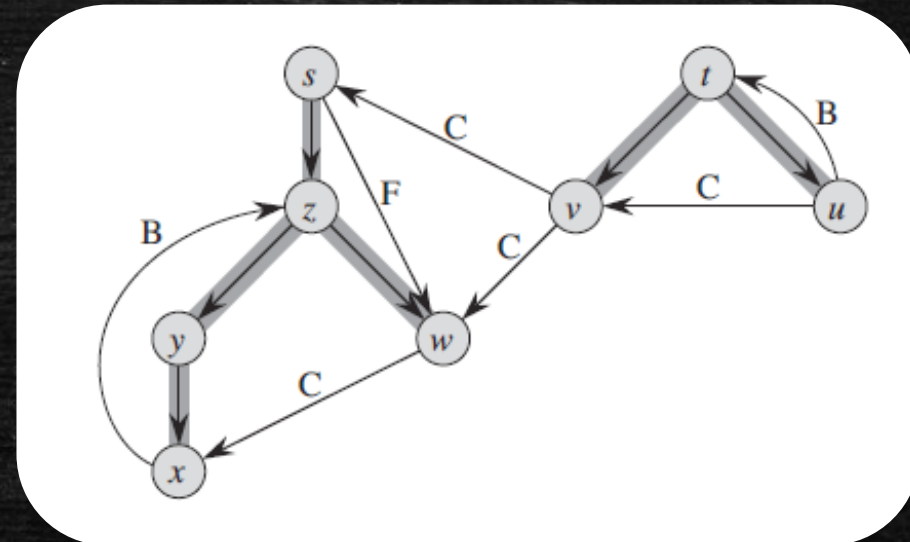


Figure 22.4 The progress of the depth-first-search algorithm DFS on a directed graph. As edges are explored by the algorithm, they are shown as either shaded (if they are tree edges) or dashed (otherwise). Nontree edges are labeled B, C, or F according to whether they are back, cross, or forward edges. Timestamps within vertices indicate discovery time/finishing times.

Depth-First Forest Edges

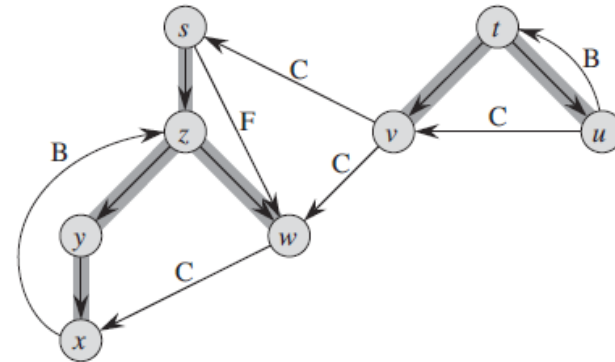
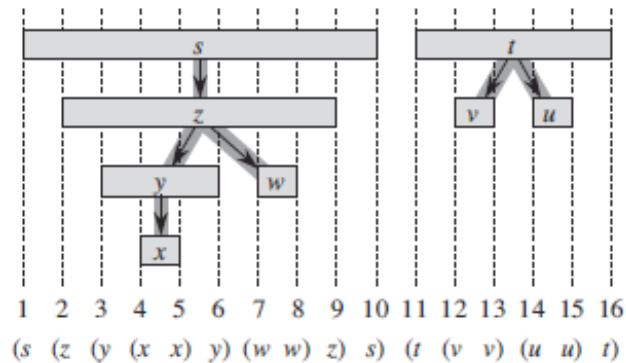
Classification of edges based on depth-first forest:

- **Tree Edges:** Edges in the depth-first forest G_π
- **Back Edges:** connecting a vertex to an ancestor in the depth-first tree
- **Forward Edges:** connecting a vertex to a descendant in the depth-first tree
- **Cross edges:** all other edges



Cycles in a Graph

- Time stamps of DFS help determine if a graph $G = (V, E)$ contains any cycles
- Consider any DFS forest of G , and consider any edge (u, v) in E :
 - If this edge is a **tree**, **forward**, or **cross** edge, then $u.f > v.f$
 - If the edge is a **back** edge then $u.f < v.f$
- G has a **cycle** if and only if the DFS forest has a **back edge**
- Checking if G is acyclic reduces to checking if it has a back edge

