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Analysis Design of Algorithms  
CS - 402

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Class - CS - B

Sem - IV

## Assignment - 1

What is ADA? What is the need to Study Algorithms? Explain in detail.

ADA :-

ADA is a branch of CS that focuses on the study of algorithms, their design, analysis of their efficiency and their application in solving computational problems.

Computational means which can be solved by an algorithm.

The need to Study algorithm are :-

- Problem Solving :- Algorithms provide systematic approaches to solving various computational problems.

By studying algorithms, you learn how to break down complex problems into smaller, more manageable steps, leading to efficient and effective solutions.

- Efficiency :- Understanding algorithms helps in developing efficient solutions.
- Scalability :- As the size of data and computational tasks continues to grow, the importance of scalable algorithms becomes paramount.
- Optimization :- Algorithm often provide multiple solutions to the same problem, with varying trade-offs in terms of time and space complexity.

- Foundation for Computer Science:

Algorithms form the backbone of computer science. They are fundamental to various subjects such as data structures, artificial intelligence.

- Problem Domain independence

Algorithms are not specific to any particular programming language or domain.

- Competitive programming and Interviews:

In fields like software engineering, algorithmic knowledge is often a key component of technical interviews.

Q2 Give the Divide and Conquer Solution for Quicksort and analyze its complexity.

Ans

Divide and conquer Technique(D&C)

D&C is a top-down approach to solve a problem. The algorithm which involves D&C technique involves 3 step →

- Divide:- The original problem into a set of sub problems.
- Conquer (or Solve) every sub-problem individually, recursive.
- combine the solution of these sub problems to get the solution of original problem.

## Quick Sort

- Pivot element based
- Partition based

It picks an element as pivot and partitions the given array around the picked pivot. There are many different versions of quick sort that picks the pivot in different ways.

Always pick first element as pivot.  
 Always pick last element as pivot.  
 Pick a random element as pivot.  
 Pick median as pivot.

Example :-       $\begin{matrix} & 1 & 2 & 3 & 4 & 5 \\ \text{Example :- } & | & | & | & | & | & | \end{matrix}$   
 $i = 0$      $j = 1$      $P = 1$ ,     $\delta = 5$

$x$  is the pivot  
 $A[\delta] = x$

So,  $[x = 4]$

II Partition Function  
 partition ( $A, p, \gamma$ )  
 {

$$x = A[\gamma]$$

$$i = p - 1$$

for ( $j = p$  to  $\gamma^4 - 1$ )

1	2	3	4	5
1	0	5	6	4

{ if ( $A[j] \leq x$ )

find the index  $y$   
 $i \neq j$

$$i = i + 1$$

Exchange

$$A[i] \leftarrow A[j]$$

}

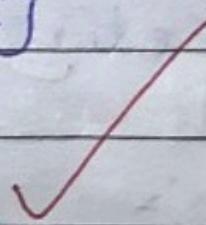
3

exchange

$$A[i+1] \leftarrow A[\gamma]$$

return  $i + 1$

3



→ 11 Algorithm

quicksort ( $A, p, r$ )

{

if ( $p < r$ )

$q = \text{partition } (A, p, r)$

quicksort ( $A, p, q-1$ )

quicksort ( $A, q+1, r$ )

}

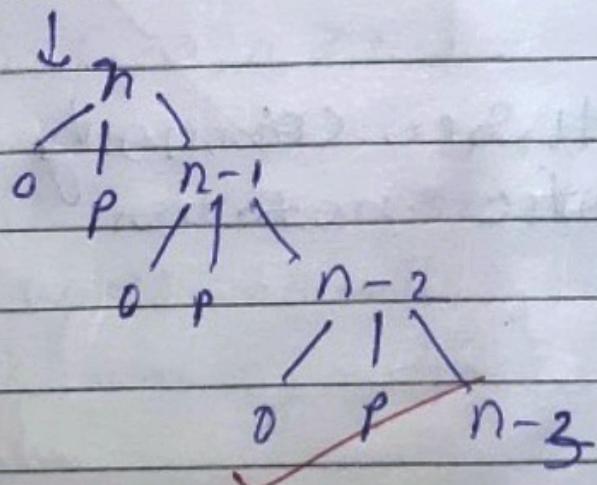
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Complexity :-

Best - case :  $O(n \log n)$

Average case :  $O(n \log n)$

worst - case :  $O(n^2)$  { when the array is already sorted }



$$T(n) = T(n-1) + n \\ O(n^2)$$

Q3 Define asymptotic notations. Give different notations used to represent the complexity of algorithms.

