

Batch Name :
Subject Name : Data Structure

DS DAY-01:

Q. Why there is a need of data structure in programming?

There is a need of data structures in programming to achieve efficiency in operations.

Q. What is a data structure?

It is a way to store data elements into the memory (i.e. into the main memory) in an **organized manner** so that operations like **addition, deletion, searching, sorting, traversal etc....** can be performed on it efficiently.

- we want to store marks 100 students:

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

- we want to arrange marks in a descending order => sorting

int arr[100]; //400 bytes

+ Array:

An array is **linear/basic data structure**, which is a **collection/ list of logically related similar type of data elements** gets stored into the memory at **contiguous locations**.

int arr[5] ;

- in an array, to convert array notation into its eq pointer notation is done by the compiler,

i.e. to maintain link between array elements in an array is the job of compiler.

- we want to store records of 100 employees:

empid : int
name : char [] / string
salary : float

- in an array we can collect/combine logically related similar type of data elements only, and hence to overcome this limitation structure data structure has been designed.

+ Structure:

It is a **linear / basic data structure**, which is a **collection / list of logically related similar and dissimilar type of data elements** gets stored into the memory collectively as a **single entity / record**.

```
struct employee
{
    int empid;//4 bytes
    char name[ 32 ];//32 bytes
    float salary;//4 bytes
};
```

sizeof structure = sum of size of all its members
sizeof(struct employee): 40 bytes

- there are 2 types of data types:

1. primitive/predefined/builtin data type: char, int, float, double, void
=> data types which are already known to the compiler

2. non-primitive / derived / user defined data type : pointer, array, structure, union, enum, function etc...
=> data types which are not already known to the compiler, i.e. need to define/derived by the programmer

user defined data type:
typedef

```
typedef int INT;//user defined => not a derived data type
typedef int bool_t;//user defined => => not a derived data type
typedef struct employee emp_t;//derived / user defined
int *iptr;//derived data type
```

- there are two types of data structures:

1. linear / basic data structures: data structures in which data elements gets stored into the memory in a linear manner/linearly and hence can be accessed linearly (i.e. one after another).

- array
- structure & union
- linked list
- stack
- queue

2. non-linear / advanced data structures: data structures in which data elements gets stored into the memory in a non-linear manner (e.g. hierarchical manner), and hence can be accessed non-linearly.

- tree
 - graph
 - binary heap
 - hash table
- etc....

structure is a derived data type =>

```
struct employee
{
    int empid;//4 bytes
    char name[ 32 ];//32 bytes
    float salary;//4 bytes
};
=> compiler
```

```
<data type> <var_name>;
int num;
```

```
struct employee emp;
```

- to learn data structures, is not to learn any programming language, it is nothing but to learn algorithms, data structure algorithms can be implemented in any programming language (C / C++ / Java / Python).

Q. What is a Program? => Machine

- A Program is a finite set of instructions written in any programming language (i.e. either in low level / high level) given to the machine to do specific task.

Q. What is an algorithm? => User/Human Beings

- An algorithm is a finite set of instructions written in any human understandable language like english, if followed, accomplishes given task.

- Program is an implementation of an algorithm.

- An algorithm is like a blue print / design of a program on paper.

Blue-print => implementation => building

- Pseudocode is a special form of an algorithm for programmer user.

Q. What is a pseudocode? => Programmer User

- An algorithm is a finite set of instructions written in any human understandable language like english with some programming constraints, if followed, accomplishes given task, this kind/form of an algorithm is referred as a pseudocode.

- an algorithm to do sum of array elements:

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

algorithm:

step-1: initially take sum as 0.

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

step-3: return final sum.

Pseudocode: => Programmer

Algorithm ArraySum(arr, n){//arr is an array of size n

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

Program: => Machine

int array_sum(int arr[], int size){

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

Example:

IT Industry:

Client (Algorithm i.e. Requirement) => Software Architect/Tech Manager

Software Architect (Pseudocode) => Software Developer

Software Developer (Program) => Machine.

- an algorithm is a solution of a given problem.
- an algorithm = solution
- one problem may has many solutions

e.g.

searching => to search/find an element (let say referred as key element), in a given collection/list/set of elements.

1. linear search
2. binary search

sorting => to arrange data elements in a collection/list of elements wither 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.....

- if one problem has many solutions, we need to select an efficient solution out of them, and to decide which solution/algorithm is an efficient one, we need to do their analysis.

