

Course Name : PG DAC
Module Name : Algorithms & Data Structures Using Java.

DAY-01:

Introduction:

Q. Why there is a need of data structures?

Q. What is data structures?

=> To store marks of 100 students:

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

int marks[100]; //400 bytes

=> Array : it is a basic/linear data structure, which is a collection/list of logically related similar type of elements gets stored into the memory at contiguous locations.

Q. Why array indexing starts from 0?

- to convert array notation into its equivalent notation is done by the compiler, (to maintain link between array elements is the job of compiler).

arr[i] ~* (arr + i)

struct student

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

<data type> <var_name>;

- data type may be any primitive/non-primitive data type

- var_name is an identifier

e.g.

int n1;

struct student s1; //abstraction => abstract data type

struct student s2;

+ class => it is a linear/basic data structure which is a collection/list of logically related similar and dissimilar type of data elements referred as data members as well as functions which can be used to perform operations on data members referred as member function/methods.

e.g.

```

class student
{
    //data members
    private int rollno;
    private String name;
    private float salary;

    //methods:
    //ctor
    //mutators
    //getter functions
    //setter functions
    //facilitators
}

```

student s1;//ADT

** to learn data structures is not to learn any specific programming language, it is nothing but to learn an algorithms, and algorithms can be implemented by using any programming language (using Java).

Q. What is an algorithm?

+ traversal of an array => to visit each array element sequentially from first element max till last element.

- Algorithm to do sum of array elements: => Any User

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 value of sum.

- Pseudocode to do sum of array elements: => Programmer User

```

Algorithm ArraySum(A, size){
    sum = 0;
    for( index = 1 ; index <= size ; index++ ){
        sum += A[ index ];
    }
    return sum;
}

```

- Program to do sum of array elements: => Machine

```

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

```

Bank Project => Bank Manager => Algorithm => Project Manager
=> Software Architect => Design Pseudocode => Developers =>
Programs => Machine

Q. What is a recursion ?

- it is a process in which function can be called within itself,
such a function is referred as recursive function.

- function call for which calling function and called function are
same, is referred as recursive function call.

- any thing can be defined in terms itself

```
main( ){  
    print("sum = "+recArraySum(arr, 0) );//first time function  
calling to the rec function  
    //calling function => main()  
    //called function => recArraySum()  
}
```

```
int recArraySum(int [] arr, int index ){  
    if( index == arr.length )  
        return 0;
```

```
    return ( arr[ index ] + recArraysum(arr, index+1) );  
    //calling function => recArraySum()  
    //called function => recArraySum()  
}
```

```
}
```

- to delete function activation record / stack frame from stack
called as stack cleanup is done either by calling function or
called function and it depends on function calling convention.

```
main(){  
    print("sum"+sum(10, 20);  
    //10 & 20 => actual params  
}
```

```
int sum(int n1, int n2){  
    int sum;  
    sum = n1 + n2;  
    return sum;  
}
```

```
//n1 & n2 => formal params  
//sum => local var
```

+ function calling conventions:

__std__
__c__
__pascal

function calling conventions decides 2 things:

1. in which order params should be passed to the function
i.e. either from L → R Or R → L
2. stack cleanup

- when any function gets called an OS creates one entry onto the stack for that function call, called as function activation record/stack frame and it gets popped or removed from the stack when an execution of that function is completed and process to remove stack frame from stack is called as stack cleanup.

City-1:

City-2:

- there may exist multiple paths between 2 cities, in this case we need to decide an optimum/efficient path
- there are some factors/measures on which efficient/optimum path can be decided:

- time
- distance
- cost
- traffic condition
- status of path
- etc...

Space Complexity:

Space Complexity = Code Space + Data Space + Stack Space

Code Space => space required for an instructions

Data Space => space required for simple vars, constants and instance vars

Stack Space (applicable only in recursive algorithm) => space required for FAR's.

- there are components of space complexity:

1. fixed component : code space + data space (space required for simple vars and constants).

2. variable component : data space (space required for instance vars) & stack space (applicable only in recursive algorithms).

Example:

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

$S = C \text{ (Code Space)} + S_p \text{ (Data Space)}$

Code Space =>

if size of an array = 5 => no. of instructions will be same
if size of an array = 10 => no. of instructions will be same
if size of an array = 100 => no. of instructions will be same

.

.

if size of an array = n => no. of instructions will be same

for any input size array no. of instructions in an algo will going to remains same => it will take constant amount of space for any input size array.

