

# Asymptotic Notations Overview

**Asymptotic notations** provide a simple, communicable way to represent the **time complexity** of algorithms using mathematical function classes.

They belong to mathematics but are used in algorithms for comparing functions and showing growth rates.

**Three main notations:**

- **Big O:** Upper bound
- **Big Omega ( $\Omega$ ):** Lower bound
- **Theta ( $\Theta$ ):** Tight/average bound

**Key principle:** Any algorithm's time complexity belongs to one (or a multiple) of these; prefer the tightest bound for usefulness.

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## Big O Notation (Upper Bound)

**Definition:**  $f(n) = O(g(n))$  if there exist positive constants  $C$  and  $n_0$  such that  $f(n) \leq C \cdot g(n)$  for all  $n \geq n_0$ .

### Step-by-Step Understanding

1. **Identify dominant term** in  $f(n)$  (highest growth rate).
2. **Choose single-term  $g(n)$**  that bounds  $f(n)$  from above (can include coefficient  $C$ ).
3. **Verify inequality** holds for all  $n \geq n_0$ .

**Example:**  $f(n) = 2n + 3$

- Convert lower terms to dominant:  $2n + 3 \leq 5n$  (for  $n \geq 1$ ).
- Here,  $C=5$ ,  $g(n)=n$ ,  $n_0=1$ .
- Thus,  $f(n) = O(n)$ .

**Alternative (looser) bounds:**

- $2n + 3 \leq 5n^2 \rightarrow f(n) = O(n^2)$  (true but less tight).

- $f(n) = O(2^n)$  also true (any faster-growing function works).
- **Wrong:**  $O(\log n)$  (grows slower, violates upper bound).

**Best practice:** Choose **tightest (closest) upper bound** for usefulness.

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## Big Omega Notation (Lower Bound)

**Definition:**  $f(n) = \Omega(g(n))$  if there exist positive constants  $C$  and  $n_0$  such that  $f(n) \geq C \cdot g(n)$  for all  $n \geq n_0$ .

**Difference from Big O:** Inequality reverses ( $\geq$  instead of  $\leq$ ).

### Step-by-Step Understanding

1. **Choose  $g(n)$**  that  $f(n)$  grows at least as fast as.
2. **Use dominant term** with minimal  $C$  (often 1).

**Example:**  $f(n) = 2n + 3$

- $2n + 3 \geq 1 \cdot n$  (for  $n \geq 0$  or  $1$ ).
- Thus,  $f(n) = \Omega(n)$ .

**Valid looser bounds:**

- $f(n) = \Omega(\log n)$  (true, since  $n$  grows faster).
- $\Omega(\sqrt{n})$  also true.

**Wrong:**  $\Omega(n^2)$  ( $f(n)$  doesn't grow that fast).

**Best practice:** Tightest lower bound is most useful.

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## Theta Notation (Tight Bound)

**Definition:**  $f(n) = \Theta(g(n))$  if there exist positive constants  $C_1$ ,  $C_2$ ,  $n_0$  such that  $C_1 \cdot g(n) \leq f(n) \leq C_2 \cdot g(n)$  for all  $n \geq n_0$ .

**Combines both bounds** using **same  $g(n)$** .

### Step-by-Step Understanding

1. **Find lower bound:**  $f(n) \geq C_1 \cdot g(n)$ .
2. **Find upper bound:**  $f(n) \leq C_2 \cdot g(n)$ .
3. **Same  $g(n)$  on both sides** → tight bound.

**Example:**  $f(n) = 2n + 3$

- Lower:  $2n + 3 \geq 1 \cdot n$
- Upper:  $2n + 3 \leq 5 \cdot n$
- Thus,  $1n \leq 2n + 3 \leq 5n \rightarrow f(n) = \Theta(n)$ .

