

### IV Merge Sort -

Merging two sorted sub-arrays

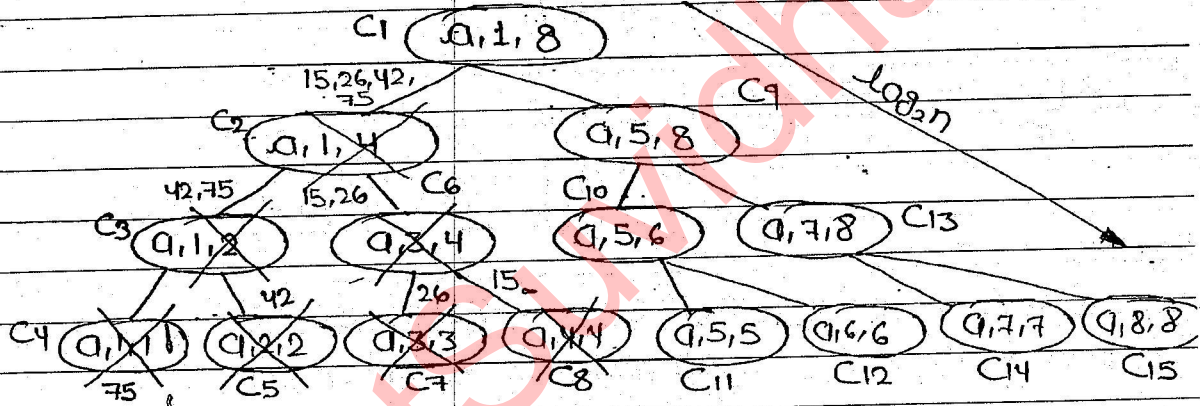
i/p: an array of n-elements

o/p: sorted array

Smaller-problem: 1-element

ex

1	2	3	4	5	6	7	8
75	42	26	15	99	84	16	21



Push - calling function  
Pop - returning f<sup>n</sup>

$\log_2 n$	<del>C4</del>	<del>C5</del>	<del>C7</del>	<del>C8</del>	C11	C12	C14	C15
	<del>C3</del>	<del>C6</del>	C10	C13				
	<del>C2</del>	C9						
	C1							

Stack representation

Merge algorithm

i/p - two sorted sub-arrays

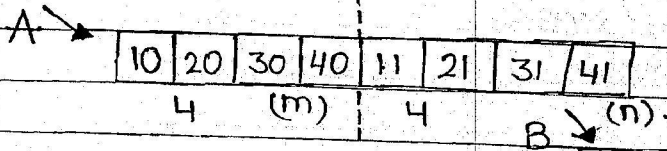
o/p - one sorted array

Merge-sort is  
a super-sort  
of merge.

Combine step.

In the algorithm -

1. If



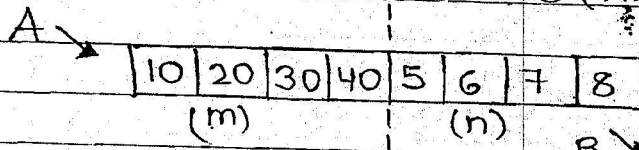
Moves =  $m + n$

Comparisons =  $m + n - 1$

[ WORST CASE ]

Out - place

Time Complexity = moves + comparisons  
=  $O(m+n)$ .



Moves =  $m + n$

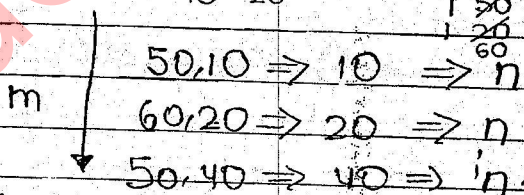
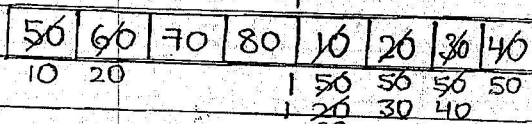
Comparisons =  $\min(m, n)$

[ BEST CASE ]

Out - place.

Time complexity = moves + comparisons  
=  $O(m+n)$

Merging two sub-arrays (sorted) each of size m and n will take  $O(m+n)$



In-place

worst-case algorithm.

⇒  $O(m \cdot n)$ .

If  $m=n$       $O(n^2)$

Merge-sort algorithm (a, i, j)

```

{
  if (i==j)
    then return a[i];
  else {
    mid = (i+j)/2;
    merge-sort (a, i, mid);      T(n/2)
    merge-sort (a, mid+1, j);   T(n/2)
    merge (a, i, mid, mid+1, j) O(n) Every case
  }
}

```

Recurrence relation - Let  $T(n)$  be time complexity of merge-sort algorithm

On  $n$ -elements

$$T(n) = \begin{cases} O(1) & \text{if } n=1 \\ 2T\left(\frac{n}{2}\right) + O(n) + O(1) & \text{if } n > 1 \end{cases}$$

$$T(n) = 2T\left(\frac{n}{2}\right) + n \rightarrow (\text{divide + combine})$$

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

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

$$= \dots$$

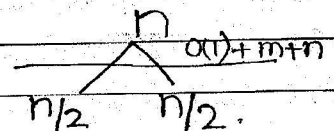
$$= 2^k T\left(\frac{n}{2^k}\right) + n \cdot k$$

$$k = \log_2 n$$

$$T(n) = 2^{\log_2 n} T(1) + n \log_2 n$$

$$= n + n \log_2 n$$

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



Merge-sort (Inplace)

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

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

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

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

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

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

$$k = \log_2 n$$

$$T(n) = 2^{\log_2 n} T(1) + n^2 \left[ \frac{1}{2^0} + \frac{1}{2^1} + \dots + \frac{1}{2^{\log_2 n - 1}} \right]$$

