

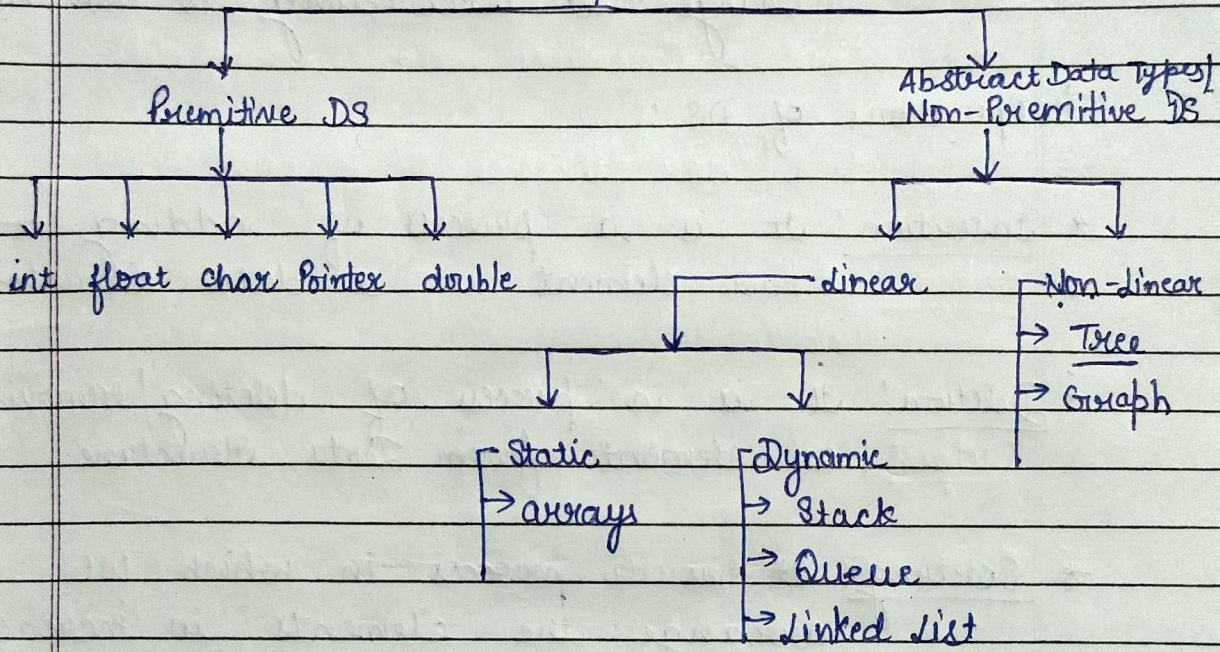
## Data Structure

Data: Anything given information is known as data. e.g.: Student's name

Structure: Representation of data.

Data Structure: It is a way of organizing data in such a way so that it can be used in an efficient way (better way).

### Types of DS



\* Primitive DS: These are basic data structure and directly operated by machine instruction

\* Non Primitive / ADT: These are derived from primitive DS.

- Tree: No cycle; Nodes are connected

- Graph: There is cycle when we connect nodes.

- \* Linear DS: In this we arrange data in a sequential way (ascending / descending order).
- \* Non linear: In this data is arranged in an hierarchical way / random way.
- \* Static: Memory is fix. We can't change it again & again.
- \* Dynamic: Memory is not fix. We can change it according to our wish.

#### ⇒ Operations of DS :

- \* Insertion: It is a process of adding a new element in a data structure.
- \* Deletion: It is a process of deleting / removing an element from Data Structure.
- \* Sorting: It is a process in which we arrange the elements in increasing or decreasing order.
- \* Merging: It is a process when we combine two data structures into a single data structure.
- \* Traversing: In this we visit each element atleast one time.

- \* Searching! Finding / Searching an element in DS.
    - ⇒ Importance of DS / Advantages
  - \* Efficient Data Processing! When we store data efficiently to save the memory / storage.
  - \* Algorithm Design: Follow steps in an efficient way.
  - \* Memory Management: Utilizing memory properly.
  - \* Scalability: Proper designing, maintaining of DS.
- ⇒ Disadvantages of DS :
- \* Complexity! Not all data types are complex but few are e.g! tree, graph etc
  - \* Memory Overhead: difficult that what we utilize where. Memory wastage also occurs.
  - \* Fixed Size! We can't change the size of arrays.
  - \* Performance tradeoff: When we use wrong data type then this leads to decrease the performance of algorithm.

