

# CSE 100: Algorithm Design and Analysis

## Chapter 22: Elementary Graph Algorithms

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# Outline

- ▶ Graph representation
- ▶ Breadth First Search
- ▶ Depth First Search. Two key theorems: Parenthesis theorem and White path theorem.
- ▶ Three applications of DFS
  - ▶ How to determine if the graph has a cycle or not
  - ▶ Topological sort
  - ▶ Computing strongly connected components

# Graph Representation

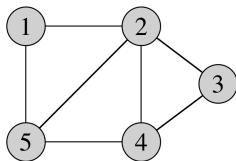
Notation. Given graph  $G = (V, E)$ , denote vertex set as  $G.V$  and edge set as  $G.E$ .

- ▶  $G$  may be either directed or undirected.
- ▶  $G$  can be represented by adjacency lists or adjacency matrix.
- ▶ Running time is often expressed in terms of  $|V|$  and  $|E|$ .

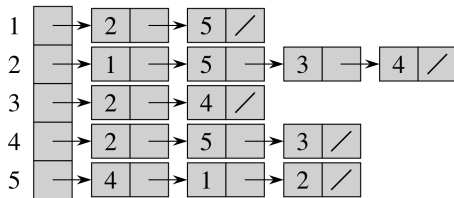
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## Adjacency Lists

- ▶ Adjacency list  $Adj[u]$  for each vertex  $u \in G.V$ .
- ▶  $Adj[u]$  has all vertices s.t.  $(u, v) \in G.E$ .
- ▶ In pseudocode.  $G.Adj[u]$ .



(a)



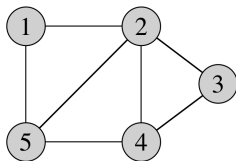
(b)

Space:

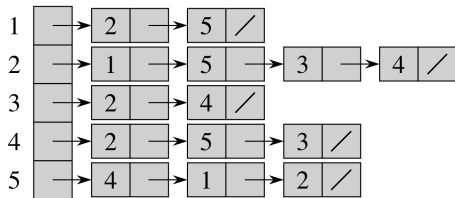
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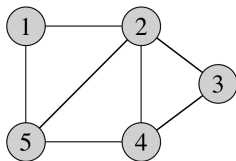
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Space:  $\Theta(|V| + |E|)$ .

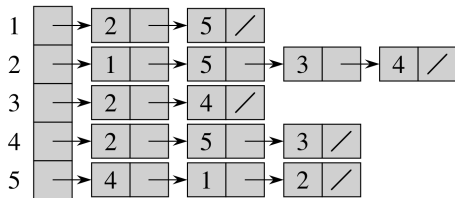
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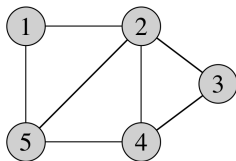
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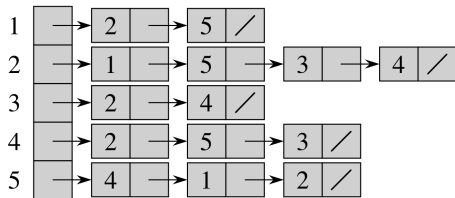
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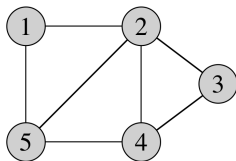
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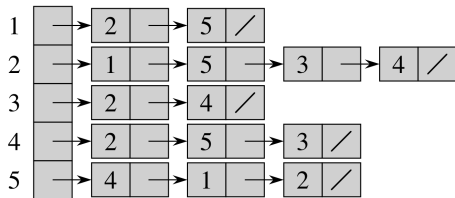
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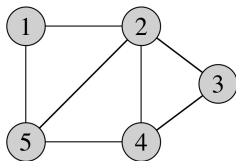
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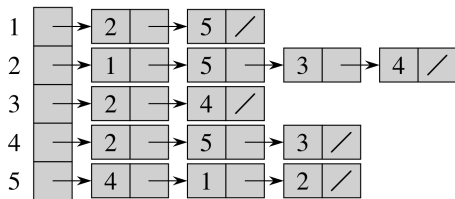
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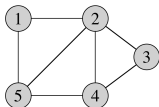
Time: to list all vertices adjacent to  $u$ :  $\Theta(deg(u))$ .

Time: to determine whether  $(u, v) \in E$  or not:  $O(deg(u))$ .

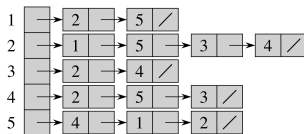
# Graph Representation

## Adjacency matrix

Represented by  $|V| \times |V|$  matrix,  $A = (a_{ij})$  where  $a_{ij} = 1$  if  $(i, j) \in E$  and 0 otherwise.