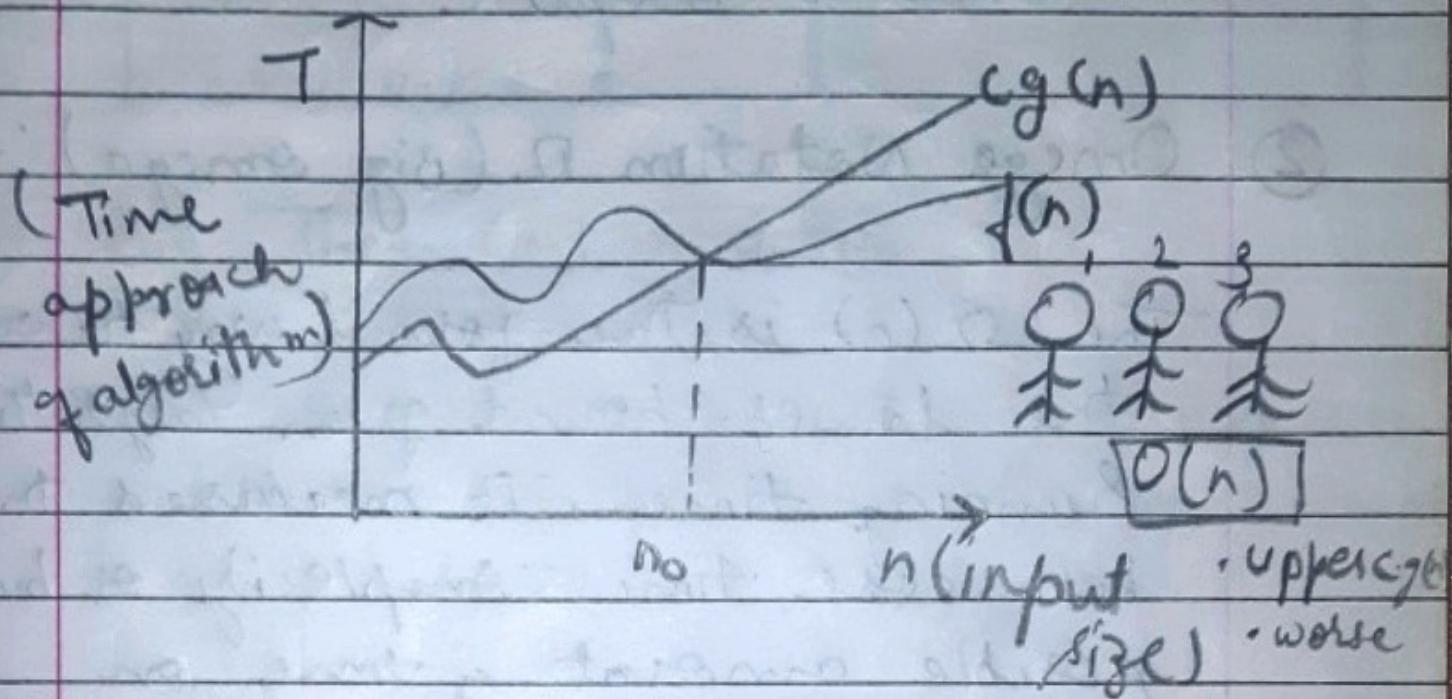
Ans ->

Asymptotic notations:-

Asymptotic notations are mathematical representations used in computer science to describe the running time or space complexity of an algorithm as the input size approaches infinity.

The are three commonly used asymptotic notation.

