

Elementary Graph Theory

Tao Hou

- Graphs: definitions (Review+New)
- Representations (Review)
- Topological sort
- DFS (mostly *New*)

- A *graph*

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- E is the set of *edges*

- ▶ an edge $e = (u, v)$ from E is a pair of vertices where $u \in V$ and $v \in V$

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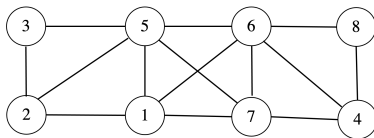
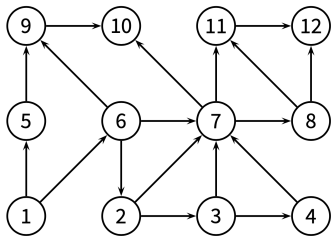
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- In this course, unless otherwise noted, we assume graphs are **simple graphs**, i.e., no *self loops* or *parallel edges*.



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- A **cycle** is a path starting and ending at the same vertex
 - ▶ A cycle is called **simple** if there are no duplicate vertices on the cycle other than the starting and ending vertices

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Connectivity (Review)

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The version of ‘connected components’ for **directed** graphs are called **strongly connected components**, which we do not touch

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 - ▶ Most ‘tree data structures’ are indeed rooted trees, e.g., binary trees, heaps, B-tree
- More on rooted tree:
 - ▶ Each vertex has exactly one in-coming edge from its **parent** except the root, which has no in-coming edges.
 - ▶ If there is a path from u to v , then u is an **ancestor** of v and v is a **descendant** of u

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 - ▶ So after adding the $n - 1$ edges, there is only one connected component.
 - ▶ This means that when we add the n -th edge, it must create a cycle.

Some facts about trees

Fact

A connected, undirected graph with n vertices and $n - 1$ edges is a tree

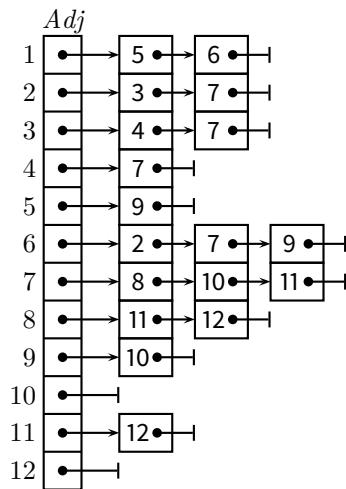
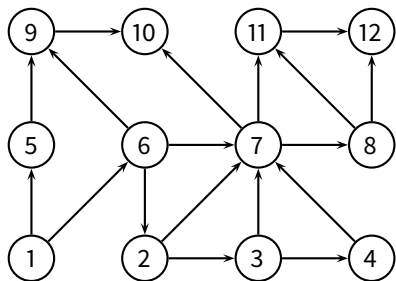
Graph Representation (Review)

- How do we represent a graph $G = (V, E)$ in a computer?

Adjacency-list representation:

- $V = \{1, 2, \dots, |V|\}$
- G consists of an array Adj
- A vertex $u \in V$ is represented by an element in the array Adj
- $Adj[u]$ is the **adjacency list** of vertex u
 - ▶ the list of the vertices that are adjacent to u
 - ▶ i.e., the list of all v such that $(u, v) \in E$
 - ▶ Notice the difference between *directed* and *undirected* graphs

Example



Using the Adjacency List (Review)

■ Iteration through E ?

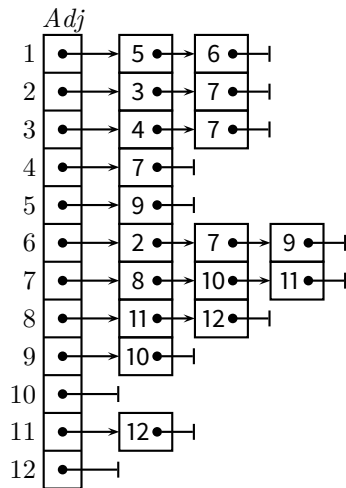
- ▶ okay (not optimal)

$$O(|V| + |E|)$$

■ Checking $(u, v) \in E$?

