

**# Module Name : Algorithms & Data Structures Using Java.**

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**# DAY-01:**

**# Introduction to data structures?**

**Q. Why there is a need of data structures?**

- if we want to store marks of 100 students

int m1, m2, m3, m4, ....., m100; //400 bytes if  
sizeof(int) = 4 bytes

int marks[ 100 ]; //400 bytes - if sizeof(int) = 4 bytes

- we want to store rollno, marks & name

rollno : int  
marks : float  
name : char [] / String / String

```
struct student
{
    int rollno;
    char name[ 32 ];
    float marks;
};
```

```
struct student s1;
```

```
class Employee
{
    //data members
    int empid;
    String empName;
    float salary;
    //member functions/methods
};
```

```
Employee e1;
```

```
Employee e2;
```

=> to learn data structures is not learn any programming language, it is a programming concept i.e. it is nothing but to learn algorithms, and algorithms learned in data structures can be implemented by using any programming language.

**# algorithm to do sum of array elements** => end user  
(human being)

step-1: initially take sum as 0.

step-2: traverse an array and add each array element into sum sequentially

from first element max till last element.

step-3: return final sum.

**# pseudocode** => programmer user

Algorithm ArraySum(A, n)//whereas A is an array of size "n"

```
{
    sum = 0;
    for( index = 1 ; index <= n ; index++ ){
        sum += A[ index ];
    }
    return sum;
}
```

**# program** => compiler => machine

```
int arraySum(int [] arr, int size){
    int sum = 0;
    for( int index = 0 ; index < size ; index++ ){
        sum += arr[ index ];
    }
    return sum;
}
```

Bank Manager => Algorithm => Project Manager => Software Architect

=> Pseudocode => Developers => Program => Machine

Problem : **"Searching"** => to search / to find an element (can be referred as a key element) into a collection/list of elements.

1. Linear Search
2. Binary Search

Problem : **"Sorting"** => to arrange data elements in a collection / list of elements either in an ascending order or in a descending order.

1. Selection Sort
  2. Bubble Sort
  3. Insertion Sort
  4. Merge Sort
  5. Quick Sort
- etc.....

- when one problem has many solutions we need to go for an efficient solution.

City-1:

City-2:

multiple paths exists => optimum path  
distance, condition, traffic situation ....

**- to traverse an array => to visit each array element sequentially from first element max till last element.**

- there are two types of algorithms:

- 1. iterative approach (non-recursive)**
- 2. recursive approach**

- recursion**
- recursive function**
- recursive function call**
- tail-recursive function**
- non-tail recursive function**

Class Employee

```
{  
    int empid;  
    String name;  
    float salary;
```

```
};
```

- object e1 is an instance of Employee class

```
Employee e1(1, "sachin", 1111.11);
```

```
Employee e2(2, "nilesh", 2222.22);
```

```
Employee e3(3, "amit", 3333.33);
```

### # Space Complexity:

for size of an array = 5 => index = 0 to 5 => only 1 mem copy of index = 1 unit

for size of an array = 10 => index = 0 to 10 => only 1 mem copy of index = 1 unit

.

.

for size of an array = n => index = 0 to n => only 1 mem copy of index = 1 unit

for any input size array we require only 1 memory copy of index var =>

simple var

+ sum:

for size of an array = 5 => sum => only 1 mem copy of sum = 1 unit

for size of an array = 10 => sum => only 1 mem copy of sum = 1 unit

.

.

for any input size array we require only 1 memory copy of sum var => simple var

+ n = input size of an array -> instance characteristics of an algo

for size of an array = 5 => if n = 5 => 5 memory copies required to store 5 ele's => 5 units

for size of an array = 10 => if n = 10 => 10 memory copies required to store 10 ele's => 10 units

for size of an array = 20 => if n = 20 => 20 memory  
copies required to store 20 ele's => 20 units

for size of an array = 100 => if n = 100 => 100 memory  
copies required to store 100 ele's => 100 units

size

- for any input size array no. of instructions inside an  
algo remains fixed i.e. space required for instructions  
i.e. code space for any size array will going to remains  
fixed or constant.

```
int sum( int n1, int n2 )//n1 & n2 are formal params
{
    int res;//local var
    res = n1 + n2;
    return res;
}
```

- When any function completes its execution control goes  
back into its calling function as an addr of next  
instruction to be executed in its calling function gets  
stored into FAR of that function as a "**return addr**".

