

CSC 317: Data Structures and Algorithm Analysis

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# Outline I

- Adjacency-Matrix Representation of Graphs
  - Adjacency-List Representation
  - Edge-List Representation
  - Comparison of Complexities of Operations for Different Graph Representations
  - Breadth-first search
  - Breadth-first trees
  - Overview of the DFS algorithm
  - DFS Algorithm
  - Complexity analysis of the DFS algorithm
  - Properties of the DFS algorithm
  - Application of DFS: Topological Sorting or Ordering Algorithm
  - Application of DFS: Strongly connected components

# Adjacency-Matrix Representation of Undirected Unweighted Graphs

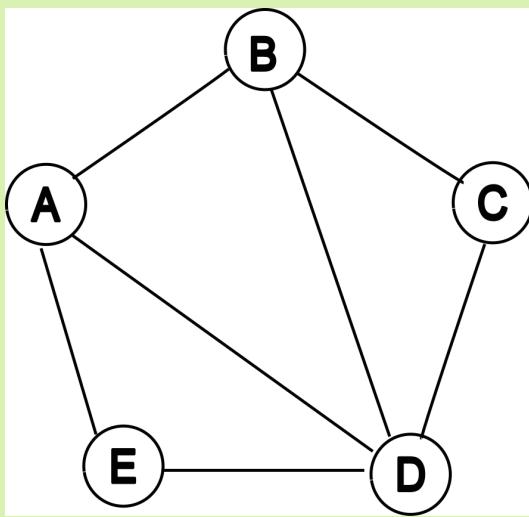


Figure: A simple undirected unweighted graph.

	A	B	C	D	E
A	-	1	0	1	1
B	1	-	1	1	0
C	0	1	-	1	0
D	1	1	1	-	1
E	1	0	0	1	-

Adjacency-matrix representation of simple (no self-loop) undirected unweighted graph.

1: an edge, 0: no edge —: no self-loop

# Adjacency-Matrix Representation of Directed Unweighted Graphs

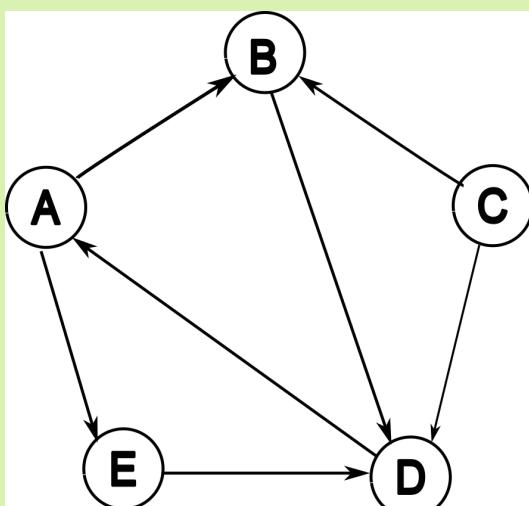


Figure: A Simple Directed Unweighted Graph.

	A	B	C	D	E
A	-	1	0	0	1
B	0	-	0	1	0
C	0	1	-	1	0
D	1	0	0	-	0
E	0	0	0	1	-

Adjacency-matrix representation of simple (no self-loop) directed unweighted graph.

1: an edge, 0: no edge —: no self-loop

# Adjacency-Matrix Representation of Directed Weighted Graphs

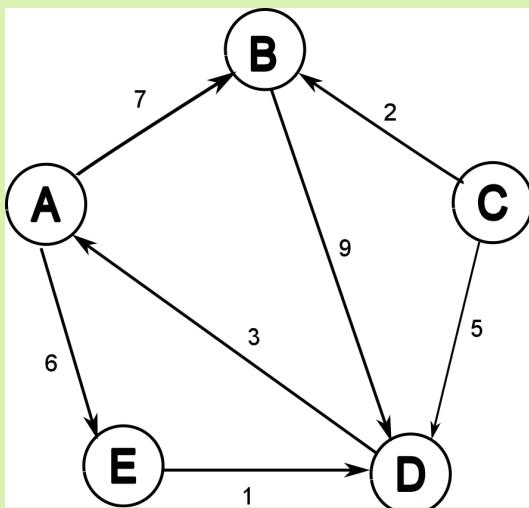


Figure: A Simple Directed Weighted Graph

	A	B	C	D	E
A	-	7	0	0	6
B	0	-	0	9	0
C	0	2	-	5	0
D	3	0	0	-	0
E	0	0	0	1	-

Adjacency-matrix representation of simple (no self-loop) directed unweighted graph.

$w \neq 0$ : an edge with weight  $w$ , 0: no edge —: no self-loop

# Adjacency-List Representation of Undirected Unweighted Graphs

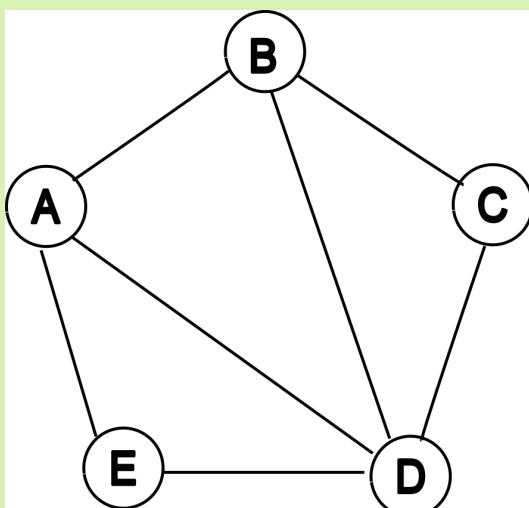


Figure: A simple undirected unweighted graph.

