

CS 374: Algorithms & Models of Computation

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Graphs, Representation, Search, DFS

Lecture 8
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Part I

Graph Basics

Why Graphs?

- 1 Graphs help model networks which are ubiquitous: transportation networks (rail, roads, airways), social networks (interpersonal relationships), information networks (web page links), and many problems that don't even look like graph problems.
- 2 Fundamental objects in Computer Science, Optimization, Combinatorics
- 3 Many important and useful optimization problems are graph problems
- 4 Graph theory: elegant, fun and deep mathematics

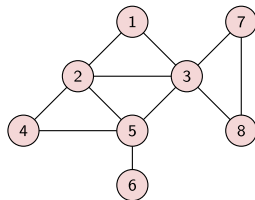
Graph

Definition

An undirected (simple) graph

$\mathbf{G} = (\mathbf{V}, \mathbf{E})$ is a 2-tuple:

- 1 \mathbf{V} is a set of vertices (also referred to as nodes/points)
- 2 \mathbf{E} is a set of edges where each edge $e \in \mathbf{E}$ is a set of the form $\{\mathbf{u}, \mathbf{v}\}$ with $\mathbf{u}, \mathbf{v} \in \mathbf{V}$ and $\mathbf{u} \neq \mathbf{v}$.



Example

In figure, $\mathbf{G} = (\mathbf{V}, \mathbf{E})$ where $\mathbf{V} = \{1, 2, 3, 4, 5, 6, 7, 8\}$ and $\mathbf{E} = \{\{1, 2\}, \{1, 3\}, \{2, 3\}, \{2, 4\}, \{2, 5\}, \{3, 5\}, \{3, 7\}, \{3, 8\}, \{4, 5\}, \{5, 6\}, \{7, 8\}\}$.

Example: Modeling Problems as Search

State Space Search

Many search problems can be modeled as search on a graph. The trick is figuring out what the vertices and edges are.

Missionaries and Cannibals

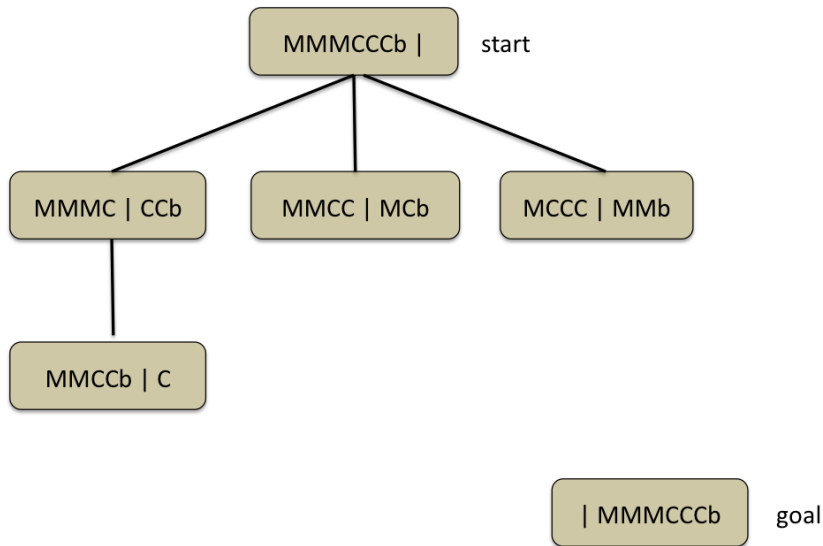
- Three missionaries, three cannibals, one boat, one river
- Boat carries two people, must have at least one person
- Must all get across
- At no time can cannibals outnumber missionaries

How is this a graph search problem?

What are the vertices?

What are the edges?

Example: Missionaries and Cannibals Graph



Notation and Convention

Notation

An edge in an undirected graphs is an *unordered* pair of nodes and hence it is a set. Conventionally we use (u, v) for $\{u, v\}$ when it is clear from the context that the graph is undirected.

- 1 u and v are the **end points** of an edge $\{u, v\}$
- 2 **Multi-graphs** allow
 - 1 *loops* which are edges with the same node appearing as both end points
 - 2 *multi-edges*: different edges between same pairs of nodes
- 3 In this class we will assume that a graph is a simple graph unless explicitly stated otherwise.