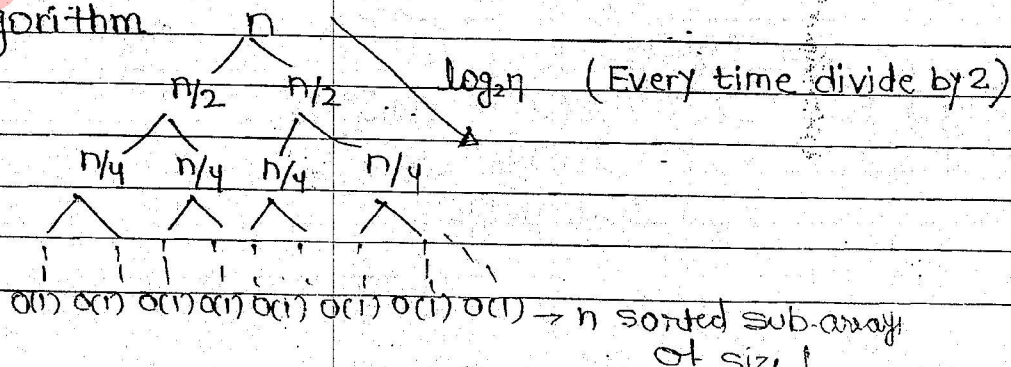
$$= n + n^2 \left[ \frac{1 - (1/2)^{\log_2 n}}{1 - 1/2} \right]$$

$$= O(n^2)$$

Merge-sort algorithm is out-place algorithm becoz in merge we are taking another array

If the array size is very large - merge sort else insertion sort is advisable.

Stable algorithm



Ex 1

i/p:  $\log n$  sorted sub-arrays each of size  $n/\log n$

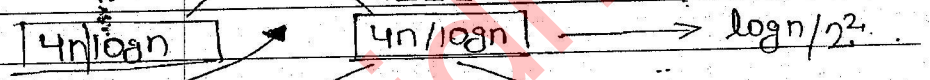
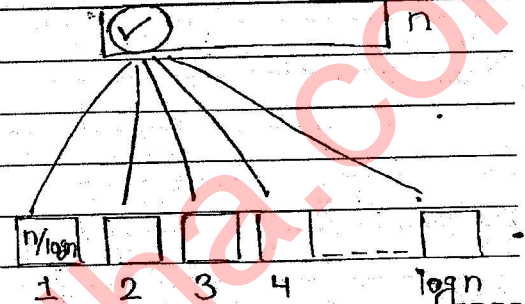
o/p: sorted array of size  $n$ .

Best algorithm, worst case.

Direct merging  
 $O(n \log n)$

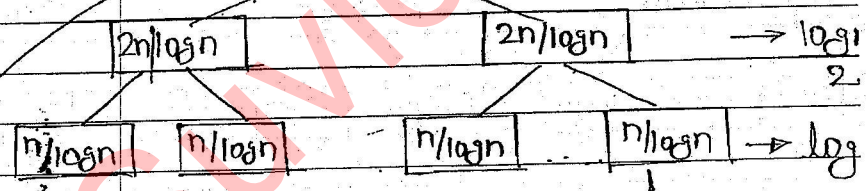
$\log n + \log n + \log n + \dots$

$n$  times



Divide & Conquer

$\log_2(\log n)$



$\frac{2n}{\log n} \times \frac{\log n}{2} = O(n)$	$\frac{\log n}{\log n} \times n = O(n)$
--	---

$K$  times.  $\frac{2^k n}{\log n} \times \frac{\log n}{2^k} = O(n)$

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

Total cost  $O(n \log_2 \log_2 n)$  [Divide & Conquer]

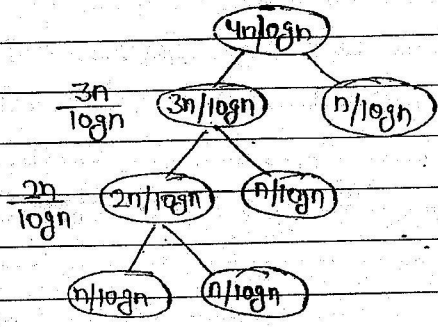
Merge algorithm.

Taking one at a time

$\frac{n}{\log n} [1 + 2 + 3 + \dots + \log n]$

$\frac{n}{\log n} \left[ \frac{\log n (\log n + 1)}{2} \right]$

$= O(n \log n)$

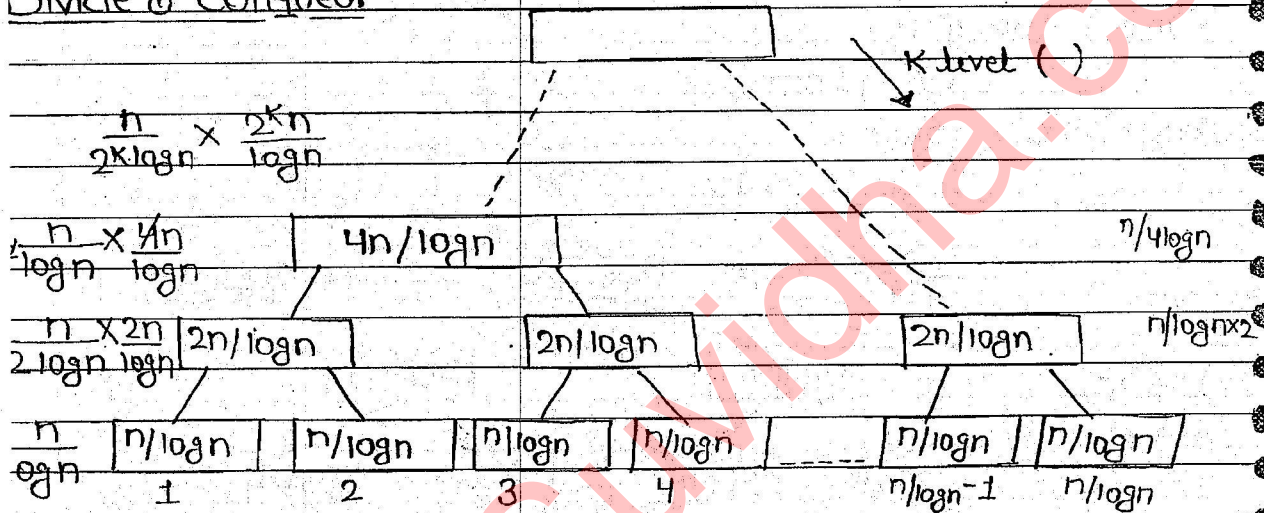


Ex-2

ip:  $\frac{n}{\log n}$  sub-arrays each of size  $\frac{n}{\log n}$

o/p: single sorted array

Divide & Conquer



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

$$\frac{n}{\log n} = 2^k \Rightarrow k = \log_2 \left[ \frac{n}{\log n} \right]$$

