

# Intro to Algorithms

①

Today: using Fibonacci sequence

- Time complexity
- Bit Complexity
- Start Big O notation.

F.B.: 0 1 1 2 3 5 8 13 ...

$F_0 = 0$   
 $F_1 = 1$  } base case.

$n^{th}$  element  
of the seq:

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

FIB 1 (n)      n is an integer  $> 0$

if  $n=0$       return 0

if  $n=1$       return 1

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

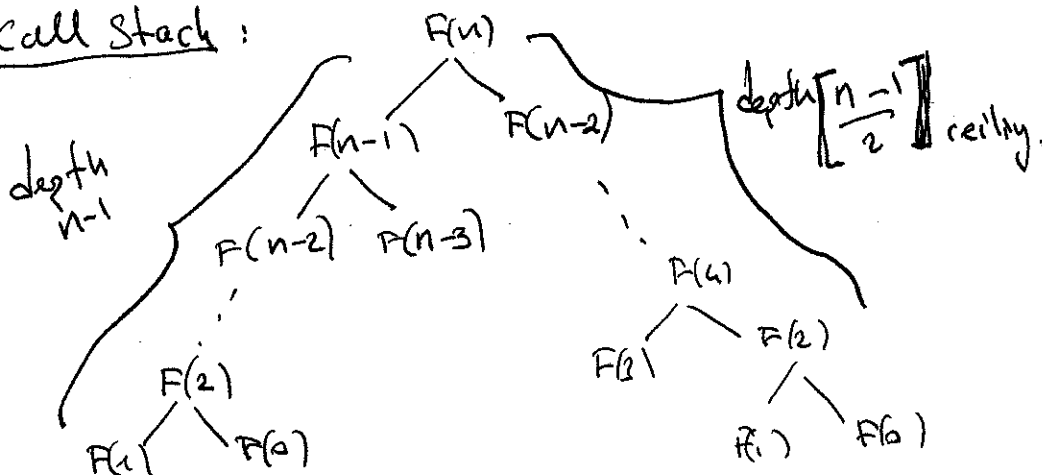
① Correctness

② efficiency.

time

space

Call Stack:



$$\underbrace{F(n-1)}_{n-1 \text{ bits}} + \underbrace{F(n-2)}_{n-2 \text{ bits}}$$

$\approx n$  operations.

$$\Rightarrow (n-2+1) \cdot n = n^2 - n \approx n^2$$

$n = 2^k$  bits to represent. } deal with large  $n$ 's.

Claim:  $F_n \geq 2^{0.5n}$  for  $n \geq 6$ .

Proof: (by induction).

base:  $F_6 = 8 \geq 2^{(0.5)6} = 2^{3} = 8 \checkmark$

Inductive Step:

for  $n \geq 6$

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

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

next: big  $O()$

## Lecture 2

④

Today: - finish basic analysis  
- Asymptotic time analysis

$O(\cdot)$ ,  $\Omega(\cdot)$ ,  $\Theta(\cdot)$

Last time:

$$F(n) \approx 2^{c \cdot n}$$

$$c \approx 0.7$$

$$0.7 \cdot n = \Theta(n)$$

$$F(n) \sim 2^n$$

$$\underline{N} = 2^K$$

$\Rightarrow$   $n$ th Fib. # requires  $n$ -bits

Fib2

$n$  times  
and  
each time  
one  
addition  
of  $n$ -bit  
#s

Fib. 1 1 1 1 1 ... 1

$$\left[ \text{for } i=2 \dots n \right. \\ \left. F[i] = \underbrace{F[i-1]}_n + \underbrace{F[i-2]}_n \right]$$

$\Theta(n^2)$

bit complexity

FIB3 ( $n$ )

based on Matrix multiplication.

$$F_0 = 0$$

$$F_1 = 1$$

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

$$\begin{bmatrix} F_1 \\ F_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} F_0 \\ F_1 \end{bmatrix}$$

$$F_1 = 0 \cdot F_0 + 1 \cdot F_1$$

$$1 \cdot F_0 + 1 \cdot F_1 = F_0 + F_1 = F_2$$

for the general case.

