

Elementary Graph Theory

Tao Hou

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

- A *graph*

$$G = (V, E)$$

- V is the set of *vertices* (also called *nodes*)

- E is the set of *edges*

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

- A **graph**

$$G = (V, E)$$

- V is the set of **vertices** (also called **nodes**)

- E is the set of **edges**

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

- *directed* graph: an edge (u, v) is from u to v and has a direction

- *undirected* graph: no directions for the edges (so $(u, v) = (v, u)$)

- A **graph**

$$G = (V, E)$$

- V is the set of **vertices** (also called **nodes**)

- E is the set of **edges**

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

- *directed* graph: an edge (u, v) is from u to v and has a direction

- *undirected* graph: no directions for the edges (so $(u, v) = (v, u)$)

- Sometimes given a graph G , we also let $V(G)$ denote the vertex set and $E(G)$ denote the edge set

- A **graph**

$$G = (V, E)$$

- V is the set of **vertices** (also called **nodes**)

- E is the set of **edges**

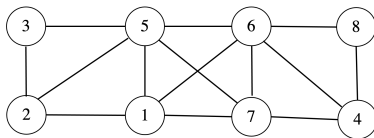
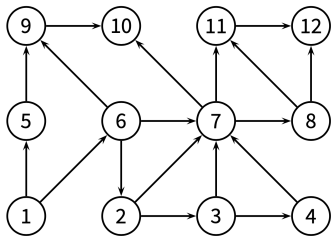
 - ▶ an edge $e = (u, v)$ from E is a pair of vertices where $u \in V$ and $v \in V$

- *directed* graph: an edge (u, v) is from u to v and has a direction

- *undirected* graph: no directions for the edges (so $(u, v) = (v, u)$)

- Sometimes given a graph G , we also let $V(G)$ denote the vertex set and $E(G)$ denote the edge set

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



Given a graph $G = (V, E)$,

- We call $G' = (V', E')$ a **subgraph** of G if $V' \subseteq V$ and $E' \subseteq E$.

Given a graph $G = (V, E)$,

- We call $G' = (V', E')$ a **subgraph** of G if $V' \subseteq V$ and $E' \subseteq E$.
- For a subset of vertices $V' \subseteq V$, the subgraph $G' = (V', E')$ **induced** by V' has an edge set consisting of all edges of G whose vertices are in V' , i.e.,

$$E' = \{(u, v) \in E \mid u \in V' \text{ and } v \in V'\}$$

Given a graph $G = (V, E)$,

- We call $G' = (V', E')$ a **subgraph** of G if $V' \subseteq V$ and $E' \subseteq E$.
- For a subset of vertices $V' \subseteq V$, the subgraph $G' = (V', E')$ **induced** by V' has an edge set consisting of all edges of G whose vertices are in V' , i.e.,

$$E' = \{(u, v) \in E \mid u \in V' \text{ and } v \in V'\}$$

- A **path** in G is a sequence of vertices v_1, v_2, \dots, v_k s.t. each (v_i, v_{i+1}) forms an edge in G
 - ▶ This applies to both directed and undirected graphs
 - ▶ Sometime a path also refers to the **sequence of edges** on the path

Given a graph $G = (V, E)$,

- We call $G' = (V', E')$ a **subgraph** of G if $V' \subseteq V$ and $E' \subseteq E$.
- For a subset of vertices $V' \subseteq V$, the subgraph $G' = (V', E')$ **induced** by V' has an edge set consisting of all edges of G whose vertices are in V' , i.e.,

$$E' = \{(u, v) \in E \mid u \in V' \text{ and } v \in V'\}$$

- A **path** in G is a sequence of vertices v_1, v_2, \dots, v_k s.t. each (v_i, v_{i+1}) forms an edge in G
 - ▶ This applies to both directed and undirected graphs
 - ▶ Sometime a path also refers to the **sequence of edges** on the path
 - ▶ A path is called **simple** if there are no duplicate vertices on the path

Given a graph $G = (V, E)$,

- We call $G' = (V', E')$ a **subgraph** of G if $V' \subseteq V$ and $E' \subseteq E$.
- For a subset of vertices $V' \subseteq V$, the subgraph $G' = (V', E')$ **induced** by V' has an edge set consisting of all edges of G whose vertices are in V' , i.e.,

$$E' = \{(u, v) \in E \mid u \in V' \text{ and } v \in V'\}$$

- A **path** in G is a sequence of vertices v_1, v_2, \dots, v_k s.t. each (v_i, v_{i+1}) forms an edge in G
 - ▶ This applies to both directed and undirected graphs
 - ▶ Sometime a path also refers to the **sequence of edges** on the path
 - ▶ A path is called **simple** if there are no duplicate vertices on the path
- 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

