Administrivia, Introduction, Graph basics and DFS

Lecture 1
January 15, 2013
The word “algorithm” comes from...

Muhammad ibn Musa al-Khwarizmi
780-850 AD
The word “algebra” is taken from the title of one of his books.
Part I

Administrivia
Instructional Staff

1. Instructor:
   - Sariel Har-Peled (sariel)
   - Alexandra Kolla (akolla)

2. Teaching Assistants:
   - Danyal Khashabi (khashab2)
   - Madan Vivek (vmadan2)
   - Hai Wang (hwang202)
   - Subhro Roy (sroy9)

3. Office hours: See course webpage

4. Email: See course webpage
Online resources

1. **Webpage:** courses.engr.illinois.edu/cs473/sp2013/
   General information, homeworks, etc.

2. **Moodle:**
   https://learn.illinois.edu/course/view.php?id=1647
   Quizzes, solutions to homeworks.

3. **Online questions/announcements:** Piazza
   https://piazza.com/#spring2013/cs473
   Online discussions, etc.
Prerequisites: CS 173 (discrete math), CS 225 (data structures) and CS 373 (theory of computation)

Recommended books:
1. Algorithms by Dasgupta, Papadimitriou & Vazirani.
   Available online for free!
2. Algorithm Design by Kleinberg & Tardos

Lecture notes: Available on the web-page after every class.

Additional References
1. Previous class notes of Jeff Erickson, Sariel HarPeled and the instructor.
2. Introduction to Algorithms: Cormen, Leiserson, Rivest, Stein.
Prerequisites

1. Asymptotic notation: $O()$, $\Omega()$, $o()$.
2. Discrete Structures: sets, functions, relations, equivalence classes, partial orders, trees, graphs
3. Logic: predicate logic, boolean algebra
4. Proofs: by induction, by contradiction
5. Basic sums and recurrences: sum of a geometric series, unrolling of recurrences, basic calculus
6. Data Structures: arrays, multi-dimensional arrays, linked lists, trees, balanced search trees, heaps
7. Abstract Data Types: lists, stacks, queues, dictionaries, priority queues
8. Algorithms: sorting (merge, quick, insertion), pre/post/in order traversal of trees, depth/breadth first search of trees (maybe graphs)
9. Basic analysis of algorithms: loops and nested loops, deriving recurrences from a recursive program
10. Concepts from Theory of Computation: languages, automata, Turing machine, undecidability, non-determinism
11. Programming: in some general purpose language
12. Elementary Discrete Probability: event, random variable, independence
13. Mathematical maturity
Grading Policy: Overview

1. Attendance/clickers: 5%
2. Quizzes: 5%
3. Homeworks: 20%
4. Midterms: 40% (2 × 20%)
5. Finals: 30% (covers the full course content)
Homeworks

1. One quiz every week: Due by midnight on Sunday.
2. One homework every week: Assigned on Tuesday and due the following Monday at noon.
3. Submit online only!
4. Homeworks can be worked on in groups of up to 3 and each group submits one written solution (except Homework 0).
   1. Short quiz-style questions to be answered individually on Moodle.
5. Groups can be changed a few times only
6. Unlike previous years no oral homework this semester due to large enrollment.
More on Homeworks

1. No extensions or late homeworks accepted.
2. To compensate, the homework with the least score will be dropped in calculating the homework average.
3. **Important:** Read homework faq/instructions on website.
Discussion Sessions

1. 50min problem solving session led by TAs
2. Four sections all in SC 1214.
   1. Tuesday
      5–5:50pm,
      6–6:50pm.
   2. Wednesday
      4–4:50pm,
      5–5:50pm.
Advice

1. Attend lectures, please ask plenty of questions.
2. Clickers...
3. Attend discussion sessions.
4. Don’t skip homework and don’t copy homework solutions.
5. Study regularly and keep up with the course.
6. Ask for help promptly. Make use of office hours.
Homeworks

1. HW 0 is posted on the class website. Quiz 0 available
2. Quiz 0 due by Sunday Jan 20 midnight
   HW 0 due on Monday January 21 noon.
3. Online submission.
4. HW 0 to be submitted in individually.
Part II