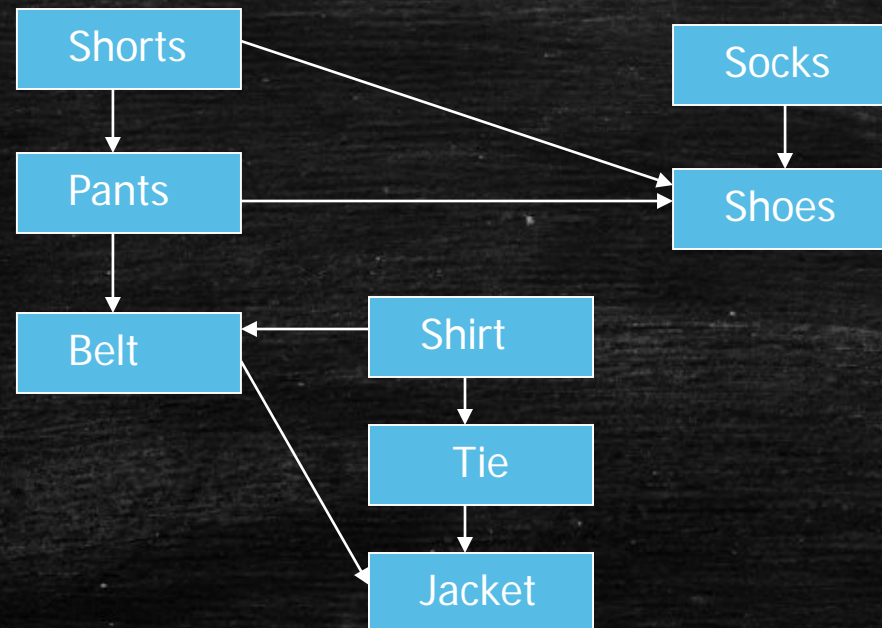


Directed Acyclic Graph

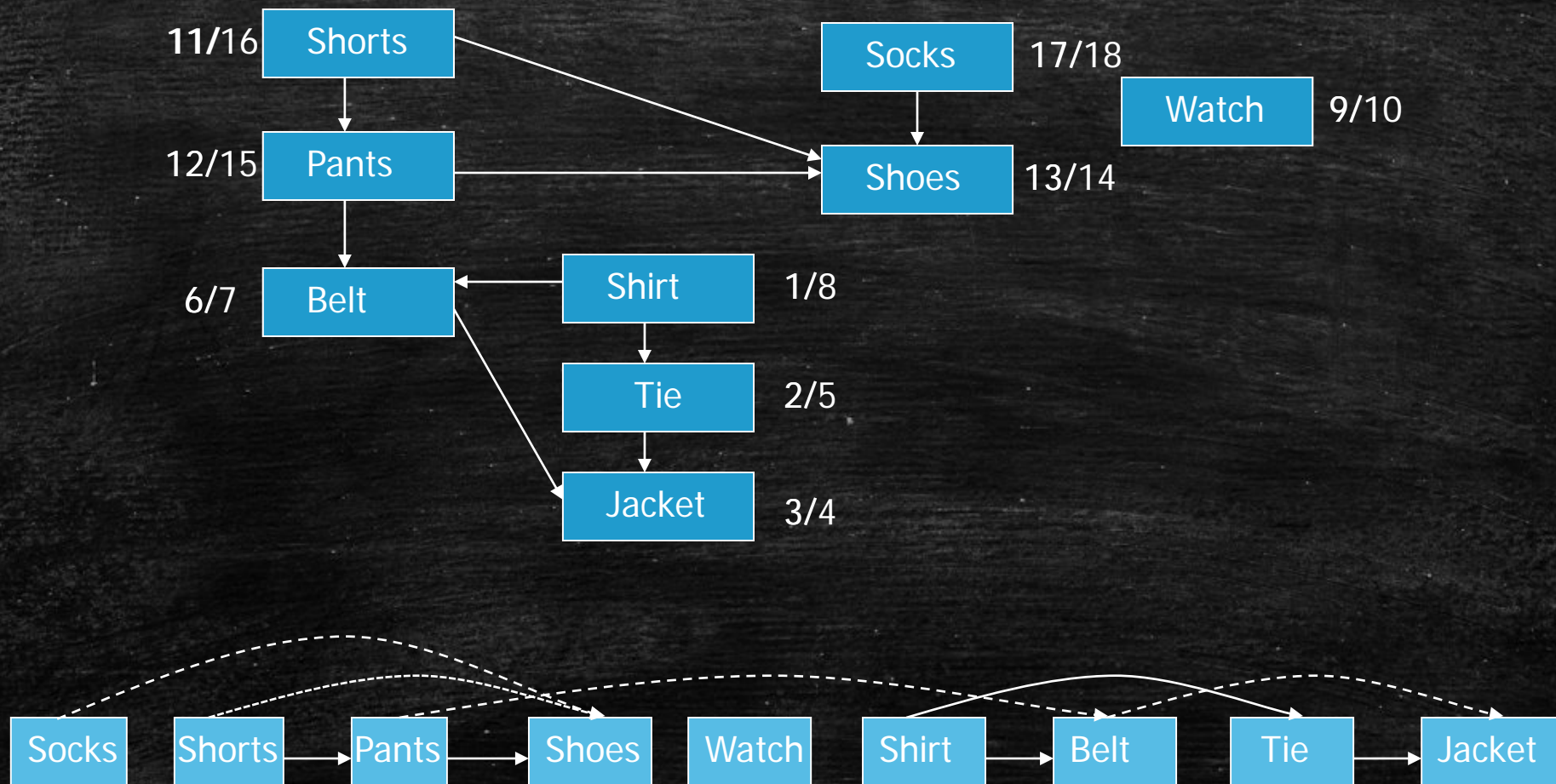
- A **directed acyclic graph** is often called a **DAG** for short.
- DAG's arise in many applications with **precedence / ordering constraints**
- In general a precedence constraint graph is a DAG in which
 - vertices are tasks and
 - the edge (u, v) means that task u must be completed before task v begins.

