

Recursive Relation

For recursive functions, recurrence relation can be established & can be used to analyze the time complexity of recursive functions.

Example:

Example: To calculate factorial of a number using recursive function

also fact (or)

Time complexity

if ($n == 0 \text{ || } n == 1$) {

return

) (base condition)

else

return $\sum_i i * \text{fact}(n-i)$

recursive call

$$T(n-1)$$

- i. Revenue relation formation

Let $f(n) \rightarrow T(n)$

$$\therefore \text{fact}(n-1) \rightarrow T(n-1)$$

$$T(n) = T(n-1) + 1, \quad n > 1$$

$$= \dots + (-1)^n n! \quad \text{for } n = 0 \text{ or } n = 1$$

For second term in $T(n-1) + 1$, always ~~term 1~~ ~~term 2~~

Asymptotic representation is used



Methods of solving recurrence relation

There are 3 methods of solving recurrence relation

- 1) substitution method } can be applied to
- 2) recursion tree method } all recurrence relt
- 3) master's theorem - cannot be applied to an
 recurrence relation
 decreasing function.

① solve $T(n) = T(n-1) + 1$, $n > 1$
 $= 1$, $n = 0 \text{ or } n = 1$

② using substitution method

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

\therefore Put $n = n-1$ in eqⁿ ①.

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

$$\therefore T(n-1) = \underline{T(n-2)} + 1 \quad \dots \quad ②$$

Put $n = n-2$ in eqⁿ ①

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

$$\therefore T(n-2) = \underline{T(n-3)} + 1 \quad \dots \quad ③$$

Put $n = n-3$ in eqⁿ ①

$$\therefore T(n-3) = T((n-3)-1) + 1$$

$$T(n-3) = \underline{T(n-4)} + 1 \quad \dots \quad ④$$



Put eqⁿ ② in eqⁿ ①.

$$\therefore T(n) = \underbrace{T(n-1)}_{T(n-2)+1} + 1$$
$$\therefore T(n) = [T(n-2)+1] + 1$$

$$\therefore T(n) = \underbrace{T(n-2)}_{T(n-3)+2} + 2 \quad \text{--- } ⑤$$

Put eqⁿ ③ in eqⁿ ⑤.

$$\therefore T(n) = [T(n-3)+1] + 2$$
$$\therefore T(n) = \underbrace{T(n-3)}_{T(n-4)+3} + 3 \quad \text{--- } ⑥$$

Put eqⁿ ④ in eqⁿ ⑥.

$$\therefore T(n) = [T(n-4)+1] + 3$$
$$\therefore T(n) = T(n-4) + 4 \quad \text{--- } ⑦$$

Assume :

$$T(n) = \underbrace{T(n-k)}_{T(1)} + k \quad \text{--- } ⑧$$

Assume $n-k=1$

$\therefore k = n-1$

Substitute value of 'k' in eqⁿ ⑧.

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

$$\therefore T(n) = T(1) + (n-1) \quad \text{--- } ⑨$$

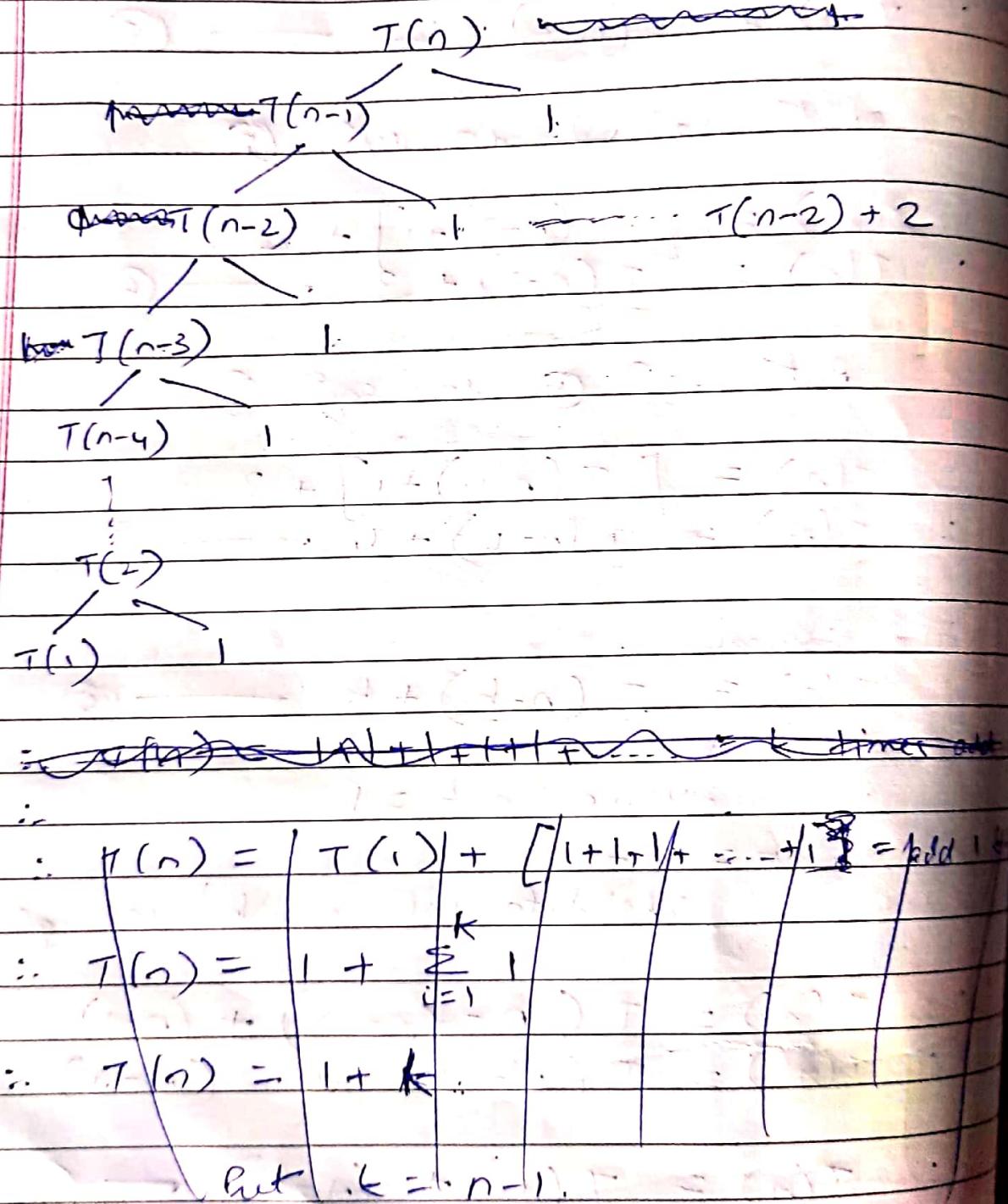
From given recursive relation, $T(1)=1$, put

