

# AOA (DK) Part - 2

## Recurrences

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Recursive function - Function calling itself.

Recurrence - Is an equation or inequality that describes a function in terms of its value on smaller inputs.

Example to understand how recurrence is counted from algorithm?

Algorithm ABC(n) ...

```
if (n > 1)
    return (ABC(n-1))
```

?

$$T(n) = 1 + T(n-1) \quad \text{if } n > 1$$

$$= O(1) \quad \text{if } n = 1$$

There are three methods to solve recurrences :-

- 1) The Substitution Method
- 2) The Recursion Tree Method
- 3) The Master Method.

## [I] The Substitution Method

There are two steps

(i) Guess the form of the solution

(ii) Use mathematical induction to find constant & show that solution works.

### Examples

I]  $T(n) = 2T(n/2) + n$  for  $n > 1$  and  $T(1) = 1$   
Determine upper bound.

Soln  $\rightarrow$  Step 1: Guess the Solution

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

Step 2: Mathematical Induction

a) To prove:  $T(n) \leq cn \log n$  --- (I) for some constant c

~~Do it~~

b) Finding Constant c:

Let us assume that bound holds for  $T(n/2)$ .

$$\therefore T(n/2) \leq c(n/2) \log(n/2) \quad \text{--- (II)}$$

Substituting equation II in recurrence relation,

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

$$T(n) = 2 \left[ c \frac{n}{2} \log\left(\frac{n}{2}\right) \right] + n$$

$$= cn \log\left(\frac{n}{2}\right) + n$$

$$= cn \log n - cn \log 2 + n$$

$$T(n) = cn \log n - cn + n \quad \dots \text{III}$$

We need to prove,

$$\text{e. } T(n) \leq cn \log n$$

$$cn \log n - cn + n \leq cn \log n$$

$$-cn + n \leq 0$$

Dividing by  $n$ ,

$$-c + 1 \leq 0$$

$$\boxed{c > 1}$$

(c) Proof of the base condition:

Suppose, put  $n = 1$  in eqn I.

$$T(1) \leq c(1) \log(1) \leq 0$$

But  $T(1) = 1$ . Hence Not satisfied.

(d) Reverse the induction

Need to find lowest value of  $n_0$  for which relation is satisfied.

Consider,  $n_0 = 2$

$$T(2) = 2T\left(\frac{1}{2}\right) + C$$

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

$$\boxed{T(2) = 4}$$

Now, for relation.

$$T(n) \leq cn \log n$$

$$T(2) \leq c(2 \log 2)$$

$$T(2) \leq 2c$$

For sufficient large value of  $c$ ,  $c \geq 2$  the condition holds the relation.

$$\therefore T(n) = O(n \log n) \text{ where } n_0 = 2$$

$$c = 2$$

For,  $n_0 = 3$ ,

$$T(n) = T(3) = 2T\left(\frac{3}{2}\right) + 3$$

$$= 2T(1) + 3$$

$$= 5$$

for relation,

$$T(n) \leq cn \log n$$

$$T(3) \leq 2 \times 3 \log 3$$

$$5 \leq 2 \times 3 (1.5)$$

$$5 \leq 9$$

$\therefore$  satisfied,  $\therefore T(n) = O(n \log n)$   $n_0 = 2$

$$c = 2$$

2]

$T(1) = 1$  and  $T(n) = T(n/2) + 1$  for  $n > 1$   
 determine the upper bound

Solution: Step 1: Guess the solution.

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

Step 2: Mathematical Induction

- (a) To prove:  $T(n) \leq c \log n$  for some constant  $c$
- (b) Pending constant  $c$ :

Let us assume that bound holds for  $T(n/2)$

$$\therefore T(n/2) \leq c \log(n/2) \quad \text{II}$$

using equation II in recurrence equation,

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

$$= c \log(n/2) + 1$$

$$= c \log n - c + 1 \quad \text{III}$$

We need to prove,

$$T(n) \leq c \log n$$

$$c \log n - c + 1 \leq c \log n$$

$$-c + 1 \leq 0$$

$$\boxed{c \geq 1}$$

(c) Prove the base condition:

Suppose, put  $n=1$  in equation I

$$T(1) \leq C \log 1 \leq 0$$

But  $T(1) = 1$ , hence Not satisfied.

(d) Reverse the induction

Need to find the lowest value of  $n_0$  for which relation is satisfied.

Consider  $n_0 = 2$

$$\begin{aligned} T(2) &= T\left(\frac{2}{2}\right) + 1 \\ &= 2 \end{aligned}$$

Now, for relation,

$$\begin{aligned} T(n) &\leq c \log n \\ T(2) &\leq c \log 2 \\ T(2) &\leq c \end{aligned}$$

Consider  $n_0 = 3$

$$\begin{aligned} T(3) &= 2T\left(\frac{3}{2}\right) + 1 \\ &= T(1) + 1 \\ &= 2 \end{aligned}$$

Now for relation

$$\begin{aligned} T(n) &\leq c \log 3 \\ T(3) &\leq 1.5c \end{aligned}$$

for sufficient larger value of  $c$ ,  $c \geq 2$  the condition holds the relation.

$$\therefore T(n) = O(\log n) \quad \text{for } n_0 = 2$$

$$c = 2$$

3] Show that solution for  $T(n) = nT(n-1) + n$  is  $O(n^2)$

Soln → Step 1: Guess the solution & find exact form  
 $T(n) = O(n^2)$

Exact form:

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

$$T(1) = 1$$

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

$$T(3) = T(3-1) + 3 = 1+2+3$$

$$T(4) = 1+2+3+4$$

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

Step 2: Mathematical Induction

(a) To prove:  $T(n) \leq c \left[ \frac{n(n+1)}{2} \right]$  — (I)

for some constant  $c$ .

