

# I. Algorithm Analysis

- Algorithm analysis is to identify the time category (or order of growth) of an algorithm
- Eight most common categories
  - $1, \log n, n, n(\log n), n^2, n^3, 2^n, n!$

Most Popular

## Three Notations

$O$  (Big Oh)  
 $O$

$\Theta$  (Big Theta)  
 $\Theta$

$\Omega$  (Big Omega)  
 $\Omega$

$O$ : Best & Worst Case differ  
 $\Theta$ : Always same  $t(n)$   
 $\Omega$ : Lower Bound

## II. Big-Oh ( $O$ ) Notation

- Ex: Algorithm SequentialAdd ( $A[0 \dots n-1]$ )

sum  $\leftarrow 0$

$i \leftarrow 0$

while ( $i < n$ ) do

sum  $\leftarrow$  sum +  $A[i]$

$i \leftarrow i + 1$

return sum

- Big  $O$  notation denotes UPPER BOUND

$$t(n) \leq c * g(n)$$

$O(n \log n)$

$n \log n$  or faster

## Rules To Simplify $T(n)$

1. Eliminate low order terms

$$T(n) = 4 * n + 5 \rightarrow 4 * n$$

2. Eliminate Constant Coefficients

$$T(n) = 4 * n \rightarrow n \rightarrow O(n)$$



## - Big-O notation with Basic Operation

• It's difficult to calculate the total number of all executions of all operations

△ To simplify the calculation, we consider only the number of executions of the basic operation

## - Big-O formal definition:

• An algorithm's running time  $t(n)$  is said to be in  $O(g(n))$ , denoted  $t(n) \in O(g(n))$ , if there exist constants  $c$  &  $n_0$  that satisfy the following condition

$$t(n) \leq c * g(n) \text{ for all } n \geq n_0$$

$g(n)$   
↓  
One of  $g$   
categories

## III. Big Omega Notation ( $\omega$ ) ( $\Omega$ )

- An algorithm's running time  $t(n)$  is said to be in  $\omega(g(n))$ , denoted  $t(n) \in \omega(g(n))$ . If there exist constants  $c$  &  $n_0$  that satisfy the following condition

$$t(n) \geq c * g(n) \text{ for all } n \geq n_0$$

- Denotes LOWER BOUND

$$\Omega(n * \log n)$$

$$t(n) \geq c * g(n)$$

$n \log n$  or slower



## IV. Big Theta Notation ( $\Theta$ )

- An algorithm's running time  $t(n)$  is said to be in  $\Theta(g(n))$ , denoted  $t(n) \in \Theta(g(n))$  if there exist constants  $c_1$ ,  $c_2$ , &  $n_0$  that satisfy the following condition

$$c_1 * g(n) \leq t(n) \leq c_2 * g(n) \text{ for all } n \geq n_0$$

Scratch:

$O$

$$f(n) = n^2 + 3n + 1 \in O(n)?$$

$$f(n) = n^2 + 3n + 1 \leq C * n \text{ for all } n \geq n_0$$

Impossible, because  $n^2$  will increase faster than  $n$

if  $f(n) = n$ , can we say  $f(n) \in O(n)$ ? Yes

if  $f(n) = n$ , can we say  $f(n) \in O(n^2)$ ? Yes

Meets upper bound  
(or faster)

$\Omega$

if  $f(n) = 2 * n + 5$  can we say  $f(n) \in \Omega(n)$ ? Yes

Can we say  $f(n) \in \Omega(n^2)$

$$f(n) = 2 * n + 5 \geq C * n^2 \text{ for all } n \geq n_0$$

not possible

$O$  = or faster

$\Omega$  = or slower

$$f(n) \geq C * (\Omega * n \log n) \text{ for all } n \geq n_0$$

$$2n + 5 \geq C * (\Omega * n \log n) \text{ for all } n \geq n_0$$

not possible



Scratch:

Theta  $\times 2$

$$C_1 \cdot n \leq 2n + 3 \leq C_2 \cdot n$$

$$C_1: 1 \quad n \leq 2n + 3 \leq 5n$$

$$C_2: 5 \quad \text{True}$$

5 minutes

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1 1 1

2 2 1

3 2 2

4 1 1

5 2 2