For a *directed* graph $G = (V, E)$,

- The ***out-degree*** of a vertex $x \in V$ is the number of edges starting with x , i.e.,

$$\text{out-deg}(x) = |\{(u, v) \in E \mid u = x\}|$$

- The ***in-degree*** of a vertex $x \in V$ is the number of edges ending with x , i.e.,

$$\text{in-deg}(x) = |\{(u, v) \in E \mid v = x\}|$$

For a *directed* graph $G = (V, E)$,

- The **out-degree** of a vertex $x \in V$ is the number of edges starting with x , i.e.,

$$\text{out-deg}(x) = |\{(u, v) \in E \mid u = x\}|$$

- The **in-degree** of a vertex $x \in V$ is the number of edges ending with x , i.e.,

$$\text{in-deg}(x) = |\{(u, v) \in E \mid v = x\}|$$

- We have

$$\sum_{v \in V} \text{out-deg}(v) = \sum_{v \in V} \text{in-deg}(v) = |E|$$

For a *directed* graph $G = (V, E)$,

- The **out-degree** of a vertex $x \in V$ is the number of edges starting with x , i.e.,

$$\text{out-deg}(x) = |\{(u, v) \in E \mid u = x\}|$$

- The **in-degree** of a vertex $x \in V$ is the number of edges ending with x , i.e.,

$$\text{in-deg}(x) = |\{(u, v) \in E \mid v = x\}|$$

- We have

$$\sum_{v \in V} \text{out-deg}(v) = \sum_{v \in V} \text{in-deg}(v) = |E|$$

For an *undirected* graph $G = (V, E)$,

- The **degree** of a vertex $x \in V$ is the number of edges having x as a vertex, i.e.,

$$\text{deg}(x) = |\{(u, v) \in E \mid u = x \text{ or } v = x\}|$$

For a *directed* graph $G = (V, E)$,

- The **out-degree** of a vertex $x \in V$ is the number of edges starting with x , i.e.,

$$\text{out-deg}(x) = |\{(u, v) \in E \mid u = x\}|$$

- The **in-degree** of a vertex $x \in V$ is the number of edges ending with x , i.e.,

$$\text{in-deg}(x) = |\{(u, v) \in E \mid v = x\}|$$

- We have

$$\sum_{v \in V} \text{out-deg}(v) = \sum_{v \in V} \text{in-deg}(v) = |E|$$

For an *undirected* graph $G = (V, E)$,

- The **degree** of a vertex $x \in V$ is the number of edges having x as a vertex, i.e.,

$$\text{deg}(x) = |\{(u, v) \in E \mid u = x \text{ or } v = x\}|$$

- We have

$$\sum_{v \in V} \text{deg}(v) = 2|E|$$

Connectivity (Review)

Given an *undirected* graph $G = (V, E)$,

- Two vertices u, v are *connected* in G if there is a path from u to v in G

Connectivity (Review)

Given an **undirected** graph $G = (V, E)$,

- Two vertices u, v are **connected** in G if there is a path from u to v in G
- A **connected component** $U \subseteq V$ of G is a **maximal** set of vertices where each pair are connected by a path in G (**maximal** means you cannot add more vertices to U anymore)

Connectivity (Review)

Given an **undirected** graph $G = (V, E)$,

- Two vertices u, v are **connected** in G if there is a path from u to v in G
- A **connected component** $U \subseteq V$ of G is a **maximal** set of vertices where each pair are connected by a path in G (**maximal** means you cannot add more vertices to U anymore)
 - ▶ Sometimes, a connected component also refers to the *subgraph induced by U* .

Connectivity (Review)

Given an **undirected** graph $G = (V, E)$,

- Two vertices u, v are **connected** in G if there is a path from u to v in G
- A **connected component** $U \subseteq V$ of G is a **maximal** set of vertices where each pair are connected by a path in G (**maximal** means you cannot add more vertices to U anymore)
 - ▶ Sometimes, a connected component also refers to the *subgraph induced by U* .
- G is called **connected** if it contains a single connected component (i.e., every two vertices are connected by a path)

Connectivity (Review)

Given an **undirected** graph $G = (V, E)$,

- Two vertices u, v are **connected** in G if there is a path from u to v in G
- A **connected component** $U \subseteq V$ of G is a **maximal** set of vertices where each pair are connected by a path in G (**maximal** means you cannot add more vertices to U anymore)
 - ▶ Sometimes, a connected component also refers to the *subgraph induced by U* .
- G is called **connected** if it contains a single connected component (i.e., every two vertices are connected by a path)

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

- A **tree** is an **acyclic**, **undirected**, **connected** graph.
 - ▶ Here 'acyclic' means having no **edge-disjoint** cycles, i.e., there is not a cycle containing distinct edges

- A **tree** is an **acyclic**, **undirected**, **connected** graph.
 - ▶ Here 'acyclic' means having no **edge-disjoint** cycles, i.e., there is not a cycle containing distinct edges
- A **forest** is an **acyclic**, **undirected** graph

- A **tree** is an **acyclic, undirected, connected** graph.
 - ▶ Here 'acyclic' means having no **edge-disjoint** cycles, i.e., there is not a cycle containing distinct edges
- A **forest** is an **acyclic, undirected** graph
 - ▶ Each connected component is a tree (so a forest nothing but a disjoint-union of trees)

- A **tree** is an **acyclic, undirected, connected** graph.
 - ▶ Here 'acyclic' means having no **edge-disjoint** cycles, i.e., there is not a cycle containing distinct edges
- A **forest** is an **acyclic, undirected** graph
 - ▶ Each connected component is a tree (so a forest nothing but a disjoint-union of trees)
- A **rooted** tree is a **directed** graph derived from a tree (which is undirected) by choosing a **root** vertex first, and then directing edges s.t. each edge points from a **parent** to its **child**.

