

(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.

we need to show that  $t_1(n) + t_2(n) \in O(\max\{g_1(n), g_2(n)\})$ . This means there means

There exists a positive constant  $C$  and  $n_0$  such that  $t_1(n) + t_2(n) \leq C$ .

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

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

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  $\max\{g_1(n), g_2(n)\}$ :

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

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.

Sol: 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  $n \log_b a$

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

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

$$f(n) = n$$

$$n \log_b a = n \cdot 1 = n$$

\*  $f(n) = O(n \log_b a)$ . 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 Relation is  $T(n) = 2T(n/2) + n$  is  $O(n \log n)$ .

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

Sol: By applying of master theorem.

$$T(n) = aT(n/b) + f(n) \text{ where } a \geq 1, b > 1$$

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

$$\text{Here } a=2, b=2, f(n)=1$$

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  $\log_b a$ :

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

$$f(n) = 1$$

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

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

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

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

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

Ans: Here, where  $n=0$

$$T(0) = 1$$

Recurrence Relation Analysis.

for  $n > 0$ :

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

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

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

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

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

from this pattern.

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

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

$$T(n) = 2^n$$

The Recurrence Relation is

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

(2) Big O Notation. Show that  $f(n) = n^2 + 3n + 5$  is  $O(n^2)$ .

Def:  $f(n) = O(g(n))$  means  $C > 0$  and  $n_0 \geq 0$

$f(n) \leq C \cdot g(n)$  for all  $n \geq n_0$ .

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

$C > 0, n_0 \geq 0$  such that  $f(n) \leq Cn^2$

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

Let's choose  $C = 9$ .

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

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

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

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