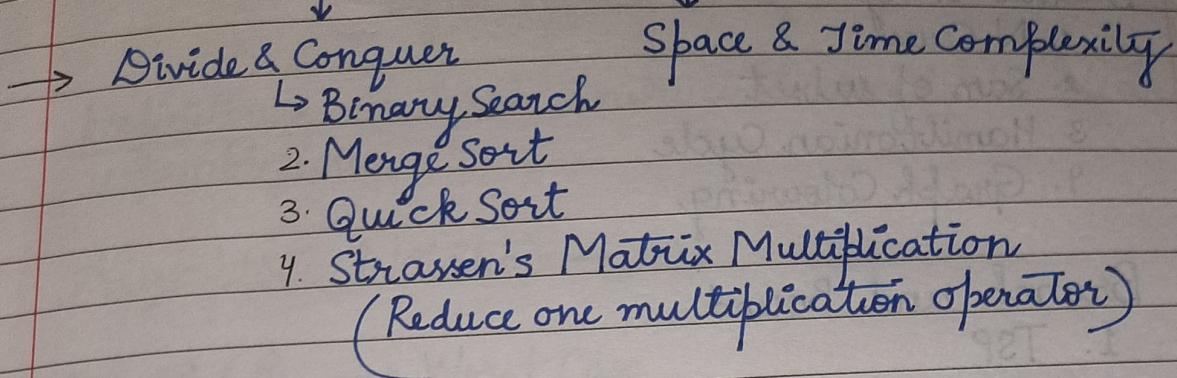


# Design & Analysis of Algorithms



## → Greedy Method

1. Task Scheduling problem
2. Travelling Sales Person problem
3. Knap Sack
4. Minimum Spanning Tree [Kruskal, Prim]  
UnDirected Graph.
5. Single Source Shortest Path  
[Dijkstra] → Directed Graph
6. Bellman Ford [-ve weights given]

## → Dynamic Programming. [Bottom Up Approach]

Subproblems dependent on each other

★★

1. Longest Common Subsequence LCS

2. Matrix Chain Multiplication MCM

3. 0/1 Knapsack

4. Travelling Sales Person problem

5. All Pair Shortest Path → (Floyd Warshall)

→ Backtracking

1. N Queen
2. Sum of subset
3. Hamiltonian Cycle
4. Graph Colouring

→ Branch and Bound

1. TSP
2. 0/1 Knapsack

UNIT 5

→ String Matching

1. KMP
2. Rabin Karp
3. Boyee Moree

What is an algorithm?

Algorithm is a finite set of instructions that if followed accomplish a particular task.

Various features of an algorithm

1. Input → At least one or more quantity is externally supplied.
2. Output → At least one quantity should produce
3. Effectiveness → Each instruction should be so basic that it can be carried by any person with help of pen & paper

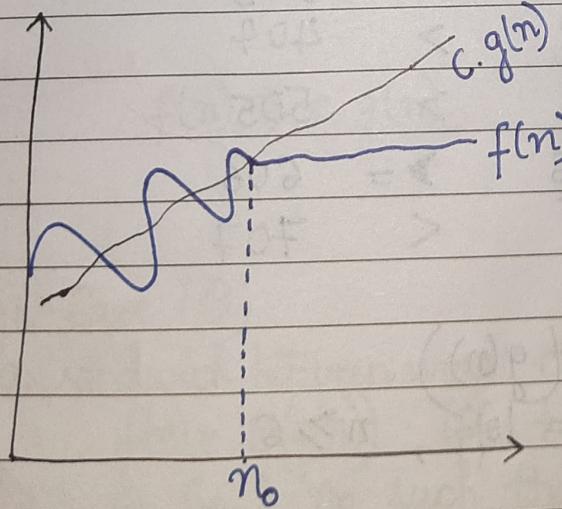
4. Finiteness → For all cases algorithm should terminate after finite no of steps.
5. Definiteness → Each instruction should be clear and unambiguous.
- Asymptotic Notation →

1. Big Oh ( $O$ ) → Worst Case
2. Omega ( $\Omega$ ) → Best Case
3. Theta ( $\Theta$ ) → Average Case

\* Big Oh ( $O$ ) → indicates max<sup>m</sup> amount of time required to solve any problem. It describes the worst case of algorithm.

