

1. If $f_1(n) = O(g_1(n))$ and $t_2(n) = O(g_2(n))$ then $t_1(n) + t_2(n) = O(\max\{g_1(n), g_2(n)\})$. Prove the assertion.

Solution:-

By definition, there exist constants C_1, n_1 such that for all $n > n_1$; $t_1(n) \leq C_1 \cdot g_1(n)$

Similarly there exist constants C_2, n_2 such that for all $n > n_2$ $t_2(n) \leq C_2 \cdot g_2(n)$

Let $n_0 = \max(n_1, n_2)$ and $C = C_1 + C_2$ for all $n > n_0$:

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

By definition of maximum:

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

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

Thus,

$$t_1(n) + t_2(n) \leq C_1 \cdot \max\{g_1(n), g_2(n)\} + C_2 \cdot \max\{g_1(n), g_2(n)\}$$

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

$$= C \cdot \max\{g_1(n), g_2(n)\}$$

$$t_1(n) + t_2(n) \leq (C_1 + C_2) \cdot \max\{g_1(n), g_2(n)\}$$

$$= C \cdot \max\{g_1(n), g_2(n)\}$$

Hence

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

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

Solution:-

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

$$n^2 + 3n + 5 \leq C \cdot n^2 \quad f(n) = n^2 + 3n + 5$$

$$\text{for } C = 2 \text{ and } n_0 = 3 \quad g(n) = n^2$$

$$n^2 + 3n + 5 \leq 2n^2$$

for all $n > 3$

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

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 $a > 1$, $b > 1$ 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 $n \log_b a$

$$f(n) = n$$

$$n \log_b a = n' = n$$

$$\log_b a = 1 \quad T(n) = O(n' \log n) = O(n \log n)$$

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

3.

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

By Applying of master's theorem

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

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

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

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 $\log_b a = \log_2 2 = 1$

$$f(n) = 1$$

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

$f(n) = O(n^c)$ 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.

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

Here, where $n=0$

$$T(0) = 1$$

Recurrence relation analysis

for $n > 0$:

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

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

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

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

from this pattern

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

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

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

The recurrence relation is

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

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