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ASSIGNMENT-2

REG NO: 192B21003

SUBJECT: DESIGN AND
ANALYSIS OF ALGORITHM

SUBJECT CODE: CSA 0088

1. Explain Asymptotic Notation with details

- * The goal of analysis of an algorithm is to compare algorithm in running time & also memory management.
- * Enlarging the complexity in terms of its relationship to known function, this type analysis called asymptotic analysis.
- * Asymptotic Notation the mathematical way of representing the time complexity.
- * The Notations we use to describe the asymptotic running time of an algorithm are defined in terms of function whose domains are set of natural numbers.

There are three types of Notations.

- * Big oh (O) Notation
- * Big omega (Ω) Notation
- * Theta (Θ) notation

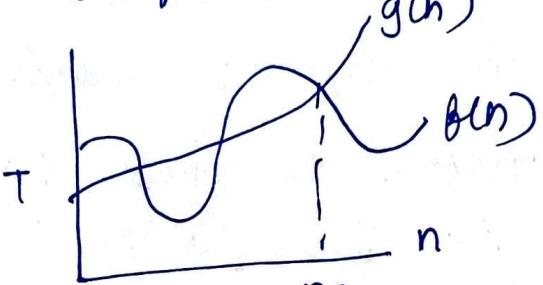
Big Oh (O) notation:

Asymptotic less than . This notation mainly represent upper bound of algorithm sum time. Big Oh notation is useful to calculate the maximum amount of time execution.

Formula:

$$f(n) \leq cg(n)$$

$O(g(n)) = \{ f(n) : \text{there exist positive constants } C \text{ and } n_0 \text{ such that } 0 \leq f(n) \leq Cg(n) \text{ for all } n \geq n_0 \}$



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

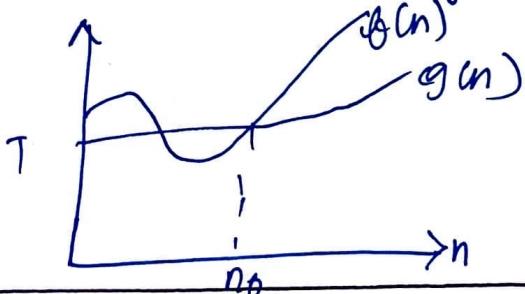
omega (Ω) notation:

Asymptotic notation "greater than" it represent lower bound of algorithm sum time by using omega notation we can calculate minimum amount of time

Formula:

$$f(n) \geq c g(n)$$

$\Omega(g(n)) = \{ f(n) : \text{exists positive constant } c > 0 \text{ and } n_0 \text{ such that } c \leq f(n) \leq g(n) \text{ for all } n \geq n_0 \}$

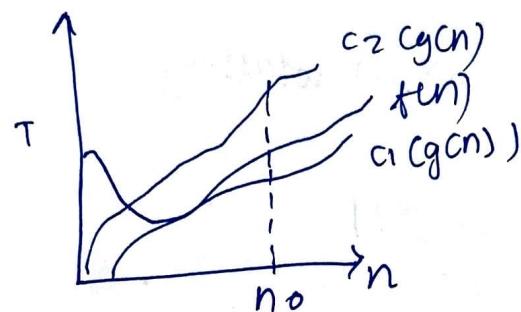


Theta(Θ) notation:

Sympotic "Equality", it represent average bound of algorithm running time. By using theta notation we can calculate average amount of time so it is called time complexity.

