Computation CS 220 — Applied Discrete Mathematics

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09 Computation

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Analogy Machines

(Inspired partly by The Most Powerful Computers You've Never Heard Of.)





Co





Tide predictor







Integrator











Breaking Analogies

A Riddle:

There are three people in a room. Five people leave the room. How many people must enter the room for it to become empty?

Breaking Analogies

A Riddle:

There are three people in a room. Five people leave the room. How many people must enter the room for it to become empty?

A Wrong Answer:

I know algebra! 3-5+x=0, so x=2.

Thus if two people enter the room it will be empty. (Because the room currently has -2 people in it.)

Respect the analogy's **limits** and **tolerances**

Computing with misrepresented data is dangerous. Interpreting meaningless data is dangerous.

Towers of Analogies



real-world physics

mathematical models

numerical solvers/algos

floating-point arithmetic

Boolean logic: T, F, \land , \lor , etc

electrical properties of silicon

Mechanical Reasoning

In the last few modules, we learned about proof systems that let us **mechanically** (aka, **formally**) "reason"—that is, derive true propositions.

Is it possible that every mathematical question could automatically be "figured out" by some effective mechanical method?

This question, called the *Entscheidungsproblem* ("decision problem"), was famously posed by David Hilbert and Wilhelm Ackermann in 1928, after advances by George Boole (1847), Gottlob Frege (1879), and Giuseppe Peano (1888) distilled logic to a symbolic language governed by deterministic rules.

What criteria do we need for an effective mechanical method?

- Each method is concerned only with a single kind of problem. The set of problem inputs and outputs should be well understood.
- It must consist of a finite arrangement of instructions.
- Each instruction must have a **definite** interpretation. That is, it must work without the aid of *luck* or *human ingenuity*.
- When applied to any problem in its domain, it must terminate in a finite number of steps with the correct answer.

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This is what we now call an **algorithm**.

A problem with a yes/no answer is called a **decision problem**, and an algorithm that solves it is called a **decision procedure**.

What does an "arrangement of instructions" look like?

- ▶ Kurt Gödel and Jacques Herbrand (1933): general recursive functions
- Alonzo Church (1936): λ -calculus terms (CS450)
- Alan Turing (1936): machines consisting of a finite state automaton paired with an unbounded read-write "tape" (CS420)

Each model provides an encoding of \mathbb{N} , and each model defines its own notion of whether a $\mathbb{N} \to \mathbb{N}$ function is **computable**.

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By 1937, Church, Kleene, and Turing proved that all three models define the same set of computable functions.

The **Church-Turing Thesis**: That's what "effectively calculable" means, then.

Is there an algorithm that can decide every statement in mathematics?

Is there an algorithm that can decide every statement in mathematics? Kurt Gödel: No. (**Gödel's Incompleteness Theorem**)

This is a big, awesome topic. I recommend *Gödel, Escher, Bach* by Douglas Hofstadter.

Algorithms

Definition (Algorithm)

An **algorithm** is a finite arrangement of precise instructions for performing a computation or for deciding a question.

An algorithm must have a definite interpretation, and it must terminate in a finite number of steps with a correct answer.

Given a procedure:

- Is it an algorithm?
 - Does it always terminate?

Show loop termination using a **termination measure**, usually a non-negative integer value that decreases each iteration.

Does it produce the correct answer?

Often, loop correctness can be shown by a **loop invariant**, a proposition that is true on each iteration of the loop (and also when the loop exits).

- How efficient is it?
 - (Running time) How long does it take to run?
 - Space) How much memory does it require?

Algorithm 1 find-max(A[1..n])

Ensure: Returns max $\{A[1], \dots, A[n]\}$. 1: $m \leftarrow A[1]$ 2: $i \leftarrow 2$ 3: while $i \le n$ do 4: if m < A[i] then 5: $m \leftarrow A[i]$ 6: end if 7: $i \leftarrow i + 1$ 8: end while

9: **return** *m*

Termination

• termination measure: n - i + 1

Correctness

- Loop Invariant: $m = \max \{A[1], \dots, A[i-1]\}$
- Before loop starts (with i = 2): m = max {A[1]}
- When the loop exits (with i = n + 1): $m = \max \{A[1], \dots, A[n]\}$

Running time

▶ n-1 loop iterations

Algorithm 2 linear-search(x,A[1...n])

- 1: $i \leftarrow 1$
- 2: while $(i \le n \text{ and } x \ne A[i])$ do
- 3: $i \leftarrow i + 1$
- 4: end while
- 5: if $i \leq n$ then
- 6: $location \leftarrow i$
- 7: **else**
- 8: $location \leftarrow 0$
- 9: end if

10: return location

Returns *i* such that A[i] = x, or 0 if A does not contain x.