Topological Sort

- A **topological sort** of a DAG is a linear ordering of the vertices of the DAG such that for each edge (u, v) , u appears before v in the ordering.



Example



Topological Sort Algorithm

TopologicalSort(G)

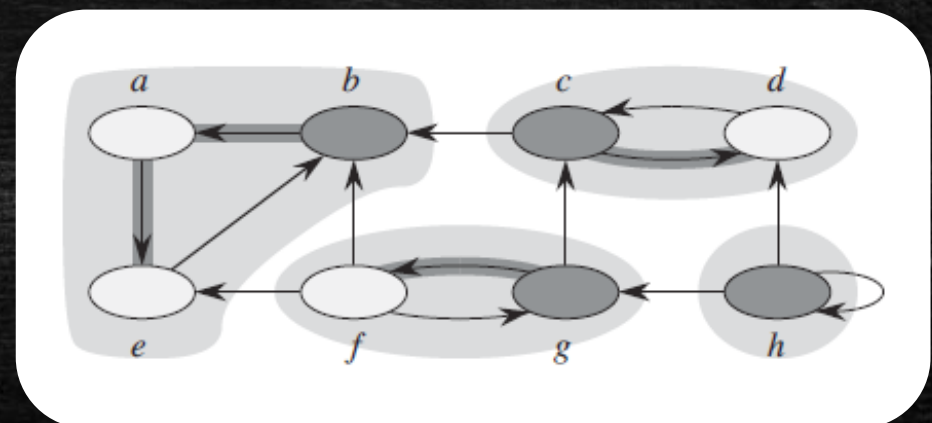
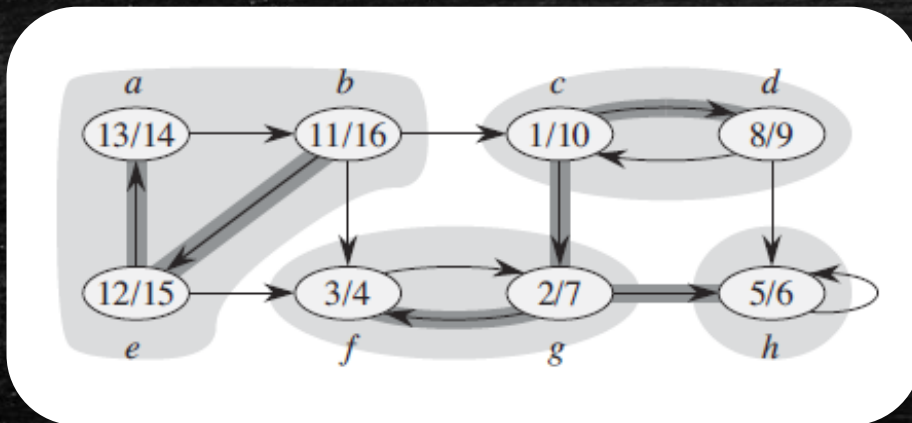
1. call DFS(G) to compute finishing times $v.f$ for each vertex v
2. as each vertex is finished, insert it onto the front of a linked list
3. return the linked list of vertices