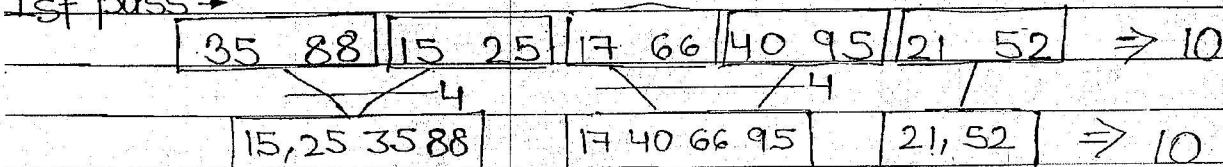
Total cost  $O\left(\frac{n^2 \log_2 n}{(\log n)^2}\right)$

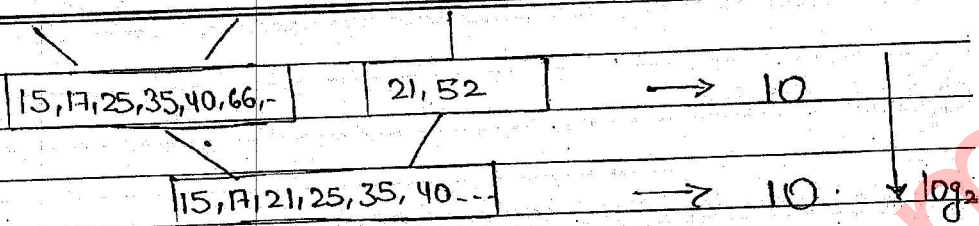
Ex-3 Consider the following array

88, 35, 25, 15, 66, 17, 95, 40, 52, 21

What will be the o/p after second pass of straight two-way merge sort algorithm.

1st pass →





Ex. 4 I/p : 2-sorted sub-arrays  $A \rightarrow m$   $B \rightarrow n$

O/p : find  $A \cap B, A \cup B$

Intersection -

I Linear search  $\rightarrow O(mn)$  II Binary search  $\rightarrow O(m \log n)$

III Merge algorithm  $\rightarrow$  skip the smaller one. if equal, then print. Complexity same.  $O(m+n)$

UNION -

I Linear search  $\rightarrow O(mn)$  II Binary search  $O(n \log m)$

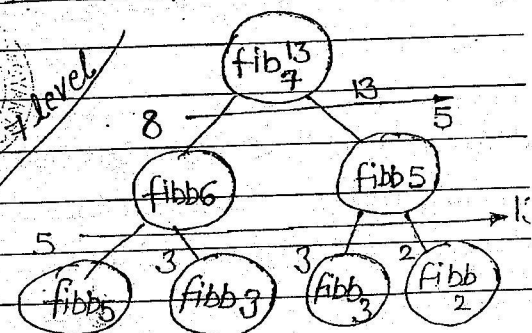
III Merge algorithm  $\rightarrow$  Print one time, skip both, if equal otherwise print both  $O(m+n)$

Ex. 5

n	0	1	2	3	4	5	6	7
fib(n)	0	1	1	2	3	5	8	13

7-level  $O(7 \cdot \text{fibb}_7)$

K-level  $O(K \cdot \text{fibb}_k)$



Divide-time - constant

fibonacci merge-sort

$\rightarrow$  divide fib series manner

Combine  $O(m+n)$

## V Quick-SORT

1. Application of Divide & Conquer.
2. Inplace algorithm (that's why Quick)
3. Not stable

before 20 10<sub>a</sub> 15 10<sub>b</sub> 45  
 after 10<sub>b</sub> 10<sub>a</sub> 15 20 45

Decreasing GP  
 series takes  
 constant time!

↗ repeated element order may change.

1. Most practical used sorting algorithm.
2. If array already sorted, then  $O(n^2)$

40 25 70 90 35 16 44 88 15 77 10 55  
 1 ( 2 3 4 5 6 7 8 9 10 11 12 )

↓ Divide meaningfully

(10 15 16 35 25) 40 (70 90 44 88 77 55)  
 ↓ sort ↓ sort (conquer)  
 [10 15 16 25 35] 40 [44 55 70 77 88 90]  
 Pivot

No combine

Partial divide & conquer application

Divide time  $n$  ( $n$  comparisons)

Conquer time  $2T(n/2)$

for merge sort  $2T(n/2) + n \Rightarrow O(n \cdot \log n)$   
 $2T(n/2) + n + c \Rightarrow O((n+c) \log n)$



### Partition algorithm

partition (a, p, q)

{

x = a[p];

i = p;

for (j = i+1; j <= q; j++)

{

if (a[j] <= x)

{

i++;

swap (a[i], a[j]);

}

}

swap (a[i], a[p]);

return (i);

}

Quicksort (A, p, q) {

m = partition (a, p, q)

comparisons  $n-1$  (BC, WC, AC). O

Quicksort (A, p, m);

Quicksort (A, m+1, q);

}

Best case

No. of swaps

1

[Partition

Worst case

No. of swaps

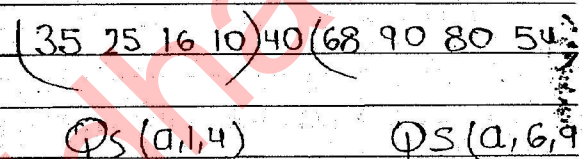
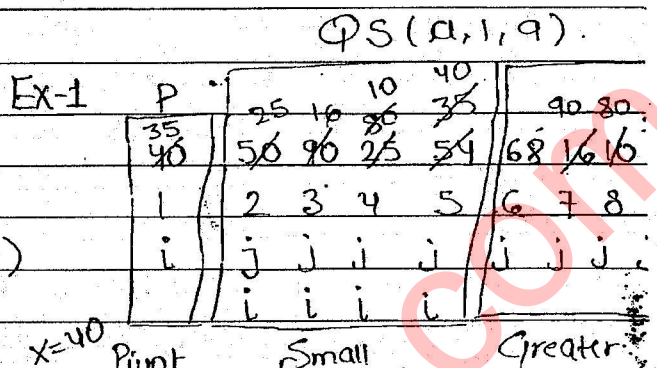
n

Algorithm]