Graph Representation I

Adjacency Matrix

Represent $G = (V, E)$ with n vertices and m edges using a $n \times n$ adjacency matrix A where

- 1 $A[i, j] = A[j, i] = 1$ if $\{i, j\} \in E$ and $A[i, j] = A[j, i] = 0$ if $\{i, j\} \notin E$.
- 2 Advantage: can check if $\{i, j\} \in E$ in $O(1)$ time
- 3 Disadvantage: needs $\Omega(n^2)$ space even when $m \ll n^2$

Graph Representation II

Adjacency Lists

Represent $G = (V, E)$ with n vertices and m edges using adjacency lists:

- 1 For each $u \in V$, $\text{Adj}(u) = \{v \mid \{u, v\} \in E\}$, that is neighbors of u . Sometimes $\text{Adj}(u)$ is the list of edges incident to u .
- 2 Advantage: space is $O(m + n)$
- 3 Disadvantage: cannot “easily” determine in $O(1)$ time whether $\{i, j\} \in E$
 - 1 By sorting each list, one can achieve $O(\log n)$ time
 - 2 By hashing “appropriately”, one can achieve $O(1)$ time

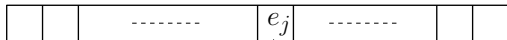
Note: In this class we will assume that by default, graphs are represented using plain vanilla (unsorted) adjacency lists.

A Concrete Representation

- Assume vertices are numbered arbitrarily as $\{1, 2, \dots, n\}$.
- Edges are numbered arbitrarily as $\{1, 2, \dots, m\}$.
- Edges stored in an array/list of size m . $E[j]$ is j 'th edge with info on end points which are integers in range 1 to n .
- Array Adj of size n for adjacency lists. $Adj[i]$ points to adjacency list of vertex i . $Adj[i]$ is a list of edge indices in range 1 to m .

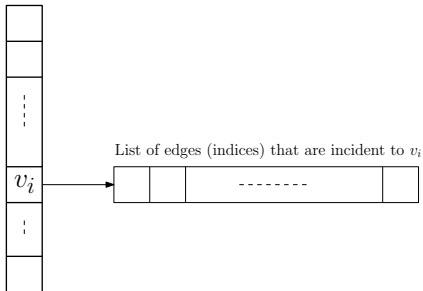
A Concrete Representation

Array of edges E



information including end point indices

Array of adjacency lists



A Concrete Representation: Advantages

- Edges are explicitly represented/numbered. Scanning/processing all edges easy to do.
- Representation easily supports multigraphs including self-loops.
- Explicit numbering of vertices and edges allows use of arrays: **$O(1)$** -time operations are easy to understand.
- Can also implement via pointer based lists for certain dynamic graph settings.

Connectivity

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

- 1 A **path** is a sequence of *distinct* vertices v_1, v_2, \dots, v_k such that $\{v_i, v_{i+1}\} \in E$ for $1 \leq i \leq k - 1$. The length of the path is $k - 1$ (the number of edges in the path) and the path is from v_1 to v_k . **Note:** a single vertex u is a path of length 0 .

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Caveat: Some times people use the term cycle to also allow vertices to be repeated; we will use the term **tour**.

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Connectivity

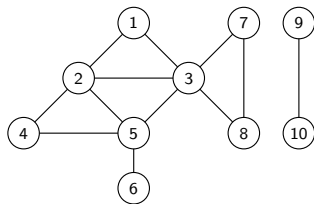
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- 3 A vertex u is **connected** to v if there is a path from u to v .
- 4 The **connected component** of u , $\text{con}(u)$, is the set of all vertices connected to u . Is $u \in \text{con}(u)$?

Connectivity contd

Define a relation \mathbf{C} on $\mathbf{V} \times \mathbf{V}$ as \mathbf{uCv} if \mathbf{u} is connected to \mathbf{v}

- 1 In undirected graphs, connectivity is a reflexive, symmetric, and transitive relation. Connected components are the equivalence classes.
- 2 Graph is **connected** if only one connected component.