- A **tree** is an **acyclic**, **undirected**, **connected** graph.
 - ▶ Here 'acyclic' means having no **edge-disjoint** cycles, i.e., there is not a cycle containing distinct edges
- A **forest** is an **acyclic**, **undirected** graph
 - ▶ Each connected component is a tree (so a forest nothing but a disjoint-union of trees)
- A **rooted** tree is a **directed** graph derived from a tree (which is undirected) by choosing a **root** vertex first, and then directing edges s.t. each edge points from a **parent** to its **child**.
 - ▶ One way to understand the 'directing' process: perform a DFS on the tree starting from the root. The directed edges always point from a vertex visited *earlier* to a vertex visited *later*
 - ▶ Specifically, the root vertex is visited the earliest, so edges are always pointing from the root to other vertices

- A **tree** is an **acyclic**, **undirected**, **connected** graph.
 - ▶ Here 'acyclic' means having no **edge-disjoint** cycles, i.e., there is not a cycle containing distinct edges
- A **forest** is an **acyclic**, **undirected** graph
 - ▶ Each connected component is a tree (so a forest nothing but a disjoint-union of trees)
- A **rooted** tree is a **directed** graph derived from a tree (which is undirected) by choosing a **root** vertex first, and then directing edges s.t. each edge points from a **parent** to its **child**.
 - ▶ One way to understand the 'directing' process: perform a DFS on the tree starting from the root. The directed edges always point from a vertex visited *earlier* to a vertex visited *later*
 - ▶ Specifically, the root vertex is visited the earliest, so edges are always pointing from the root to other vertices
 - ▶ Most 'tree data structures' are indeed rooted trees, e.g., binary trees, heaps, B-tree

- A **tree** is an **acyclic**, **undirected**, **connected** graph.
 - ▶ Here ‘acyclic’ means having no **edge-disjoint** cycles, i.e., there is not a cycle containing distinct edges
- A **forest** is an **acyclic**, **undirected** graph
 - ▶ Each connected component is a tree (so a forest nothing but a disjoint-union of trees)
- A **rooted** tree is a **directed** graph derived from a tree (which is undirected) by choosing a **root** vertex first, and then directing edges s.t. each edge points from a **parent** to its **child**.
 - ▶ One way to understand the ‘directing’ process: perform a DFS on the tree starting from the root. The directed edges always point from a vertex visited *earlier* to a vertex visited *later*
 - ▶ Specifically, the root vertex is visited the earliest, so edges are always pointing from the root to other vertices
 - ▶ 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

Observation

A tree with n vertices has $n - 1$ edges.

Observation

A tree with n vertices has $n - 1$ edges.

Proof:

- Consider that initially we only have the n vertices of the tree, and we add each of the $n - 1$ edges one by one.

Observation

A tree with n vertices has $n - 1$ edges.

Proof:

- Consider that initially we only have the n vertices of the tree, and we add each of the $n - 1$ edges one by one.
- When we add each edge, exactly one of the following two things can happen:
 - (1) Two connected components in the graph merge into one connected component (number of connected components decrease by 1);
 - (2) A cycle is created (number of connected components stays the same).

Observation

A tree with n vertices has $n - 1$ edges.

Proof:

- Consider that initially we only have the n vertices of the tree, and we add each of the $n - 1$ edges one by one.
- When we add each edge, exactly one of the following two things can happen:
 - (1) Two connected components in the graph merge into one connected component (number of connected components decrease by 1);
 - (2) A cycle is created (number of connected components stays the same).
- Number of edges cannot be $< n - 1$:

Observation

A tree with n vertices has $n - 1$ edges.

Proof:

- Consider that initially we only have the n vertices of the tree, and we add each of the $n - 1$ edges one by one.
- When we add each edge, exactly one of the following two things can happen:
 - (1) Two connected components in the graph merge into one connected component (number of connected components decrease by 1);
 - (2) A cycle is created (number of connected components stays the same).
- Number of edges cannot be $< n - 1$:
 - ▶ If the number of edges is $< n - 1$, then the number of connected components cannot decrease to one (adding each edge decreases the number of connected component by at most one)

Observation

A tree with n vertices has $n - 1$ edges.

Proof:

- Consider that initially we only have the n vertices of the tree, and we add each of the $n - 1$ edges one by one.
- When we add each edge, exactly one of the following two things can happen:
 - (1) Two connected components in the graph merge into one connected component (number of connected components decrease by 1);
 - (2) A cycle is created (number of connected components stays the same).
- Number of edges cannot be $< n - 1$:
 - ▶ If the number of edges is $< n - 1$, then the number of connected components cannot decrease to one (adding each edge decreases the number of connected component by at most one)
- Number of edges cannot be $> n - 1$:

Observation

A tree with n vertices has $n - 1$ edges.

Proof:

- Consider that initially we only have the n vertices of the tree, and we add each of the $n - 1$ edges one by one.
- When we add each edge, exactly one of the following two things can happen:
 - (1) Two connected components in the graph merge into one connected component (number of connected components decrease by 1);
 - (2) A cycle is created (number of connected components stays the same).
- Number of edges cannot be $< n - 1$:
 - ▶ If the number of edges is $< n - 1$, then the number of connected components cannot decrease to one (adding each edge decreases the number of connected component by at most one)
- Number of edges cannot be $> n - 1$:
 - ▶ If the number of edges is $> n - 1$, consider adding the first $n - 1$ edges.

Observation

A tree with n vertices has $n - 1$ edges.

Proof:

- Consider that initially we only have the n vertices of the tree, and we add each of the $n - 1$ edges one by one.
- When we add each edge, exactly one of the following two things can happen:
 - (1) Two connected components in the graph merge into one connected component (number of connected components decrease by 1);
 - (2) A cycle is created (number of connected components stays the same).
- Number of edges cannot be $< n - 1$:
 - ▶ If the number of edges is $< n - 1$, then the number of connected components cannot decrease to one (adding each edge decreases the number of connected component by at most one)
- Number of edges cannot be $> n - 1$:
 - ▶ If the number of edges is $> n - 1$, consider adding the first $n - 1$ edges.
 - ▶ Since the tree has no cycle, only situation (1) can happen.