The function  $f(n) = O(g(n))$  if & only if there exists +ve constants  $c$  and  $n_0$  such that

$$f(n) \leq c g(n) \text{ for all } n, n \geq n_0$$



$$f(n) = 3n+2 \quad g(n) = n \quad \text{Prove that } f(n) = O(g(n))$$

$$f(n) \leq c g(n)$$

$$3n+2 \leq c \cdot n, n \geq n_0$$

$n$	$f(n) = 3n+2$	$g(n) = n$	$c \cdot n$	$c=4$
1	5	>	$4n = 4$	
2	8	=	8	
3	11	<	12	

$$f(n) = O(g(n))$$

for  $c=4$ ,  $n \geq 2$

$$f(n) = 100n + 6$$

$$g(n) = n$$

$$\text{Prove } f(n) = O(g(n))$$

$$f(n) \leq g(n)$$

$$100n + 6 \leq n$$

$n$	$f(n)$	$c \cdot n$	$101n$	$c=101$
1	106	>	101	
2	206	>	202	
3	306	>	303	
4	406	>	404	
5	506	>	505	
6	606	=	606	
7	706	<	707	

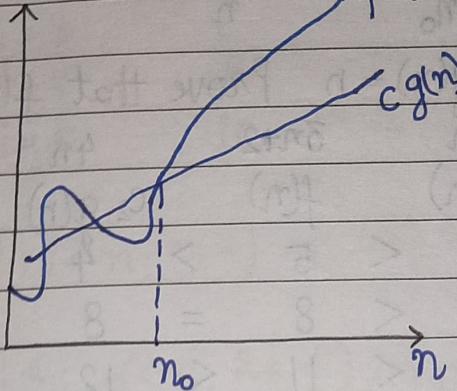
$$f(n) = O(g(n))$$

for  $c=101$ ,  $n \geq 6$

\* Omega Notation ( $\Omega$ ).  $\rightarrow$  defines best case of an algorithm

The function  $f(n) = \Omega g(n)$  iff there exists two +ve constants such  $c$  and  $n_0$  such that

$$\boxed{f(n) \geq c g(n)} \quad \text{for all } n, n \geq n_0$$



$$f(n) = 3n+2 \quad g(n) = n \quad \text{Prove that } f(n) = \Omega g(n).$$

$$f(n) \geq c \cdot g(n)$$

$n$	$3n+2$	$3n$	Let $c=3$
1	5	> 3	
2	8	> 6	

$$f(n) \geq c \cdot g(n) \quad f(n) = \Omega g(n)$$

\* Average Case ( $\Theta$ )

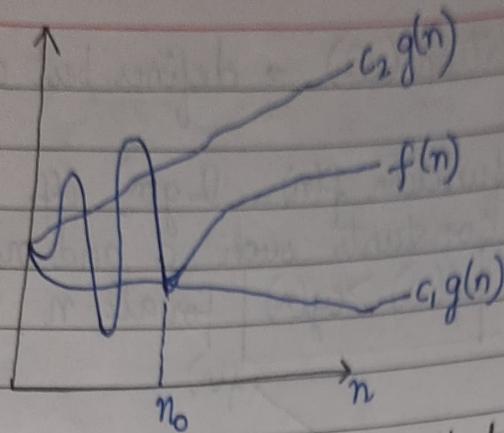
It is a sandwich between worst case & best case

$f(n) = \Theta(g(n))$  iff there exist +ve constants  $c_1, c_2$  &  $n_0$  such that

$$\boxed{c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)}$$

Best Case

Worst Case.



$f(n) = 3n+2 \quad g(n) = n$  Prove that  $f(n) = \Theta(g(n))$

$n$	$c_1 g(n)$	$f(n)$	$c_2 g(n)$
1	3	< 5	> 4
2	6	< 8	= 8
3	9	< 11	< 12

$$n_0 = 3 \quad c_1 = 3 \quad c_2 = 4 \quad n \geq 2$$

## Type of Complexity

- |                            |                         |
|----------------------------|-------------------------|
| 1. Constant $O(1)$         | 4. Quadratic $O(n^2)$   |
| 2. Logarithmic $O(\log n)$ | 5. Cubic $O(n^3)$       |
| 3. Linear $O(n)$           | 6. Exponential $O(2^n)$ |

①  $\text{for } (i=0; i<100; i++)$       100 times  
     {  $O(100)$  constant  
     {  $\text{printf}("*)";$   
     }

②  $\text{for } (i=0; i<n; i++)$   
     {  $\text{for } (j=0; j<n; j++)$        $n^2$   
     {  $\text{printf}(" ");$   
     }

$\text{for } (i=0; i<n; i++)$   
     {  $\text{for } (j=0; j<n; j++)$   
     {  $\text{printf}(" ");$   
     }  
 $n+n = 2n \Rightarrow O(n)$

Q for ( $i=1$ ;  $i < n$ ;  $2*i$ )  
 { printf ("")  
 }

$i = 1$	2	$2^i < n$
$i = 2$	4	$\log 2^i < \log n$
$i = 3$	6	$i \log 2 < \log n$
$i = 4$	8	
$i = 5$	10	$i < \log n$

Q for ( $i=1$ ;  $i < n$ ;  $i++$ ) n  
 {  
 for ( $j=1$ ;  $j < n$ ;  $2*j$ )  $\log_2 n$   
 { printf ("")  
 }  
 }  
 } Independent loops  
 $O(n + \log n)$   
 $O(n)$   
 $O(n * \log n)$



## Recurrence.

1. Master's Method
2. Iterative Method
3. Recursion tree
4. Substitution Method.

Generalised form  $T(n) =$

$$\begin{cases} O(1), & \text{if } n \leq 1 \\ a.T\left(\frac{n}{b}\right) + D(n) + C(n), & \text{if } n > 1 \end{cases}$$

Imp

$a \rightarrow$  no of subproblems

$n/b \rightarrow$  size of subproblem.