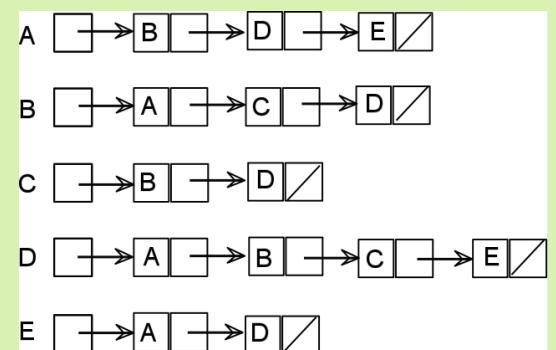


Figure: Adjacency-list representation of a graph

# Adjacency-List Representation of Directed Weighted Graphs

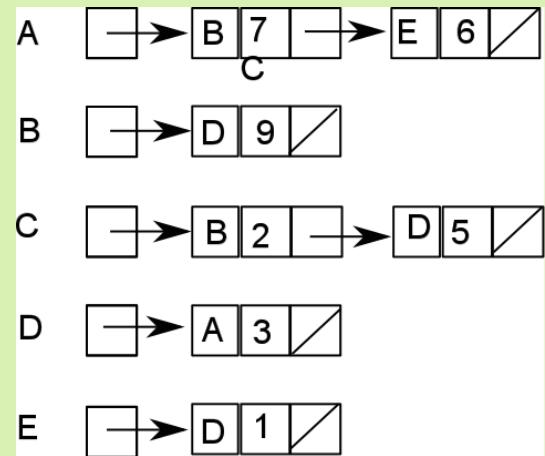
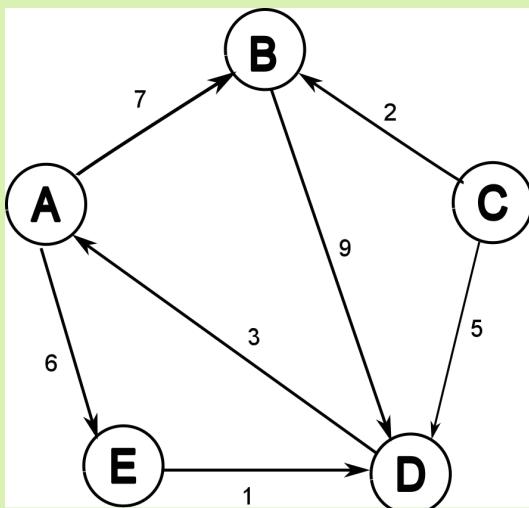
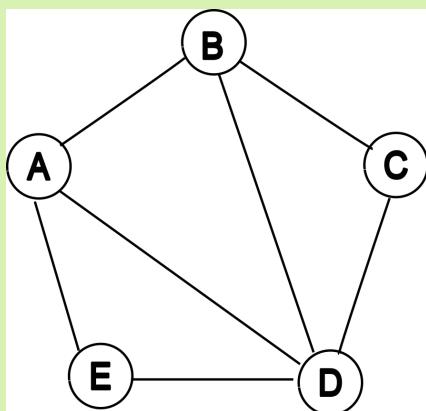


Figure: A Simple Directed Weighted Graph

Figure: Adjacency-list representation of directed weighted graph

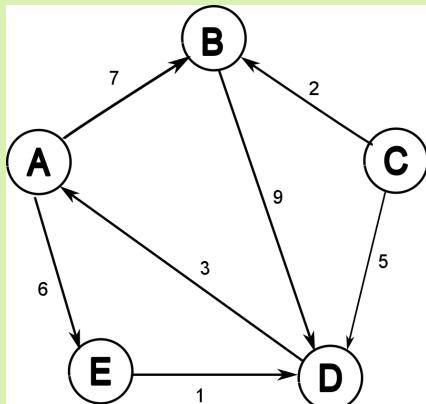
## Edge-List Representation of Undirected Unweighted Graphs



End 1	A	A	A	B	B	C	D
End 2	B	D	E	C	D	D	E
Weight	1	1	1	1	1	1	1

Figure: A simple undirected unweighted graph.

# Edge-List Representation of Directed Weighted Graphs



End 1	A	A	B	C	C	D	E
End 2	B	E	D	B	E	A	D
Weight	7	6	9	2	5	3	1

## Figure: A Simple Directed Weighted Graph

## Complexity Comparison

Operation	Adjacency matrix	Adjacency list	Edge list
Find an edge	$O(1)$	$O(n)$	$O(m)$
Degree of a node (or # of neighbors)	$O(n)$	$O(n)$	$O(m)$
Minimum weight neighbor	$O(n)$	$O(n)$	$O(m)$
Maximum weight neighbor	$O(n)$	$O(n)$	$O(m)$

Table: Comparison of Complexities of Graph Representations

# Graph Search Algorithms

- **Notations:**  $G(V, E)$  denotes a **graph** with a set of  $V$  **vertices or nodes** and  $E$  is a set of **edges**.

$V = \{v_1, v_2, \dots, c_n\}$  denotes the set of  $n$  edges and

$E = \{e_1, e_2, \dots, e_m\}$  denotes the set of  $m$  edges such that  $e_i = (v_{i_1}, v_{i_2})$ .

- A graph  $G(V, E)$  is a generalization of a tree, because

A tree with  $n$  nodes has exactly  $n - 1$  edges and

Between two nodes  $v_i$  and  $v_j$ , there is only one path.

- A graph with  $n$  nodes can have 0 to  $\frac{n(n-1)}{2}$  edges.

- **Graph Search:** Two possibilities: Visit every node and visit every edge; a third would be visit every node **and** every edge.

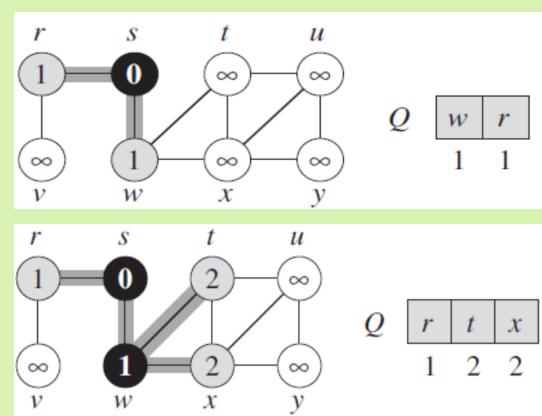
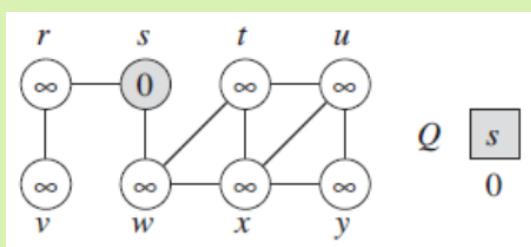
- We will learn two elementary graph search algorithms and their applications:

- **Breadth-first-search:** Shortest path and minimum spanning tree for unweighted graphs.
  - **Depth-first-search:** Minimal spanning tree, Topological sorting, strongly connected components in a graph, AI search algorithms.

