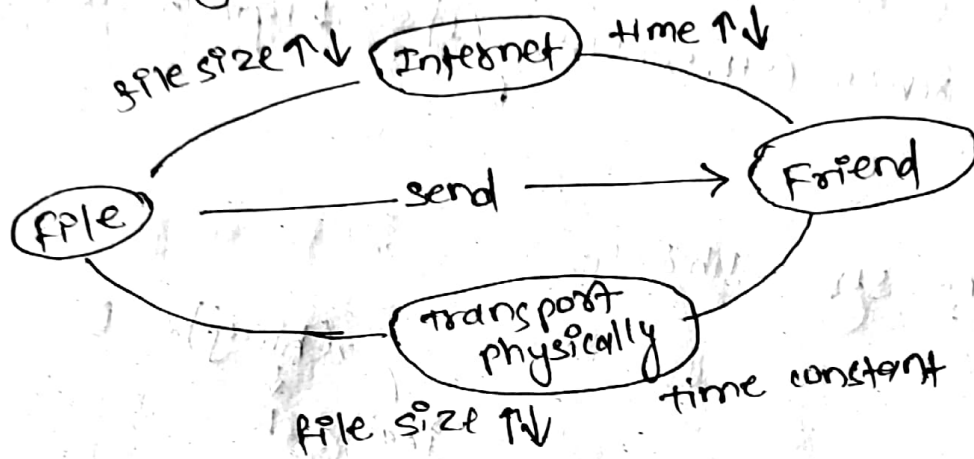


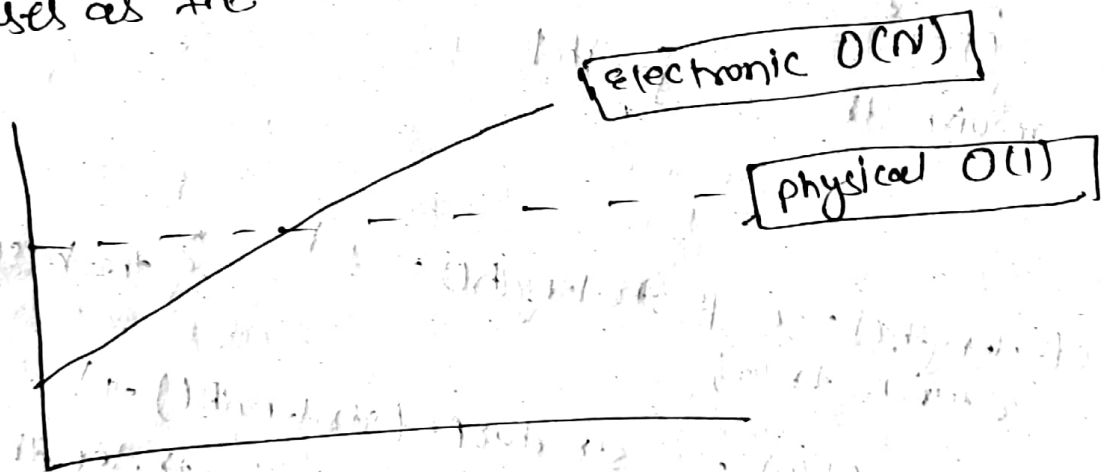
Big O notation

Big O is the language and metric we use to describe the efficiency of algorithms

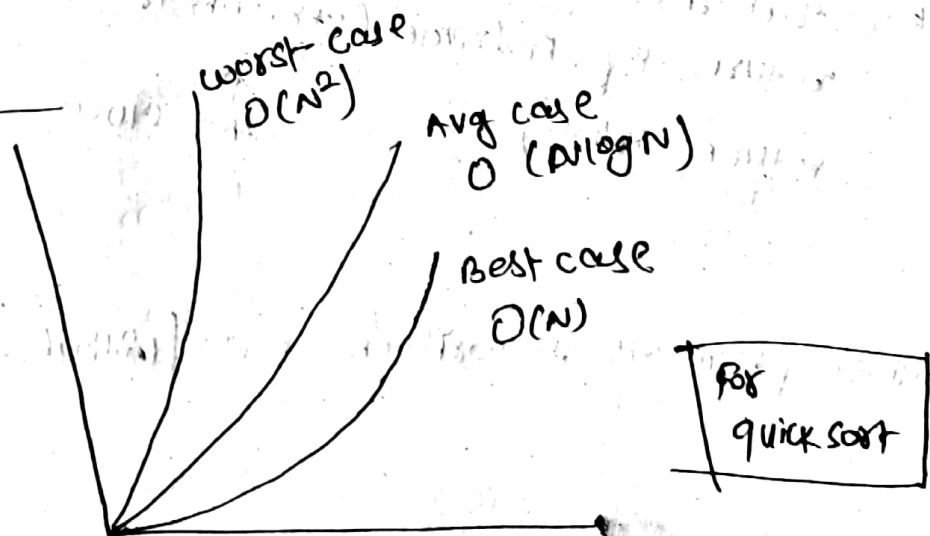


time complexity

A way of showing how the runtime of a function increases as the size of input increases.