$D(n) \rightarrow$  Time reqd for dividing any array

$C(n) \rightarrow$  Time reqd for combining soln

$$f(n) = D(n) + C(n)$$

\*

## Binary Search

Right sublist discarded. At a time one

$$T(n) = 1 \cdot T\left(\frac{n}{2}\right) + O(1) + O(1)$$

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

\*

## Linear Search

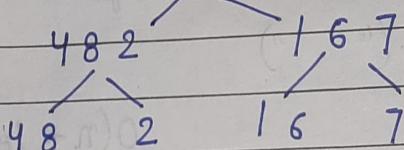
$$T(n) = T(n-1) + O(1) + O(1)$$

↓      ↓  
divide    combine

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

## Merge Sort

4 8 2 1 6 7



$\frac{n}{2}$  → size of  
subproblems

$$T(n) = 2 \cdot T\left(\frac{n}{2}\right) + O(1) + O(n)$$

just cal mid    compare, swap &  
combine

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

## Master's Method

$$T(n) = \Theta(n^{\log_b a}) \text{ if } f(n) = O(n^{\log_b a - \epsilon}) \text{ for } \epsilon > 0$$

2.

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

3.

$$T(n) = \Theta(f(n)) \text{ if } f(n) = O(n^{\log_b a + \epsilon}) \text{ for } \epsilon > 0 \text{ and}$$

if  $a f\left(\frac{n}{b}\right) \leq f(n)$  for  $c < 1$

and for large  $n$

## Master's Method

Case 1:  $T(n) = \Theta(n^{\log_b a})$  if  $f(n) = O(n^{\log_b a - \epsilon})$  for  $\epsilon > 0$

Case 2:  $T(n) = \Theta(n^{\log_b a} \cdot \log n)$  if  $f(n) = O(n^{\log_b a})$

Case 3:  $T(n) = \Theta(f(n))$   $f(n) = O(n^{\log_b a + \epsilon})$  and  $af(\frac{n}{b}) \leq cf(n)$

$c < 1$  and for large value of  $n$

$$[a, b \geq 1]$$

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

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

$$\begin{aligned} n^{\log_b a} &= 9 \\ n^{\log_3 9} &= n^2 \end{aligned}$$

Case 1:  $n^{\log_b a} > f(n)$  Case 2:  $n^{\log_b a} = f(n)$

Case 3:  $n^{\log_b a} < f(n)$

$\therefore n^{\log_b a} > f(n)$  Case 1 holds

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

$$f(n) = O(n^2)$$

$$n = O(n^{2-1}), \epsilon > 0 \quad \epsilon = 1$$

$$T(n) = \underline{\Theta(n^{\log_b a})} = \underline{\Theta(n^2)}$$

Q  $T(n) = T\left(\frac{n}{2}\right) + 1$  // Binary Search

$$T(n) = aT\left(\frac{n}{b}\right) + f(n) \quad a=1 \quad b=2 \quad f(n)=1$$

$$n^{\log_b a} = n^{\log_2 1} = n^0 = 1$$

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

$1 = 1$  Case 2 holds

$$T(n) = \Theta\left(n^{\log_b a} \cdot \log n\right) = \Theta(n^0 \log n) = \underline{\Theta(\log n)}$$

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

$$a=1 \quad b=2 \quad f(n)=n^2 \quad n^{\log_b a} = n^{\log_2 1} = n^0 = 1$$

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

Case 3 holds

$$f(n) = O(n^{0+2}) \quad \epsilon = 2$$

$$n^2 = O(n^{0+2})$$

Regularity condition  $a f\left(\frac{n}{b}\right) \leq c f(n)$

$$a=1 \quad b=2$$

$$f(n)=n^2 \quad 1 \cdot f\left(\frac{n}{2}\right) \leq c f(n)$$

$$\begin{aligned} f(n) &= n^2 \\ f\left(\frac{n}{2}\right) &= \left(\frac{n}{2}\right)^2 \end{aligned}$$

$$\left(\frac{n}{2}\right)^2 \leq c \cdot n^2$$

$$\frac{n^2}{4} \leq c \cdot n^2 \Rightarrow c = \frac{1}{4}$$

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

Merge Sort

Q  $T(n) = 2T\left(\frac{n}{2}\right) + n$   $\rightarrow$   $a=2 \quad b=2 \quad f(n)=n$

$$n^{\log_b a} = n^{\log_2 2} = n = f(n) \quad \text{Case 2 holds}$$

$$\begin{aligned} T(n) &= \Theta(n^{\log_b a} \cdot \log n) \\ &= \Theta(n \log n) \end{aligned}$$

