

4. All about common Complexities

Big(1) - Constant Time

Big(1) can also be called as $O(1)$ or $O(c)$

- No matter size of array/input , accessing an array/input same amount of time.
- Example : Accessing a specific element in an array by index.

```
1 package com.dsa;
2
3 public class Complex {
4
5     static void constantTime() { 1 usage
6         int array[] = {
7             1, 2, 4, 5, 6
8         };
9
10        /**
11         * here the statement which is accessing our array will take constant time no matter we increase the input size.
12         * i.e. Complexity is  $O(1)$ 
13         */
14        System.out.println(array[0]);
15    }
16
17    public static void main(String[] args) {
18        constantTime();
19    }
20 }
21
```

Equation for time : $f(n) = C$

$O(n)$ - Linear Time

- Time grows directly proportional to the size of input.
- Example : Traversing an array of size.

```

1 package com.dsa;
2
3 public class Complex_LinearTime {
4
5     static void linearTime() { 1 usage
6         int array[] = {
7             1, 2, 4, 5, 6
8         };
9
10        for(int i = 0; i <= array.length - 1; i++) {
11
12            /**
13             * here the statement which is accessing our array will take more time when input is increased.
14             * i.e. time will grow proportionally when input is grown
15             * i.e. Complexity is O(n)
16             */
17            System.out.println((i+1) + " time \t");
18            System.out.println(array[i] + " ");
19        }
20    }
21
22    public static void main(String[] args) {
23        linearTime();
24    }
25 }
26

```

Equation for time : $f(n) = a(n) + C$ {Linear Equation $\rightarrow y = mx+c$ }

$O(n^2)$ - Quadratic Time

- As the input grows, time taken increases quadratically.
- Example : Bubble Sort or checking all pairs in an array.

```

1  package com.dsa;
2
3  ► public class Complex_QuadtraticTime_BubbleSort {
4      static void bubbleSort() { 1 usage
5          int array[] = {
6              4, 2, 1, 3
7          };
8
9          System.out.println("Before bubble sort: ");
10         for(int a : array){
11             System.out.print(a + ",");
12         }
13
14         int i,j,temp;
15         boolean flag;
16
17         for(i = 0; i < array.length-1; i++){
18             flag = false;
19             for(j = 0; j < array.length-i-1; j++){
20                 if(array[j] > array[j+1]){
21                     temp = array[j];
22                     array[j] = array[j+1];
23                     array[j+1] = temp;
24                     flag = true;
25                 }
26             }
27             if(!flag){
28                 break;
29             }
30         }
31         System.out.println("\nAfter bubble sort: ");
32         for(int s : array){
33             System.out.print(s + ",");
34         }
35     }
36
37  ► public static void main(String[] args) {
38      bubbleSort();
39  }
40
41  }

```

Equation for time :

Here A, B and C are constants

$$f(n) = An^2 + Bn + C$$

As compared to An^2 , $Bn + C$ is far smaller so ignoring constants we get :

$$f(n) = An^2$$

Again ignoring constant A we get :

$$f(n) = n^2$$

$O(\log(n))$ - Logarithmic Time

- In this the algorithm cuts problem in half with each step, so the time grows logarithmically.
- Example : Binary Search

Equation for time : $\log(n) = K$

$O(n \log n)$ - Linearithmic Time

- The algorithm divides the input into subproblems and processes each subproblem linearly.
- Example : Merge Sort, Quick Sort in average cases.

$O(2^n)$ Exponential Time Complexity

- The time grows exponentially with the size of input, meaning it doubles with each additional input.
- Recursive algorithm that solves a problem by breaking into multiple smaller sub problems.
- Example : Fibonacci Series using recursion.

$O(n!)$ Factorial Time Complexity

- Factorial Time Complexity occurs in algorithm that involve generating all possible permutations of a set.
- Example : Brute-Force solution for the travelling salesman problem.

Overall Comparison of all Time Complexities

$$n(c) < O(\log \log n) < O(\log n) < O(\log n^{1/2}) < O(n) < O(n \log n) < O(n^2) < O(n^3) < O(n^k) < O(2^n)$$