its value in eqⁿ ⑨.

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

$$\therefore \boxed{T(n) = n} \leftarrow \text{Time complexity}$$

- ② using recursion tree method
calculate the total running time at each level of tree.



* Time complexity = sum of work done at each level

$$T(n) = T(n-k) + [1+1+ \dots k \text{ times}]$$

$$\therefore T(n) = T(n-k) + \sum_{i=1}^k 1$$

$$\therefore T(n) = T(n-k) + k.$$

$$\text{But } n-k=1, \therefore k=n-1$$

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

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

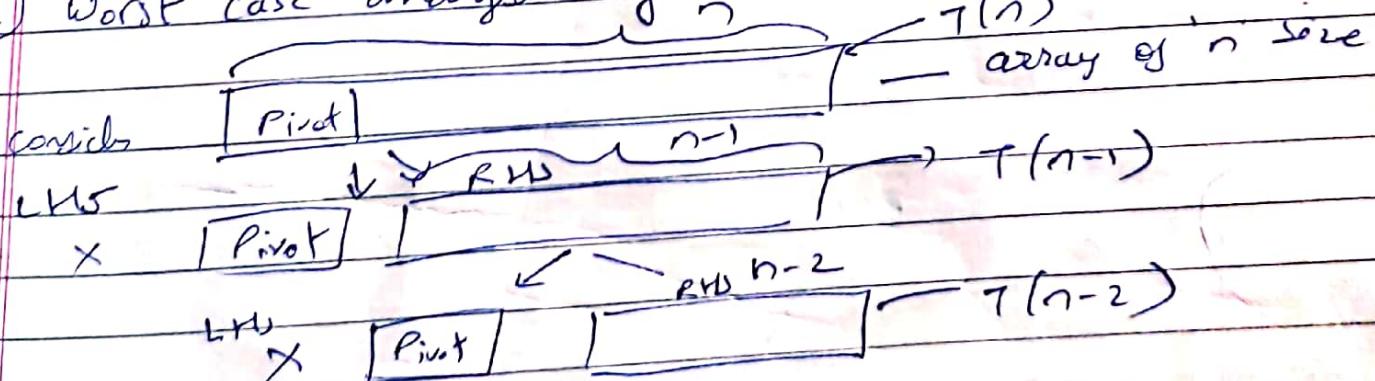
$$\therefore T(n) = T(1) + (n-1).$$

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

$$\therefore T(n) = O(n) \quad | \text{--- Time complexity}$$

~~O(n)~~

(2) Worst case analysis of Quicksort algorithm.



$$\therefore T(n) = T(n-1) + n, n > 1$$

$$= 1, n = 1$$

here $T(n-1) \rightarrow$ time to partition $(n-1)$ elements

$n \rightarrow$ time required for comparisons of pivot with other elements.