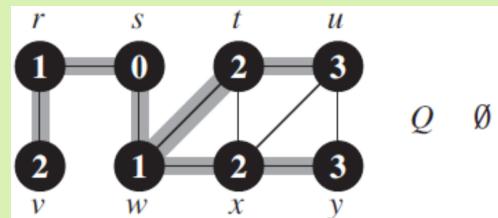
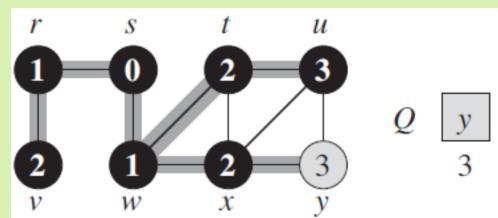
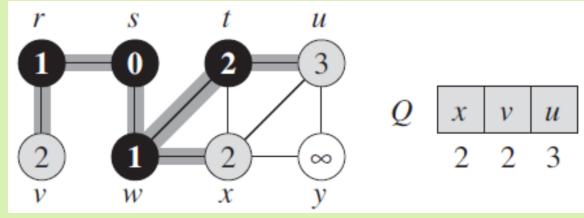
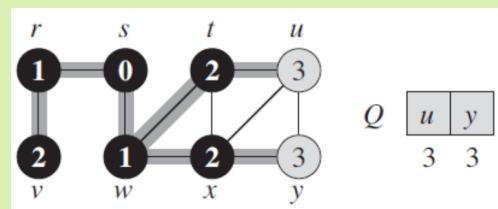
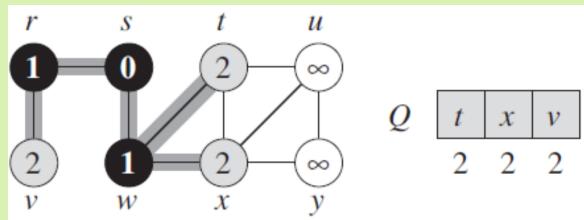
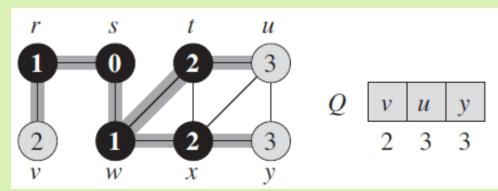
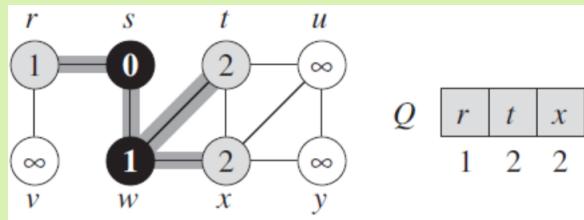
## Breadth-first search (BFS) An example – I

## Trace of execution of a BFS algorithms on a graph

- Two data structures are used:
    - Representation of graph adjacency list
    - A FIFO (first-in-first-out) queue,  $Q$ , for nodes to be considered next



## Breadth-first search (BFS): An example – II



# Breadth-first search (BFS) Algorithm

### BFS( $G, s$ )

```

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

```

- Data structures:
    - Representation of graph
    - adjacency list
    - A FIFO (first-in-first-out) queue,  
 $Q$  for nodes to be considered next

# Complexity Analysis I

```

BFS( $G, s$ )
1   for each vertex  $u \in G.V - \{s\}$ 
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13          if  $v.color == \text{WHITE}$ 
14               $v.color = \text{GRAY}$ 
15               $v.d = u.d + 1$ 
16               $v.\pi = u$ 
17              ENQUEUE( $Q, v$ )
18       $u.color = \text{BLACK}$ 

```

- Initialization: Lines 1 to 4 time complexity  $O(n)$ , because the loop is executed  $(n - 1)$  times  
Lines 5 to 9 are executed only once.
  - While** loop is executed  $O(n)$  time,  
Because a node is enqueued only once at Line 17
  - The **for** loop for each node is executed for each edge  
Each edge is visited twice, once from each node it connects
  - Total complexity is thus  $m$ , the number of edges.
  - Complexity of BSF** is  $O(|V| + |E|) = O(n + m)$

# Complexity Analysis II

- **Notations:** Shortest-path distance - number of edges - from  $s$  to  $v$  is denoted as  $\delta(s, v)$

### Lemma (Lemma 22.1)

Let  $G = (V, E)$  be a directed or undirected graph, and let  $s \in V$  be an arbitrary vertex. Then, for any edge  $(u, v) \in E$

$$\delta(s, v) \leq \delta(s, u) + 1.$$

Proof.

Outline: Algorithm Lines 11 and 12: we reach  $v$  from  $u$ . Thus, distance from  $s$  to  $v$  is at least one more than that for  $u$ .  $\square$

# Complexity Analysis III

Lemma (Lemma 22.2)

Let  $G = (V, E)$  be a directed or undirected graph, and suppose that BFS is run on  $G$  from a given source vertex  $s \in V$ . Then upon termination, for each vertex  $v \in V$  the value of  $v.d \geq \delta(s, v)$ .

## Proof.

Outline: Proof is by induction on the number of ENQUEUE operations. At Line 2  $v.d$  is assigned to inf and at Line 6  $v.d$  is assigned '0'.

Induction hypothesis holds for  $s$ , because  $s.d = 0 = \delta(s, s)$ , and  $v.d = \inf \geq \delta(s, v)$  for all  $v \in (V - \{s\})$ .

- From Line 15 and Lemma 22.1, we have
  - $v.d = u.d + 1 \geq \delta(s, u) + 1 \geq \delta(s, v)$



# Complexity Analysis IV

### Lemma (Lemma 22.3)

Suppose that during the execution of the BFS on a graph  $G(V, E)$ , the  $Q$  contains the vertices  $\langle v_1, v_2, \dots, v_r \rangle$ , where  $v_1$  is the head of  $Q$  and  $v_r$  is the tail. Then,  $v_r.d < v_1.d + 1$  and  $v_i.d < v_{i+1}.d$  for  $i = 1, 2, \dots, r - 1$ .