Course Goals and Overview
Topics

1. Some fundamental algorithms
2. Broadly applicable techniques in algorithm design
   1. Understanding problem structure
   2. Brute force enumeration and backtrack search
   3. Reductions
   4. Recursion
      1. Divide and Conquer
      2. Dynamic Programming
   5. Greedy methods
   6. Network Flows and Linear/Integer Programming (optional)
3. Analysis techniques
   1. Correctness of algorithms via induction and other methods
   2. Recurrences
   3. Amortization and elementary potential functions
4. Polynomial-time Reductions, NP-Completeness, Heuristics
Goals

1. Algorithmic thinking
2. Learn/remember some basic tricks, algorithms, problems, ideas
3. Understand/appreciate limits of computation (intractability)
4. Appreciate the importance of algorithms in computer science and beyond (engineering, mathematics, natural sciences, social sciences, ...)
5. Have fun!!!
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Part III

Some Algorithmic Problems in the Real World
Directions to Chicago, IL
136 mi – about 2 hours 20 mins
Champaign, IL to Chicago, IL - Google Maps http://maps.google.com/maps?f=d&saddr=Champaign,+IL&da...
Shortest Paths - Paris to Berlin
Compression: reduce size for storage and transmission
Coding: add redundancy to protect against errors in storage and transmission

Efficient algorithms for compression/coding and decompressing/decoding part of most modern gadgets (computers, phones, music/video players ...)

Sariel, Alexandra (UIUC)
Web search: Google, Yahoo!, Microsoft, Ask, ...

Text search: Text editors (Emacs, Word, Browsers, ...)

Regular expression search: grep, egrep, emacs, Perl, Awk, compilers
RSA Crypto-system: generate key $n = pq$ where $p, q$ are primes

Primality: Given a number $N$, check if $N$ is a prime or composite.

Factoring: Given a composite number $N$, find a non-trivial factor
Public-Key Cryptography

Foundation of Electronic Commerce

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**Primality:** Given a number \( N \), check if \( N \) is a prime or composite.

**Factoring:** Given a composite number \( N \), find a non-trivial factor
Programming: Parsing and Debugging

[godavari: /temp/test] chekuri % gcc main.c

**Parsing:** Is main.c a syntactically valid C program?

**Debugging:** Will main.c go into an infinite loop on some input?

**Easier problem ???** Will main.c halt on the specific input 10?
Optimization

Find the cheapest of most profitable way to do things

1. Airline schedules - AA, Delta, ...
2. Vehicle routing - trucking and transportation (UPS, FedEx, Union Pacific, ...)
3. Network Design - AT&T, Sprint, Level3 ...

Linear and Integer programming problems
Part IV

Algorithm Design
Important Ingredients in Algorithm Design

1. What is the problem (really)?
   1. What is the input? How is it represented?
   2. What is the output?

2. What is the model of computation? What basic operations are allowed?

3. Algorithm design

4. Analysis of correctness, running time, space etc.

5. Algorithmic engineering: evaluating and understanding of algorithm’s performance in practice, performance tweaks, comparison with other algorithms etc. (Not covered in this course)
Primality testing

**Problem**

Given an integer $N > 0$, is $N$ a prime?

**SimpleAlgorithm:**

```
for i = 2 to $\lfloor \sqrt{N} \rfloor$ do
  if i divides N then
    return ‘‘COMPOSITE’’
return ‘‘PRIME’’
```

**Correctness?** If $N$ is composite, at least one factor in $\{2, \ldots, \sqrt{N}\}$

**Running time?** $O(\sqrt{N})$ divisions? Sub-linear in input size! Wrong!
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Primality testing

...Polynomial means... in input size

How many bits to represent \( N \) in binary? \( \lceil \log N \rceil \) bits.
Simple Algorithm takes \( \sqrt{N} = 2^{(\log N)/2} \) time.
Exponential in the input size \( n = \log N \).

1. Modern cryptography: binary numbers with 128, 256, 512 bits.
2. Simple Algorithm will take \( 2^{64}, 2^{128}, 2^{256} \) steps!
3. Fastest computer today about 3 petaFlops/sec: \( 3 \times 2^{50} \) floating point ops/sec.