(b) Finding constant  $c$ :

Let us assume that bound holds for  $n(n-1)$ .

$$\therefore T(n-1) \leq c \left[ \frac{(n-1)n}{2} \right] \dots \text{--- (II)}$$

Using equation II, in recurrence equation,  
 $T(n) = T(n-1) + n$

$$= c \left[ \frac{n(n-1)}{2} \right] + n$$

$$= \frac{c}{2} n^2 - \frac{c}{2} n + n$$

We need to prove,

$$T(n) \leq c \left[ \frac{n^{(n+1)}}{2} \right]$$

$$\frac{cn^2}{2} - \frac{cn+n}{2} \leq \cancel{\frac{cn^2}{2}} + \frac{c}{2} n$$

$$n \leq cn$$

Dividing by  $n$ ,

$$\boxed{c \geq 1}$$

(c) Prove the base condition:

Suppose, put  $n=1$  in equation T.

$$\begin{aligned} T(1) &\leq c \left( \frac{1^{(1+1)}}{2} \right) \\ &\leq c \left( 1 \left( \frac{1+1}{2} \right) \right) \\ &\leq c \\ &\leq 1 \quad [c = 1] \end{aligned}$$

which is satisfied.

$\therefore T(n) = O(n^2)$  for  $c = 1$

$$n_0 = 1$$

4)

Show that solution for  $T(n) = 2T\left(\frac{n}{2}\right) + cn$  is  $O(n \log n)$ . 48

Solution:

Step 1: Guess the solution  
 $\therefore T(n) = O(n \log n)$

Step 2: Mathematical Induction

(a) To prove:  $T(n) \leq cn \log n$  ... (I)

(b) Finding constant c:

Let us assume that the bound holds true for  $T\left(\frac{n}{2} + 17\right)$

$$\therefore T\left(\frac{n}{2} + 17\right) = c\left(\frac{n}{2} + 17\right) \log\left(\frac{n}{2} + 17\right) \quad (\text{--- II})$$

using equation (II) in recurrence equation.

$$\therefore T(n) = 2 \left[ c\left(\frac{n}{2} + 17\right) \log\left(\frac{n}{2} + 17\right) \right] + n.$$

$$= cn \log\left(\frac{n}{2} + 17\right) + 34c \log\left(\frac{n}{2} + 17\right) + n$$

$$= cn \log\left(\frac{n+34}{2}\right) + 34c \log\left(\frac{n+34}{2}\right) + n$$

$$= cn \log(n+34) - cn + 34c \log(n+34) - 34c + n$$

$$= cn \log\left(n\left(1+\frac{34}{n}\right)\right) + 34c \log\left(n\left(1+\frac{34}{n}\right)\right) - c(n+34) + n$$

$$= \underline{\underline{cn \log n}} + \underline{\underline{cn \log\left(1+\frac{34}{n}\right)}} + \underline{\underline{34c \log n}} + \underline{\underline{34c \log\left(1+\frac{34}{n}\right)}} \\ - c(n+34) + n$$

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$$= c(n+34)\log n + c(n+34)\log\left(1+\frac{34}{n}\right) - c(n+34)$$

+ n

When  $n$  is very large, then  $\frac{34}{n}$  tends to zero  
and  $n+34 \approx n$ .

$$T(n) = cn\log n + cn\log(1) = nc + n$$

We need to prove,

$$T(n) \leq cn\log n$$

$$cn\log n - cn + n \leq cn\log n$$

Dividing by  $n$ ,

$$-c + 1 \leq 0$$

$$\therefore \boxed{c \geq 1}$$

(c) Prove the base condition

Suppose : put  $n=1$  in eqn ①

$$T(1) \leq c(1)\log(1) \leq 0$$

But  $T(1) = 1$ . Hence not satisfied.

(d) Reverse the induction

For larger values of  $n$ ,  $n_{\frac{n}{2}} + 17 \approx n_{\frac{n}{2}}$ .

$$\begin{aligned} T(n) &= 2T\left(\frac{n}{2}\right) + n \\ &\approx 2T\left(\frac{n}{2}\right) + n \end{aligned}$$

We need to find lowest value of  $n_0$  for which relation is satisfied.

Consider,

$$n_0 = 2$$

$$\begin{aligned} T(2) &= 2T(1) + 2 \\ &= 4 \end{aligned}$$

For Relation,

$$\begin{aligned} T(2) &\leq C_2 \log 2 \\ &\leq 2C \end{aligned}$$

$$n_0 = 3$$

$$\begin{aligned} T(3) &= 2T\left(\lceil \frac{2}{3} \rceil\right) + 3 \\ &= 2T(2) + 3 \\ &= 8 + 3 \\ &= 11 \end{aligned}$$

For Relation

$$\begin{aligned} T(3) &\leq C_3 \log 3 \\ &\leq 4.5C \end{aligned}$$

For sufficient larger value of  $C$ ,  $C \geq 3$  then condition holds the relation.

$$\therefore T(n) = O(n \log n) \quad \text{for } n_0 = 2$$

$$C = 3$$

5] Solve the recurrence  $T(n) = 2T(\lfloor \sqrt{n} \rfloor) + \log n$   
by making change of variable

Solution : 1] Changing the Variable

$$\text{Let } m = \log n \quad \rightarrow ①$$

$$\therefore n = 2^m \quad \rightarrow ②$$

$$\therefore \sqrt{n} = 2^{m/2} \quad \rightarrow ③$$

$\therefore$  substituting ①, ②, ③ in recurrence equation,

$$T(2^m) = 2T(2^{m/2}) + m$$

Now Rename,  $s(m) = T(2^m)$

$$\therefore s(m) = 2s(m/2) + m \quad \rightarrow ④$$

2] Prove :

(a) Guess the Solution :

$$s(m) = O(m \log m) \quad \rightarrow ⑤$$

(b) Mathematical induction

To prove :  $s(m) \leq cm \log m \quad \rightarrow ⑥$

Finding c :

Let us assume bound holds for  $s(m/2)$

$$\therefore s(m/2) \leq cm/2 \log m/2 \quad \rightarrow ⑦$$

Substituting eqn ⑦ in ④

$$s(m) = 2 \left[ \frac{cm}{2} \log \frac{m}{2} \right] + m \\ = cm \log m - cm + m$$

We need to prove,

$$s(m) \leq cm \log m$$

$$cm \log m - cm + m \leq cm \log m$$

Multiplying by  $m$ ,

$$-c + 1 \leq 0$$

$$\Rightarrow c \geq 1$$

Prove the base condition:

put  $m = 1$ , in equation ⑥

$$s(1) \leq c(1) \log(1) = 0$$

But  $s(1) = 1$ , Hence not satisfied.