Formula:

$$\begin{aligned} c_1 g(n) &\leq f(n) = c_2 g(n) \\ \Theta(g(n)) &= f(n) \text{ there exist positive constant } c_1, c_2 \\ 0 \leq c_1 g(n) &\leq f(n) \leq c_2 g(n) \end{aligned}$$



Q.) Master theorem problem:

$$T(n) = 3T(n/4) + n$$

$$T(n) = 3T(n/4) + n \quad \text{using Master theorem}$$

$$a = 3, b = 4$$

$$\begin{aligned} a(n) &= n \\ \Rightarrow n &\log_4 3 \\ f(n) &\in n \log_4 3 \\ &= O(n) \end{aligned}$$

Hence proved

$$3) f(n) = 4n^3 + 2n^2 + n$$
$$g(n) = n^3$$
$$PT f(n) \text{ is } O(g(n))$$

$n=0:$

$$f(0) = 4(0)^3 + 2(0)^2 + 0$$
$$g(0) = 0^3$$

$O=0$
 $O(n) \text{ notation} //$

$n=1$

$$f(1) = 4(1)^3 + 2(1)^2 + 1$$

$$g(1) = 1^3$$

$7 < 1$
 $O(n) \text{ notation} //$

$n=2$

$$f(2) = 4(2)^3 + 2(2)^2 + 1$$

$$g(2) = 2^3$$

$16 < 8$
 $O(n) \text{ notation} //$

$n=3$

$$f(3) = 4(3)^3 + 2(3)^2 + 1$$

$$g(3) = 3^3$$

$64 < 27$
 $O(n) \text{ notation} //$

Apply the solution method to solve the recursive iteration method.

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

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

$$T(n) = 2(2T(n/4) + n/2) + n \rightarrow ①$$

$$T(n) = 2(2T(n/8) + n/4) + n \rightarrow ②$$

$$T(n) = 2(2T(n/16) + n/8) + n \rightarrow ③$$

$$T(n) = 2^k T(n/2^k) + kn$$

$$\frac{n}{2^k} = 1$$

$$n = 2^k$$

$$k = \log_2 n$$

$$T(n) = 2 \log_2 n + Cn(\log_2 n) + \log_2 n$$

$$n(1)$$

$$= nT(1) + (1) \log_2 1$$

$$= O(n \log n)$$

If near

proved //

5) $h(n) = 5n^4 + 3n^3 + n$ P.T $h(n)$ is $\Theta(n^4)$. Explain
the significance of constant factors in your proof.

$$n=0$$

$$\begin{aligned} h(0) &= 5(0)^4 + 3(0)^3 + 0 & h(0) &= 0 \\ &= 0 & &= 0 \end{aligned}$$

$$n=1$$

$$\begin{aligned} h(1) &= 5(1)^4 + 3(1)^3 + 1 & h(1) &= 1 \\ &= 9 & &= 4 \\ h(1) &= 9 > 4 & &= 4 \\ & & &= 4 \end{aligned}$$

condition satisfy

$$\begin{aligned} n=2 \\ h(2) &= 5(2)^4 + 3(2)^3 + 2 \\ &= 105 \end{aligned}$$

$$\begin{aligned} n=2 \\ h(2) &= 24 \\ &= 16 \end{aligned}$$

$$h(2) = 105 > 16$$

condition satisfy

$$n=3$$

$$\begin{aligned} h(3) &= 5(3)^4 + 3(3)^3 + 3 \\ &= 459 \end{aligned}$$

$$h(3) = 459 > 81$$

$$\begin{aligned} n=3 \\ h(3) &= 3^4 \\ &= 81 \end{aligned}$$

condition
satisfy

Explore Master theorem in detail.

The Master theorem is a tool used to solve recurrence relation that arise in the analysis of divide & conquer algorithm

This theorem provides a systematic way of solving recurrence relations

It's used to determine running time of algorithm in terms of asymptotic notations

$$T(n) = a(T)(n/b) + f(n)$$

$$f(n) = \Theta(n^k \log n^q)$$

case 1: if $\log b^a > k$ then $\Theta(n \log^q b)$

$$= T(n) = \Theta(n \log b^q)$$

case 2: if $\log b^a = k$ then $T(n) = \Theta(n \log b^k, \log n)$

case 3: if $f(n) > n \log^a b$ then $T(n) = \Theta(f(n))$

where :

n = size of input

a = no. of subproblems in the recursion

n/b = size of each problem

$f(n)$ = cost of work done outside the recursive call.