

Fundamental of Algorithms

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Introduction:-Complexity of recursive algorithms →

- When an algorithm contains a recursive call to itself, its running time is described by a recurrence equation or recurrence.
- A recurrence describes the overall running time of a problem of size  $n$ .
- Mathematical tools are used to solve these recurrences.
- These tools are:-
  - 1) Substitution Method.
  - 2) Recursion tree Method.
  - 3) Master Method.

1) Substitution Method →

- The solution is guessed and mathematical induction is used to prove that the guess is correct or incorrect.
- This method is powerful, but it can be applied only in cases when it is easy to guess the form of the answer.

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eg) Consider a recurrence relation  
 $T(n) = T(n-1) + n$   
with initial condition  $T(0) = 0$ .

sol<sup>n</sup>:  $T(n) = T(n-1) + n$

If  $n=1$ , then  $T(1) = T(0) + 1$   
 $= 0 + 1$   
 $= 1$

If  $n=2$ , then  $T(2) = T(1) + 2$   
 $= 1 + 2$   
 $= 3$

If  $n=3$ ,  $T(3) = T(2) + 3$   
 $= 3 + 3$   
 $= 6$

Thus,

$n$	$T(n)$
1	1
2	3
3	6
$\vdots$	$\vdots$
$\vdots$	$\vdots$

$$T(n) = \frac{n(n+1)}{2}$$

This can be denoted in terms of Big-Oh notation as -

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

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eg) Solve the following recurrence relation using substitution method.

$$T(n) = T(n-1) + 1 \quad \text{where } T(0) = 0.$$

sol<sup>n</sup>:- If  $n=1$ ,  $T(1) = T(0) + 1$   
 $= 0 + 1$   
 $= 1$

If  $n=2$ ,  $T(2) = T(1) + 1$   
 $= 1 + 1$   
 $= 2$

If  $n=3$ ,  $T(3) = T(2) + 1$   
 $= 2 + 1$   
 $= 3$

Thus,

$n$	$T(n)$
1	1
2	2
3	3

$$T(n) = n$$

ie.  $T(n) = O(n)$ .

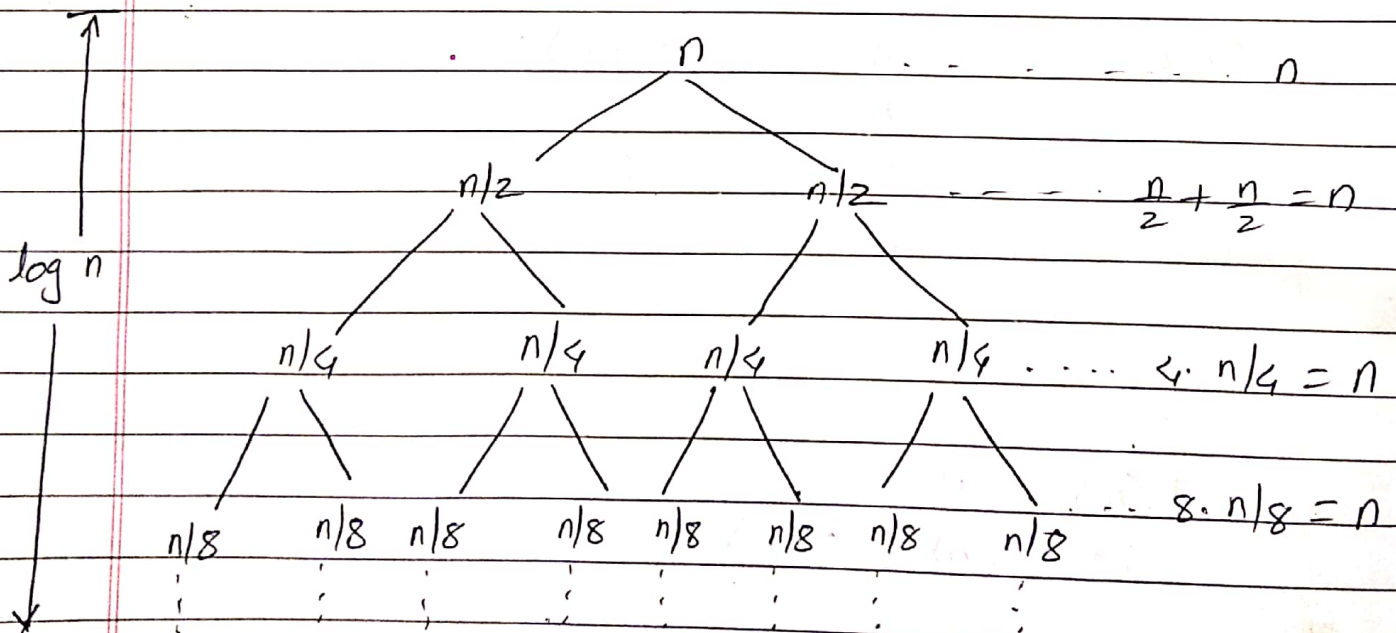


## 2) Recursion tree method →

- In this method, we draw a recurrence tree and calculate the time taken by every level of the tree.
- Finally, we sum the work done at all the levels.
- To draw the recurrence tree, we start from the given recurrence relation and keep drawing till we find a pattern among all the levels. This pattern is typically a arithmetic or geometric series.

eg) Solve the recurrence relation using recursion tree method.

