

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.

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

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

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

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

$$\text{consider } t_1(n) + t_2(n) \text{ For all } n \geq n_0$$

$$t_1(n) + t_2(n) \leq c_1 g_1(n) + c_2 g_2(n)$$

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

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

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

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

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

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

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

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.

A. 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 $n^{\log_b a}$:

$$f(n) = n$$

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

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

In our case:

$$\log_b a = 1$$

$$T(n) = O(n^1 \log n) = O(n \log n)$$

The time complexity of recurrence Relation is $T(n) = 2T(n/2) + n$

is $O(n \log n)$

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

A. By Apply of master theorem

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

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

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

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 $T(n) = O(n^{\log_b a} \cdot \log n)$

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

Lets calculate $\log_b a$:

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

$$F(n) = 1$$

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

$F(n) = O(n^c)$ with $c < \log_b a$ (Case 1)

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

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

Time complexity of Recurrence relation

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

4)

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

A.

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

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

The recurrence relation is $T(n) = 2T(n-1)$ for $n > 0$ and

$$T(0)=1 \text{ is } T(n)=2^n$$

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

A. $F(n) = O(g(n))$ means $C > 0$ and $n_0 \geq 0$

$$F(n) \leq C \cdot g(n) \text{ for all } n \geq n_0$$

$$\text{Given is } F(n) = n^2 + 3n + 5$$

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

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

Let choose $C = 2$

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

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

$$\text{So, } C = 9, n_0 = 1 \quad F(n) \leq 9n^2 \text{ for all } n \geq 1$$

$$F(n) = n^2 + 3n + 5 \text{ is } O(n^2)$$