Observation

A tree with n vertices has $n - 1$ edges.

Proof:

- Consider that initially we only have the n vertices of the tree, and we add each of the $n - 1$ edges one by one.
- When we add each edge, exactly one of the following two things can happen:
 - (1) Two connected components in the graph merge into one connected component (number of connected components decrease by 1);
 - (2) A cycle is created (number of connected components stays the same).
- Number of edges cannot be $< n - 1$:
 - ▶ If the number of edges is $< n - 1$, then the number of connected components cannot decrease to one (adding each edge decreases the number of connected component by at most one)
- Number of edges cannot be $> n - 1$:
 - ▶ If the number of edges is $> n - 1$, consider adding the first $n - 1$ edges.
 - ▶ Since the tree has no cycle, only situation (1) can happen.
 - ▶ So after adding the $n - 1$ edges, there is only one connected component.

Observation

A tree with n vertices has $n - 1$ edges.

Proof:

- Consider that initially we only have the n vertices of the tree, and we add each of the $n - 1$ edges one by one.
- When we add each edge, exactly one of the following two things can happen:
 - (1) Two connected components in the graph merge into one connected component (number of connected components decrease by 1);
 - (2) A cycle is created (number of connected components stays the same).
- Number of edges cannot be $< n - 1$:
 - ▶ If the number of edges is $< n - 1$, then the number of connected components cannot decrease to one (adding each edge decreases the number of connected component by at most one)
- Number of edges cannot be $> n - 1$:
 - ▶ If the number of edges is $> n - 1$, consider adding the first $n - 1$ edges.
 - ▶ Since the tree has no cycle, only situation (1) can happen.
 - ▶ 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

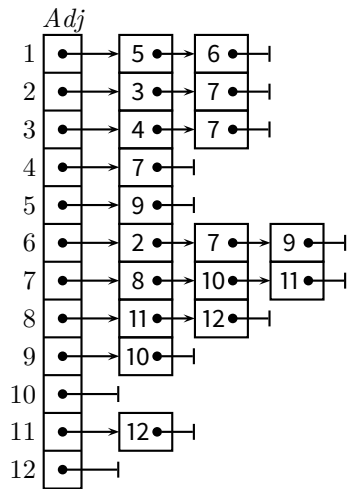
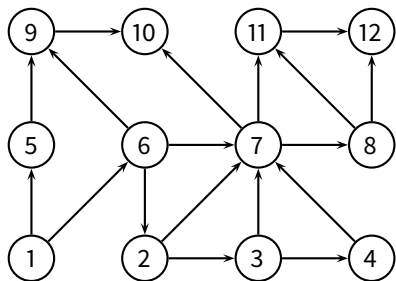
Graph Representation (Review)

- 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 ?

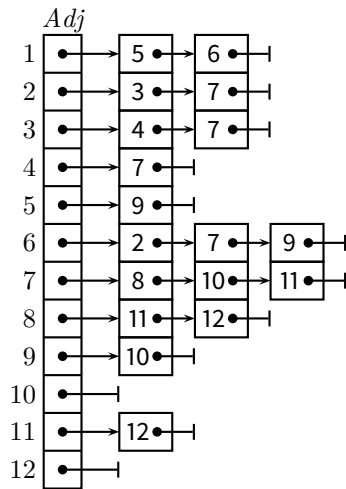
- ▶ okay (not optimal)

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

■ Checking $(u, v) \in E$?

- ▶ looks bad, but it depends

$$O(|V|)$$



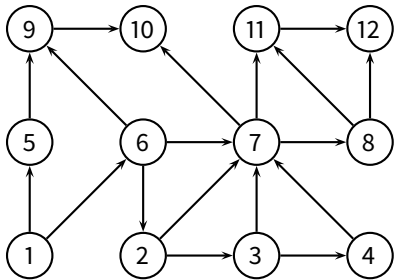
Adjacency-Matrix Representation (Review)

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)

■ **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)
- ▶ **Note:** The 'acyclic' here is for directed graphs and therefore means only 'no cycles' (we don't need to say 'no edge-disjoint cycles' here)

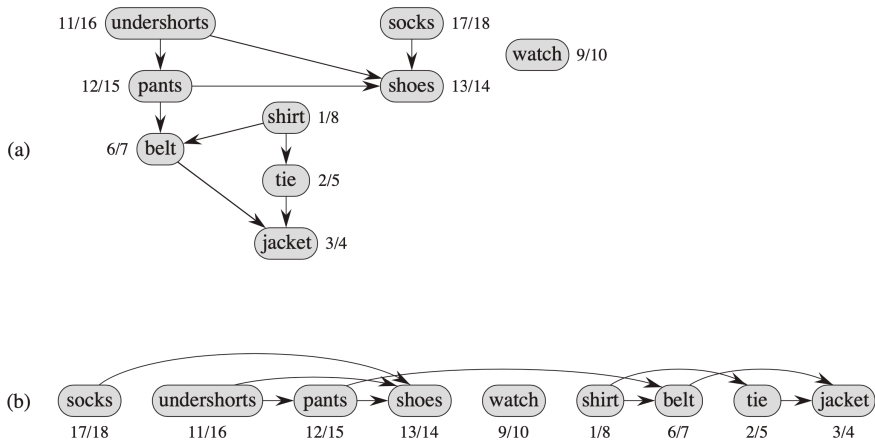
■ **Problem:** (topological sort)