(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

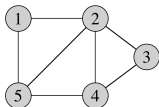
(c)

Space:

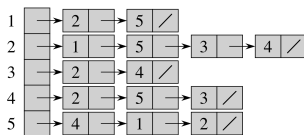
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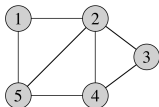
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Space:  $\Theta(|V|^2)$ .

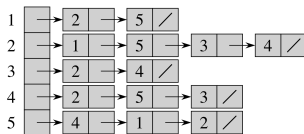
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(c)

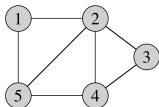
Space:  $\Theta(|V|^2)$ .

Time: to list all vertices adjacent to  $u$ :  $\Theta(|V|)$ .

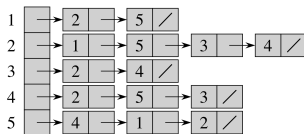
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(c)

Space:  $\Theta(|V|^2)$ .

Time: to list all vertices adjacent to  $u$ :  $\Theta(|V|)$ .

Time: to determine whether  $(u, v) \in E$  or not:  $\Theta(1)$ .

# Graph Representation

Q: We say that a graph is dense if  $|E|$  is much larger than  $|V|$  and sparse otherwise. If the graph is sparse, would you use adjacency lists or adjacency matrix?

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Q: If a graph is 'almost' complete, would you use adjacency lists or adjacency matrix?

Q: If a graph  $G$  is undirected, what is  $\sum_{u \in G.V} |G.Adj[u]| (= \sum_{u \in G.V} deg(u))$ ?



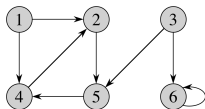
# Graph Terminology

If  $(u, v) \in G.E$  and  $G$  is undirected, we say that  $u$  is adjacent to  $v$ , or equivalently  $v$  is adjacent to  $u$ .

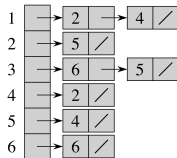
If  $(u, v) \in G.E$  and  $G$  is directed, we say that  $u$  is adjacent to  $v$ , or equivalently  $v$  is adjacent from  $u$ .

# Graph Representation

Example of directed graph



(a)



(b)

	1	2	3	4	5	6
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2	0	0	0	0	1	0
3	0	0	0	0	1	1
4	0	1	0	0	0	0
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(c)

# Graph Representation

Q: Say  $A$  is an adjacency matrix for an undirected graph. Then, it must be the case that  $a_{ij} = a_{ji}$  for all  $1 \leq i, j \leq |V|$ . True or False?

# Graph Representation

Q: Say  $A$  is an adjacency matrix for an undirected graph. Then, it must be the case that  $a_{ij} = a_{ji}$  for all  $1 \leq i, j \leq |V|$ . True or False? True, meaning that  $A = A^T$ .

# Graph Search Algorithms

Breadth-First-Search vs. Depth-First-Search

Both work for both undirected and directed graphs.

# Breadth-First-Search

Input: a graph  $G = (V, E)$  and a source  $s$ .

Output:

A tree consisting of vertices reachable from  $s$  encoding distance from  $s$ .

More precisely, the tree can be represented by:

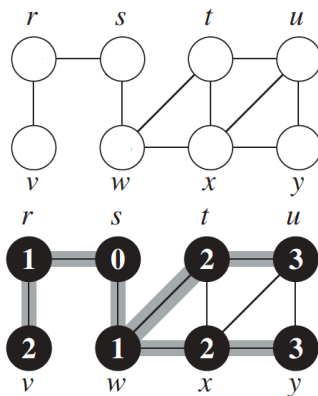
$v.d$ : distance (smallest # of edges) from  $s$  to  $v$ , for all  $v \in V$ .

$v.\pi$ :  $v$ 's predecessor. Edges  $\{(v.\pi, v) \mid v \neq s\}$  forms a tree.

The distance from  $s$  to  $v$  on the tree formed by  $\pi$  must be equal to  $v.d$ .

# Breadth-First-Search

example



\* The tree output may not be unique. But  $v.d$  remains the same.

# Breadth-First-Search

## Implementation

Intuitively, it's like sending a wave from  $s$ .

We simulate the 'parallel' wave propagation using FIFO queue  $Q$ .

$v \in Q$  if and only if wave has hit  $v$  but has not come out of  $v$  yet.



