

Asymptotic Growth of Functions: Big-Oh Notation

Defⁿ Let $f: [0, \infty) \rightarrow \mathbb{R}$
 $g: [0, \infty) \rightarrow \mathbb{R}$ eventually positive (ie. positive for $x > \text{some bound}$)

We say $f(x) = O(g(x))$ if there are constants $C > 0$ and $M > 0$
such that $|f(x)| \leq C g(x)$ for $x > M$.

Intuitively: $f = O(g)$ if g dominates f if allowed ① a constant scaling (C)
② we look at the tail/
far enough out (M)

Write $f = O(g)$
a kind of "upper bd"

Examples. $7x^2 + 3 = O(x^2)$
 $300x^2 = O(x^2)$ $x^2 = O(x^3)$
silly

Examples.

$$f = 500 \log x$$

$$g = x$$

$$f = O(g)$$

$$g \neq O(f)$$

$$f = |\sin x|$$

$$g = |\cos x|$$

$$f \neq O(g)$$

$$g \neq O(f)$$

$$f = |\sin x|$$

$$g = \frac{1}{2}$$

Runtimes .

time complexity = runtime is the # of operations performed in an algorithm as a function of input size

space complexity is # of bits of memory needed to run an algorithm as a function of input size .

Ex. Counting . Input: an integer n in decimal notation

Algorithm: print "Hello!" n times.

n steps, each consist of a constant # of operations (near-)

$$\text{runtime} = \Theta(n) = O(10^M)$$

$$\begin{aligned} \text{input size} &= \# \text{ digits} \\ M &:= \lceil \log_{10}(n) \rceil \end{aligned}$$

exponential runtime

a

First Assessment, Friday, September 16th in class. Covers everything so far (details on website).

Trial Division . for factoring.

Input = n , input size = $M = \log_{10}(n)$
(integer to factor)

Algorithm: ^{trial} divide n by m for all
 $2 \leq m \leq \sqrt{n}$

Goal: find a factor.

Runtime = $O(\sqrt{n})$ trial divisions

Runtime of one trial division is polynomial in M

Runtime = $O(\sqrt{n}) \cdot \text{poly}(M) = O(10^{M/2} \text{poly}(M))$

exponential! (bad)

Runtimes. polynomial: $O(p(M))$ for some polynomial p

$\left(\begin{array}{l} M = \text{input} \\ \text{size} \end{array} \right)$ exponential: $O(e^{p(M)})$ for some polynomial p

Subexponential: runtime $< e^{p(M)}$ for all poly p
"in between"

Examples:

$$e^{c \sqrt{M \log M}}$$

comes up a lot

$$M^{\log M} = e^{(\log M)^2}$$

both are not polynomial, but are faster than exponential

Modular Exponentiation $g^e \pmod{n}$

Input: base g , exponent e , modulus n .

Size: $O(\log(n))$ bits

Steps: $\log(n)$ squarings $g, g^2, g^4, g^8, \dots$

+ $\log(n)$ multiplications to build $g^e = g^1 g^4 g^{32} \dots$

$O(\log n)$ multiplications

→ are each polynomial time ie. $\text{poly}(\log(n))$

↙ school algorithm

So overall: $O(\log n) \cdot \text{poly}(\log(n)) = \text{poly}(\log(n))$

polynomial time.

Discrete Logarithm Problem

Notation: $h = g^a \pmod{p}$ assume g is invertible.

then $a = L_g(h)$ "discrete log base g of $h \pmod{p}$ "
(mod $p-1$)

often we take $0 \leq a < \text{ord}(g) \leq p-1$.

Example. mod 5.

$$2^0 = 1, 2^1 = 2, 2^2 = 4, 2^3 = 3, 2^4 = 1$$

primitive root

$$4^0 = 1, 4^1 = 4, 4^2 = 1$$

order 2

$$\text{So } L_2(3) = 3$$

$$L_2(4) = 2$$

$$L_2(1) = 0$$

$$L_4(3) = \text{not defined.}$$

Logs are exponents so:

$$L_g(h_1 h_2) = L_g(h_1) + L_g(h_2)$$