

① If $t_1(n) = O(g_1(n))$ 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 assertions.

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

Similarly there exist constants C_2, n_2 such that for all $n \geq n_2$:

$$t_2(n) \leq C_2 g_2(n)$$

Let $n_0 = \max\{n_1, n_2\}$ and $C = C_1 + C_2$. For all $n \geq n_0$: $t_1(n) + t_2(n) \leq C_1 g_1(n) + C_2 g_2(n)$

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,

$$t_1(n) + t_2(n) \leq C_1$$

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

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

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

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

Hence

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

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

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

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

For $C = 2$ and $n_0 = 3$

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

for all $n \geq 3$

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

$$g(n) = 2n^2$$

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

② Find the time complexity of the recurrence equation.

Let us consider such that recurrence for merge sort

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

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

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

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

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

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

compare $f(n)$ with $n^{\log_b a}$

$$f(n) = n$$

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

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

In our case

$$\log_b a = 1$$

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

Then time complexity of recurrence relation is

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

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

By applying of master theorem

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

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

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

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

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

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

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

lets calculate $\log_b a$.

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

$$f(n) = 1$$

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

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

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

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

The complexity of recurrence relation

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

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

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

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

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

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

The recurrence relation is

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

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