## Proof

Outline: Proof is by induction on the number of ENQUEUE operations.



# Complexity Analysis V

Theorem (Theorem 2.5)

Let  $G = (V, E)$  be a directed or undirected graph, and suppose that BFS is run on  $G$  from a given source node  $s \in V$ . Then, during its execution, BFS discovers every vertex  $v \in V$  that is reachable from the source node  $s$ , and upon termination,  $v.d = \delta(s, v)$  for all  $v \in V$ .

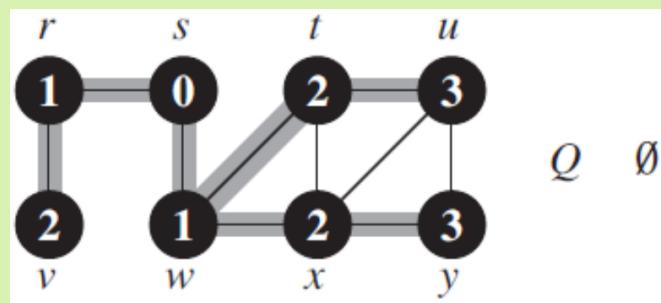
Moreover, for any vertex  $v \neq s$  that is reachable from  $s$ , one of the shortest paths from  $s$  to  $v$  is a shortest path from  $s$  to  $v.\pi$  followed by the edge  $(v.\pi, v)$ .

## Proof.

Outline: Proof is by contradiction. Assume that some vertex receives a  $d$  value not equal to its shortest-path distance and show that this contradicts following the steps of the algorithm.

# Breadt-first trees

Corresponding to the  $\pi$  attributes we have tree.



The highlighted edges corresponds the BFS tree.

- Formally, for a given graph  $G(V, E)$  with source node  $s$ , we define the **predecessor subgraph** of  $G$  as  $G_\pi(V_\pi, E_\pi)$ , where

$$V_\pi = \{v \in V : v.\pi \neq \text{NIL}\} \cup \{s\}$$

and

$$E_\pi = \{(v.\pi, v) : V.\pi \neq \text{NIL}\} - \{s\}$$

- The subgraph  $G_\pi$  is a *breadth-first tree*.

# DFS: Depth-first search

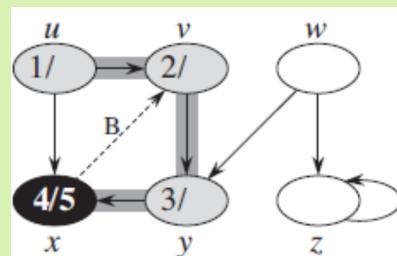
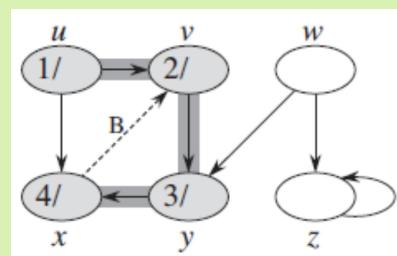
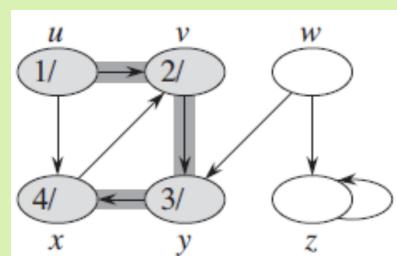
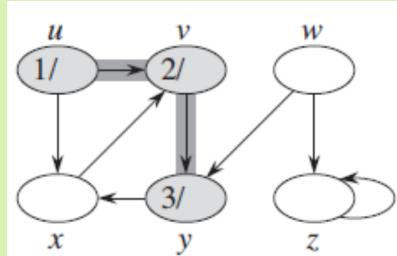
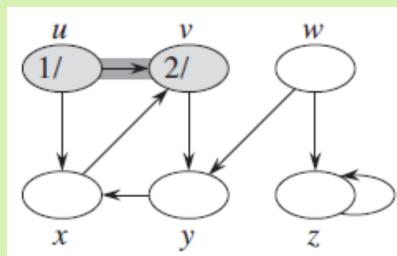
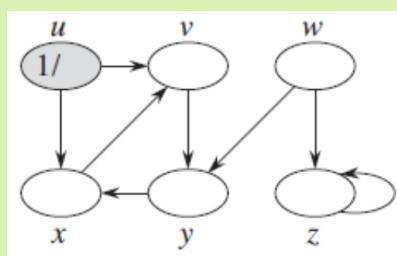
- Search **deeper** in the graph, if possible
  - Explore edges from the **most recently** discovered vertex  $v$
  - Once **all edges** from the vertex  $v$  **has been explored**
  - **Backtrack** to the vertex  $u$ , from where vertex  $v$  was discovered
  - **Repeat** DFS from all undiscovered vertices

