

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 $|E|$.

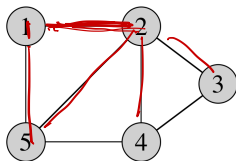
"
of vertices # of edges.

for brevity, V instead of $|V|$
 E "
 $|E|$

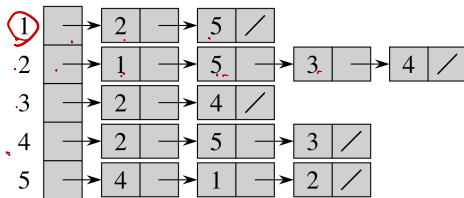
Graph Representation

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)



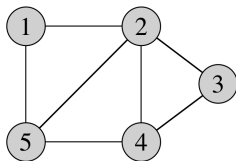
(b)

Space:

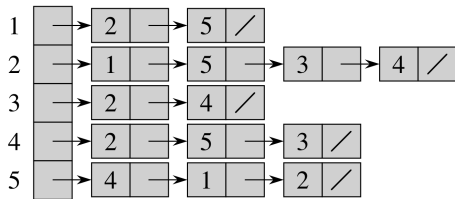
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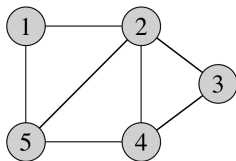
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Space: $\Theta(|V| + |E|)$.

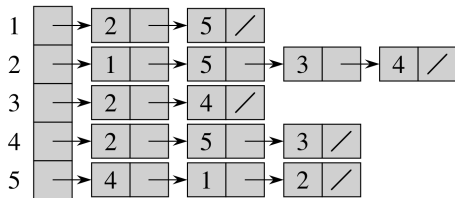
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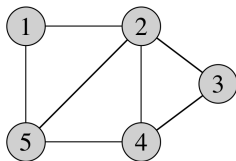
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Time: to list all vertices adjacent to u :

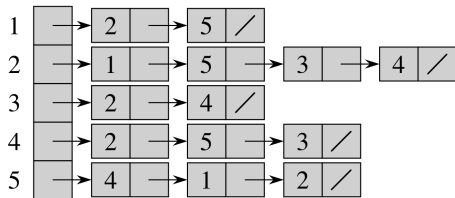
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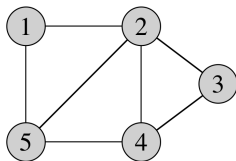
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Time: to list all vertices adjacent to u : $\Theta(deg(u))$.

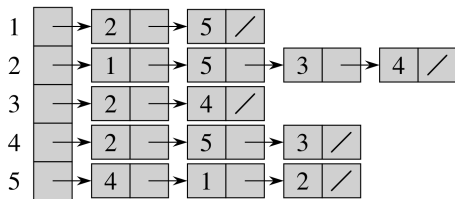
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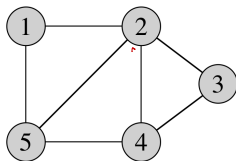
Time: to list all vertices adjacent to u : $\Theta(deg(u))$.

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

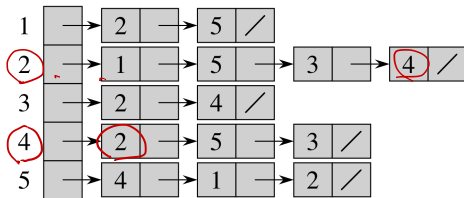
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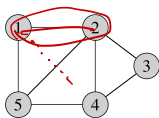
Time: to list all vertices adjacent to u : $\Theta(deg(u))$.

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

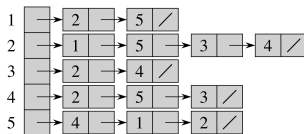
Graph Representation

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	1	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

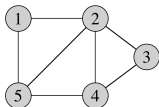
(c)

Space:

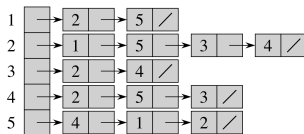
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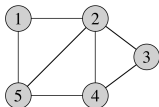
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Space: $\Theta(|V|^2)$.

