

9-3-2024

Date	/ /
Page	

## Assignment - 1

Q What is ADA? What is the need to study Algorithm. 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.

\* Need to Study Algorithm:-

- Problem-Solving:- Algorithm provide systematic approaches to solving various computational problems. By studying algorithms, you learn how to break complex problem.

- 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:- Algorithms often provide multiple solutions to the same problem, with varying trade-offs in terms of time space.
- Foundation for Computer Science:- Algorithms forms the backbone of Computer Science. They are fundamental to various subfields such that data structures.

- Problem Domain independence Algorithms:- Algorithms are not specific to any particular programming language or domain.
- Competitive programming and interviews  
In fields like software engineering algorithms knowledge is often a key component of technical interviews.
- Q) Give the Divide and Conquer. For Quick sort and analyse its Complexity?
- ⇒ Divide and Conquer Technique:-
- DSC is a top down approach to solve a problem the algorithm which involves DSC techniques involves 3 steps:-
- Divide:- The Original problems into a set of sub problems.



Date     /     /      
 Page    

• Conquer (or Solve):- every sub-problem recursive.

• Combine:- The solutions of these sub-problems to get the solution of original problem.

\* QUICK SORT (For Sorting)

- Pivot element based
- Partition based.

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

- 
- 1) Always pick first element as pivot
  - 2) Pick last element as Pivot
  - 3) Pick a random element as Pivot
  - 4) Pick median as pivot.

Ex:-

	1	2	3	4	5
	1	5	0	6	4

↑  
 $i=0$     $P=1$   
 $J=4$

$x$  is the pivot  
 $A[x] = x$   
 i.e.,  $x=4$

→ // Partition Function:-

Partition(A, P, r)

```

{
  x = A[r]
  i = P-1
  for (j = P to r-1)
    if (A[j] ≤ x)

```

```

      i = i+1
      exchange
      A[i] & A[j]

```

```

}
exchange
A[i+1] & A[r]
return (i+1)
}

```

→ // Algorithm:-

quickSort(A, P, r)

{  
if (P < r)

q = partition(A, P, r)

quickSort(A, P, q-1)

quickSort(A, q+1, r)

}

}

\* Complexity:-

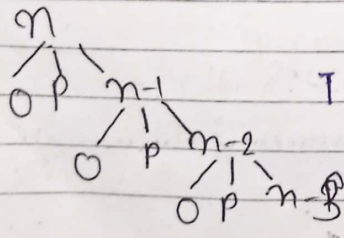
• Best case:-  $O(n \log n)$

• Average case:-  $O(n \log n)$

• Worst-Case:-  $O(n^2)$

- 2 when the array is already sorted

↓



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

$$O(n^2)$$

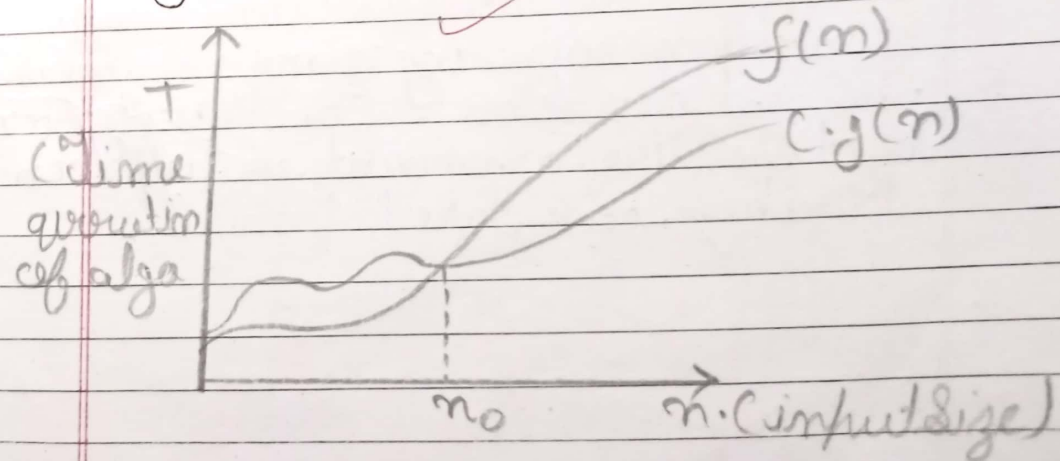
Q Define asymptotic notations. Give different notations used to represent the complexity of algorithm.

