

## 5. ASYMPTOTIC NOTATIONS AND ITS PROPERTIES

Asymptotic notation is a notation, which is used to take meaningful statement about the efficiency of a program.

The efficiency analysis framework concentrates on the order of growth of an algorithm's basic operation count as the principal indicator of the algorithm's efficiency.

To compare and rank such orders of growth, computer scientists use three notations, they are:

- $O$  - Big oh notation
- $\Omega$  - Big omega notation
- $\Theta$  - Big theta notation

Let  $t(n)$  and  $g(n)$  can be any nonnegative functions defined on the set of natural numbers. The algorithm's running time  $t(n)$  usually indicated by its basic operation count  $C(n)$ , and  $g(n)$ , some simple function to compare with the count.

### Example 1:

|                         |   |                                     |
|-------------------------|---|-------------------------------------|
| $n \in O(n^2)$ ,        | $100n + 5 \in O(n^2)$ ,                 | $\frac{1}{2}n(n - 1) \in O(n^2)$ .  |
| $n^3 \notin O(n^2)$ ,   | $0.00001n^3 \notin O(n^2)$ ,            | $n^4 + n + 1 \notin O(n^2)$ .       |
| $n^3 \in \Omega(n^2)$ , | $\frac{1}{2}n(n - 1) \in \Omega(n^2)$ , | but $100n + 5 \notin \Omega(n^2)$ . |

where  $g(n) = n^2$ .

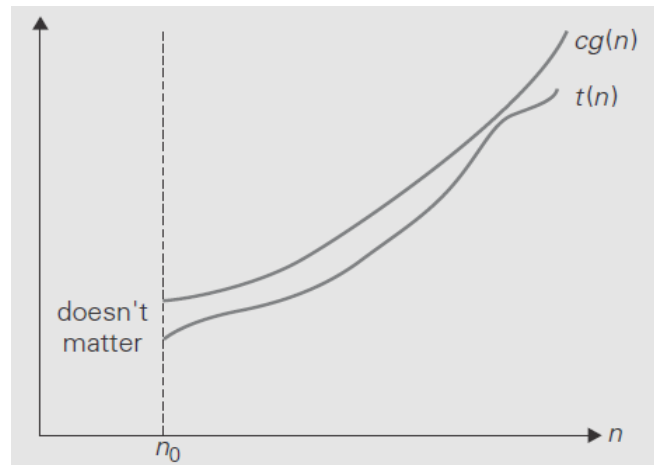
### (I) $O$ - Big oh notation

A function  $t(n)$  is said to be in  $O(g(n))$ , denoted  $t(n) \in O(g(n))$ , if  $t(n)$  is bounded above by some constant multiple of  $g(n)$  for all large  $n$ , i.e., if there exist some positive constant  $c$  and some nonnegative integer  $n_0$  such that

$$t(n) \leq c \cdot g(n) \text{ for } n \geq n_0.$$

Where  $t(n)$  and  $g(n)$  are nonnegative functions defined on the set of natural numbers.

$O$  = Asymptotic upper bound = Useful for worst case analysis = Loose bound



**FIGURE 1.5** Big-oh notation:  $t(n) \in O(g(n))$ .

**Example2:** Prove the assertions  $100n + 5 \in O(n^2)$ .

$$\begin{aligned} \text{Proof: } 100n + 5 &\leq 100n + n \text{ (for all } n \geq 5) \\ &= 101n \\ &\leq 101n^2 \quad (\forall n \leq n^2) \end{aligned}$$

Since, the definition gives us a lot of freedom in choosing specific values for constants  $c$  and  $n_0$ . We have  $c=101$  and  $n_0=5$

**Example3:** Prove the assertions  $100n + 5 \in O(n)$ .

$$\begin{aligned} \text{Proof: } 100n + 5 &\leq 100n + 5n \text{ (for all } n \geq 1) \\ &= 105n \\ \text{i.e., } 100n + 5 &\leq 105n \\ \text{i.e., } t(n) &\leq cg(n) \end{aligned}$$

