

Introduction to Algorithms

Lecture 8 Graph and DFS

Xue Chen

xuechen1989@ustc.edu.cn

2025 spring in



Course Logistic

- ① Midterm: 13:30 — 15:30 on April 17, no make-up exam
- ② Office Hour: 3A103 (this week and next week)

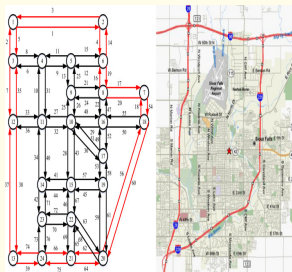
Outline

Overview

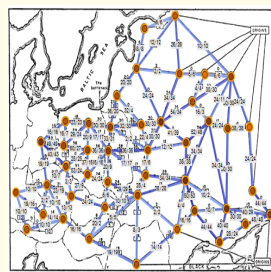
Graphs are fundamental objects in many areas:



(a) Social Network



(b) Traffic Network

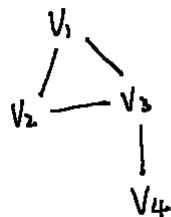


(c) Network Flow

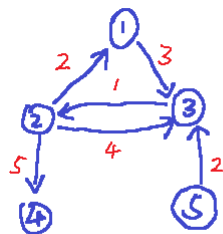
Overview (II)

Basic notation

- 1 Graph G is specified by its vertex set V and edge set E
- 2 n vertices (a.k.a. nodes, points, terminals) in $V: \{v_1, \dots, v_n\}$
- 3 m edges in E : could be undirected or directed, unweighted or weighted, ...



(a) undirected



(b) directed & weighted

Its flexibility captures abstract models of many practical problems

Plan

Many interesting problems

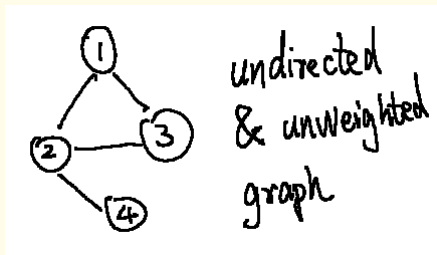
- 1 Connectivity: directed vs undirected, connected components, minimal spanning trees,
- 2 Shortest paths: distance, negative costs, single-source vs all pairs, . . .
- 3 Max flow, min cut, matching, . . .
- 4 (optional) max cut, random walk, page-rank, effective resistance, . . .

Our focus in the next 3 weeks!

Outline

Introduction

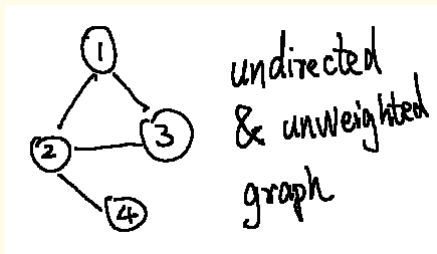
Two standard ways to store/access a graph: consider **unweighted undirected graph** for convenience



- 1 Adjacency matrix $A \in \mathbb{R}^{n \times n}$: $A[i, j] = 1$ if $(i, j) \in E$; o.w. $A[i, j] = 0$.

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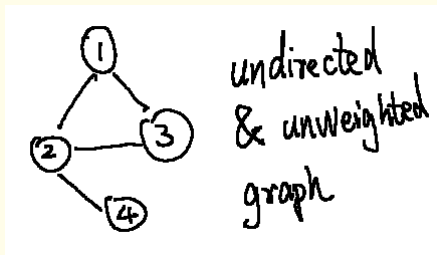


① Adjacency matrix $A \in \mathbb{R}^{n \times n}$: $A[i, j] = 1$ if $(i, j) \in E$; o.w. $A[i, j] = 0$.

$$A = \begin{bmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}$$

Introduction

Two standard ways to store/access a graph: consider **unweighted undirected graph** for convenience



