

CS 331, Fall 2025

Lecture 1 (8/25)

Today: - Logistics

- Background
 - Induction
 - Asymptotics
 - Data Structures
- Orders of magnitude
- Recursion
- Multiplication

Logistics

We use the following websites.

1) Course website

(kjtian.github.io/cs331-fa25.html)

- Lecture notes, HW posted

- Syllabus, course info

- Feedback forms

Lecture feedback

What made the least sense?

TAs will go over in sections.

Notes feedback

Types, confusing examples

2) Ed

- Announcements, practice tests

- Ask questions about lecture, HW

3) (2022)

- HW turn-in, grades released
- Lecture videos

Grade breakdown

40% HW (6x lowest dropped)

30% Midterms (2x)

30% Final exam

Some notes

- No AI on HW! Best opportunity to practice for tests.

Acceptable use: asking to explain lecture notes, examples

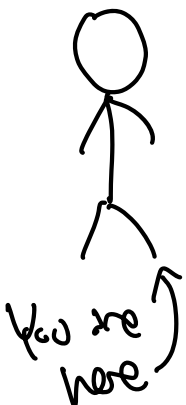
- Coding component

≤ 1 coding question / HW, Python w/ starter code

Let me know early if concerns about background.

- Tip: ask questions early, use OI + discussions (Ed, feedback forms)

Paradigms	Toolkits	Hardness	Wilderness
<ul style="list-style-type: none">- Recursive- Dynamic programming- Greedy	<ul style="list-style-type: none">- Graphs- Continuous- Randomized	<ul style="list-style-type: none">- Complexity	<ul style="list-style-type: none">- Interview- Job- Research- ...

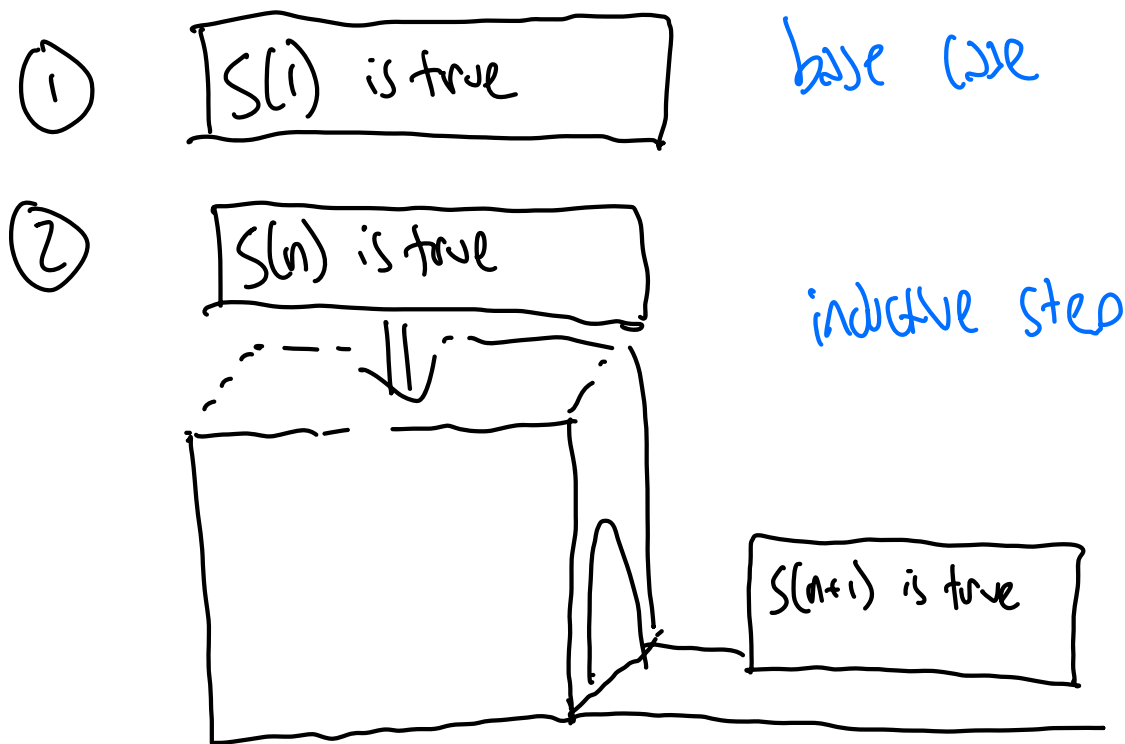


Background: Induction (part I, Section 2)