(Q) using substitution method

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

Put $n = n-1$ in ①

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

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

Put $n = n-2$ in ①

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

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

Put $n = n-3$ in ①

$$\therefore T(n-3) = T((n-3)-1) + (n-3)$$

$$\therefore T(n-3) = T(n-4) + (n-3) \quad \text{--- } ④$$

Put ④ in ①

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

$$\therefore T(n) = \cancel{T(n-2)} + (n-1) + n \quad \text{--- } ⑤$$

Put ③ in ⑤

$$\therefore T(n) = [T(n-3) + (n-2)] + (n-1) + n$$

$$\therefore T(n) = \cancel{T(n-3)} + (n-2) + (n-1) + n \quad \text{--- } ⑥$$

Put ④ in ⑥

$$\therefore T(n) = [T(n-4) + (n-3)] + (n-2) + (n-1) + n$$

$$\therefore T(n) = T(n-4) + (n-3) + (n-2) + (n-1) + n \quad \dots$$

$$\therefore T(n) = T(1) + \cancel{[T(n-4) + (n-3) + (n-2) + (n-1) + n]}$$

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

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

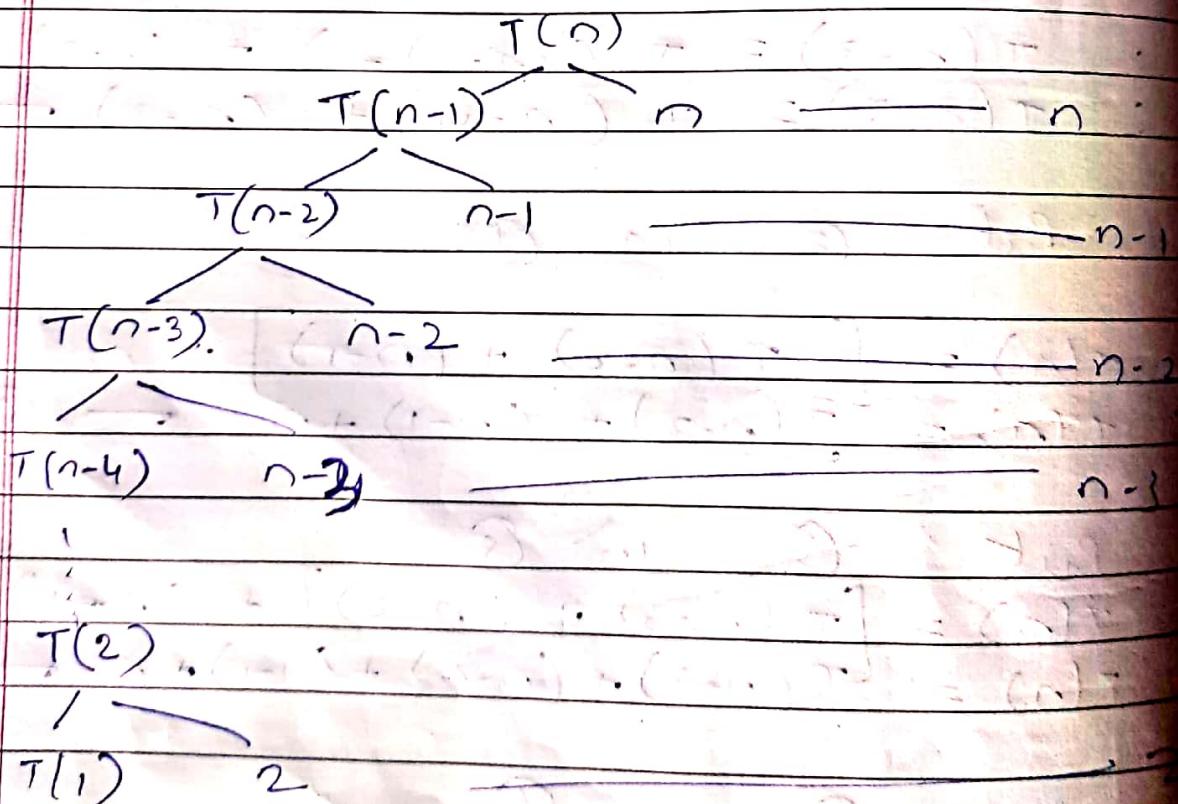
$$\therefore T(n) = \frac{n(n+1)}{2}$$

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

$\therefore T(n) = O(n^2)$ Fastest growing term is $\frac{n^2}{2}$

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

(b) recursion tree method



Time complexity = sum of work done at each level

$$\therefore T(n) = T(1) + [2 + \dots + (n-3) + (n-2) + (n-1) + n]$$

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

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

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

Fastest growing term is $\frac{n^2}{2}$.

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

$$(3) \text{ solve: } T(n) = T\left(\frac{n}{2}\right) + T\left(\frac{n}{2}\right) + n, n > 1$$

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

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

$$\begin{aligned} \text{dividing function} \quad & \therefore T(n) = 2T\left(\frac{n}{2}\right) + n, n > 1 \\ & = 1, n = 1 \end{aligned}$$

For merge sort, $T(n)$ algo mergesort(A,

- represents the recursive relation of
- merge sort algorithm
- best case for quick sort algorithm

For merge sort -

Time complexity

algo mergesort(A, l, h) — $T(n)$

{ if ($l < h$)

mid = $(l + h)/2$

mergesort(A, l, mid) — $T(\frac{n}{2})$

mergesort(A, mid+1, h) — $T(\frac{n}{2})$

merge(A, l, mid, h) — n

 { }

 3

$$\Rightarrow T(n) = 2T\left(\frac{n}{2}\right) + n, \quad n \geq 1. \quad f.$$

$$= 1, \quad n=1$$

For quick sort

Time complexity

algo quicksort(A, l, h) — $T(n)$

{ if ($l < h$)

pos = partition(A, l, h) — n

quicksort(A, l, pos-1) — $T\left(\frac{n}{2}\right)$

quicksort(A, pos+1, h) — $T\left(\frac{n}{2}\right)$

B

for best case

} :

solve,

a) using substitution method

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

Put $n = \frac{n}{2}$ in ①

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

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

$$\therefore T\left(\frac{n}{2}\right) = 2 \left(T\left(\frac{n}{4}\right) + \frac{n}{2} \right) \quad \dots \text{--- } ②$$

Put $n = \frac{n}{4}$ in ①

$$\therefore T\left(\frac{n}{4}\right) = 2 T\left(\frac{n}{8}\right) + \frac{n}{4} \quad \dots \text{--- } ③$$

Put $n = \frac{n}{8}$ in ①

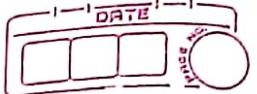
$$\therefore T\left(\frac{n}{8}\right) = 2 T\left(\frac{n}{16}\right) + \frac{n}{8} \quad \dots \text{--- } ④$$