Revised the production

Consider,  $m_0 = 2$

$$s(2) = 2s(1) + 2$$

$$= 4$$

For Relation,

$$s(2) \leq c \cdot 2 \log 2$$

$$\leq 2 \cdot c$$

for sufficient larger value of  $c$ ,  
 $c \geq 2$ , relation holds.

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$$\therefore S(m) = O(m \log m) \quad \text{for } \frac{c=2}{m_0=2}$$

3] Changing Back from  $S(m)$  to  $T(n)$

$$\begin{aligned} T(n) &= T(2^m) \\ &= S(m) \\ &= O(m \log m) \end{aligned}$$

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

$$\begin{aligned} n_0 &= 2^{m_0} \\ &= 2^2 \end{aligned}$$

$$n_0 = 4$$

$$\therefore T(n) = O(\log n \log \log n) \quad \text{for } \frac{c=2}{n_0=4}$$

## The Recursion Tree Method

- In this method, recursion tree is constructed.
- Using the tree, recurrence relation is solved.
- In the tree, each node represents the cost of single subproblem somewhere in the set of recursive function / invocation.
- The method is effectively used in divide and conquer strategy.
- It is used to generate good guess.
- After solution is found by recursion tree method, it is verified by substitution method.

### Example Steps:

- i. Draw the Recursion tree
2. Cost of each level
3. Find the depth of tree
4. Find the number of leaves (No. of nodes at last level)
5. Find total cost last level.
6. Total cost for entire tree.

### 3 cases :

1. Cost of Root node is Maximum
  2. Cost of Leaf node is Maximum
  3. Cost of Each ~~node~~ is Same
- } All leaf nodes are at same level

### Problem on Case 1

$$\text{Ex: } T(n) = 3T\left(\frac{n}{4}\right) + \Theta(n^2)$$

$$\therefore T(n) = 3T\left(\frac{n}{4}\right) + cn^2$$

The above statement says that the complete problem of size 'n' is divided into three subproblems of size ' $n/4$ ' and the operations cost is  $cn^2$ .

### Step 1: Recursion Tree

Recursive call	Number of Nodes	Tree	Row Sum
$T(n)$	1	$cn^2$	$cn^2$
$T(n/4)$	3	$\frac{cn^2}{16}$ $\frac{cn^2}{16}$ $\frac{cn^2}{16}$	$\frac{3}{16} cn^2$
$T(n/16)$ $= T(n/4^2)$	$3^2$	$\frac{(cn)^2}{(16)^2}$ $\frac{(cn)^2}{(16)^2}$ $\frac{(cn)^2}{(16)^2}$ $\frac{(cn)^2}{(16)^2}$ $\frac{(cn)^2}{(16)^2}$ $\frac{(cn)^2}{(16)^2}$ $\frac{(cn)^2}{(16)^2}$ $\frac{(3^2)}{(16)} cn^2$	
$T(n/4^i)$	$3^i$	$c\left(\frac{n}{16^i}\right)^2$	$\left(\frac{3^i}{16}\right) cn^2$

Step 2: Cost of each level =  $\left(\frac{3}{16}\right)^i cn^2$

Step 3: Depth of tree i.e last level of tree,  
where

$$\frac{n}{4^i} = 1$$

$$n = 4^i$$

$$\log_4 n = \log_4 4^i$$

$$\log_4 n = i$$

$$i = \log_4 n$$

Step 4: Number of leaves i.e No. of nodes  
in the last level

$$= 3^i$$

$$= 3^{\log_4 n}$$

$$= n^{\log_4 3} = n^{0.79248}$$

Step 5: Total cost at last level

$$= n^{\log_4 3} \times T(1)$$

$$= \Theta(n^{\log_4 3})$$

Step 6: Total cost of entire tree,

$$T(n) = cn^2 + \frac{3}{16} cn^2 + \left(\frac{3}{16}\right)^2 cn^2 + \dots + \left(\frac{3}{16}\right)^{\log_4 n - 1} cn^2 + \Theta(n^{\log_4 3})$$

$$T(n) = \sum_{i=0}^{\log_4 n - 1} \left(\frac{3}{16}\right)^i cn^2 + \Theta(n^{\log_4 3})$$

Approximate equation ①,

$$T(n) < \sum_{i=0}^{\infty} \left(\frac{3}{16}\right)^i cn^2 + \Theta(n^{\log_4 3})$$

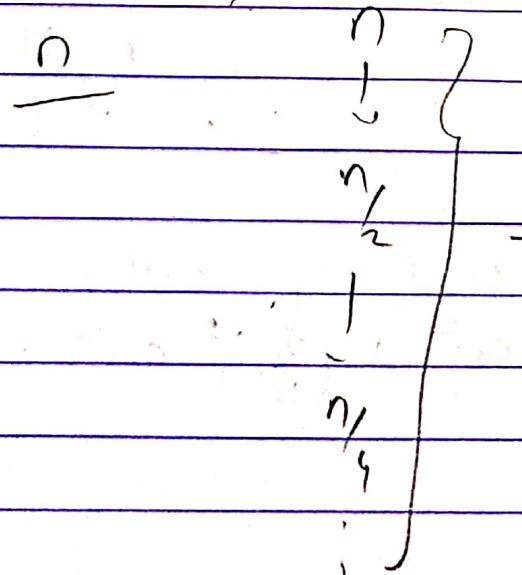
$$= \frac{1}{1 - \left(\frac{3}{16}\right)} cn^2 + \Theta(n^{\log_4 3})$$

$$\sum_{k=0}^{\infty} x^k = \frac{1}{1-x}$$

$$T(n) = \frac{16}{13} cn^2 + \Theta(n^{\log_4 3})$$

first term has higher growth in comparison with second term.