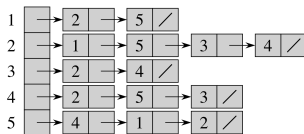
Graph Representation

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(b)

	1	2	3	4	5
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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)

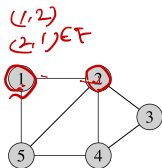
Space: $\Theta(|V|^2)$.

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

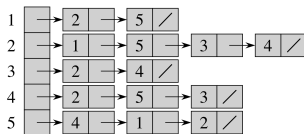
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Represented by $|V| \times |V|$ matrix, $A = (a_{ij})$ where $a_{ij} = 1$ if $(i, j) \in E$ and 0 otherwise.



(a)



(b)

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

(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?

Graph Representation

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

Graph Representation



u is v 's neighbor so (u, v) is counted in $\deg(v)$.

v is u 's neighbor so (v, u) is counted in $\deg(u)$.

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?

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

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

$$\deg(u) = |G.Adj[u]|$$

= # of u 's neighbors.

$$= 2|E|$$

Graph Terminology



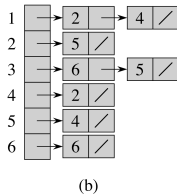
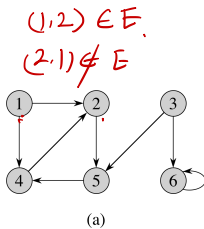
If $(u, v) \in G.E$ and G is undirected, we say that u is adjacent to v , or equivalently v is adjacent to u .

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



	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)

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?

$$A = A^T ?$$

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.

focus

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:

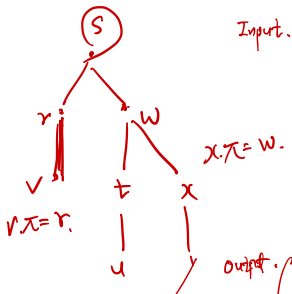
$v.d$: distance (smallest # of edges) from s to v , for all $v \in V$.

$v.\pi$: 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 π must be equal to $v.d$.

Breadth-First-Search

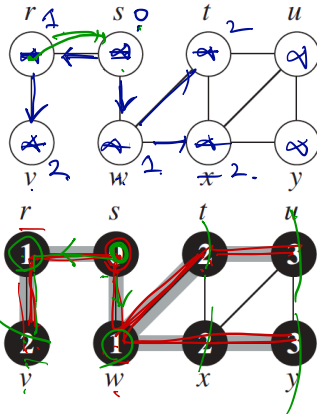
example



Input.

$x.x = w.$

Output.



def: tree: i) connected
ii) no cycle.

FIFO.

~~s~~ ~~v~~ ~~w~~ v t x



* 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 $u \in V - \{s\}$

$u.d = \infty$

$s.d = 0$

$Q = \emptyset$

\Rightarrow FIFO queue.

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 $v.\pi.$ \Rightarrow v 's parent in the output.

Breadth-First-Search

Running Time

$O(\underline{E} + \underline{V})$. 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:

- ▶ π 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

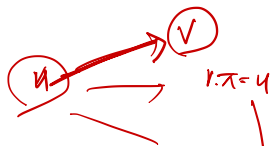
We use colors to indicate the status of each vertex.