# 1 Big O notation (-upper bound function $g(f(n))$ )



$$f(n) \leq c \cdot g(n)$$

$$n \geq n_0$$

$$c > 0, n_0 \geq 1$$

$$\Rightarrow$$

$$f(n) = O(g(n))$$

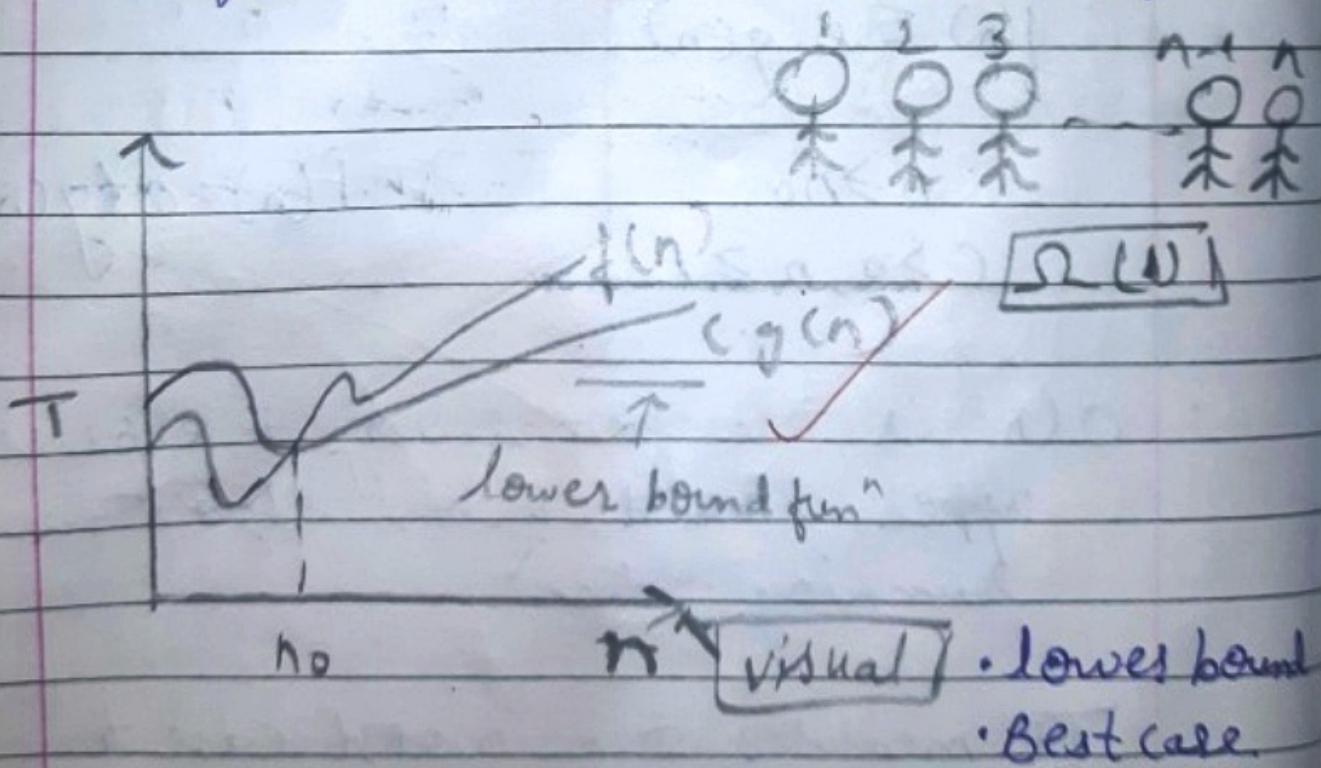
- $O(n)$  is the formal way to express the upper bound of an algorithm running time.

- It measures the worst case time complexity or longest amount of

time an algorithm can possibly take to complete.

## ② Omega Notation $\Omega$ (Big Omega) :-

The  $\Omega(n)$  is the formal way to express the lower bound of an algorithm's running time. It measures the best case time complexity or but possible amount of time an algorithm can take to complete.

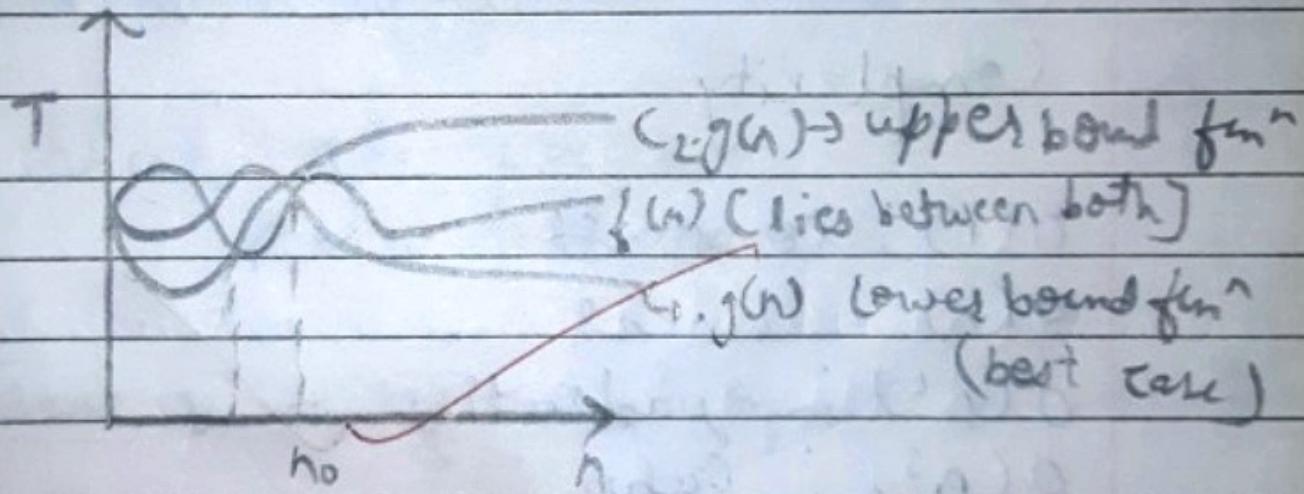


$$\begin{aligned} f(n) &\geq c \cdot g(n) \\ n &> n_0 \\ c &> 0, n_0 \geq 1 \end{aligned}$$

$$\Rightarrow f(n) = \Omega(g(n))$$

### ③ Big Theta ( $\Theta$ ) Notation

The  $\Theta(n)$  is the formal way to express both the lower bound and upperbound of an algorithm's running time