**FAR contains:**

**local vars**

**formal params**

**return addr** => addr of next instruction to be executed in  
its calling function

**old frame pointer** => an addr of its prev stack frame/FAR.  
etc...

## # Linear Search:

```
for( index = 1 ; index <= n ; index++ ){  
    if( key == arr[ index ] )  
        return true;  
}  
  
return false;
```

- In Linear Search best case "if key found at very first position"

for size of an array = 10, no. of comparisons = 1

for size of an array = 20, no. of comparisons = 1

for size of an array = 50, no. of comparisons = 1

.  
.

for any input size array => no. of comparisons = 1

Running Time =>  $O(1)$

- In Linear Search worst case "if either key found at last first position or key does not exists"

for size of an array = 10, no. of comparisons = 10

for size of an array = 20, no. of comparisons = 20

for size of an array = 50, no. of comparisons = 50

.  
.

for size of an array =  $n$ , no. of comparisons =  $n$

Running Time =>  $O(n)$

Lab Work => Implement Linear Search => by non-rec as well  
rec way

# DAY-02:

- linear search
- binary search
- comparison between searching algo
- sorting algorithms:

basic sorting algo's : selection, bubble & insertion

**assumption-1:**

**if running time of an algo is having any additive / subtractive / divisive / multiplicative constant then it can be neglected.**

e.g.

$O(n + 3) \Rightarrow O(n)$

$O(n - 2) \Rightarrow O(n)$

$O(n / 5) \Rightarrow O(n)$

$O(6 * n) \Rightarrow O(n)$

# Binary Search:

by means of calculating mid position big size array gets divided logically into two subarray's:

**left subarray => left to mid-1**

**right subarray => mid+1 to right**

**for left subarray => value of left remains same, whereas value of right = mid-1**

**for right subarray => value of right remains same, whereas value of left = mid+1**

best case occurs in binary search if key is found in very first iteration in only 1 comparison.

if size of an array = 10, no. Of comparisons = 1

if size of an array = 20, no. Of comparisons = 1

.

.

for any input size array, no. Of comparisons = 1

=> running time =>  $O(1)$

- in this algo, **in every iteration 1 comparison takes place and search space gets divided by half** i.e. array gets divided logically into two subarray's and in next iteration we will search key either into left subarray or into right subarray.

**# Substitution method to calculate time complexity of binary search:**

Size of an array = 1000

$1000 = n$

$500 = n/2$

$250 = n/4$

$125 = n/8$

.  
. .

$n/2 / 2 \Rightarrow n / 4$

$n/4 / 2 \Rightarrow n / 8$

.  
.

for iteration-1 input size of an array  $\Rightarrow n$

after iteration-1:  $n/2 + 1 \Rightarrow n / 2^1 + 1$

after iteration-2:  $n/4 + 2 \Rightarrow n / 2^2 + 2$

after iteration-3:  $n/8 + 3 \Rightarrow n / 2^3 + 3$

.  
. .

after iteration-k:  $n / 2^k + k$  .... eq-I

**$T(n) = n / 2^k + k$  .... eq-I**

**lets assume,  $n = 2^k$**

**$n = 2^k$**

**$\log n = \log 2^k$  .... [ by taking log on both sides ]**

**$\log n = k \log 2$**



$\log n = k \quad \dots [ \text{as } \log 2 \approx 1 ]$   
 $k = \log n$

put value of  $n = 2^k$  and  $k = \log n$  in eq-I, we get

$T(n) = n / 2^k + k$   
 $\Rightarrow T(n) = 2^k / 2^k + \log n$   
 $\Rightarrow T(n) = 1 + \log n$   
 $\Rightarrow T(n) = O(1 + \log n)$   
 $\Rightarrow T(n) = O(\log n + 1)$   
 $\Rightarrow \underline{T(n) = O(\log n)}$ .

