

Algorithms: Complexity (Big Omega and Big Theta)

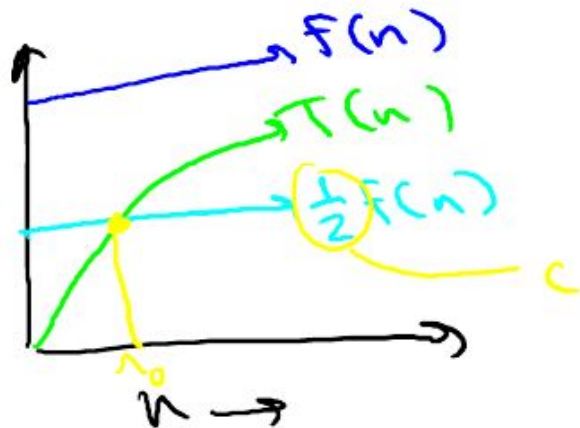
Omega Notation

Definition : $T(n) = \Omega(f(n))$

If and only if there exist constants c, n_0 such that

$$T(n) \geq c \cdot f(n) \quad \forall n \geq n_0.$$

Picture



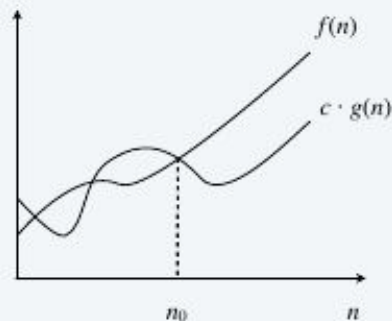
$$T(n) = \Omega(f(n))$$

Big Omega notation

Lower bounds. $f(n)$ is $\Omega(g(n))$ if there exist constants $c > 0$ and $n_0 \geq 0$ such that $f(n) \geq c \cdot g(n) \geq 0$ for all $n \geq n_0$.

Ex. $f(n) = 32n^2 + 17n + 1$.

- $f(n)$ is both $\Omega(n^2)$ and $\Omega(n)$. ← choose $c = 32, n_0 = 1$
- $f(n)$ is not $\Omega(n^3)$.



Typical usage. Any compare-based sorting algorithm requires $\Omega(n \log n)$ compares in the worst case.

Vacuous statement. Any compare-based sorting algorithm requires at least $O(n \log n)$ compares in the worst case.

n $f(n)$	$\lg n$	n	$n \lg n$	n^2	2^n	$n!$
10	0.003 μs	0.01 μs	0.033 μs	0.1 μs	1 μs	3.63 ms
20	0.004 μs	0.02 μs	0.086 μs	0.4 μs	1 ms	77.1 years
30	0.005 μs	0.03 μs	0.147 μs	0.9 μs	1 sec	8.4×10^{15} yrs
40	0.005 μs	0.04 μs	0.213 μs	1.6 μs	18.3 min	
50	0.006 μs	0.05 μs	0.282 μs	2.5 μs	13 days	
100	0.007 μs	0.1 μs	0.644 μs	10 μs	4×10^{13} yrs	
1,000	0.010 μs	1.00 μs	9.966 μs	1 ms		
10,000	0.013 μs	10 μs	130 μs	100 ms		
100,000	0.017 μs	0.10 ms	1.67 ms	10 sec		
1,000,000	0.020 μs	1 ms	19.93 ms	16.7 min		
10,000,000	0.023 μs	0.01 sec	0.23 sec	1.16 days		
100,000,000	0.027 μs	0.10 sec	2.66 sec	115.7 days		
1,000,000,000	0.030 μs	1 sec	29.90 sec	31.7 years		

Figure 2.4: Growth rates of common functions measured in nanoseconds

$\Omega(\dots)$ means a lower bound

- We say “ $T(n)$ is $\Omega(g(n))$ ” if $T(n)$ grows at least as fast as $g(n)$ as n gets large.
- Formally,

$$T(n) = \Omega(g(n))$$
$$\Leftrightarrow$$
$$\exists c, n_0 > 0 \text{ s.t. } \forall n \geq n_0,$$

$$0 \leq c \cdot g(n) \leq T(n)$$

Switched these!!

