

# DSA

## Order of growth

A function  $f(n)$  is said to be growing faster than  $g(n)$  if

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty //$$

OR

$$\lim_{n \rightarrow \infty} \frac{g(n)}{f(n)} = 0$$

$$\frac{\quad}{\quad} \quad n \geq 0$$

Directway:

- ① Ignore lower order terms
- ② Ignore leading constant

Remember:

$$c < \log(\log n) < \log n < n^{1/3} < n^{1/2} < n < n^2 < n^3 < n^4 < 2^n < n^n$$

Check for best, Average and worst case

eg:

```
int getSum(int arr[], int n)
{
    if (n % 2 != 0) return 0;
    for (int i = 0; i < n; i++)
        sum = sum + arr[i];
    return sum;
}
```

Worst case:  $n$

Best case:  $1$

# Asymptotic Notations

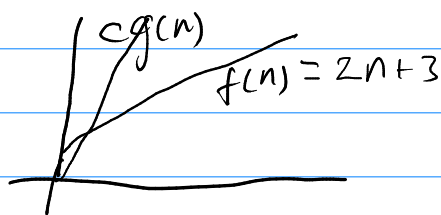
Big O: Represents exact bound or upper bound

Theta: Represent exact bound.

Omega: Represent exact or lower bound.

## Big O Notation (Upper bound or Order of growth)

We say  $f(n) = O(g(n))$  if there exist constants  $C$  and  $n_0$  such that  $f(n) \leq Cg(n)$  for all  $n \geq n_0$



### Example

$$f(n) = 2n + 3$$

$$2n + 3 \leq 3n \text{ for } n \geq 3$$

$$\left\{ \frac{n}{4}, 2n + 3, \frac{n}{100} + \log n, n + 10000, n/10000, 100, \log n + 100 \right\} \in O(n)$$

$$\{ n^2 + n, 2n^2 + n^2 + 100n, n^2 + 2\log n \} \in O(n^2)$$

$$\{ 1000, 2, 3 \} \in O(1)$$

Application

```
int linearsearch(int arr[], int n, int x)
{
    for(int i=0; i<n; i++)
        if(arr[i] == x)
            return i;
    return -1;
}
```

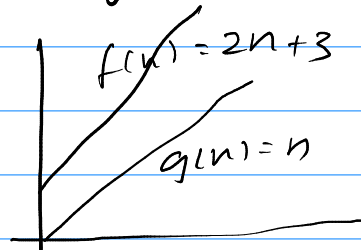
Time complexity :  $O(n)$

### Omega Notation lower bound

$f(n) = \Omega(g(n))$  if there exist positive constant  $C$  and  $n_0$  such that  $0 \leq Cg(n) \leq f(n)$  for all  $n \geq n_0$

example:

$$f(n) = 2n + 3 \\ = \Omega(n)$$



①  $\{n/4, n/2, 2n, 3n, 2n+3, n^2, 2n^2, \dots, n^n\} \in \Omega(n)$

② If  $f(n) = \Omega(g(n))$   
then  $g(n) = O(f(n))$

③ Omega notation is useful when we have lower bound on time complexity

# Theta Notation

$f(n) = \Theta(g(n))$  if there exist positive constants  $C_1, C_2$  and  $n_0$  such that  $0 \leq C_1 g(n) \leq f(n) \leq C_2 g(n)$  for all  $n \geq n_0$

Example

$$f(n) = 2n + 3 \quad \text{order of growth} \\ = \Theta(n)$$

$$C_1 g(n) \leq f(n) \leq C_2 g(n) \quad \text{for all } n \geq n_0$$

① If  $f(n) = \Theta(g(n))$

then  $f(n) = O(g(n))$  and  $f(n) = \Omega(g(n))$

and  $g(n) = O(f(n))$  and  $g(n) = \Omega(f(n))$

② Theta is useful to represent time complexity when we know exact bound. For example time complexity to find sum, max and min in an array is  $\Theta(n)$

③  $\{n^2, n^2/4, \dots, 2n^2, \dots, 4n^2 + 2n \log n, \dots\} \in \Theta(n^2)$

# Analysis of Common loops

① for (int i=0; i<n; i=i+c)  
 { // Some  $\Theta(1)$  work  
 }

$$i=0, i=c, i=2c$$

$$\Rightarrow i_k = kC$$

$$kC < n \Rightarrow \boxed{k < \frac{n}{c}}$$

$$\text{No of iterations} = n/c$$

Time complexity :  $\Theta(n)$

② for (int i=n; i>0; i=i-c)  
 {  
 // some  $\Theta(1)$  work  
 }

Time complexity =  $\Theta(n)$

③ for (int i=1; i<n; i=i\*c)  
 // some  $\Theta(1)$  work;

$$c^{k-1} < n$$

$$\ln_c n = k-1$$

$$k < \ln_c n + 1$$

④ for (int i=n; i>0; i=i/c)

$$n \frac{n}{c} \frac{n}{c^2} \dots$$

$$\Rightarrow \ln_c n \geq k-1$$

$$\frac{n}{c^{k-1}} \geq 1$$

$$k \leq \log_c n + 1 \quad n \geq c^{k-1}$$

⑤ for (int i=2; i<n; i=pow(i,c))

$$2 \cdot 2^c \cdot (2^c)^c \cdot ((2^c)^c)^c$$

$$2^k < n$$

$$c = 2 \quad (2^2)^2$$

$$2^2 \quad 2^4 \quad 2^8$$

$$c^k \leq \log_2 n$$

$$k \leq \log(\log_2 n)$$

⑥ void func(int n)