Put ④ in ①

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

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

$$\therefore T(n) = 4 T\left(\frac{n}{4}\right) + 2n \quad \dots \text{--- } ⑤$$



Put ③ in ⑤

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

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

$$\therefore T(n) = 8T\left(\frac{n}{8}\right) + 3n \quad \text{--- } ⑥$$

Put ④ in ⑥

$$\therefore T(n) = 8 \left[2T\left(\frac{n}{16}\right) + \frac{n}{8} \right] + 3n$$

$$\therefore T(n) = 16T\left(\frac{n}{16}\right) + n + 3n$$

$$\therefore T(n) = 16T\left(\frac{n}{16}\right) + 4n \quad \text{--- } ⑦$$

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

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

$\therefore 2^k = n$

$\therefore k = \log_2 n$

Put in ⑧

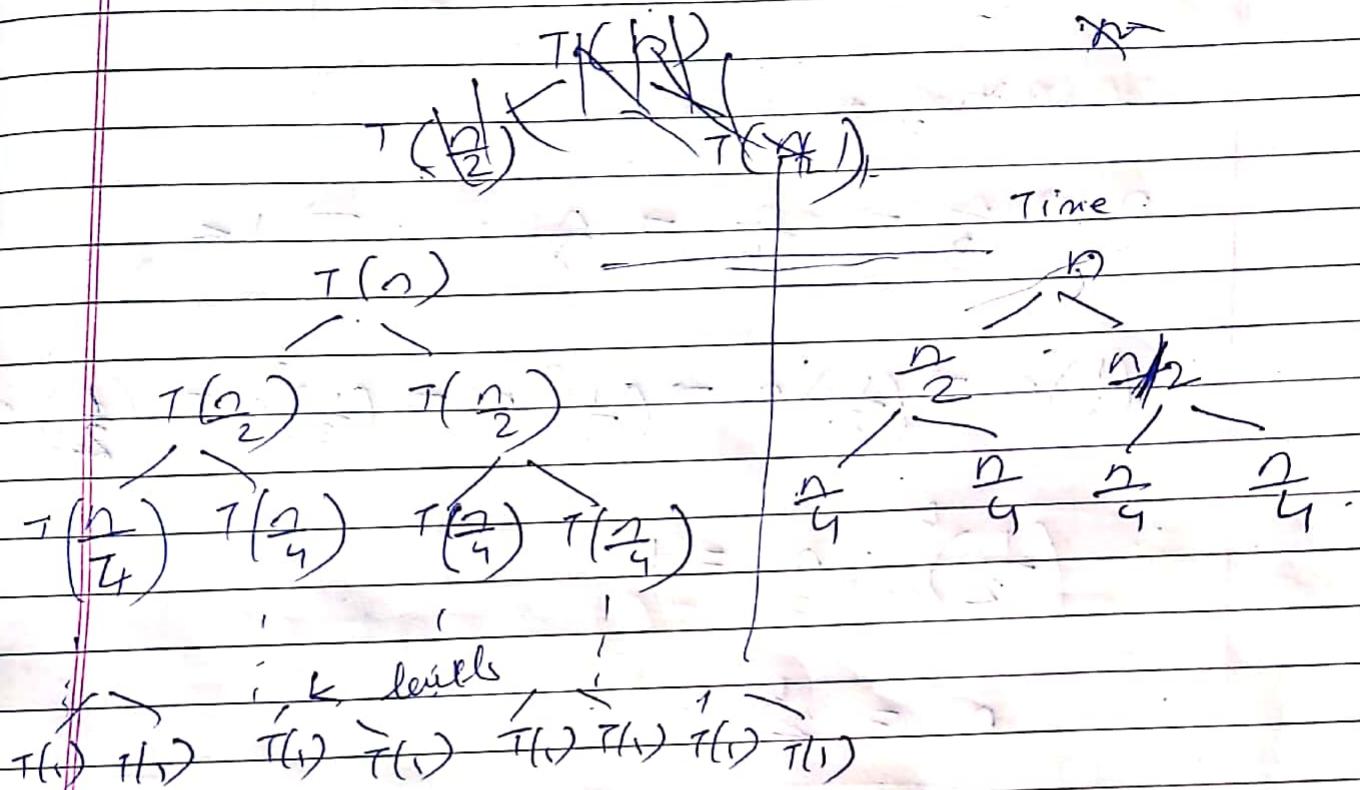
$$\therefore T(n) = nT(1) + (\log_2 n) \cdot n$$