\* Difficult in choosing right DS!

\* Data Duplication: When we use 2 data types at a same time leads to data duplication.

⇒ Algorithm: Step by step instruction / process for solving any problem.  
OR

It is well defined procedure that takes some input & produce some output.

⇒ Properties / characteristics of Algorithm:

1. Input: It should take 0 or more than 1 input.
2. Output: It must produce an output.
3. Finiteness: It must have finite number of steps.
4. Unambiguous: Each & every step must be clearly defined. No duplicacy.
5. Effectiveness: Each & Every step must be followed.

→ Different types of Algorithms!

\* Selection: Decision making + Control statements

# Sequence: No control statements & no decision making

-# Iteration

□ Sequence eg: Add 2 no's.

Step 1 : Start

2 : Read a,b

3 : Sum = a+b

4 : display Sum

5 : Stop.

□ Selection eg: Eligible for Vote :

Step 1 : Start

2 : Read age

3 : if age  $\geq 18$ ; go to step IV else Step V

4 : Write eligible for vote

5 : Write not eligible for vote

6 : Stop .

□ Iteration Eg : Fibonacci Series, Factorial.

Ques: Write an algorithm to check whether a no. is even or odd.

Ques: Write an algorithm to find greatest among 5 no's.

Ques: Write an algorithm to find factorial of a no.

Solution

1. Step 1: Start

2: Read a

3: if  $a \% 2 == 0$ ; go to step 4 else

4: Write even

5: Write odd.

6: Stop.

2. Step 1: Start

2: Read a, b, c, d, e

3: if  $(a > b) \& (a > c) \& (a > d) \& (a > e)$ :

~~if (a > c)~~ write a is greater.

4: else if  $(b > a) \& (b > c) \& (b > d) \& (b > e)$ :  
write b is greater.

5: else if  $(c > a) \& (c > b) \& (c > d) \& (c > e)$ :  
write c is greater.

6: else if  $(d > a) \& (d > b) \& (d > c) \& (d > e)$ :  
write d is greater.

7: else

write e is greater.

8: Stop

3. Step 1: Start

2: Read a

3: if  $(a == 0) \& (a == 1)$ :

    Write 1

4: else:

~~fact = 1~~ while  $a > 0$ :

~~fact = fact \* a~~

~~fact = fact \* a~~

$a--$

- 5! write fact.
- 6! Stop.

→ Flow chart:

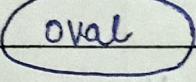
↳ Graphical representation of an algorithm.

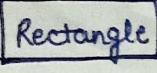
Algorithm : time & memory should be less  
amount of

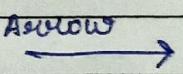
\* Time Complexity: Execution time of 1 Algo.

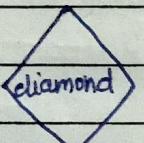
\* Space/Memory Complexity: Space that an algo take