Example

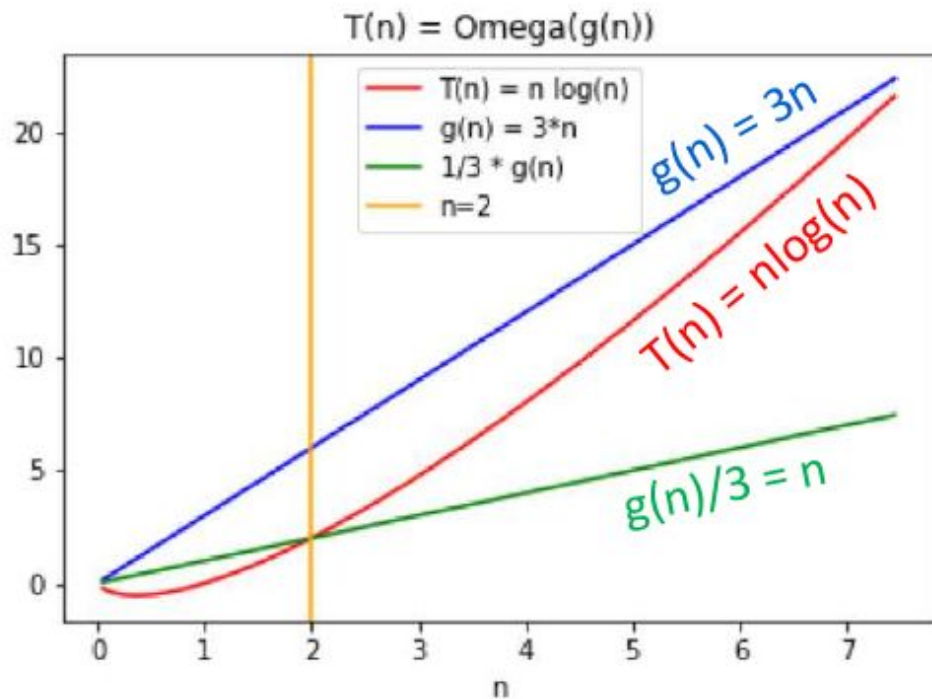
$n \log_2(n) = \Omega(3n)$

$$T(n) = \Omega(g(n))$$

\Leftrightarrow

$$\exists c, n_0 > 0 \text{ s.t. } \forall n \geq n_0,$$

$$0 \leq c \cdot g(n) \leq T(n)$$



- Choose $c = 1/3$
- Choose $n_0 = 2$
- Then

$$\forall n \geq 2,$$

$$0 \leq \frac{3n}{3} \leq n \log_2(n)$$

Example: $\sqrt{n} = \Omega(\lg n)$, with $c = 1$ and $n_0 = 16$.

Examples of functions in $\Omega(n^2)$:

$$n^2$$

$$n^2 + n$$

$$n^2 - n$$

$$1000n^2 + 1000n$$

$$1000n^2 - 1000n$$

Also,

$$n^3$$

$$n^{2.00001}$$

$$n^2 \lg \lg \lg n$$

$$2^{2^n}$$

Theta Notation

Definition : $T(n) = \theta(f(n))$ if and only if
 $T(n) = O(f(n))$ and $T(n) = \Omega(f(n))$

Equivalent : there exist constants c_1, c_2, n_0 such that

$$c_1 f(n) \leq T(n) \leq c_2 f(n)$$

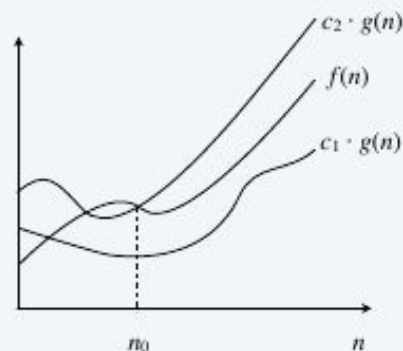
$$\forall n \geq n_0$$

Big Theta notation

Tight bounds. $f(n)$ is $\Theta(g(n))$ if there exist constants $c_1 > 0$, $c_2 > 0$, and $n_0 \geq 0$ such that $0 \leq c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)$ for all $n \geq n_0$.

Ex. $f(n) = 32n^2 + 17n + 1$.

- $f(n)$ is $\Theta(n^2)$. ← choose $c_1 = 32$, $c_2 = 50$, $n_0 = 1$
- $f(n)$ is neither $\Theta(n)$ nor $\Theta(n^3)$.



Typical usage. Mergesort makes $\Theta(n \log n)$ compares to sort n elements.

↗
between $\frac{1}{2} n \log_2 n$
and $n \log_2 n$

$\Theta(\dots)$ means both!