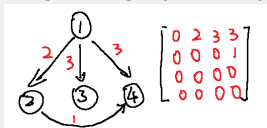
- 1 Adjacency matrix $A \in \mathbb{R}^{n \times n}$: $A[i, j] = 1$ if $(i, j) \in E$; o.w. $A[i, j] = 0$.
- 2 For each vertex v , store all its edges in an adjacency-list $Adj[v]$:

$$Adj[1] = \{2, 3\}, Adj[2] = \{3, 1, 4\}, Adj[3] = \{1, 2\}, Adj[4] = \{2\}.$$

Adjacency Matrix

General Graphs

- 1 For directed graph, $A[i, j] = 1$ only if $(i, j) \in E \Rightarrow A[i, j] \neq A[j, i]$
- 2 For weighted graph, $A[i, j] = w_e$ for edge $e = (i, j)$

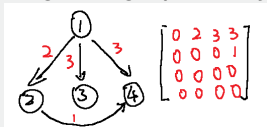


- 3 Many variants: A^T , normalized $A[i, j] = \frac{A[i, j]}{\deg(i)}$, symmetrical normalized $A[i, j] = \frac{A[i, j]}{\sqrt{\deg(i) \cdot \deg(j)}}$, ...

Adjacency Matrix

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Pro: (1) Easy to access and maintain; (2) Elegant math expression; (3) Rich tools from matrix theory;

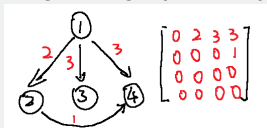
Example

What does $A^{100}[i, j]$ stand for?

Adjacency Matrix

General Graphs

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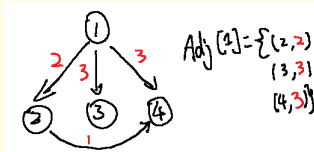
Pro: (1) Easy to access and maintain; (2) Elegant math expression; (3) Rich tools from matrix theory;

Con: n^2 space is wasteful when $|E|$ is small

Adjacency List

General Graphs

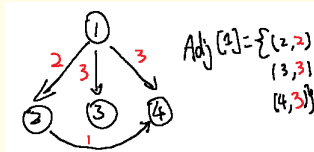
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- 2 Maintain another list for incoming edges
- 3 Record weights



Adjacency List

General Graphs

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Pro: (1) Save space; (2) Improve running time, $O(n^2) \rightarrow O(m)$, for sparse graphs

Con: (1) Hard to maintain and access — how to determine $(i, j) \in E$?
(2) No clean math notation

Summary

Usually we use adjacency matrix for dense graphs and adjacency list for sparse graphs.

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Usually we use **adjacency matrix** for dense graphs and **adjacency list** for sparse graphs.



Pick the correct representation based on applications!

Outline

Exploring Graphs

Given an **undirected** graph $G = (V, E)$, explore all reachable nodes?



Exploring Graphs

Given an **undirected** graph $G = (V, E)$, explore all reachable nodes?



Basic idea:

- 1 Keep track of all nodes discovered;
- 2 While there is an unexplored path, follow it

Algorithm Description

Keep track of (1) discovered vertices; (2) which edge to follow.

procedure DFS-EXPLORE(u)

// visited(v) = False for all v in the initial stage

visited(u) = True

for each $v \in \text{Adj}[u]$ **do**

if *visited(v) = False* **then**

$v.\pi = u$

// Record v 's parent node

 DFS-EXPLORE(v)

Analysis: Correctness

Claim

When all vertices have $visited(v) = False$, DFS-EXPLORE(u) will mark all vertices reachable from u to *True*.

Analysis: Correctness

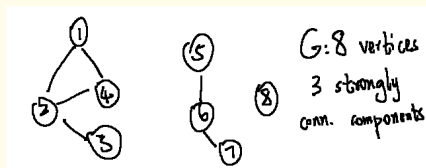
Claim

When all vertices have $visited(v) = False$, DFS-EXPLORE(u) will mark all vertices reachable from u to *True*.

Proof Sketch: For contradiction, suppose v is not marked and \exists a path from u to v . Consider the previous node w in front of v on this path ...