Connectivity Problems

Algorithmic Problems

- 1 Given graph G and nodes u and v , is u *connected* to v ?
- 2 Given G and node u , find all nodes that are connected to u .
- 3 Find all connected components of G .

Connectivity Problems

Algorithmic Problems

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Can be accomplished in $O(m + n)$ time using **BFS** or **DFS**.

Basic Graph Search

Given $\mathbf{G} = (\mathbf{V}, \mathbf{E})$ and vertex $\mathbf{u} \in \mathbf{V}$:

Explore(\mathbf{u}):

Initialize $\mathbf{S} = \{\mathbf{u}\}$

while there is an edge (\mathbf{x}, \mathbf{y}) with $\mathbf{x} \in \mathbf{S}$ and $\mathbf{y} \notin \mathbf{S}$ **do**
 add \mathbf{y} to \mathbf{S}

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Proposition

Explore(u) *terminates with* $S = \text{con}(u)$.

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Running time: depends on implementation

- 1 Naive: $O(mn)$ with $O(m)$ time for each scan.
- 2 Breadth First Search (**BFS**): use **queue** data structure
- 3 Depth First Search (**DFS**): use **stack** data structure
- 4 DFS/BFS run in $O(m + n)$ time. Review CS 225 material!

Part II

DFS

Depth First Search

DFS is a very versatile graph exploration strategy. Hopcroft and Tarjan (Turing Award winners) demonstrated the power of **DFS** to understand graph structure. **DFS** can be used to obtain linear time ($O(m + n)$) algorithms for

- 1 Finding cut-edges and cut-vertices of undirected graphs
- 2 Finding strong connected components of directed graphs
- 3 Linear time algorithm for testing whether a graph is planar

DFS in Undirected Graphs

Recursive version.

DFS(G)

Mark all nodes as unvisited

while there is an unvisited node **u do**

DFS(u)

DFS(u)

Mark **u** as visited

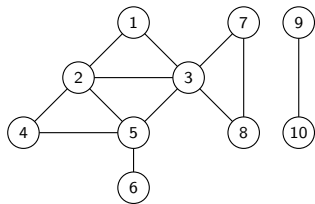
for each edge (u,v) in $\text{Adj}(u)$ **do**

if **v** is not marked

DFS(v)

Implemented using a global array **Mark** for all recursive calls.

Example



DFS Tree/Forest

DFS(G)

Mark all nodes unvisited

Set **T** to be empty

while \exists unvisited node **u** **do**

DFS(u)

Output **T**

DFS(u)

Mark **u** as visited

for **uv** in **Adj(u)** **do**

if **v** is not marked

 add **uv** to **T**

DFS(v)

DFS Tree/Forest

DFS(G)

Mark all nodes unvisited

Set T to be empty

while \exists unvisited node u **do**

DFS(u)

Output T

DFS(u)

Mark u as visited

for uv in $\text{Adj}(u)$ **do**

if v is not marked

 add uv to T

DFS(v)

Edges classified into two types: $uv \in E$ is a

- 1 **tree edge**: belongs to T
- 2 **non-tree edge**: does not belong to T

Properties of DFS tree

Proposition

- 1 \mathbf{T} is a forest
- 2 connected components of \mathbf{T} are same as those of \mathbf{G} .
- 3 If $\mathbf{uv} \in \mathbf{E}$ is a non-tree edge then, in \mathbf{T} , either:
 - 1 \mathbf{u} is an ancestor of \mathbf{v} , or
 - 2 \mathbf{v} is an ancestor of \mathbf{u} .

Question: Why are there no *cross-edges*?

DFS with Predecessors

Keep track of predecessors.

DFS(G)

```
for all  $u \in V(G)$  do
    Mark  $u$  as unvisited
    Set  $\text{pred}(u)$  to null
 $T$  is set to  $\emptyset$ 
while  $\exists$  unvisited  $u$  do
    DFS(u)
Output  $T$ 
```

DFS(u)

```
Mark  $u$  as visited
for each  $uv$  in  $\text{Out}(u)$  do
    if  $v$  is not marked then
        add edge  $uv$  to  $T$ 
        set  $\text{pred}(v)$  to  $u$ 
    DFS(v)
```

DFS with Visit Times

Keep track of when nodes are visited.