- **analysis of an algorithm** is a work of calculating/determining how much **time** i.e. computer time and **space** i.e. computer memory it needs to run to completion.

- there are 2 measures of analysis of an algorithm:

1. **time complexity** of an algorithm is the amount of **time** i.e. **computer time** it needs to run to completion.

2. **space complexity** of an algorithm is the **amount of space** i.e. **computer memory** it needs to run to completion.

1. **linear search / sequential search:**

algorithm:

step-1: accept key from user (key = element which is to be search)

step-2: start traversal of an array from first element and compare value of key element with each array element sequentially till match is not found or max till last element.

step-3: if the value of key matches with any of the array element then return true other wise return false.

pseudocode:

Algorithm LinearSearch(A, key, n)

```
{  
    for( index = 1 ; index <= n ; index++ ) {  
        if( key == A[ index ] )//if key matches with any array ele  
            return true;//key is found  
    }  
  
    //if key do not matches with any of array element  
    return false;//key is not found  
}
```

best case occurs : if key is found at first position => $O(1)$

if size of an array = 10 => no. of comparisons = 1

if size of an array = 20 => no. of comparisons = 1

if size of an array = 50 => no. of comparisons = 1

if size of an array = 100 => no. of comparisons = 1

.

.

.

if size of an array = n => no. of comparisons = 1

worst case occurs : if either key is found at last position or key does not exist : $O(n)$.

if size of array = 10 \Rightarrow no. of comparisons = 10

if size of array = 20 \Rightarrow no. of comparisons = 20

if size of array = 100 \Rightarrow no. of comparisons = 100

.

.

if size of array = $n \Rightarrow$ no. of comparisons = n

best case time complexity = if an algo takes min amount of time to run to completion.

worst case time complexity = if an algo takes max amount of time to run to completion.

average case time complexity = if an algo takes neither min nor max amount of time to run to completion.

+ Asymptotic analysis: it is a **mathematical** way to calculate time complexity and space complexity of an algorithm without implementing it in any programming language.

- in this kind of analysis, focus is on basic operation in that algorithm

e.g. searching \Rightarrow basic operation is comparison, and hence analysis can be done depends on no. of comparisons takes places in different cases.

sorting \Rightarrow basic operation is comparison, and hence analysis can be done depends on no. of comparisons takes places in different cases.

Addition of matrices \Rightarrow basic operation addition, and hence analysis can be done depends on number addition instructions.

- there are some notations and few assumptions that we need to follow:

there are 3 asymptotic notations:

1. **Big Omega (Ω)** - this notation is used to represent **best case time complexity**.

- big omega is referred as asymptotic lower bound

- running of an algo should not be less than its **asymptotic lower bound**.

2. **Big Oh (O)** - this notation is used to represent **worst case time complexity**

- big oh is referred as asymptotic upper bound.

- running of an algo should not be greater than its asymptotic upper bound.

3. **Big Theta (θ)** - this notation is used to represent an **average case time complexity**.

- big theta is referred as asymptotic tight bound.

Assumption:

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

e.g.

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

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

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

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

- if an algo follows divide-and-conquer approach then we get time complexity in terms of log.

DS DAY-02:

2. Binary Search

algorithm:

step-1: accept key from user

step-2:

- calculate mid pos by the formula $\Rightarrow \text{mid} = (\text{left} + \text{right}) / 2$

- by means of calculating mid position, big size array gets divided logically into two subarray's \Rightarrow left subarray & right subarray

- left subarray is from left to mid-1, and right subarray is from mid+1 to right.

for left subarray \Rightarrow value of left remains as it is, value of right = mid-1

for right subarray \Rightarrow value of right remains as it is, value of left = mid+1

step-3: compare value of key with ele at mid pos, if key matches with ele at mid pos return true

step-4: if key do not matches then search key either into the left subarray or into the right subarray

step-5: repeat step-2, step-3 & step-4, till either key not found, or max till subaarray is valid, if subarray is invalid then return false indicates key not found.

if(left <= right) \Rightarrow subarray is valid

if(left > right) \Rightarrow subarray is invalid

- in a binary tree, any element is at either one of the following 3 positions:

1. root pos

2. leaf pos

3. non-leaf pos

root node/root pos \Rightarrow first pos

node which is not having further child node \Rightarrow leaf node

node which is having child node \Rightarrow non-leaf node

if key is found at root pos => best case => $O(1)$
no. of comparisons for input size array = 1
time complexity = $\Omega(1)$.

if key is found at non-leaf pos => average case
 $O(\log n)$
time complexity = $\theta(\log n)$

if either key is found at leaf pos or key is not found => worst case
 $O(\log n)$
time complexity = $O(\log n)$

+ Sorting Algorithms:

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

- when we say sort array elements, by default we need to sort array elements in an ascending order.

