

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 exists a positive constant  $c$

$$t_1(n) + t_2(n) \leq c$$

$$t_1(n) \leq c_1 g_1(n) \text{ for all } n \geq n_1$$

$$t_2(n) \leq c_2 g_2(n) \text{ for all } n \geq n_2$$

Let  $n_0 = \max\{n_1, n_2\}$  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  $\max\{g_1(n), g_2(n)\}$ .  
 $g_1(n) \leq \max\{g_1(n), g_2(n)\}$  and  $g_2(n) \leq \max\{g_1(n), g_2(n)\}$

Thus

$$c_1 g_1(n) \leq c_1 \max\{g_1(n), g_2(n)\}$$

$$c_2 g_2(n) \leq c_2 \max\{g_1(n), g_2(n)\}$$

$$c_1 g_1(n) + c_2 g_2(n) \leq \max\{g_1(n), g_2(n)\} + c_2 \max\{g_1(n), g_2(n)\}$$

$$c_1 g_1(n) + c_2 g_2(n) \leq (c_1 + c_2) \max\{g_1(n), g_2(n)\}$$

$$t_1(n) + t_2(n) \leq (c_1 + c_2) \max\{g_1(n), g_2(n)\} \text{ for all } n \geq n_0$$

By the definition of Big O Notation

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

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

Hence proved.

2. Find the time complexity of the recurrence equation.

Let us consider such that recurrence for merge sort

$$T(n) = 2T\left(\frac{n}{2}\right) + n$$

By using master theorem

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

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

$$\text{Ex: } T(n) = 2T\left(\frac{n}{2}\right) + n$$

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

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

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

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

$$f(n) = n$$

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

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

In our case

$$\log_b a = 1$$

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

Time complexity of recurrence is

$$T(n) = 2T\left(\frac{n}{2}\right) + n$$

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



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

By applying of master theorem

$$T(n) = aT(n/b) + f(n) \text{ where } \begin{matrix} a \geq 1 \\ b \geq 1 \end{matrix}$$

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

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  $T(n) = O(f(n))$

Lets calculate  $\log_b a$ :

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

$$f(n) = 1$$

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

$$f(n) = O(n^c) \text{ with } c \leq \log_b a \text{ (case 1)}$$

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

$$c < 1, \text{ so } T(n) = O(n \log_b a) = O(n^1) = O(n)$$

Time complexity of Recurrence Relation

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

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

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

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

Since  $T(0) = 1$  we have 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 s.t  $f(n) = n^2 + 3n + 5$  is  $O(n^2)$

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

$f(n) \leq c(g(n))$  for all  $n > n_0$

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

$c > 0$   $n_0 > 0$  such that  $f(n) \leq c \cdot n^2$

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

$$c = 2$$

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

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

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

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