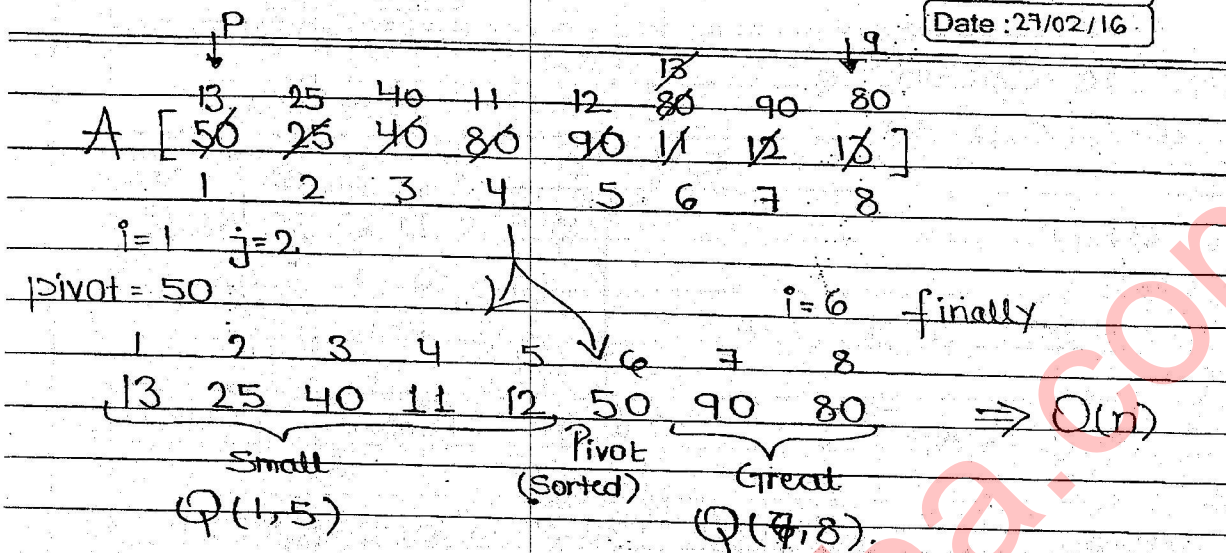
Time complexity of n elements on partition algorithm is  $O(n)$  [every case] and it is in-place.

Ex 2. A [50, 25, 40, 80, 90, 11, 12, 13]

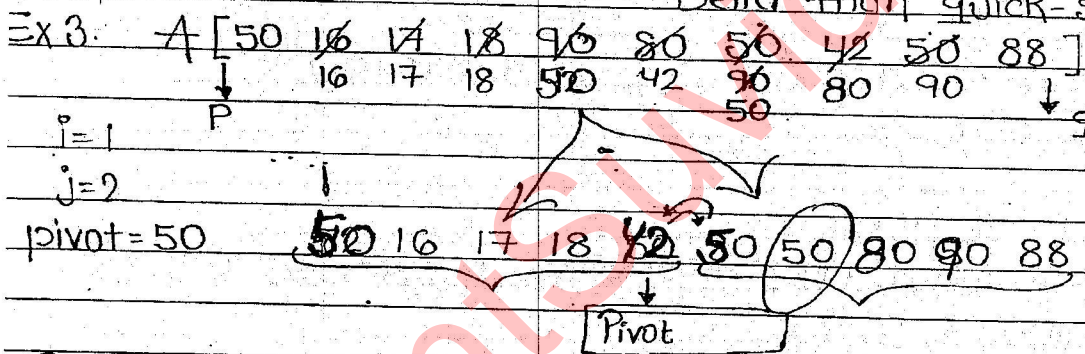
x=50 i=1 j=2 p=1 q=8



return (a, 1, 9)



Randomized Quick-sort  $\rightarrow$  pivot selection random  
 $\rightarrow$  better than quick-sort



```

QuickSort (A, p, q) // Algorithm
{
  if (p == q) // small problem.
    return A[p];
  else {
    int m = partition (A, p, q); // Divide
    QuickSort (A, p, m-1); // Conquer
    QuickSort (A, m+1, q);
    return (A); // No combine
  }
}

```

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

Let  $T(n)$  be time complexity of quick-sort algorithm then recurrence relation

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

In general,

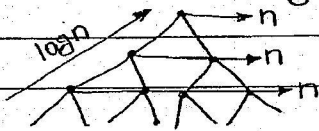
$$T(n) = \begin{cases} O(1) & \text{if } n=1 \\ O(n) + T(m-p) + T(q-m) & \text{if } n > 1 \end{cases}$$

$\begin{matrix} & \swarrow & \searrow \\ P & & q \end{matrix}$

**Best case**

- Balanced tree, exactly two partitions

$$\begin{aligned} T(n) &= O(n) + 2T(n/2) \\ &= 2T(n/2) + n \\ &= O(n \log n) \end{aligned}$$

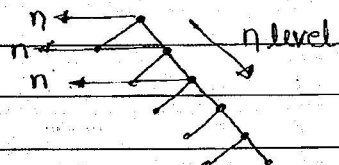


Stack space -  $\log_2 n$   
Total space -  $n + \log_2 n$

**Worst case**

- Unbalanced tree, one to  $n-1$  partition

$$\begin{aligned} T(n) &= O(n) + T(n-1) \\ &= T(n-1) + n \\ &= O(n^2) \end{aligned}$$



Stack space -  $n$   
Using better program -  $\log_2 n$

**Average case**

$$\begin{aligned} T(n) &= 2T(n/2) + n \\ &= 2[T(n/2) + n/2] + n \end{aligned}$$

$$= 2T(n/2) + 2n \quad \left[ \text{Small constants does not matter} \right]$$

$$= O(n \log n) = \text{Best case (same recurrence rel)}$$

Worst case if array is already sorted or almost sorted.

Ex- Quicksort algorithm is run on following two i/p to keep in ascending order.

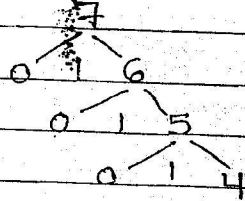
1st i/p - 1, 2, 3, ..., n      Left Gap

2nd i/p - n, n-1, n-2, ..., 1      Alternate Gap

Let  $C_1$  and  $C_2$  be the no. of comparisons made for the i/p I and II respectively. Then relation between  $C_1$  &  $C_2$ .