$$\therefore T(n) = n + n \cdot \log_2 n$$

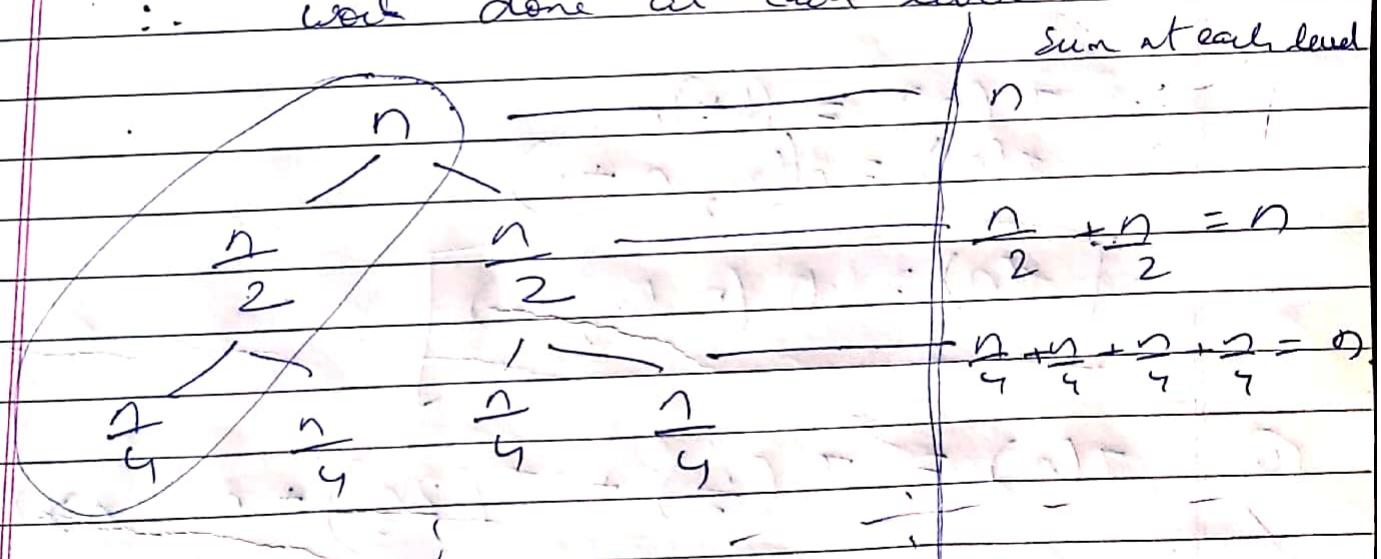
$$\therefore T(n) = n + n \cdot \log_2 n \rightarrow \text{fastest growing term}$$

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

(b) using recursion tree method



\therefore work done at each level



\therefore up to $k+1$ levels; where $k = \text{height of tree}$ $\therefore \text{Total sum} = (k+1)n$

To calculate height i.e. k , consider longest path here both, left hand side and right hand side represent longest path.



lets consider left hand side Path

~~Commonly used~~

$n \rightarrow \frac{n}{2} \rightarrow \frac{n}{4} \rightarrow \frac{n}{8} \dots \rightarrow \frac{n}{2^k}$