- Running time
 - DFS: $\Theta(n + m)$
 - Insertion to linked list: $\Theta(n)$

Strongly Connected Components

Strongly Connected Components(G)

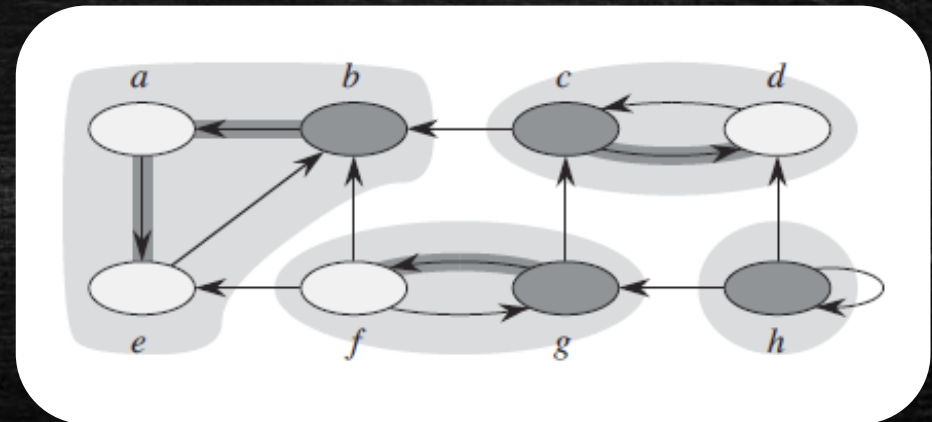
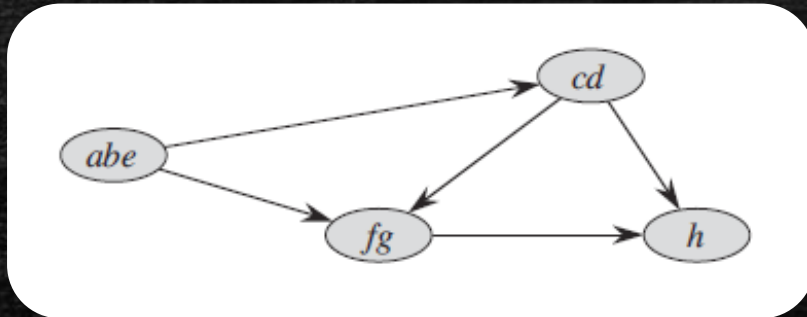
1. call DFS(G) to compute finishing times $v.f$ for each vertex v
2. compute G^T
3. call DFS(G^T), but in the main loop of DFS, use decreasing order w.r.t. $v.f$
4. return the vertices of each tree in the depth-first forest formed in line 3 as a separate strongly connected component



Strongly Connected Components

Strongly Connected Components(G)

1. call DFS(G) to compute finishing times $v.f$ for each vertex v
2. compute G^T
3. call DFS(G^T), but in the main loop of DFS, use decreasing order w.r.t. $v.f$
4. return the vertices of each tree in the depth-first forest formed in line 3 as a separate strongly connected component



Practice Problems

- A root of a DAG is a vertex r such that every other vertex of the DAG can be reached from r using a directed path. Give an algorithm that determines whether a given DAG has a root.
- Provide an algorithm that, given a directed acyclic graph $G = (V, E)$ and two vertices $s, t \in V$, returns the number of simple paths from s to t in G .