

• Recurrence

It refers to a way of describing how time complexity of an algo. depends on the size of its input.

Solving Recurrences: Substitution, Iteration, Master methods.

• Substitution Method [Guessing]

Make an educated guess about form of soln. and prove it correct through mathematical induction. Can prove both upper bounds $O()$ and lower bounds $\Omega()$. Examples:-

1 Solve the equation by Substitution Method $T(n) = T(n/2) + 1$

We have to show that it is asymptotically bound by $O(\log n)$

→ For $T(n) = O(\log n)$, we have to show that for some constant c
 $T(n) \leq c \log n$: Put this in given recurrence eqⁿ.

$$T(n) \leq c \log(n/2) + 1$$

$$T(n) \leq c \log(n/2) - c \log_2 2 + 1$$

$$T(n) \leq c \log n \quad \text{for } c \geq 1$$

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

→ $T(n) = T(n/2) + c$ — (1)

$$T(n) = T(n/4) + c + c$$

$$T(n/2) = T(n/4) + c$$
 — (2)

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

$$T(n/4) = T(n/8) + c$$
 — (3)

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

$$n = 2^k \rightarrow \log n = k \log 2 \quad \begin{matrix} \nearrow k = \log n \\ \text{k times} \end{matrix}$$

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

$$\begin{aligned} 1 + kc &: \text{Time complexity} \\ &= 1 + \log n \cdot c = O(\log n) \end{aligned}$$

3 $T(n) = \begin{cases} 1 & \text{if } n = 1 \\ n \cdot T(n-1) & \text{if } n > 1 \end{cases}$

→ $T(n) = n \cdot T(n-1)$ — (1)

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

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

$$= (n-2)T(n-3)$$
 — (3)

$$= (n-1)T(n-2)$$
 — (2)

$$T(n) = n(n-1)T(n-2) \quad \text{: (2) in (1)}$$

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

eliminate this

$$T(n) = n(n-1)(n-2) \dots T(n-3)$$

(n-1) steps

$$= n(n-1)(n-2)(n-3) \dots T(n-(n-1))$$

$$= n(n-1)(n-2)(n-3) \dots T(1)$$

$$= n(n-1)(n-2) \dots 3 \times 2 \times 1$$

$$= n \times n(1-1/n) \times n(1-2/n) \times \dots n(3/n) \times n(2/n) \times n(1/n)$$

$$= O(n^n)$$

$$n \times n \times n \dots = n^n$$

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

$$\rightarrow T(n/2) = 2T(n/4) + n/2 \quad \text{--- (2)}$$

$$T(n/4) = 2T(n/8) + n/4 \quad \text{--- (3)}$$

$$T(n/8) = 2T(n/16) + n/8$$

$$(2) \text{ in } (1) \quad T(n) = 2 \left[2T\left(\frac{n}{4}\right) + \frac{n}{2} \right] + n = 2^2 T\left(\frac{n}{2^2}\right) + n + n$$

$$T(n) = 2^2 [T(n/2^2)] + 2n \quad \text{Substitute (3) in this}$$

$$T(n) = 2^2 [2T(n/8) + n/4] + 2n$$

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

 $\therefore k$ times

$$= 2^k T(n/2^k) + kn$$

 \rightarrow to $T(1)$

$$= 2^k T(1) + n \log n$$

$$= n + n \log n$$

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

dominating

$$5 \quad T(n) = \begin{cases} 1 & \text{if } n=1 \end{cases}$$

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

$$\rightarrow T(n-1) = T(n-2) + \log(n-1) \quad \text{--- (2)}$$

$$T(n-2) = T(n-3) + \log(n-2) \quad \text{--- (3)}$$

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

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

 $\therefore k$ times

$$= T(n-k) + \log(n-(k-1)) + \log(n-(k-2)) + \dots + \log n$$

and this

$$n-k=1 \rightarrow k=n$$

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

$$\begin{aligned} T(n) &= 1 + \log(n-(n-1)) + \log(n-(n-2)) + \dots + \log n \\ &= 1 + \log 1 + \log 2 + \dots + \log n \\ &= 1 + \log(1 \cdot 2 \cdot 3 \cdot 4 \dots n) = 1 + \log(n!) = 1 + \log n^n \\ &= O(n \log n) \end{aligned}$$

$$\begin{aligned} n(n-1)(n-2)(n-3)\dots \\ = n \times n \times n \times n \dots = n^n \end{aligned}$$

→ Cookbook approach

• Master Theorem / Method [solve subproblems, combine their soln.]