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



$$\frac{n}{2^i} = 1$$

$$n = 2^i$$

$$\log_2 n = \log_2 2^i$$

Problem on case 2

Ex.

$$T(n) = 4T\left(\frac{n}{2}\right) + \theta(n)$$

$$\therefore T(n) = 4T\left(\frac{n}{2}\right) + cn$$

Step 1: Recursion tree

Recursive call	No. of nodes	Tree	Row sum
$T(n)$	1	$Cn$	$Cn$
$T\left(\frac{n}{2}\right)$	4	$\frac{Cn}{2}$ $\frac{Cn}{2}$ $\frac{Cn}{2}$ $\frac{Cn}{2}$	$2cn$
$T\left(\frac{n}{4}\right)$ $= T\left(\frac{n}{2^2}\right)$	$4^2$	$\frac{Cn}{2^2} \frac{Cn}{2^2} \frac{Cn}{2^2} \frac{Cn}{2^2} \frac{Cn}{2^2} \frac{Cn}{2^2} \frac{Cn}{2^2} \frac{Cn}{2^2}$	$2^2 cn$
$T\left(\frac{n}{8}\right)$	$4^3$	$\frac{Cn}{2^3} \dots$	$2^3 cn$

Step 2: Cost of each level =  $2^i cn$

Step 3: Depth of tree i.e last level of tree,

$$\frac{n}{2^i} = 1$$

$$\log_2 n = \log_2 2^i$$

$$\boxed{T = \log_2 n}$$

Step 4: No. of leaf nodes i.e. No. of nodes at last level =  $4^{\log_2 n}$

$$\begin{aligned} &= n^{\log_2 4} \\ &= n^{\log_2 2^2} \\ &= n^2 \end{aligned}$$

Step 5: Total cost at last level. =  $n^2 \times T(1)$   
 $= \Theta(n^2)$

Step 6: Total cost of entire tree,

$$\begin{aligned} T(n) &= cn + 2cn + 2^2 cn + \dots + 2^{\log_2 n - 1} cn \\ &\quad + \Theta(n^2) \\ &= \sum_{i=0}^{\log_2 n - 1} 2^i cn + \Theta(n^2) \end{aligned}$$

Approximating equation ④ ~

$$< \sum_{i=0}^{\infty} 2^i cn + \Theta(n^2)$$

$$= \frac{1}{1-2} cn + \Theta(n^2)$$

$$\left[ \because \sum_{i=0}^{\infty} x^i = \frac{1}{1-x} \right]$$

$$= cn + \Theta(n^2)$$

Second term has higher growth in comparison to first

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

OR

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Step 6: Total cost of entire tree,

$$T(n) = cn + 2cn + 2^2 cn + \dots + 2^{\log_2 n} cn$$

$\approx 2^0 + 2^1$

$$= \sum_{i=0}^{\log_2 n} 2^i cn \quad \approx cn \sum_{i=0}^{\log_2 n} 2^i$$

$$= cn \left[ \frac{2^{\log_2 n} - 1}{2 - 1} \right]$$

[using Geometric progression]  
 $a=1 \quad r=2$

$$= cn \left[ n^{\log_2 2} - 1 \right]$$

$$\left[ S_n = \frac{a(r^n - 1)}{r - 1} \right]$$

$$= cn[n - 1]$$

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

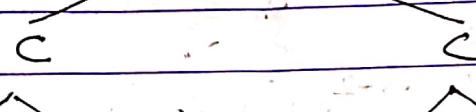
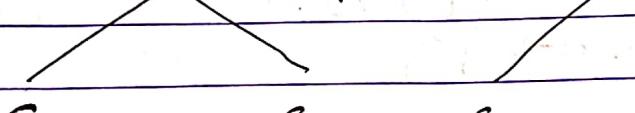
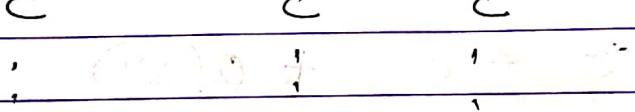
First term has higher growth

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

$$\text{Ex 3 } T(n) = 2T(n-1) + 1$$

$$\therefore T(n) = 2T(n-1) + C$$

Step 1 : Recursion Tree

Recursive Call	No. of nodes	Tree	Row Sum
$T(n)$	1		$C$
$T(n-1)$	2		$2C$
$T(n-2)$	$2^2$		$2^2 C$
$T(n-i)$	$2^i$		$2^i C$

Step 2 : Cost at each level =  $2^i C$

Step 3 : Depth of tree i.e last level of tree

$$\begin{aligned} n-i &= 1 \\ \boxed{i} &= n-1 \end{aligned}$$

Step 4: No. of leaf nodes =  $2^{\frac{n}{2}}$   
 $= 2^{n-1}$

Step 5: Total cost at last level =  $2^{n-1} \times T(1) = \Theta(2^{n-1})$   
 $= \Theta(2^n)$

Step 6: Total cost of entire tree,

$$T(n) = c + 2c + 2^2c + 2^3c + \dots + 2^{n-2}c + \Theta(2^n)$$

$$= c \sum_{i=0}^{n-2} 2^i + \Theta(2^n)$$

[Using Geometric Progression]

$$= c \left( \frac{2^{n-2} - 1}{2 - 1} \right)$$

$$\left\{ S_n = \frac{a(r^n - 1)}{r - 1} \right.$$

$$= c(2^{n-2} - 1)$$

$$T(n) = 2^{n-2}c - c + \Theta(2^n)$$

and third term have same order  
 first term & last higher than second term.

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

OR

$$\leq c \sum_{i=0}^{\infty} 2^i + \Theta(2^n)$$

$$\leq c \left[ \frac{1}{1-2} \right] + \Theta(2^n)$$

$$\leq -c + \Theta(2^n)$$

Problem on case 3

Ex:  $T(n) = 2T(n/2) + \Theta(n)$

$\therefore T(n) = 2T\left(\frac{n}{2}\right) + cn$

Step 1: Recursion Tree.

