

Transition: DS + Algorithms

Runtime
Analysis



Last time: Big-O Proofs

$$\sum_{i=1}^n i$$

Worksheet review:

$$\sum_{i=1}^n \left(\sum_{j=1}^i 1 \right) = \sum_{i=1}^n \left(\underset{j=1}{1} + \underset{j=2}{1} + \dots + \underset{j=i}{1} \right)$$

for $i \leftarrow 1$ to n

for $j \leftarrow 1$ to i
add 1

$$= \sum_{i=1}^n \binom{i}{1}$$

$$= \sum_{i=1}^n i$$

$$= 1 + 2 + 3 + \dots + n$$

$$= \frac{n(n+1)}{2} = \frac{n^2}{2} + \frac{n}{2}$$

$$= O(n^2)$$

$$\sum_{i=1}^n i^2 = 1^2 + 2^2 + \dots + n^2$$

↑
aside

$$\lg 2^n = \log_2 (2^n) = n$$

dim of $\lg$ = exponent we raise 2 to in order = 2^n

$$2^k = 2^n$$

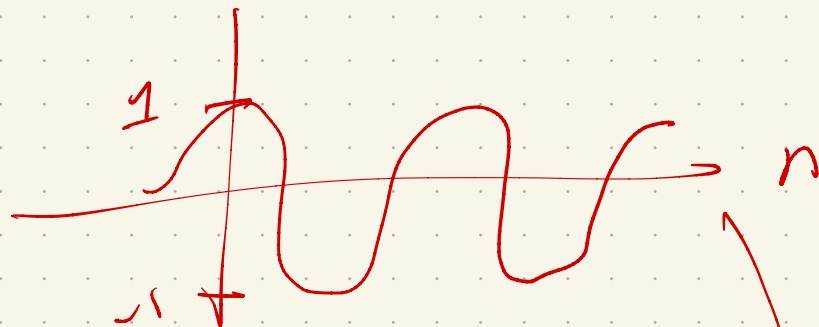
exponents: $x^a \cdot x^b = x^{a+b}$

$$(x^a)^b = x^{ab}$$

Next: $2^{2 \lg n}$

$$= (2^{\lg 2^n})^2 = n^2 = \Theta\left(\sum_{i=1}^n \sum_{j=1}^n 1\right)$$

$$\cos n + 2$$



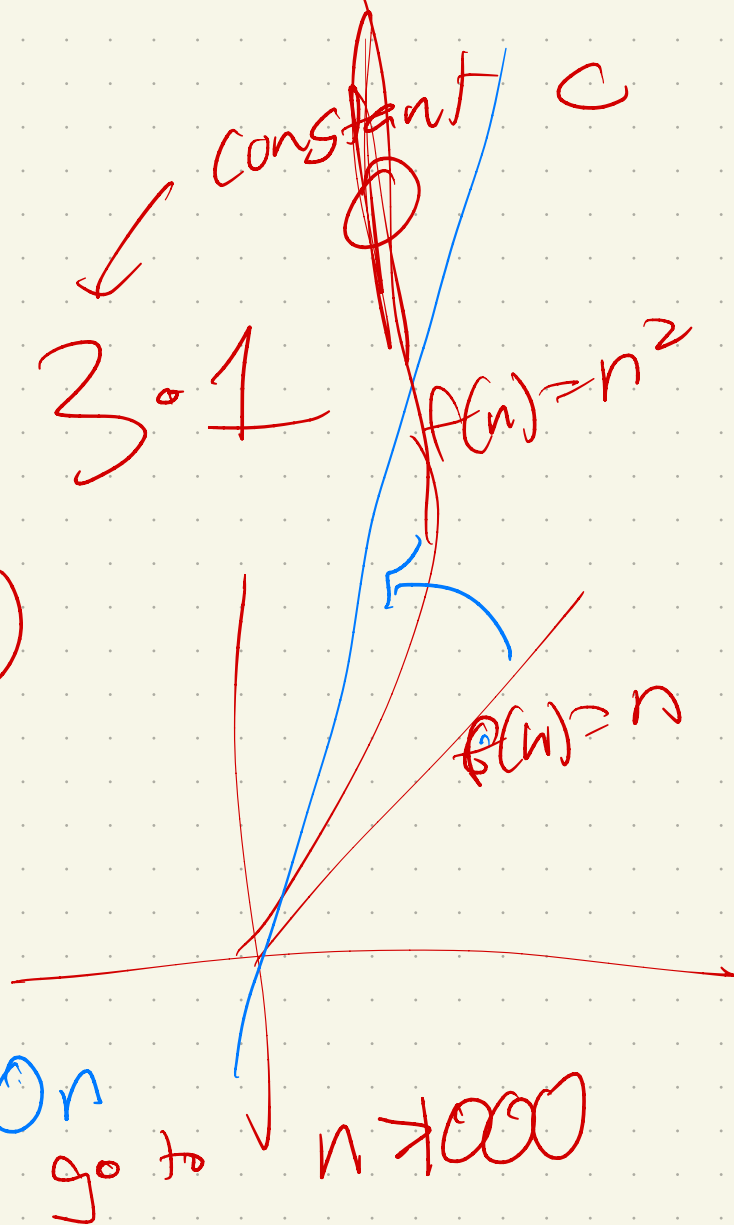
$$\cos n \leq 1$$

$$= O(1)$$

$$\cos n + 2 \leq 3 \approx 3 \cdot 1$$

$$\Rightarrow O(1)$$

$$\cos n + 2 \neq 0$$



Recap Let f and g be functions from $\mathbb{Z}^+$ to $\mathbb{R}^+$ $\rightarrow$ counting things