Solution - Take any i/p of the form to apply partition

1st I/P      10   20   30   40   50   60   70



7 elements  
6 comparisons, 1 swap

10 (20 30 40 50 60 70)

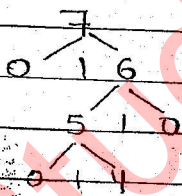
6 elements  
5 comparisons, 1 swap

2 elements  
1 comparison, 1 swap

In general  $(n-1) + (n-2) + (n-3) + \dots + 2 + 1$

$\frac{n(n-1)}{2} = O(n^2)$  Comparisons

2nd I/P      10   n-1      40   30   20   10   70   Swaps  
70   60   50   40   30   20   10



7 elements, 6 comparisons, 7 swaps

10 (60 50 40 30 20) 70

6 elements, 5 comparisons  
1 swap

10 (60 50 40 30 20) 70

4 elements, 3 comparisons  
4 swaps

Comparisons =  $O(n^2)$

Swaps =  $n+1 + n+1 + \dots$

$= \frac{n \times n}{2} + \frac{n \times 1}{2} = O(n^2)$       3 elements, 1 swap, 2 comparisons

2 elements, 2 swap, 1 comparison

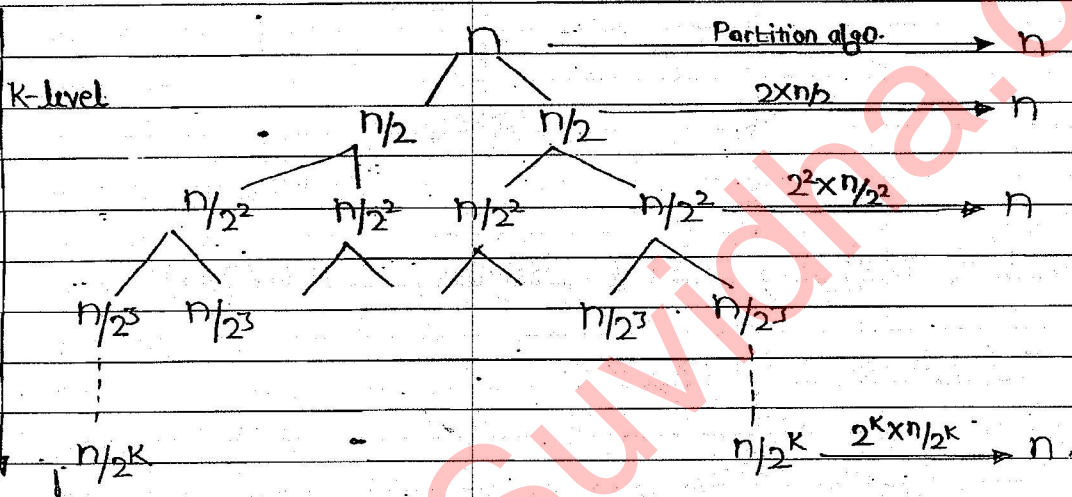
So,  $C_1 = C_2$  (Answer)       $S_1 < S_2$  (swaps)

∴ Quick-sort gives best case to average case performance if the array is not sorted.