- ▶ looks bad, but it depends

$$O(|V|)$$



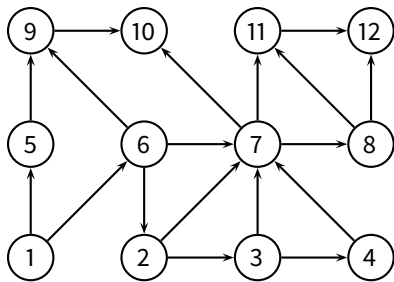
Adjacency-Matrix Representation (Review)

Adjacency-matrix representation:

- $V = \{1, 2, \dots, |V|\}$
- G consists of a $|V| \times |V|$ matrix A
- $A = (a_{ij})$ such that

$$a_{ij} = \begin{cases} 1 & \text{if } (i, j) \in E \\ 0 & \text{otherwise} \end{cases}$$

Example

[illegible]

Using the Adjacency Matrix (Review)

- Iteration through E ?
 - ▶ possibly very bad
- Checking $(u, v) \in E$?
 - ▶ optimal

$O(|V|^2)$

$O(1)$

[illegible]

- Adjacency-list representation

$$O(|V| + |E|)$$

optimal

- Adjacency-matrix representation

$$O(|V|^2)$$

possibly very bad

Choosing a Graph Representation (Review)

■ Adjacency-list representation

- ▶ generally good, especially for its optimal space complexity
- ▶ bad for **dense** graphs and algorithms that require random access to edges
- ▶ preferable for **sparse** graphs or graphs with **low degree**

■ Adjacency-matrix representation

- ▶ suffers from a bad space complexity
- ▶ good for algorithms that require random access to edges
- ▶ preferable for **dense** graphs

■ Sparse vs. dense graph

- ▶ **Sparse** graph: $|E| = O(|V|)$
- ▶ **Dense** graph: $|E| = \Theta(|V|^2)$

■ **Problem:** (topological sort)

Given a *directed acyclic graph* (DAG)

- ▶ find an ordering of vertices such that you only end up with *forward edges*
- ▶ in another word, if there is an edge (u, v) , then u appears before v in the ordering (that's also the reason why we can do this *only* on DAG instead of general graphs)

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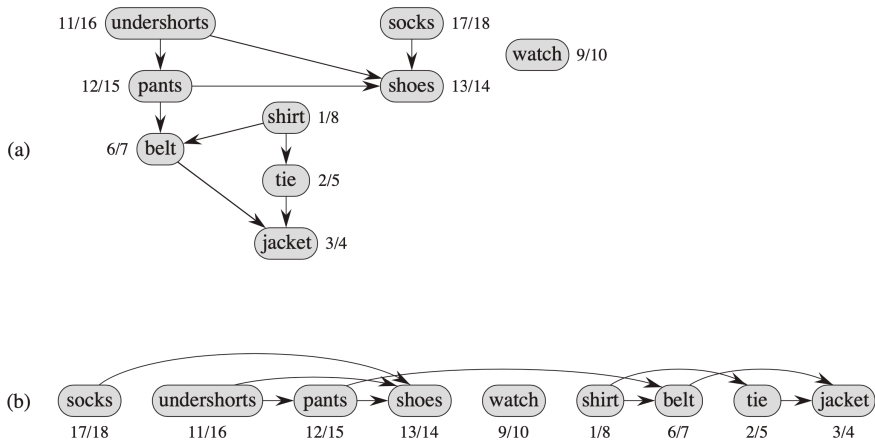
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■ **Example:** dependencies in software packages

- ▶ find an installation order for a set of software packages
- ▶ such that every package is installed only after all the packages it depends on



(Example from CLRS)

Topological Sort Algorithm

TOPOLOGICAL-SORT(G)

- 1 **while** $\exists v \in V$ s.t. $in-deg(v) = 0$
- 2 output v
- 3 remove v and all its out-going edges from G

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Argument of correctness:

- We remove an edge only when its starting vertex has been output in the order
- Thus, when a vertex v has in-degree 0, this means that all vertices pointing to v (if any) have been output, so that we can also safely output v

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Question:

- Why should there always be a vertex with 0 in-degree?

Topological Sort: Alternative Algorithm

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We will see why this algorithm works later on.

