

ASSIGNMENT-2

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SUB: DESIGN AND

ANALYSIS OF ALGORITHM

SUB CODE: CSA 0685

1. Explain Asymptotic Notation with details.

- The goal of analysis of an algorithm is to compare algorithm in running time & also memory management.
- Expressing 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 notation we use to describe the asymptotic running time of an algorithm are defined in terms of function whose domain we set of natural numbers.

There are three types of Notations:

- Big oh (O) Notation.
- Big omega (Ω) Notation.
- Theta (Θ) Notation.

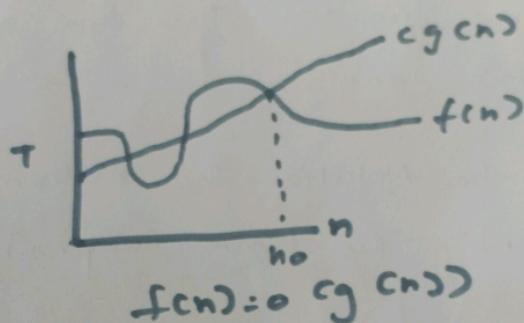
Big oh (O) notation:

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

Formula:

$$f(n) \leq Cg(n)$$

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

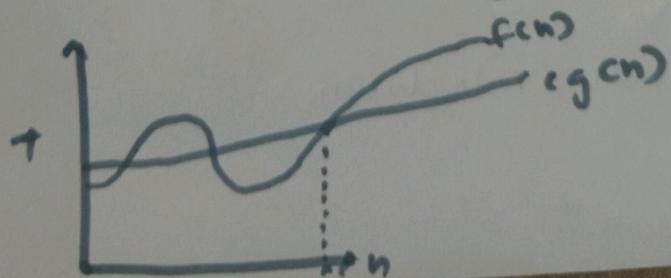


omega (Ω) notation:

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

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

$\Omega(g(n)) = \{ f(n) : \text{exist positive constants } c, n_0 \text{ such that } 0 \leq cg(n) \leq f(n) \text{ for all } n \geq n_0 \}$



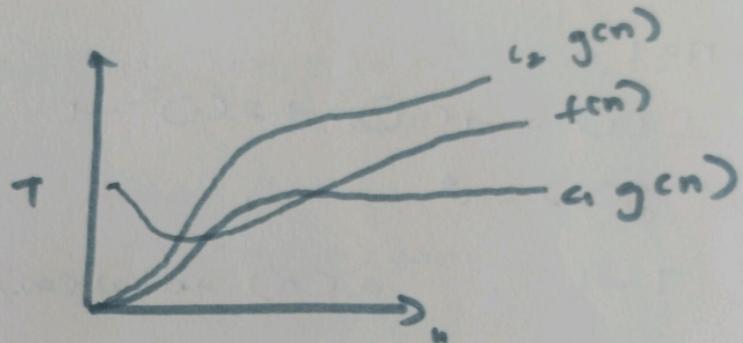
Theta(θ) notation:

Asymptotic "Equality" at 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:

$$c_1 g(n) \leq f(n) \leq c_2 g(n)$$

$\Theta(g(n)) = \{f(n) : \text{There exist positive constants } c_1, c_2\}$
 $0 < c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \in \mathbb{N}$



Master theorem problem:

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

$T(n) = 3T(n/4) + n$ using Master theorem.

$$a=3, b=4$$

$$f(n) = n$$

$$\rightarrow n \cdot \log_4 3$$

$$f(n) < n \log_4 3$$

$$= O(n)$$

Hence proved.

$$3. f(n) = 4n^3 + 2n^2 + n$$

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

P.T $f(n)$ is $\Theta(g(n))$

$n=0:$

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

$$g(0) = 0^3$$

$0 = 0.$

$\Theta(n)$ notation //

$n=1:$

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

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

$7 < 1 \quad \Theta(n)$ notation //

$n=2$

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

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

$33 < 8$

$\Theta(n)$ notation //

$n = 3$

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

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

$163 < 27 \quad \Theta(n)$ notation.

4. Apply the substitution method to solve the recurrence

$$T(n) = 2T(n/2) + n \quad \text{iteration method}$$

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

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

$$T(n) = 2(2T(n/4) + n/2) + n$$

$\hookrightarrow ①$

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

$\hookrightarrow ②$

$$T(n) = 2(2T(n/16) + n/8) + n$$

:

$\hookrightarrow ③$

times.

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

$$n = 4T(n/4) + 2n$$

$$n = 8T(n/8) + 3n$$

$$n = 2T(n/16) + 16n$$

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

$$n = 2^k$$

$$k = \log_2^n$$

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

$n \in \mathbb{C}$

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

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

Linear

Prooved //

5) $h(n) = 5n^4 + 3n^3 + n$ P.T. $h(n)$ is $\Theta(n^4)$

$n=0$

$$h(0) = 5(0)^4 + 3(0)^3 + 0 \quad n(0) 0^4 \\ = 0 \quad = 0$$

$n=1$

$$h(1) = 5(1)^4 + 3(1)^3 + 1 \quad n(1) = 1^4 \\ = 5 + 3 + 1 = 9 \quad = 1^4 \\ = 9$$

$$n(-2) = 9 > 4$$

$n=2$

$$h(2) = 5(2)^4 + 3(2)^3 + 2 \quad \text{condition satisfy} \\ = 160 + 24 + 2 = 186 \quad n=2 \\ = 186 \quad h(2) = 2^4 \\ = 16$$

$$n(-2) = 186 > 16$$

$n=3$

$$h(3) = 5(3)^4 + 3(3)^3 + 3 \quad n=3 \\ = 405 + 81 + 3 = 489 \quad h(3) = 3^4 \\ = 81$$

$$\therefore n(-2) = 489 > 81$$

condition satisfy

$$n=3 \\ h(3) = 3^4 \\ = 81$$

Condition satisfied

6. Explain Master theorem with details:

The Master theorem is a tool used to solve recurrence relation that arise in the analysis of divide and conquer algorithms.

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

$$f(n) = \Theta(n^k \log_n^3)$$

case 1 :

$$\text{If } \log_b^a > k \text{ Then } \Theta(n \log^a b)$$
$$= T(n) = \Theta(n \log_b^a)$$

case 2 :

$$\text{If } \log_b^a = k \text{ then } T(n) = \Theta(n \log_b^a, \log n)$$

case 3 :

$$f(n) > n \log_b^a \text{ then } T(n) = \Theta(f(n))$$

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
runtime call.