Lesson:
Pay attention to representation size in analyzing efficiency of algorithms. Especially in \textit{number} problems.
How many bits to represent $N$ in binary? $\lceil \log N \rceil$ bits.

Simple Algorithm takes $\sqrt{N} = 2^{(\log N)/2}$ time. *Exponential* in the input size $n = \log N$.

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Efficient algorithms

So, is there an *efficient/good/effective* algorithm for primality?

**Question:**
What does efficiency mean?

In this class *efficiency* is broadly equated to *polynomial time.* 
$O(n), O(n \log n), O(n^2), O(n^3), O(n^{100}), \ldots$ where $n$ is size of the input.

Why? Is $n^{100}$ really efficient/practical? Etc.

Short answer: polynomial time is a robust, mathematically sound way to define efficiency. Has been useful for several decades.
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Lincoln’s tour

Circuit court - ride through counties staying a few days in each town.

Lincoln was a lawyer traveling with the Eighth Judicial Circuit.

Picture: travel during 1850.

Very close to optimal tour.

Might have been optimal at the time.
Solving TSP by a Computer

Is it hard?

1. \( n \) = number of cities.
2. \( n^2 \) = size of input.
3. Number of possible solutions is
   \[
   n \times (n - 1) \times (n - 2) \times \ldots \times 2 \times 1 = n!.
   \]
4. \( n! \) grows very quickly as \( n \) grows.
   \[\begin{align*}
   n = 10: & \quad n! \approx 3628800 \\
   n = 50: & \quad n! \approx 3 \times 10^{64} \\
   n = 100: & \quad n! \approx 9 \times 10^{157}
   \end{align*}\]
Solving TSP by a Computer

Fastest super computer can do (roughly)

\[2.5 \times 10^{15}\]

operations a second.

Assume: computer checks \(2.5 \times 10^{15}\) solutions every second, then...

1. \(n = 20 \implies 2\text{ hours.}\)
2. \(n = 25 \implies 200\text{ years.}\)
3. \(n = 37 \implies 2 \times 10^{20}\text{ years!!!}\)
What is a good algorithm?

Running time...

<table>
<thead>
<tr>
<th>Input size</th>
<th>(n^2) ops</th>
<th>(n^3) ops</th>
<th>(n^4) ops</th>
<th>(n!) ops</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>0 secs</td>
<td>0 secs</td>
<td>0 secs</td>
<td>0 secs</td>
</tr>
<tr>
<td>20</td>
<td>0 secs</td>
<td>0 secs</td>
<td>0 secs</td>
<td>0 secs</td>
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<td>30</td>
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<td>0 secs</td>
<td>0 secs</td>
<td>0 secs</td>
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<tr>
<td>100</td>
<td>0 secs</td>
<td>0 secs</td>
<td>0 secs</td>
<td>0 secs</td>
</tr>
<tr>
<td>8000</td>
<td>0 secs</td>
<td>0 secs</td>
<td>1 sec</td>
<td>3 \cdot 10^9 \text{ years}</td>
</tr>
<tr>
<td>16000</td>
<td>0 secs</td>
<td>0 secs</td>
<td>26 secs</td>
<td>never</td>
</tr>
<tr>
<td>32000</td>
<td>0 secs</td>
<td>0 secs</td>
<td>6 mins</td>
<td>never</td>
</tr>
<tr>
<td>64000</td>
<td>0 secs</td>
<td>0 secs</td>
<td>111 mins</td>
<td>never</td>
</tr>
<tr>
<td>200,000</td>
<td>0 secs</td>
<td>3 secs</td>
<td>7 days</td>
<td>never</td>
</tr>
<tr>
<td>2,000,000</td>
<td>0 secs</td>
<td>53 mins</td>
<td>202.943 years</td>
<td>never</td>
</tr>
<tr>
<td>(10^8)</td>
<td>4 secs</td>
<td>12.6839 years</td>
<td>(10^9) years</td>
<td>never</td>
</tr>
<tr>
<td>(10^9)</td>
<td>6 mins</td>
<td>12683.9 years</td>
<td>(10^{13}) years</td>
<td>never</td>
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</table>
What is a good algorithm?