$S_p = \text{space required for simple vars} + \text{space required for constants}$
 $+ \text{space required for instance vars}$

simple vars: A, sum, index => 3 units

1 unit of memory => A

1 unit of memory => sum

1 unit of memory => index

instance var => n =>

for size of an array is 5 i.e. n = 5 => 5 units

for size of an array is 10 i.e. n = 10 => 10 units

for size of an array is 100 i.e. n = 100 => 100 units

.

.

for size of an array is n => n units => instance var

```
index++ => index = index + 1;
```

- quick sort
- merge sort
- implementation of advanced data structures algo's like - traversal methods in tree.

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

- for any input size array, no. of instructions will be remain same, hence compilation time also remains same for any input size array.

Asymptotic Analysis:

Searching => to search/find key element in a given collection/list of elements.

- there are basic two searching algorithms:

1. Linear Search:
2. Binary Search:

1. Linear Search/ Sequential Search:

```
Algorithm LinearSearch(A, n, key){ //A is an array of size n  
    for( index = 1 ; index <= n ; index++ ){  
        if( key == A[ index ] )  
            return true;  
    }  
    return false;  
}
```

Best Case : if key is found in an array at first position
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 = n => no. of comparisons = 1

for any input size array no. of comparisons in best case = 1 and hence in this case linear search algo takes $O(1)$ time.

Worst Case : if either key is found in an array at last position or key does not exists.

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

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

.

.

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

in worst case no. of comparisons depends on an input size of an array, hence running time of linear search algo in worst case is $O(n)$.

Rule:

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

e.g.

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

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

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

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

Home Work : to implement Linear Search \Rightarrow by using recursion as well as non-recursive method.

DAY-01:

- introduction to ds:

why there is a need of ds ?

what is a ds and its types

what is an algorithm, program, pseudocode

types of algorithms: iterative/non-recursive and recursive

what is a recursion? recursive function, rec function call

types of rec functions: tail & non-tail recursive

analysis of an algo: space complexity & time complexity

asymptotic analysis \Rightarrow it is a mathematical way to calculate time complexity and space complexity of an algo without implementing it in any programming language.

DAY-02:

2. Binary Search:

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

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

best case : if key is found in an array in very first iteration
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 = 100 => no. of comparisons = 1

.

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

for any input size array, in best case no. Of comparisons = 1,
hence running time of an algo in best case = $O(1)$
time complexity of binary search in best case => $\Omega(1)$.

if(left <= right) => subarray is valid
OR
if(left > right) => subarray is invalid

n = 1000

iteration-1:
search space = 1000 = n

mid=500
[0..... 499] 500 [501 1000] => no. Of comparisons=1

iteration-2:
search space = 500 = n/2

[0...249] 250 [251 ... 499] => no. Of comparisons=1

iteration-3:
search space = 250 => n / 4
[0 ...124] 125 [126 ... 250] => no. Of comparisons=1

.

if size of an array = 1 => trivial case => $T(1) = O(1)$

if size of an array > 1 i.e. size of an array = n

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

after iteration-1:

$$T(n) = T(n/2) + 1 \Rightarrow T(n/2^1) + 1$$

after iteration-2:

$$T(n) = T(n/4) + 2 \Rightarrow T(n/2^2) + 2$$

after iteration-3:

$$T(n) = T(n/8) + 3 \Rightarrow T(n/2^3) + 3$$

.

lets assume, k no. of iterations takes place

after k iterations:

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

lets assume kth iteration is the last iteration

assume $\Rightarrow n \Rightarrow 2^k$

$$\Rightarrow n \Rightarrow 2^k$$

$$\Rightarrow \log n = \log 2^k \dots \dots \text{by taking log on both sides}$$

$$\Rightarrow \log n = k \log 2$$

$$\Rightarrow \log n = k \dots \dots [\log 2 \approx 1]$$

$$\Rightarrow \underline{k = \log n}$$

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

substitute values of $n = 2^k$ & $k = \log n$ in above equation, we get

$$T(n) = T(2^k/2^k) + \log n$$

$$T(n) = T(1) + \log n$$

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

$$T(n) = O(\log n)$$

+ Sorting Algorithms:

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

- basic sorting algorithms:

1. selection sort
2. bubble sort
3. insertion sort

- advanced sorting algorithm:

4. quick sort
5. merge sort

1. selection sort:

for size of an array = n

total no. of comparisons = $(n-1) + (n-2) + (n-3) + (n-4) + \dots + 1$

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

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

$T(n) = O(n^2 - n)$ divisive can be neglected

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

Rule:

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

e.g.

$O(n^3 + n^2 + 4) \Rightarrow O(n^3)$

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