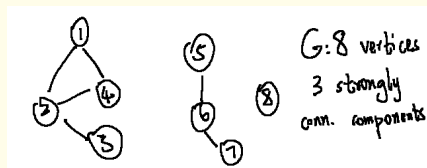
Connected Components

```
procedure DFS( $G$ )  
  for each  $v \in V$  do  
     $visited(v) = False$   
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  for each  $v \in V$  do  
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Connected Components

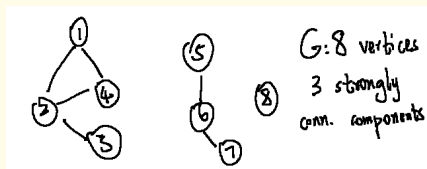
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- 1 To explore all vertices, restart DFS at any vertex that has not yet been visited.

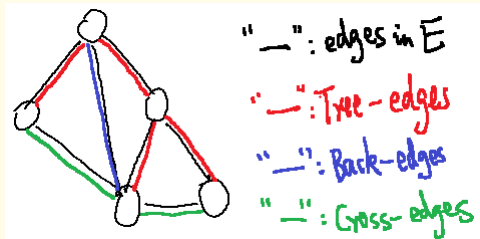
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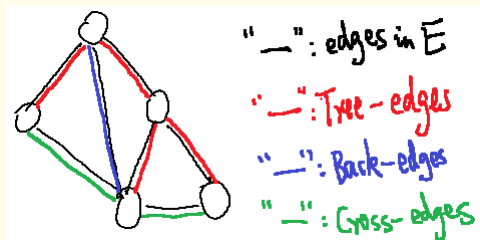
- 1 To explore all vertices, restart DFS at any vertex that has not yet been visited.
- 2 **Strongly Connected Components:** a *maximal* set of vertices $C \subseteq V$ s.t. for every $u, v \in C$, $u \rightarrow v$ and $v \rightarrow u$.
- 3 Each DFS-EXPLORE call finds a strongly connected component

Types of Edges



Define 3 types of edges according to a given **tree**

Types of Edges



Define 3 types of edges according to a given **tree**

Theorem 22.10 in **CLRS** for **Undirected Graphs**

Consider the DFS-tree based on $v.\pi$, all edges in E are either tree-edges or back-edges — there is no **cross-edges**

Running Time

Theorem

The running time is $O(n + m)$.

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DFS-EXPLORE(u) is applied at most once for each u and $\sum_u \deg(u) = 2m$.

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$visited(u) = \text{True}$

for each $v \in Adj[u]$ **do**

if $visited(v) = \text{False}$ **then**

$v.\pi = u$

 DFS-EXPLORE(v)

// Time: $O(\deg(u))$

procedure DFS(G)

for each $v \in V$ **do**

$visited(v) = \text{False}$

$v.\pi = \text{NIL}$

for each $v \in V$ **do**

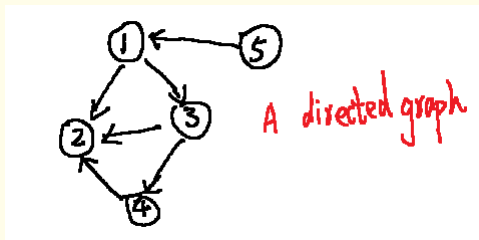
if $visited(v) = \text{False}$ **then**

 DFS-EXPLORE(v)

Outline

Introduction

While everything looks straightforward for undirected graphs, it becomes more involved in directed graphs.



- ① DFS
- ② Strongly connected components
- ③ Directed Acyclic Graph

Basic DFS

Consider the **same** DFS procedure:

```
procedure DFS-EXPLORE( $u$ )
```

```
   $visited(u) = True$ 
```

```
  for each  $v \in Adj[u]$  do
```

```
    if  $visited(v) = False$  then
```

```
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```

```
      DFS-EXPLORE( $v$ )
```