Running time...
Primes is in $\mathbb{P}$!

**Theorem (Agrawal-Kayal-Saxena’02)**

*There is a polynomial time algorithm for primality.*

First polynomial time algorithm for testing primality. Running time is $O(\log^{12} N)$ further improved to about $O(\log^6 N)$ by others. In terms of input size $n = \log N$, time is $O(n^6)$.


Neeraj Kayal and Nitin Saxena were undergraduates at IIT-Kanpur!
What about before 2002?

Primality testing a key part of cryptography. What was the algorithm being used before 2002?

Miller-Rabin *randomized* algorithm:

1. runs in polynomial time: $O(\log^3 N)$ time
2. if $N$ is prime correctly says “yes”.
3. if $N$ is composite it says “yes” with probability at most $1/2^{100}$ (can be reduced further at the expense of more running time).

Based on Fermat’s little theorem and some basic number theory.
Modern public-key cryptography based on RSA (Rivest-Shamir-Adelman) system.

Relies on the difficulty of factoring a composite number into its prime factors.

There is a polynomial time algorithm that decides whether a given number $N$ is prime or not (hence composite or not) but no known polynomial time algorithm to factor a given number.

**Lesson**

Intractability can be useful!
Factoring

1. Modern public-key cryptography based on RSA (Rivest-Shamir-Adelman) system.

2. Relies on the difficulty of factoring a composite number into its prime factors.

3. There is a polynomial time algorithm that decides whether a given number $N$ is prime or not (hence composite or not) but no known polynomial time algorithm to factor a given number.

Lesson

Intractability can be useful!
Three variants of problems.

1. **Decision problem**: answer is yes or no.
   
   **Example**: Given integer $N$, is it a composite number?

2. **Search problem**: answer is a feasible solution if it exists.
   
   **Example**: Given integer $N$, if $N$ is composite output a non-trivial factor $p$ of $N$.

3. **Optimization problem**: answer is the *best* feasible solution (if one exists).
   
   **Example**: Given integer $N$, if $N$ is composite output the *smallest* non-trivial factor $p$ of $N$.

For a given underlying problem:

Optimization $\geq$ Search $\geq$ Decision
Three variants of problems.

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For a given underlying problem:

Optimization $\geq$ Search $\geq$ Decision
Quantum Computing

**Theorem (Shor’1994)**

*There is a polynomial time algorithm for factoring on a quantum computer.*

RSA and current commercial cryptographic systems can be broken if a quantum computer can be built!

**Lesson**

Pay attention to the model of computation.
Theorem (Shor’1994)

*There is a polynomial time algorithm for factoring on a quantum computer.*

RSA and current commercial cryptographic systems can be broken if a quantum computer can be built!

**Lesson**

Pay attention to the model of computation.
Many many different problems.

1. Adding two numbers: efficient and simple algorithm
2. Sorting: efficient and not too difficult to design algorithm
3. Primality testing: simple and basic problem, took a long time to find efficient algorithm
5. Halting problem: important problem in practice, undecidable!
Problem  Given two \( n \)-digit numbers \( x \) and \( y \), compute their product.

Grade School Multiplication

Compute “partial product” by multiplying each digit of \( y \) with \( x \) and adding the partial products.

\[
\begin{array}{c}
3141 \\
\times 2718 \\
\hline
25128 \\
3141 \\
\hline
21987 \\
6282 \\
\hline
8537238
\end{array}
\]
Time analysis of grade school multiplication

1. Each partial product: $\Theta(n)$ time
2. Number of partial products: $\leq n$
3. Adding partial products: $n$ additions each $\Theta(n)$ (Why?)
4. Total time: $\Theta(n^2)$
5. Is there a faster way?
Fast Multiplication

Best known algorithm: \( O(n \log n \cdot 2^{O(\log^* n)}) \) time [Furer 2008]

Previous best time: \( O(n \log n \log \log n) \) [Schonhage-Strassen 1971]

**Conjecture:** there exists and \( O(n \log n) \) time algorithm

We don’t fully understand multiplication!
Computation and algorithm design is non-trivial!
Fast Multiplication

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**Conjecture:** there exists and $O(n \log n)$ time algorithm

We don’t fully understand multiplication!
Computation and algorithm design is non-trivial!
Algorithm design requires a mix of skill, experience, mathematical background/maturity and ingenuity.