# Breadth-First-Search

## Implementation

```
BFS( $V, E, s$ )  
  for each  $u \in V - \{s\}$   
     $u.d = \infty$   
   $s.d = 0$   
   $Q = \emptyset$   
  ENQUEUE( $Q, s$ )  
  while  $Q \neq \emptyset$   
     $u = \text{DEQUEUE}(Q)$   
    for each  $v \in G.\text{Adj}[u]$   
      if  $v.d == \infty$   
         $v.d = u.d + 1$   
        ENQUEUE( $Q, v$ )
```

# Breadth-First-Search

Question: For every vertex  $v$ ,  $v.d$  changes at most once during the execution of BFS. Correct?

# Breadth-First-Search

## Implementation

```
BFS( $V, E, s$ )  
  for each  $u \in V - \{s\}$   
     $u.d = \infty$   
   $s.d = 0$   
   $Q = \emptyset$   
  ENQUEUE( $Q, s$ )  
  while  $Q \neq \emptyset$   
     $u = \text{DEQUEUE}(Q)$   
    for each  $v \in G.\text{Adj}[u]$   
      if  $v.d == \infty$   
         $v.d = u.d + 1$   
        ENQUEUE( $Q, v$ )
```

Change the code so that it computes  $v.\pi$ .

# Breadth-First-Search

## Running Time

$O(E + V)$ . Each vertex is enqueued and dequeued exactly once.  
Edge  $(u, v)$  is explored once when  $u$  is dequeued before  $v$ .

# Depth-First-Search

DFS picks an arbitrary *undiscovered* vertex as a *starting* vertex if there is any, and repeat the following:

- ▶ explores edges out of the most recently discovered vertex  $v$  that still has unexplored edges leaving it.
- ▶ backtracks to explore edges leaving the vertex from which  $v$  was discovered once all of  $v$ 's edges have been explored.

When  $v$  is discovered from  $u$  (when exploring  $(u, v)$ ), edge  $(u, v)$  becomes a tree edge.

In the end, we may have one or more *depth-first* trees. That is, a depth-first forest.

# Depth-First-Search

Input:

- ▶  $G = (V, E)$ , either directed or undirected.
- ▶ No source vertex is given.

Output:

- ▶  $\pi$  to record predecessors (to encode the resulting DFF) .
  - ▶ If  $v.\pi \neq NIL$ , then  $(v.\pi, v)$  is an edge of the DFF.
- ▶ two timestamps on each vertex  $v$ :
  - ▶  $v.d$  = discovery time
  - ▶  $v.f$  = finishing time

# Depth-First-Search

We use colors to indicate the status of each vertex.

- ▶ Initially,  $v$  is white.
- ▶ When  $v$  is discovered,  $v$  becomes gray.
- ▶ When  $v$  is finished, i.e. all edges out of  $v$  were explored and the token is moved up to  $v$ 's parent,  $v$  becomes black.

Time stamps.

- ▶ All timestamps are distinct ( $1$  to  $2|V|$ ).
- ▶  $v.d$  and  $v.f$  are recorded when they are discovered and finished, respectively.

