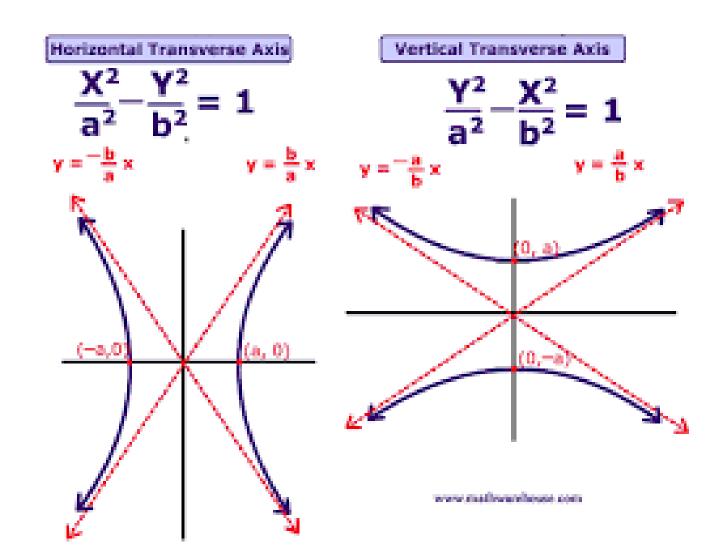
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- > Usually, an algorithm that is asymptotically more efficient will be the best choice for all but very small inputs.
- > We will use asymptotic notation primarily to describe the running times of algorithms.

- > Which are different asymptotic notations?
 - Θ-notation
 - O-notation
 - Ω -notation
 - o-notation
 - ω -notation

 \succ The notations we use to describe the asymptotic running time of an algorithm are defined in terms of functions whose domains are the set of natural numbers N = $\{0, 1, 2, ...\}$.

$\triangleright \Theta$ -notation

For a given function g(n), we denote by $\Theta(g(n))$ the set of functions

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

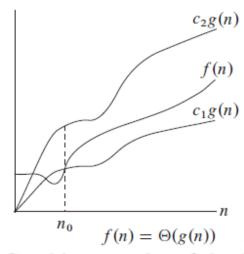


Figure 3.1 Graphic examples of the Θ , O, and Ω notations.

In each part, the value of n_0 shown is the minimum possible value; any greater value would also work. (a) Θ -notation bounds a function to within constant factors. We write $f(n) = \Theta(g(n))$ if there exist positive constants n_0 , c_1 , and c_2 such that at and to the right of n_0 , the value of f(n) always lies between $c_1g(n)$ and $c_2g(n)$ inclusive.

> Θ-notation

A function f(n) belongs to the set $\Theta(g(n))$ if there exist positive constants c_1 and c_2 such that it can be "sandwiched" between $c_1g(n)$ and $c_2g(n)$, for sufficiently large n.

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- " $f(n) = \Theta(g(n))$ " to express the same notion. Instead, we will usually write
- You might be confused because we abuse equality in this way, but we shall see later in this section that doing so has its advantages.

> Θ-notation

Figure 3.1(a) gives an intuitive picture of functions f(n) and g(n), where $f(n) = \theta(g(n))$.

> Θ-notation

- Figure 3.1(a) gives an intuitive picture of functions f(n) and g(n), where $f(n) = \theta(g(n))$.
- For all values of n at and to the right of n_0 , the value of f(n) lies at or above $c_1g(n)$ and at or below $c_2g(n)$. In other words, for all $n \ge n_0$, the function f(n) is equal to g(n) to within a constant factor. We say that g(n) is an asymptotically tight bound for f(n).

> Θ-notation

In Chapter 2, we introduced an informal notion of Θ -notation that amounted to throwing away lower-order terms and ignoring the leading coefficient of the highest-order term. Let us briefly justify this intuition by using the formal definition to show that $\frac{1}{2}n^2 - 3n = \Theta(n^2)$. To do so, we must determine positive

constants c_1 , c_2 , and n_0 such that

$c_1 n^2$	<	$\frac{1}{2}n^2$	-3n	<	c_2n^2
•		2		_	_

for all $n \ge n_0$. Dividing by n^2 yields

$$c_1 \leq \frac{1}{2} - \frac{3}{n} \leq c_2$$
.

n	f(n)	n	f(n)
1	-5/2	6	0
2	-1	7	1/14
3	-1/2	8	1/8
4	-1/4	9	1/6
5	-1/10	10	1/5

We can make the right-hand inequality hold for any value of $n \ge 1$ by choosing any constant $c_2 \ge 1/2$. Likewise, we can make the left-hand inequality hold for any value of $n \ge 7$ by choosing any constant $c_1 \le 1/14$. Thus, by choosing $c_1 = 1/14$, $c_2 = 1/2$, and $n_0 = 7$, we can verify that $\frac{1}{2}n^2 - 3n = \Theta(n^2)$. Certainly, other choices for the constants exist, but the important thing is that *some* choice exists. Note that these constants depend on the function $\frac{1}{2}n^2 - 3n$; a different function belonging to $\Theta(n^2)$ would usually require different constants.

> Θ-notation

We can also use the formal definition to verify that $6n^3 \neq \Theta(n^2)$. Suppose for the purpose of contradiction that c_2 and n_0 exist such that $6n^3 \leq c_2n^2$ for all $n \geq n_0$. But then dividing by n^2 yields $n \leq c_2/6$, which cannot possibly hold for arbitrarily large n, since c_2 is constant.