- We say “ $T(n)$ is $\Theta(g(n))$ ” iff both:

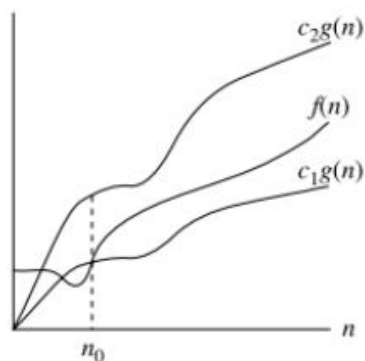
$$T(n) = O(g(n))$$

and

$$T(n) = \Omega(g(n))$$

Θ -notation

$\Theta(g(n)) = \{f(n) : \text{there exist positive constants } c_1, c_2, \text{ and } n_0 \text{ such that}$
 $0 \leq c_1g(n) \leq f(n) \leq c_2g(n) \text{ for all } n \geq n_0\}.$




$g(n)$ is an *asymptotically tight bound* for $f(n)$.


Example: $n^2/2 - 2n = \Theta(n^2)$, with $c_1 = 1/4$, $c_2 = 1/2$, and $n_0 = 8$.

Let $T(n) = \frac{1}{2}n^2 + 3n$. Which of the following statements are true? (Check all that apply.)

$T(n) = O(n)$.

 $T(n) = \Omega(n)$. $[n_0 = 1, c = \frac{1}{2}]$

 $T(n) = \Theta(n^2)$. $[n_0 = 1, c_1 = 1/2, c_2 = 4]$

 $T(n) = O(n^3)$. $[n_0 = 1, c = 4]$

Take-away from examples

- To prove $T(n) = O(g(n))$, you have to come up with c and n_0 so that the definition is satisfied.
- To prove $T(n)$ is **NOT** $O(g(n))$, one way is **proof by contradiction**:
 - Suppose (to get a contradiction) that someone gives you a c and an n_0 so that the definition *is* satisfied.
 - Show that this someone must be lying to you by deriving a contradiction.

Big Oh Examples

$3n^2 - 100n + 6 = O(n^2)$ because $3n^2 > 3n^2 - 100n + 6$

$3n^2 - 100n + 6 = O(n^3)$ because $.01n^3 > 3n^2 - 100n + 6$

$3n^2 - 100n + 6 \neq O(n)$ because $c \cdot n < 3n^2$ when $n > c$

Think of the equality as meaning *in the set of functions*.

Big Omega Examples

$3n^2 - 100n + 6 = \Omega(n^2)$ because $2.99n^2 < 3n^2 - 100n + 6$

$3n^2 - 100n + 6 \neq \Omega(n^3)$ because $3n^2 - 100n + 6 < n^3$

$3n^2 - 100n + 6 = \Omega(n)$ because $10^{10^{10}}n < 3n^2 - 100n + 6$

Big Theta Examples

$3n^2 - 100n + 6 = \Theta(n^2)$ because O and Ω

$3n^2 - 100n + 6 \neq \Theta(n^3)$ because O only

$3n^2 - 100n + 6 \neq \Theta(n)$ because Ω only

Yet more examples

- $n^3 + 3n = O(n^3 - n^2)$
- $n^3 + 3n = \Omega(n^3 - n^2)$
- $n^3 + 3n = \Theta(n^3 - n^2)$

- 3^n is **NOT** $O(2^n)$
- $\log(n) = \Omega(\ln(n))$
- $\log(n) = \Theta(2^{\log\log(n)})$

remember that $\log = \log_2$ in this class.

More Big Oh relatives

Little-Oh Notation

Definition: $T(n) = o(f(n))$ if and only if for all constants $c > 0$, there exists a constant n_0 such that

$$T(n) \leq c \cdot f(n) \quad \forall n \geq n_0$$

Exercise: $\forall k \geq 1, n^{k-1} = o(n^k)$

***o*-notation**

The asymptotic upper bound provided by O -notation may or may not be asymptotically tight. The bound $2n^2 = O(n^2)$ is asymptotically tight, but the bound $2n = O(n^2)$ is not. We use o -notation to denote an upper bound that is not asymptotically tight. We formally define $o(g(n))$ (“little-oh of g of n ”) as the set

$$o(g(n)) = \{f(n) : \text{for any positive constant } c > 0, \text{ there exists a constant } n_0 > 0 \text{ such that } 0 \leq f(n) < cg(n) \text{ for all } n \geq n_0\}.$$

For example, $2n = o(n^2)$, but $2n^2 \neq o(n^2)$.

The definitions of O -notation and o -notation are similar. The main difference is that in $f(n) = O(g(n))$, the bound $0 \leq f(n) \leq cg(n)$ holds for *some* constant $c > 0$, but in $f(n) = o(g(n))$, the bound $0 \leq f(n) < cg(n)$ holds for *all* constants $c > 0$. Intuitively, in o -notation, the function $f(n)$ becomes insignificant relative to $g(n)$ as n approaches infinity; that is,

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0. \tag{3.1}$$

Some authors use this limit as a definition of the o -notation; the definition in this book also restricts the anonymous functions to be asymptotically nonnegative.