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

$$\therefore \left(\frac{1}{2}\right)^k \cdot n = 1$$

$$\therefore n = 2^k$$

$$\therefore k = \log_2 n$$

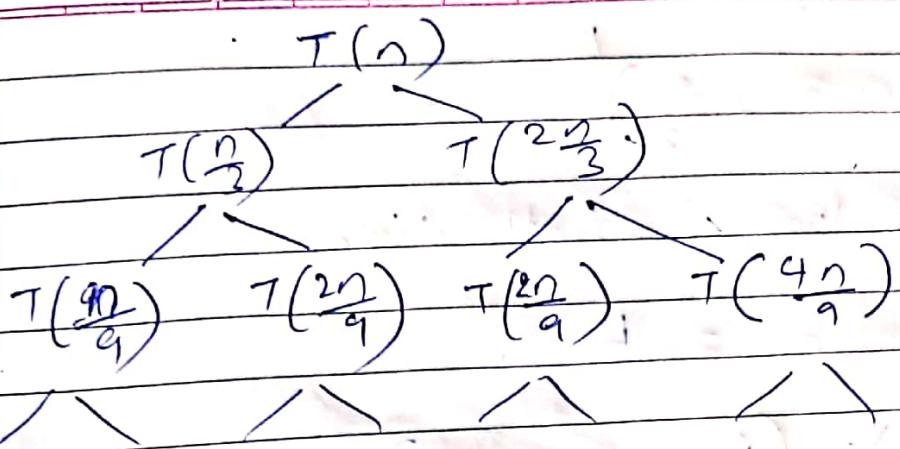
$$\therefore T(n) = (k+1)n$$

$$\therefore T(n) = n(\log_2 n + 1)$$

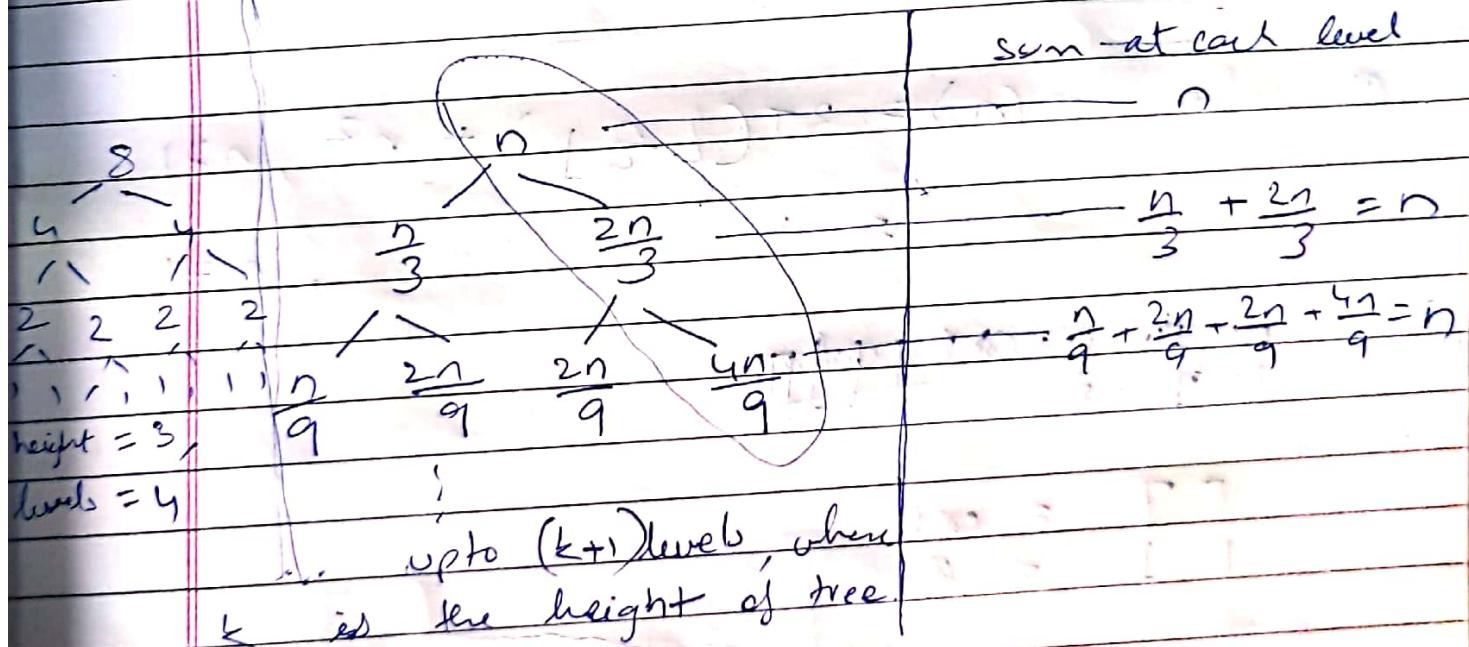
$$\therefore T(n) = O(n \log_2 n)$$

$$(4) \quad T(n) = T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right) + n, n \geq 1$$

Solve using recursion tree method



f. work done at each level



$$\therefore T(n) = \cdot (k+1) \cdot n$$

consider the longest path in the recursion tree

$$\begin{aligned} &: n \rightarrow \frac{2n}{3} \rightarrow \frac{4n}{9} \cdots \rightarrow 15 \\ &\therefore \left(\frac{2}{3}\right)^n \rightarrow \left(\frac{2}{3}\right)^n \rightarrow \left(\frac{2}{3}\right)^2 n \cdots \rightarrow \left[\left(\frac{2}{3}\right)^n\right] \end{aligned}$$