You want to prove a statement $S(n)$ for all $n \in \mathbb{N}$.

e.g. $S(n)$: There are 2^n subsets of $[n] = \{1, 2, \dots, n\}$

Induction builds "machines" to generate new statements out of old ones. Example: "basic induction"



This can prove any $S(n)$: $S(1) \Rightarrow S(2) \Rightarrow S(3) \dots \Rightarrow S(\infty) \dots$

Ex. $S(1) : |\{\emptyset, \{1\}\}| = 2 = 2^1 \checkmark$

$S(n+1)$: Subsets of $[n+1]$ containing $n+1$
not containing $n+1$

2^n
 2^n \checkmark

Example

Claim: For all $x \in \mathbb{R}$, $n \in \mathbb{N}$,

$$x^n - 1 = (x - 1) \underbrace{\left(\sum_{i=0}^{n-1} x^i \right)}_{1 + x + \dots + x^{n-1}}$$

Proof: Base case $S(1)$: $x - 1 = (x - 1)(1)$ ✓

Induction Assume $S(n)$ is true, so we know

$$x^n - 1 = (x - 1) \sum_{i=0}^{n-1} x^i$$

$$\text{Then, } x^{n+1} - 1 = x^{n+1} - x^n + x^n - 1$$

$$= (x^{n+1} - x^n) + (x - 1) \sum_{i=0}^{n-1} x^i$$

$$= (x - 1) \left(x^n + \sum_{i=0}^{n-1} x^i \right) \quad \checkmark \quad \blacksquare$$

Consequence: $a_0 + a_0 r + a_0 r^2 + \dots + a_0 r^k$

$$= a_0 (1 + r + r^2 + \dots + r^k) = a_0 \cdot \frac{r^{k+1} - 1}{r - 1}$$

Sum of geometric series.

Another version: Strong induction.

① $S(1)$ is true base case

② $S(1)$ is true
 $S(2)$ is true
...
 $S(n)$ is true inductive step



We get this
stronger assumption
for free during
normal induction!

Still proves all $S(n)$.

Example

Claim: All $n \geq 2$, $n \in \mathbb{N}$ can be written
as a product of primes.

Proof: Base case $S(2)$: $2 = 2$ ✓

Induction If $n \geq 2$, either...

n is prime

$n = n$ ✓

n is composite

$n = \underbrace{2}_{p_1 p_2 \dots p_i} \cdot \underbrace{b}_{q_1 q_2 \dots q_j}$ ✓

Strong induct:

$p_1 p_2 \dots p_i q_1 q_2 \dots q_j$

Background: Asymptotics (Part I, Section 3)

In math as long as you can prove all $S(n)$, happy.

You can use as many steps as you want.