- ▶ 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       $u.color = WHITE$ 
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  $v \in G.Adj[u]$ 
5      if  $v.color == WHITE$ 
6           $v.\pi = u$ 
7          DFS-VISIT( $G, v$ )
8   $u.color = BLACK$ 
9   $time = time + 1$ 
10  $u.f = time$ 
```


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 $G = (V, E)$.

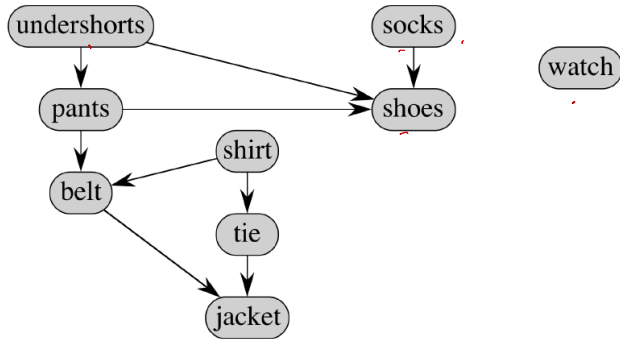
Output: **A** linear ordering of all vertices such that for any $(u, v) \in E$, u appears before v in the ordering.

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

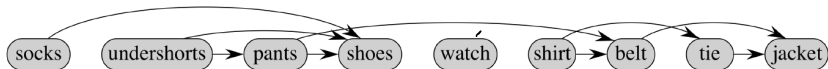
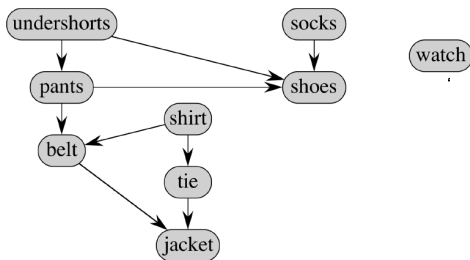
* total order: for all two distinct vertices a, b , either $a > b$ or $b > a$.

Topological sort

Input:



Topological sort



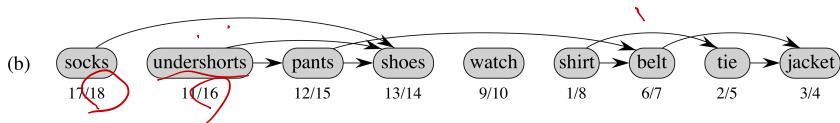
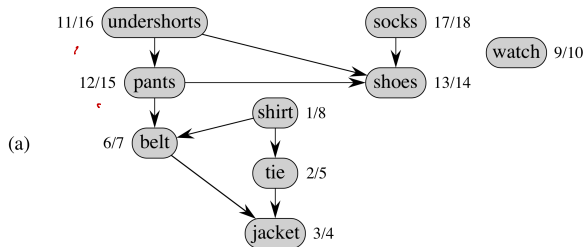
Topological sort

TOPOLOGICAL-SORT(G)

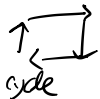
- 1 call DFS(G) to compute finishing times $v.f$ for each vertex v
- 2 as each vertex is finished, insert it onto the front of a linked list
- 3 **return** the linked list of vertices

Order them in decreasing order of their finish time.

Topological sort



Topological sort



$$V_0 \rightarrow V_1 \rightarrow V_2 \rightarrow \dots \rightarrow V_n$$

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

Lemma (22.11)

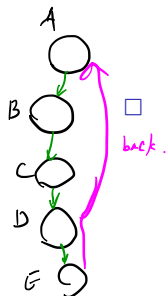
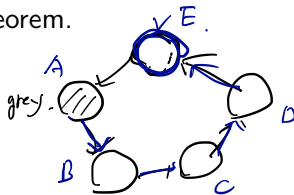
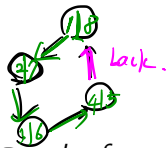
A directed graph G is acyclic if and only if a depth-first search of G yields no back edge.

Proof.

(\Rightarrow): Back edge implies a cycle.

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

cycle \Rightarrow back edge.



Topological sort

Theorem (22.12)

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



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Proof.

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

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WTS if $(u, v) \in E$, then $v.f < u.f$. Cases to consider based on v 's color when exploring (u, v) .

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- v is gray.

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- ▶ v is gray. Impossible since otherwise, (u, v) is a back edge.

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- ▶ v is gray. Impossible since otherwise, (u, v) is a back edge.
- ▶ 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 black.

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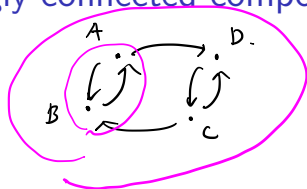
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WTS if $(u, v) \in E$, then $v.f < u.f$. Cases to consider based on v 's color when exploring (u, v) .

- ▶ 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 $v.f < u.f$.
- ▶ v is black. v is already finished. u is still gray. So, $v.f < u.f$.



Strongly connected components



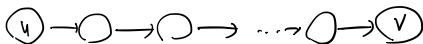
$A \rightsquigarrow B, B \rightsquigarrow A$.
But $\{A, D\}$ is not a SCC
because it is not maximal.

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 $\underline{u \rightsquigarrow v}$ and $\underline{v \rightsquigarrow u}$.

(there is a path from u to v).



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 $y \in C_j$.