Approach in this class and many others:

1. Improve skills by showing various tools in the abstract and with concrete examples
2. Improve experience by giving many problems to solve
3. Motivate and inspire
4. Creativity: you are on your own!
What model of computation do we use?

Turing Machine?
What model of computation do we use?

Turing Machine?
Turing Machines: Recap

1. Infinite tape
2. Finite state control
3. Input at beginning of tape
4. Special tape letter "blank" ⊣
5. Head can move only one cell to left or right
Turing Machines

1. Basic unit of data is a bit (or a single character from a finite alphabet)
2. Algorithm is the finite control
3. Time is number of steps/head moves

Pros and Cons:
1. Theoretically sound, robust and simple model that underpins computational complexity.
2. Polynomial time equivalent to any reasonable “real” computer: Church-Turing thesis
3. Too low-level and cumbersome, does not model actual computers for many realistic settings
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“Real” Computers vs Turing Machines

How do “real” computers differ from TMs?

1. random access to memory
2. pointers
3. arithmetic operations (addition, subtraction, multiplication, division) in constant time

How do they do it?

1. basic data type is a word: currently 64 bits
2. arithmetic on words are basic instructions of computer
3. memory requirements assumed to be \( \leq 2^{64} \) which allows for pointers and indirect addressing as well as random access
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Unit-Cost RAM Model

Informal description:

1. Basic data type is an integer/ floating point number
2. Numbers in input fit in a word
3. Arithmetic/comparison operations on words take constant time
4. Arrays allow random access (constant time to access \( A[i] \))
5. Pointer based data structures via storing addresses in a word
Example

**Example**

**Sorting**: input is an array of \( n \) numbers

1. input size is \( n \) (ignore the bits in each number),
2. comparing two numbers takes \( O(1) \) time,
3. random access to array elements,
4. addition of indices takes constant time,
5. basic arithmetic operations take constant time,
6. reading/writing one word from/to memory takes constant time.

We will usually not allow (or be careful about allowing):

1. bitwise operations (and, or, xor, shift, etc).
2. floor function.
3. limit word size (usually assume unbounded word size).
Caveats of RAM Model

Unit-Cost RAM model is applicable in wide variety of settings in practice. However it is not a proper model in several important situations so one has to be careful.

1. For some problems such as basic arithmetic computation, unit-cost model makes no sense. Examples: multiplication of two \( n \)-digit numbers, primality etc.

2. Input data is very large and does not satisfy the assumptions that individual numbers fit into a word or that total memory is bounded by \( 2^k \) where \( k \) is word length.

3. Assumptions valid only for certain type of algorithms that do not create large numbers from initial data. For example, exponentiation creates very big numbers from initial numbers.
Models used in class

In this course:

1. Assume unit-cost **RAM** by default.
2. We will explicitly point out where unit-cost RAM is not applicable for the problem at hand.
Part V

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) etc etc.

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
An undirected (simple) graph \( G = (V, E) \) is a 2-tuple:

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

Example

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

Notation

An edge in an undirected graph 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

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

- By sorting each list, one can achieve $O(\log n)$ time
- 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.
Connectivity

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

1. A path is a sequence of distinct vertices $v_1, v_2, \ldots, v_k$ 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 $v_k$.

2. A cycle is a sequence of distinct vertices $v_1, v_2, \ldots, v_k$ such that $\{v_i, v_{i+1}\} \in E$ for $1 \leq i \leq k - 1$ and $\{v_1, v_k\} \in E$.

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

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.

\[ \text{Connected Graphs} \]

\[
\{1, 2, 3, 4, 5, 6, 7, 8\} \quad \text{and} \quad \{9, 10\}
\]
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 $G$.