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

$$a=2 \quad b=4 \quad f(n)=\sqrt{n}$$

$$n^{\log_b a} \Rightarrow n^{1/2}$$

Case 2 holds

$$\Theta(n^{1/2} \log n)$$

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

$$a=1 \quad b=\frac{3}{2} \quad f(n)=1$$

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

$$T(n) = T(\sqrt{n}) + 1$$

$$a=1 \quad b=1$$

$$= T(n^{1/2}) + 1 \quad \text{--- } ①$$

By changing of variable method  
Suppose  $m = \log_2 n$

$$n = 2^m$$

Substitute value of  $n$  in eq<sup>n</sup> ①

$$T(2^m) = T(2^{m/2}) + 1$$

$$S(m) = S\left(\frac{m}{2}\right) + 1$$

$$m^{\log_b a} = m^{\log_2 1} = m^0 = 1 = f(n) \quad \text{Case 2 holds}$$

$$S(m) = \Theta(\log m)$$

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

$$\begin{aligned}
 T(n) &= 2T(\sqrt{n}) + \log n \\
 &= 2T(n^{1/2}) + \log n \quad \text{--- ①}
 \end{aligned}$$

By changing of variable

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

Substitute in ①

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

$$S(m) = 2T\left(\frac{m}{2}\right) + m.$$

$$m^{\log_b a} = m^{\log_2 2} = m^1 = m \quad f(m) = m$$

Case 2 holds

$$\begin{aligned}
 &\Theta(m^{\log_b a} \cdot \log m) \\
 &= \Theta(m \cdot \log(\log_2 n))
 \end{aligned}$$

$$\Theta(\log_2(\log(\log_2 n)))$$

Master's Theorem  
 $f(n)$  is the cost of dividing solving a problem

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

$$a = 3, b = 4 \\ n^{\log_b a} = n^{\log_4 3} = n^{0.793}$$

$$f(n) = n \log n \quad \text{Case 3 holds}$$

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

Regularity Cond<sup>n</sup>  $a f\left(\frac{n}{b}\right) \leq c \cdot f(n)$

$$3 f\left(\frac{n}{4} \log \frac{n}{4}\right) \leq c \cdot \log n$$

$$\left(3 \frac{n}{4} \log \frac{n}{4} - 3 \frac{n}{4} \log_2 4\right) \leq c \log n$$

$$\left(\frac{3}{4} n \log n - \frac{3}{4} n 2^2\right) \leq c \log n$$

$$c = \frac{3}{4}, c < 1$$

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

$$a = 2, b = 2$$

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

$$f(n) = n \log n$$

$$n < n \log n$$

Case 3 holds

$$af\left(\frac{n}{b}\right) \leq c \cdot f(n)$$

$$2\left(\frac{n \log \frac{n}{2}}{2}\right) \leq c \cdot n \log n.$$

$$n \log n - n \log_2 2 \leq c \cdot n \log n$$

$$n \log n - n \leq c \cdot n \log n$$

$$\boxed{c=1}$$

{ Case 3 does not hold  
as Regularity cond'n fails }

- \* If Polynomial term is left after dividing, then we can apply Master's method

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

$$n^{0.793} < n \log n$$

Bigger term up.

$$\frac{n \log n}{n^{0.793}} = n^{0.293} \log n$$

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

$$n < n \log n$$

$$\frac{n \log n}{n} = \log n$$

- \*  $f(n)$  should grow asymptotically & exponentially polynomially. Only then Case 3.