Used to analyze time complexity of divide and conquer algorithms (break down problem into subproblems).

$$T(n) = aT(n/b) + f(n) \quad \therefore \text{form.}$$

a: no. of subproblems

n/b: size of each subproblem

From
yt

$$\begin{aligned} a \geq 1, b > 1 \\ \text{Soln: } T(n) &= n^{\log_b a} [u(n)] \\ R(n) &= \frac{f(n)}{n^{\log_b a}} \end{aligned}$$

$R(n)$	$u(n)$
$n^r, r > 0$	$O(n^r)$
$n^r, r < 0$	$O(1)$
$(\log n)^i, i \geq 0$	$\frac{(\log_2 n)^{i+1}}{i+1}$

From
notes

to get these	check these
$\left\{ \begin{aligned} \Theta(n^{\log_b a}) \\ \Theta(n^{\log_b a} \log n) \\ \Theta(f(n)) \end{aligned} \right.$	$\left\{ \begin{aligned} f(n) &= O(n^{\log_b a - \epsilon}) \\ f(n) &= \Theta(n^{\log_b a}) \\ f(n) &= \Omega(n^{\log_b a + \epsilon}) \end{aligned} \right.$
↓ Answers	AND $a f(n/b) < c f(n)$ for large n

1 $T(n) = 9T(n/3) + n$

→ $a=9 \quad b=3 \quad f(n)=n$
 $n^{\log_b a} = n^{\log_3 9} = \Theta(n^2)$

Since $f(n) = O(n^{\log_3 9 - \epsilon})$ where $\epsilon=1$, case 1 applies

$T(n) = \Theta(n^{\log_b a})$ when $f(n) = O(n^{\log_b a - \epsilon})$

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

yt
→

2 $T(n) = 8T(n/2) + n^2$

$a=8 \quad b=2 \quad f(n)=n^2$ $T(n) = n^{\log_b a} u(n) = n^3 u(n)$
 $= n^3 \times 1 = O(n^3)$

3 $T(n) = T(n/2) + c$

→ $a=1$ $b=2$ $f(n)=c$ $T(n) = n^{\log_b a} u(n)$

$T(n) = n^{\log_2 1} u(n) = n^0 u(n) = u(n)$

$R(n) = f(n) = (\log_2 n)^0 \cdot c$

$(\log_2 n)^{0+1} = (\log_2 n) \cdot c = \log_2 n \cdot c = O(\log_2 n)$

4 $T(n) = 8T(n/2) + 1000n^2$ apply master theorem on it.

→ $T(n) = 8T(n/2) + 1000n^2$ $a=8$ $b=2$ $f(n) = 1000n^2$

$a \geq 1$ $b > 1$ $\log_b a = \log_2 8 = 3$

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

$1000n^2 = O(n^{3-\epsilon}) = O(n^2)$

$T(n) = \Theta(n^3)$ as $T(n) = \Theta(n^{\log_b a})$

1st case applied

5 $T(n) = 2T(n/2) + 10n$ solve the recurrence using master method

→ $a=2$ $b=2$ $f(n) = 10n$

$\log_b a = \log_2 2 = 1$

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

$10n = \Theta(n \log^k n) = \Theta(n)$

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

Case 2 applied

6 Solve the recurrence relation:

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

→ Compare with $T(n) = aT(n/b) + f(n)$ with $a \geq 1$ and $b > 1$

$a=2$ $b=2$ $f(n) = n^2$ $\log_b a = \log_2 2 = 1$

Put all the values in $f(n) = \Omega(n^{\log_b a + \epsilon})$

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

$\epsilon = 1$

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

1st condⁿ

$2 \times \left(\frac{n}{2}\right)^2 \leq cn^2 \rightarrow \frac{1}{2}n^2 \leq cn^2$

$c = 1/2$

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

Iteration Method / Recursive Tree Method

It means to expand the recurrence and express it as a summation of terms of n and initial condition.