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)
- ▶ **Note:** The 'acyclic' here is for directed graphs and therefore means only 'no cycles' (we don't need to say 'no edge-disjoint cycles' here)

■ **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

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  $G$ 
```

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

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  $G$ 
```

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

Question:

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

Topological Sort: Alternative Algorithm

TOPOLOGICAL-SORT(G)

- 1 **DFS**(G)
- 2 output V sorted in reverse order of $f[\cdot]$

Topological Sort: Alternative Algorithm

TOPOLOGICAL-SORT(G)

- 1 **DFS**(G)
- 2 output V sorted in reverse order of $f[\cdot]$

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

- *Input: $G = (V, E)$, which can be *directed* or *undirected**
- Explores the graph starting from s , touching all vertices that are reachable from s
 - ▶ We also enumerate *all possible* seeds and traverse the entire graph eventually

- *Input*: $G = (V, E)$, which can be *directed* or *undirected*
- Explores the graph starting from s , touching all vertices that are reachable from s
 - ▶ We also enumerate *all possible* seeds and traverse the entire graph eventually
- Visiting of vertices is done in *recursive* fashion:
 - ▶ When we visit a vertex u , we immediately visit an adjacent vertex v of u without finishing the visiting of u
 - ▶ We finish visiting u when all adjacent vertices has been visited (hence the finishing of the visiting is defined *recursively*)
 - ▶ We backtrack when we finish visiting a vertex (done automatically by recursion)

- *Input*: $G = (V, E)$, which can be *directed* or *undirected*
- Explores the graph starting from s , touching all vertices that are reachable from s
 - ▶ We also enumerate *all possible* seeds and traverse the entire graph eventually
- Visiting of vertices is done in *recursive* fashion:
 - ▶ When we visit a vertex u , we immediately visit an adjacent vertex v of u without finishing the visiting of u
 - ▶ We finish visiting u when all adjacent vertices has been visited (hence the finishing of the visiting is defined *recursively*)
 - ▶ We backtrack when we finish visiting a vertex (done automatically by recursion)
- Produces a **DFS forest**, consisting of all the **DFS trees** rooted at the seeds

- *Input: $G = (V, E)$, which can be *directed* or *undirected**
- Explores the graph starting from s , touching all vertices that are reachable from s
 - ▶ We also enumerate *all possible* seeds and traverse the entire graph eventually
- Visiting of vertices is done in *recursive* fashion:
 - ▶ When we visit a vertex u , we immediately visit an adjacent vertex v of u without finishing the visiting of u
 - ▶ We finish visiting u when all adjacent vertices has been visited (hence the finishing of the visiting is defined *recursively*)
 - ▶ We backtrack when we finish visiting a vertex (done automatically by recursion)
- Produces a **DFS forest**, consisting of all the **DFS trees** rooted at the seeds
- Coloring for vertices:
 - ▶ **white**: not yet visited
 - ▶ **grey**: being visited, but haven't finished visiting
 - ▶ **black**: finished visiting

- *Input*: $G = (V, E)$, which can be *directed* or *undirected*
- Explores the graph starting from s , touching all vertices that are reachable from s
 - ▶ We also enumerate *all possible* seeds and traverse the entire graph eventually
- Visiting of vertices is done in *recursive* fashion:
 - ▶ When we visit a vertex u , we immediately visit an adjacent vertex v of u without finishing the visiting of u
 - ▶ We finish visiting u when all adjacent vertices has been visited (hence the finishing of the visiting is defined *recursively*)
 - ▶ We backtrack when we finish visiting a vertex (done automatically by recursion)
- Produces a **DFS forest**, consisting of all the **DFS trees** rooted at the seeds
- Coloring for vertices:
 - ▶ **white**: not yet visited
 - ▶ **grey**: being visited, but haven't finished visiting
 - ▶ **black**: finished visiting
- Associates **two time-stamps** to each vertex
 - ▶ $d[u]$ records when DFS starts visiting u (turns *grey*)
 - ▶ $f[u]$ records when DFS finishes visiting u and therefore backtracks from u (turns *black*)

DFS(G)

```
1  for each vertex  $u \in V(G)$ 
2       $color[u] = WHITE$ 
3       $\pi[u] = NIL$ 
4   $time = 0$  // “global” variable
5  for each vertex  $u \in V(G)$ 
6      if  $color[u] == WHITE$ 
7          DFS-VISIT( $u$ )
```

DFS-VISIT(u)

```
1   $color[u] = GREY$ 
2   $time = time + 1$ 
3   $d[u] = time$ 
4  for each  $v \in Adj[u]$ 
5      if  $color[v] == WHITE$ 
6           $\pi[v] = u$ 
7          DFS-VISIT( $v$ )
8   $color[u] = BLACK$ 
9   $time = time + 1$ 
10  $f[u] = time$ 
```

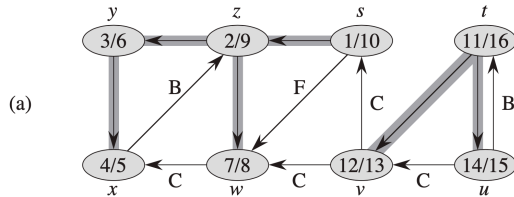
DFS(G)

```
1  for each vertex  $u \in V(G)$ 
2       $color[u] = WHITE$ 
3       $\pi[u] = NIL$ 
4   $time = 0$  // “global” variable
5  for each vertex  $u \in V(G)$ 
6      if  $color[u] == WHITE$ 
7          DFS-VISIT( $u$ )
```