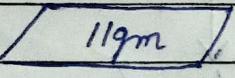
→ Symbols

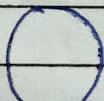
1.  → Start/Stop

2.  → Process/operation

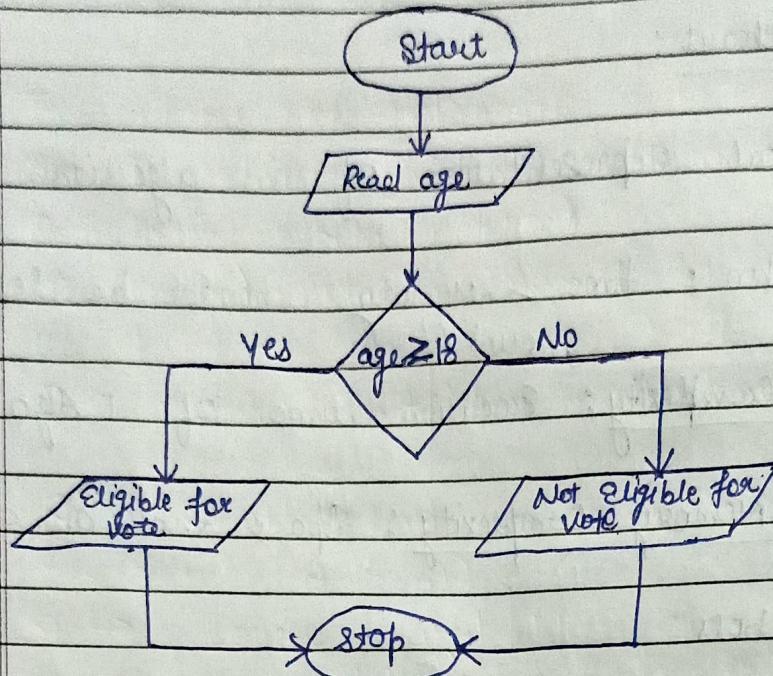
3.  → indicates the flow b/w steps

4.  → condition

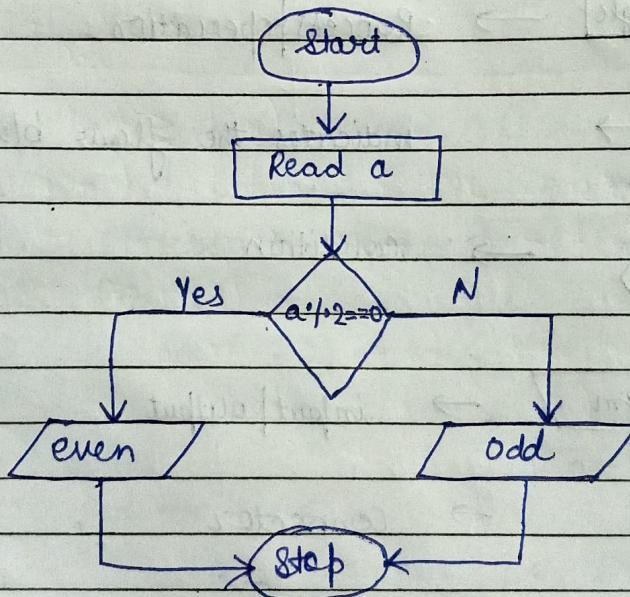
5.  → input/output

6.  → connector

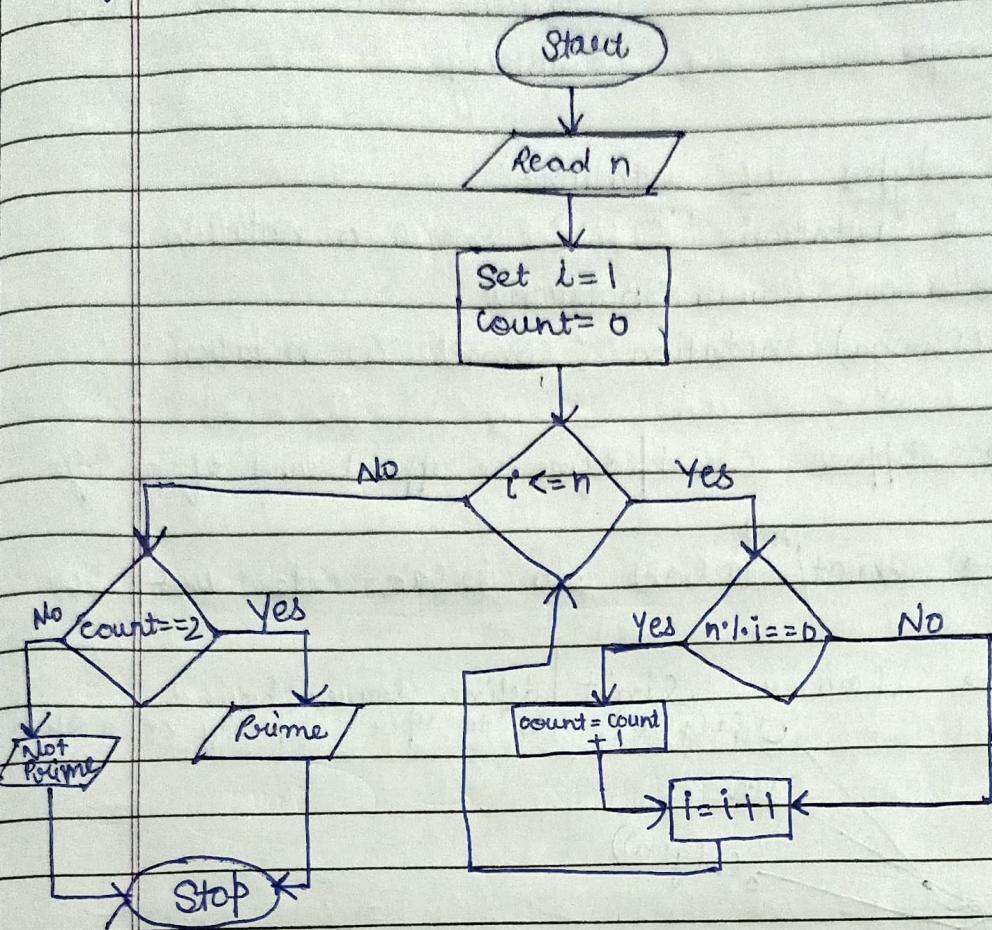
Eg.: Whether a person is eligible for vote or not.