Recursive Tree method →

Two recursion calls

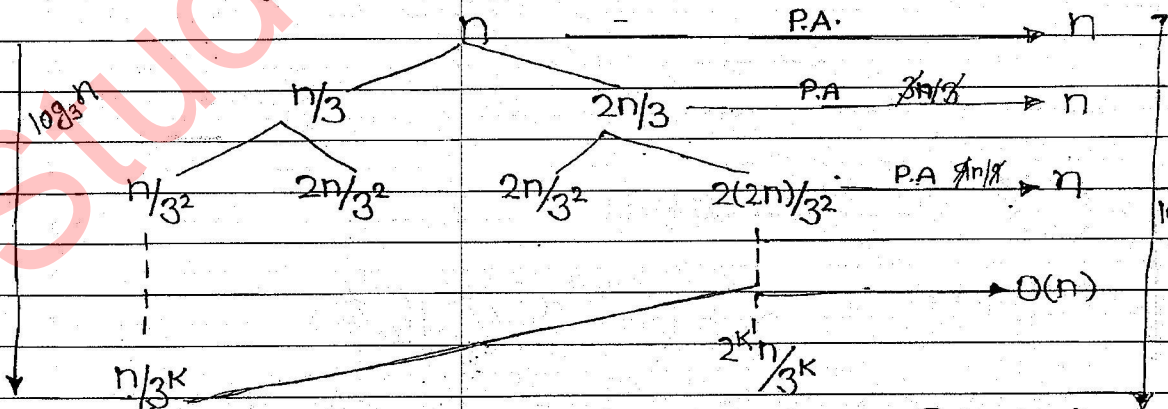
$$1. T(n) = T(n/2) + T(n/2) + n$$



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

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

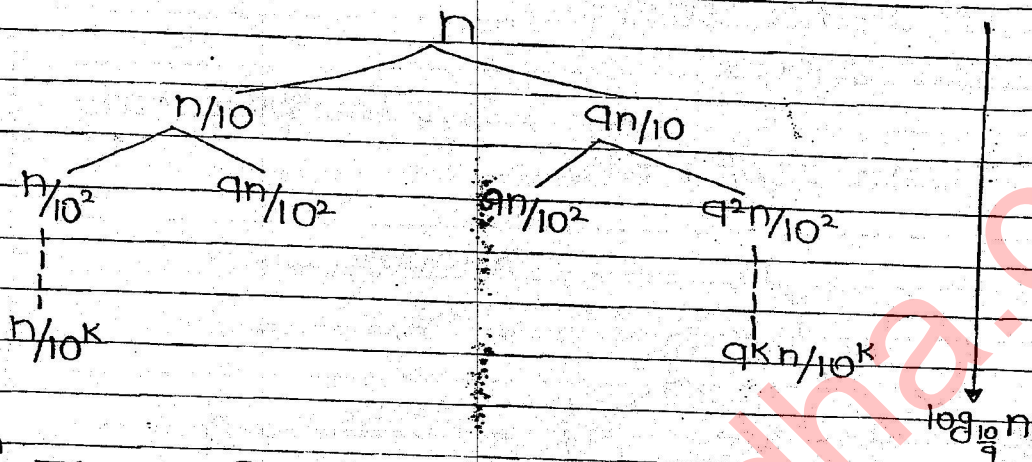
$$2. T(n) = T(n/3) + T(2n/3) + n$$



$$\frac{n}{(3/2)^k} = 1 \Rightarrow k = \log_{3/2} n$$

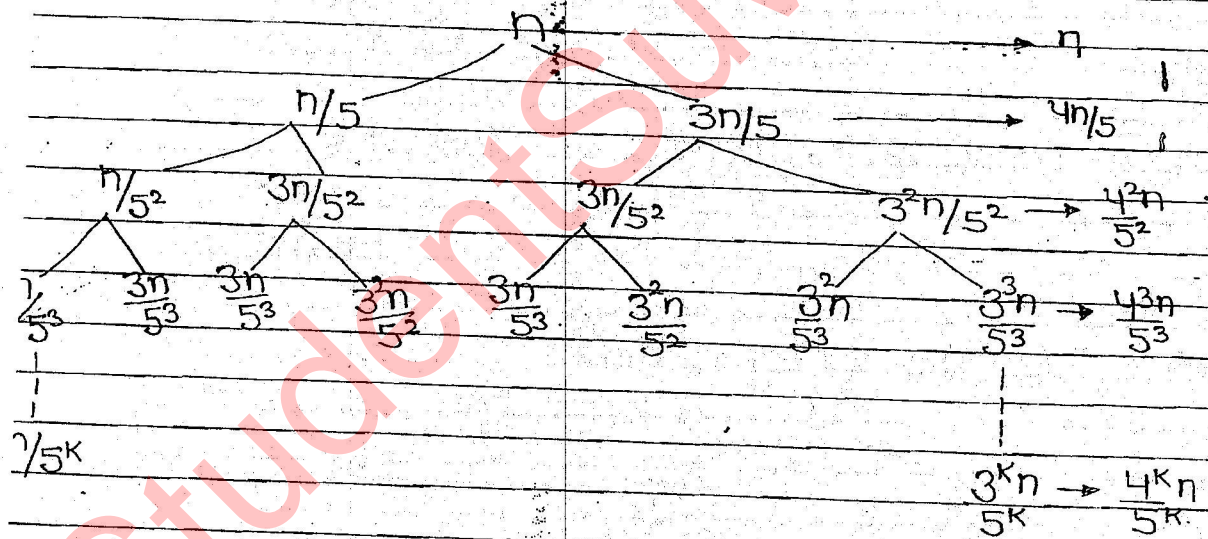
$$T(n) = O(n \log_{3/2} n) \\ T(n) = \Omega(n \log_3 n) \\ = \Theta(n \log n)$$

$$T(n) = T(n/10) + T(9n/10) + n$$



$$T(n) = O(n \log_{10} n), \Omega(n \log_{10} n), \Theta(n \log n)$$

$$T(n) = T(n/5) + T(3n/5) + n$$



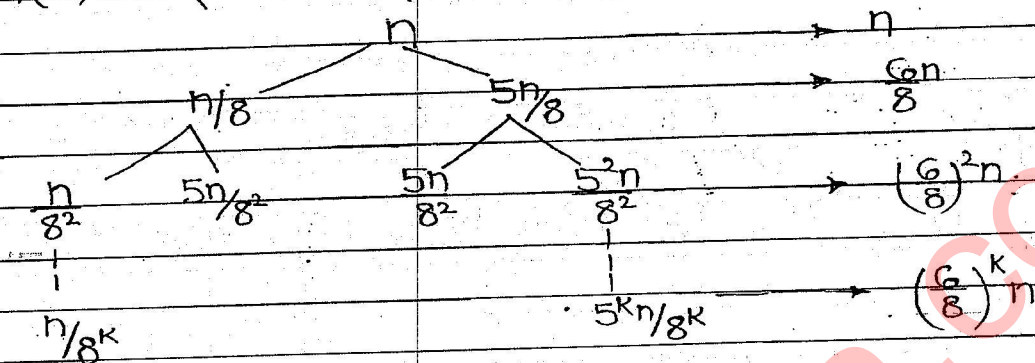
$$n = \left(\frac{5}{3}\right)^k \Rightarrow k = \log_{5/3} n$$

$$T(n) = \left[ n + \frac{4n}{5} + \frac{4^2n}{5^2} + \dots + \frac{4^{\log_{5/3} n} n}{5^{\log_{5/3} n}} \right] \times \log_{5/3} n$$

$$= n \log_{5/3} n \left[ \frac{1 - (4/5)^{\log_{5/3} n}}{1/5} \right] = O(n \log_{5/3} n) = O(n)$$

$$\Omega(n) = \Theta(n)$$

5.  $T(n) = T(n/8) + T(5n/8) + n$

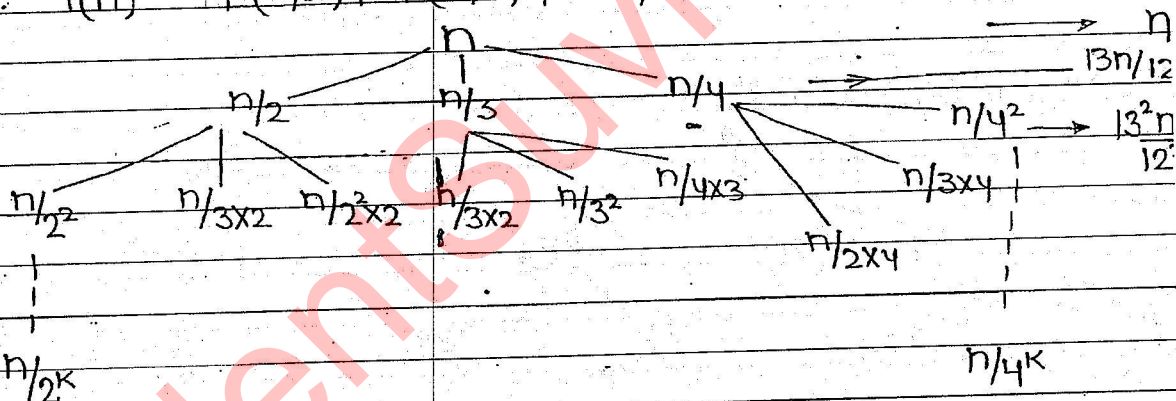


$$T(n) = n \left[ 1 + \frac{2}{4} + \left(\frac{3}{4}\right)^2 + \dots + \left(\frac{3}{4}\right)^{\log_{8/5} n} \right] = O(n)$$