Recursive call	No. of nodes	Tree	Row sum
$T(n)$	1	$cn$	$cn$
$T\left(\frac{n}{2}\right)$	2	$\frac{cn}{2}$	$cn$
$T\left(\frac{n}{4}\right)$	$2^2$	$\frac{cn}{2^2}$	$cn$
$T\left(\frac{n}{2^i}\right)$	$2^i$	$\frac{cn}{2^i}$	$cn$

Step 2: Cost at each level =  $cn$

Step 3 : Depth of tree i.e. last level of tree,

$$\frac{n}{2^l} = 1$$

$$l = \log_2 n$$

Step 4 : No. of leaf nodes =  $2^l$

$$= 2^{\log_2 n}$$

$$= n^{\log_2 2}$$

$$= n.$$

Step 5 : Total cost at last level =  $n * T(1)$   
=  $\Theta(n) = cn$

Step 6 : Total cost of entire tree ,

$$T(n) = 1cn + 2cn + 3cn + \dots + cn + \Theta(n)$$

$$= \sum_{i=1}^{\log_2 n} cn + \Theta(n)$$

$$= cn \sum_{i=0}^{\log_2 n - 1} 1 + \Theta(n).$$

$$= cn (\log_2 n) + \Theta(n)$$

$$= cn \log_2 n + \Theta(n)$$

First term is higher than second term ,

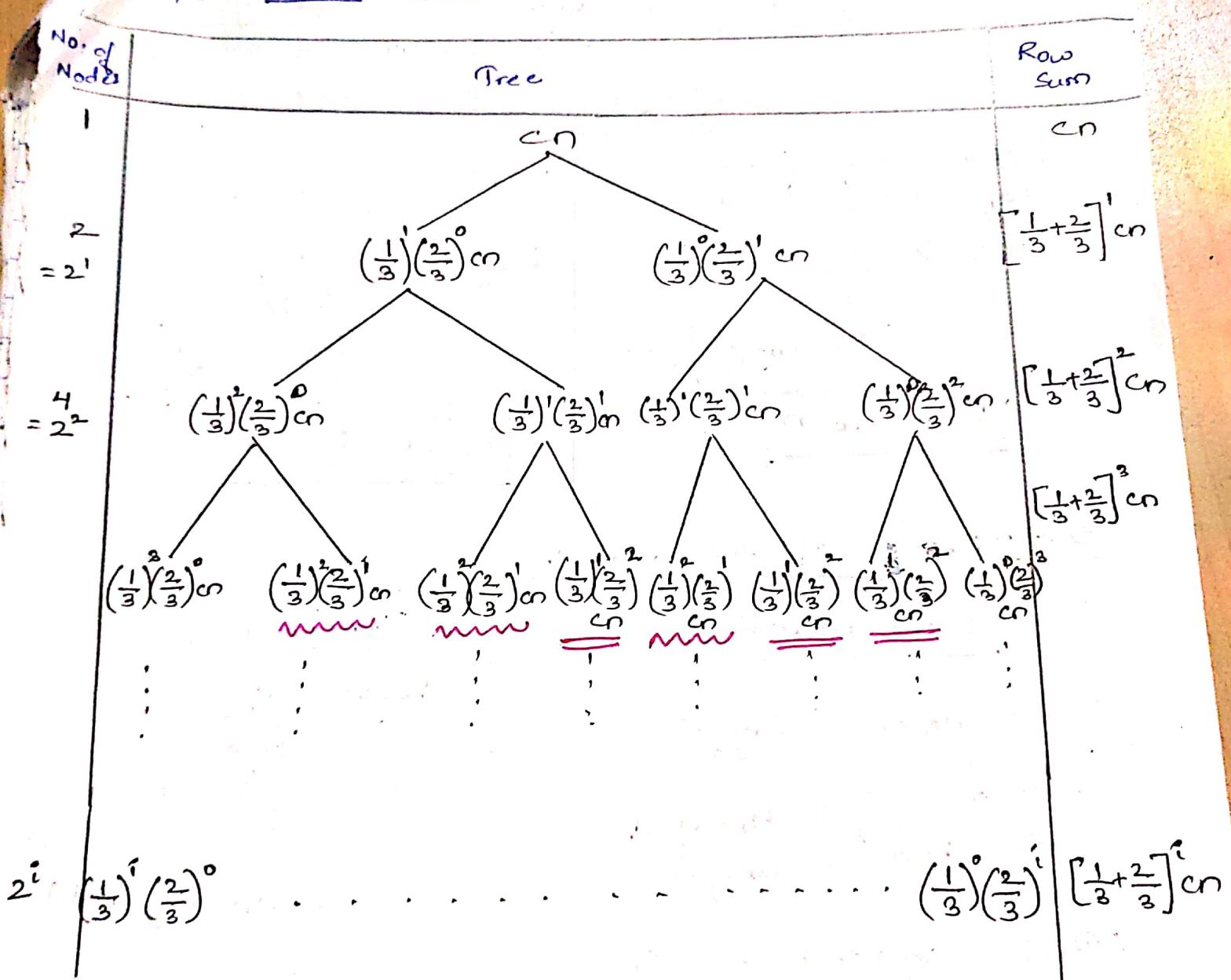
$$\therefore T(n) = \Theta(n \log_2 n)$$

$$\therefore Tcn = \pi\left(\frac{n}{3}\right) + \pi\left(\frac{2n}{3}\right) + cn$$

(Not so  
say Nature)

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Step 1 : Recursion Tree



Step 2 : Cost at each level = cn

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### Step 3 : Depth of the tree

$\therefore \frac{2n}{3} > \frac{n}{3}$  i.e Right subtree is doing more work than the left subtree.