Eg.: Make a flow chart to check whether a no. is even or odd.



eg whether a no. is prime or not.



⇒ Algorithm analysis :

Problem has more than one solution then which technique we are using to solve that particular problem is called as algo. analysis.

Analysis is done on two parameters

- \* Time Complexity
- \* Space Complexity

Two types of algo. analysis :

- \* A priori analysis : Exact time & space consume
- \* A posteriori analysis : Approx

2b Define algorithm complexity and its types along with algorithmic notations.

Ans Complexity in algorithms refers to the amount of resources (such as time or memory) required to solve a problem or perform a task. The most common measure of complexity is time complexity, which refers to the amount of time an algorithm takes to produce a result as a function of the size of the input. Memory complexity refers to the amount of memory used by an algorithm.

Types of complexity:

- (i) Constant Complexity: If the function or method of the program takes negligible execution time. Then that will be considered as constant complexity.
- (ii) Logarithmic complexity: It imposes a complexity of  $O(\log(N))$ . It undergoes

the execution of the order of  $\log(N)$  steps. To perform operations on  $N$  elements, it often takes the logarithmic base as 2.

3. Linear Complexity: It imposes the complexity of  $O(N)$ . It encompasses the same number of steps as that of the total number of elements to implement an operation on  $N$  elements.
4. Quadratic Complexity: It imposes the complexity of  $O(n^2)$ . For  $N$  input data size, it undergoes the order of  $N^2$  count of operations on  $N$  number of elements for solving a given problem.
5. Exponential complexity: It imposes a complexity of  $O(2^N)$ ,  $O(N!)$ ,  $O(nk)$ . For  $N$  elements, it will execute the order of the count of operations that is exponentially dependable on the input data size.

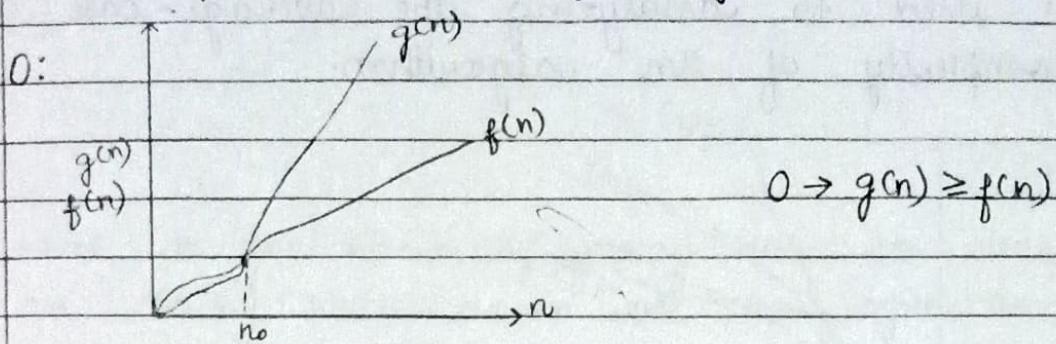
### $\Rightarrow$ Algorithmic Notations:

- Asymptotic Notation: It is a mathematical notation that are used to represent time complexity

Three types of notations:

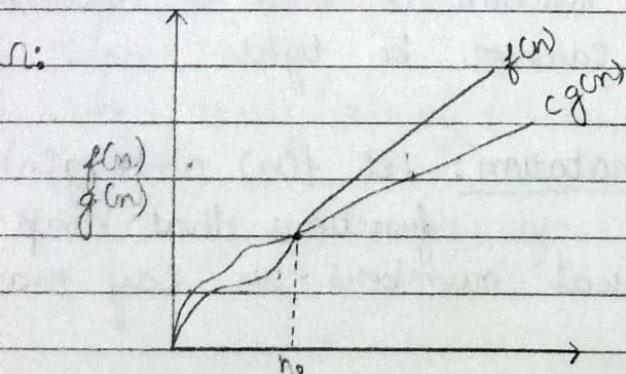
1. Big O notation, little o notation
2. Omega notation, little omega notation
3. Theta notation

Big O Notation: Big - O notation represents the upper bound of the running time of an algorithm. By using big O-notation, we can asymptotically limit the expansion of a running time a range of constant factors above and below. It gives the worst case complexity of an algorithm. It is a model for quantifying algorithm performance.



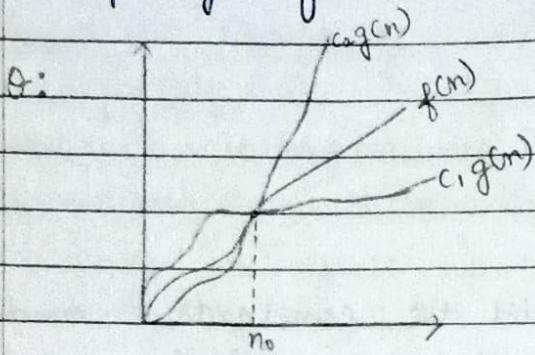
$O(g(n)) = \{f(n)\}$ : there exist +ve constants  $c$  and  $n_0$  such that  $0 \leq f(n) \leq cg(n)$  for all  $n \geq n_0$

Omega Notation: It represents the lower bound of the running time of an algorithm. Thus, it provides the best-case complexity of an algorithm. The execution time serves as a lower bound on the algorithms time complexity. It is defined as the condition that allows an algorithm to complete statement execution in the shortest amount of time.



$\Omega(g(n)) = \{ f(n) : \text{there exist +ve constants } c \text{ and } n_0 \text{ such that } 0 < f(n) \geq cg(n) \text{ for all } n \geq n_0\}$

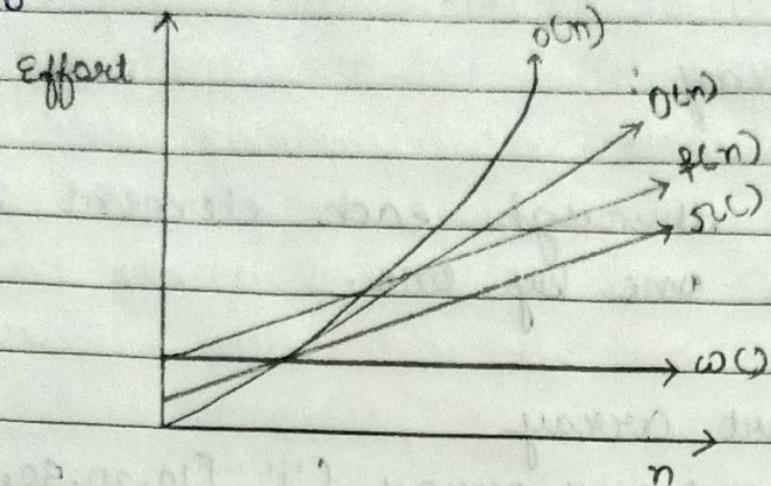
- Theta Notation: Theta notation encloses the function from above and below. Since it represents the upper and the lower bound of the running time of an algorithm, it is used to analyzing the average-case complexity of an algorithm.



