

Asymptotic Notation: $O()$, $o()$, $\Omega()$, $\omega()$, and $\Theta()$

The Idea

“Big-O” notation was introduced in P. Bachmann’s 1892 book *Analytische Zahlentheorie*. He used it to say things like “ x is $O(\frac{n}{2})$ ” instead of “ $x \approx \frac{n}{2}$.” The notation works well to compare algorithm efficiencies because we want to say that the growth of effort of a given algorithm approximates the shape of a standard function.

The Definitions

Big-O ($O()$) is one of five standard asymptotic notations. In practice, Big-O is used as a tight upper-bound on the growth of an algorithm’s effort (this effort is described by the function $f(n)$), even though, as written, it can also be a loose upper-bound. To make its role as a tight upper-bound more clear, “Little-o” ($o()$) notation is used to describe an upper-bound that cannot be tight.

Definition (Big-O, $O()$): Let $f(n)$ and $g(n)$ be functions that map positive integers to positive real numbers. We say that $f(n)$ is $O(g(n))$ (or $f(n) \in O(g(n))$) if there exists a real constant $c > 0$ and there exists an integer constant $n_0 \geq 1$ such that $f(n) \leq c * g(n)$ for every integer $n \geq n_0$.

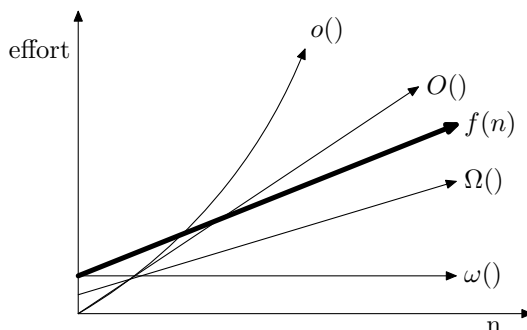
Definition (Little-o, $o()$): Let $f(n)$ and $g(n)$ be functions that map positive integers to positive real numbers. We say that $f(n)$ is $o(g(n))$ (or $f(n) \in o(g(n))$) if for any real constant $c > 0$, there exists an integer constant $n_0 \geq 1$ such that $f(n) < c * g(n)$ for every integer $n \geq n_0$.

On the other side of $f(n)$, it is convenient to define parallels to $O()$ and $o()$ that provide tight and loose lower bounds on the growth of $f(n)$. “Big-Omega” ($\Omega()$) is the tight lower bound notation, and “little-omega” ($\omega()$) describes the loose lower bound.

Definition (Big-Omega, $\Omega()$): Let $f(n)$ and $g(n)$ be functions that map positive integers to positive real numbers. We say that $f(n)$ is $\Omega(g(n))$ (or $f(n) \in \Omega(g(n))$) if there exists a real constant $c > 0$ and there exists an integer constant $n_0 \geq 1$ such that $f(n) \geq c \cdot g(n)$ for every integer $n \geq n_0$.

Definition (Little-Omega, $\omega()$): Let $f(n)$ and $g(n)$ be functions that map positive integers to positive real numbers. We say that $f(n)$ is $\omega(g(n))$ (or $f(n) \in \omega(g(n))$) if for any real constant $c > 0$, there exists an integer constant $n_0 \geq 1$ such that $f(n) > c \cdot g(n)$ for every integer $n \geq n_0$.

This graph should help you visualize the relationships between these notations:



(Continued...)

These definitions have far more similarities than differences. Here's a table that summarizes the key restrictions in these four definitions:

Definition	$\exists c > 0$	$\exists n_0 \geq 1$	$f(n) \leq c \cdot g(n)$
$O()$	\exists	\exists	\leq
$o()$	\forall	\exists	$<$
$\Omega()$	\exists	\exists	\geq
$\omega()$	\forall	\exists	$>$

While $\Omega()$ and $\omega()$ aren't often used to describe algorithms, we can build on them (on $\Omega()$ in particular) to define a notation that describes a combination of $O()$ and $\Omega()$: "Big-Theta" ($\Theta()$). When we say that an algorithm is $\Theta(g(n))$, we are saying that $g(n)$ is both a tight upper-bound and a tight lower-bound on the growth of the algorithm's effort.

Definition (Big-Theta, $\Theta()$): Let $f(n)$ and $g(n)$ be functions that map positive integers to positive real numbers. We say that $f(n)$ is $\Theta(g(n))$ (or $f(n) \in \Theta(g(n))$) if and only if $f(n) \in O(g(n))$ and $f(n) \in \Omega(g(n))$.

Application Examples

Here are a few examples that show how the definitions should be applied.

1. Let $f(n) = 7n + 8$ and $g(n) = n$. Is $f(n) \in O(g(n))$?

For $7n + 8 \in O(n)$, we have to find c and n_0 such that $7n + 8 \leq c \cdot n$, $\forall n \geq n_0$. By inspection, it's clear that c must be larger than 7. Let $c = 8$.

Now we need a suitable n_0 . In this case, $f(8) = 8 \cdot g(8)$. Because the definition of $O()$ requires that $f(n) \leq c \cdot g(n)$, we can select $n_0 = 8$, or any integer above 8 – they will all work.

We have identified values for the constants c and n_0 such that $7n + 8$ is $\leq c \cdot n$ for every $n \geq n_0$, so we can say that $7n + 8$ is $O(n)$.

(But how do we know that this will work for every n above 7? We can prove by induction that $7n + 8 \leq 8n$, $\forall n \geq 8$. Be sure that you can write such proofs if asked!)

2. Let $f(n) = 7n + 8$ and $g(n) = n$. Is $f(n) \in o(g(n))$?

In order for that to be true, for any c , we have to be able to find an n_0 that makes $f(n) < c \cdot g(n)$ asymptotically true.

However, this doesn't seem likely to be true. Both $7n + 8$ and n are linear, and $o()$ defines loose upper-bounds. To show that it's not true, all we need is a counter-example.

Because any $c > 0$ must work for the claim to be true, let's try to find a c that won't work. Let $c = 100$. Can we find a positive n_0 such that $7n + 8 < 100n$? Sure; let $n_0 = 10$. Try again!

Let's try $c = \frac{1}{100}$. Can we find a positive n_0 such that $7n + 8 < \frac{n}{100}$? No; only negative values will work. Therefore, $7n + 8 \notin o(n)$, meaning $g(n) = n$ is not a loose upper-bound on $7n + 8$.

3. Is $7n + 8 \in o(n^2)$?

Again, to claim this we need to be able to argue that for any c , we can find an n_0 that makes $7n + 8 < c \cdot n^2$. Let's try examples again to make our point, keeping in mind that we need to show that we can find an n_0 for any c .

If $c = 100$, the inequality is clearly true. If $c = \frac{1}{100}$, we'll have to use a little more imagination, but we'll be able to find an n_0 . (Try $n_0 = 1000$.) From these examples, the conjecture *appears* to be correct.

To prove this, we need calculus. For $g(n)$ to be a loose upper-bound on $f(n)$, it must be the case that $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$. Here, $\lim_{n \rightarrow \infty} \frac{7n+8}{n^2} = \lim_{n \rightarrow \infty} \frac{7}{2n} = 0$ (by l'Hôpital). Thus, $7n + 8 \in o(n^2)$.