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

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

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

**Explore**($u$):

- Initialize $S = \{u\}$
- **while** there is an edge $(x, y)$ with $x \in S$ and $y \not\in S$ **do**
  - add $y$ to $S$

**Proposition**

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

Running time: depends on implementation

1. Breadth First Search (**BFS**): use queue data structure
2. Depth First Search (**DFS**): use stack data structure
3. Review CS 225 material!
Basic Graph Search

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

**Explore**($u$):

- Initialize $S = \{u\}$
- **while** there is an edge $(x, y)$ with $x \in S$ and $y \notin S$ **do**
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**Proposition**

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

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Basic Graph Search

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

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Running time: depends on implementation

1. Breadth First Search (**BFS**): use queue data structure
2. Depth First Search (**DFS**): use stack data structure
3. Review CS 225 material!
Part VI

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 ($O(m + n)$) time 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 $u$ as unvisited

while there is an unvisited node $u$ do

**DFS(u)**

Mark $u$ as visited

for each edge $(u,v)$ in $Ajd(u)$ do

if $v$ is not marked

**DFS(v)**

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

The set of connected components of a graph is the set \( \{ \text{component} \} \), where \( u \in V \).

The connected components in the above graph are \( \{1, 2, 3, 4, 5, 6\} \) and \( \{7, 8, 9, 10\} \).

A graph is said to be connected when it has exactly one connected component. In other words, every pair of vertices in the graph are connected.
DFS(G)

Mark all nodes as unvisited

$T$ is set to $\emptyset$

while $\exists$ unvisited node $u$ do

$\text{DFS}(u)$

Output $T$

DFS(u)

Mark $u$ as visited

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

if $v$ is not marked

add $uv$ to $T$

$\text{DFS}(v)$

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

1. tree edge: belongs to $T$

2. non-tree edge: does not belong to $T$
**DFS (G)**
Mark all nodes as unvisited
\( T \) is set to \( \emptyset \)
while \( \exists \) unvisited node \( u \) do
\( \text{DFS}(u) \)
Output \( T \)

**DFS (u)**
Mark \( u \) as visited
for \( uv \) in \( \text{Adj}(u) \) do
if \( v \) is not marked
add \( uv \) to \( T \)
\( \text{DFS}(v) \)

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

**Question:** Why are there no cross-edges?
DFS with Visit Times

Keep track of when nodes are visited.

\[
\text{DFS}(G) \\
\text{for all } u \in V(G) \text{ do} \\
\quad \text{Mark } u \text{ as unvisited} \\
\quad T \text{ is set to } \emptyset \\
\quad \text{time} = 0 \\
\quad \text{while } \exists \text{ unvisited } u \text{ do} \\
\quad \quad \text{DFS}(u) \\
\quad \text{Output } T
\]

\[
\text{DFS}(u) \\
\quad \text{Mark } u \text{ as visited} \\
\quad \text{pre}(u) = ++\text{time} \\
\quad \text{for each } uv \text{ in } \text{Out}(u) \text{ do} \\
\quad \quad \text{if } v \text{ is not marked then} \\
\quad \quad \quad \text{add edge } uv \text{ to } T \\
\quad \quad \text{DFS}(v) \\
\quad \text{post}(u) = ++\text{time}
\]
Example

Connected Graphs

1. The set of connected components of a graph is the set \( \{ \text{component} \} | u \in V \} \).

2. The connected components in the above graph are \( \{1, 2, 3, 4, 5, 6, 7, 8\} \) and \( \{9, 10\} \).

3. A graph is said to be connected when it has exactly one connected component. In other words, every pair of vertices in the graph are connected.

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**pro 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.*

**Proof.**

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

**pre and post numbers useful in several applications of DFS—soon!**
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pre and post numbers useful in several applications of DFS—soon!
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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- Assume without loss of generality that \(\text{pre}(u) < \text{pre}(v)\). Then v visited after u.
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pre and post numbers useful in several applications of DFS- soon!
Part VII

Directed Graphs and Decomposition
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 $y$. 
Graph $G = (V, E)$ with $n$ vertices and $m$ edges:


2. **Adjacency Lists**: for each node $u$, $\text{Out}(u)$ (also referred to as $\text{Adj}(u)$) and $\text{In}(u)$ store out-going edges and in-coming edges from $u$.

Default representation is adjacency lists.
Directed Connectivity

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

1. A **(directed) path** is a sequence of distinct vertices $v_1, v_2, \ldots, v_k$ 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 $v_k$.