1. Selection Sort:

for size of an array is n ,

in iteration-1: no. of comparisons = $n-1$

in iteration-2: no. of comparisons = $n-2$

in iteration-3: no. of comparisons = $n-3$

.

.

iterations ($n-1$):

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

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

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

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

$\Rightarrow O(n^2)$

rule/assumption: if running time of an algo is having a polynomial, then in its time complexity only leading term will be considered.

e.g.

$O(n^3 + n + 4) \Rightarrow O(n^3)$

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

$O(n^3 + n^2 + n - 3) \Rightarrow O(n^3)$

DS DAY-03:
2. Bubble Sort:

for size of an array is n,
in iteration-1: no. of comparisons = n-1
in iteration-2: no. of comparisons = n-2
in iteration-3: no. of comparisons = n-3
.
.

iterations (n-1):

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

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

```
for( pos = 0 ; pos < SIZE-it-1 ; pos++ )  
for it=0; pos=0,1,2,3,4  
for it=1; pos=0,1,2,3  
for it=2; pos=0,1,2  
.  
.
```

best case:
input array \Rightarrow 10 20 30 40 50 60
flag= 0

for it=0
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

for it=1
iteration-1: all next iterations will be skipped

- if all pairs are in order
 \Rightarrow if there is no need of swaaping in first iteration, it means all pairs are already in order \Rightarrow array is already sorted \rightarrow no need to go to next iteration and we are reducing no. of comparisons at an implementation level.

=> array ele's are already sorted

- in best case this algo takes only one iteration and

total no. of comparisons = $n-1$

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

best case time complexity = $\Omega(n)$.

3. Insertion Sort:

```
for( i = 1 ; i < SIZE ; i++){  
    j = i-1;  
    key = arr[ i ];  
  
    while( j >= 0 && key < arr[ j ] ){  
        arr[ j+1 ] = arr[ j ]; //shift ele towards its right by 1  
        j--; //goto the pre element  
    }  
  
    //insert key into the left hand i.e. into an array at its appropriate pos  
    arr[ j+1 ] = key;  
}
```

best case : if array ele's are already sorted:

iteration-1: no. of iterations=1

10 20 30 40 50 60

10 20 30 40 50 60

iteration-2: no. of iterations=1

10 20 30 40 50 60

10 20 30 40 50 60

iteration-3: no. of iterations=1

10 20 30 40 50 60

10 20 30 40 50 60

iteration-4: no. of iterations=1

10 20 30 40 50 60

10 20 30 40 50 60

iteration-5: no. of iterations=1

10 20 30 40 50 60

10 20 30 40 50 60

insertion sort algo takes max $(n-1)$ no. of iterations, and in each iteration only one comparison takes place, and hence
total no. of comparisons = $1 * (n-1) = n-1$
 $T(n) = O(n-1) \Rightarrow O(n) \Rightarrow \Omega(n)$.

4. Merge Sort:

- this algo follows divide-and-conquer approach
- there are 2 steps in this algo:

step-1: divide big size array logically into smallest size subarray's i.e. subarray having size 1.

- in each iteration by means of calculating mid pos, we are dividing big size array logically into two subarray's, left subarray will be from left to mid, and right subarray will be from mid+1 to right.

for left subarray, value of left remains as it is and right = mid
for right subarray, value of right remains as it is and left = mid+1

step-2: merge two already sorted subarray's into a single array in a sorted manner.

- if subarray contains only 1 ele \Rightarrow subarray is already sorted.

DS DAY-04:

5. Quick Sort:

- this algo follows divide-and-conquer approach.
- the basic logic of this algo is partitioning

partitioning:

step-1: to select pivot element (in an array we can select either **leftmost** or rightmost or middlemost element as a pivot element)

step-2: shift all ele's which are smaller than pivot towards its left and shift all ele's which are greater than pivot towards its right, by means of this shifting pivot element gets fixed at its appropriate pos and big size array gets divided logically into two partitions, **left partition & right partition**.

step-3: apply partitioning on left partition as well as right partition till the size of partition is greater than 1.

worst case(quick Sort): if either array elements are exists in already sorted manner or exactly in a reverse order $\Rightarrow O(n^2)$
- rarely occurs.

SunBeam