This is DIFFERENT from BFS

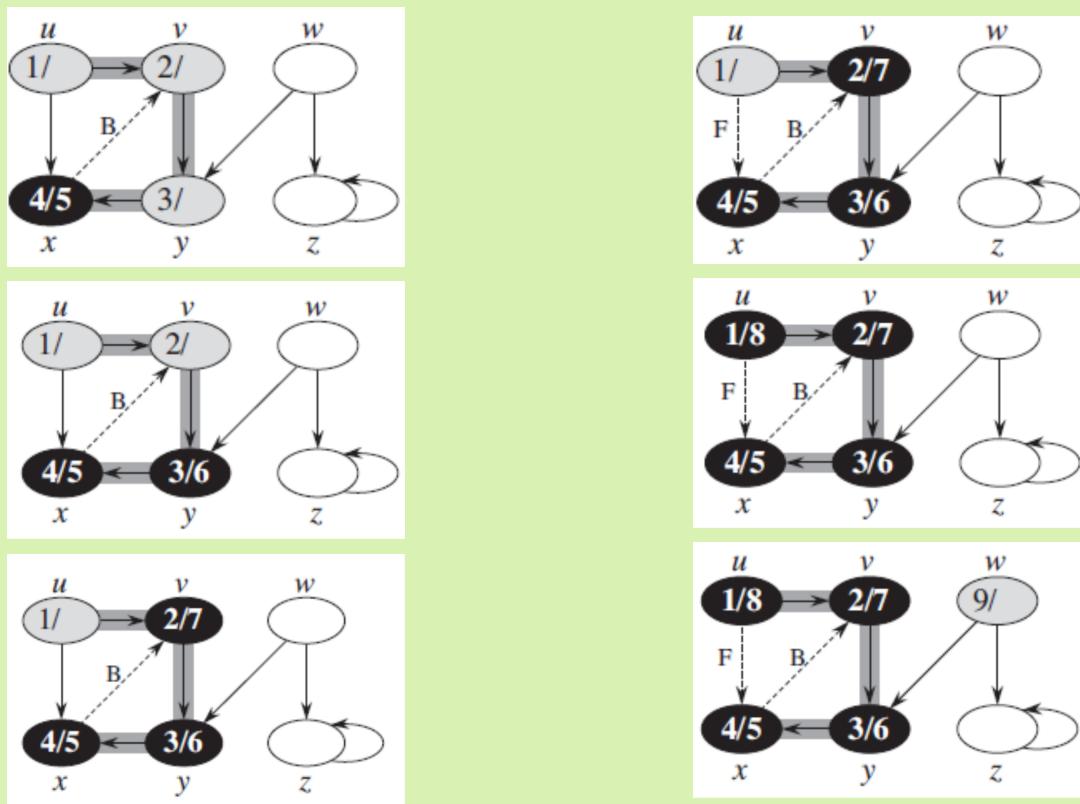
- For disconnected graphs DFS may produce multiple DFS-trees  
Called DFS forest
  - Every node will be marked twice:
    - $v.d$  stores 'time/step' when  $v$  is discovered
    - $v.f$  stores 'time/step' when DFS finish visiting all nodes reachable from  $v$

It is obvious that  $v.d < v.f$
  - Time stamps at a node stores important information about structure of the graph.

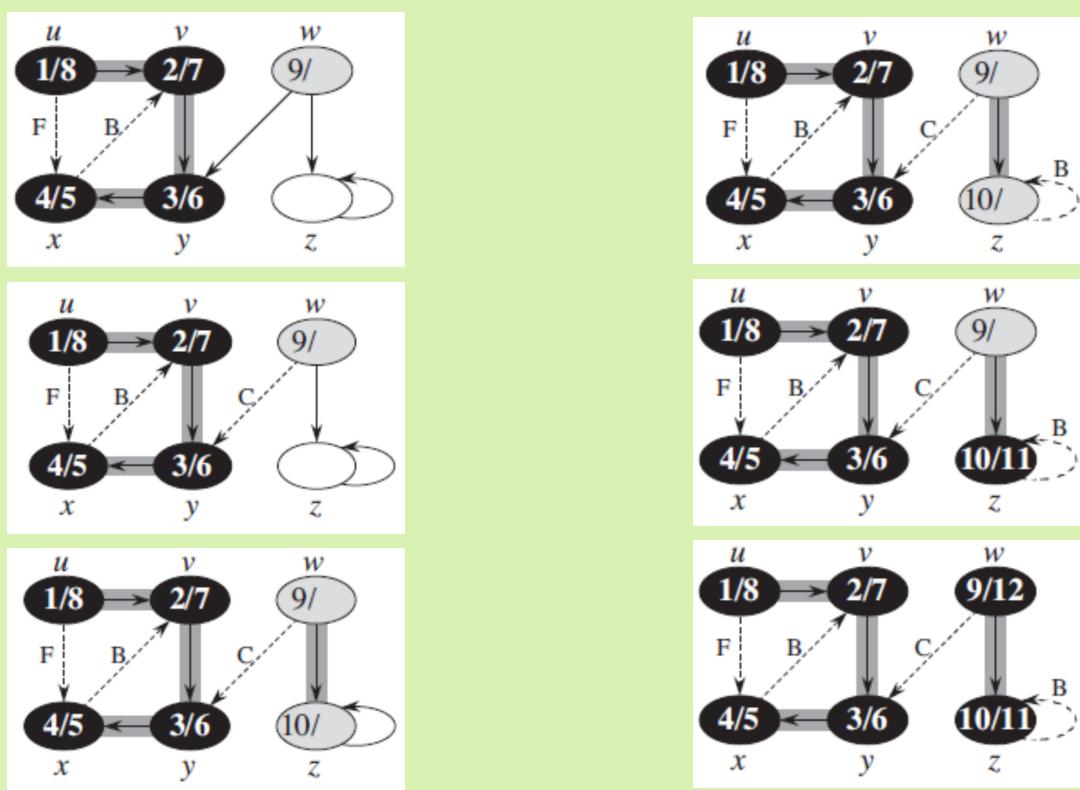
## Trace of execution of a DFS on a graph I



# Trace of execution of a DFS on a graph II



## Trace of execution of a DFS on a graph III



# DFS Algorithm

**DFS**( $G$ )

```

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

```

**DFS-VISIT**( $G, u$ )

```

1   time = time + 1
2   u.d = time
3   u.color = GRAY
4   for each v ∈ G.Adj[u]
5       if v.color == WHITE
6           v.π = u
7           DFS-VISIT(G, v)
8   u.color = BLACK
9   time = time + 1
10  u.f = time

```

# Complexity analysis of the DFS algorithm

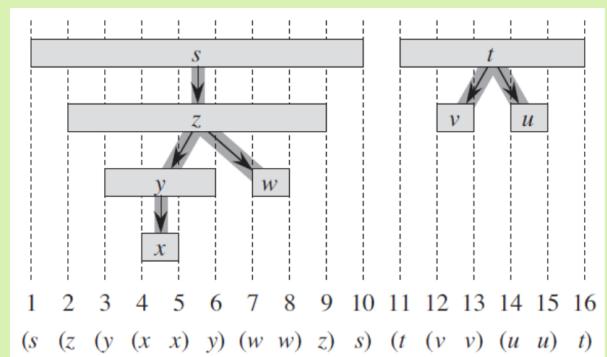
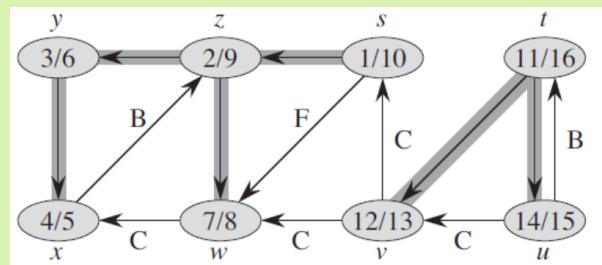
- Initialization Steps 1 to 4 in  $\text{DFS}(G)$  takes  $O(n)$  time, where  $n$  is the number of vertices.
  - For** loop in Steps 5 to 7 in  $\text{DFS}(G)$  and Steps 4 to 7 in  $\text{DFS}(G, v)$  visits each edge only once.
  - Thus, complexity of the DFS algorithm is  $O(n + m)$ , where  $m$  is the number of edges in  $G$ .