Termination

• termination measure: n - i + 1

Correctness

- Loop Invariant: x is not in A[1..(i - 1)]
- When the loop exits, either
 - i > n: Then by LI, x is not in A[1...n]. So return 0.
 - x = A[i].
 So return *i*.

Running time

up to n loop iterations

Example: Binary Search

- If the terms in a sequence are ordered, a binary search algorithm is more efficient than linear search.
- The binary search algorithm iteratively restricts the relevant search interval until it closes in on the position of the element to be located.
- **Example:** Binary search for the number 8.



Algorithms

Algorithm 3 binary-search $(x, A[1n])$ Require: A is sorted in ascending order				
2:	while $(i < j)$ do			
3:	$m \leftarrow \lfloor (i+j)/2 \rfloor$			
4:	if $x > A[m]$ then	1		
5:	$i \leftarrow m+1$	search $A[(m+1)j]$		
6:	else			
7:	$j \leftarrow m$	search $A[i . . m]$		
8:	end if			
9:	end while			
10:	if $(x = A[i])$ then			
11:	$location \leftarrow i$			
12:	else			
13:	$location \leftarrow 0$			
14:	end if			
15:	return location			

Returns *i* such that A[i] = x, or 0 if A does not contain x.

Termination

• termination measure: j - i

Correctness

- $\blacktriangleright A[i] \le x \le A[j]$
- Loop Invariant: if x is in A, then x is in A[i...j].
- When the loop exits, i = j, and either A[i] = x or x is not in A.
- Relies on precondition: sorted.

Running time

• up to $\left\lceil \log_2(n) \right\rceil$ loop iterations

How much time will an algorithm take?

On what hardware? What language and compiler? Is the VM "hot"? Are there other threads running? ...

We want* to abstract these details away.

Furthermore, we disregard the time usage for small inputs. Instead, we care how the time *grows* as a function of input size.

(We can also ask the same question about **space** usage.)

Computational Complexity

Input	Alg. A	Alg. B
n	5000n	$\lceil 1.1^n \rceil$
10	50,000	3
100	5×10^5	13,781
1000	5×10^6	$2.5 imes 10^{41}$
10^{6}	5×10^9	4.8×10^{41392}

This means that algorithm B cannot be used for large inputs, while running algorithm A is still feasible.

The growth class of the time (or space) usage as a function of input size is

- a mathematically stable basis for comparing algorithms, and
- often a useful approximation to an algorithm's performance in practice.

Counterpoint: Designing a Fast, Efficient, Cache-friendly Hash Table, Step by Step

Big-O Notation

Definition (Big-O)

Suppose $t, f : \mathbb{R}^+ \to \mathbb{R}^+$.

Then t is O(f) if there are constants c > 0and $n_0 > 0$ such that for all $n \ge n_0$:

 $t(n) \leq c \cdot f(n)$

That is, there is some multiple of f(n) that eventually bounds t(n) from above.

Example

▶
$$2n + 4$$
 is $O(n)$
▶ $n^2 + 2n + 1$ is $O(n^2)$



Useful Nomenclature

Function	Name	
c	Constant	
$\log n$	Logarithmic	
$\log^2 n$	Log-squared	
n	Linear	
$n\log n$	$n\log n$	
n^2	Quadratic	
n^3	Cubic	
2^n	Exponential	



Euclid's Algorithm

Recall:

Definition

The greatest common divisor (GCD) of a and b, where $a, b \in \mathbb{Z}^+$, written gcd(a, b), is the greatest $d \in \mathbb{Z}^+$ such that $d \mid a$ and $d \mid b$.

Update: We can also allow one of the arguments to be zero:

$$gcd(a,0) = gcd(0,a) = a \qquad \text{if } a > 0$$

But gcd(0,0) is undefined.

Euclid's Algorithm

There is a better algorithm for finding the GCD of two integers that dates back to Euclid's *Elements*, from around 300 BC.

Euclid's Algorithm

$$gcd(a,b) = \begin{cases} gcd(b, a \mod b) & \text{if } b > 0\\ a & \text{if } b = 0 \text{ and } a > 0 \end{cases}$$

Example

Suppose we want to find gcd(287, 91).

$$gcd(287,91) = gcd(91, 14)$$

= $gcd(14, 7)$
= $gcd(7, 0)$
= 7

because $287 = 3 \cdot 91 + 14$ because $91 = 6 \cdot 14 + 7$ because $14 = 2 \cdot 7 + 0$ **Algorithm 4** euclid-gcd(a, b)

Require: $a, b \in \mathbb{Z}^+$ **Ensure:** Returns gcd(a, b).