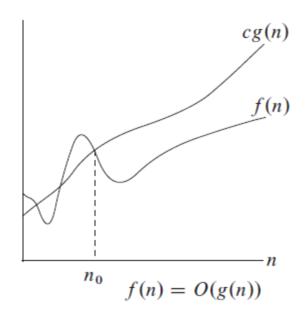
> Θ-notation

Since any constant is a degree-0 polynomial, we can express any constant function as $\Theta(n^0)$, or $\Theta(1)$. This latter notation is a minor abuse, however, because the expression does not indicate what variable is tending to infinity.² We shall often use the notation $\Theta(1)$ to mean either a constant or a constant function with respect to some variable.

> O-notation

The Θ -notation asymptotically bounds a function from above and below. When we have only an *asymptotic upper bound*, we use O-notation. For a given function g(n), we denote by O(g(n)) (pronounced "big-oh of g of n" or sometimes just "oh of g of n") the set of functions

 $O(g(n)) = \{f(n) : \text{ there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \le f(n) \le cg(n) \text{ for all } n \ge n_0 \}$.



> *O*-notation

> f(n) = O(g(n)) means that a function f(n) belongs to the set O(g(n)).

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- > So, any quadratic function an² + bn + c, where a > 0 is not only in $\theta(n^2)$ but also in $O(n^2)$.
- > It may be surprising that when a > 0, any linear function an + b is in $O(n^2)$. This can be verified by taking c = a + |b| and $n_0 = \max(1, -b/a)$.

> O-notation

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- > So, any quadratic function an² + bn + c, where a > 0 is not only in $\theta(n^2)$ but also in $O(n^2)$.
- > It may be surprising that when a > 0, any linear function an + b is in $O(n^2)$. This can be verified by taking c = a + |b| and $n_0 = \max(1, -b/a)$.
- When we write f(n) = O(g(n)), we are merely claiming that some constant multiple of g(n) is an asymptotic upper bound on f(n), with no claim of how tight an upper bound it is.

> *O*-notation

> Using O-notation, we can often describe the running time of an algorithm merely by inspecting the algorithm's overall structure.

> *O*-notation

- > Using O-notation, we can often describe the running time of an algorithm merely by inspecting the algorithm's overall structure.
- For example, the doubly nested loop structure of the insertion sort algorithm from Chapter 2 immediately yields an $O(n^2)$ upper bound on the worst-case running time: the cost of each iteration of the inner loop is bounded from above by O(1) (constant), the indices i and j are both at most n, and the inner loop is executed at most once for each of the n^2 pairs of values for i and j.

> *O*-notation

Since O-notation describes an upper bound, when we use it to bound the worstcase running time of an algorithm, we have a bound on the running time of the algorithm on every input—the blanket statement we discussed earlier.

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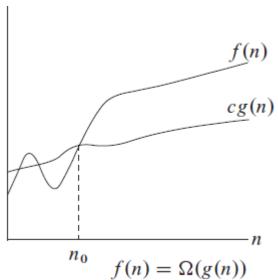
The $\Theta(n^2)$ bound on the worst-case running time of insertion sort, however, does not imply a $\Theta(n^2)$ bound on the running time of insertion sort on *every* input. For example, we saw in Chapter 2 that when the input is already sorted, insertion sort runs in $\Theta(n)$ time.

$\triangleright \Omega$ -notation

Just as O-notation provides an asymptotic *upper* bound on a function, Ω -notation provides an *asymptotic lower bound*. For a given function g(n), we denote by $\Omega(g(n))$ (pronounced "big-omega of g of n" or sometimes just "omega of g of n") the set of functions

$$\Omega(g(n)) = \{f(n) : \text{ there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \le cg(n) \le f(n) \text{ for all } n \ge n_0 \}$$
.

Figure 3.1(c) shows the intuition behind Ω -notation. For all values n at or to the right of n_0 , the value of f(n) is on or above cg(n).



> Theorem 3.1

For any two functions f(n) and g(n), we have $f(n) = \Theta(g(n))$ if and only if f(n) = O(g(n)) and $f(n) = \Omega(g(n))$.

> Theorem 3.1

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For any two functions f(n) and g(n), we have f(n) = \Theta(g(n)) if and only if f(n) = O(g(n)) and f(n) = \Omega(g(n)).
```

> When best case running time of some algorithm is $\Omega(g(n))$, it means that running time of that algorithm for every input is also $\Omega(g(n))$

> Asymptotic notation in equations and inequalities

We have already seen how asymptotic notation can be used within mathematical formulas. For example, in introducing O-notation, we wrote " $n = O(n^2)$." We might also write $2n^2 + 3n + 1 = 2n^2 + \Theta(n)$. How do we interpret such formulas?

> Asymptotic notation in equations and inequalities

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When the asymptotic notation stands alone (that is, not within a larger formula) on the right-hand side of an equation (or inequality), as in $n = O(n^2)$, we have already defined the equal sign to mean set membership: $n \in O(n^2)$. In general, however, when asymptotic notation appears in a formula, we interpret it as standing for some anonymous function that we do not care to name. For example, the formula $2n^2 + 3n + 1 = 2n^2 + \Theta(n)$ means that $2n^2 + 3n + 1 = 2n^2 + f(n)$, where f(n) is some function in the set $\Theta(n)$. In this case, we let f(n) = 3n + 1, which indeed is in $\Theta(n)$.

\triangleright 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

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

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

\triangleright ω -notation

By analogy, ω -notation is to Ω -notation as σ -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

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

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

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

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