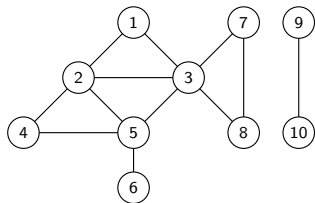
DFS(G)

```
for all  $u \in V(G)$  do
    Mark  $u$  as unvisited
T is set to  $\emptyset$ 
time = 0
while  $\exists$  unvisited  $u$  do
    DFS(u)
Output T
```

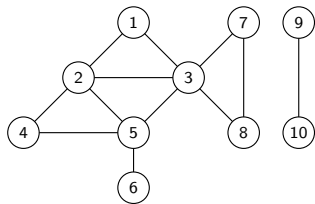
DFS(u)

```
Mark  $u$  as visited
pre(u) = ++time
for each  $uv$  in Out(u) do
    if  $v$  is not marked then
        add edge  $uv$  to T
        DFS(v)
post(u) = ++time
```


Example



Example



pre and post numbers

Node u is **active** in time interval $[\text{pre}(u), \text{post}(u)]$

Proposition

For any two nodes u and v , the two intervals $[\text{pre}(u), \text{post}(u)]$ and $[\text{pre}(v), \text{post}(v)]$ are disjoint or one is contained in the other.

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Proof.

- Assume without loss of generality that $\text{pre}(u) < \text{pre}(v)$. Then v visited after u .
- If **DFS**(v) invoked before **DFS**(u) finished, $\text{post}(v) < \text{post}(u)$.

pre and post numbers

Node **u** is **active** in time interval $[\text{pre}(\mathbf{u}), \text{post}(\mathbf{u})]$

Proposition

*For any two nodes **u** and **v**, the two intervals $[\text{pre}(\mathbf{u}), \text{post}(\mathbf{u})]$ and $[\text{pre}(\mathbf{v}), \text{post}(\mathbf{v})]$ are disjoint or one is contained in the other.*

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- Assume without loss of generality that $\text{pre}(\mathbf{u}) < \text{pre}(\mathbf{v})$. Then **v** visited after **u**.
- If **DFS(v)** invoked before **DFS(u)** finished, $\text{post}(\mathbf{v}) < \text{post}(\mathbf{u})$.
- If **DFS(v)** invoked after **DFS(u)** finished, $\text{pre}(\mathbf{v}) > \text{post}(\mathbf{u})$ \square

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- If $\text{DFS}(v)$ invoked after $\text{DFS}(u)$ finished, $\text{pre}(v) > \text{post}(u)$ \square

pre and **post** numbers useful in several applications of **DFS**- soon!

Part III

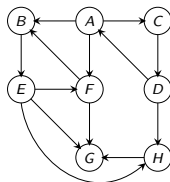
Directed Graphs and Decomposition

Directed Graphs

Definition

A directed graph $G = (V, E)$ consists of

- 1 set of vertices/nodes V and
- 2 a set of edges/arcs $E \subseteq V \times V$.



An edge is an *ordered* pair of vertices. (u, v) different from (v, u) .

Examples of Directed Graphs

In many situations relationship between vertices is asymmetric:

- 1 Road networks with one-way streets.
- 2 Web-link graph: vertices are web-pages and there is an edge from page p to page p' if p has a link to p' . Web graphs used by Google with PageRank algorithm to rank pages.
- 3 Dependency graphs in variety of applications: link from x to y if y depends on x . Make files for compiling programs.
- 4 Program Analysis: functions/procedures are vertices and there is an edge from x to y if x calls y .