$\Theta(g(n)) = \{ f(n) : \text{there exist +ve constants } c_1, c_2 \text{ and } n_0 \text{ such that } 0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \geq n_0\}$

- little o asymptotic notation: Big-O is used as a tight upper bound on the growth of an algorithm's effort (this effort is described by the function  $f(n)$ , even though, as written, it can also be a loose upper bound. "little-o" ( $o(\cdot)$ ) notation is used to describe an upper bound that cannot be tight.
- Little omega asymptotic notation: Let  $f(n)$  and  $g(n)$  be functions that map +ve integers to +ve real numbers. We say that

$f(n)$  is  $\omega(g(n))$  (or  $f(n) \in \omega(g(n))$ ) if for any real constant  $c > 0$ , there exist an integer constant  $n_0 \geq 1$  such that  $f(n) > c g(n) \geq 0$  for every integer  $n \geq n_0$ .



## → Properties of Asymptotic notation

\* Reflexive property: Only  $O$ ,  $\Omega$ ,  $\Theta$

$$f(n) = O f(n)$$

$$f(n) = \Omega f(n)$$

$$f(n) = \Theta f(n)$$

\* Symmetric property: Only theta shows symmetric property

$$f(n) = \Theta(g(n)) \text{ if and only if } g(n) = \Theta(f(n))$$

$a = b$

\* Transpose Symmetry:

$f(n) = O(g(n))$  iff  $g(n) = \Omega(f(n))$ ,  
 $f(n) = \Theta(g(n))$  iff  $g(n) = \omega(f(n))$ .

\* Transitivity property: All notation

If  $f(n) = O(g(n))$  and  $g(n) = O(h(n))$   
then  $f(n) = O(h(n))$

If  $f(n) = \Omega(g(n))$  &  $g(n) = \Omega(h(n))$  then  
 $f(n) = \Omega(h(n))$ .

If  $f(n) = O(g(n))$  &  $g(n) = O(h(n))$  then  
 $f(n) = O(h(n))$ .

If  $f(n) = O(g(n))$  &  $g(n) = O(h(n))$  then  
 $f(n) = O(h(n))$ .

If  $f(n) = \omega(g(n))$  &  $g(n) = \omega(h(n))$  then  
 $f(n) = \omega(h(n))$

⇒ Types of Complexities:

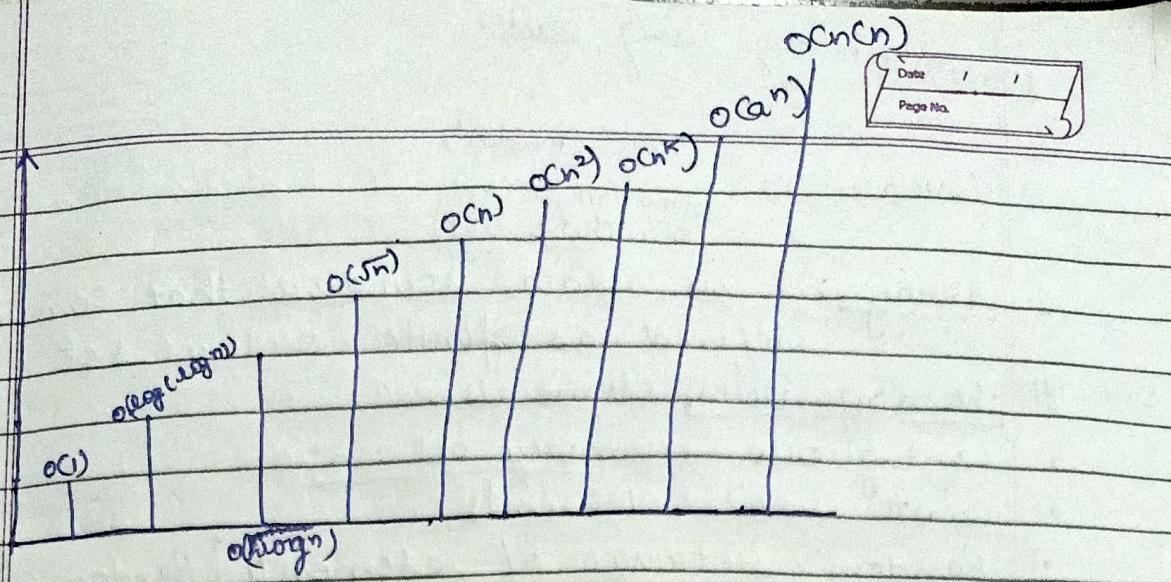
1. Constant time complexity =  $O(1)$

2. Logarithmic " " =  $O(\log(\log n))$ ,  
 $O(\sqrt{\log n})$ ,  $O(\log n)$ .

3. Linear time complexity =  $O(\sqrt{n})$  and  $O(n)$

4. Polynomial " " =  $O(n^k)$  where  $k =$   
constant &  $k > 1$

5. Exponential time complexity =  $O(a^n)$  where  
 $a > 1$ .



Lab Inherit function: String Concatenation  $\rightarrow$  struct  
 Kush part string da concat karao & Strncat  
 $\text{strncat}(\text{Str1}, \text{Str2}, \text{len}) \rightarrow \text{index}$   
 Fine letter jodne a

$\Rightarrow$  Difference b/w Linear and Non-Linear DS.

### Linear DS

### Non-Linear DS

- |  |  |
|--|--|
| * The elements are arranged in a sequence such as array, stack or queue. | Elements are organized arranged in Random/ hierarchical order. |
| * In linear ds single level is involved.                                 | Multiple levels are involved.                                  |
| * Easy to implement  | Difficult to implement   |
| * Data elements can be traversed in a single run.                        | Cannot be traversed in a single run.                           |
| * Memory is not used in an efficient manner.                             | Memory used in efficient manner.                               |

Ques: What do you mean by array? How to declare an array? Write down all the operations of array.

Ans: array: It is a linear data structure that stores a collection of items of same data type in contiguous memory location. Each item in an array is indexed starting with 0. We can directly access an array element by using its index value.

⇒ Declaration of array

Arrays can be declared in various ways in different languages, we declare array in python.

→ Using array module:

Import array

arr = array.array('i',[1,2,3,4,5])

Here 'i' specifies the type of element in array.

→ Using Numpy Module  
Import numpy as np  
arr = np.array ([1, 2, 3, 4, 5])

⇒ operations in array :

1. Traversing: Going through each element in the array one by one.

For example: Import array  
arr = array.array ('i', [10, 20, 30, 40])  
for i in range (len(arr)):  
    print (arr[i])

2. Insertion: Adding a new element to the array

For example: arr = [10, 20, 30, 40]  
arr.append (50)

3. Deletion: Removing the element from array

For example: arr = [10, 20, 30, 40]  
arr.pop (1) # Remove element at index 1

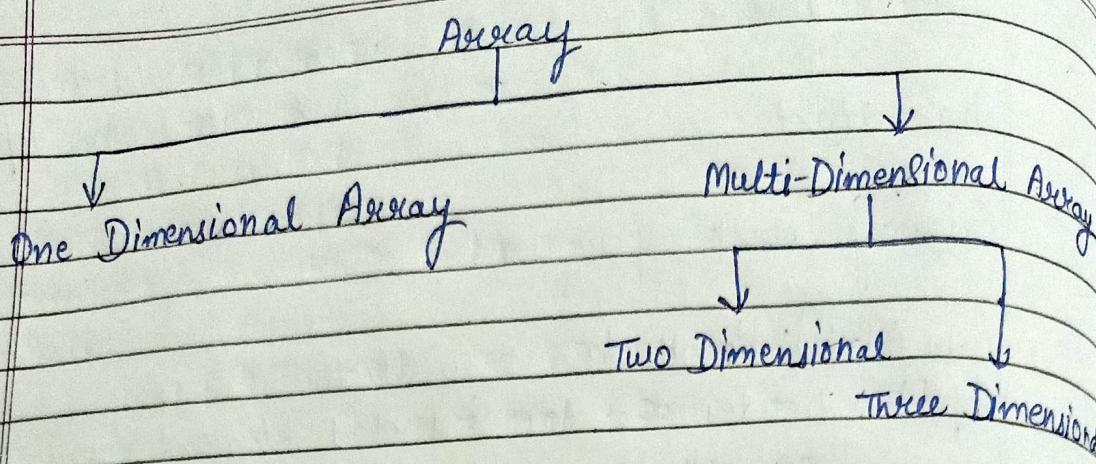
4. Searching: Locating an element in the array

For example: arr = [10, 20, 30, 40]  
found = False  
for i in range (len(arr)):  
    if arr[i] == 30:  
        found = True  
        break

5. Update: Changing the value of element at a particular index.

For Example: arr = [10, 20, 30, 40]

arr[3] = 50 # change the element at index 3 to 50.



