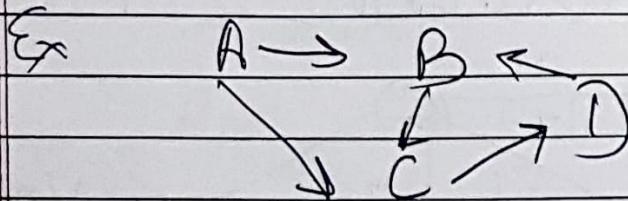


Adjacency Matrix : Better for dense graph

Adjacency List : Better for sparse graph

→ Directed graph :
Only one direction edge



BS502

9. HASHING

Storing data by dynamic DS
called Hash Table

Table organization and searching
technique in which

Insert/Delete/Search operations
done in O(1)

Uses HASH FUNCTION to put
data in Hash table.

Enables to store & retrieve
data quickly. Can be implemented
as an array

Every entry in hash table is associated w/ some KEY.
Hash Function transforms key to an ~~an~~ table index.

Key(k) \rightarrow [Hash Function] \rightarrow Address $h(k)$

Ex:- Name	key(k)	$h(k)$ key % B
DLD A	985926	6
EDLC	970876	10
DSF	986047	11
CDA	970428	5

D	1	2	3	4	5	6	7	8	9	10	11	12
					CDA	D2 D A				EDL C DSF		

Characteristics of good hash fnctr
 → Minimize collision i.e. distribute keys uniformly across the table
 → Compute $h(k)$ efficiently
 → Use all info in the key.

→ Collision:- Diff keys when hash to same locations.
 Colliding keys are known as Synonyms.

Popular Hash Fns

- Division Remainder Method
(Modulo Division Method)
- Mid Square Method
- Folding Method
- Multiplication Method

① Division Remainder Method :-

$$h(k) = k \bmod n \text{ or } k \bmod (n+1)$$

Decide a n .

10

HUFFMAN ALGORITHM

Used for text compression
lossy compression - Quality degraded

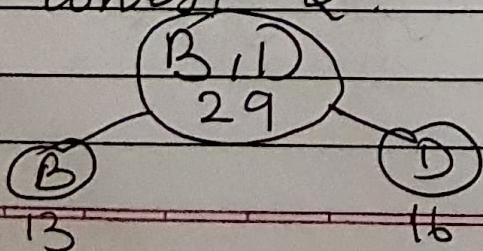
Ex:-

Character →	A	B	C	D	E	∅
Frequency →	22	13	18	16	31	

Sort acc to frequency

B	D	C	A	E
13	16	18	22	31

Combine lowest 2 :-

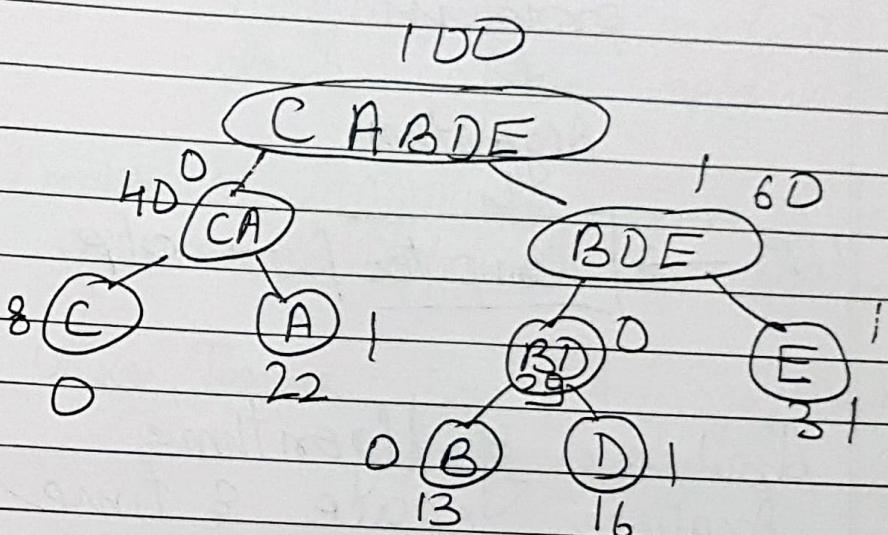


Recorded

C	A	B, D	E
18	22	29	31

BD	E	C, A
29	31	4D

CA	BDE
4D	60



Code for each alphabet

A	0	1	22×2
B	1	0	0
C	0	0	13×3
D	1	0	1
E	1	1	16×3
			31×2

229 bits

Most frequently occurring character has fewer number of bits.