$$T(n) = 8T\left(\frac{n}{2}\right) + n^2 \log n$$

$$a=8 \quad b=2$$

$$n \log_b a = n \log_2 8 = n^3$$

$$n^3 > n^2 \log n \quad \text{Case 1 holds}$$

$$\frac{n^3}{n^2 \log n} = \frac{n}{\log n}$$

$$T(n) = 7T\left(\frac{n}{2}\right) + n^2 \log n$$

$$T(n) = 6T\left(\frac{n}{2}\right) + n^2 \log n$$

$$T(n) = 5T\left(\frac{n}{2}\right) + n^2 \log n$$

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

$$n^2 = n^2 \log n \quad \frac{n^2 \log n}{n^2} = \log n$$

Case 2. fn grows asymptotically

There is a gap b/w Case 1 & Case 2

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

$$n^{\frac{1}{2}} - \frac{n^2 \log n}{n^{\frac{1}{2}}} = n^{-}$$

Case 3

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

$$T(n) = 1T\left(\frac{n}{2}\right) + n^2 \log n$$

## \* Iterative Method

$$T(n) = 3T\left(\frac{n}{4}\right) + n \quad \text{--- (1)}$$

No of subproblems      size of each subproblem

$$T\left(\frac{n}{4}\right) = 3T\left[\frac{n}{4 \cdot 4}\right] + \frac{n}{4}$$

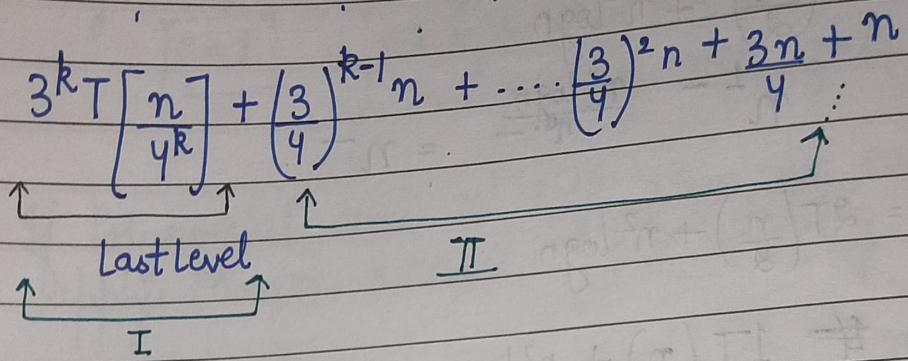
Putting value of  $\frac{n}{4}$  in (1)

$$T(n) = 3 \left[ 3T\left(\frac{n}{16}\right) + \frac{n}{4} \right] + n$$

$$= 3^2 T\left[\frac{n}{16}\right] + \frac{3n}{4} + n$$

$$3^2 \left[ 3T\left[\frac{n}{4^3}\right] + \frac{n}{4^2} \right] + \frac{3n}{4} + n$$

$$3^3 \left[ \frac{n}{4^3} \right] + \left( \frac{3}{4} \right)^2 n + \frac{3n}{4} + n$$



$$I \quad 3^k T\left(\frac{n}{4^k}\right)$$

Running Time  $T(1) = 1$

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

$$T\left(\frac{n}{4^k}\right) = 1 \Rightarrow k = \log_4 n$$

$$3^k(1) \Rightarrow 3^{\log_4 n} \Rightarrow n^{\log_4 3} \Rightarrow n \leftarrow 1 \quad n \leftarrow 1$$

Changing variables  $n^{<1}$

$$II \quad \left(\frac{3}{4}\right)^{k-1} n \dots \dots \left(\frac{3}{4}\right)^2 n + \frac{3n}{4} + n \rightarrow \underline{\underline{GP}}$$

In this avg case is in form of  $\Theta$

But We assume that we have  $\infty$  terms.

$$\left( \frac{3}{4} \right)^0 n + \left( \frac{3}{4} \right)^1 n + \left( \frac{3}{4} \right)^2 n + \dots$$

$$\frac{1}{1-a} = \frac{1}{1-\frac{3}{4}} = 4n$$

$$I + II \Rightarrow n^{<1} + 4n = \underline{\underline{O(n)}} \rightarrow \text{Worst Case}$$

D)  $T(n) = 2T\left(\frac{n}{2}\right) + n \quad \text{--- } ①$

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

Putting the value of  $\frac{n}{2}$  in ①

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

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

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

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

$$= 2^k T\left(\frac{n}{2^k}\right) + nk$$

I

$$2^k T\left(\frac{n}{2^k}\right) \quad T(1) = 1 \quad \frac{n}{2^k} = 1$$

$$k = \log_2 n$$

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

$$n + n \log n \approx \underline{\underline{n \log n}} \quad \text{Merge Sort}$$



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

$O(n^2)$

Worst Case of QuickSort

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

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

Putting the value in ①

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

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

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

:

:

:

$$T(n-k) + \underbrace{(n-(k-1)) + \dots + (n-2) + (n-1)}_{\text{Sum of } n \text{ natural nos}} + n$$

$$\text{Let } n-k=1$$

$$\underline{k=n-1.}$$

↓  
Sum of  $n$  natural

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

$$1 + n^2 + n = O(n^2)$$

Simpl 8

$$T(n) = 8T\left(\frac{n}{2}\right) + n^2 \quad \text{--- ①.}$$

$$T\left(\frac{n}{2}\right) = 8T\left(\frac{n}{2^2}\right) + \left(\frac{n}{2}\right)^2$$

