CSI - 3105 Design & Analysis of Algorithms Course 5

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Fall 2019

Exercise #10 (Chapter 1)

Write an algorithm that finds the m smallest numbers in a list of n numbers (where $1 \le m \le n$) and analyze its running time in the worst case.

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What do you think of the following solution?

Sort the list of numbers using Merge Sort. Then scan the list and return the first m numbers.

Sorting using Merge Sort takes $O(n \log(n))$ time and scanning the list takes O(n) time. So in total, this algorithm takes $O(n \log(n)) + O(n) = O(n \log(n))$ time.



A Faster Solution

What do you think of the following solution?

Let L be the list of numbers. Find the m-th smallest element of L using the Select Algorithm with Select(L, m). Let this m-th smallest element be x. Scan the list and return all numbers that are $\leq x$.

A call to the Select Algorithm takes O(n) time and scanning the list takes O(n) time. So in total, this algorithm takes O(n) + O(n) = O(n) time.

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Nothing in the question says that the numbers are all different. So we cannot assume that they are!

A Faster Solution

What do you think of the following solution?

Let L be the list of numbers. Find the m-th smallest element of L using the Select Algorithm with Select(L,m). Let this m-th smallest element be x. Initialize a counter cnt to 0. Scan the list and return all numbers that are < x. Every time you return a number, increment cnt by 1. When you are done scanning the list, return the number x exactly m-cnt times.

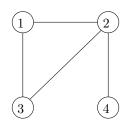
A call to the Select Algorithm takes O(n) time and scanning the list takes O(n) time. Returning the number x exactly m-cnt times takes O(n) time since $m-cnt \le n$. So in total, this algorithm takes O(n)+O(n)+O(n)=O(n) time.



Chapter 3: Graph Algorithms

A graph G is made of a set V of vertices (or nodes) together with a set E of edges. We write G = (V, E).

A graph is *undirected* if each edge in E is a pair $\{u, v\}$, where $u, v \in V$ and $u \neq v$.



(5)

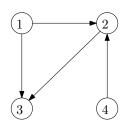
$$V = \{1, 2, 3, 4, 5\}$$

$$E = \{\{1,2\},\{1,3\},\{2,3\},\{2,4\}\}$$

Chapter 3: Graph Algorithms

A graph G is made of a set V of vertices (or nodes) together with a set E of edges. We write G = (V, E).

A graph is *directed* is each edge in E is an **ordered** pair (u, v), where $u, v \in V$ and $u \neq v$.



 $\overbrace{5}$

$$V = \{1, 2, 3, 4, 5\}$$

$$E = \{(1,2), (1,3), (2,3), (4,2)\}$$

Examples:

- A road map
- Facebook. Vertices are users. There is an edge $\{A, B\}$ if and only if A and B are "friends".
- WWW. Vertices are web pages. There is a directed edge (A, B) if and only if A has a link to B.

Examples:

Scheduling exams!

V = set of all courses taught this term

There is an edge $\{u, v\}$ if and only if there is at least one student taking both courses u and v.

Let C be the minimum number of colors needed such that

- each vertex gets one color.
- For each edge $\{u, v\}$, u and v have different colors.

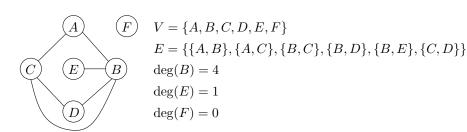
Then we can make an exam schedule with C time slots 1, 2, ..., C: all vertices (i.e., courses) with color i have their exam in time slot i. In this way, there are no conflicts!

But computing C is very difficult...

In a graph G = (V, E), two vertices $u, v \in V$ are adjacent if there is an edge between u and v.

A vertex $u \in V$ is *incident* to an edge $e \in E$ if one of the two vertices of e is u.

The *degree* of a vertex $u \in V$ is equal to the number of edges incident to u.





9 / 17

When the graph is oriented, the *outdegree* of a vertex $u \in V$ is equal to the number of edges $e \in E$ such that u is the starting point of e. The *indegree* of a vertex $u \in V$ is equal to the number of edges $e \in E$ such that u is the endpoint of e.

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Theorem (Handshaking Lemma)

Let G = (V, E) be a graph. then

$$\sum_{u\in V}\deg(u)=2|E|.$$

Proof:



When the graph is oriented, the *outdegree* of a vertex $u \in V$ is equal to the number of edges $e \in E$ such that u is the starting point of e. The *indegree* of a vertex $u \in V$ is equal to the number of edges $e \in E$ such that u is the endpoint of e.

Theorem (Handshaking Lemma)

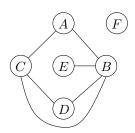
Let G = (V, E) be a graph. then

$$\sum_{u\in V} \deg(u) = 2|E|.$$

PROOF: Each edge is counted twice!