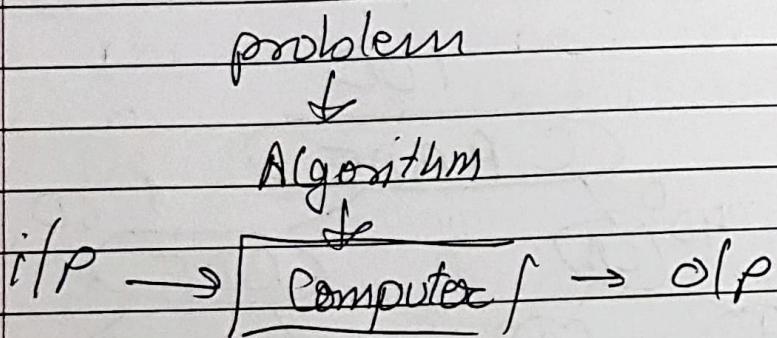
Before compression ($100 + 40 + 60 + 29 + 31 + 18 + 22 + 31 + 13 + 16$) = 800 bit

After compression = 229 bits

Analysis of Algorithms

1 Introduction

Algorithm - Well defined computational procedure that takes some value or set of values as input & produces value(s) as output



→ Analysis of Algorithms
Analyze space & time complexity

Algorithms can be specified in natural language or pseudo code

Time Complexity - Amt of computer time an algo needs to execute & get the result

Space Complexity - Am't of memory needed to run an algo.

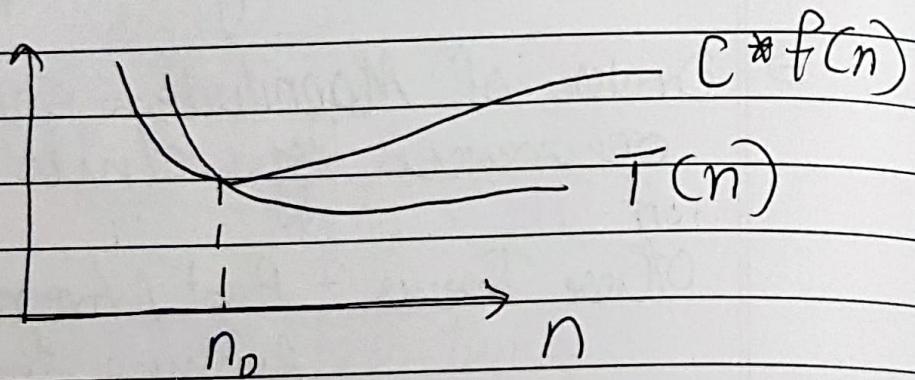
- Priori Analysis
 - Machine independent,
 - done before implementing
- Posteriori Analysis
 - Target machine dependent
 - done after implementing
- Order of Magnitude - Sum of no. of occurrences of statements contained in it.
- Other Terms :- Best / Average / Worst case running time
- Asymptotic Notations - How running time of an algo increases w/ size of input (bounded)