DFS-VISIT(u)

```
1   $color[u] = GREY$ 
2   $time = time + 1$ 
3   $d[u] = time$ 
4  for each  $v \in Adj[u]$ 
5      if  $color[v] == WHITE$ 
6           $\pi[v] = u$ 
7          DFS-VISIT( $v$ )
8   $color[u] = BLACK$ 
9   $time = time + 1$ 
10  $f[u] = time$ 
```

- 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:

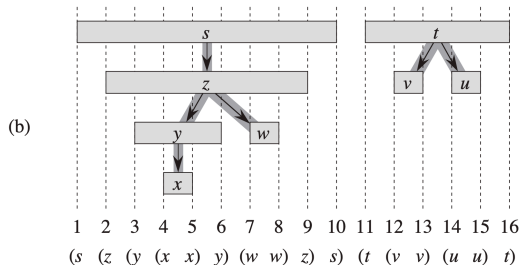
1. The intervals $[d[u], f[u]]$ and $[d[v], f[v]]$ are entirely disjoint, and neither one is a descendant of the other in the DFS forest
2. One interval is entirely contained in the other interval, and the vertex is a descendant of another (e.g., $[d[v], f[v]] \subseteq [d[u], f[u]]$ and v is a descendant of u)

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:

1. The intervals $[d[u], f[u]]$ and $[d[v], f[v]]$ are entirely disjoint, and neither one is a descendant of the other in the DFS forest
2. One interval is entirely contained in the other interval, and the vertex is a descendant of another (e.g., $[d[v], f[v]] \subseteq [d[u], f[u]]$ and v is a descendant of u)

Example (from CLRS):



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:

1. The intervals $[d[u], f[u]]$ and $[d[v], f[v]]$ are entirely disjoint, and neither one is a descendant of the other in the DFS forest
2. One interval is entirely contained in the other interval, and the vertex is a descendant of another (e.g., $[d[v], f[v]] \subseteq [d[u], f[u]]$ and v is a descendant of u)

proof:

- Without loss of generality, assume $d[u] < d[v]$

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:

1. The intervals $[d[u], f[u]]$ and $[d[v], f[v]]$ are entirely disjoint, and neither one is a descendant of the other in the DFS forest
2. One interval is entirely contained in the other interval, and the vertex is a descendant of another (e.g., $[d[v], f[v]] \subseteq [d[u], f[u]]$ and v is a descendant of u)

proof:

- Without loss of generality, assume $d[u] < d[v]$
- Then, by comparing $d[v]$ with $f[u]$, we have two case: (1) $d[v] < f[u]$; (2) $d[v] > f[u]$

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:

1. The intervals $[d[u], f[u]]$ and $[d[v], f[v]]$ are entirely disjoint, and neither one is a descendant of the other in the DFS forest
2. One interval is entirely contained in the other interval, and the vertex is a descendant of another (e.g., $[d[v], f[v]] \subseteq [d[u], f[u]]$ and v is a descendant of u)

proof:

- Without loss of generality, assume $d[u] < d[v]$
- Then, by comparing $d[v]$ with $f[u]$, we have two case: (1) $d[v] < f[u]$; (2) $d[v] > f[u]$
- First consider $d[v] < f[u]$ (aka. $d[u] < d[v] < f[u]$)

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:

1. The intervals $[d[u], f[u]]$ and $[d[v], f[v]]$ are entirely disjoint, and neither one is a descendant of the other in the DFS forest
2. One interval is entirely contained in the other interval, and the vertex is a descendant of another (e.g., $[d[v], f[v]] \subseteq [d[u], f[u]]$ and v is a descendant of u)

proof:

- Without loss of generality, assume $d[u] < d[v]$
- Then, by comparing $d[v]$ with $f[u]$, we have two case: (1) $d[v] < f[u]$; (2) $d[v] > f[u]$
- First consider $d[v] < f[u]$ (aka. $d[u] < d[v] < f[u]$)
- 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*

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:

1. The intervals $[d[u], f[u]]$ and $[d[v], f[v]]$ are entirely disjoint, and neither one is a descendant of the other in the DFS forest
2. One interval is entirely contained in the other interval, and the vertex is a descendant of another (e.g., $[d[v], f[v]] \subseteq [d[u], f[u]]$ and v is a descendant of u)

proof:

- Without loss of generality, assume $d[u] < d[v]$
- Then, by comparing $d[v]$ with $f[u]$, we have two case: (1) $d[v] < f[u]$; (2) $d[v] > f[u]$
- First consider $d[v] < f[u]$ (aka. $d[u] < d[v] < f[u]$)
- 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

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:

1. The intervals $[d[u], f[u]]$ and $[d[v], f[v]]$ are entirely disjoint, and neither one is a descendant of the other in the DFS forest
2. One interval is entirely contained in the other interval, and the vertex is a descendant of another (e.g., $[d[v], f[v]] \subseteq [d[u], f[u]]$ and v is a descendant of u)

proof:

- Without loss of generality, assume $d[u] < d[v]$
- Then, by comparing $d[v]$ with $f[u]$, we have two case: (1) $d[v] < f[u]$; (2) $d[v] > f[u]$
- First consider $d[v] < f[u]$ (aka. $d[u] < d[v] < f[u]$)
- 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

White-Path Theorem

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

White-Path Theorem

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

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 $d[u]$

White-Path Theorem

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 consisting of *only* white vertices

proof (continue):

- “ \Leftarrow ”: Use proof by contradiction, suppose that there is a “white path” from u to v at time $d[u]$, but v is not a descendant of u in the DFS tree

White-Path Theorem

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 consisting of *only* white vertices

proof (continue):

- “ \Leftarrow ”: Use proof by contradiction, suppose that there is a “white path” from u to v at time $d[u]$, but v is not a descendant of u in the DFS tree
- Let x be the first vertex on the path that is not a descendant of u (why such an x exists?)