Representation

Graph $\mathbf{G} = (\mathbf{V}, \mathbf{E})$ with \mathbf{n} vertices and \mathbf{m} edges:

- 1 **Adjacency Matrix**: $\mathbf{n} \times \mathbf{n}$ *asymmetric* matrix \mathbf{A} . $\mathbf{A}[\mathbf{u}, \mathbf{v}] = 1$ if $(\mathbf{u}, \mathbf{v}) \in \mathbf{E}$ and $\mathbf{A}[\mathbf{u}, \mathbf{v}] = 0$ if $(\mathbf{u}, \mathbf{v}) \notin \mathbf{E}$. $\mathbf{A}[\mathbf{u}, \mathbf{v}]$ is not same as $\mathbf{A}[\mathbf{v}, \mathbf{u}]$.
- 2 **Adjacency Lists**: for each node \mathbf{u} , $\mathbf{Out}(\mathbf{u})$ (also referred to as $\mathbf{Adj}(\mathbf{u})$) and $\mathbf{In}(\mathbf{u})$ store out-going edges and in-coming edges from \mathbf{u} .

Default representation is adjacency lists. Concrete representation discussed previously for undirected graphs easily extends to directed graphs.

Directed Connectivity

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

- 1 A **(directed) path** is a sequence of *distinct* vertices v_1, v_2, \dots, v_k such that $(v_i, v_{i+1}) \in E$ for $1 \leq i \leq k - 1$. The length of the path is $k - 1$ and the path is from v_1 to v_k . By convention, a single node u is a path of length 0 .

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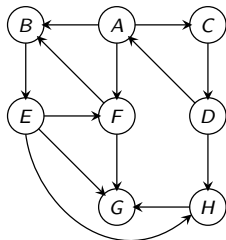
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- 4 Let **rch**(u) be the set of all vertices reachable from u .

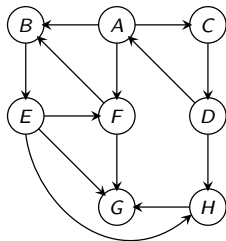
Connectivity contd

Asymmetry: **D** can reach **B** but **B** cannot reach **D**



Connectivity contd

Asymmetry: **D** can reach **B** but **B** cannot reach **D**



Questions:

- 1 Is there a notion of connected components?
- 2 How do we understand connectivity in directed graphs?

Connectivity and Strong Connected Components

Definition

Given a directed graph G , u is strongly connected to v if u can reach v and v can reach u . In other words $v \in \text{rch}(u)$ and $u \in \text{rch}(v)$.

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Proposition

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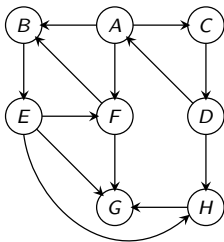
Proposition

\mathbf{C} is an equivalence relation, that is reflexive, symmetric and transitive.

Equivalence classes of \mathbf{C} : *strong connected components* of \mathbf{G} .
They *partition* the vertices of \mathbf{G} .

$\text{SCC}(\mathbf{u})$: strongly connected component containing \mathbf{u} .

Strongly Connected Components: Example



Directed Graph Connectivity Problems

- 1 Given \mathbf{G} and nodes \mathbf{u} and \mathbf{v} , can \mathbf{u} reach \mathbf{v} ?
- 2 Given \mathbf{G} and \mathbf{u} , compute $\text{rch}(\mathbf{u})$.
- 3 Given \mathbf{G} and \mathbf{u} , compute all \mathbf{v} that can reach \mathbf{u} , that is all \mathbf{v} such that $\mathbf{u} \in \text{rch}(\mathbf{v})$.
- 4 Find the strongly connected component containing node \mathbf{u} , that is $\text{SCC}(\mathbf{u})$.
- 5 Is \mathbf{G} strongly connected (a single strong component)?
- 6 Compute *all* strongly connected components of \mathbf{G} .

Directed Graph Connectivity Problems

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First five problems can be solved in $\mathbf{O}(n + m)$ time by adapting **BFS/DFS** to directed graphs. The last one requires a clever **DFS** based algorithm.