i) \mathcal{O} Notation :- Worst Case running time

for functions $T(n)$ and $f(n)$
nonnegative

$T(n) = \mathcal{O}(f(n))$ if there are positive constants c, n_0 such that

$$T(n) \leq c * f(n) \text{ for all } n, n \geq n_0 \quad (c > 0, n_0 \geq 1)$$



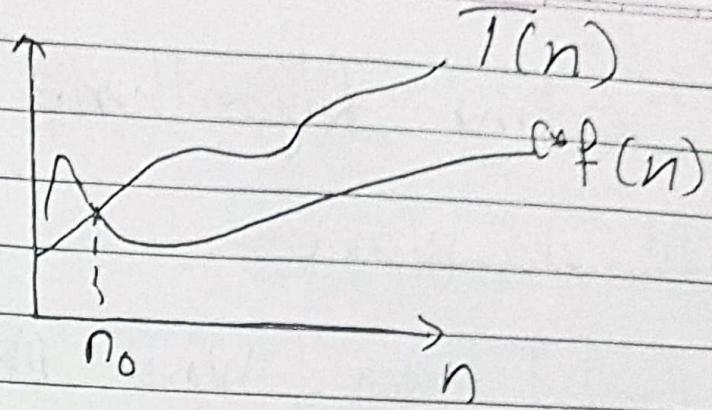
$f(n)$ is the upper bound

ii) $\mathcal{\Omega}$ Notation :- Best Case

$T(n) = \mathcal{\Omega}(f(n))$ if there is c and n_0 such that

$$T(n) \geq c * f(n) \text{ for all } n, n \geq n_0 \quad (c > 0, n_0 \geq 1)$$

$f(n)$ is the lower bound



(iii) Θ notation - Average Case
 $T(n) = \Theta(g(n))$ if there is c_1, c_2, n_0 such that

$$c_1 * g(n) \leq T(n) \leq c_2 * g(n)$$

for all $n, n \geq n_0$
 $(c_1, c_2 > 0, n_0 \geq 1)$

→ Algorithms of two types

① Iterative - Count no. of times instructions are executed

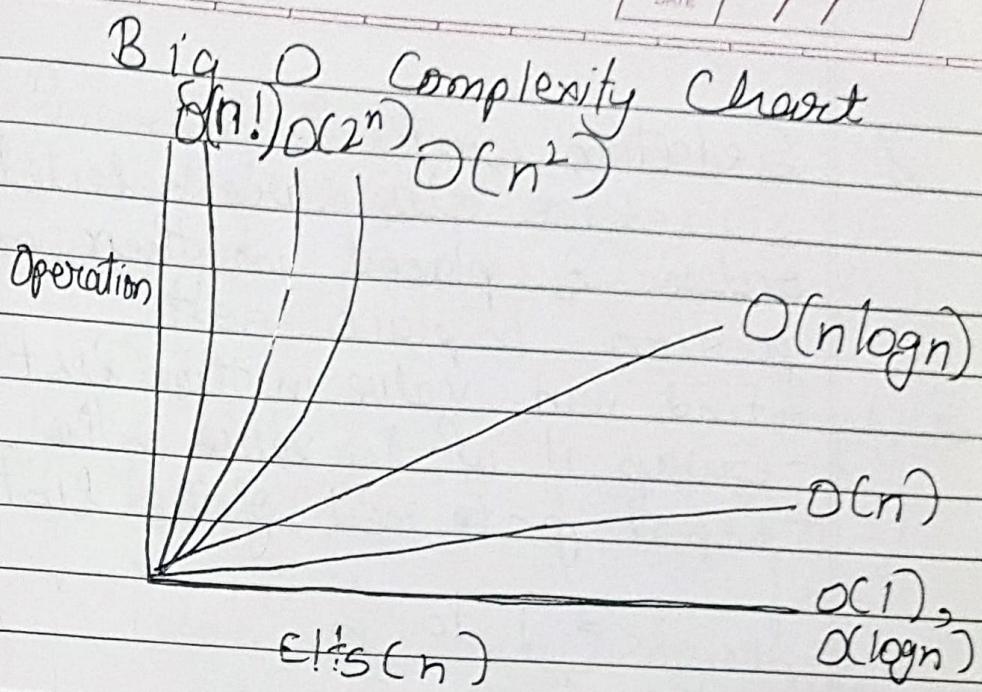
② Recursive - Recurrence equation

2. Sorting Algorithme

At a glance

Name (Sort)	Best Case	Avg Case	Worst Case	Memory	Method	Stable
Quick	$n \log n$	$n \log n$	n^2	$\log n$	Partition	N
Merge	$n \log n$	$n \log n$	$n \log n$	n	Merge	Y
Heap	$n \log n$	$n \log n$	$n \log n$	1	Selection	N
Insertion	n	n^2	n^2	1	Insertions	Y
Selection	n^2	n^2	n^2	1	Selection	N
Bubble	n	n^2	n^2	1	Exchange	Y
Shell	$n \log n$	$n^{4/3}$	$n^{3/2}$	1	Insertions	N
Tree	$n \log n$	$n \log n$	$n \log n$	n	Insertions	Y
Cycle	n^2	n^2	n^2	1	Selection	N
TimSort	n	$n \log n$	$n \log n$	n	Merge + Selection	Y
Used by Python						

TimSort is hybrid of merge + insertion sort



1. Bubble Sort

Simplest algo. Repeatedly swaps adjacent elements if they are in wrong order.