{ for (int i=0; i<n; i++)  
// some  $\Theta(1)$  work }  $\Theta(n)$

for (int i=1; i<n; i=i\*2)  $\Theta(\log n)$   
// some  $\Theta(1)$  work

for (int i=0; i<100; i++)  $\Theta(1)$   
// some  $\Theta(1)$  work  
}

⑦ for (int i=0; i<n; i++)  
for (int j=1; j<n; j=j\*2)  
// same  $\Theta(1)$  work

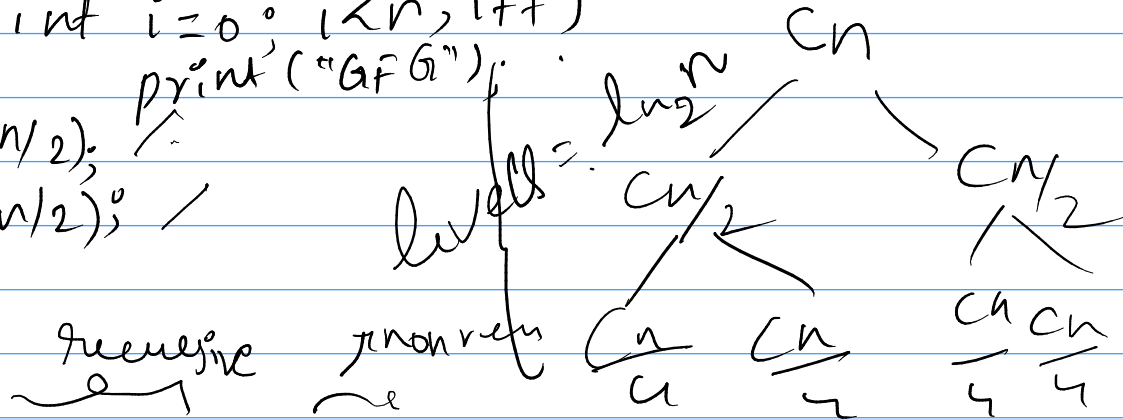
$$\log n + \log n + \dots + n \text{ times}$$

$$\Theta(N \log N)$$

# Analysis of Recursion

```

void fun(int n)
{
    if (n <= 1)
        return;
    for (int i = 0; i < n; i++)
        print("GFG");
    fun(n/2);
    fun(n/2);
}
    
```



$$T(n) = 2T(n/2) + cn$$

$$T(1) = c$$

work at each level =  $cn$

Time complexity ( $N \log N$ )

## Recursion Tree Method

- we work non recursive part as root of tree and recursive part as children
- we keep expanding children until we see a pattern

$$T(n) = 2T(n-1) + c$$

$$T(1) = c$$

$$\Rightarrow c + 2c + 4c + \dots + \frac{a(r^n - 1)}{r - 1} + \frac{a \cdot 2^{n-1}}{1}$$

$$= O(2^n)$$

→ Recursion tree method doesn't give exact bound always.

## Space Complexity

Order of growth of memory (OR RAM) space in terms of input size

```
int arrSum(int arr[], int n)
{
    int sum = 0;
    for (int i = 0; i < n; i++)
        sum = sum + arr[i];
    return sum;
}
```

Auxiliary space : Order of growth of extra space or temporary space in terms of input size

```
int arrSum(int arr[], int n)
{
    int sum = 0;
    for (int i = 0; i < n; i++)
        sum = sum + arr[i];
    return sum;
}
```

Auxiliary space:  
 $\Theta(1)$   
space complexity  
 $\Theta(n)$

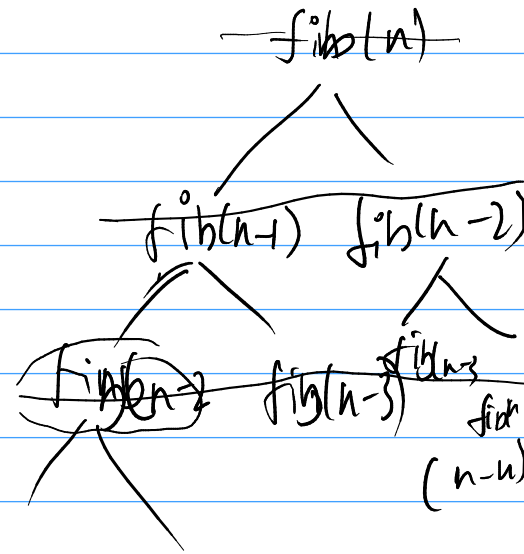


```

int fib(int n)
{
    if (n == 0 || n == 1)
        return n;
    return fib(n-1) + fib(n-2);
}

```

Auxiliary space : *height of tree*



Q) `int fib(int n)`

```

{
    int f[n+1];
    f[0] = 0;
    f[1] = 1;
    for (int i = 2; i <= n; i++) {
        f[i] = f[i-1] + f[i-2];
    }
    return f[n];
}

```

$\Theta(n)$