**Restrictions:**

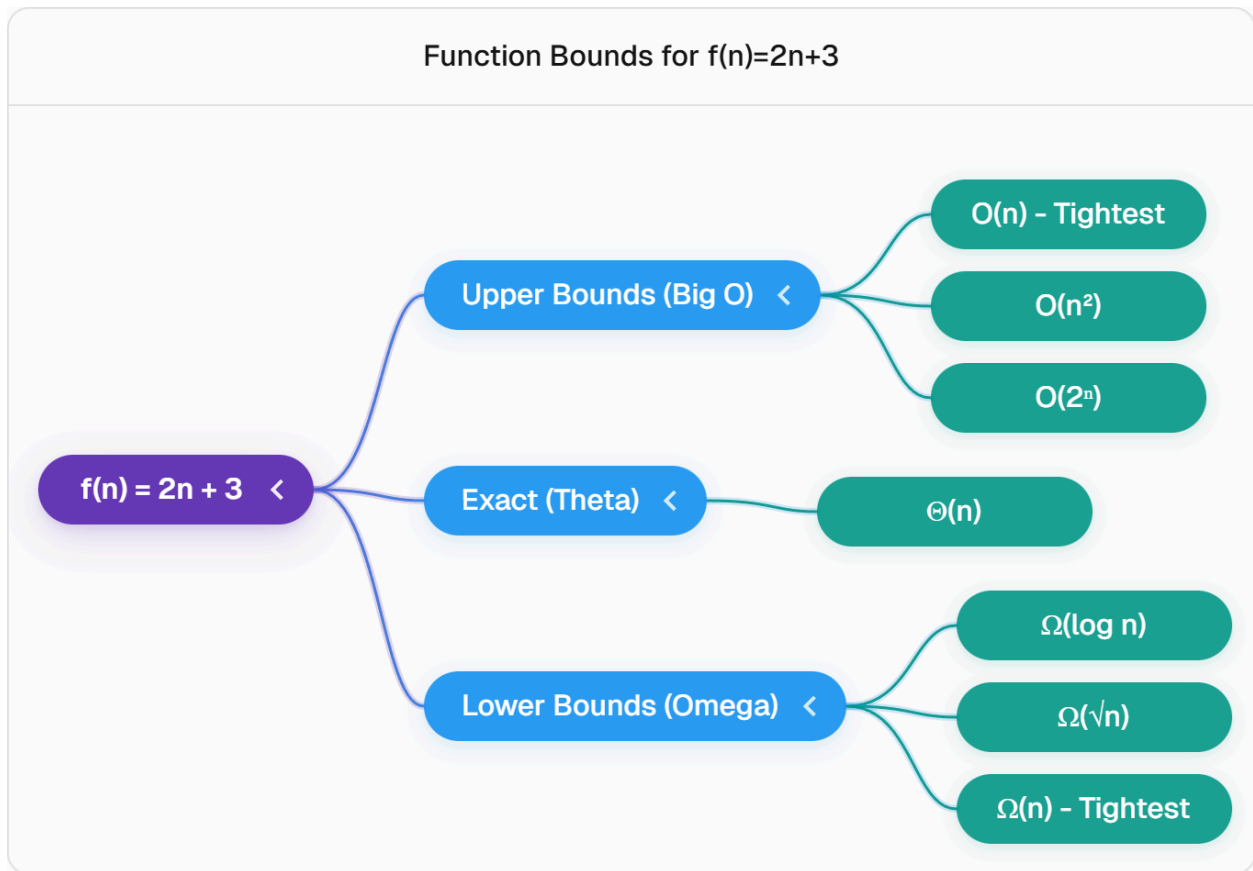
- Cannot use  $n^2$  or  $\log n$  (bounds don't match).
- **Most precise and preferred** when possible.

## Comparison of Notations

Notation	Bound Type	Inequality	Example for $f(n)=2n+3$	Tightness
<b>Big O</b>	Upper	$f(n) \leq C \cdot g(n)$	$O(n)$ , $O(n^2)$ , $O(2^n)$	Loose to tight
<b>Omega</b>	Lower	$f(n) \geq C \cdot g(n)$	$\Omega(n)$ , $\Omega(\log n)$	Loose to tight
<b>Theta</b>	Tight	$C_1 g(n) \leq f(n) \leq C_2 g(n)$	$\Theta(n)$ only	Exact match

**Visual function relationships** (for  $f(n)=2n+3$ ):

**Rule:** Prefer **Theta** → **Big O** → **Omega** (in that order of precision).



## Common Misconceptions

- **Not related to best/worst case:** These are function bounds, not algorithm cases (e.g., Big O can describe best or worst case).
- **Multiple valid representations:** Always true for looser bounds, but choose closest for analysis.
- **Single-term focus:** Simplify to one dominant term with coefficient.