Not suitable for large datasets.

```

for(i=0; i < n-1; i++)
    for(j=0; j < n-i-1; j++)
        if (arr[j] > arr[j+1])
            swap(arr[j], arr[j+1])
    
```

Useful for almost sorted datasets.

2. Selection Sort

Successive elts are selected in order & placed in their proper position. In place sort.

- Find min. value in the list
- Swap it w/ the value in the 1^{st} posⁿ
- Repeat for rest of the list

Step	Cost
for $i = 1$ to n	$n+1$
for $j = i$ to n	$n(n+1)/2$
if ($A[i] < A[j]$) then	$n(n+1)/2 - 1$
$j = k$	$n(n+1)/2 - 1$
swap ($A[i], A[j]$)	1
swap ($A[i], A[j]$)	n

Use when list is small or memory is limited.

3. Insertion Sort : Places an unsorted element and place in its right position. (like cards)

Code

for $j = 2$ to n

key = $A[j]$

$i = j - 1$

while $i > 0$ and $A[i] > \text{key}$

$A[i + 1] = A[i]$

$i = i - 1$

$A[i + 1] = \text{key}$

Analysis

n

$n - 1$

$n - 1$

$\sum_{i=2}^n t_i$

$n - 1$

$n - 1$

$n - 1$

When list is sorted (nearly sorted).
Space complexity of O(1)

3. Merge Sort (Divide & Conquer)

Divides input array in half repeatedly until only one elt left, and merges the two sorted halves.

Merge-Sort (A , low , high)

if ($\text{low} < \text{high}$)

$\text{mid} = (\text{low} + \text{high}) / 2$

Merge-Sort (A , low , mid)

Merge-Sort (A , $\text{mid} + 1$, high)

Merge (A , low , mid , high)

Merge (A, low, mid, high)

$h = \text{low}$, $i = \text{low}$, $j = \text{mid} + 1$

Auxiliary array $B[]$

while ($h \leq \text{mid}$ & $j \leq \text{high}$)

if ($A[h] \leq A[j]$)

$B[i] = A[h]$

$h = h + 1$

else

$B[i] = A[j]$

$j = j + 1$

$i = i + 1$

if ($h > \text{mid}$)

for $k = j$ to high

$B[i] = A[k]$

$i = i + 1$

else

for $k = h$ to mid

$B[i] = A[k]$

$i = i + 1$

for $k = \text{low}$ to high

$A[k] = B[k]$

Uses:- When data structure doesn't allow random access. (linked list ex)

Used in external sorting.

Sequential access

5

Quick Sort (Divide & Conquer)
 Fastest known sorting algo.
 Aka Partition Exchange Sort.
 No merge step involved.

Quicksort (A, lb, ub)
 if ($lb \leq ub$)

 pivot = Partition (A, lb, ub)

 Quicksort ($A, lb, pivot - 1$)

 Quicksort ($A, pivot + 1, ub$)

Partition (A, lb, ub)

$x = A[lb]$

 up = ub

 down = lb

 while ($down < up$)

 while ($A[down] \leq x \& down < up$)

$down = down + 1$

 while ($A[up] > x$)