// Time: $O(deg(u))$

Basic DFS

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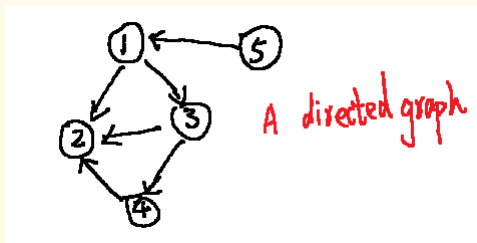
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      DFS-EXPLORE( $v$ )
```

Fact

In directed graphs, it finds all vertices reachable from u .

Next question: How about strongly connected components?

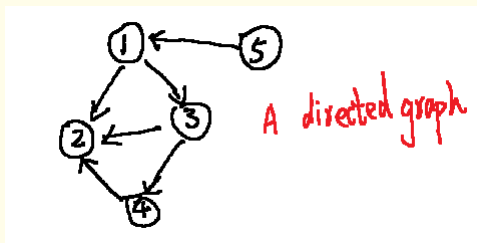
Strongly Connected Components



Question

How many strongly connected components?

Strongly Connected Components

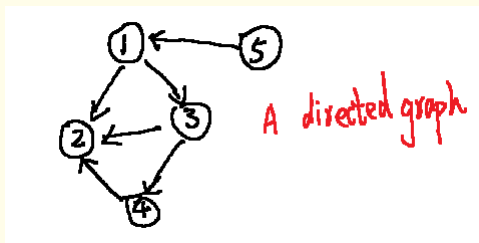


Question

How many strongly connected components?

Ans: 5, each vertex contribute a component.

Strongly Connected Components



A directed graph

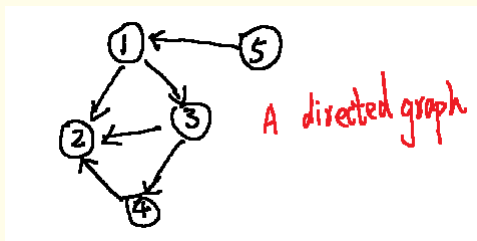
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- 1 DFS-EXPLORE does not find a connected component every time.
- 2 Discuss forward/backward edges and cross edges again.

Strongly Connected Components



Question

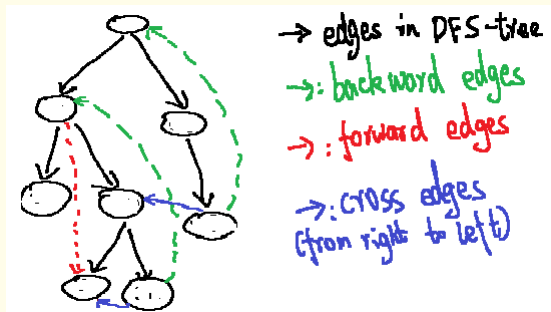
How many strongly connected components?

Ans: 5, each vertex contribute a component.

- 1 DFS-EXPLORE does not find a connected component every time.
- 2 Discuss forward/backward edges and cross edges again.
- 3 Define time-stamps for vertices in DFS
- 4 Show the algorithm to find connected components in **linear time**

Types of Edges

In a directed graph, the DFS procedure labels all edges with 4 types:



Exception: No edge from left to right.

Observation

Strongly conn. components are defined by **backward edges**.

Time Stamps

Define time-stamps to (1) determine the type of each edge and (2) use **backward edges** to find conn. components in **linear time** $O(n + m)$

procedure DFS-EXPLORE(u)

$t = t + 1; u.d = t;$

// Discover u at t

$visited(u) = True$

for each $v \in Adj[u]$ **do**

if $visited(v) = False$ **then**

$v.\pi = u$

 DFS-EXPLORE(v)

$t = t + 1; u.f = t;$

// Finish at t

procedure DFS(G)

$t = 0$

for each $v \in V$ **do**