$100n + 5 \in O(n)$  with  $c=105$  and  $n_0=1$

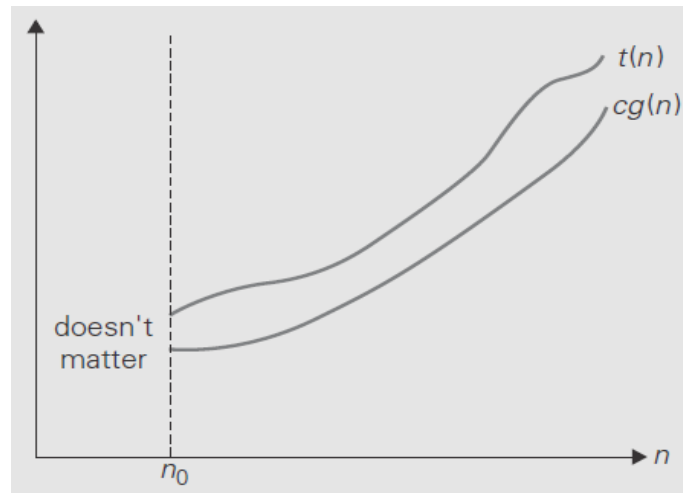
### (i) $\Omega$ - Big omega notation

A function  $t(n)$  is said to be in  $\Omega(g(n))$ , denoted  $t(n) \in \Omega(g(n))$ , if  $t(n)$  is bounded below by some positive constant multiple of  $g(n)$  for all large  $n$ , i.e., if there exist some positive constant  $c$  and some nonnegative integer  $n_0$  such that

$$t(n) \geq cg(n) \text{ for all } n \geq n_0.$$

Where  $t(n)$  and  $g(n)$  are nonnegative functions defined on the set of natural numbers.

$\Omega$  = Asymptotic lower bound = Useful for best case analysis = Loose bound



**FIGURE 1.6** Big-omega notation:  $t(n) \in \Omega(g(n))$ .

**Example4:** Prove the assertions  $n^3+10n^2+4n+2 \in \Omega(n^2)$ .

Proof:  $n^3+10n^2+4n+2 \geq n^2$  (for all  $n \geq 0$ )

i.e., by definition  $t(n) \geq cg(n)$ , where  $c=1$  and  $n_0=0$

**(ii)  $\Theta$  - Big theta notation**

A function  $t(n)$  is said to be in  $\Theta(g(n))$ , denoted  $t(n) \in \Theta(g(n))$ , if  $t(n)$  is bounded both above and below by some positive constant multiples of  $g(n)$  for all large  $n$ , i.e., if there exist some positive constants  $c_1$  and  $c_2$  and some nonnegative integer  $n_0$  such that

$$c_2g(n) \leq t(n) \leq c_1g(n) \text{ for all } n \geq n_0.$$

Where  $t(n)$  and  $g(n)$  are non-negative functions defined on the set of natural numbers.

$\Theta$  = Asymptotic tight bound = Useful for average case analysis

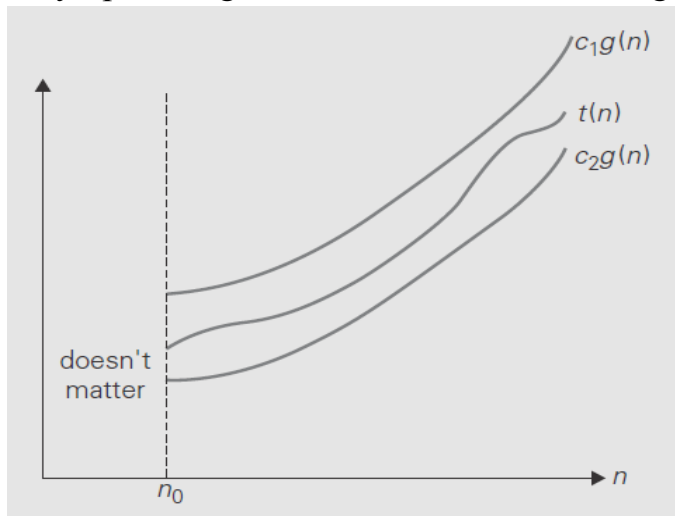


FIGURE 1.7 Big-theta notation:  $t(n) \in \Theta(g(n))$ .