# Properties of the DFS algorithm I

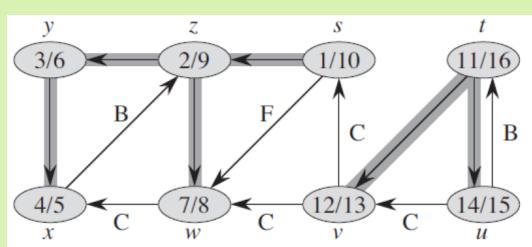
## Theorem (Parenthesis theorem)

In any depth-first search of a (directed or undirected) graph  $G(V, E)$ , for any two vertices  $u$  and  $v$ , exactly one of the following three conditions holds:

- the intervals  $[u.d, u.f]$  and  $[v.d, v.f]$  are entirely disjoint, and neither  $u$  nor  $v$  is a descendant of the other in the depth-first forest,
  - the interval  $[u.d, u.f]$  is contained entirely within the interval  $[v.d, v.f]$ , and  $u$  is a descendant of  $v$  in the depth-first tree, or
  - the interval  $[v.d, v.f]$  is contained entirely within the interval  $[u.d, u.f]$ , and  $v$  is a descendant of  $u$  in the depth-first tree.



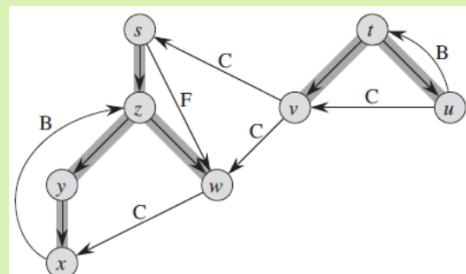
## Classification of Edges I



- **Tree Edges** are the edges in the DFS forest.

Edge  $(u, v)$  is a tree edge if  $v$  was first *discovered* from  $u$  by exploring edge  $(u, v)$ .
  - **Back edges** are those edges  $(u, v)$  connecting a vertex  $u$  to an *ancestor*  $v$  in a DFS tree.

We consider self-loops, to be a back edges.



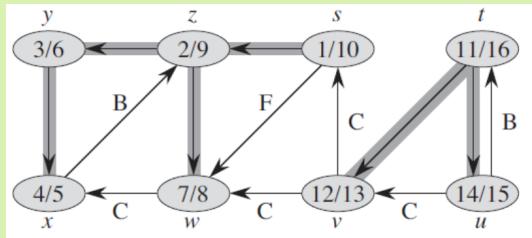
## Examples of Tree edges

$(s, z)$ ,  $(z, y)$ ,  $(y, x)$ ,  $(t, v)$  and  $(t, u)$ .

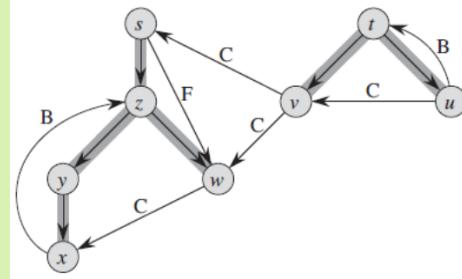
**Examples of Back edges** ( $x, z$ )  
and ( $u, t$ ).

- The **key idea**: the color of a node when explored from an *edge*
    - The color is WHITE  $\Rightarrow$  a tree edge.
    - The color is GRAY  $\Rightarrow$  a back edge.

## Classification of Edges II



- **Forward edges** are those *nontree* edges  $(u, v)$  connecting a vertex  $u$  to a descendant  $v$  in the DFS tree.
- **Cross edges** are all other edges. They can go between vertices in the same DFS tree, as long as one vertex is not an ancestor of the other, or they can go between vertices in different DFS trees.



**Examples of a Forward edge**  
 $(s, w)$ .

**Examples of Cross edges**  
 $(w, x), (v, w), (v, s)$  and  $(u, v)$ .

- The **key idea**: the color of a node when explored from an edge
  - The color is BLACK  $\Rightarrow$  a forward or a cross edge.

## Motivation for Topological Sorting or Ordering Algorithm I

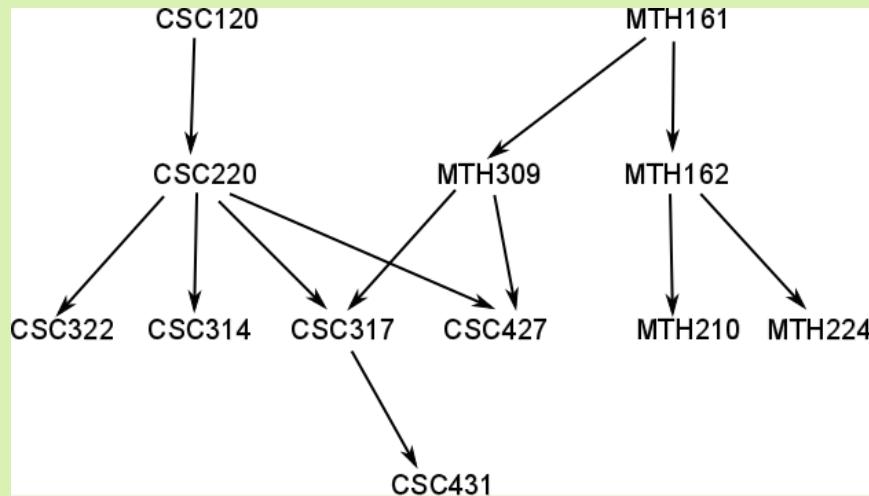
- Why we need topological sorting or ordering?
- The list of **required** course for a CS major;
  - Computer Science courses CSC120, CSC220, CSC314, CSC322, CSC317, CSC427, and CSC 431
  - Math courses MTH161, MTH162, MTH210, MTH224, and MTH309
  - CSC120 and MTH161 have no prerequisites (assuming a student has done enough math in high school)
  - A student wants to prepare a plan for completing both requirements

Table: Prerequisites for required CSC and MTH courses.