$visited(v) = False$

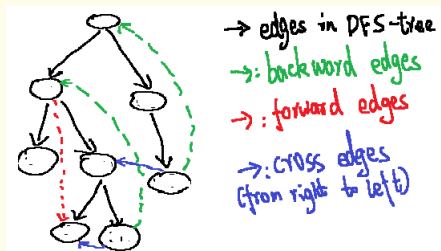
$v.\pi = NIL$

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Example

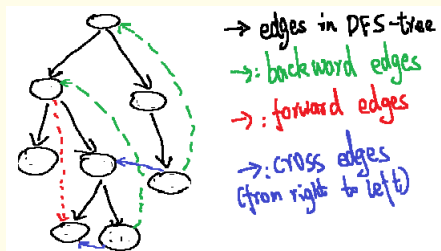


Theorem 22.7 in CLRS

For two vertices u and v ,

- ① $[v.d, v.f] \subset [u.d, u.f]$: v is a descendant of u
- ② $[u.d, u.f] \subset [v.d, v.f]$: u is a descendant of v
- ③ $[u.d, u.f]$ and $[v.d, v.f]$ are disjoint: Neither u nor v is a descendant of the other in DFS

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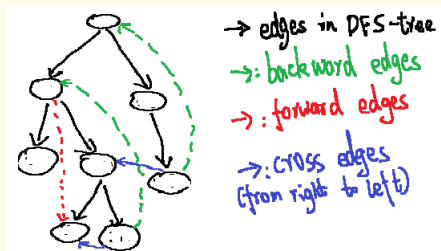


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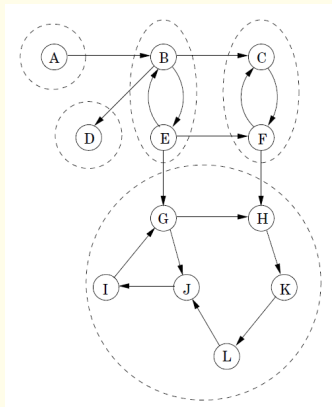
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These properties hold for both directed and undirected DFS.

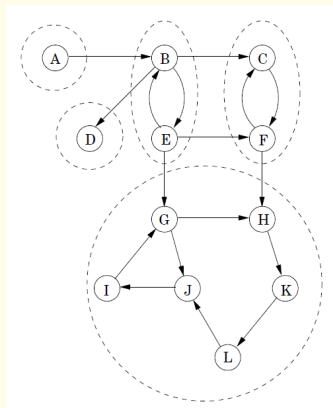
Find Connected Components



Intuition:

- 1 If the DFS starts with a **sink** conn. component like D or $\{G, H, K, I, J, L\}$, it will find the correct component

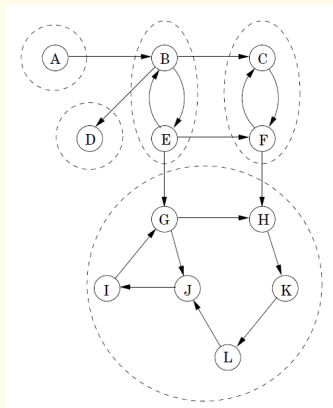
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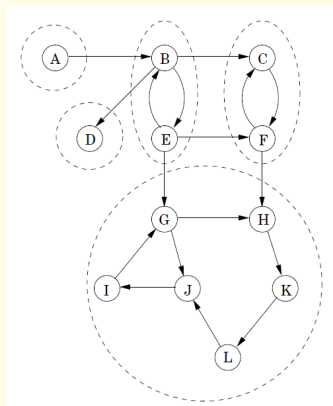
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- 4 Question: Find a way to start from D or $\{G, H, K, I, J, L\}$?

Find Connected Components



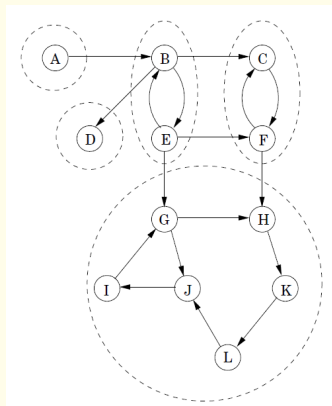
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Observation

After Procedure DFS marks all vertices, the node u that receives the highest $u.f$ must lie in a **source** conn. component.

Find Connected Components



Intuition:

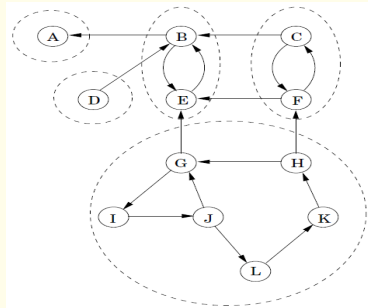
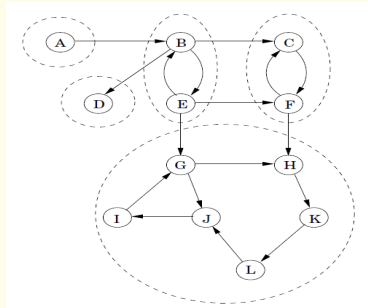
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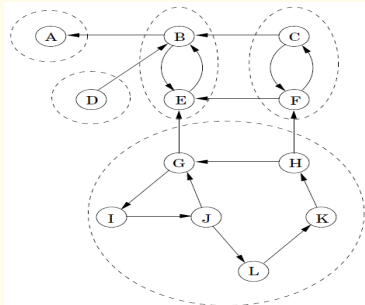
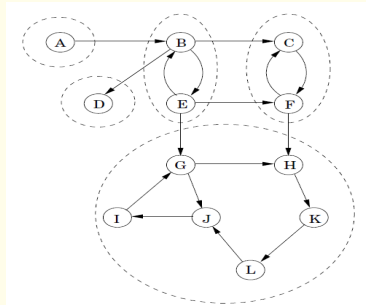
After Procedure DFS marks all vertices, the node u that receives the highest $u.f$ must lie in a **source** conn. component.

The last node must be A

Find **source** conn. component by considering the reverse graph G^T