Recursion Tree : iteration method which is in the form of a tree, where each level's nodes are expanded. 2 term = root
Each root and child represents cost of single subproblem
Each level costs' sum = pre-level costs. divide till you get 1.
Pre-level costs' sum = total cost.

1 Consider the Recurrence $T(n) = \begin{cases} 1 & \text{if } n=1 \\ 2T(n-1) & \text{if } n>1 \end{cases}$

→ $T(n) = 2T(n-1) = 2[2T(n-2)] = 2^2 T(n-2)$ } From
 $= 4[2T(n-3)] = 2^3 T(n-3) = 2^4 T(n-4)$ } substitution

Repeat the procedure i times

$T(n) = 2^i T(n-i)$ ^{→ terminate} $n-i=1 \rightarrow i=n-1$

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

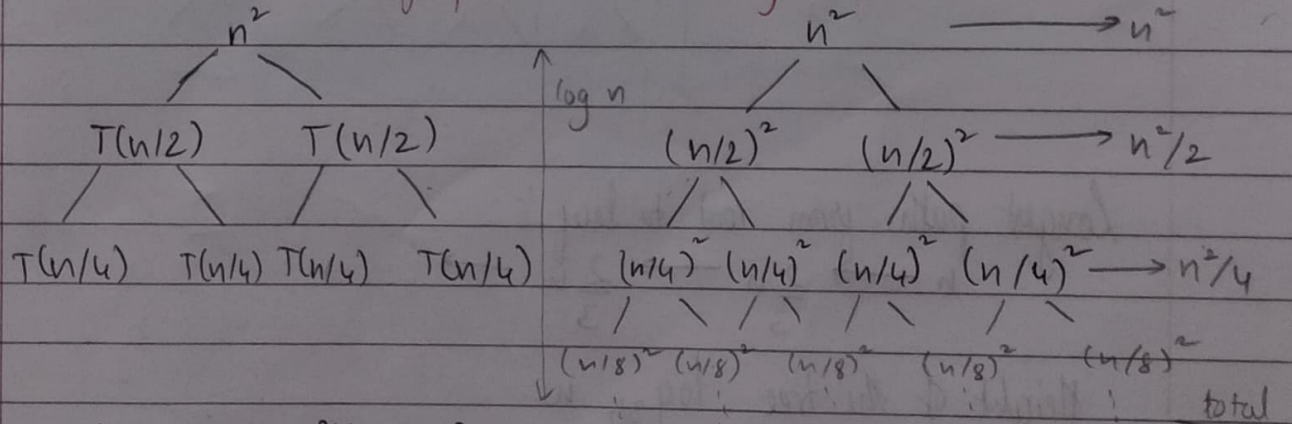
2 $T(n) = T(n-1) + 1$ and $T(1) = \theta(1)$ Substitution.

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

$T(n-k) = T(1) = \theta(1) \rightarrow T(n) = \theta(1) + (n-1) = 1 + n - 1 = n = \theta(n)$

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

We have to obtain asymptotic bound using recursion tree method.



$T(n) = n^2 + n^2/2 + n^2/4 + \dots$
 $\leq n^2 \sum_{i=0}^{\infty} 1/2^i$
 $= n^2 / (1 - 1/2)$