$\therefore$  leaf nodes will not appear at the same level.

leftmost leaf of the left subtree is at.

$$h_{left} \Rightarrow \frac{n}{3^i} = 1$$

$$i = \log_3 n$$

$$h_{left} = \log_3 n$$

Rightmost leaf of the right subtree is at

$$h_{right} \Rightarrow \frac{n}{(\frac{3}{2})^i} = 1$$

$$i = \log_{\frac{3}{2}} n$$

$$h_{right} = \log_{\frac{3}{2}} n$$

### Step 4 : Cost of the entire tree

#### Lower bound

$$h_{left} = \log_3 n$$

Consider a full binary tree with  $h_{left}$  as height.

$$\therefore T(n) = cn + cn + \dots + cn$$

$$= \sum_{i=0}^{\log_3 n} cn$$

$$= cn (\log_3 n + 1)$$

$$= cn \log_3 n + cn$$

$$\therefore T(n) = \Omega(n \log n)$$

#### Upper bound

$$h_{right} = \log_{\frac{3}{2}} n$$

$$\therefore T(n) = \sum_{i=0}^{\log_{\frac{3}{2}} n} cn$$

$$= cn (\log_{\frac{3}{2}} n + 1)$$

$$= cn \log_{\frac{3}{2}} n + cn$$

$$\therefore T(n) = \Theta(n \log_{\frac{3}{2}} n)$$

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

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For larger values of  $n$ , (Not so significant)

$$n-1 \approx n$$

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

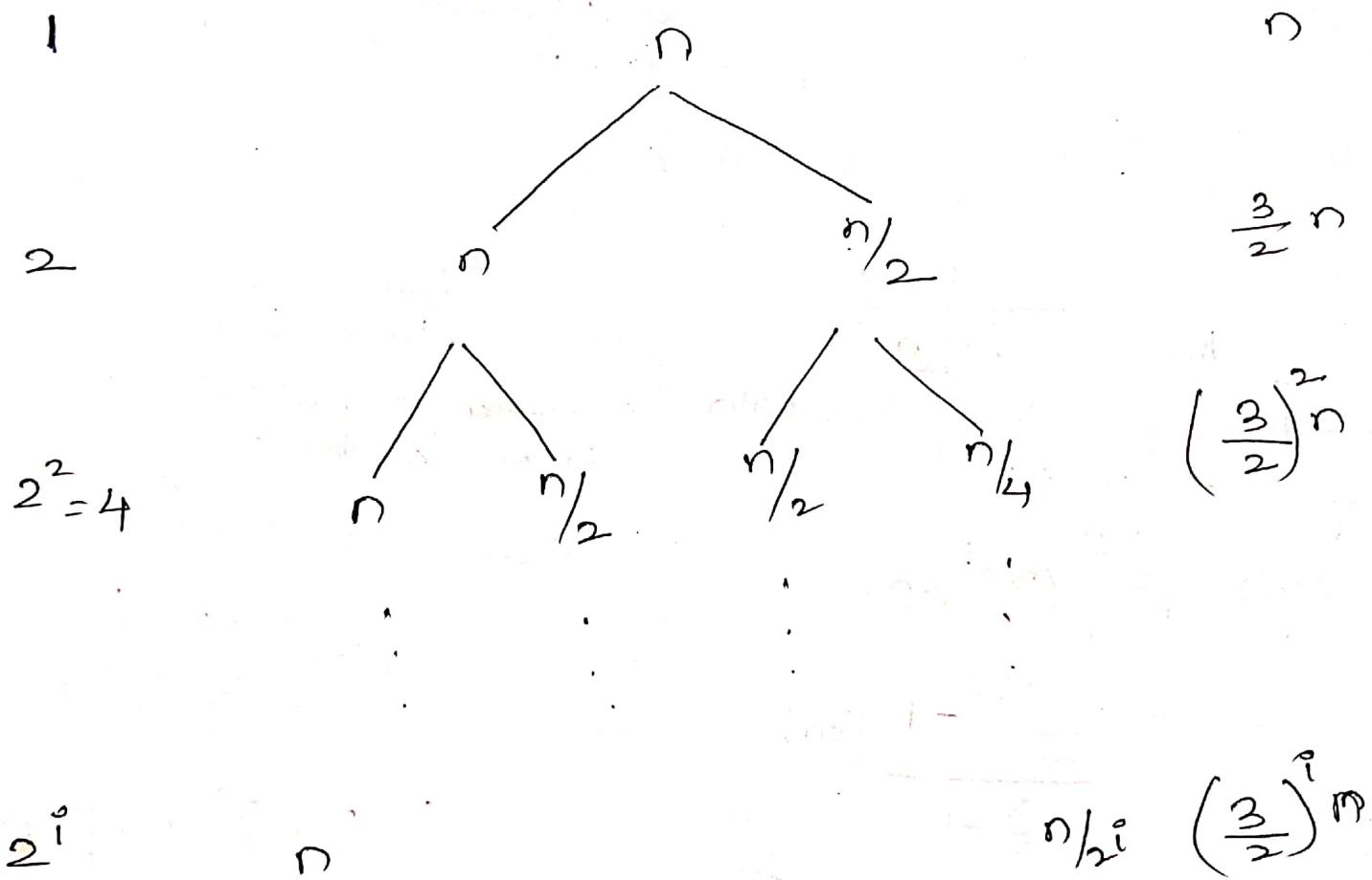
Step 1:

Rewrsion      Tree

No. of  
Nodes.

Tree

Row  
Sum



Step 2: Cost at each level =  $\left(\frac{3}{2}\right)^k cn$

### Step 3: Depth of the tree

$\therefore n > n_{1/2}$  i.e. left subtree is doing more work than right subtree.

$\therefore$  leaf nodes will appear at the same level.