**Example5:** Prove the assertions  $\frac{1}{2}n(n-1) \in \Theta(n^2)$ .

Proof: First prove the right inequality (the upperbound):

$$\frac{1}{2}n(n-1) = \frac{1}{2}n^2 - \frac{1}{2}n \leq \frac{1}{2}n^2 \text{ for all } n \geq 0.$$

Second, we prove the left inequality (the lower bound):

$$\frac{1}{2}n(n-1) = \frac{1}{2}n^2 - \frac{1}{2}n \geq \frac{1}{2}n^2 - \lceil \frac{1}{2}n \rceil \lceil \frac{1}{2}n \rceil \text{ for all } n \geq 2.$$

$$\begin{aligned} \Theta \quad & \frac{1}{2}n(n-1) \leq \frac{1}{2}n^2 \\ \text{i.e.,} \quad & \frac{1}{4}n^2 \leq \frac{1}{2}n(n-1) \leq \frac{1}{2}n^2 \\ \text{Hence,} \quad & \frac{1}{4}n(n-1) \in \Theta(n^2) \end{aligned}$$

where  $c_2 = \frac{1}{2}$ ,  $c_1 = \frac{1}{4}$ , and  $n_0 = 2$

**Note:** asymptotic notation can be thought of as "relational operators" for functions similar to the corresponding relational operators for values.

$$\Rightarrow \Theta(), \quad \leq \Theta(), \quad \geq \Theta(), \quad < \Theta(), \quad > \Theta()$$

### Useful Property Involving the Asymptotic Notations

The following property, in particular, is useful in analyzing algorithms that comprise two consecutively executed parts.

**THEOREM:** If  $t_1(n) \in O(g_1(n))$  and  $t_2(n) \in O(g_2(n))$ , then  $t_1(n) + t_2(n) \in O(\max\{g_1(n), g_2(n)\})$ .  
(The analogous assertions are true for the  $\Omega$  and  $\Theta$  notations as well.)

**PROOF:** The proof extends to orders of growth the following simple fact about four arbitrary real numbers  $a_1, b_1, a_2, b_2$ : if  $a_1 \leq b_1$  and  $a_2 \leq b_2$ , then  $a_1 + a_2 \leq 2 \max\{b_1, b_2\}$ .

Since  $t_1(n) \in O(g_1(n))$ , there exist some positive constant  $c_1$  and some nonnegative integer  $n_1$  such that

$$t_1(n) \leq c_1 g_1(n) \text{ for all } n \geq n_1.$$

Similarly, since  $t_2(n) \in O(g_2(n))$ ,

$$t_2(n) \leq c_2 g_2(n) \text{ for all } n \geq n_2.$$

Let us denote  $c_3 = \max\{c_1, c_2\}$  and consider  $n \geq \max\{n_1, n_2\}$  so that we can use both inequalities. Adding them yields the following:

$$\begin{aligned} t_1(n) + t_2(n) &\leq c_1 g_1(n) + c_2 g_2(n) \\ &\leq c_3 g_1(n) + c_3 g_2(n) \\ &= c_3 [g_1(n) + g_2(n)] \\ &\leq c_3 2 \max\{g_1(n), g_2(n)\}. \end{aligned}$$

Hence,  $t_1(n) + t_2(n) \in O(\max\{g_1(n), g_2(n)\})$ , with the constants  $c$  and  $n_0$  required by the definition  $O$  being  $2c_3 = 2 \max\{c_1, c_2\}$  and  $\max\{n_1, n_2\}$ , respectively.

The property implies that the algorithm's overall efficiency will be determined by the part with a higher order of growth, i.e., its least efficient part.

**THEOREM:** If  $t_1(n) \in O(g_1(n))$  and  $t_2(n) \in O(g_2(n))$ , then  $t_1(n) + t_2(n) \in O(\max\{g_1(n), g_2(n)\})$ .

### Basic rules of sum manipulation

$$\sum_{i=l}^u c a_i = c \sum_{i=l}^u a_i, \quad \text{(R1)}$$

$$\sum_{i=l}^u (a_i \pm b_i) = \sum_{i=l}^u a_i \pm \sum_{i=l}^u b_i, \quad \text{(R2)}$$

### Summation formulas

$$\sum_{i=l}^u 1 = u - l + 1 \quad \text{where } l \leq u \text{ are some lower and upper integer limits, (S1)}$$

$$\sum_{i=0}^n i = \sum_{i=1}^n i = 1 + 2 + \dots + n = \frac{n(n+1)}{2} \approx \frac{1}{2}n^2 \in \Theta(n^2). \quad \text{(S2)}$$