Assume $\left(\frac{2}{3}\right)^k \cdot n = 1$

$$\therefore n = \left(\frac{3}{2}\right)^k$$

taking log on both sides

$$\log_{3/2} n = k$$

$$\therefore T(n) = (\log_{3/2} n + 1)n$$

$$\therefore T(n) = O(n \cdot \log_{3/2} n)$$

③ solve $T(n) = 3T\left(\lfloor \frac{n}{2} \rfloor\right) + n^2, n > 1$

$$= 1, n = 1$$

~~ceil value~~

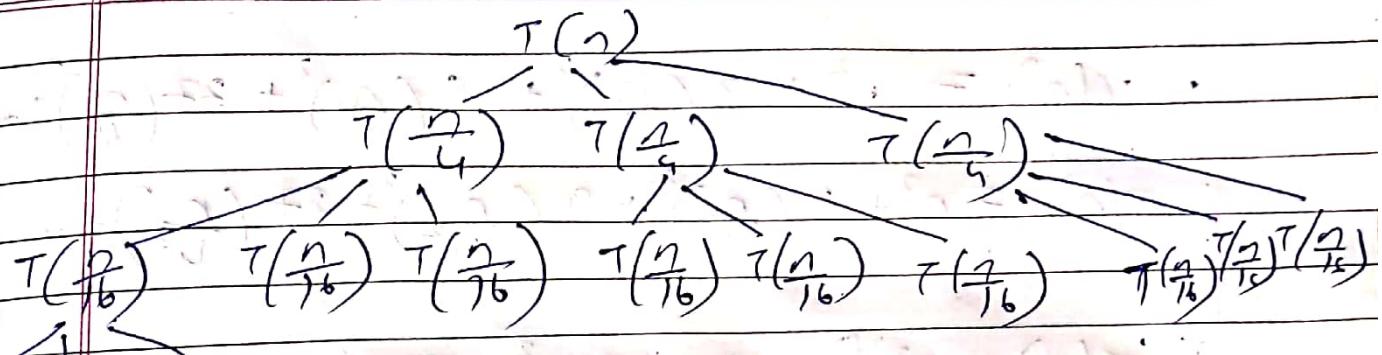
$\lceil \cdot \rceil$ = ceil

$\lfloor \cdot \rfloor$ = floor

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

$$= 1, n = 1$$

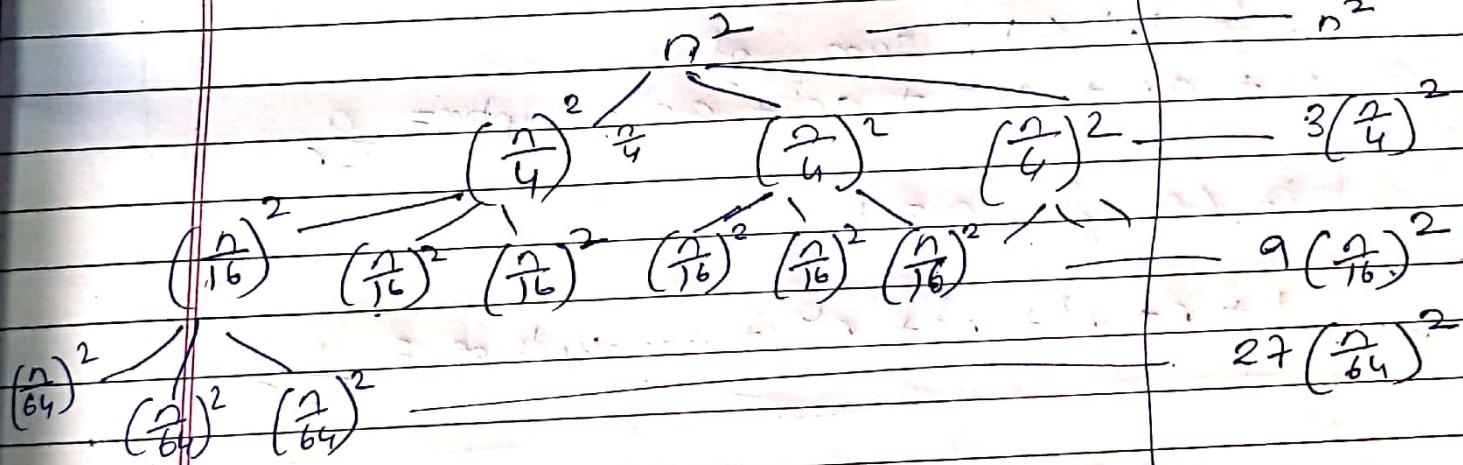
using recursion tree.



upto $(k+1)$ levels

\therefore work done at each level

sum of work done
at each level



Time complexity = sum of work at each level

$$\therefore T(n) = n^2 + \beta \cdot \left(\frac{n}{4}\right)^2 + 9 \cdot \left(\frac{n}{16}\right)^2 + 27 \cdot \left(\frac{n}{64}\right)^2 + \dots$$

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

$$\therefore T(n) = n^2 \left[1 + \frac{3}{16} + \frac{9}{16^2} \right]$$

$$\therefore T(n) = n^2 + \frac{3n^2}{16} + \frac{9n^2}{16^2} + \frac{27n^2}{64} + \dots$$

$$\therefore T(n) = n^2 + \frac{3n^2}{4^2} + \frac{9n^2}{16}$$

$$\therefore T(n) = n^2 + 3\left(\frac{n}{2}\right)^2 + 9\left(\frac{n}{4}\right)^2 + 27\left(\frac{n}{8}\right)^2$$

$$\therefore T(n) = n^2 + 3\left(\frac{n}{2}\right)^2 + 9\left(\frac{n}{4^2}\right)^2 + 27\left(\frac{n}{8^2}\right)^2$$

$$\therefore T(n) = n^2 \left[1 + \frac{3}{16} + \left(\frac{3}{16}\right)^2 + \left(\frac{3}{16}\right)^3 + \dots \right]$$

Geometric Progression with common ratio ($r = \frac{3}{16}$)

\therefore for $|r| < 1$, sum of $a + ar + ar^2 + ar^3 + ar^4 + \dots$ upto $\infty = \frac{a}{1-r}$

Take for $a = 1$:

$$1 + r + r^2 + r^3 + r^4 + \dots \text{ upto } \infty = \frac{1}{1-r}$$

$$\therefore T(n) = n^2 \left[\frac{1}{1 - \frac{3}{16}} \right] \quad \because \left[\frac{3}{16} \right] < 1$$

$$\therefore T(n) = n^2 \left[\frac{1}{\frac{13}{16}} \right]$$

$$\therefore T(n) = n^2 \left(\frac{16}{13} \right)$$

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

constant

master's Theorem:

- It is used to solve recurrence relation.
- it cannot be applied to all recurrence relations.

$$\text{if } T(n) = aT(\frac{n}{b}) + f(n)$$

where $f(n) = \Theta(n^k \log^p n)$ and

$a \geq 1$, $b > 1$, $k \geq 0$ and p is real number

case 1:

if $\log_b a > k$, then $T(n) = \Theta(n^{\log_b a})$

case 2:

if $\log_b a = k$.

a) if $p > -1$, then $T(n) = \Theta(n^k \log^{p+1} n)$

b) if $p = -1$, then $T(n) = \Theta(n^k \log \log n)$

c) if $p < -1$, then $T(n) = \Theta(n^k)$

case 3:

if $\log_b a < k$.

a) if $p \geq 0$, then $T(n) = \Theta(n^k \log^p n)$

b) if $p < 0$, then $T(n) = \Theta(n^k)$

Solving examples using master's theorem.

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

$$a = 9, b = 3, k = 1, p = 0$$

$\because a > 1, b > 1, k > 0 \text{ & } p \text{ is real}$
master's theorem can be applied

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

$$\therefore \log_b a > k.$$

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

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

$$\textcircled{2} \quad T(n) = 2T\left(\frac{n}{2}\right) + an \quad (\text{merge sort } p \\ \text{quick sort } q)$$

$$a = 2, b = 2, k = 1, p = 0$$

$\because a > 1, b > 1, k > 0 \text{ & } p \text{ is real,}$
master's theorem can be applied

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

$$\therefore \log_b a = k \text{ and } p > -1.$$

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

$$= \Theta(n \log^{p+1} n)$$

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

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

$a = 2^n$, $b = \frac{5}{2}$, $k = 1$, ρ is real.

~~Master's theorem~~

$\because a$ is not a constant, master's theorem cannot be applied.

④ $T(n) = 0.5 T\left(\frac{n}{2}\right) + n$

$\because a < 1$, master's theorem cannot be applied

Strassen's matrix multiplication.