Putting in ①

$$T(n) = 8\left[8T\left(\frac{n}{2^3}\right)\right] + n^2 + \left(\frac{n}{2}\right)^2$$

$$T(n) = 8\left[8T\left(\frac{n}{2^2}\right) + \left(\frac{n}{2}\right)^2\right] + n^2$$

$$T(n) = 8^3 [T$$

$$8^2 T\left[\frac{n}{2^2}\right] + \frac{8n^2}{4} + n^2$$

$$= 8^3 T\left[\frac{n}{2^3}\right] + 4n^2 + 2n^2 + n^2$$

$$= 8^k T\left[\frac{n}{2^k}\right] + (2)^{k-1} n^2 + \dots + 2n^2 + n^2$$

↓  
 I

$$\frac{n}{2^k} = 1 \Rightarrow k = \log_2 n$$

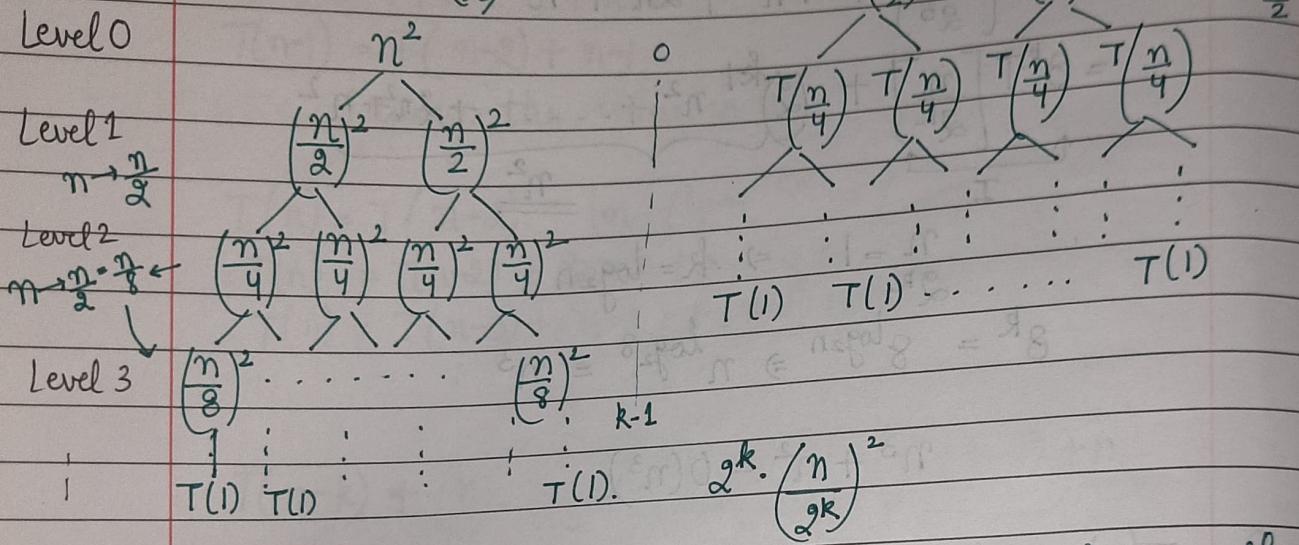
$$8^k = 8^{\log_2 n} \Rightarrow n^{\log_2 8} = n^3$$

$$n^3 + n^2 = O(n^3)$$

$$8 T(n) = 2 T\left(\frac{n}{2}\right) + n \log n.$$

## \* Recursion Tree Method -

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



$$\text{Ht of tree} = \log n$$

$$\text{Level 1 No of nodes} \times \text{Cost} \\ 2 \times \left(\frac{n}{2}\right)^2$$

$$\begin{aligned} \text{Level 0 Cost} &\rightarrow n^2 & 2^0 n^2 \\ \text{Level 1 Cost} &\rightarrow \left(\frac{n}{2}\right)^2 + \left(\frac{n}{2}\right)^2 & 2^1 \cdot \left(\frac{n}{2}\right)^2 \\ &= 2^1 \cdot \frac{n^2}{2^2} = \frac{n^2}{2} \\ \text{Level 2} &\rightarrow 2^2 \times \left(\frac{n}{4}\right)^2 & 2^2 \cdot \frac{n^2}{4^2} = \frac{n^2}{4} \\ &\vdots & \vdots \\ \text{Level k} &\rightarrow 2^k \cdot T(1) & \end{aligned}$$

$$\frac{n}{2^k} = 1 \\ \Rightarrow n = 2^k$$

$$k = \log_2 n$$

Running time at last level = no of nodes at last level \* Computing time at that level