This graph represents the time growth of an algorithm with respect to input size  $n$

$$c_2 g(n) \geq f(n) \quad c_1 g(n)$$

$$n \geq n_0$$

$$c_1, c_2 > 0 \quad n_0 \geq 1$$

$\Rightarrow$

$$f(n) = O(g(n))$$

O

Upper bound

worst case

$\Omega$

Lower bound

Best case

$\Theta$

Between

Average case

## Complexity.

$O(1)$ : constant

$O(n)$ : Linear

$O(n^2)$ : quadratic

$O(n^3)$ : Cubic

$O(2^n)$ : exponential

$O(\log n)$ : log.

Q4 write all the three cases of master Theorem for the equation

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

Sol: Master Theorem :- for solving recurrence relation.

$$T(n) = aT\left(\frac{n}{b}\right) + \Theta(n^k \log^p n) \quad (\text{can be } \{1\} \text{ or two})$$

check  $a \geq 1, b \geq 1, k \geq 0$

$p$  is a real number

case 1 :- If  $a > b^k$  then  $T(n) = \Theta(n^{1/\rho})$

case 2  $\rightarrow$  If  $a = b^k$

(a) If  $p > -1$ , then  $T(n) = \Theta(n^{k+\frac{1}{p}} \log^{p+1} n)$

(b) If  $p = -1$  then  $T(n) = \Theta(n^{1/\rho} \log \log n)$

(c) If  $p < -1$ , then  $T(n) = \Theta(n \log^p n)$

case 3 If  $a < b^k$

(a) If  $p \geq 0$ , then  $T(n) = \Theta(n^k \log^p n)$

(b) If  $p < 0$ , then  $T(n) = \Theta(n^k)$

Q5 How can we prove that Strassen's matrix multiplication is advantages over ordinary matrix multiplication

Ans. It is used to solve the matrix multiplication problem

The usual matrix multiplication method multiplies each row with each column to achieve the product matrix time complexity taken by this approach is  $O(n^3)$ . Since it takes three loops to

multiply.

This method was introduced to reduce the time complexity from  $O(n^3)$  to  $O(n \log n)$

### \* Naive method of matrix multiplication

Let  $x$  is a matrix of order  $p \times q$   
 $y$  is a matrix of order  $q \times r$ .

$$z = [x]_{p \times q} [y]_{q \times r}$$

$z$  has the order  $p \times r$

Algorithm ~~matrix~~ - multiplication  
 $(x, y, z)$

for  $i = 1$  to  $p$

$j = 1$  to  $r$

$z[i, j] = 0$

for  $k=1$  to  $q$  do

$$z[i, j] = z[i, j] + x[i, k] \times y[k, j]$$

complexity  $\approx O(n^3)$

SMM Algorithm :-

With this algorithm the time  
complexity can be improved  
a little bit.

$$x = \begin{bmatrix} A & B \\ C & D \end{bmatrix} \quad y = \begin{bmatrix} E & F \\ G & H \end{bmatrix}$$

$$z = \begin{bmatrix} I & J \\ K & L \end{bmatrix}$$

$$M_1 := (A+C) \times (E+F)$$

$$M_2 := (B+D) \times (G+H)$$

$$M_3 := (A-D) \times (E-H)$$

$$M_4 = A \times (E - H)$$

$$M_5 = (C + D) \times E$$

$$M_6 = (A + B) \times H$$

$$M_7 = D \times (G - E)$$

Then,

$$I = M_2 + M_3 - M_6 - M_7$$

$$J = M_4 + M_6$$

$$K = M_5 + M_7$$

$$L = M_1 - M_3 - M_4 - M_5$$

Analysis :-

$$T(n) = ? \quad , \text{ if } n=1 \}$$

$$\left[ \cancel{(T \times T(\frac{n}{2}) \times d \times n^2)}, \text{ otherwise} \right]$$

where  $c$  and  $d$  are constants

~~By induction we get  $T(n) = O(n \log n)$ .~~