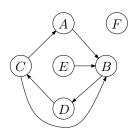
$$G = (V, E)$$
 $V = \{v_1, v_2, ..., v_n\}$



	A	B	C	D	E	F
A	0	1	1	0	0	0
B	1	0	1	1	1	0
C	1	1	0	1	0	0
D	0	1	1	0	0	0
E	0	1	0	0	0	0
F	0	0	0	D 0 1 1 0 0	0	0

Adjacency matrix

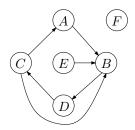
$$G = (V, E)$$
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	A	B	C	D	E	F
A	0	1	0	0	0	0
B	0	0	0	1	0	0
C	1	1	0	0	0	0
D	0	0	1	0	0	0
E	0	1	0	0	0	0
F	0	0	0	D 0 1 0 0 0	0	0

Adjacency matrix

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 $V = \{v_1, v_2, ..., v_n\}$



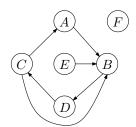
	A	B	C	D	E	F
A	0	1	0	0	0	0
B	0	0	0	1	0	0
C	1	1	0	0	0	0
D	0	0	1	0	0	0
E	0	1	0	0	0	0
F	0	0	0	D 0 1 0 0 0	0	0

Adjacency matrix

Advantage:

• In O(1) time, we can test if there is an edge between two given vertices.

$$G = (V, E)$$
 $V = \{v_1, v_2, ..., v_n\}$



	A	B	C	D	E	F
A	0	1	0	0	0	0
B	0	0	0	1	0	0
C	1	1	0	0	0	0
D	0	0	1	0	0	0
E	0	1	0	0	0	0
F	0	0	0	D 0 1 0 0 0	0	0

Adjacency matrix

Advantage:

• In O(1) time, we can test if there is an edge between two given vertices.

Disadvantage:

- Uses $\Theta(n^2)$ space for any graph.
- Finding all neighbours of a given vertex takes O(n) time.

$$G = (V, E) \qquad V = \{v_1, v_2, ..., v_n\}$$

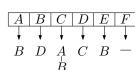
$$A \qquad F \qquad \begin{array}{c|cccc} A & B & C & D & E & F \\ \hline & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow \\ & B & A & A & B & B & - \\ & C & C & B & C & & \text{Adjacency list} \\ & D & D & & E & & \end{array}$$
Adjacency list

$$G = (V, E)$$

$$A \qquad F$$

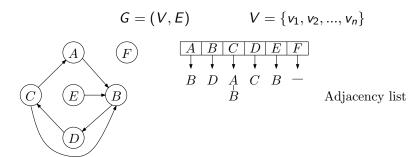
$$A \qquad B$$

$$B \qquad D$$



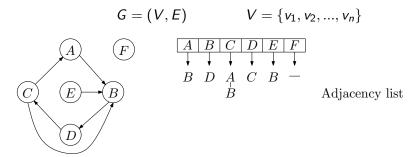
 $V = \{v_1, v_2, ..., v_n\}$

Adjacency list



Advantage:

- Uses $\Theta(|V| + |E|)$ space.
- Finding all neighbours of a vertex $u \in V$ takes $O(1 + \deg(u))$ time.



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- Uses $\Theta(|V| + |E|)$ space.
- Finding all neighbours of a vertex $u \in V$ takes $O(1 + \deg(u))$ time.

Disadvantage:

• Testing if $\{u, v\}$ (or (u, v)) is an edge takes $O(1 + \deg(u))$ time.

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Section 3.1: Exploring an Undirected Graph

Let G = (V, E) be an undirected graph.

Task: Find all vertices that can be reached from a given vertex $v \in V$.

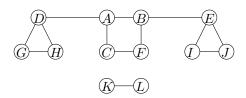
Algorithm explore(v)

8: postvisit(v)

```
    visited(v) = TRUE
    previsit(v)
    for each edge {u, v} ∈ E do
    if visited(u) = FALSE then
    call explore(u)
    end if
    end for
```

// See later

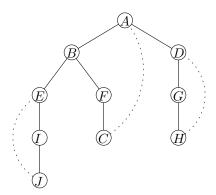
// See later



Run explore(A). In the for-loop, use alphabetical order (i.e., adjacency lists are sorted alphabetically). Each time an edge $\{u, v\}$ is traversed (because visited(u) = FALSE): u is discovered for the first time.

- Draw $\{u, v\}$ as a solid edge.
- All other edges: dotted.

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The solid edges form a *tree* (connected, no cycle). These edges are called *tree edges*. The dotted edges are called *back edges*.

Why is algorithm explore(v) correct?

First, how can we explain that it always terminates?

Algorithm explore(v)1: visited(v) = TRUE

```
2: previsit(v)
3: for each edge {u, v} ∈ E do
4: if visited(u) = FALSE then
5: call explore(u)
6: end if
```

7: end for

8: postvisit(v)

// See later

// See later

Why is algorithm explore(v) correct?

First, how can we explain that it always terminates?

The number of vertices u such that "visited(u) = FALSE" decreases in each recursive call.

```
Algorithm explore(v)
```

```
    visited(v) = TRUE
    previsit(v)
    for each edge {u, v} ∈ E do
    if visited(u) = FALSE then
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    end if
    end for
```

8: postvisit(v)

// See later

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How can we explain that it does visit all vertices that are reachable from v?

```
\overline{\mathbf{Algorithm}} explore(v)
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    visited(v) = TRUE
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```

// See later

// See later

8: postvisit(v)

How can we explain that it does visit all vertices that are reachable from v?

Lemma

Assume that, initially, visited(u) = FALSE. After explore(v) has terminated,

$$visited(u) = TRUE$$

there is a path from v to u.

Algorithm explore(v)

```
1: visited(v) = TRUE
2: previsit(v)
3: for each edge \{u, v\} \in E do
      if visited(u) = FALSE then
4:
5:
         call explore(u)
      end if
7: end for
```

// See later

// See later

8: postvisit(v)