Some comments:

- The first algorithm is mainly of theoretical value (helps you to understand the whole procedure)
- In practice, you should utilize DFS to compute topological sorting for DAGs because it's much simpler (you don't need to bother to delete the edges)
- So topological sort can be done in $O(|V| + |E|)$ time

- *Input: $G = (V, E)$, which can be *directed* or *undirected**
- Explores the graph starting from s , touching all vertices that are reachable from s

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DFS(G)

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1  for each vertex  $u \in V(G)$ 
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4   $time = 0$  // “global” variable
5  for each vertex  $u \in V(G)$ 
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```
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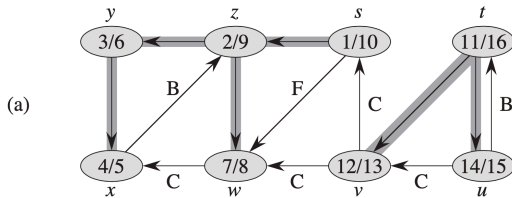
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- A first very silly question: Can DFS ever end?



(Example from CLRS)

Complexity of DFS

- The loop in **DFS-VISIT**(u) (lines 4–7) executes for $O(\text{out-deg}(u))$ times
- We call **DFS-VISIT**(u) *once* for each vertex u
 - ▶ either in **DFS**, or recursively in **DFS-VISIT**
 - ▶ because we call it only if $\text{color}[u] = \text{WHITE}$, but then we immediately set $\text{color}[u] = \text{GREY}$
- So, the overall complexity is $\Theta(|V| + |E|)$

Parenthesis Theorem

In a DFS on a (directed or undirected) graph G , for any two vertices u and v , exactly one of the following two holds:

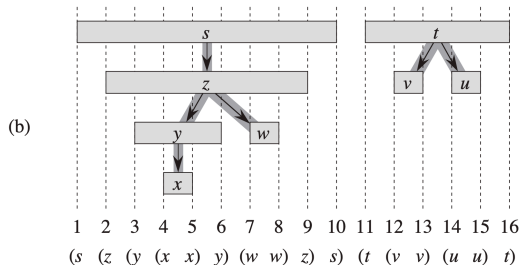
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- Observe: *the grey vertices in DFS always form a linear chain of descendants (in the DFS tree) corresponding to the stack of active DFS-VISIT invocations*
- This means that v is a descendant of u in the DFS forest
- Also, the visiting of u cannot finish before we finish visiting u (this is how recursive calls work), so $f[v] < f[u]$ (aka. $d[u] < d[v] < f[v] < f[u]$)

- Now consider $d[v] > f[u]$
- Obviously, $d[u] < f[u] < d[v] < f[v]$, so the two intervals are disjoint

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In a DFS forest of a (directed or undirected) graph G , a vertex v is a descendant of a vertex u if and only if at time $d[u]$, there is a path from u to v on G consisting of *only* white vertices

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proof:

- “ \Rightarrow ”: let w be any descendant of u in the DFS tree
- By the previous Parenthesis Theorem, we have that $d[u] < d[w]$, so when u is discovered, w is still white
- Notice that on the path from u to v in the DFS tree, all vertices are descendants of v , so all of them are white at time $d[u]$

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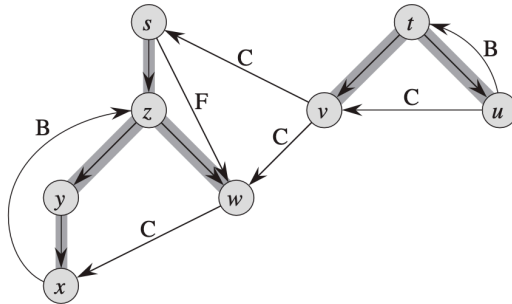
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- But if this is true, then x must be a descendant of w and in turn a descendant of u (a contradiction)

Four Types of Edges in DFS on Directed Graphs

- **Tree edge:** Edges on the DFS forest
- **Back edge:** Connecting a vertex to its *ancestor* in the DFS forest
- **Forward edge:** Non-tree edges connecting a vertex to its *descendant* in the DFS forest
- **Cross edge:** all other edges



(Example from CLRS)

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- Therefore, (u, v) is a back edge

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Proof:

- Consider a path P connecting u, v in G
- Let x be the first vertex on P visited by DFS. Apparently, we can reach u and v from x
- By the description of DFS, the DFS visit on x will touch all vertices that are reachable from x . So we will reach u and v from visiting x .
- Therefore, u, v, x are all in the same DFS tree.