Find **source** conn. component by considering the reverse graph G^T



Description

STRONGLY-CONNECTED-COMPONENTS (G)

- 1 call $\text{DFS}(G)$ to compute finishing times $u.f$ for each vertex u
- 2 compute G^T
- 3 call $\text{DFS}(G^T)$, 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

Analysis

Running time: $O(n + m)$.

Correctness: Lemma 22.14 in [CLRS](#)

If C and C' are two strongly connected components with an edge from C to C' , then the highest $u.f$ in C is bigger than the highest $u'.f$ in C' .

Analysis

Running time: $O(n + m)$.

Correctness: Lemma 22.14 in [CLRS](#)

If C and C' are two strongly connected components with an edge from C to C' , then the highest $u.f$ in C is bigger than the highest $u'.f$ in C' .

Proof: 2 cases

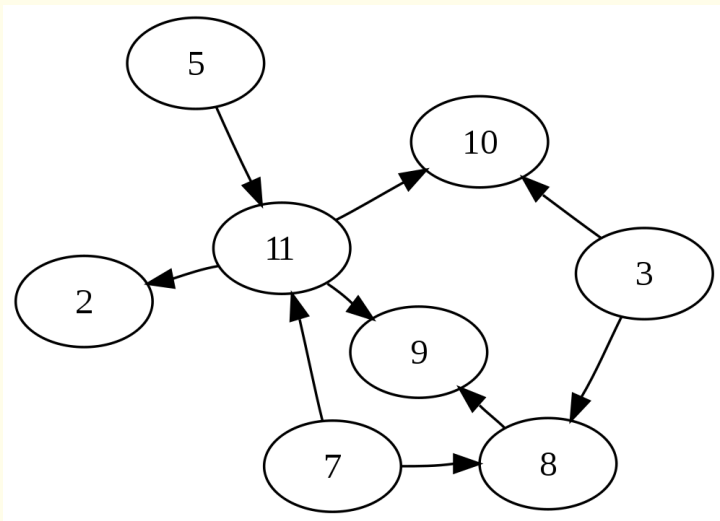
- 1 Visit C before C'
- 2 Visit C' before C

This lemma indicates that Line 3 picks a **source** conn. component in G , which is a **sink** conn. component in G^T .

Outline

Directed Acyclic Graph

If there is no cycle in G , all conn. components are of size 1 — called acyclic.



Topological Order

A order σ on all vertices such that u appears before v in σ whenever $(u, v) \in G$.

Applications

- 1 Dependency relation: compiler, resource managements, ...
- 2 Time order: Data processing, sociology, ...
- 3 Biology: Evolution, ...

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Applications

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Question: Given a DAG, how to compute the order?

Algorithm

- 1 Call $DFS(G)$ to compute finishing time $v.f$
- 2 Once finish processing a node v , add it to the front of the list
- 3 Output the list

Algorithm

- 1 Call $DFS(G)$ to compute finishing time $v.f$
- 2 Once finish processing a node v , add it to the front of the list
- 3 Output the list

Analysis

- 1 Running time: $O(n + m)$
- 2 Correctness: If $\exists (u, v) \in E$, $v.f < u.f$ is always true.

Questions?