Different Scenarios



Types

Big O : It is the complexity that is going to be less or equal to the worst case
 Big O - $O(N)$ {max time?}

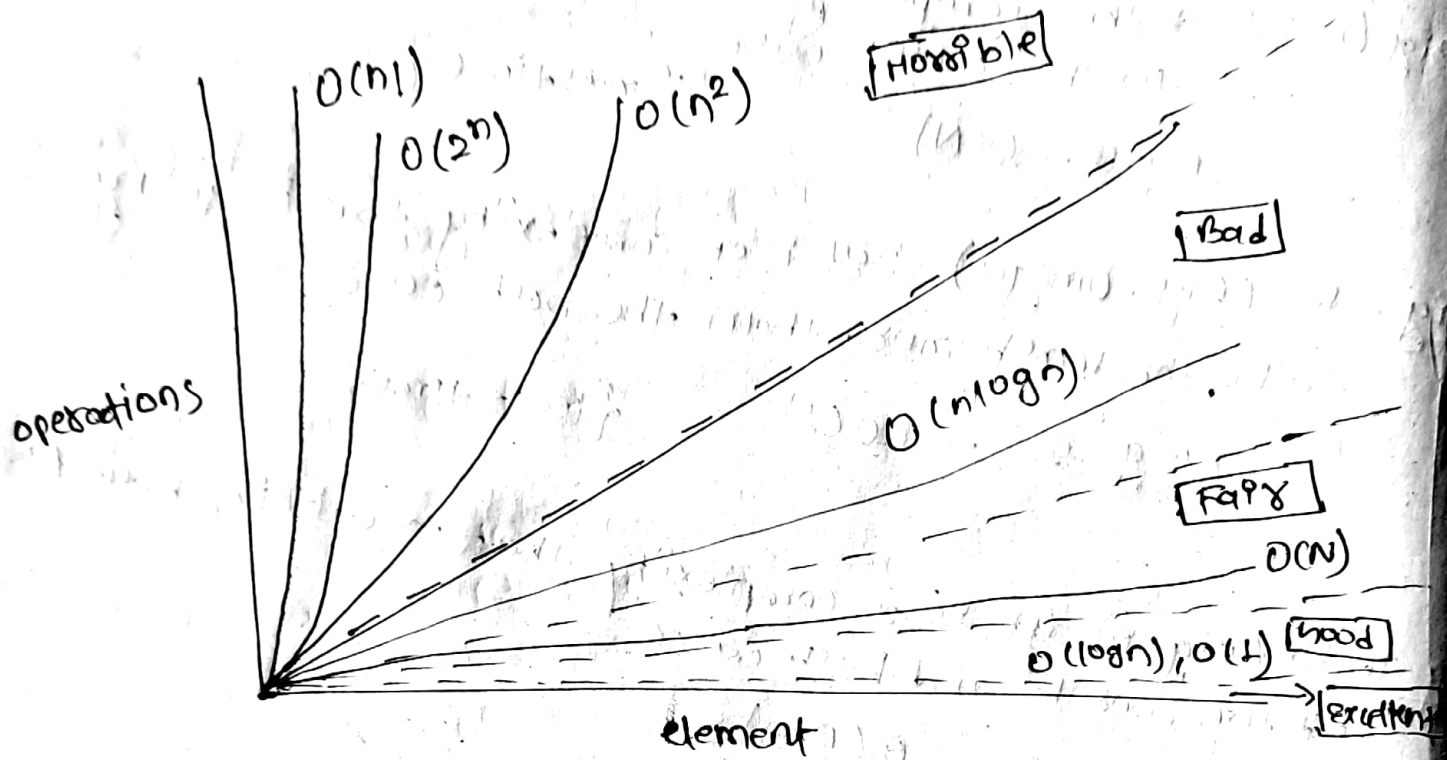
Big Ω (Big - Omega) : It is a complexity that is going to be the least more than the best case
 Big Ω - $\Omega(1)$ {least time?}