In algos we care about efficiency along with correctness

In this class, n = size of input to problem

How to compare Algo 1: solves in $f(n)$ time

Algo 2: solves in $g(n)$ time

Philosophy

If n is small, Algo 1, 2 both fast

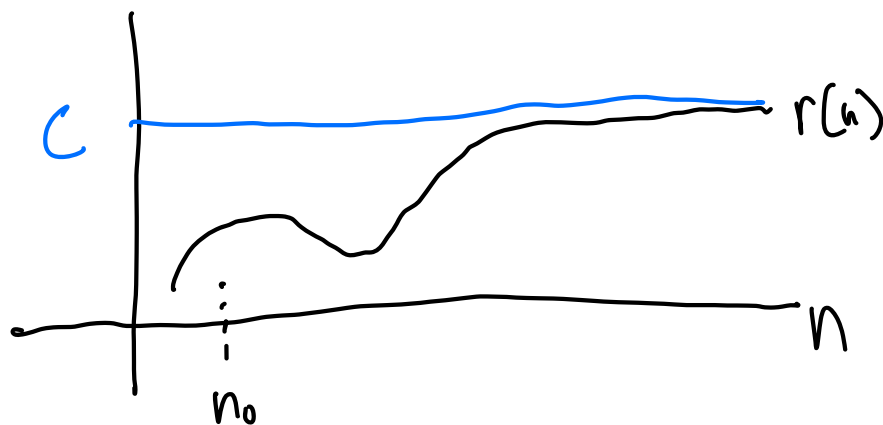
We care about performance as $n \rightarrow \infty$
"asymptotics"

The zoo:

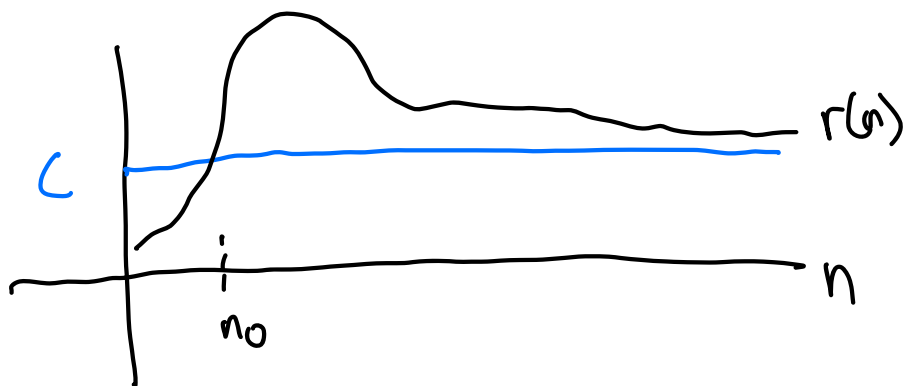
$$\begin{array}{l} f(n) = \begin{array}{l} O(g(n)) \\ \Omega(g(n)) \\ \Theta(g(n)) \\ o(g(n)) \\ \omega(g(n)) \end{array} \iff \frac{f(n)}{g(n)} = \begin{array}{l} O(1) \\ \Omega(1) \\ \Theta(1) \\ o(1) \\ \omega(1) \end{array} \end{array}$$

When is a function $r(n)$...

... $O(1)$? $\forall n \geq n_0, r(n) \leq C$ (constant independent of n , like 100)



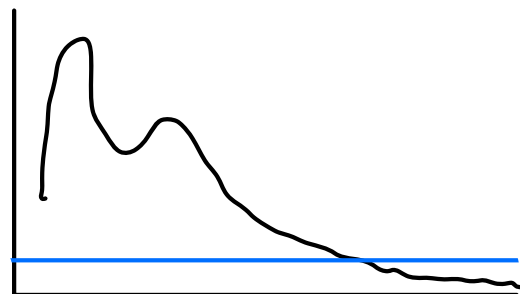
... $\Omega(1)$? $\forall n \geq n_0, r(n) \geq C$



... $\Theta(1)$? It just means $O(1)$ and $\Omega(1)$.

... $o(1)$? It just means not $\Omega(1)$.

