

ASYMPTOTICS AND FUNCTIONS

Note: This handout summarizes highlights of CLR Chapter 2. See the book for more details.

Motivation

Fine-grained bean counting exposes too much detail for comparing functions.

Want a course-grained way to compare functions that ignores constant factors and focuses on their relative growth in the limit as input sizes get large.

For example, consider:

	n = 1	n = 1,000	n = 1,000,000
$p(n) = 100n + 1000$			
$q(n) = 3n^2 + 2n + 1$			
$r(n) = 0.1n^2$			

How Do Your Functions Grow?

Asymptotic notation is a way of characterizing functions that facilitates comparing their growth in the limit of large inputs. Here is an informal summary of the notation:

Notation	Pronunciation	Loosely
$f \in \omega(g)$	f is way bigger than g	$f > g$
$f \in \Omega(g)$	f is at least as big as g	$f \geq g$
$f \in \Theta(g)$	f is about the same as g	$f = g$
$f \in O(g)$	f is at most as big as g	$f \leq g$
$f \in o(g)$	f is way smaller than g	$f < g$

Notes:

- Each of $\omega(g)$, $\Omega(g)$, $\Theta(g)$, $O(g)$, $o(g)$ denotes a *set* of functions. Thus, $\omega(g)$ is the set of all functions way bigger than g, $\Omega(g)$ is the set of all functions at least as big as g, etc.
- The notation $f \in \Theta(g)$ is really shorthand for $f \in \Omega(g)$ and $f \in O(g)$.
- The phrases “is at least $O(\dots)$ ” and “is at most $O(\dots)$ ” are non-sensical. “Is at least” should be written Ω , and “is at most” should be written O .

Relating the Notations

Here are some of the relationships between the notations:

If $f = O(g)$, then $f = O(g)$.

If $f = o(g)$, then $f = O(g)$.

$O(g) = O(g) = O(g)$

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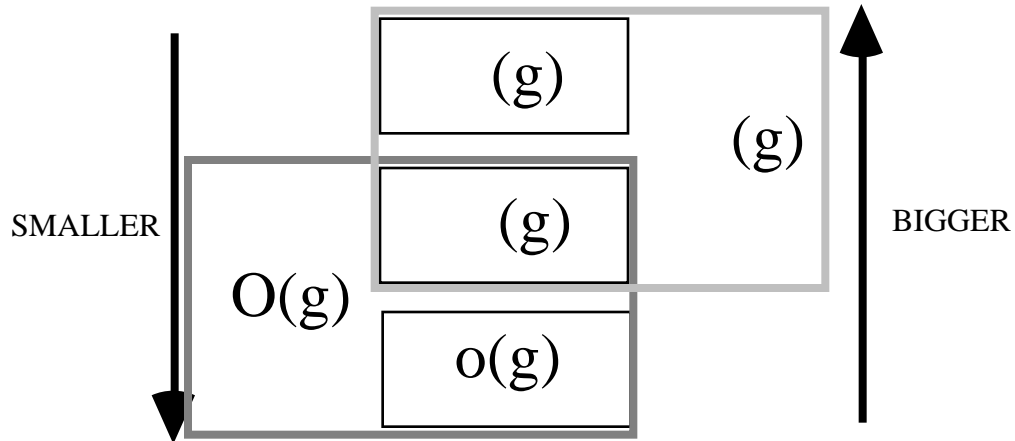
$f = O(g)$ if and only if $g = O(f)$

$f = o(g)$ if and only if $g = O(f)$

$f = O(g)$ if and only if $g = O(f)$

Warning: unlike numbers, not every pair of functions is comparable!

The following diagram depicts some of these relationships:



Formalizing the Non-tight Bounds (o and ω)

$$f = o(g) \text{ if } \lim_n \frac{f(n)}{g(n)} = 0$$

$$f = \omega(g) \text{ if } \lim_n \frac{f(n)}{g(n)} = \infty$$

Formalizing the Tight Bounds (Θ , Ω , and \mathcal{O})

$$\mathcal{O}(g) = \{f \mid \text{there exist positive constants } c, n_0 \text{ such that}$$
$$0 \leq f(n) \leq cg(n) \text{ for all } n \geq n_0.\}$$

Think of this as a game. Suppose you claim that $f = \mathcal{O}(g)$. Then you select c and n_0 , but I try to find a particular n that defeats your claim.

$$\Omega(g) = \{f \mid \text{there exist positive constants } c, n_0 \text{ such that}$$
$$0 < cg(n) \leq f(n) \text{ for all } n \geq n_0.\}$$

$$\Theta(g) = \{f \mid \text{there exist positive constants } c_1, c_2, n_0 \text{ such that}$$
$$c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \geq n_0.\}$$

Relating Exponentials and Logarithms

Key Identities:

- $b^{(\log_b a)} = a = \log_b(b^a)$

Examples:

$$\lg(\text{Error!}) =$$

$$32^{(\lg n)} =$$

Asymptotics Involving Exponentials and Logarithms

How do $\log_2 n$ and $\log_3 n$ compare?

How do 2^n and 3^n compare?

Fact 1: if $a > 0$, $\lim_n \frac{a^n}{n^b} =$

Fact 1 implies $a^n \gg (n^b)$.

In other words: *Any positive exponential grows faster than any polynomial.*

Substituting $\lg n$ for n and 2^a for a in Fact 1 yields:

Fact 2: if $a > 0$, $\lim_n \frac{n^a}{\lg^b n} =$

Fact 2 implies $n^a \gg (\lg^b n)$.

In other words: *Any positive polynomial grows faster than any polylogarithmic function.*

Factorials

Definition: $n! = 1 \cdot 2 \cdot 3 \cdots n$

Stirling's approximation: $n! \sim \sqrt{2\pi n} \frac{n^n}{e^n}$

Asymptotics derivable from Stirling's approximation:

- $n! = o(n^n)$

- $n! = (2^n)$
- $\lg(n!) = (n \lg n)$