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



The depth of the tree is  $\log n$ .  
Hence, the total cost =  $n \log n$ .  $T(1)$   
 $= n \log n$

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3) Master Method →

- It is a direct way to get the solution.
- Master method works only for the following type of recurrences.

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

where,  $a$  is constant and  $a \geq 1$   
 $n \geq d$  and  $d$  is some constant

[ $F(n)$  must be positive]

case 1: If  $F(n)$  is  $n^d$  where  $d \geq 0$ , then

a) if  $a < b^d$ ,  $T(n) = O(n^d)$

b) if  $a = b^d$ ,  $T(n) = O(n^d \log n)$

c) if  $a > b^d$ ,  $T(n) = O(n^{\log_b a})$

case 2: If  $F(n)$  is  $(n^{\log_b a})$ , then

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

case 3: If  $F(n)$  is  $(n^{\log_b a} \log^k n)$ , then

$$T(n) = O(n^{\log_b a} \log^{k+1} n)$$

case 4: If  $F(n)$  is  $(n^{\log_b a + c})$ , then

$$T(n) = O(F(n))$$



eg) Solve using Master Method →

1)  $T(n) = 4T(n/2) + n$

Sol<sup>n</sup>: we have,  $T(n) = aT(n/b) + F(n)$

Here,  $a=4$ ,  $b=2$  and  $F(n)=n$  [case 1]  
ie.  $n^1$ , hence  $d=1$

Now,  $b^d = 2^1 = 2$

Thus,  $a > b^d$  [case 1c]

$$\begin{aligned}\therefore T(n) &= O(n^{\log_b a}) \\ &= O(n^{\log_2 4}) \\ &= O(n^2)\end{aligned}$$

Thus, the time complexity is  $O(n^2)$

2)  $T(n) = 2T(n/2) + n \log n$

Sol<sup>n</sup>: Here,  $a=2$ ,  $b=2$  and  $F(n)=n \log n$  [case 3]

$$\therefore T(n) = O(n^{\log_b a} \log^{k+1} n)$$

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

$$\log^{k+1} = \log^{1+1} = \log^2$$

$$\begin{aligned}\therefore T(n) &= O(n^{\log_b a} \log^{k+1} n) \\ &= O(n^1 \log^2 n) \\ &= O(n \log^2 n)\end{aligned}$$

3)  $T(n) = 4T(n/2) + n^2$

sol<sup>n</sup>:-

Here,

$a=4$ ,  $b=2$  and  $F(n)=n^2$  [case 1]  
ie.  $d=2$

Now,  $b^d = 2^2 = 4$

Here,  $a=b$  (case 1b)

$$\therefore T(n) = \Theta(n^d \log n)$$
$$= \Theta(n^2 \log n)$$

4)  $T(n) = 2T(n/2) + n^3$

sol<sup>n</sup>:-

Here,

$a=2$ ,  $b=2$ ,  $F(n)=n^3$  [case 1]  
ie.  $d=3$

Now,  $b^d = 2^3 = 8$

Here,  $a < b^d$  [case 1a]

$$\therefore T(n) = \Theta(n^d)$$
$$= \Theta(n^3)$$

5)  $T(n) = 2^n T(n/2) + n^n$

Here,  $a=2^n$ ,  $b=2$  and  $d=n$

$a$  and  $d$  should be constant.

Hence, master method is not applicable.

$$6) \quad T(n) = 2T(n/4) + n^{0.51}$$

Here,  $a=2$ ,  $b=4$ ,  $d=0.51$

$$b^d = 4^{0.51} = 2.03$$

$$a < b^d \quad (\text{case 1a})$$

$$\therefore T(n) = O(n^{0.51})$$

$$7) \quad T(n) = 0.5T(n/2) + 1/n$$

Here,  $a=0.5$ ,  $b=2$ ,  $F(n)=1/n$

Since,  $a < 1$ , master method is not applicable

$$8) \quad T(n) = 9T(n/3) + n^2 \log n$$

Here,  $a=9$ ,  $b=3$  and  $F(n)=n^2 \log n$  [case 3]

$$\therefore T(n) = O(n^{\log_b a} \log^{k+1} n)$$

$$\log_b a = \log_3 9 = 2$$

$$\log^{k+1} = \log^{1+1} = \log^2$$

$$\therefore T(n) = O(n^2 \log^2 n)$$

$$9) \quad T(n) = 64T(n/8) - n^2 \log n$$

Here,  $a=64$ ,  $b=8$  and  $F(n)=-n^2 \log n$

Master method not applicable, since  $F(n)$  is negative.