DFS in Directed Graphs

DFS(G)

Mark all nodes u as unvisited

T is set to \emptyset

$time = 0$

while there is an unvisited node u **do**

DFS(u)

Output T

DFS(u)

Mark u as visited

$pre(u) = ++time$

for each edge (u, v) in $Out(u)$ **do**

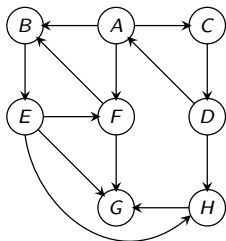
if v is not marked

 add edge (u, v) to T

DFS(v)

$post(u) = ++time$

Example



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Generalizing ideas from undirected graphs:

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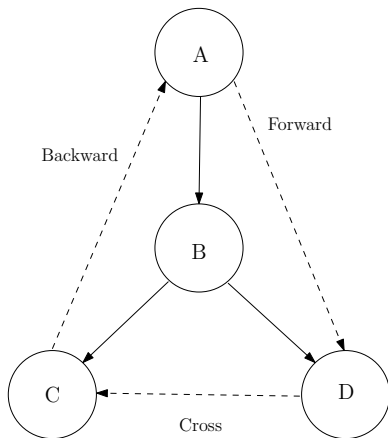
Note: Not obvious whether **DFS(G)** is useful in dir graphs but it is.

DFS Tree

Edges of **G** can be classified with respect to the **DFS** tree **T** as:

- 1 **Tree edges** that belong to **T**
- 2 A **forward edge** is a non-tree edges (x, y) such that $\text{pre}(x) < \text{pre}(y) < \text{post}(y) < \text{post}(x)$.
- 3 A **backward edge** is a non-tree edge (y, x) such that $\text{pre}(x) < \text{pre}(y) < \text{post}(y) < \text{post}(x)$.
- 4 A **cross edge** is a non-tree edges (x, y) such that the intervals $[\text{pre}(x), \text{post}(x)]$ and $[\text{pre}(y), \text{post}(y)]$ are disjoint.

Types of Edges



Directed Graph Connectivity Problems

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Algorithms via DFS- I

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- 2 Given \mathbf{G} and \mathbf{u} , compute $\text{rch}(\mathbf{u})$.

Use $\text{DFS}(\mathbf{G}, \mathbf{u})$ to compute $\text{rch}(\mathbf{u})$ in $\mathbf{O}(n + m)$ time.

Algorithms via DFS- II

- 1 Given \mathbf{G} and \mathbf{u} , compute all \mathbf{v} that can reach \mathbf{u} , that is all \mathbf{v} such that $\mathbf{u} \in \text{rch}(\mathbf{v})$.

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Definition (Reverse graph.)

Given $\mathbf{G} = (\mathbf{V}, \mathbf{E})$, \mathbf{G}^{rev} is the graph with edge directions reversed
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Compute $\text{rch}(\mathbf{u})$ in \mathbf{G}^{rev} !

- 1 **Correctness:** exercise
- 2 **Running time:** $\mathbf{O}(n + m)$ to obtain \mathbf{G}^{rev} from \mathbf{G} and $\mathbf{O}(n + m)$ time to compute $\text{rch}(\mathbf{u})$ via **DFS**. If both $\text{Out}(\mathbf{v})$ and $\text{In}(\mathbf{v})$ are available at each \mathbf{v} then no need to explicitly compute \mathbf{G}^{rev} . Can do **DFS**(\mathbf{u}) in \mathbf{G}^{rev} implicitly.

Algorithms via DFS- III

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Hence, $SCC(\mathbf{G}, \mathbf{u})$ can be computed with two **DFS**es, one in \mathbf{G} and the other in \mathbf{G}^{rev} . Total $O(n + m)$ time.

Why can $\text{rch}(\mathbf{G}, \mathbf{u}) \cap \text{rch}(\mathbf{G}^{\text{rev}}, \mathbf{u})$ be done in $O(n)$ time?

Algorithms via DFS- IV

- 1 Is G strongly connected?

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Pick arbitrary vertex u . Check if $SC(G, u) = V$.

Algorithms via DFS- V

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```
for each vertex  $u \in V$  do  
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```

Algorithms via DFS- V

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```
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```

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

Algorithms via DFS- V

- 1 Find *all* strongly connected components of G .

```
for each vertex  $u \in V$  do  
  find  $SC(G, u)$ 
```

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

Q: Can we do it in $O(n + m)$ time?