Big Theta (Big Θ) : It is a complexity that is within bounds of the worst and best cases.
 Big Θ - $\Theta(N/2)$

Runtime complexities

complexity	name	sample
$O(1)$	constant	Accessing a specific element in array
$O(N)$	Linear	Loop through array elements
$O(\log N)$	logarithmic	Find an element in sorted array
$O(N^2)$	Quadratic	looking every index in the array twice
$O(2^N)$	exponential	double recursion in Fibonacci

Big-O complexity chart



#Space Complexity

linear array
SC: $O(N)$

2-D array
SC: $O(N^2)$

```
static int sum(int n){
    if (n <= 0)
        return 0;
    return n + sum(n-1);
}
```

1. sum(3)
2. sum(2)
3. sum(1)
4. sum(0)

space complexity : $O(N)$.

#1 Drop constants and Non-Dominant terms

Drop constant

$$O(2N) \rightarrow O(N)$$

Non-Dominant terms

$$O(N^2 + N) \rightarrow O(N^2)$$

$$O(N + \log N) \rightarrow O(N)$$

$$O(2 \times 2^N + 1000 N^{100}) \rightarrow O(2 \times 2^N) \rightarrow O(2^N)$$

Add vs multiply

- If algorithm is in the form, 'do this, then when you are all done, do that', then you add the run times

ex

```
for (a=0; arrayA.length; a++) {  
    system.out.println(arrayA[a]);  
}  
for (b=0; arrayB.length; b++) {  
    system.out.println(arrayB[b]);  
}
```

⇒ Add the run times: $O(A+B)$

- If algorithm is in the form, 'do this for each time you do that', then you multiply the runtimes.

ex

```

for (a=0; arrayA.length; a++) {
    for (b=0; arrayB.length; b++) {
        System.out.println (arrayA[a] + arrayB[b]);
    }
}

```

→ multiply the runtimes : $O(A * B)$

measuring the Big O time complexities

no.	description	complexity
Rule 1	Any assignment statements and if statements that are executed once regardless of the size of the problem	$O(1)$
Rule 2	A simple 'for' loop from 0 to n	$O(N)$
Rule 3	A nested loop of same type	$O(N^2)$
Rule 4	A loop, in which the controlling parameter is divided by two at each step	$O(\log N)$
Rule 5	when dealing with multiple statements just add up	!

Example

```

public static int biggestElement (int a) {
    int big = a[0];
    for (int i=1; i < a.length; i++) {
        if (big < a[i])
            big = a[i];
    }
    cout ("big : " + big);
}

```

Annotations for complexity:

- `int big = a[0];` → $O(1)$
- `for (int i=1; i < a.length; i++)` → $O(N)$
- `if (big < a[i])` → $O(1)$
- `big = a[i];` → $O(1)$
- The loop body (if and assignment) is grouped by a brace → $O(1)$
- The entire loop is grouped by a brace → $O(N)$
- The final `cout` statement → $O(1)$

total time complexity
 $= O(1) + O(n) + O(1)$
 $= O(n)$ A

measure of Recursive Algorithm

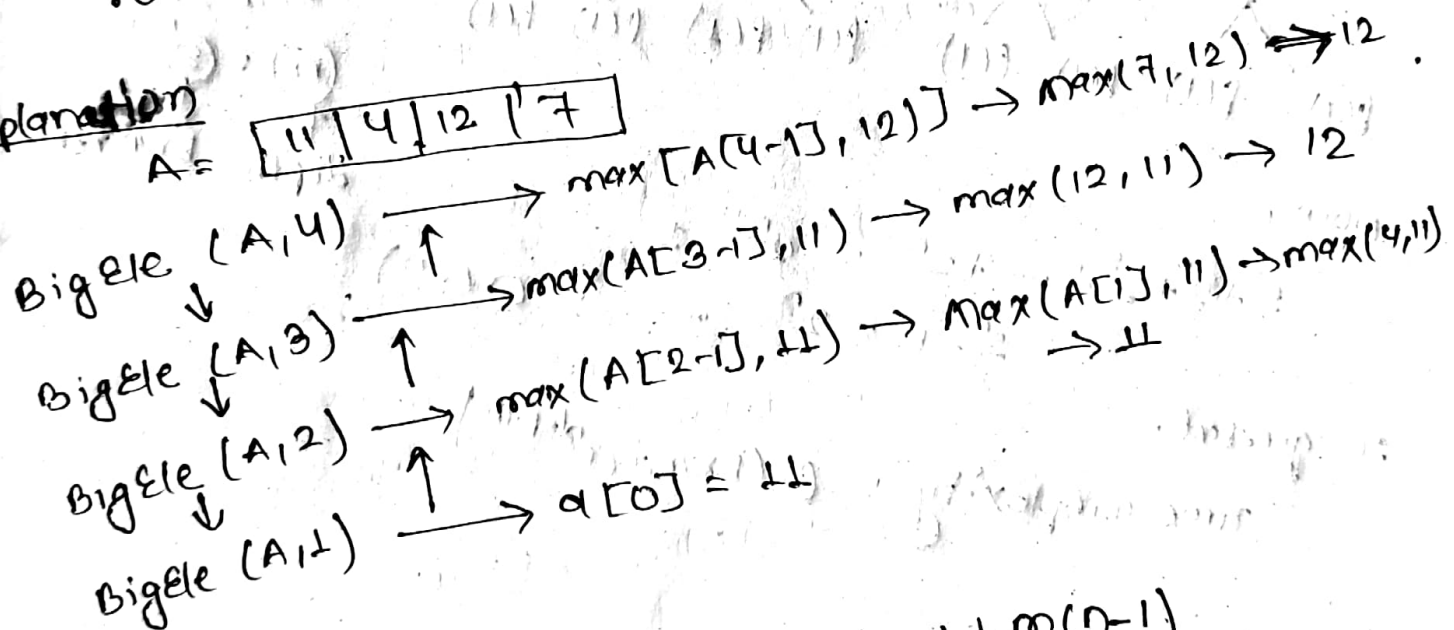
sample Array

5	4	10	...	8	11	...	6	68	87	10
---	---	----	-----	---	----	-----	---	----	----	----

static int BigEle ([] a, int n) {
 if (n == 1)
 return a [0];
 return max (a [n], BigEle (a, n - 1));
}

$\rightarrow m(n)$
 $\rightarrow O(1)$
 $\rightarrow O(1)$
 $\rightarrow m(n-1) + O(1)$

Explanation



now,

$$\left. \begin{aligned} M(n) &= O(1) + m(n-1) \\ m(n) &= O(1) + m(n-1) - 1 \\ &= O(1) + m(n-2) \\ m(n-2) &= O(1) + m(n-2) - 1 \\ m(1) &= O(1) \end{aligned} \right\} \Rightarrow$$

$$\begin{aligned} m(n) &= 1 + m(n-1) \\ &= 1 + (1 + m(n-1) - 1) \\ &= 1 + 1 + (m(n-2) - 1) \\ &= 2 + (1 + m(n-2) - 1) \\ &= 3 + m(n-3) \\ &= a + m(n-a) \\ &= n-1 + m(n-(n-1)) \quad \{ a = n-1 \} \\ &= n-1 + m(1) \\ &= n-1 + 1 = n = O(n) \end{aligned}$$