$$= \Omega(n)$$

$$= \Theta(n)$$

6.  $T(n) = T(n/2) + T(n/3) + T(n/4) + n$



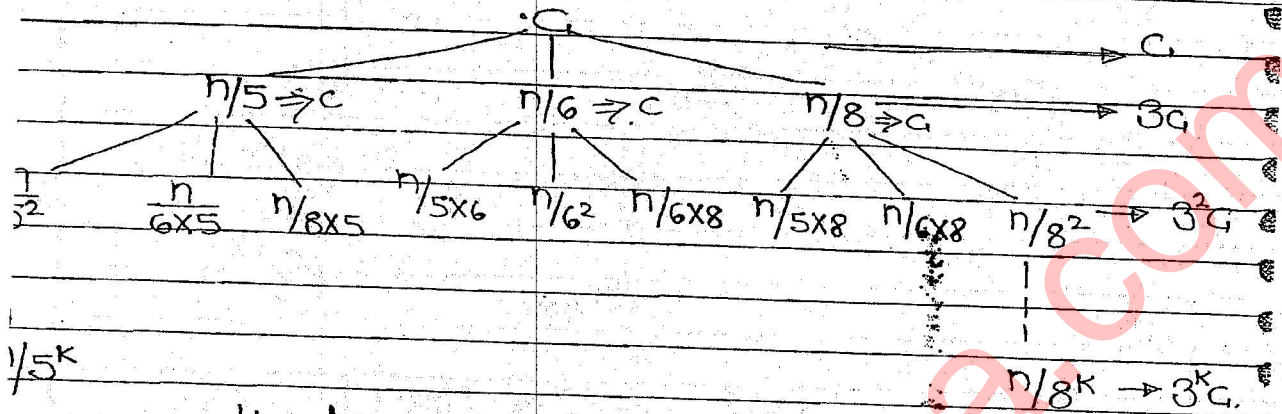
$$T(n) = n \left[ 1 + \frac{13}{12} + \frac{13^2}{12^2} + \dots + \left(\frac{13}{12}\right)^{\log_4 n} \right]$$

$$= n \left[ \frac{1 - \left(\frac{13}{12}\right)^{\log_4 n}}{1/13} \right] = 13n \left[ \left(\frac{13}{12}\right)^{\log_4 n} - 1 \right]$$

$$= 13n \left[ n^{\log_4 13 - \log_4 12} - 1 \right] = 13n \cdot [ - ]$$

$= \Omega( - )$   
for upper bound, take  $\log_2 n$

8.  $T(n) = T(n/5) + T(n/6) + T(n/8) + C.$



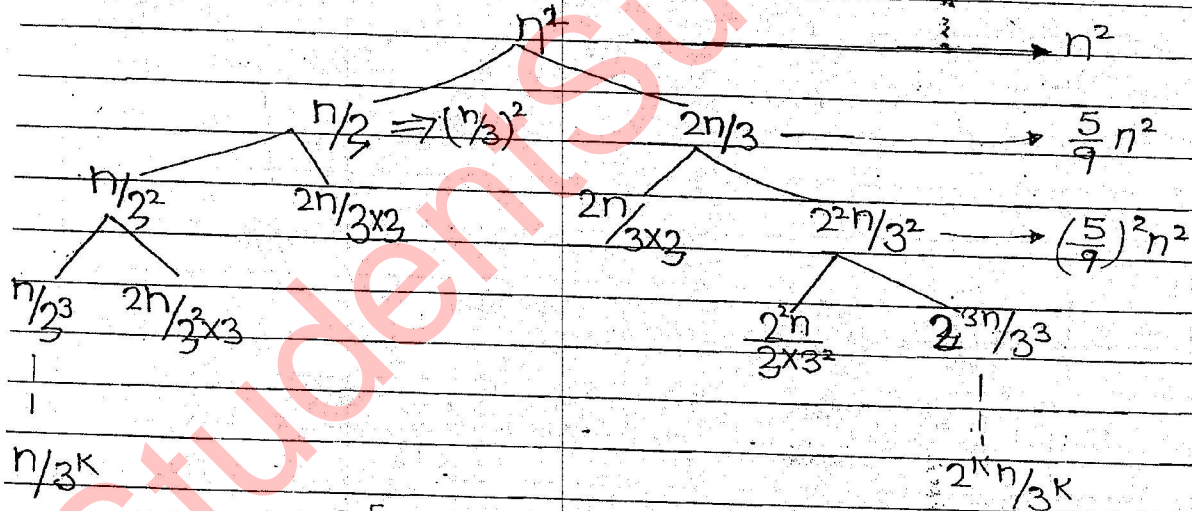
$1/5^k$

$k = \log_5 n$  Upper-bound

$$T(n) = (1 + 3 + 3^2 + \dots + 3^{\log_5 n}) C$$

$$= \frac{C}{2} (3^{\log_5 n} - 1) = O(3^{\log_5 n})$$

9.  $T(n) = T(n/2) + T(2n/3) + n^2$

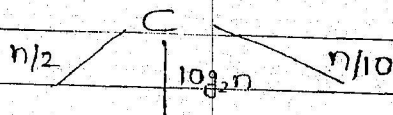


$n/3^k$

$$T(n) = n^2 \left[ 1 + \frac{5}{9} + \frac{5^2}{9^2} + \dots + \left(\frac{5}{9}\right)^{\log_3 n} \right]$$

$$= O(n^2)$$

10.  $T(n) = T(n-2) + T(n/2) + T(n-10) + C$





Best & Average case -

$$T(n) = n + T(n/2) + T(n/2) \Rightarrow O(\log_2 n \cdot n)$$

$$T(n) = n + T(n/5) + T(4n/5) \xrightarrow{\uparrow} \xrightarrow{\uparrow} \xrightarrow{\uparrow} \text{base depend}$$

$$T(n) = n + T(n/10) + T(9n/10) \xrightarrow{\uparrow} \xrightarrow{\uparrow} \text{on } \alpha$$