Course	Prerequisite(s)
CSC120	NONE
CSC220	CSC120
CSC322	CSC220
CSC314	CSC220
CSC317	CSC220, MTH309
CSC427	CSC220, MTH309
CSC431	CSC317
MTH161	NONE
MTH162	MTH161
MTH210	MTH162
MTH224	MTH162
MTH309	MTH161

## Motivation for Topological Sorting or Ordering Algorithm II

- How to make a sequential list so that if a student take courses from left to right, then prerequisites are satisfied?
- Draw a prerequisite graph.



- Notice, this is a *directed acyclic graph* (DAG).
- If the graph had a cycle, no solution exists.

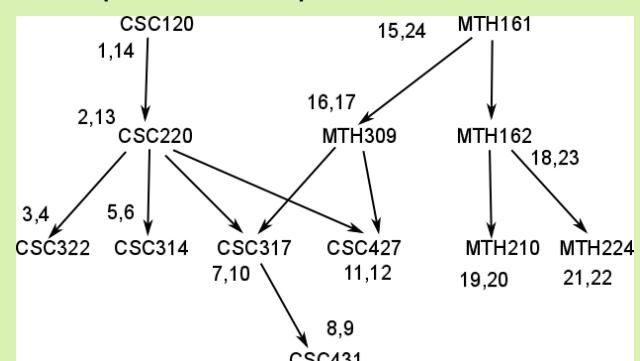
## Topological Sorting or Ordering Algorithm

**TOPOLOGICAL-SORT( $G$ )**

- 1  $L$ : an empty linked-list
- 2 call  $\text{DFS}(G)$
- 3 as each vertex is finished,  
insert it onto the front of  $L$

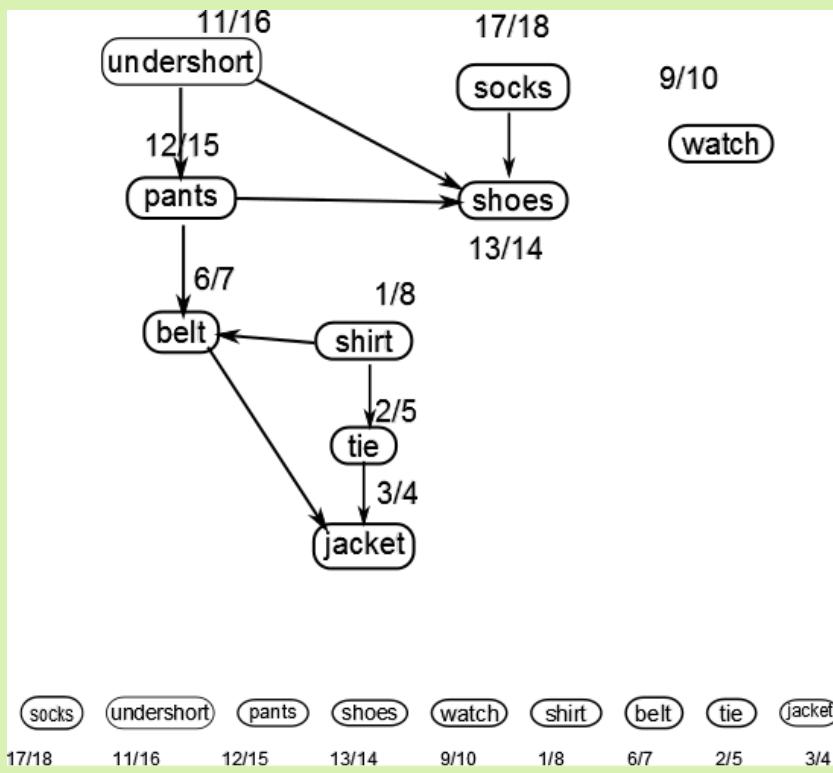
- 4 **return** the linked-list,  $L$ , of vertices

- Prerequisite Example:



- $L = \text{MTH161}, \text{MTH162}, \text{MTH224}, \text{MTH210}, \text{MTH309}, \text{CSC120}, \text{CSC220}, \text{CSC427}, \text{CSC317}, \text{CSC431}, \text{CSC314}, \text{CSC322}$

Topological Sorting or Ordering Algorithm: Example in the book



- Dressing dependence graph
    - Each node has two numbers:
      - first visit time
      - final visit time
  - The list below is sorted in decreasing order final visit time
  - The order may not be unique for a DAG
  - If you start with undershort, you will get a different list

Topological Sorting or Ordering Algorithm: Example in the book

### Lemma (Lemma 22.11)

A directed graph  $G$  is acyclic if and only if a DFS of  $G$  yields no back edges.

## Proof.

Outline:  $\Rightarrow$  If there is a back edge  $(u, v)$ , then  $v$  is an ancestor of  $u$ . Thus, there is a path from  $v$  to  $u$  and a directed edge from  $u$  to  $v$ , which completes a cycle.

$\Leftarrow$  If  $G$  contains a cycle, we have to show that there is a back edge.  $\square$

### Theorem (Theorem 22.12)

**TOPOLOGICAL-SORT** produces a topological sort of the directed acyclic graph provided as its input.

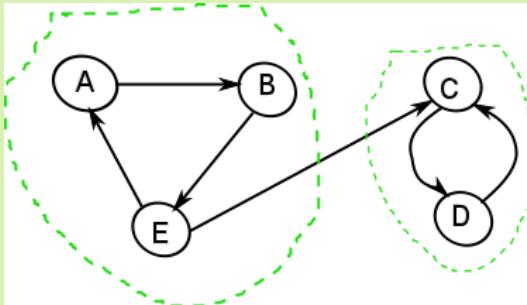
## Proof.

Please read the book.