① normal matrix multiplication

algo mul (a, b, n)

{
for ($i = 0; i \leq n-1; i++$)

{
for ($j = 0; j \leq n-1; j++$)

$c[i][j] = 0$

{
for ($k = 0; k \leq n-1; k++$)

n^3

$c[i][j] = c[i][j] +$
 $a[i][k] * b[k][j]$

$\frac{3}{3}$

$\frac{3}{3}$

$\frac{3}{3}$

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

e.g.: $\begin{bmatrix} 1 & 6 & 9 \\ 2 & 8 & 3 \end{bmatrix} \begin{bmatrix} 1 & 7 \\ 2 & 4 \\ 3 & 5 \end{bmatrix} = \begin{bmatrix} - & - & - \\ - & - & - \end{bmatrix}$

matrix a 2×3 $P \ 9$ 3×3 $R \times S$ 2×3
matrix b matrix c

(2) using divide + conquer technique.

- assumes following

1) can be applied to square matrix

2) n is a power of 2 i.e. $n = 2^k$

consider 2×2 matrix multiplication:

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \times \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} = \begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix}$$

where

$$\begin{aligned} c_{11} &= a_{11} \times b_{11} + a_{12} \times b_{21} \\ c_{12} &= a_{11} \times b_{12} + a_{12} \times b_{22} \\ c_{21} &= a_{21} \times b_{11} + a_{22} \times b_{21} \\ c_{22} &= a_{21} \times b_{12} + a_{22} \times b_{22} \end{aligned} \quad \left. \begin{array}{l} \text{It consists of} \\ 8 \text{ multiplication} \\ \text{operations \&} \\ 4 \text{ addition} \\ \text{operations.} \end{array} \right\}$$

using divide + conquer, consider $n = 4$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}_{4 \times 4} \times \begin{bmatrix} b_{11} & b_{12} & b_{13} & b_{14} \\ b_{21} & b_{22} & b_{23} & b_{24} \\ b_{31} & b_{32} & b_{33} & b_{34} \\ b_{41} & b_{42} & b_{43} & b_{44} \end{bmatrix}_{4 \times 4} =$$

$$\begin{bmatrix} c_{11} & c_{12} & c_{13} & c_{14} \\ c_{21} & c_{22} & c_{23} & c_{24} \\ c_{31} & c_{32} & c_{33} & c_{34} \\ c_{41} & c_{42} & c_{43} & c_{44} \end{bmatrix}_{4 \times 4}$$

Divide each matrix into 4 sub-matrices ($\frac{n}{2}$)
 such as matrix A divided as

$$\text{matrix } (A_{11}) = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, \quad \text{matrix } (A_{12}) = \begin{bmatrix} a_{13} & a_{14} \\ a_{23} & a_{24} \end{bmatrix}$$

$$\text{matrix } (A_{21}) = \begin{bmatrix} a_{31} & a_{32} \\ a_{41} & a_{42} \end{bmatrix}, \quad \text{matrix } (A_{22}) = \begin{bmatrix} a_{33} & a_{34} \\ a_{43} & a_{44} \end{bmatrix}$$

1) by divide matrix B + matrix C.

then, the 3×3 matrices A, B, C can be represented as

$$\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} * \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} = \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}$$

$$\therefore C_{11} = A_{11} \times B_{11} + A_{12} \times B_{21}$$

$$C_{12} = A_{11} \times B_{12} + A_{12} \times B_{22}$$

$$C_{21} = A_{21} \times B_{11} + A_{22} \times B_{21}$$

$$C_{22} = A_{21} \times B_{12} + A_{22} \times B_{22}$$

it consists of 8 matrix multiplication operations of 4 matrix addition.

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

since asymptotic notation is used for second term:

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

= 1

$n = 2$

using master's theorem,

$$a = 8, b = 2, k = 2, p = 0$$

since $a > 1, b > 1, k > 0$ & p is real

master's theorem can be applied

$$\therefore \log_b a = \log_2 8 = 3$$

since $\log_b a > k$

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

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

(3) using strassen's matrix multiplication

$C = \alpha I_d$.

$$\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} = \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}$$

where $A_{11}, A_{12}, A_{21}, A_{22}$ are submatrices of matrix A , $B_{11}, B_{12}, B_{21}, B_{22}$ are submatrices of matrix B and matrix C

using strassen's method, the values of submatrices C_{11}, C_{12}, C_{21} & C_{22} can be calculated using 7 matrix multiplication operations, $\underline{\underline{B}}$ as follows.

	shortest
$P_1 = A * (F - H)$	A row +
$P_2 = (A + B) * H$	H col +
$P_3 = (C + D) * E$	E
$P_4 = D * (G - E)$	D
$P_5 = (A + D) * (E + H)$	diagonal
$P_6 = (B - D) * (G + H)$	last CR
$P_7 = (A - C) * (E + F)$	First CR

$$\therefore C_{11} = P_4 + P_5 + P_6 - P_2$$

$$C_{12} = P_1 + P_2$$

$$C_{21} = P_3 + P_4$$

$$C_{22} = P_1 - P_3 + P_5 - P_7$$

∴ using strassen's method

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

$$= 1$$

$$n \geq 2$$

using master's theorem.

$$T(n) = \Theta(n^{\log_2 7})$$

$$T(n) = \Theta(n^{2.81})$$

$$T(n) = \Theta(n^{2.81})$$

↳ Strassen's algorithm is more complex

than matrix multiplication but it is faster than standard multiplication

↳ Strassen's algorithm is used in computer graphics and image processing

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