$$2^k \cdot 1 = 2^{\log_2 n} = n$$

$$\left( \frac{n^2 + n^2 + \frac{n^2}{4} + \frac{n^2}{8} + \dots}{2} \right)$$

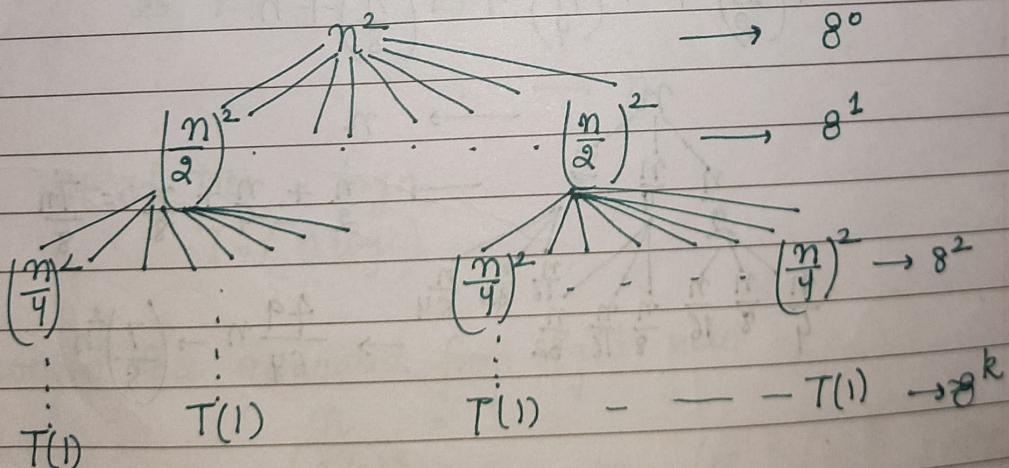
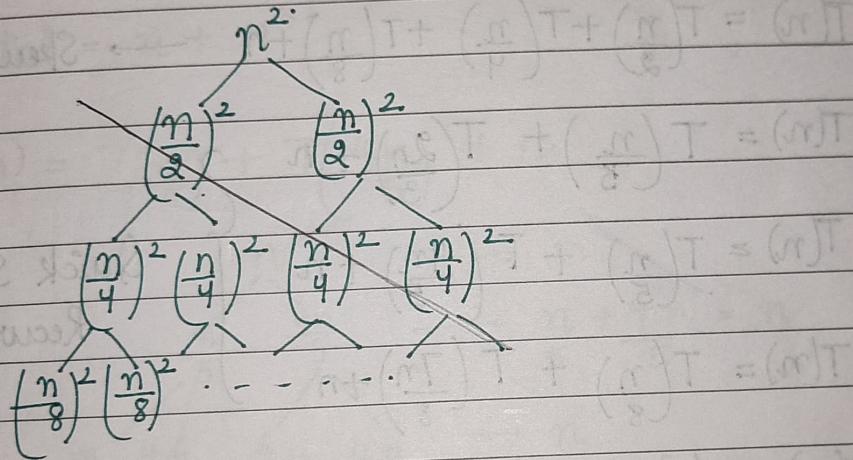
$$n^2 \left( \frac{1}{2^0} + \frac{1}{2^1} + \frac{1}{2^2} + \frac{1}{2^3} + \dots \right)$$

$$= n^2$$

$$\text{Total complexity} = n^2 + n$$

$$= O(n^2)$$

$$T(n) = 8T\left(\frac{n}{2}\right) + n^2$$



$$\text{Level 0 Cost} \rightarrow 8^0 \times n^2$$

$$\text{" 1 " } \rightarrow 8^1 \times \left(\frac{n}{2}\right)^2$$

$$\text{" 2 " } \rightarrow 8^2 \times \left(\frac{n}{4}\right)^2$$

$$8^k \cdot \left(\frac{n}{2^k}\right)^2$$

$$8^0 \cdot \left(\frac{n}{2^0}\right)^2$$

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

$$\left(\frac{n}{2^2}\right)^2$$

$$\text{Let } \frac{n}{2^k} = 1 \Rightarrow n = 2^k \\ k = \log_2 n$$

$$\text{Last Level } 8^k \cdot 1 = 8^{\log_2 n} \Rightarrow n^{\log_2 8} = n^3$$

$$n^2 + 2n^2 + 4n^2 + \dots \\ n^2(1+2+4+\dots) \\ n^2$$

$$\text{Total complexity} = n^2 + n^3 \\ = O(n^3)$$

$$T(n) = T\left(\frac{n}{2}\right) + T\left(\frac{n}{4}\right) + T\left(\frac{n}{8}\right) + n \quad \rightarrow \text{Skewed Tree}$$