why?

We say $f(n) = O(g(n))$ if $\exists$
constants c and n_0 s.t. $\forall n > n_0$,
 $f(n) \leq c \cdot g(n)$

Pick a worksheet one:

That limit definition

I did $f(n) = O(g(n))$ if $\exists n_0 \neq c$ s.t.
 $\forall n > n_0, f(n) \leq C \cdot g(n)$

Take that inequality:

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

$\Rightarrow$

$$\frac{f(n)}{g(n)} \leq C \quad \text{if } n > n_0$$

$$\text{So } f(n) = O(g(n)) \Leftrightarrow \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} \leq C$$

Big Ω Theorem

$\exists c, n_0$ s.t. $\forall n > n_0, f(n) \leq c \cdot g(n)$

Let f and g be functions from $\mathbb{Z}^+ \rightarrow \mathbb{R}^{\geq}$

We say f is $\Omega(g(n))$ if

$\exists$ positive constants c and n_0
such that $\forall n > n_0,$

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

" f is big-omega of g "

S_n is $O(n)$
pick $c=6$
 $S_n \leq 6n$

Example: $5 \cdot n$ is $\Omega(n)$:

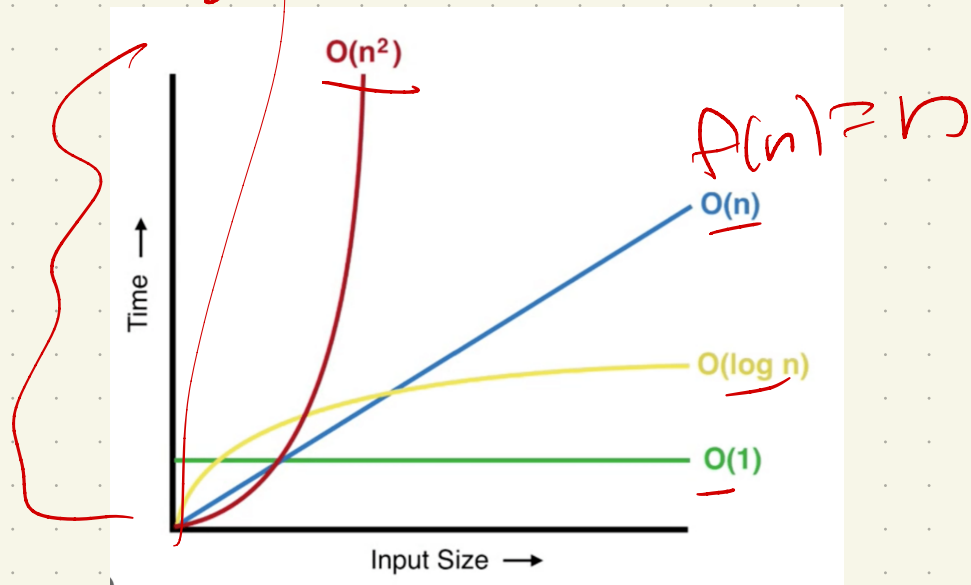
Pick c and n_0 s.t. $5n \geq c \cdot n$
pick $c=2, \forall n > 1,$