⇒ Definition:-

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

\* Different types notations:-

(i) Big O Notation:-





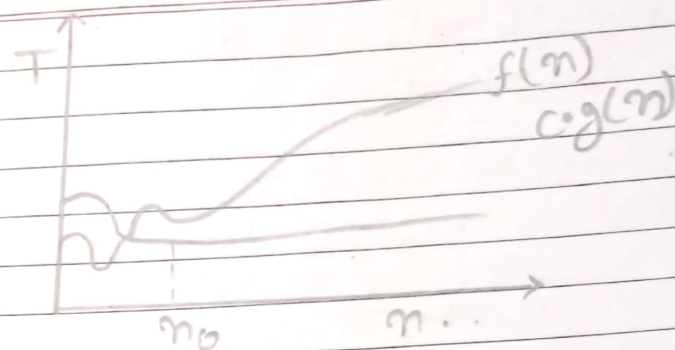
$$\boxed{\begin{array}{l} f(n) \leq C \cdot g(n) \\ n \geq n_0 \\ C > 0, n_0 \geq 1 \end{array}} \Rightarrow f(n) = O(g(n))$$

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

• It measures the <sup>worst</sup> case time complexity or longest amount of time an algorithm can possibly take to complete.

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

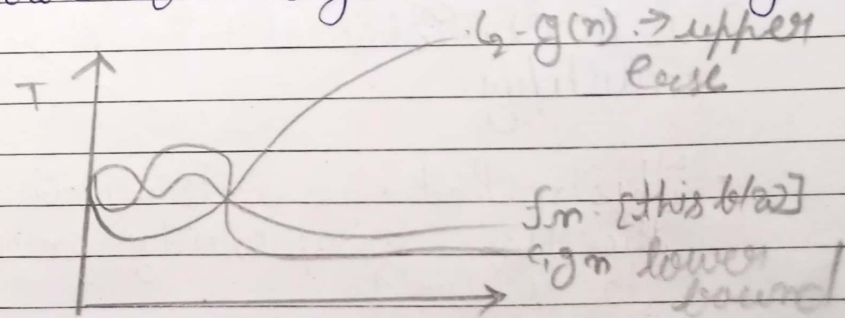
The  $\Omega(n)$  is the lower bound of an algorithm running time. It measures the best case time complexity or least possible amount of time an algorithm can take to complete.



$$\boxed{\begin{array}{l} f(n) \geq C \cdot g(n) \\ n > n_0 \\ C > 0, n_0 \geq 1 \end{array}} \Rightarrow \boxed{\frac{f(n)}{g(n)} \geq C}$$

## (3) Big Theta ( $\Theta$ ) Notation:-

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



- This graph represents the time growth of an algorithm

$$\boxed{\begin{matrix} C_2 g(n) \geq f(n) \geq C_1 g(n) \\ n \geq n_0 \\ C_1, C_2 > 0, n_0 \geq 1 \end{matrix}} \Rightarrow f(n) = O(g(n))$$

Q5) How can we prove that Strassen's Matrix multiplication is advantages over ordinary Matrix multiplication.

⇒

- 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  $O(n^{\log_2 7})$

\* Naive Method of Matrix Multiplication:-

Let  $X$  is a matrix of order  $p \times q$   
 $Y$  is a matrix of order  $q \times r$

$$Z = [X]_{p \times q} \cdot [Y]_{q \times r}$$

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

Algorithm Matrix-multiplication  
 $(X, Y, Z)$

for  $j = 1$  to  $p$

for  $i = 1$  to  $r$

$Z[j, i] := 0$

for  $k = 1$  to  $q$  do

$Z[j, i] = Z[j, i] + X[j, k] \times Y[k, i]$

Complexity =  $O(n^3)$

SMM Algorithm:- With this algo, the time consumption can be improved a little bit.



Date  /  /   
Page

$$X = \begin{bmatrix} A & B \\ C & D \end{bmatrix} \cdot Y = \begin{bmatrix} E & F \\ G & H \end{bmatrix}$$

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

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

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

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

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

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

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

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

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:-

Date  /  /   
Page

$$T(n) = \begin{cases} c & \text{if } n=1 \\ d \times T\left(\frac{n}{2}\right) \times 2 \times n^2 & \text{otherwise} \end{cases}$$

where  $c$  and  $d$  are constants  
We get  $T(n) = O(n^{\log})$

Q4) Write the three cases of Master theorem for the equation:

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

Master theorem:- for solving recurrence relation

$$T(n) = aT\left(\frac{n}{b}\right) + O(n^k \log n)$$

Check  $a \geq 1, b > 1, k \geq 0$   
 $p$  is a real number

Case 1:- If  $a > b^k$  then  $T(n) = O(n \log^p b)$

Case 2:- If  $a = b^k$  ✓

If  $p > -1$  then  $T(n) = O(n^{\log_b a} \log^{p+1} n)$

If  $p = -1$  then  $T(n) = O(n^{\log_b a} \log \log n)$

If  $p < -1$  then  $T(n) = O(n^{\log_b a})$

If  $a < b^k$

If  $p \geq 0$  then  $T(n) = O(n^k \log^p n)$

If  $p < 0$  then  $T(n) = O(n^k)$

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