$$\begin{bmatrix} F_2 \\ F_3 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} F_1 \\ F_2 \end{bmatrix}$$

$$\begin{bmatrix} F_2 \\ F_3 \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}}_{M^2} \begin{bmatrix} F_0 \\ F_1 \end{bmatrix}$$

(5)

$$\begin{aligned} \begin{bmatrix} F_n \\ F_{n+1} \end{bmatrix} &= M^n \cdot \begin{bmatrix} F_0 \\ F_1 \end{bmatrix} = M^n \begin{bmatrix} 0 \\ 1 \end{bmatrix} \\ &= \underbrace{\begin{bmatrix} a & b \\ c & d \end{bmatrix}}_{M^n} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} b \\ d \end{bmatrix} \Rightarrow F_n = b \end{aligned}$$

FIB3(n)

if  $n=0$  return 0  
 if  $n=1$  return 1

$$M = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \quad M^1 = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}$$

for  $i = 2 \dots n$ 

$$M^i = M^{i-1} \times M$$

$$M^i = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

n times

 $O(n^2)$  $\Rightarrow O(n^3)$ 

return b

— let's consider a better analysis

Q: how many matrix multiplications does it take to compute  $M^n$ 

$$n = 2^k$$

$$M^n = M^{2^k}$$

$$M^{2^k} = M^{2^{k-1}}$$

we can compute recursively

$$M^{2^2} \cdot M^{2^2} = M^{2^3}$$

$$M^{2^2} = M^{2^1} \cdot M^{2^1}$$

!

Note: every squaring we are doubling the exponent  
 $k = \log n$



$$M^n = \begin{cases} (M^{\lfloor n/2 \rfloor})^2 & \text{if } n \text{ is even} \\ M (M^{\lfloor n/2 \rfloor})^2 & n \text{ is odd.} \end{cases}$$

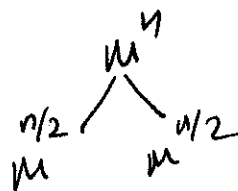
⑥

$\Rightarrow \log n$  squares,  $\log n$  multiplications by  $M$ .

[general rule: if the problem size is halved at each iteration then it takes  $O(\log n)$  steps.]

Formally we write a recurrence formula for the running time of an algorithm.

$$T(n) = T\left(\frac{n}{2}\right) + \underbrace{\text{mul.}\left(\frac{n}{2}\right)}$$



$$\widetilde{\text{Rib3}(n)} = T\left(\frac{n}{2}\right) + \left(\frac{n}{2}\right)^2$$

$$= \left[ T\left(\frac{n}{4}\right) + O\left(\frac{n}{4}\right)^2 \right] + \frac{n^2}{4}$$

$$= T\left(\frac{n}{4}\right) + O\left(\frac{n^2}{16}\right) + O\left(\frac{n^2}{4}\right)$$

$$= \frac{n^2}{4} + \frac{n^2}{16} + \frac{n^2}{64} \dots$$

$$= \frac{n^2}{4} \left( 1 + \frac{1}{4} + \frac{1}{16} \dots \right)$$

$T(n) \leq \frac{n^2}{4} \cdot 2$  geometric series  $\leq 2$

$$\leq \frac{n^2}{2} \Rightarrow \boxed{O(n^2)}$$

$\text{Rib3}(n): O(n^3) \rightarrow O(n^2 \log n) \rightarrow O(n^2)$

but: we can do faster matrix multiplication

$$O(n^{1.6}) \approx \boxed{O(n \cdot \sqrt{n})}$$

Rib3

Rib2  $O(n^2)$

Fib4(n)

$$F_n = \frac{1}{\sqrt{5}} (\lambda_1^n - \lambda_2^n) \quad (*)$$

$\lambda_1$  and  $\lambda_2$  are the eigen values of matrix  $M$  (7)

$$\lambda_1 = \frac{1+\sqrt{5}}{2}$$

$$\lambda_2 = \frac{1-\sqrt{5}}{2}$$

$$M = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}$$

eigen values and vectors to express a matrix

$$M = V \Lambda V^{-1}$$

$\underbrace{V}_{\text{eigen vector}}$   $\underbrace{\Lambda}_{\text{eigen values}}$  are at the diagonal

$$M \cdot v = \lambda v$$

for any value of  $\lambda$  that solves this equation is an eigen value of  $M$ , and corresponding  $v$  is the eigen vector.