2. A **cycle** is a sequence of distinct vertices $v_1, v_2, \ldots, v_k$ such that $(v_i, v_{i+1}) \in E$ for $1 \leq i \leq k - 1$ and $(v_k, v_1) \in E$.

3. A vertex $u$ can reach $v$ if there is a path from $u$ to $v$. Alternatively $v$ can be reached from $u$.

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


Asymmetricity: \textbf{A} can reach \textbf{B} but \textbf{B} cannot reach \textbf{A}

Questions:
1. Is there a notion of connected components?
2. How do we understand connectivity in directed graphs?
Asymmetricity: A can reach B but B cannot reach A

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 $v \in rch(u)$ and $u \in rch(v)$.

Define relation $C$ where $uCv$ if $u$ is (strongly) connected to $v$.

Proposition

$C$ is an equivalence relation, that is reflexive, symmetric and transitive.

Equivalence classes of $C$: strong connected components of $G$. They partition the vertices of $G$.

$SCC(u)$: strongly connected component containing $u$. 
Connectivity and Strong Connected Components

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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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$C$ is an equivalence relation, that is reflexive, symmetric and transitive.

Equivalence classes of $C$: strong connected components of $G$. They *partition* the vertices of $G$.

$\text{SCC}(u)$: strongly connected component containing $u$. 
Strongly Connected Components: Example

A directed graph (also called a digraph) is $G = (V, E)$, where $V$ is a set of vertices or nodes and $E \subseteq V \times V$ is set of ordered pairs of vertices called edges.
Directed Graph Connectivity Problems

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

First four problems can be solved in $O(n + m)$ time by adapting BFS/DFS to directed graphs. The last one requires a clever DFS based algorithm.
Directed Graph Connectivity Problems

1. Given $G$ and nodes $u$ and $v$, can $u$ reach $v$?
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DFS in Directed Graphs

**DFS(G)**

- Mark all nodes \( u \) as unvisited
- \( T \) is set to \( \emptyset \)
- \( \text{time} = 0 \)
- while there is an unvisited node \( u \) do
  - DFS\(_u\)
  - Output \( T \)

**DFS(u)**

- Mark \( u \) as visited
- \( \text{pre}(u) = ++\text{time} \)
- for each edge \((u, v)\) in Out\(_u\) do
  - if \( v \) is not marked
    - add edge \((u, v)\) to \( T \)
    - DFS\(_v\)
  - \( \text{post}(u) = ++\text{time} \)
Example

iClicker Question: What are the strong connected components?

(A) \{A, B, C, D, E, F, G, H\}
(B) \{A, B, C\} , \{D, E, F\} , \{G\} , \{H\}
(C) \{A, C, D\} , \{B, E, F, H\} , \{G\}
(D) \{A, C, D\} , \{B, E, F\} , \{G\} , \{H\}
(E) \{A, C, D\} , \{B, E, F\} , \{G, H\}
A directed graph (also called a digraph) is $G = (V, E)$, where $V$ is a set of vertices or nodes and $E \subseteq V \times V$ is a set of ordered pairs of vertices called edges.
DFS Properties

Generalizing ideas from undirected graphs:

1. **DFS**(u) outputs a directed out-tree \( T \) rooted at \( u \)
2. A vertex \( v \) is in \( T \) if and only if \( v \in rch(u) \)
3. For any two vertices \( x, y \) the intervals \([\text{pre}(x), \text{post}(x)]\) and \([\text{pre}(y), \text{post}(y)]\) are either disjoint or one is contained in the other.
4. The running time of **DFS**(u) is \( O(k) \) where \( k = \sum_{v \in rch(u)} |\text{Adj}(v)| \) plus the time to initialize the Mark array.
5. **DFS**(G) takes \( O(m + n) \) time. Edges in \( T \) form a disjoint collection of out-trees. Output of **DFS**(G) depends on the order in which vertices are considered.
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 $(x, y)$ such that $\text{pre}(y) < \text{pre}(x) < \text{post}(x) < \text{post}(y)$.