Leftmost leaf of the left subtree is at,

$$h_{left} \Rightarrow n = 2^i$$

$$\boxed{2^i = n}$$

$$\therefore h_{left} = i$$

Rightmost leaf of the right subtree is at,

$$h_{right} \Rightarrow \frac{n}{2^i} = 1$$

$$\therefore n = 2^i$$

$$\boxed{2^i = \log_2 n}$$

$$\therefore \boxed{h_{right} = \log_2 n}$$

### Step 4: Cost of the entire tree

#### Lower Bound

$$h_{right} = \log_2 n$$

Consider a full binary tree with height as height

$$\therefore T(n) = \sum_{i=0}^{\log_2 n} \left(\frac{3}{2}\right)^i cn$$

$$\begin{aligned} & \left\langle \sum_{i=0}^{\log_2 n} \left(\frac{3}{2}\right)^i cn \right\rangle = \frac{\frac{3}{2}^{\log_2 n} - 1}{\frac{3}{2} - 1} [cn] \\ & = \frac{1}{1 - \frac{3}{2}} cn \\ & \therefore - 2cn \\ & \therefore T(n) = \Omega(n^{1.58}) \end{aligned}$$

#### Upper Bound

$$h_{left} = n$$

Consider a full BT with height as height

$$\therefore T(n) = \sum_{i=0}^n \left(\frac{3}{2}\right)^i cn$$

$$= cn \left[ \frac{\frac{3}{2}^n - 1}{\frac{3}{2} - 1} \right]$$

$$= 2 \left[ \frac{\frac{3}{2}^n - 1}{\frac{3}{2} - 1} \right] cn$$

$$= [3^n - 2] cn$$

$$= 3^n cn - 2cn$$

$$\therefore T(n) = O(n 3^n)$$

# THE MASTER METHOD

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- The master method is used for solving recurrences of the form :

$$T(n) = aT\left(\frac{n}{b}\right) + f(n) \quad \text{--- (I)}$$

where  $a \geq 1$  and  $b > 1$  are constants and  $f(n)$  is an asymptotically positive function.

- The recurrence in equation (I) describes the running time of an algorithm that divides a problem of size  $n$  into  $a$  subproblems, each of size  $\frac{n}{b}$ , where  $a$  and  $b$  are positive constants.
- function  $f(n)$  includes cost of dividing the problem and combining the results of subproblem.

## MASTER THEOREM

Let  $a \geq 1$  and  $b > 1$  be constants, let  $f(n)$  be a function, and let  $T(n)$  be defined a recurrence defined as,

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

then  $T(n)$  has the following asymptotic bounds

case 1:

If  $f(n) = O(n^{\log_b a - \epsilon})$  for some constant  $\epsilon > 0$   
 i.e.  $f(n) < n^{\log_b a}$

then,

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

case 2:

If  $f(n) = \Theta(n^{\log_b a})$   
 if  $f(n) = n^{\log_b a}$

then,

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

case 3:

i) If  $f(n) = \Omega(n^{\log_b a + \epsilon})$  for some constant  $\epsilon > 0$ ,  
 i.e.  $f(n) > n^{\log_b a}$

and

ii) if  $a f(n/b) \leq c f(n)$  for some constant  $c < 1$

then,

$$T(n) = \Theta(f(n))$$

EXAMPLES

$$1. T(n) = 9T\left(\frac{n}{3}\right) + n$$

Step 1 : Comparing with standard form

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

$$\therefore a = 9 \quad b = 3 \quad f(n) = n$$

Step 2 : calculate  $n^{\log_b a}$

$$n^{\log_b a} = n^{\log_3 9} = n^2$$

Step 3 : Identifying the Case

$$f(n) = n = O(n^{\log_b a}) = O(n^2)$$

$$\therefore [f(n) < n^{\log_b a}]$$

$$\text{i.e. } f(n) = O(n^{2-\epsilon}) \text{ where } \epsilon = 1$$

By Case 1 of master theorem,

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

$$[T(n) = O(n^2)]$$

$$g. \quad T(n) = T\left(\frac{2n}{3}\right) + 1$$

Step 1: Comparing with standard form

i.e.  $T(n) = aT\left(\frac{n}{b}\right) + f(n)$

$$a = 1$$

$$b = \frac{3}{2}$$

$$f(n) = 1$$

Step 2: Calculate  $n^{\log_b a}$

$$n^{\log_b a} = n^{\log_{\frac{3}{2}} 1} = n^0$$

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

Step 3: Identifying the case

$$f(n) = 1 \quad n^{\log_b a} = 1$$

$$\therefore f(n) = n^{\log_b a}$$

$$\text{i.e. } f(n) = \Theta(n^{\log_b a})$$

By case 2 of master theorem,

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

$$\boxed{T(n) = \Theta(\log n)}$$

$$3. \quad T(n) = 3T\left(\frac{n}{4}\right) + n \log n$$

Step 1: Comparing with standard form

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

$$\therefore a = 3 \quad b = 4 \quad f(n) = n \log n$$

Step 2: calculate  $n^{\log_b a}$

$$n^{\log_b a} = n^{\log_4 3} = n^{0.393}$$

Step 3: Identifying the case

$$(i) \quad f(n) = n \log n \Rightarrow cn^{\log_b a} = n^{0.393}$$

$$\therefore [f(n)]^a > c n^{\log_b a}$$

$$\text{i.e } f(n) = \Omega(n) \\ = \Omega(n^{0.393+\epsilon}) \text{ where } \epsilon \approx 0.2$$

$$(ii) \quad af\left(\frac{n}{b}\right) = 3 \left[ \frac{n}{4} \log \frac{n}{4} \right]$$

$$= \frac{3}{4} n \log n - \frac{3}{4} \log 4$$

$$af\left(\frac{n}{b}\right) \leq cn \log n \quad \text{where } c = \frac{3}{4}$$

By case 3 of Master theorem,

$$\therefore T(n) = \Theta(f(n)) = \Theta(n \log n)$$

Teacher's Sign.: \_\_\_\_\_

$$4. T(n) = 4T\left(\frac{n}{2}\right) + n$$

Step 1: Comparing with standard form

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

$$\therefore a = 4 \quad b = 2 \quad f(n) = n$$

Step 2: Calculate  $n^{\log_b a}$

$$n^{\log_b a} = n^{\log_2 4} = n^2$$

Step 3: Identifying the case

$$f(n) = n \quad n^{\log_b a} = n^2$$

$$\therefore \boxed{f(n) < n^{\log_b a}}$$

$$\text{i.e. } f(n) = O(n^{2-\epsilon}) \text{ where } \epsilon = 1$$

By case 1 of master theorem,

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

$$\boxed{T(n) = O(n^2)}$$

5.