log n times
GP sum = $\frac{1}{1 - 1/2}$

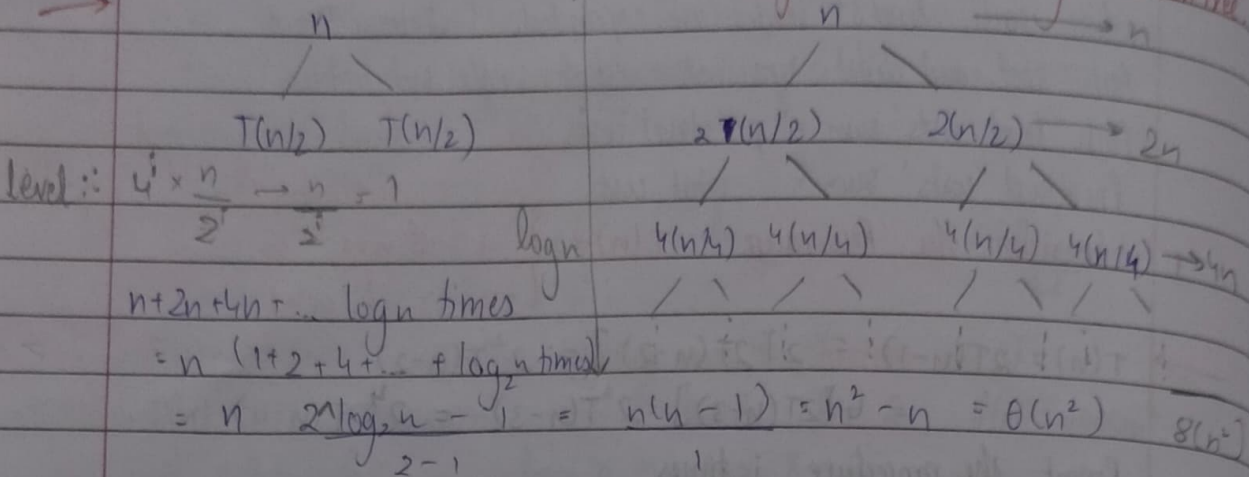
$T(n) \leq n^2 (1 / (1 - 1/2))$
 $\leq 2n^2$

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

$\frac{n}{2} = 1 \rightarrow \log n = i \log_2 2$
 $i = \log n$

$$\frac{1}{1-r} = \text{G.P. sum} = \frac{a(1-r^n)}{(1-r)}$$

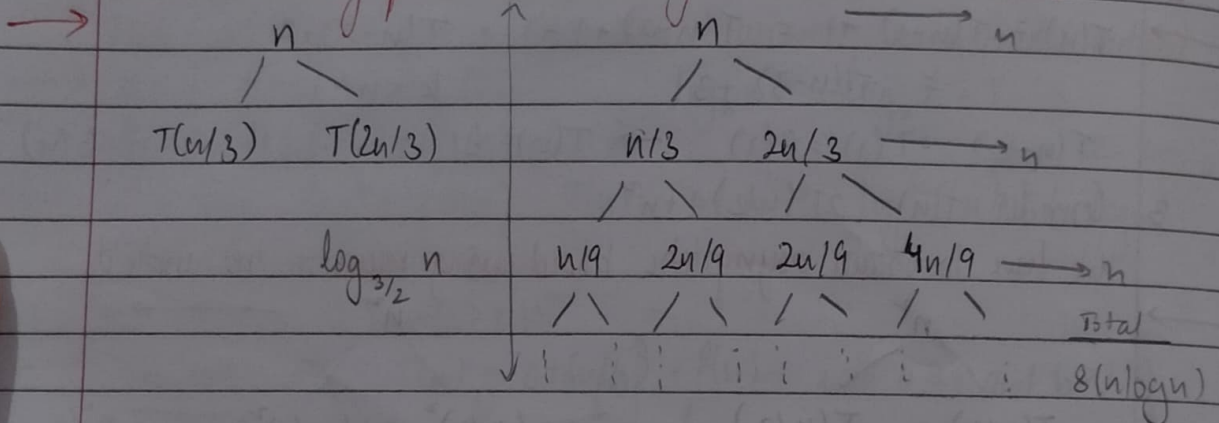
4 $T(n) = 4T(n/2) + n$: obtain the asymptotic bound using recursion tree.



5 Consider the following recurrence

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

obtain the asymptotic bound using recursion tree method



Longest path from root to leaf:

$$n \rightarrow \frac{2n}{3} \rightarrow \frac{2n}{3} \rightarrow \dots$$

$$\frac{n}{3} : \text{level} : \frac{n}{3} = 1$$

Height of the tree $\log_{3/2} n$

$$T(n) = n + n + n + \dots \log_{3/2} n \text{ times}$$

$$= \theta(n \log n)$$