In general,

$$T(n) = n + T(\alpha n) + T((1-\alpha)n) \quad 0 < \alpha < 1$$

fraction of  $n$  one side, remaining another side

$$\text{base} = \max \left[ \log_{\alpha} n \text{ or } \log_{\frac{1}{1-\alpha}} n \right]$$

$$\text{Stack size} = \max [ \quad ]$$

Worst case -

$$T(n) = n + T(n-1) + T(0) = O(n^2)$$

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

$$= O(n^2)$$

$$T(n) = n + T(9) + T(n-10)$$

$$= \frac{n \times O(n)}{10} = O(n^2) = O(n^2)$$

In general

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

$$= \frac{n \times O(n)}{c} = O(n^2)$$

Constant remains one side & remaining another side

$$\text{Stack size} = O(n/c) = O(n)$$

$$\text{better programming} = O(\log n)$$

In quick sort, the sorting of  $n$  numbers, a pivot the  $n/5$ th element is selected as pivot using  $O(n)$  time (smallest) complexity algorithm. Then what will be time complexity of quick sort in worst case.  
 (a)  $O(n \log n)$  (b)  $O(n^2)$  (c)  $O(n)$  (d)  $O(n^3)$

Solution:  $T(n) = O(1) + O(n) + T(n-1) + T(0)$

$\swarrow$  Pivot selection       $\swarrow$  Partition algorithm       $\downarrow$  Worst case

$$T(n) = O(n) + O(n) + T\left(\frac{n-1}{5}\right) + T\left(\frac{n-n}{5}\right)$$

(given)
Left
Right

$$= 2n + T(n/5) + T(4n/5)$$

(given by problem)

$$= 2n \log_{5/4} n = O(n \log n)$$

In quicksort, the sorting of  $n$  numbers, the  $(n/5)$ th element is selected as pivot using  $O(\log n)$  time complexity. Then what will be best case time complexity of quick sort.  
 i)  $O(n \log n)$  ii)  $O(n^2)$  iii)  $O(n)$  iv)  $O(n^3)$

Pivot selection (not constant) 1 2 3 4 5  $n/5$   $n$

If 1st minimum is selected as pivot, it will go to the first place (its correct position)

But  $(n/5)$ th element can go any-where.

$$T(n) = O(\log n) + O(n) + T(n/5) + T(4n/5)$$

$\underbrace{\hspace{10em}}$ 
  
 Pivot      Partition      Best case (given) (any)

$$= n + T(n/5) + T(4n/5)$$

$$= O(n \log_{5/4} n)$$

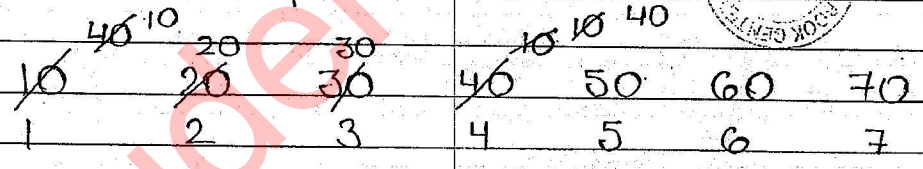
3. In quick-sort, the sorting of  $n$  numbers, the  $(\frac{n}{10})^{\text{th}}$  largest element is taken as pivot using  $O(n^2)$ . What will be the best case time complexity of quick-sort.

$$\begin{aligned}
 T(n) &= O(n^2) + O(n) + T\left(\frac{n-n}{10}\right) + T\left(\frac{n-1}{10}\right) \\
 &= O(n^2) + O(n) + T\left(\frac{n-1}{10}\right) + T\left(\frac{n-n}{10}\right) \\
 &= n^2 + T\left(\frac{n}{10}\right) + T\left(\frac{9n}{10}\right) \\
 &= O(n^2) \quad \text{[By recursive tree method]}
 \end{aligned}$$

The problem is one-way, no other cases.  
Median is nothing but middle person in sorted array  
↳  $(\frac{n}{2})^{\text{th}}$  smallest element

Middle element is  $(\frac{n}{2})^{\text{th}}$  element.  
[Best/Worst case depends on the selection of the pivot]. (middle element as pivot in sorted array)

Randomized Quick-Sort -



$r = \text{RG}(1,7)$   
Swap  $(a[p], a[r])$

Apply partition



$r = \text{RG}(1,3)$

$r = \text{RG}(5,7)$

Swap  $(a[p], a[r])$

Swap  $(a[p], a[r])$

Apply partition

Apply partition.

Even if the array is ~~not~~ sorted, randomized quick sort will not give the worst case.

Selecting pivot element randomly is called randomized quick sort.

Always gives the best / average performance.

Worst case when there is a gap on either side. always (chances are very less)  $O(n^2)$

## VI SELECTION PROCEDURE -

i/p: An array of  $n$  elements and integer  $K$

O/p: find  $K$ th smallest element.

Case I - Apply merge sort to access  $a[K]$

Case II - Apply  $K$  passes of selection sort.

Case III - If array sorted, return  $a[K]$ .

Case IV - Divide & Conquer

Example

$A [50 \ 25 \ 40 \ 70 \ 15 \ 88 \ 92 \ 64 \ 85 \ 80 \ 20 \ 17]$   $K=3$   
 1 2 3 4 5 6 7 8 9 10 11 12

Partition

7  
 $m (25 \ 40 \ 15 \ 35 \ 20) (50) (70 \ 88 \ 92 \ 64 \ 80)$   $K=3$   
 17

Partition

4  
 $m (15, 20) (25) (35 \ 40)$   
 17

Partition

$(15) (20) (17) m \Rightarrow m=3 (17) 20$  return  $a[K]$

Selection (a, k, i, j)

```

{
  if (i == j)
    return a[i] // small
  else {
    m = partition(a, i, j) // n (for partition)
    if (m == k)
      then return a[m] } ~ O(1) // Best case
    else if (m < k)
      Selection(a, k, m+1, j)
    else
      Selection(a, k, i, m-1)
  }
}

```

Recurrence relation - Let  $T(n)$  be the time complexity of above algorithm on  $n$  elements

$$T(n) = \begin{cases} O(1) & \text{if } n=1 \\ T(m-i) + O(1) + O(n) & \text{if } n > 1 \\ \text{or} \\ T(j-m) \end{cases}$$

(Average to worst case)

Best case / Average case

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

$$\Downarrow \\ O(n)$$

Worst case

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

$$\Downarrow \\ O(n^2)$$

1. Better partitions

2. Immediately matching

fraction of  $n$

Stack space  $O(n)$

using better program  $O(\log n)$