# Strongly connected components I

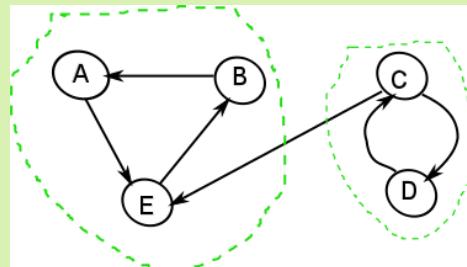
**Transpose graph**  $G(V, E^T)$ : For a directed graph  $G(V, E)$  if direction of every edge  $e = (u, v)$  is reversed, that is,  $e^T = (v, u)$ . And  $E^T = \{(v, u) : \text{such that } (u, v) \in E\}$ . We will denote  $G(V, E^T)$  by  $G^T$ .

- A directed graph G



- In the graph  $G$ , there exists paths from vertices A, B, and E to vertices C or D. But no path to vertices A, B, and E from vertices C or D.

- Transpose graph  $G^T$  of the graph  $G$



- Direction of every edge of  $G$  is reversed to obtain  $G^T$ .
  - In  $G^T$  there is no path from vertices A, B, and E to vertices C or D.
  - But in  $G^T$  there is path to vertices A, B, and E from vertices C or D.

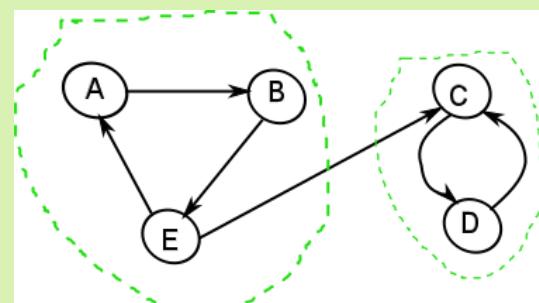
## Strongly connected components II

**Strongly connected graph (SCG):**  
A directed graph is **strongly connected**, if there is a directed path from  $u$  to  $v$ , then there is also a directed path from vertex  $v$  to  $u$  for every vertex pair of  $u$  and  $v$ .

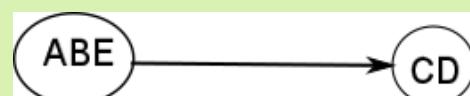
All directed graph may not be SCGs. There are many examples: social networks, software module dependency graph, food web in a ecosystem, and connections of neurons in our brain.

**Strongly connected component** (SSC) of a directed graph  $G(V, E)$  is maximal set of vertices  $C \subseteq V$ , such that  $C$  forms a SCG.

## A directed graph with two SCC



## SCC graph



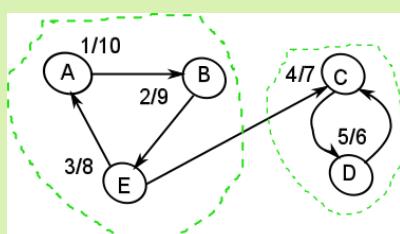
# An Algorithm for finding strongly connected components

STRONGLY-CONNECTED-COMPONENTS( $G$ )

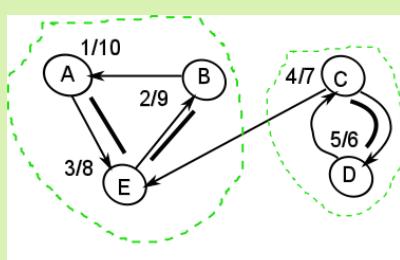
- 1 call DFS( $G$ ) to compute finishing times  $u.f$  for each vertex  $v$
  - 2 compute  $G^T$
  - 3 call DFS( $G^T$ , but in the main loop of DFS, CONSIDER THE VERTICES  
in order of decreasing  $u.f$  (as computed in line 1)
  - 4 output the vertices of each tree in the depth-first forest formed in line 3  
as a separate strongly connected components

## Trace of Execution of SCC algorithm

Graph  $G$  showing discovery times  $u.d$  and finishing times  $u.f$



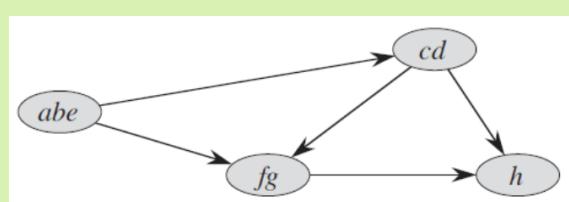
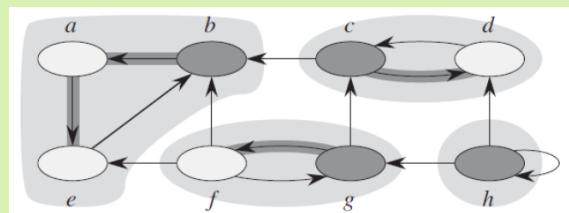
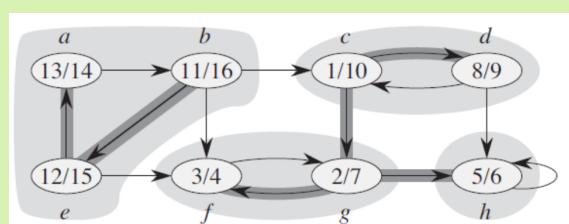
Transpose graph  $G^T$  with highlighted DFS forest; vertices are searched with decreasing  $u.f.$



### Strongly connected components

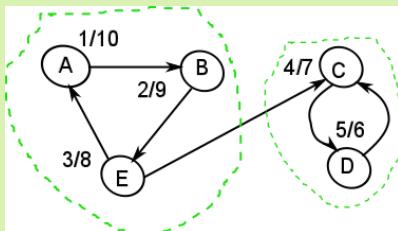


### Example in the textbook

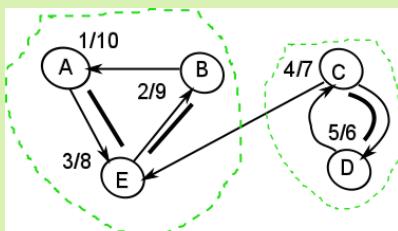


## Trace of Execution of SCC algorithm

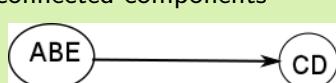
Graph  $G$  showing discovery times  $u.d$  and finishing times  $u.f$



Transpose graph  $G^T$  with highlighted DFS forest; vertices are searched with decreasing  $u.f.$



### Strongly connected components



## Example in the textbook