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

A
B
C D
Cross
Forward
Backward

Backward

Forward

Cross
Directed Graph Connectivity Problems

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

1. Given $G$ and nodes $u$ and $v$, can $u$ reach $v$?
2. Given $G$ and $u$, compute $rch(u)$.

Use $\text{DFS}(G, u)$ to compute $rch(u)$ in $O(n + m)$ time.
Given $G$ and $u$, compute all $v$ that can reach $u$, that is all $v$ such that $u \in rch(v)$.

**Definition (Reverse graph.)**

Given $G = (V, E)$, $G^{rev}$ is the graph with edge directions reversed $G^{rev} = (V, E')$ where $E' = \{(y, x) \mid (x, y) \in E\}$

Compute $rch(u)$ in $G^{rev}$!

1. **Correctness:** exercise

2. **Running time:** $O(n + m)$ to obtain $G^{rev}$ from $G$ and $O(n + m)$ time to compute $rch(u)$ via DFS. If both $Out(v)$ and $In(v)$ are available at each $v$ then no need to explicitly compute $G^{rev}$. Can do it $DFS(u)$ in $G^{rev}$ implicitly.
Given \( G \) and \( u \), compute all \( v \) that can reach \( u \), that is all \( v \) such that \( u \in \text{rch}(v) \).

**Definition (Reverse graph.)**

Given \( G = (V, E) \), \( G^{\text{rev}} \) is the graph with edge directions reversed \( G^{\text{rev}} = (V, E') \) where \( E' = \{(y, x) \mid (x, y) \in E\} \)

Compute \( \text{rch}(u) \) in \( G^{\text{rev}} \)!

1. **Correctness:** exercise

2. **Running time:** \( O(n + m) \) to obtain \( G^{\text{rev}} \) from \( G \) and \( O(n + m) \) time to compute \( \text{rch}(u) \) via \( \text{DFS} \). If both \( \text{Out}(v) \) and \( \text{In}(v) \) are available at each \( v \) then no need to explicitly compute \( G^{\text{rev}} \). Can do it \( \text{DFS}(u) \) in \( G^{\text{rev}} \) implicitly.
Given $G$ and $u$, compute all $v$ that can reach $u$, that is all $v$ such that $u \in rch(v)$.

**Definition (Reverse graph.)**

Given $G = (V, E)$, $G^{rev}$ is the graph with edge directions reversed $G^{rev} = (V, E')$ where $E' = \{(y, x) \mid (x, y) \in E\}$

Compute $rch(u)$ in $G^{rev}$!

1. **Correctness**: exercise

2. **Running time**: $O(n + m)$ to obtain $G^{rev}$ from $G$ and $O(n + m)$ time to compute $rch(u)$ via DFS. If both $Out(v)$ and $In(v)$ are available at each $v$ then no need to explicitly compute $G^{rev}$. Can do it $DFS(u)$ in $G^{rev}$ implicitly.
\( SC(G, u) = \{ v \mid u \text{ is strongly connected to } v \} \)

1. Find the strongly connected component containing node \( u \). That is, compute \( SCC(G, u) \).

\[ SCC(G, u) = rch(G, u) \cap rch(G^{rev}, u) \]

Hence, \( SCC(G, u) \) can be computed with two DFSes, one in \( G \) and the other in \( G^{rev} \). Total \( O(n + m) \) time.
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Hence, SCC(G, u) can be computed with two DFSes, one in G and the other in G^{rev}. Total \(O(n + m)\) time.
Is \( G \) strongly connected?

Pick arbitrary vertex \( u \). Check if \( SC(G, u) = V \).
Is $G$ strongly connected?

Pick arbitrary vertex $u$. Check if $SC(G, u) = V$. 
Find all strongly connected components of \( G \).

\[
\text{for each vertex } u \in V \text{ do find } SC(G, u)
\]

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

Q: Can we do it in \( O(n + m) \) time?
Find all strongly connected components of $G$.

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

Q: Can we do it in $O(n + m)$ time?
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?
Reading and Homework 0

Chapters 1 from Dasgupta et al book, Chapters 1-3 from Kleinberg-Tardos book.

Proving algorithms correct - Jeff Erickson’s notes (see link on website)