# Depth-First-Search

## 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   $u.color = \text{GRAY}$ 
4  for each  $v \in G.Adj[u]$ 
5      if  $v.color == \text{WHITE}$ 
6           $v.\pi = u$ 
7          DFS-VISIT( $G, v$ )
8   $u.color = \text{BLACK}$ 
9   $time = time + 1$ 
10  $u.f = time$ 
```



# Depth-First Search

Running time

# Depth-First Search

## Running time

$\Theta(V + E)$ .

\* DFS-VISIT is called on each vertex exactly once, when it is white—then, it immediately becomes grey.

# Topological sort

Directed acyclic graph (DAG): A directed graph with no cycles.  
Good for modeling processes and/or structures that have a **partial order**.

- ▶ Transitive.  $a > b$  and  $b > c \Rightarrow a > c$ .
- ▶ But not all comparisons of two nodes/elements are known.

# Topological sort

Input: DAG  $G = (V, E)$ .

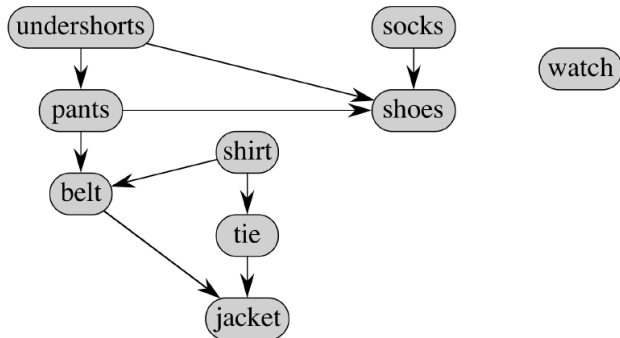
Output: **A** linear ordering of all vertices such that for any  $(u, v) \in E$ ,  $u$  appears before  $v$  in the ordering.

Or equivalently, find a total order that is consistent with a given partial order.

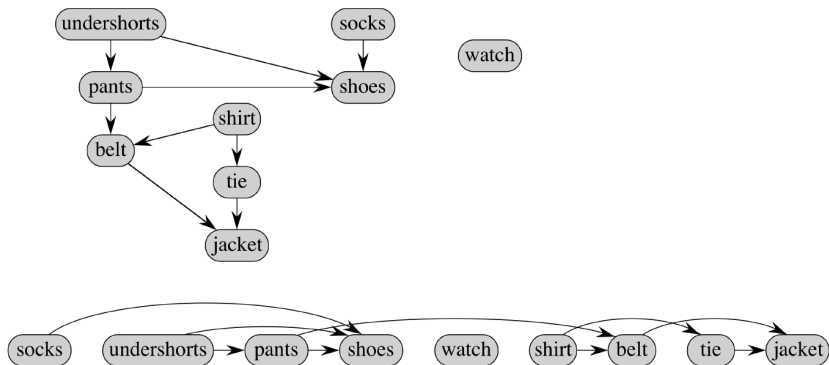
\* total order: for all two distinct vertices  $a, b$ , either  $a > b$  or  $b > a$ .

# Topological sort

Input:



# Topological sort

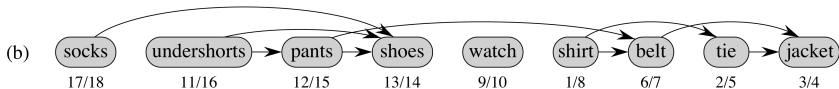
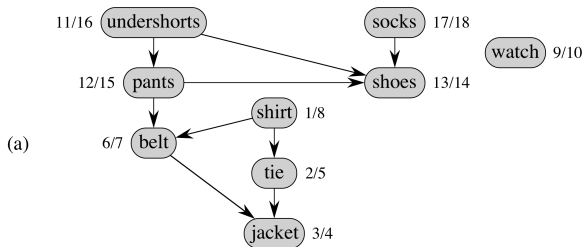


# Topological sort

## TOPOLOGICAL-SORT( $G$ )

- 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

# Topological sort





# Topological sort

How do we know if a given graph is a DAG or not?

## Lemma (22.11)

*A directed graph  $G$  is acyclic if and only if a depth-first search of  $G$  yields no back edge.*

## Proof.

( $\Rightarrow$ ): Back edge implies a cycle.

( $\Leftarrow$ ): Use the white-path theorem.



# Topological sort

## Theorem (22.12)

*Topological-Sort gives a topological sort of the input DAG.*

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WTS if  $(u, v) \in E$ , then  $v.f < u.f$ .

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- ▶  $v$  is white.

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- ▶  $v$  is gray. Impossible since otherwise,  $(u, v)$  is a back edge.
- ▶  $v$  is white.  $v$  becomes a descendant of  $u$ . Then, by parenthesis theorem  $v.f < u.f$ .



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- ▶  $v$  is white.  $v$  becomes a descendant of  $u$ . Then, by parenthesis theorem  $v.f < u.f$ .
- ▶  $v$  is black.  $v$  is already finished.  $u$  is still gray. So,  $v.f < u.f$ .

□

# Strongly connected components

Input: a directed graph  $G = (V, E)$ .

Output: all strongly connected components (SCCs) of  $G$ .

A SCC of  $G$  is a maximal set of vertices  $C \subseteq V$  such that for all  $u, v \in C$ , both  $u \rightsquigarrow v$  and  $v \rightsquigarrow u$ .

# Strongly connected components

$G^T = (V, E^T)$ : transpose of  $G$  where  $E^T = \{(u, v) : (v, u) \in E\}$ .  
Running time for creating  $G^T$ ?

## Strongly connected components

$G^T = (V, E^T)$ : transpose of  $G$  where  $E^T = \{(u, v) : (v, u) \in E\}$ .  
Running time for creating  $G^T$ ?  $\Theta(V + E)$  using adjacency lists.

# Strongly connected components

## Observation

$G$  and  $G^T$  have the same SCCs.

# Strongly connected components

## STRONGLY-CONNECTED-COMPONENTS( $G$ )

- 1 call DFS( $G$ ) to compute finishing times  $u.f$  for each vertex  $u$
- 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 component

# Strongly connected components

Component graph  $G^{SCC} = (V^{SCC}, E^{SCC})$  of  $G = (V, E)$ :  
 $v_i \in V^{SCC}$  iff  $C_i$  is a SCC of  $G$ .  
 $(v_i, v_j) \in E^{SCC}$  iff  $(x, y) \in E$  for some  $x \in C_i$  and  $y \in C_j$ .