$$M = \begin{bmatrix} | & | \\ v_1 & v_2 \\ | & | \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} \begin{bmatrix} | & | \\ & \\ | & | \end{bmatrix}^{-1}$$

$$M^n = \begin{bmatrix} | & | \\ v_1 & v_2 \\ | & | \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}^n \begin{bmatrix} | & | \\ & \\ | & | \end{bmatrix}^{-1} \quad (*)$$

we write characteristic equation

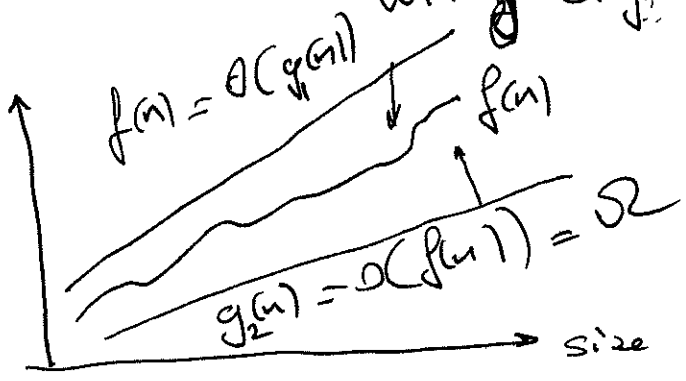
$$|M - \lambda I| = 0$$

Solve this

# Asymptotic Time Analysis with big-Oh notation.

$$\Theta(n) = \Omega(n) \Rightarrow \Theta(n)$$

Time it takes.

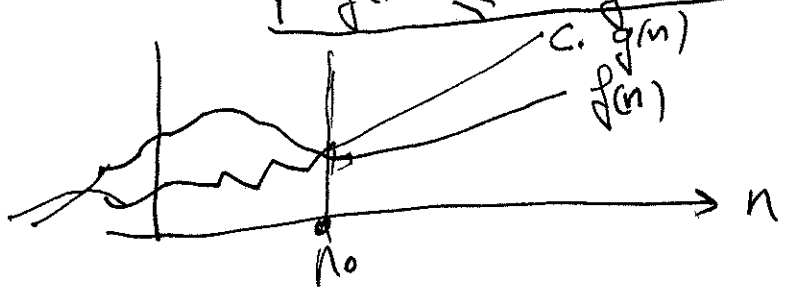


size of the problem in n

## Definition:

we say that  $f(n) = O(g(n))$  if there exists a constant  $c > 0$  such that for all  $n > n_0$ .

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



$\underbrace{2n^2}_g$  vs.  $\underbrace{5n^2 + 10^6 n + 5}_f$   $n > 0$

$$g = O(f)$$

$$g \leq c \cdot f$$

$1000 n^2$  v.s.  $n^3$

# Asymptotic analysis using limits - order of growth. (9)

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \begin{cases} 0: \text{case I } f(n) \text{ has a smaller order of growth.} \\ c: \text{case II } f(n) \text{ has the same order of growth as } g(n) \\ \infty: \text{case III } f(n) \text{ has larger order of growth. w.r.t. } g(n) \end{cases}$$

If case I and case II are true then  $f(n) \in O(g(n))$   
 case II and case III are true then  $f(n) \in \Omega(g(n))$   
 case II  $f(n) \in \Theta(g(n))$ .

example

$$f(n): n^2 + 5n \quad g(n): 10n^2 + 2n + 100$$

$$\lim_{n \rightarrow \infty} \frac{n^2 + 5n}{10n^2 + 2n + 100} = \frac{1}{10} = 0.1 \text{ constant}$$

$$f(n) \in \Theta(g(n)) \quad \checkmark$$