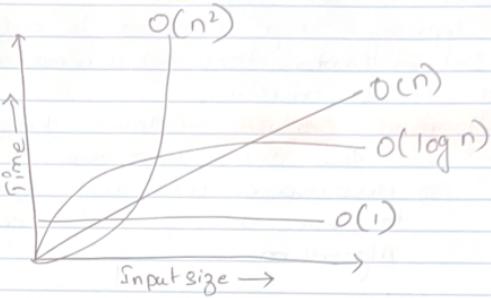


Big O Notation.



Time Complexities:-

- ① $O(1)$ or Constant time.
- ② $O(\log n)$ or Logarithmic time.
- ③ $O(n)$ or Linear time.
- ④ $O(n^2)$ or Exponential time
- ⑤ $O(n!)$ or Factorial time.

① $O(1)$ → Constant time. The Algorithm takes the same amount of time regardless of the input size.

② $O(\log n)$ → Logarithmic time. The algorithm time grows logarithmically as the input size increases, often due to dividing the problem into half.

③ $O(n)$ → Linear time. The time grows linearly with the input size.

④ $O(n \log n)$ → Log-Linear time. The time grows in proportion to ' n ' times ' $\log n$ ', often seen in efficient sorting.

Algorithm
Time complexity
Time complexity of an algorithm is the amount of time required to execute it.

5) $O(n^2)$ → Quadratic time. The time grows quadratically with the input size; common in algorithms with nested loops.

A COMPARISON OF SOME COMMON FUNCTIONS:-

Function	Name	Function	Name
① Any constant	Constant	⑦ n^2	Quadratic
② $\log n$	Logarithmic	⑧ n^3	Cubic
③ $\log^2 n$	Log-square	⑨ n^4	Quartic
④ \sqrt{n}	Root-n	⑩ 2^n	Exponential
⑤ $\frac{1}{n}$	Linear	⑪ e^n	Exponential
⑥ $n \log n$	Linearithmic	⑫ $n!$	n -Factorial

FORMULAE

List of handy formulas which can be helpful when calculating the time complexity of an algorithm.

Summation

Equation

$$① (\sum_{i=1}^n c) = c+c+c+\dots+c$$

$$cn$$

$$② (\sum_{i=1}^n i) = 1+2+3+\dots+n$$

$$\frac{n(n+1)}{2}$$

$$③ (\sum_{i=1}^n i^2) = 1+4+9+\dots+n^2$$

$$\frac{n(n+1)(2n+1)}{6}$$

$$④ (\sum_{i=0}^n r^i) = r^0+r^1+r^2+\dots+r^n$$

$$\frac{(r^{n+1}-1)}{r-1}$$

General Tips:

- ① Every time a list or array gets iterated over c₁ length times, it is most likely in O(n) time.
- ② When you see a problem where the number of elements in the problem space gets halved each time, it will most probably be in O(log n) runtime.
- ③ Whenever you have a single nested loop, the problem is most likely in quadratic (n²) time.

* Example. → Running Time Complexity. (for-loop).

class sum
{
 int n=10; - O(1)
 int sum=0; - O(1)
 for (int i=0; i<n; i++)
 {
 sum += 1; - O(n)
 }
 System.out.println(sum);
}

Output:- 10.

Detailed Step-by-step process for Run-time Complexity:

int n=10; Running time complexity = 1
int sum=0; " " " = 1
int i=0; " " " = 1

i < n → The test operation i < n comprises of three steps
① Access the value of i. $\frac{1}{3}$
② Access the value of n. $\frac{1}{3}$
③ Compare the values of i and n. $\frac{1}{3}$

→ On careful analysis you will notice that the ~~for~~ statement is executed for each value of i.

* Therefore, the statement is executed n+1 times.

* Therefore the total statement occurs for 3(n+1) times.

sum += 1; - 3 (Accessing, adding, updating)

This statement is executed n times because of for loop.
Therefore, in total this statement accounts for 3n primitive operations.

i++ - 3 (Accessing, adding, updating) occurs n times → 3n

S.O.P(sum); - 2 (Accessing, displaying)

COMMON COMPLEXITY SCENARIOS:-

① Simple for-loop with an increment of size 1.

for (int x=0; x<n; x++) {

// Statement(s) take constant time

Explanation: ① The initialization $x=0$; runs once - $O(1)$

② The comparison Statement

$x < n$; runs $n+1$ times.

③ The constant time statement

in the loop itself runs n times.