$up = up - 1$

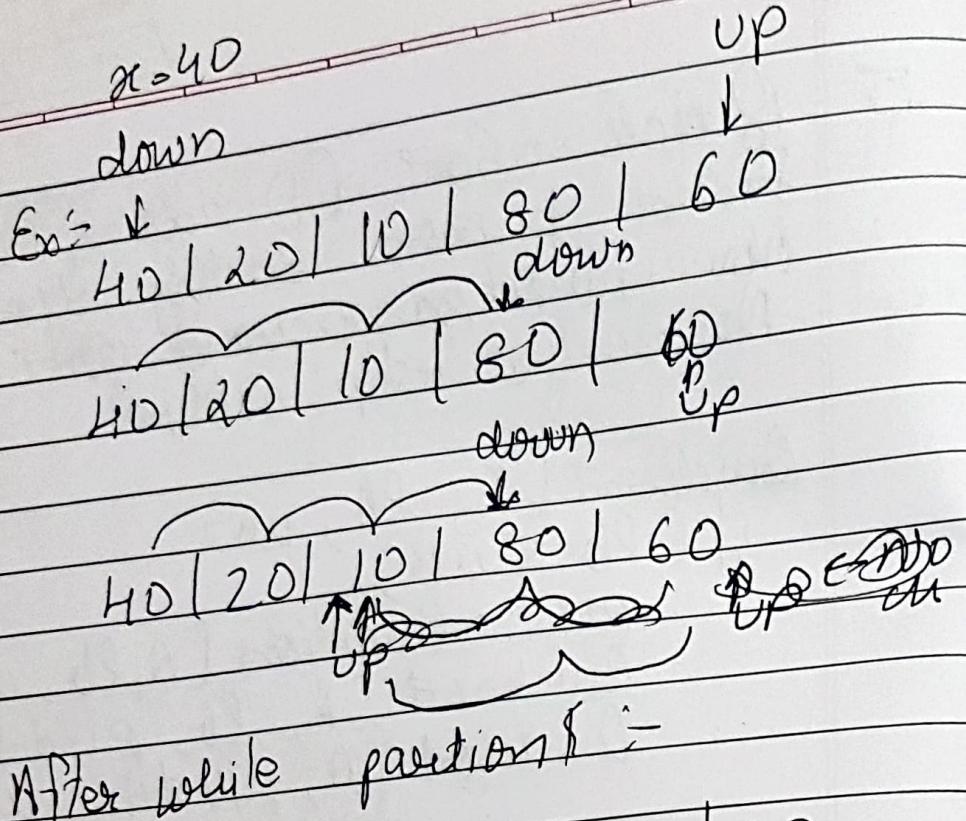
 if ($down \leq up$)

 swap ($A[down], A[up]$)

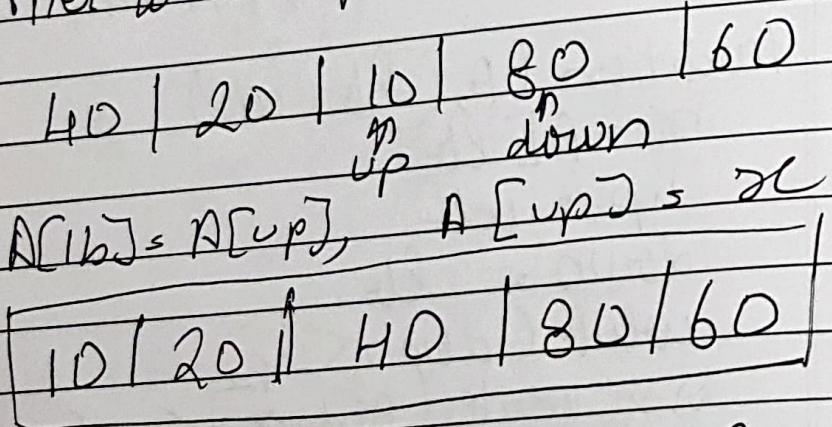
$A[lb] = A[up]$

$A[up] = x$

 Return up



After while partition :-



partition will return 3 as op

When to use :-

More effective when data fits in memory. It is an in-place sort.

Worst performance on sorted array.

Solution :- Randomized quicksort
 - Choose pivot randomly.
 - Make random shuffling.

5.

Searching

1.

Linear Search

linear-search(seq , A)

```
for ( $i = 0$  to  $n$ )
    if ( $A[i] = \alpha$ )
```

return i

return null

Time Complexity: $O(n)$ (Avg/Worst)
 Best: $O(1)$

2.

Binary Search (Divide & Conquer)

bin-search(A , n , target)

left = 1

right = n

while ($left \leq right$)

$mid = \lfloor \log_2 (left + right) / 2 \rfloor$

if ($\text{target} = A[mid]$)

return mid

else if ($\text{target} > A[mid]$)

low = $mid + 1$

else

high = $mid - 1$

Time Complexity Avg/Worst = $O(\log n)$

5. Approximating Time Complexity

→ Iterative :- Count number of times instructions are executed

Ex:- for i = 1 to n
 for j = 1 to n
 for k = 1 to n
 ...

Time Complexity = $O(n^3)$

for i = 1 to n n
 ...
 for j = 1 to n n
 ...

