

Monday (January 4th)

Algorithms

UCP program, let Yener know.

Wear a mask around Yener.

get on the piazza → homework posting

Handwritten notes scanned & uploaded;
all announcements in piazza

TAs & I will use submitly for HW

recordings online, "as the need arises" → ~~get access~~ to the recordings.

Yener will give permission to the ^{all} recordings → Webex recordings → email him.

The reason he isn't giving permission

to everyone for the recordings is

because no one shows up to class when

they have recordings.

This is supposed to be fun &
interactive. Yener wants people to
come & talk to him.

What we will learn: ~~how to solve~~

given a problem ~~how to~~ solve it

Find a solution that is correct.

Monday January 9th 2023

Algorithms 12

goals: minimize space, time complexity
which is really minimizing energy.

1. how many steps it will take to run a given algorithm

eg: given input size N , when you are solving a problem

how many steps will it take and what
is the cost?

Cost \rightarrow the operation

(addition vs multiplication vs exponentiation)

given

64 bit word

if a number doesn't fit in the word then it cannot be stored.

So the size of the number is important.

be clear on:

number of steps on input N is a gross approximation
used to determine the amount of bits to operate on N .

N is a countable number of things.

Big-O notation gives an upper bound on time complexity.

"The program is not going to exceed this bound."

$\Omega \rightarrow$ Lower bound, Omega

"No matter which algorithm you use, it takes at least
this amount of resources."

Algorithms Monday January 4th 2023

An Algorithm can give an upper bound.

To get the lower bound a theorem must be developed.

One goal is to optimize, make the upper bound match the lower bound.

A gap emerges between the upper and lower bound.

There is a constant approximation for NP problems.

If you cannot close the gap \rightarrow this problem does not exist a solution in polynomial time.

It takes exponential or super exponential time at

which point you approximate based on the biggest factor.
*you stay up all night at the disco tech when ^{fall} the firm!

"in greedy algorithms there is no regret"

Heisenberg Uncertainty Principle

If you observe a system you Δ in Δ change the value.

Lab assignment every week - "Simple". Show your work, get checked.

Will give homework with modifications from popular questions online.

Will give exams. closed book.

Yener doesn't trick students.

He will tell you what is in the exam during lecture.

Yener does late hours for Zoom meetings.

Algorithms Monday January 9th

Today: using Fibonacci sequence

- Time complexity

- Bit complexity

- Big-O notation

Fib = [0, 1, 1, 2, 3, 5, 8, 13, ...]

$F_0 = 0$
 $F_1 = 1$ } Base case

n th element of the sequence

$F_n = F_{n-1} + F_{n-2}$ } recursive definition

if you have a formula like this

you can turn this into an algorithm

FIB1(n) n is an integer > 0

if $n = 0$ return 0

if $n = 1$ return 1

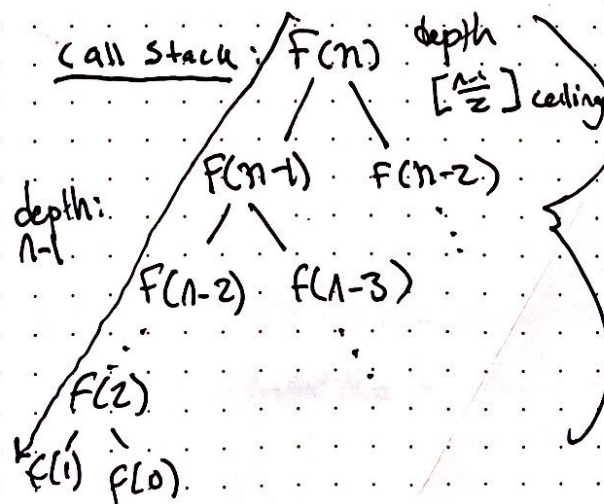
return FIB1($n-1$) + FIB1($n-2$)

① Correctness \rightarrow is this algo, given n , going to return something & is that something the answer we are looking for?

② efficiency \rightarrow

time

space



The leaves of the executing graph will be recursively to the base case.

Algorithms Monday January 9th 2023

Continued...

of recursive calls < # of nodes in a complete binary tree of depth $n-1$.

time complexity of this recursive algorithm can be bounded by calls, which is bounded by the number of nodes in a complete binary tree.

↓
 $\# = 2^{n-1} = 2^n$

because it is a geometric series

$$\sum_{i=0}^K r^i = \frac{r^{K+1} - 1}{r - 1}$$

in Fib case:

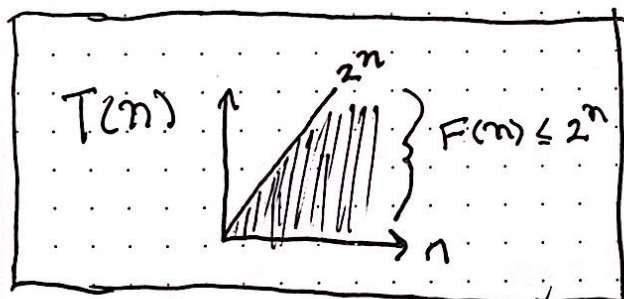
$$r = 2$$

$$K = n-1$$

$$F(10) \leq 2^{10}$$

Time complexity of $FIB1(n)$

in general $\rightarrow F(n) \leq 2^n$: exponential in n .



Improvement! Stop recomputing and eliminate redundancy in the call stack.

Use memorization to

Cache the results of $F(n)$

Storing it in a linear array, size of n .

$FIB2(n)$

$i = 0, 1, 2, 3 \dots n$

$F = 0 \quad 1 \quad 1 \quad 2 \quad 3 \quad 5 \dots$

$$F[0] = 0$$

$$F[1] = 1$$

for $i = 2 \dots n$

$$F[i] = F[i-1] + F[i-2] \rightarrow n+1 \text{ steps} \rightarrow \text{linear time algorithm.}$$

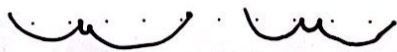
We assume that access is constant time so its $O(1)$ is cost.

Monday, January 9th 2023

Algorithms

Continued...

$$F(n-1) + F(n-2)$$



$n-1$
bits

$n-2$
bits



$$n \text{ operations} \Rightarrow (n-2+1) * n = n^2 - n \approx \boxed{n^2}$$

$n = 2^k$ bits to represent. $\left\{ \begin{array}{l} k \text{ is a linear function.} \\ \text{deal with larger numbers} \end{array} \right.$

Claim: $F_n \geq 2^{0.5n}$ For $n \geq 6$ $F(n) \leq 2^n$

Proof: (by induction)

Base Case: $F_6 = 8 \geq 2^{(0.5) \times 6} = 2^{6/2} = 8 \checkmark$

Inductive Step: For $n \geq 6$

$$F_{n+1} = F_n + F_{n-1} \geq 2^{n/2} + 2^{(n-1)/2} = 2^{(n-1)/2} \cdot (2^{1/2} + 1)$$

$$\geq 2^{(n-1)/2} \cdot 2 \geq 2^{(n+1)/2} \checkmark$$

$$F(n) \geq 2^{0.5n} \geq 2^n$$

Next time: ~~for~~ Big O Notation.