AKA $\lim_{n \rightarrow \infty} r(n) = 0$



... $\omega(1)$? It just means not $O(1)$

AKA $\lim_{n \rightarrow \infty} r(n) = \infty$

Three main function types in this class:

Exponential $(2^{100n}, e^n, \dots)$

Polynomial $(n^6, 12n^3, n^2 + 3, \dots)$

Polylogarithmic $(\log^3 n, 3 \log^4 n, \log n + 2, \dots)$

Helpful rule

Exponential \gg Polynomial $c^n = \omega(n^b) \quad \forall \begin{matrix} c > 1 \\ b > 0 \end{matrix}$

Polynomial \gg Polylogarithmic $n^b = \omega(\log^a(n)) \quad \forall \begin{matrix} b > 0 \\ a > 0 \end{matrix}$

Exercise

Rank the following asymptotically

$n^{100} e^n$, 1000000, $\log^{1000}(n)$, n^{-100} , $n^2 3^n$

Background: Data Structures (Part I, Section 7)

Assumed background, notes provide proofs.

Make sure you're comfortable: Very useful in algos.

General rule: In this class, you can use anything in notes as true without proof. Use following APIs "for free"

Array

- Init $O(1)$
- Insert $O(1)$
- Delete $O(1)$
- Query $O(1)$

Heap

- Init $O(n)$
- Insert $O(\log(n))$
- Extract Min $O(\log(n))$
- Delete $O(\log(n))$

LinkedList

- Init $O(1)$
- Insert $O(1)$
(changes index)
- Delete $O(1)$
(changes index)
- Query $O(1)$
(needs address)

BST

- Insert $O(\log(n))$
- Delete $O(\log(n))$
- Query $O(\log(n))$
- Search $O(\log(n))$

Hash Table

- Init $O(1)$
- Insert $O(1)$ (only in expectation...)
- Search $O(1)$
- Delete $O(1)$

Stack/Queue

Based on
LinkedList

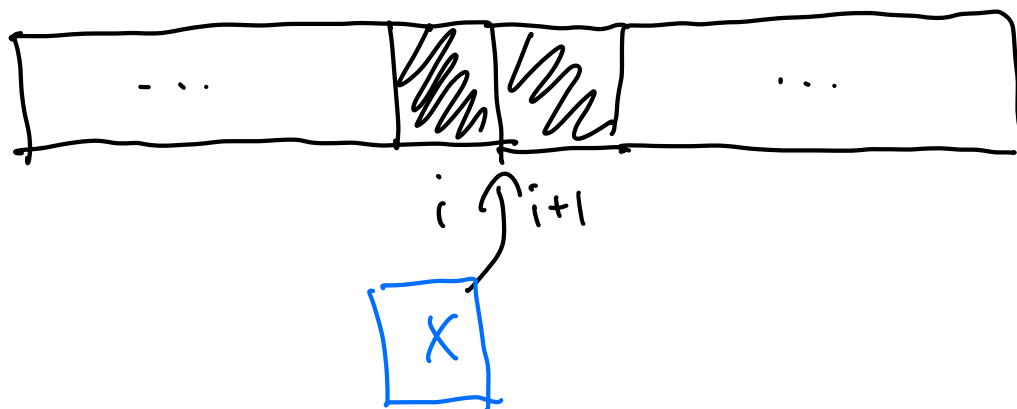
not used
until Part VII

Exercise

Which is preferable?

Array vs. Linked list

- You want to insert in between adjacent elements



- You want to query the i^{th} slot in the list

Heap vs. BST

- You want to find the i^{th} largest element of unsorted list

14	-3	200	0.5	0.1
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$i = 3$ Ans: 0.5

- You want to perform \uparrow task for many ($\gg n$) values of i , possibly different

Orders of magnitude (Part II, Section 4)

Taxonomy of runtimes:

polylog \ll (near-)linear \ll polynomial \ll exponential

$$n < n \log(n) < n \log^2(n) < n^2 < n^3$$

Sublinear
algs

$$f(n) = o(n)$$

Part VII only

e.g. $n = 4 \cdot 10^9$

Google searches
every day

We will care about
log factors in this class.

Caution: what constants?

$$\log^3(n) > \sqrt{n} \text{ if } n \leq 10^7$$

We will be upfront about
when constants are large,
usually no problem.