1:
$$x \leftarrow a$$

2: $y \leftarrow b$
3: while $y \neq 0$ do

4:
$$r \leftarrow x \mod y$$

5:
$$x \leftarrow y$$

6:
$$y \leftarrow r$$

- 7: end while
- 8: **return** *x*

Termination

while loop: y decreases

Correctness

- Loop Invariant: gcd(x,y) = gcd(a,b)
- After each loop iteration: see lemma
- When the loop exits (with y = 0): gcd(x, 0) = gcd(a, b)

Running time

O(log(max(a, b))) iterations

Lemma

Let $a, b \in \mathbb{N}$ with $b \neq 0$. Then $gcd(a, b) = gcd(b, a \mod b)$.

Proof.

First we show that a, b have the same common divisors as $a, a \mod b$. Recall this property: for all $x, y, z \in \mathbb{Z}$, if $x \mid y$ and $x \mid z$, then $x \mid (y + z)$. By division, there are q, r such that a = qb + r, where $r = a \mod b$.

- By applying the divisibility property above to the equation for a, we see that any common divisor of b and a mod b must also be a divisor of a.
- ▶ By rewriting the equation to a mod b = a + (-q)b and applying the divisibility property again, we see that any common divisor of a and b is also a divisor of a mod b.

Since a, b have exactly the same common divisors as $b, a \mod b$, they have the same *greatest* common divisor. That is, $gcd(a, b) = gcd(b, a \mod b)$.

The Extended Euclidean Algorithm

The GCD of a and b can be expressed as a linear combination of a and b. That is, gcd(a, b) = sa + tb for some $s, t \in \mathbb{Z}$.

Example

Previously: gcd(287,91) = gcd(91,14) = gcd(14,7) = gcd(7,0) = 7.

$287 = 3 \cdot 91 + 14$	\rightarrow	$14 = 287 - 3 \cdot 91$
$91 = 6 \cdot 14 + 7$	\rightarrow	$7=91-6\cdot 14$
$14 = 2 \cdot 7 + 0$		

If we substitute backwards using the equations on the right:

 $7 = 91 - 6 \cdot 14$ by 2nd equation

 $= 91 - 6 \cdot (287 - 3 \cdot 91)$ by 1st equation

 $= -6 \cdot 287 + 19 \cdot 91$ by algebra

(In fact, every intermediate remainder can be expressed as some sa + tb ...)

The Extended Euclidean Algorithm



Termination

while loop: *y* decreases

Correctness

Loop Invariants:

$$\blacktriangleright \gcd(x,y) = \gcd(a,b)$$

$$x = s_x a + t_x b$$

$$y = s_y a + t_y b$$

- After each loop iteration: see lemma; algebra
- When the loop exits (with y = 0):
 - $\blacktriangleright \gcd(x,0) = \gcd(a,b)$

$$\bullet \ x = s_x a + t_x b$$

Running time

O(log(max(a,b))) iterations

The Multiplicative Inverse in Modular Arithmetic

Definition (Multiplicative Inverse)

Let $m \in \mathbb{Z}^+$, and let $x \in \mathbb{N}$. The **multiplicative inverse (mod** m) of x is a number $y \in \{0, ..., m - 1\}$ such that $xy \equiv 1 \pmod{m}$.

The multiplicative inverse does not always exist. It exists if and only if x and m are relatively prime.

Examples

- the multiplicative inverse (mod 10) of 7 is 3 because 3 · 7 = 21 ≡ 1 (mod 10)
- ► the multiplicative inverse (mod 8) of 7 is 7 because 7 · 7 = 49 ≡ 1 (mod 8)
- ► the multiplicative inverse (mod 6) of 4 does not exist because 4 times anything is even, and no even number is = 1 (mod 6)

Calculating the Multiplicative Inverse

We can use the extended Euclidean Algorithm to find the multiplicative inverse of $x \pmod{m}$:

- If $gcd(x,m) \neq 1$ then the multiplicative inverse does not exist.
- Otherwise (if x and m are relatively prime), the algorithm computes s and t such that sx + tm = 1.

Thus sx - 1 = -tm, and thus $sx \equiv 1 \pmod{m}$ (by definition).

Example

We could use the extended Euclidean algorithm to calculate

$$\gcd(31, 43) = 1 = -18 \cdot 31 + 13 \cdot 43$$

If we take -18 and normalize it to the proper range: $(-18) \mod 43 = 25$. So the multiplicative inverse (mod 43) of 31 is 25.

Check: $31 \cdot 25 = 775 = 18 \cdot 43 + 1 \equiv 1 \pmod{43}$.