

1) If  $t_1(n) \in O(g_1(n))$  and  $t_2(n) \in O(g_2(n))$  then  $t_1(n) + t_2(n) \in O(\max\{g_1(n), g_2(n)\})$ . prove the assertions.

Sol: We need to show that  $t_1(n) + t_2(n) \in O(\max\{g_1(n), g_2(n)\})$ . This means there exists a positive constant  $C$  and  $n_0$  such that  $t_1(n) + t_2(n) \leq C$

$$t_1(n) \leq C_1 g_1(n) \text{ for all } n \geq n_1$$

$$t_2(n) \leq C_2 g_2(n) \text{ for all } n \geq n_2$$

$$\text{Let } n_0 = \max\{n_1, n_2\} \text{ for all } n \geq n_0$$

consider  $t_1(n) + t_2(n)$  for all  $n \geq n_0$

$$t_1(n) + t_2(n) \leq C_1 g_1(n) + C_2 g_2(n)$$

we need to relate  $g_1(n)$  and  $g_2(n)$  to  $\max\{g_1(n), g_2(n)\}$ .

$$g_1(n) \leq \max\{g_1(n), g_2(n)\} \text{ and } g_2(n) \leq \max\{g_1(n), g_2(n)\}$$

Thus.

$$C_1 g_1(n) \leq C_1 \max\{g_1(n), g_2(n)\}$$

$$C_2 g_2(n) \leq C_2 \max\{g_1(n), g_2(n)\}$$

$$C_1 g_1(n) + C_2 g_2(n) \leq C_1 \max\{g_1(n), g_2(n)\} + C_2 \max\{g_1(n), g_2(n)\}$$

$$C_1 g_1(n) + C_2 g_2(n) \leq (C_1 + C_2) \max\{g_1(n), g_2(n)\}$$

$$t_1(n) + t_2(n) \leq (C_1 + C_2) \max\{g_1(n), g_2(n)\} \text{ for all } n \geq n_0$$

By the definition of Big O Notation

$$t_1(n) + t_2(n) \in O(\max\{g_1(n), g_2(n)\})$$

$$t_1(n) + t_2(n) \in O(\max\{g_1(n), g_2(n)\})$$

Thus, the assertion is proved.

(2) find the time complexity of the recurrence equation  
let us consider such that recurrence for merge sort

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

By using master theorem

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

where  $a \geq 1$ ,  $b \geq 1$  and  $f(n)$  is positive function

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

$$a = 2, b = 2, f(n) = n$$

By comparing of  $f(n)$  with  $\log_b a$

$$\log_b a = \log_2 2 = 1$$

Compare  $f(n)$  with  $n \log_b a$

$$f(n) = n$$

$$n \log_b a = n^1 = n$$

\*  $f(n) = O(n \log_b a)$ , then  $T(n) = O(n \log_b a \log n)$

In our case

$$\log_b a = 1$$

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

— then — time complexity of recurrence is

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

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

$$(3) \quad T(n) = \begin{cases} 2T(n/2) + 1 & \text{if } n > 1 \\ 1 & \text{otherwise} \end{cases}$$

Sol: By Applying of master theorem  
 $T(n) = aT(n/b) + f(n)$  where  $a \geq 1$   
 $b > 1$

$$T(n) = 2T(n/2) + 1$$

Here  $a=2$ ,  $b=2$ ,  $f(n)=1$

By comparison of  $f(n)$  and  $n \log_b a$

If  $f(n) = O(n^c)$  where  $c < \log_b a$ , then  $T(n) = O(n \log_b a)$

If  $f(n) = O(n \log_b a)$ , then  $T(n) = O(n \log_b a \log n)$

If  $f(n) = \Omega(n^c)$  where  $c > \log_b a$ , then  $T(n) = O(f(n))$

Let's calculate  $\log_b a$ :

$$\log_b a = \log_2 2 = 1$$

$$f(n) = 1$$

$$n^{\log_b a} = n^1 = n$$

$f(n) = O(n^c)$  with  $c < \log_b a$  (case 1)

In this case  $c=0$  and  $\log_b a = 1$

$c < 1$ . So  $T(n) = O(n \log_b a) = O(n^1) = O(n)$

Time complexity of Recurrence Relation

$$T(n) = 2T(n/2) + 1 \text{ is } O(n)$$



$$(4) T(n) : \begin{cases} 2T(n-1) & \text{if } n > 0 \\ 1 & \text{otherwise} \end{cases}$$

sol:- Here, where  $n=0$

$$T(0) = 1$$

Recurrence Relation Analysis

-for  $n > 0$ :

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

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

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

$$T(n-2) = 2T(n-3)$$

$$T(1) = 2T(0)$$

from this pattern

$$T(n) = 2 \cdot 2 \cdot 2 \dots 2 \cdot T(0) = 2^n T(0)$$

Since  $T(0) = 1$ , we have

The recurrence relation is

$$T(n) = 2T(n-1) \text{ for } n > 0 \text{ and } T(0) = 1 \text{ is } T(n) = 2^n$$

5) Big O notation show that  $f(n) = n^2 + 3n + 5$  is  $O(n^2)$

$f(n) = O(g(n))$  means  $c > 0$  and  $n_0 > 0$

$$f(n) \leq c g(n) \text{ for all } n \geq n_0$$

Given is  $f(n) = n^2 + 3n + 5$

$$c > 0, n_0 \geq 0 \text{ such that } f(n) \leq c \cdot n^2$$

$$f(n) = n^2 + 3n + 5$$

Let's choose  $c = 9$

$$f(n) \leq 9 \cdot n^2$$

$$f(n) = n^2 + 3n + 5 \leq n^2 + 3n^2 + 5n^2 = 9n^2$$

So,  $c = 9, n_0 = 1$   $f(n) \leq 9n^2$  for all  $n \geq 1$

$f(n) = n^2 + 3n + 5$  is  $O(n^2)$