Exponential-time
algs

Characteristic of
"brute-force" solution
 $f(n) = c^n, c > 1$

Part VIII: When
is there nothing
better to do?

e.g. $c = 2$
 $n = 250$

$$c^n > \# \text{ particles in universe}$$

Recursive algorithms (Part II, section 1)

To analyze an algo:

- 1) Proof of correctness
- 2) Analyze runtime (next lecture)

How to analyze correctness?

Complex Algo (inputs)

Step 1

Step 2

Subroutine₁(inputs')

Subroutine₂(inputs'')

Step 3

...

1) Assume all subroutines are correct. Are they "stitched together" correctly?

2) Are all the subroutines correct? Prove it.

e.g. to get ready for class,

put on clothes

pack bag

start coffee machine

...

This unit's theme: particular type of algo

Recursion key idea: only use algo itself as subroutine

Recursive Algo (size n input):

Step 1

Step 2

Recursive Algo (size $n' < n$ input)

Step 3

Recursive Algo (size $n'' < n$ input)

...

~~2) Are all the subroutines
correct? Prove it.~~

no need: strong induction!
AKA "recursion fairy"

e.g. to paint a fence, paint left half, paint right half

Recursive algorithm is like an efficient
proof by strong induction.

Multiplication (Part II, Section 2)

Basic model: Add, subtract, multiply 1-digit #'s in $O(1)$.

Warmup

How expensive is adding n -digit #'s?

Runtime:
 $O(n)$

$$\begin{array}{r} a = 123456 \\ + b = 987654 \\ \hline = 1111110 \end{array}$$

grade-school
also: each
output takes $O(1)$

How about multiplication?

2 = 1 2 3 4 5 6
x b = 9 8 7 6 5 4

Runtime:
 $O(n^2)$

$$\begin{array}{r} \text{we:} \\ n^2) \quad \begin{array}{r} 49\ 3\ 8\ 2\ 4 \\ 6\ 1\ 7\ 2\ 8\ 0 \\ 7\ 4\ 0\ 7\ 3\ 6 \\ 8\ 6\ 4\ 1\ 9\ 2 \\ 9\ 8\ 7\ 6\ 4\ 8 \\ +\ 11\ 11\ 10\ 4 \\ \hline = 121931812224 \end{array} \end{array}$$

n numbers
each $O(n)$ -long

Can recursion help? Want: Smaller subproblems
that if solved, make progress

Idea 1: divide-and-conquer

e.g.

$$a = a_1 \cdot 10^{n/2} + a_0$$

$$b = b_1 \cdot 10^{n/2} + b_0$$

$$\begin{array}{cccccc} 1 & 2 & 3 & 4 & 5 & 6 \\ \hline & & a_1 & & a_0 & \\ & & & & & \\ 9 & 8 & 7 & 6 & 5 & 4 \\ \hline & & b_1 & & b_0 & \end{array} \quad n=6$$

By "FOIL"

$$ab = a_1 b_1 \cdot 10^n + (a_1 b_0 + a_0 b_1) \cdot 10^{n/2} + a_0 b_0$$

4 $\frac{n}{2}$ -digit multiplications
+ Carries & bitshifts $O(n)$

$$\begin{aligned} T(n) &= 4T\left(\frac{n}{2}\right) + O(n) \\ &\quad \underbrace{\text{runtime on}}_{\text{input length } n} \\ &= O(n^2) \end{aligned}$$

Idea 2: Karatsuba recursion

we'll see why
next time!

Rewrite $(a_1 b_0 + a_0 b_1)$

$$\begin{aligned} &= \underbrace{(a_1 + a_0)(b_1 + b_0)}_{\text{one more } \frac{n}{2}\text{-digit multiply}} - \underbrace{a_1 b_1}_{\text{precomputed}} - \underbrace{a_0 b_0}_{\text{precomputed}} \end{aligned}$$

$$\begin{aligned} T(n) &= 3T\left(\frac{n}{2}\right) \\ &\quad + O(n) \\ &= O(n^{1.59}) \end{aligned}$$