White-Path Theorem

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 consisting of *only* white vertices

proof (continue):

- “ \Leftarrow ”: Use proof by contradiction, suppose that there is a “white path” from u to v at time $d[u]$, but v is not a descendant of u in the DFS tree
- Let x be the first vertex on the path that is not a descendant of u (why such an x exists?)
- Let w be the predecessor of x on the path (so that w is a descendant of u ; notice that w could be u itself)

White-Path Theorem

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 consisting of *only* white vertices

proof (continue):

- “ \Leftarrow ”: Use proof by contradiction, suppose that there is a “white path” from u to v at time $d[u]$, but v is not a descendant of u in the DFS tree
- Let x be the first vertex on the path that is not a descendant of u (why such an x exists?)
- Let w be the predecessor of x on the path (so that w is a descendant of u ; notice that w could be u itself)
- Since $d[u] < d[x]$, by the Parenthesis Theorem, we must have $d[u] < f[u] < d[x]$ (because x is not descendant of u)

White-Path Theorem

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 consisting of *only* white vertices

proof (continue):

- “ \Leftarrow ”: Use proof by contradiction, suppose that there is a “white path” from u to v at time $d[u]$, but v is not a descendant of u in the DFS tree
- Let x be the first vertex on the path that is not a descendant of u (why such an x exists?)
- Let w be the predecessor of x on the path (so that w is a descendant of u ; notice that w could be u itself)
- Since $d[u] < d[x]$, by the Parenthesis Theorem, we must have $d[u] < f[u] < d[x]$ (because x is not descendant of u)
- Consider the time the search visits w , we must have that x is white during the whole process (because if you haven't finish visiting w , you definitely haven't finished visiting u)

White-Path Theorem

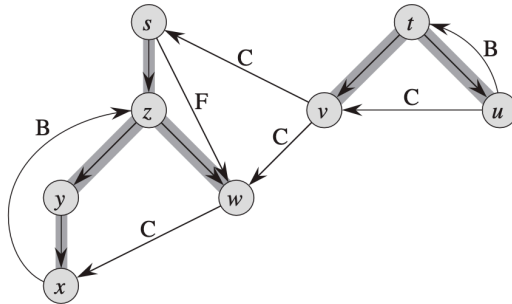
In a DFS forest of a (directed or undirected) graph G , a vertex v is a descendant of a vertex u if and only if at time $d[u]$, there is a path from u to v consisting of *only* white vertices

proof (continue):

- “ \Leftarrow ”: Use proof by contradiction, suppose that there is a “white path” from u to v at time $d[u]$, but v is not a descendant of u in the DFS tree
- Let x be the first vertex on the path that is not a descendant of u (why such an x exists?)
- Let w be the predecessor of x on the path (so that w is a descendant of u ; notice that w could be u itself)
- Since $d[u] < d[x]$, by the Parenthesis Theorem, we must have $d[u] < f[u] < d[x]$ (because x is not descendant of u)
- Consider the time the search visits w , we must have that x is white during the whole process (because if you haven't finish visiting w , you definitely haven't finished visiting u)
- 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)

Detecting cycles in an undirected graph using DFS

Lemma

A directed graph G has a cycle if and only if a depth-first search on G yields a back edge

Detecting cycles in an undirected graph using DFS

Lemma

A directed graph G has a cycle if and only if a depth-first search on G yields a back edge

proof:

- “ \Leftarrow ”: easy

Detecting cycles in an undirected graph using DFS

Lemma

A directed graph G has a cycle if and only if a depth-first search on G yields a back edge

proof:

- “ \Leftarrow ”: easy
- Now we try to show the forward direction “ \Rightarrow ”

Detecting cycles in an undirected graph using DFS

Lemma

A directed graph G has a cycle if and only if a depth-first search on G yields a back edge

proof:

- “ \Leftarrow ”: easy
- Now we try to show the forward direction “ \Rightarrow ”
- Suppose that G contains a cycle c

Detecting cycles in an undirected graph using DFS

Lemma

A directed graph G has a cycle if and only if a depth-first search on G yields a back edge

proof:

- “ \Leftarrow ”: easy
- Now we try to show the forward direction “ \Rightarrow ”
- Suppose that G contains a cycle c
- Without loss of generality, assume c is a simple cycle

Detecting cycles in an undirected graph using DFS

Lemma

A directed graph G has a cycle if and only if a depth-first search on G yields a back edge

proof:

- “ \Leftarrow ”: easy
- Now we try to show the forward direction “ \Rightarrow ”
- Suppose that G contains a cycle c
- Without loss of generality, assume c is a simple cycle
- Let v be the first vertex on c discovered by DFS , and let u be the vertex pointing to v on c

Detecting cycles in an undirected graph using DFS

Lemma

A directed graph G has a cycle if and only if a depth-first search on G yields a back edge

proof:

- “ \Leftarrow ”: easy
- Now we try to show the forward direction “ \Rightarrow ”
- Suppose that G contains a cycle c
- Without loss of generality, assume c is a simple cycle
- Let v be the first vertex on c discovered by DFS, and let u be the vertex pointing to v on c
- When v is discovered, we have that all vertices on path from v to u (on c) are white (undiscovered)