Total = $n + n = 2n$
 Time Complexity = $O(n)$

→ Recursive Algorithms :-

Three methods :-

(i) Recursion Relation/Substitution
 Ex:- $A(n)$

if ($n \geq 1$)

return $A(n/2) + A(n/2)$

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

If return was $n-1$, it would
be

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

* Some Common Recurrences

Recurrence	Solution
$T(n) = T(n/2) + d$	$T(n) = O(\log_2 n)$
$T(n/2) + n$	$O(n)$
$2T(n/2) + d$	$O(n)$
$2T(n/2) + n$	$O(n \log n)$
$T(n-1) + d$	$O(n)$

(ii) Master's Theorem : For divide & conquer algo;

if problem of size n is divided into ~~into~~ a problems of size n/b
let cost of each stage to divide &
conquer be $f(n)$

Then, according to Master's theorem

$$\text{If } T(n) = aT(n/b) + f(n) \quad a \geq 1, b \geq 1 \text{ then,}$$

$$T(n) = \begin{cases} \Theta(n^{\log_b a}) & f(n) = O(n^{\log_b a - \epsilon}) \\ \Theta(n^{\log_b a} \log n) & f(n) = \Theta(n^{\log_b a}) \\ \Theta(f(n)) & f(n) = \Omega(n^{\log_b a + \epsilon}) \end{cases}$$

$$\text{and } af(n/b) < c_f(n) \quad \epsilon > 0, \epsilon > 0, c < 1$$

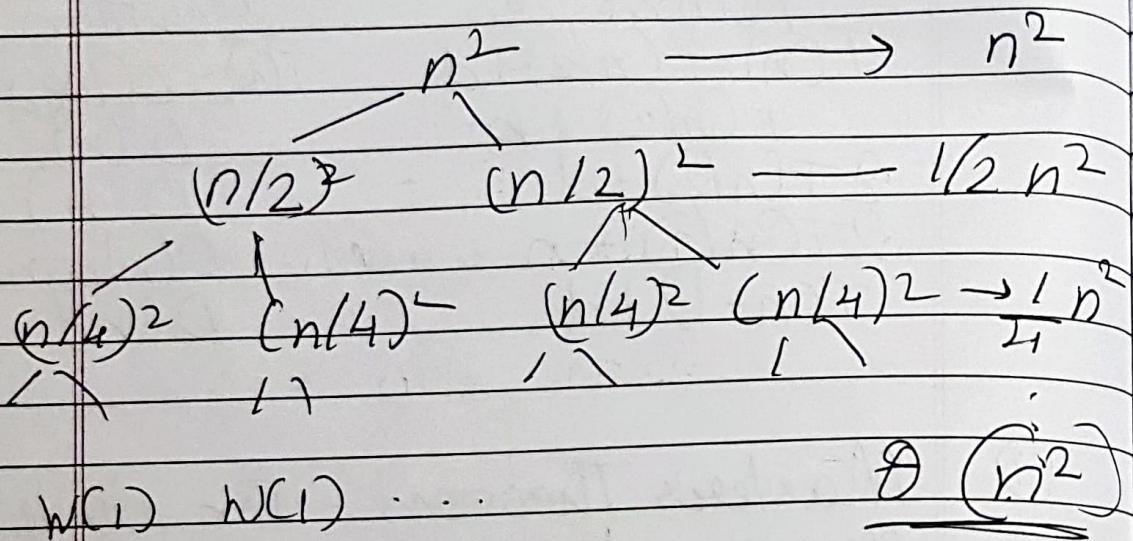
(iii)

Recurrence Tree

Node - Cost at various levels
Sum up cost of all levels

Ex :-

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



5.

DIVIDE & CONQUER

Divide problem to subproblems
 Solve Subproblem - Conquer
 Combine to get solution



Previous examples:-

- (1) - Merge Sort
- (2) - Quick Sort
- (3) - Binary Search

(4)

Strassen's Matrix Multiplication

Standard Matrix Multiplication: $O(n^3)$
 Strassen's - $\sim O(n^{\log 7})$

where n is $n \times n$ matrix

MatMul(A, B, C, n)

if $n = 1$

$$C = C + A * B$$

else

MatMul(A, B, C, $n/4$)

MatMul(A, $(B + n/4)$, $(C + n/4)$, $n/4$)

MatMul($A + 2 \times (n/4)$, B, $C + 2 \times n/4$, $n/4$)

MatMul($A + 2 \times (n/4)$, $B + n/4$, $C + 3 \times n/4$, $n/4$)

MatMul($A + (n/4)$, $B + 2 \times (n/4)$, P, $n/4$)