- One Dimensional Array ? Single Row and multiple Columns.

eg: array = [10 | 20 | 30 | 40 | 50]

→ ~~Two~~ ways to import array module?

1. import array - This will import the entire array module.
2. from array import \* - This will import all class, objects, variables etc from array module.

3 import numpy

→ Creating & Initializing 1-D Array?

import array as arr

array\_name = arr.array('type\_code', [elements])

eg import array as arr

marks = arr.array('i', [10, 20, 30, 40, 50])

- Two-Dimensional Array: list of lists which represent a table like structure with rows & columns

import numpy as np.

array name = np.array ([[elements], [elements]])

eg: a = np.array ([[1, 2, 3], [1, 2, 3]])

- Three-Dimensional Array:

import numpy as np

array name = np.array ([[ [elements], [elements]], [[elements], [elements]]]])

OR

array name = np.arange(1, 25).reshape(2, 3, 4)

→ output: [[[1, 2, 3, 4],  
[5, 6, 7, 8],  
[9, 10, 11, 12]],

[[13, 14, 15, 16],  
[17, 18, 19, 20],  
[21, 22, 23, 24]]])

\*

- Numericals on Array: 1D array.

- Address of the element at  $k^{th}$  index.

$a[k] = \text{base address} + \text{size of 1 element} * [\text{element}]^{\text{No. of elements}}$   
 before the element whose index is given (given  
 bound - lower bound)] which

$$\text{eg 1. } A = [0, 1, 2, 3, \dots, 100] \\ \text{size} = 4 \text{ Byte}$$

$$\text{Size} = 4 \text{ bytes}$$

address = 1001

Size = 100  
Base address = 1001

Base address = 1001  
Find the address of the element that is present on 50<sup>th</sup> index.  
 $\therefore \text{Address} = 1001 + 4(50)$

$$\Rightarrow 1001 + 4(50 - 0) = 1001 + 4(50) \\ = 1001 + 200 = 1201 \text{ Ans}$$

ege Consider the linear array A [16 : 30]. If base address is 100 and space required to store each element is 4 words then the address of A[27] is given by.

$$\begin{aligned} \Rightarrow A[27] &= 100 + 4(27 - 16) \\ &= 100 + 4(11) \\ &= 100 + 44 = \boxed{144} \text{ Ans} \end{aligned}$$

## Numericals in 2D Array

Two Methods  $\rightarrow$  Row Major Storage.

→ Column Major Storage: |||||

B - Base Address    W = Storage size of an element

Lee-Lower bound of now

Lc - Lower bound of column

$m$  - No. of rows in matrix

$N$  - No. of columns in matrix

#

# Numericals of 2D Array

Date \_\_\_\_\_  
Page No. \_\_\_\_\_

1. Row Major:

$$\text{Address of } A[i][j] = B + W * [N * (i - 1_{\alpha}) + (j - 1_{\alpha})]$$

eg:  $A[2][1] = 104 + 2 * [4 * (2 - 0) + (1 - 0)]$   
 $= 104 + 2 * [8 + 1]$   
 $= 106(?) = 104 + 2(9)$   
 $= 104 + 18 = \boxed{122} \text{ Ans}$

2. Column Major:

$$\text{Add of } A[i][j] = B + W * [(i - 1_{\alpha}) + M * (j - 1_{\alpha})]$$

eg:  $A[1][2] = 104 + 2 * [(1 - 0) + 3(2 - 0)]$   
 $= 104 + 2 * [1 + 6]$   
 $= 104 + 2 * [7]$   
 $= 104 + 14 = \boxed{118} \text{ Ans}$

	0	1	2	3
0	5	7	4	3
1	1	0	9	8
2	6	2	1	4

R C

18 OB LB UB

Ques An array,  $M[1:10, 1:10]$  required 2 bytes of the storage for each element and the beginning location is 500 so determine the location of  $M[5][6]$ .

⇒ Row major:  $M[5][6] = 500 + 2 * [10(5 - 1) + (6 - 1)]$   
 $= 500 + 2[40 + 5]$   
 $= 500 + 2(45)$   
 $= 500 + 90$   
 $= 590 \text{ Ans}$