$$5n \geq 2n \checkmark$$

Big- Θ

If f is $O(g)$ and $\Omega(g)$ then f is $\Theta(g)$.

This gives a way to nicely "bucket" function size:



Ex: S_n is $\Theta(n)$

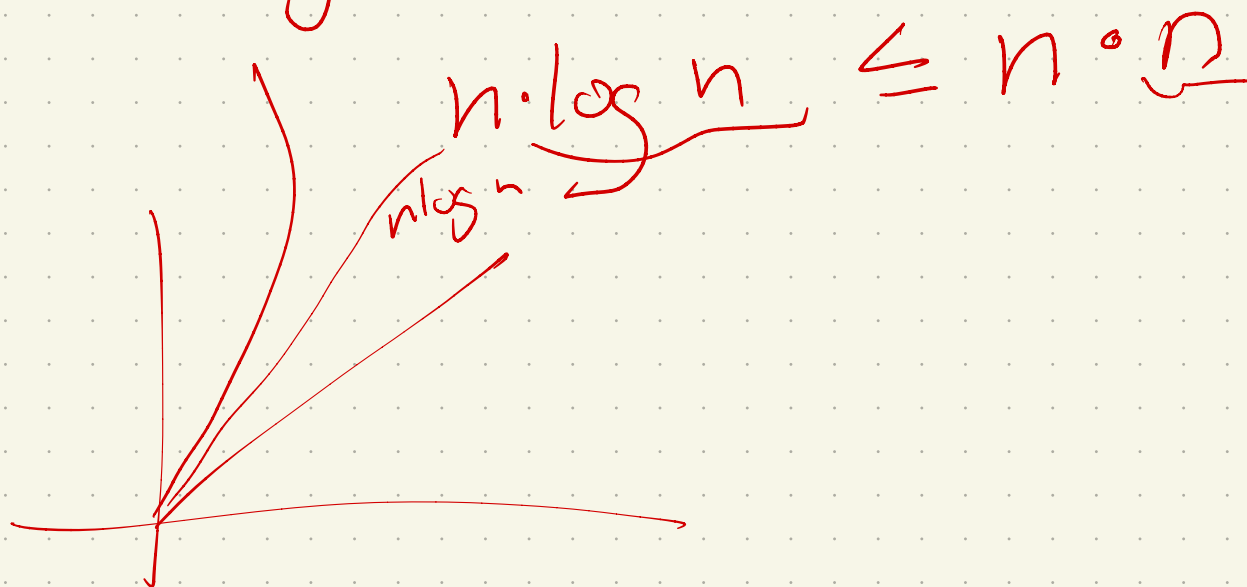
Take $n \log n$:

$n \log_2 n$ is $\Omega(n)$:

$$\log_2 n \geq 1 \quad \forall n \geq 2$$

$$\text{so } n \cdot \log_2 n \geq n \cdot 1$$

$n \log_2$ is $O(n^2)$



Theorem

$$3x^2 + 2x - 5$$

Let $p(n)$ be a degree k polynomial, so

$$p(n) = \underline{a_k} n^{\underline{k}} + \underline{a_{k-1}} n^{\underline{k-1}} + \dots + \underline{a_1} n + \underline{a_0}$$

for constants a_i and $a_k \neq 0$.