measuring Recursive Algorithm with multiple calls

```
public int f(int n) {
```

```
    if (n == 1)
        return 1;
```

```
    return f(n-1) + f(n-1);
```

```
}
```

branches

level node

0 = 2^0

1 = 2^1

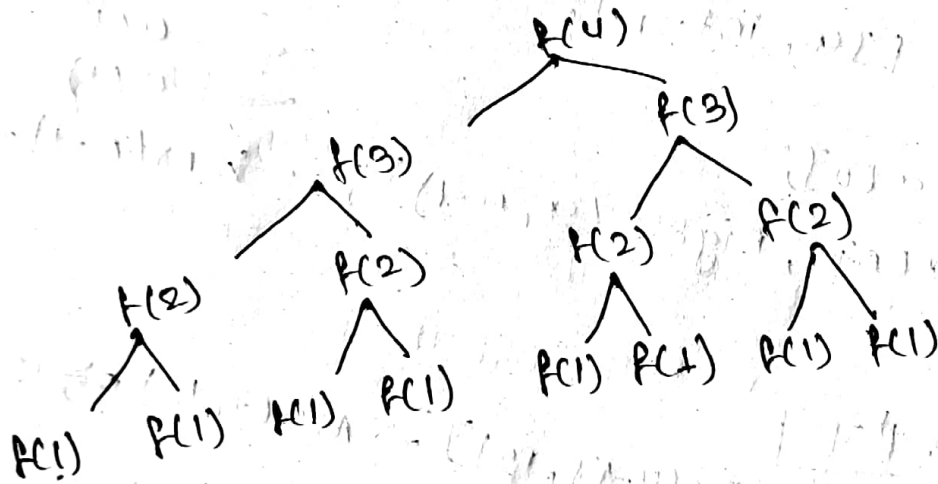
2 = 2^2

3 = 2^3

$\vdots = 2^n$

depth

branches



Now,

$$2^0 + 2^1 + 2^2 + \dots + 2^n = 2^{n+1} - 1$$

$$= 2^n - 1 = 2^n \rightarrow O(2^n)$$

In general.

Time complexity : $O(\text{branches}^{\text{depth}})$

Q.3 Find the time complexity?

```
void printUnorderedPairs (int [] array) {
    for (int i = 0; i < array.length; i++) {
        for (int j = i+1; j < array.length; j++) {
            cout << a[i] + " " + a[j] << endl;
        }
    }
}
```

1. counting iterations

1st $\rightarrow n-1$
 2nd $\rightarrow n-2$
 \vdots
 1

$$\Rightarrow 1 + 2 + \dots + (n-2) + (n-1)$$

$$= \frac{n(n-1)}{2} = \frac{n^2}{2} - \frac{n}{2}$$

$$= O(n^2)$$

2. Average work.
 Outer loop - $O(N)$
 Inner loop - ?

step 1 $\rightarrow 10$
 step 2 $\rightarrow 9$
 step 3 $\rightarrow 8$
 \vdots
 1

Avg = $\frac{10+9+8+\dots+1}{n} = \frac{n+1}{2} \approx \frac{n}{2}$

So, outer inner $\Rightarrow n \times \frac{n}{2} \Rightarrow \frac{n^2}{2}$
 $= O(n^2)$

Q.5 $O(mn)$

Q.4 $O(m \cdot N)$

$m = \text{arrayA.length}$
 $N = \text{arrayB.length}$

Q.7

which of the following are equivalent to $O(N)$? why?

1. $O(N+p)$, where $p < N/2$ ✓
2. $O(2N)$ ✓
3. $O(N + \log N)$ ✓
4. $O(N + N \log N)$
5. $O(N+m)$

Q.8

Time complex of factorial code used in recursion.

$O(N)$. method \rightarrow Q.3

Q.9 Fibonacci

$O(2^N)$ \Rightarrow

branches \downarrow
given (2).
(depth) $\rightarrow N$

Q.10 Power of 2

$O(\log(N))$