1. Selection Sort:

total no. of comparisons =  $(n-1) + (n-2) + (n-3) + \dots + 1$   
 $\Rightarrow n(n-1) / 2$

hence

$\Rightarrow T(n) = O(n(n-1) / 2)$   
 $\Rightarrow T(n) = O((n^2 - n) / 2)$   
 $\Rightarrow T(n) = O(n^2 - n)$   
 $\Rightarrow \mathbf{T(n) = O(n^2)}$

**assumption:**

if running time of an algo is having a polynomial then in its time complexity only leading term will be considered.  
e.g.

$O(n^3 + n^2 + 5) \Rightarrow O(n^3)$ .

**assumption:**

if an algo contains a nested loops and no. of iterations of outer loop and inner loop don't know in advanced then running time of such algo will be whatever time required for statements which are inside inner loop.

**for( i = 0 ; i < n ; i++ ){**

**for( j = 0 ; j < n ; j++ ){**

**statement/s  $\Rightarrow n*n$  no. of times  $\Rightarrow O(n^2)$  times**

**}**

**}**

**+ features of sorting algorithms:**

**1. inplace** => if a sorting algo do not takes extra space (i.e. space required other than actual data ele's and constant space) to sort data elements in a collection/list of elements.

**2. adaptive** => if a sorting algorithm works efficiently for already sorted input sequence then it is referred as an adaptive.

**3. stable** => if in a sorting algorithm, relative order of two elements having same key value remains same even after sorting then such sorting algorithm is referred as stable.

Input array => 10 40 20 30 10' 50

After Sorting:

Output => 10 10' 20 30 40 50 => stable

Output => 10' 10 20 30 40 50 => not stable

**# Design & Analysis of an Algorithm By Coreman**

**2. Bubble Sort:**

### # DAY-03:

#### 3. Bubble Sort:

total no. of comparisons =  $(n-1) + (n-2) + (n-3) + \dots + 1$

$\Rightarrow n(n-1) / 2$

hence

$\Rightarrow T(n) = O(n(n-1) / 2)$

$\Rightarrow T(n) = O((n^2 - n) / 2)$

$\Rightarrow T(n) = O(n^2 - n) \dots$  [as 2 is a divisive constant it can be neglected]

$\Rightarrow \mathbf{T(n) = O(n^2)}$  ....

for it=0  $\Rightarrow$  pos=0,1,2,3,4

for it=1  $\Rightarrow$  pos=0,1,2,3

for it=2  $\Rightarrow$  pos=0,1,2

for it=3  $\Rightarrow$  pos=0,1

for( pos = 0 ; pos < arr.length-1-it ; pos++ )

Best Case : if array elements are already sorted

flag = false

iteration-0:

10 20 30 40 50 60

10 20 30 40 50 60

10 20 30 40 50 60

10 20 30 40 50 60

10 20 30 40 50 60

if all pairs are in order => array is already sorted =>  
no need of swapping => no need to goto next iteration

in best case only 1 iteration takes places and in  
iteration total (n-1) no. Of comparisons takes place

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

**$T(n) = O(n)$ .**

3. Insertion Sort:

```
for( i = 1 ; i < SIZE ; i++ ){//for loop for iterations
    key = arr[ i ];
    j = i-1;
```

```
    /* if index is valid && compare value of key with an
    ele at that index */
```

```
    while( j >= 0 && key < arr[ j ] ){
        //shift ele towards its right hand side by 1 pos
        arr[ j+1 ] = arr[ j ];
        j--;//goto prev ele
    }
```

```
    //insert key at its appropriate pos
    arr[ j+1 ] = key;
```

}

Best Case:

Iteration-1:

10 20 30 40 50 60

10 20 30 40 50 60

no. of comparisons = 1

Iteration-2:

10 20 30 40 50 60

10 20 30 40 50 60

no. of comparisons = 1

Iteration-3:

10 20 30 40 50 60

10 20 30 40 50 60

no. of comparisons = 1

Iteration-4:

10 20 30 40 50 50

10 20 30 40 50 50

no. of comparisons = 1

Iteration-5:

10 20 30 40 50 50

10 20 30 40 50 50

no. of comparisons = 1

in best case total  $(n-1)$  no. of iterations are required  
and in each iteration only 1 comparison takes place  
total no. Of comparisons =  $1 * (n-1) \Rightarrow n-1$   
 $T(n) = O(n - 1)$   
 $T(n) = O(n) \Rightarrow \Omega(n)$ .

#### + **Linked List:**

#### **Q. Why Linked List ?**

##### + **Limitations of an array data structure:**

1. in an array we can collect/combine logically related only similar type of data element  $\Rightarrow$  to overcome this limitation **structure** data structure has been designed.

2. array is **static** i.e. size of an array is fixed , its size cannot be either grow or shrink during runtime.

**int arr[ 100 ];**

3. **addition & deletion** on an array are not efficient as it takes  **$O(n)$**  time.

- while adding ele into an array we need to shift elements towards its right hand side by one-one pos till depends on size of an array, whereas while deleting an ele from an array we need to shift elements towards its left hand side by one-one pos till depends on size of an array

- to overcome 2<sup>nd</sup> & 3<sup>rd</sup> limitations of an array data structure **linked list** data structure has been designed.

- **linked list must be dynamic**

- **addition & deletion operations on linked list must be performed efficiently i.e. expected in  $O(1)$  time.**

Q. What is a Linked List ?

Linked List is a basic/linear data structure, which is a collection/list of logically related similar type of data elements in which an addr(ref) of first element in it always kept into a pointer/ref referred as head, and each element contains actual data and an addr/link of its next element (as well as link of its prev element) in it.

- Element in a Linked List is also called as a node.
- Basically there are 2 types of linked list:
  1. singly linked list : it is a type of linked list in which each node in it contains link to its next node only i.e. in each node no. of links = 1
    - there are 2 subtypes of sll:
      - i. singly linear linked list
      - ii. singly circular linked list
  2. doubly linked list : it is a type of linked list in which each node in it contains link to its next node as well as link to its prev node i.e. in each node no. of links = 2
    - there are 2 subtypes of dll:
      - i. doubly linear linked list
      - ii. doubly circular linked list

### **i. singly linear linked list =>**

- on linked list we can perform basic 2 operations

**1. addition : to add/insert node into the linked list**

**2. deletion : to delete/remove node from the linked list**

**1. addition : to add/insert node into the linked list:**

- we can add node into the linked list by 3 ways:

**i. add node into the linked list at last position**

**ii. add node into the linked list at first position**

**iii. add node into the linked list at specific position (in between position).**

**i. add node into the linked list at last position:**

- we can add as many as we want number of nodes into the **slll** at last position in  **$O(n)$  time.**

Best Case :  $\Omega(1)$  - if list is empty

Worst Case :  $O(n)$

Average Case :  $\Theta(n)$

- to traverse a linked list => to visit each node in a linked list sequentially from first node max till last node.

- we can start traversal from first node and we get an addr/ref of first node always from head

**ii. add node into the linked list at first position:**

- we can add as many as we want number of nodes into the **slll** at first position in  **$O(1)$  time.**

Best Case :  $\Omega(1)$  - if list is empty

Worst Case :  $O(1)$

Average Case :  $\Theta(1)$



**iii. add node into the linked list at specific position:**

- we can add as many as we want number of nodes into the **slll** at first position in  **$O(n)$  time.**

Best Case :  $\Omega(1)$  - if pos = 1

Worst Case :  $O(n)$  - if pos is last position

Average Case :  $\Theta(n)$

- In a Linked List Programming remember one rule:  
**make before break => always create new links (links which are associative with newly created node) first and then only break old links.**

**Lab Work : Convert SLLL program into a menu driven program.**