MatMul($A + (n/4)$, $B + 3 \times (n/4)$, $C + (n/4)$, $n/4$)

MatMul($A + 3 \times (n/4)$, $B + 2 \times (n/4)$, $C + 2 \times (n/4)$, $n/4$)

MatMul($A + 3 \times (n/4)$, $B + 3 \times (n/4)$, C + 3 × (n/4), n/4)

There are 18 $n^{1/2} \times n^{1/2}$ matrix
addⁿ subⁿ $O(n^2)$

7 $n^{1/2} \times n^{1/2}$ multiplications

$$T(n) = 7T(n/2) + O(n^2)$$

$$= O(n^{2+81})$$

$$T(n) = O(n^{\log 7})$$

(5)

Max & Min in a list

- Divide problem into $n^{1/2}$ subproblems & find min and max recursively-

Time Complexity = $O(n)$

MaxMin(i, j, max, min)

if $i = j$

max = min = $A[i]$

else if ($i = j - 1$)

if $(A[i] < A[j])$

min = $A[i]$; max = $A[j]$

else

min = $A[j]$; max = $A[i]$

else

mid = $\lceil (i+j)/2 \rceil$

MaxMin(i, mid, max, min)

MaxMin(mid+1, j, max, min)

if ($\max < \max_1$)

max = \max_1

if ($\min < \min_1$)

min = \min_1

6.

DYNAMIC PROGRAMMING

Method used when problem to be solved is a sequence of decisions. Final solution by combining solution to sequence of subproblems.

Steps :-

- Characterize structure of optimal soln
- Recursively define value of optimal soln
- Compute value of optimal soln, typically in a bottom-up fashion.
- Construct optimal soln from computed info.

(1)

All Pairs Shortest Path

Floyd-Warshall Algorithm

A graph $G(V, E)$, find shortest path between every pair of vertices

$w(i,j) = D$ if $i=j$

$w(i,j) = \infty$ if no connecting edge

$w(i,j)$ is wt of edge.

No of iterations = no of vertices

Algorithm \downarrow No of vertices

~~for~~ $n \leftarrow \text{rows}(W)$

$D(0) \leftarrow n$ ~~# input~~

~~for~~ $k \leftarrow 1$ to n

~~for~~ $i \leftarrow 1$ to n

~~for~~ $j \leftarrow 1$ to n

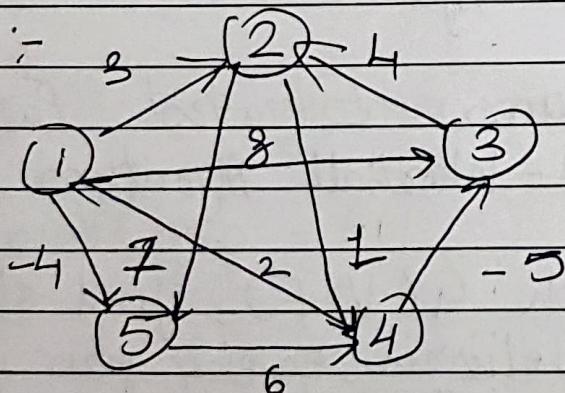
$$D_{ij} \leftarrow \min(D_{ij}, D_{ik} + D_{kj})$$

return D

$$\cancel{D_0 = W_0}$$

Time complexity = $O(n^3)$

Ex :-



	1	2	3	4	5
1	0	3	8	∞	-4
2	∞	0	∞	1	not
3	∞	4	0	∞	∞
4	2	∞	-5	0	∞
5	∞	∞	∞	6	0

Iteration 1 :- D_1 -> Distance from to all nodes via node 1

$D^{(1)}_s$	1	2	3	4	5
1	0	3	8	∞	-4
2	∞	0	∞	1	7
3	∞	4	0	∞	∞
4	2	5	-5	0	-2
5	∞	∞	∞	6	0

Expl:-

$i = 4, j = 2$ To go from node 4 to 2 via 1, total cost
 $= \min(\infty, 2 + 3 = 5) = 5$

Similarly $i = 4, j = 3$

$$W^b = \min(-5, 2 + 8 = 10) = -5$$

Iteration k=2

	1	2	3	4	5
1	0	3	8	4	-4
$D^{(2)}_s$	∞	0	∞	1	7
2	∞	4	0	5	11
3	2	5	-5	0	-2
4	∞	∞	∞	6	0