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

Step 1: Comparing with standard form

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

$$\therefore a = 2 \quad b = 2 \quad f(n) = n \log n$$

Step 2: calculate  $n^{\log_b a}$

$$n^{\log_b a} = n^{\log_2 2} = n$$

Step 3: Identifying the case

$$f(n) = n \log n \quad n^{\log_b a} = n$$

$\therefore f(n)$  is asymptotically larger than  $n^{\log_b a}$ ,  
it is not polynomially larger.

$\therefore$  The ratio  $f(n)/n^{\log_b a} = \log n$  and  $\log n$  is asymptotically less than  $n^\epsilon$  for any positive constant  $\epsilon$ .

$\therefore$  Hence, Not solvable using master method because recurrence falls in the gap between case 2 & case 3.

6.  $T(n) = 2T(n/2) + \Theta(n)$   
 - case 2  $f(n) = n$   $n^{\log_b a} = n$   
 $\therefore T(n) = \Theta(n \log n)$

7.  $T(n) = 8T(n/2) + \Theta(n^2)$   
 $f(n) = n^2$   $n^{\log_b a} = n^{\log_2 8} = n^3$   
 - case 1  
 $\therefore T(n) = \Theta(n^3)$

8.  $T(n) = 3T(n/2) + \Theta(n^2)$   
 $f(n) = n^2$   $n^{\log_b a} = n^{\log_2 3} = n^{2.80}$   
 - case 1  
 $\therefore T(n) = \Theta(n^{\log_2 3})$

9.  $T(n) = 2T(n/4) + 1$   
 $f(n) = 1$   $n^{\log_b a} = n^{\log_4 2} = n^{0.5} = n^{1/2}$   
 - case 1  
 $\therefore T(n) = \Theta(n^{\log_4 2}) = \Theta(n^{0.5}) = \Theta(\sqrt{n})$

10.  $T(n) = 2T(n/4) + \sqrt{n}$   
 $f(n) = \sqrt{n}$   $n^{\log_b a} = \sqrt{n} = n^{0.5}$   
 - case 2  
 $\therefore T(n) = \Theta(\sqrt{n} \log n)$

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11.  $T(n) = 2T(n/4) + n$

$f(n) = n$        $n^{\log_2 2} = n^{\log_4 4} = \sqrt{n}$   
 D - case 3

$$T(n) = \Theta(n)$$

12.  $T(n) = 2T(n/4) + n^2$

$f(n) = n^2$        $n^{\log_2 2} = \cancel{n} \sqrt{n}$   
 $f(n) > n^{\log_2 2}$

$a^b f(n/b) = 2 \left[ \frac{n^2}{16} \right] = \frac{2}{16} n^2 \leq c n^2$   
 $c = \frac{3}{16} < 1$

case 3

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

$$12. T(n) = 4T(n/2) + n^2 \log n$$

Step 1: Comparing with standard form

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

$$a = 4 \quad b = 2 \quad f(n) = n^2 \log n$$

Step 2: calculate  $n^{\log_b a}$

$$n^{\log_b a} = n^{\log_2 4} = n^2$$

Step 3: Identifying the case

$$f(n) = n^2 \log n \quad n^{\log_b a} = n^2$$

$\therefore f(n)$  is asymptotically larger than  $n^{\log_b a}$ ,  
it is not polynomially larger

$\therefore$  The ratio  $f(n)/n^{\log_b a} = \log n$  and  $\log n$

is asymptotically less than  $n^\epsilon$  for any positive constant  $\epsilon$ .

$\therefore$  Hence, Not Solvable using master method

because recurrence falls in the gap between case 2 & case 3.

# THE MASTER METHOD

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OR

(FOR DECREASING FUNCTION)

## MASTER THEOREM

Let  $a > 0$ ,  $b > 0$  and  $k \geq 0$  be constants and  $f(n)$  be a function such that  $f(n) = O(n^k)$

Let  $T(n)$  be a recurrence defined as,

$$T(n) = aT(n-b) + f(n) \quad n > 1$$

$$= c \cdot r^{n/b} + f(n) \quad n \leq 1$$

where  $c$  is constant.

then  $T(n)$  has following asymptotic bound:

Case 1:

If  $a < 1$  then

$$T(n) = O(n^k) \text{ i.e. } O(f(n))$$

Case 2:

If  $a = 1$  then

$$T(n) = O(n^{k+1}) \text{ i.e. } O(n + f(n))$$

Case 3:

If  $a > 1$  then

$$T(n) = O(n^k a^{n/b}) \text{ i.e. } O(f(n) \cdot a^{n/b})$$

$$\text{Ex. } ① \quad T(n) = 3T(n-2) + n$$

Step 1: comparing with standard form  
 $T(n) = aT(n-b) + f(n)$

$$\therefore a = 3 \quad b = 2 \quad f(n) = n \quad k = 1$$

Step 2: Identifying case.

$$\therefore \cancel{a > 1} \quad a = 3 > 1$$

By case 3 of master theorem,

$$T(n) = O(f(n) a^{n/b}) \text{ or } O(n^k a^{n/b})$$

$$\boxed{T(n) = O(n \times 3^{n/2})}$$

$$\textcircled{2} \quad T(n) = T(n-1) + n$$

Step 1: comparing with standard form  
 $T(n) = aT(n-b) + f(n)$

$$\therefore a = 1 \quad b = 1 \quad f(n) = n \quad k = 1$$

Step 2: Identifying case.

$$\therefore a = 1$$

By case 2 of master theorem,

$$\begin{aligned} T(n) &= O(n \times f(n)) \text{ or } O(n \times n^k) \\ &= O(n^2) \end{aligned}$$

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(3)

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

Step 1: Comparing with standard form  
 $T(n) \leq aT(n-b) + f(n)$

$$\therefore a = 2 \quad b = 2 \quad f(n) = n^2 \quad k = 2$$

Step 2: Identifying case.  
 $\therefore a = 2 > 1$

By case 3 of master theorem,

$$T(n) = O(f(n) a^{n/b}) \quad \text{OR} \quad O(n^k a^{n/b})$$

$$= O(n^2 2^{n/2})$$