Detecting cycles in an undirected graph using DFS

Lemma

A directed graph G has a cycle if and only if a depth-first search on G yields a back edge

proof:

- “ \Leftarrow ”: easy
- Now we try to show the forward direction “ \Rightarrow ”
- Suppose that G contains a cycle c
- Without loss of generality, assume c is a simple cycle
- Let v be the first vertex on c discovered by DFS, and let u be the vertex pointing to v on c
- When v is discovered, we have that all vertices on path from v to u (on c) are white (undiscovered)
- By the White-Path Theorem, u must be a descendant of v in the depth-first forest

Detecting cycles in an undirected graph using DFS

Lemma

A directed graph G has a cycle if and only if a depth-first search on G yields a back edge

proof:

- “ \Leftarrow ”: easy
- Now we try to show the forward direction “ \Rightarrow ”
- Suppose that G contains a cycle c
- Without loss of generality, assume c is a simple cycle
- Let v be the first vertex on c discovered by DFS, and let u be the vertex pointing to v on c
- When v is discovered, we have that all vertices on path from v to u (on c) are white (undiscovered)
- By the White-Path Theorem, u must be a descendant of v in the depth-first forest
- Therefore, (u, v) is a back edge

Topological Sort: Alternative Algorithm

TOPOLOGICAL-SORT(G)

- 1 **DFS**(G)
- 2 output V sorted in reverse order of $f[\cdot]$

Proof of correctness:

Topological Sort: Alternative Algorithm

TOPOLOGICAL-SORT(G)

- 1 **DFS**(G)
- 2 output V sorted in reverse order of $f[\cdot]$

Proof of correctness:

- It suffices to show that for any edge $(u, v) \in G$, $f[v] < f[u]$

Topological Sort: Alternative Algorithm

TOPOLOGICAL-SORT(G)

- 1 **DFS**(G)
- 2 output V sorted in reverse order of $f[\cdot]$

Proof of correctness:

- It suffices to show that for any edge $(u, v) \in G$, $f[v] < f[u]$
- First 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

Topological Sort: Alternative Algorithm

TOPOLOGICAL-SORT(G)

- 1 **DFS**(G)
- 2 output V sorted in reverse order of $f[\cdot]$

Proof of correctness:

- It suffices to show that for any edge $(u, v) \in G$, $f[v] < f[u]$
- First 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
- When we are visiting u and exploring the edge (u, v) in DFS, we have that v cannot be grey because otherwise (u, v) would be a back edge (notice that u must be at the stack top when we are exploring (u, v)), contradicting the previous Lemma saying that DFS on DAG yields no back edges

Topological Sort: Alternative Algorithm

TOPOLOGICAL-SORT(G)

- 1 **DFS**(G)
- 2 output V sorted in reverse order of $f[\cdot]$

Proof of correctness:

- It suffices to show that for any edge $(u, v) \in G$, $f[v] < f[u]$
- First 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
- When we are visiting u and exploring the edge (u, v) in DFS, we have that v cannot be grey because otherwise (u, v) would be a back edge (notice that u must be at the stack top when we are exploring (u, v)), contradicting the previous Lemma saying that DFS on DAG yields no back edges
- Then v must be white or black

Topological Sort: Alternative Algorithm

TOPOLOGICAL-SORT(G)

- 1 **DFS**(G)
- 2 output V sorted in reverse order of $f[\cdot]$

Proof of correctness:

- It suffices to show that for any edge $(u, v) \in G$, $f[v] < f[u]$
- First 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
- When we are visiting u and exploring the edge (u, v) in DFS, we have that v cannot be grey because otherwise (u, v) would be a back edge (notice that u must be at the stack top when we are exploring (u, v)), contradicting the previous Lemma saying that DFS on DAG yields no back edges
- Then v must be white or black
- If v is white, then we shall visit v as a result of exploring the edge (u, v) . By DFS, we cannot finish visiting u before finishing visiting v . So $f[v] < f[u]$.

Topological Sort: Alternative Algorithm

TOPOLOGICAL-SORT(G)

- 1 **DFS**(G)
- 2 output V sorted in reverse order of $f[\cdot]$

Proof of correctness:

- It suffices to show that for any edge $(u, v) \in G$, $f[v] < f[u]$
- First 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
- When we are visiting u and exploring the edge (u, v) in DFS, we have that v cannot be grey because otherwise (u, v) would be a back edge (notice that u must be at the stack top when we are exploring (u, v)), contradicting the previous Lemma saying that DFS on DAG yields no back edges
- Then v must be white or black
- If v is white, then we shall visit v as a result of exploring the edge (u, v) . By DFS, we cannot finish visiting u before finishing visiting v . So $f[v] < f[u]$.
- If v is black, we have already finished visiting v . But the visiting of u is not finished. So we obviously have $f[v] < f[u]$.

Observation

If there is a path from a vertex u to a vertex v in an ***undirected*** graph G (aka. u, v are in the same connected component), then u, v must be in the same DFS tree after performing a depth-first search on G .

Observation

If there is a path from a vertex u to a vertex v in an **undirected** graph G (aka. u, v are in the same connected component), then u, v must be in the same DFS tree after performing a depth-first search on G .

Comment: The opposite is also true. Think about what these observations implies

Observation

If there is a path from a vertex u to a vertex v in an **undirected** graph G (aka. u, v are in the same connected component), then u, v must be in the same DFS tree after performing a depth-first search on G .

Comment: The opposite is also true. Think about what these observations implies

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.