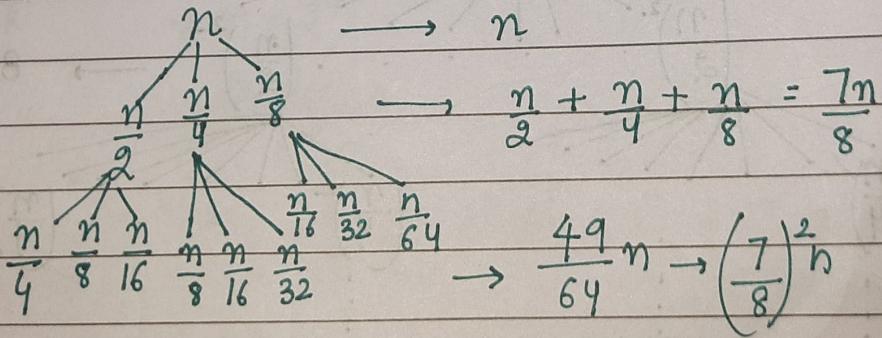
$$T(n) = T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right) + n$$

$$T(n) = T\left(\frac{n}{5}\right) + T\left(\frac{4n}{5}\right) + n$$

$$T(n) = T\left(\frac{n}{8}\right) + T\left(\frac{7n}{8}\right) + n$$

Quick Sort Avg Case  
Recurrence

$$T(n) = T\left(\frac{n}{2}\right) + T\left(\frac{n}{4}\right) + T\left(\frac{n}{8}\right) + n.$$



$$T(1) \rightarrow T(1) \rightarrow \left(\frac{7}{8}\right)^k \cdot 1$$

$$\frac{n}{2^k} = 1 \Rightarrow n = 2^k \Rightarrow k = \log_2 n$$

$$\text{Last level } \left(\frac{7}{8}\right)^k = \left(\frac{7}{8}\right)^{\log_2 n} = n^{\log_2 7/8} \Rightarrow n < 1 \quad n^{<1}$$

Level 0 to  $k-1$

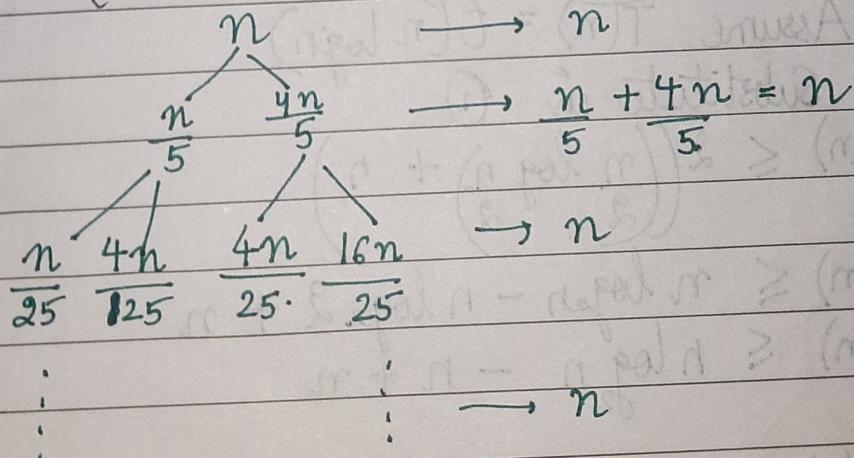
$$n + \frac{7}{8}n + \left(\frac{7}{8}\right)^2 n + \dots$$

$$n \left[ 1 + \frac{7}{8} + \left(\frac{7}{8}\right)^2 + \dots \right] \rightarrow \text{GP}$$

$$= n$$

$$\text{Total} = n + n^{<1}$$

$$\Theta T(n) = T\left(\frac{n}{5}\right) + T\left(\frac{4n}{5}\right) + n$$



$$\text{Ht of tree} = \log n$$

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

Longest Tree generated from right side

$$\frac{n}{2^k} = 1 \Rightarrow n = (5/4)^k \quad k = \log_{5/4} n$$

$$\Theta = \left( \frac{k}{n \log_{5/4} n} \right) \quad \text{Complexity at each level} * \text{no of levels.}$$

~~If f(n) is n, logn, n logn guess  $O(n \log n)$~~

~~If f(n) is cons, 1, c~~  $T(n) = O(n \log n)$   $\rightarrow T(n) = O(\log n)$ .

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

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

### \* Substitution Method

1. Make a guess.
2. Prove by Induction.

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

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

$$T(n) \leq \left(2T\left(\frac{n}{2}\right) + n\right) \quad \text{--- (1)}$$

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

Substitute in (1)

$$T(n) \leq 2\left(\left(\frac{n}{2} \log \frac{n}{2}\right) + n\right)$$

$$T(n) \leq n \log_2 n - n \log_2 2 + n$$

$$T(n) \leq n \log_2 n - n + n$$

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

$T(n) = O(n \log n) \rightarrow$  Hence Proved

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

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

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

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

$$T(n) \leq 2\left(\frac{n \log\left(\frac{n}{2}\right)}{2} + n\right).$$

$$T(n) \leq n \log_2 n - n \log_2 2 + n$$

$$T(n) \leq n \log_2 n - n + n$$

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

$T(n) = O(n \log n)$  Hence Proved.