Then $p(n) = \Theta(n^k)$.

Proof:

Big- Θ :

Idea: $C = \sum a_i$

use this $C: p(n) \leq C n^k$

Big-Ω:

Choose some fraction
"small enough"

$$p(n) \geq (\text{frac}) n^k$$

$$\underbrace{n^2 - 2n + 5}$$

$$\geq C \cdot n^2$$

$$\frac{1}{2} n^2 \quad \text{pick } n_0 > 5$$

Using this theorem;

Give big- Θ estimates for:

$$\bullet f(x) = \frac{1}{8} x^5 + x^3 + 2 \\ = \Theta(x^5)$$

$$\bullet f(x) = 20000 x^2 - 100000 x \\ = \Theta(x^2)$$

$$\bullet f(x) = \frac{2x^3}{3000} - x + 2 \\ = \Theta(x^3)$$

Other tricks:

Sometimes $f(n) = \sum_{i=1}^n \log_2 n$

How to get Θ -bound?

$$f(n) = \sum_{i=1}^n \log_2 n$$

$$\Rightarrow (\log_2 n + \log_2 n + \dots + \log_2 n)_{i=1}^{i=n}$$

$$= n \log_2 n$$

Another: $\log_2 n$ versus $\log_{10}(n)$.

What is a log?

exp you raise
base to in order
to get ~~thing inside~~
argument

$$x^a \cdot x^b = x^{a+b}$$

$$\log_2 a + \log_2 b = \log_2 abc$$

$$\log_2 2^3 + \log_2 2^4 = 3 + 4 = 7$$

$$\log_2 2^3 \cdot 2^4 = \log_2 2^7$$

Identities:

$$\log_2 a^b = b \log_2 a$$

$$a \leftarrow (x^a)^b = (x^b)^a$$

$$\log_b a = \frac{\log_x a}{\log_x b}$$

$\log_2 n$ versus $\log_{10} n$
goal: big-O!

$$\log_2 n \stackrel{\text{identity}}{=} \frac{\log_{10} n}{\log_{10} 2}$$

$$= \left(\frac{1}{\log_{10} 2} \right) \cdot \log_{10} n$$
$$= O(\log_{10} n)$$

$\log_7 n$ and $\log_{13} n$

$$\hookrightarrow \log_7 n = \frac{\log_{13} n}{\log_3 7}$$

Another:

?

$$\sum_{i=1}^n \log_2 i$$

$$= \log_2 1 + \log_2 2 + \dots + \log_2 n$$

$$= \log_2 (1 \cdot 2 \cdot 3 \cdot \dots \cdot n)$$

$$= \log_2 (n!)$$

$$\leq \log_2 n + \log_2 n + \dots + \log_2 n$$

$$\leq n \log_2 n$$

$$= \underline{\underline{O(n \log n)}}$$

Next time:

Analyzing algorithms with $\log-O$.

Reading: recaps the definitions

→ then uses $\log-O$

Example: ^{assignment} Function with input $A[1..n]$
sum $\leftarrow 0$
For $i \leftarrow 1$ to n
 $sum \leftarrow sum + A[i]$
return sum

$x = 0$
 $x == 0?$

$A[1..n]$

if $(x = 0)$

array
 $A[1], A[2], \dots, A[n]$

Runtime: Count operations!

Can't count directly - depends on A, & how large.

So, use a function:

input array $A[1..n]$
+ $a_1, \dots, a_n$

sum $\leftarrow 0 \leftarrow O(1)$

For $i \leftarrow 1$ to n

sum $\leftarrow$ sum + $A[i]$ $\leftarrow$ one + one store! $O(1)$

return sum

End result:

Runtime

$$T(n) = O(1) + \left(\sum_{i=1}^n 2 \right) + 1$$
$$= O(1) + \underline{O(n)} + O(1) = \underline{O(n)}$$

main:

function (^A send in array)

fu