④ The loop increment statement

$x++$; runs $n+1$ times.

$$\therefore i+(n+1)+cn+n \Rightarrow 2n+cn+2$$

$\therefore O(n(2+c)+2)$

(by dropping lower order terms)

2. For-Loop with Increments of K.

for (int x=0; x<n; x+=k) {

// Statement(s) take constant time

$$\begin{aligned} &\text{Running Time Complexity} \\ &\downarrow \\ &2 + n \frac{(2+c)}{k} = O(n) \end{aligned}$$

Explanation: ① The initialization $x=0$; runs once.

② Then x is incremented by k until it reaches n .

$$[0, k, 2k, 3k, \dots, nk] \leq n \rightarrow \frac{n}{k} + 1$$

③ Hence, the incrementation part $x+=k$ for the loop takes $\frac{n}{k}$ times.

④ Constant $c \cdot \frac{n}{k}$ times $\rightarrow c \cdot \frac{n}{k}$ times

$$= 1 + \frac{n}{k} + 1 + \frac{n}{k} + \frac{cn}{k} \Rightarrow \frac{2n}{k} + \frac{cn}{k} + 2$$

$$\therefore \frac{n}{k} (2+c) + 2 \quad [\because \text{Here } k \text{ is constant}]$$

$$\therefore O(n)$$

3. Simple Nested For Loop.

```
for(int i=0; i<n; i++) {  
    for(int j=0; j<m; j++) {  
        // Statement(s) that take(c) constant  
        // time  
    }  
}
```

Running Time Complexity

$$nm(2+c) + 2 + cn = O(nm)$$

Explanation:- The inner loop is a simple for loop that takes $(2m+2) + cm$ time and the outer loop runs in n times so it takes $n((2m+2) + cm)$ time. Additionally, the initialization, increment and test for the outer loop take $2n+2$ time \Rightarrow So in total, the running time is.

$$n((2m+2) + cm) + 2n+2 \Rightarrow 2nm + 4n + cmn + 2$$

$$\Rightarrow nm(2+c) + 4n + 2, \text{ which is } O(nm).$$

4. Nested For-Loop with dependent variables.

```
for(int i=0; i<n; i++) {  
    for(int j=0; j<i; j++) {  
        // Statement(s) that take(c) constant  
        // time  
    }  
}
```

Running Time Complexity: - $O(n^2)$

5. Nested For-Loop with Index modification.

```
for( int i=0; i<n; i++) {
```

```
    for( int j=2; j<i; j++) {
```

// Statements that take constant time.

Running Time Complexity = $O(n)$

6. Loops with $\log n$ time complexity.

$i = \text{constant}$

$i < n = \text{constant}$

$i = \text{constant}$

while ($i < n$) {

$i = k;$

// Statement(s) that take(s) constant time

}

Running Time Complexity = $\log_k(n) = O(\log_k(n))$

8. Big(O) of Nested loop with addition.

```
class Nestedloop {
```

```
public static void main (String args) {
```

```
    int n = 10;
```

```
    int sum = 0;
```

```
    double pie = 3.14;
```

```
    for( int var=0; var<n; var = var + 3) {
```

```
        System.out.println ("pie" + pie);
```

```
        for( int j=0; j<n; j = j + 2) {
```

```
            System.out.println ("j" + j);
```

Statement	no. of Executions
$int n=10;$	1
$int sum=0;$	1
$double pie=3.14;$	1

$$\text{Sum} += \frac{2}{3} \times \frac{1}{2}$$

System.out.println("Sum: " + Sum);

$$= \frac{2}{3} \times \frac{1}{2} = \frac{1}{3}$$

so it adds what elements of the array.

$$\text{Running Time Complexity} = O(n^2)$$

$$\text{For each element } i = 1 + 1 + 1 + \frac{n}{3} + 1 + \frac{n}{3} + \frac{n}{3} + \frac{n}{3} \times \left(\frac{n}{2} + 1 \right) + \left(\frac{n}{3} \times \frac{n}{2} \right) + \left(\frac{n}{3} \times \frac{n}{2} \right) +$$

$$\left(\frac{n}{3} \times \frac{n}{2} \right)$$

$$= \frac{15 + 5n + 2n^2}{3}$$

Big(O) Time complexity: $O(n^2)$

Complexity when a small loop > big loop

RECURSION

Recursion is a method of solving problems by dividing them into smaller problems.

in recursion we divide the problem into smaller problems.

and solve those smaller problems by dividing them into even smaller problems.

and solve those smaller problems by dividing them into even smaller problems.

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