ω -notation

By analogy, ω -notation is to Ω -notation as o -notation is to O -notation. We use ω -notation to denote a lower bound that is not asymptotically tight. One way to define it is by

$f(n) \in \omega(g(n))$ if and only if $g(n) \in o(f(n))$.

Formally, however, we define $\omega(g(n))$ (“little-omega of g of n ”) as the set

$\omega(g(n)) = \{f(n) : \text{for any positive constant } c > 0, \text{ there exists a constant } n_0 > 0 \text{ such that } 0 \leq cg(n) < f(n) \text{ for all } n \geq n_0\}$.

For example, $n^2/2 = \omega(n)$, but $n^2/2 \neq \omega(n^2)$. The relation $f(n) = \omega(g(n))$ implies that

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty,$$

if the limit exists. That is, $f(n)$ becomes arbitrarily large relative to $g(n)$ as n approaches infinity.

***o*-notation**

$o(g(n)) = \{f(n) : \text{for all constants } c > 0, \text{ there exists a constant } n_0 > 0 \text{ such that } 0 \leq f(n) < cg(n) \text{ for all } n \geq n_0\}.$

Another view, probably easier to use: $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0.$

$$n^{1.9999} = o(n^2)$$

$$n^2 / \lg n = o(n^2)$$

$$n^2 \neq o(n^2) \text{ (just like } 2 \neq 2)$$

$$n^2 / 1000 \neq o(n^2)$$

***\omega*-notation**

$\omega(g(n)) = \{f(n) : \text{for all constants } c > 0, \text{ there exists a constant } n_0 > 0 \text{ such that } 0 \leq cg(n) < f(n) \text{ for all } n \geq n_0\}.$

Another view, again, probably easier to use: $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty.$

$$n^{2.0001} = \omega(n^2)$$

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

$$n^2 \neq \omega(n^2)$$

Comparing functions

Many of the relational properties of real numbers apply to asymptotic comparisons as well. For the following, assume that $f(n)$ and $g(n)$ are asymptotically positive.

Transitivity:

$$f(n) = \Theta(g(n)) \text{ and } g(n) = \Theta(h(n)) \quad \text{imply} \quad f(n) = \Theta(h(n)) ,$$

$$f(n) = O(g(n)) \text{ and } g(n) = O(h(n)) \quad \text{imply} \quad f(n) = O(h(n)) ,$$

$$f(n) = \Omega(g(n)) \text{ and } g(n) = \Omega(h(n)) \quad \text{imply} \quad f(n) = \Omega(h(n)) ,$$

$$f(n) = o(g(n)) \text{ and } g(n) = o(h(n)) \quad \text{imply} \quad f(n) = o(h(n)) ,$$

$$f(n) = \omega(g(n)) \text{ and } g(n) = \omega(h(n)) \quad \text{imply} \quad f(n) = \omega(h(n)) .$$

Reflexivity:

$$f(n) = \Theta(f(n)) ,$$

$$f(n) = O(f(n)) ,$$

$$f(n) = \Omega(f(n)) .$$

Symmetry:

$f(n) = \Theta(g(n))$ if and only if $g(n) = \Theta(f(n))$.

Transpose symmetry:

$f(n) = O(g(n))$ if and only if $g(n) = \Omega(f(n))$,

$f(n) = o(g(n))$ if and only if $g(n) = \omega(f(n))$.

Because these properties hold for asymptotic notations, we can draw an analogy between the asymptotic comparison of two functions f and g and the comparison of two real numbers a and b :

$f(n) = O(g(n))$ is like $a \leq b$,

$f(n) = \Omega(g(n))$ is like $a \geq b$,

$f(n) = \Theta(g(n))$ is like $a = b$,

$f(n) = o(g(n))$ is like $a < b$,

$f(n) = \omega(g(n))$ is like $a > b$.

We say that $f(n)$ is *asymptotically smaller* than $g(n)$ if $f(n) = o(g(n))$, and $f(n)$ is *asymptotically larger* than $g(n)$ if $f(n) = \omega(g(n))$.

Where Does Notation Come From?

“On the basis of the issues discussed here, I propose that members of SIGACT, and editors of computer science and mathematics journals, adopt the O , Ω , and Θ notations as defined above, unless a better alternative can be found reasonably soon”.

-D. E. Knuth, “Big Omicron and Big Omega and Big Theta”, SIGACT News, 1976. Reprinted in “Selected Papers on Analysis of Algorithms.”

Suggested Reading

- Algorithms (CLRS)
 - ◆ Chapter 3
 - Section 3.1
- Algorithm illuminated (Part 1) by Tim Roughgarden
 - ◆ Chapter 2
 - Section 2.4

