

Data Structure

- organising data into memory for efficient processing, along with few operations like add, delete, edit, search etc that we can perform on that data
eg. stack - linearly (sequential) operations - push, pop, peek
array - contiguous arrangement
- used to achieve
 1. Abstraction
 - how data is organised into memory, is hidden from outside
 - how data is processed into memory, is also hidden from outside
 - Abstract Data Types (ADT)
 2. Reusability
 - reused in our applications as per our need
 - reused to implement some another data structure
 - reused to implement to few algorithms eg. traversal into Tree, graph
 3. Efficient processing
 - efficiency is measured in two terms
 1. Time - time required to execute
 2. Space -space required inside memory to execute

Types of Data structure

Linear Data Structures

- Data is organised sequentially



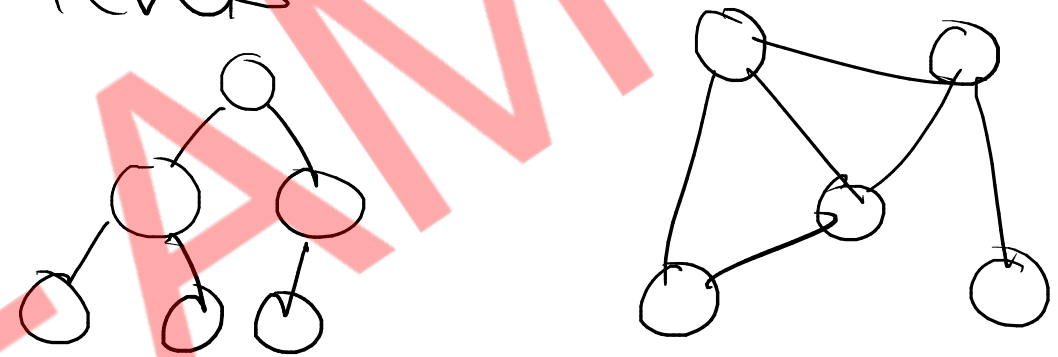
- data can be accessed linearly/sequentially

- basic data structures

- 1) Array
- 2) Struct/class
- 3) stack
- 4) Queue
- 5) Linked List

Non-linear (hierachical) Data Structures

- Data is organised in multiple levels



- data can not be accessed linearly/sequentially

- Advanced data structures

- 1) Tree (heap)
- 2) Graph

1) Hash Table

Algorithm

Program - set of instructions to machine (CPU)

Algorithm - set of instructions to human (developer)

- step by step solution of given problem statement

Search key inside collection of values (array)

step1 : take key from user

step2 : traverse array elements one by one

step3 : compare key with every element of array

step4 : if key is found return True or not found return False

- always written into human understandable languages**
- algorithms are templates/blue prints**
- algorithms can be implemented in any of the programming languages**
- Algorithm - template**
- Program - implementation**

eg searching algorithms

sorting algorithms

Algorithm analysis / Efficiency measurement / Complexities

- finding time and space requirement of an algorithm
 1. Time - time required to execute the algorithm (ns, us, ms, s)
 2. Space - space required to excute the algorithm inside memory (bytes, kb, mb, ..)

1. Exact analysis

- finding exact space and time of the algorithm
- it depends on some exeternal factors
- time is dependent on type of machine(cpu), no of processes running at that time
- space is dependent on type of machine(architecture), data types

2. Approximate analysis

- finding approximate time and space of the algorithm
- mathematical approach is used to find time and space complexity of the algorithm and it is known as "Asymptotic analysis"
- it also tells about behavior of the algorithm when input is changed or sequence of input is changed
- behaviour of algorithm can be observed into three cases
 1. Best case
 2. Average case
 3. Worst case

to denote time and space complexity
we use Big-O notation

Time Complexity

- count the number of iterations for the loop which is used inside the algorithm
- time required is directly proportional to the iterations of the loop

1. print 1D array on console

```
void print1DArray(int arr[], int n){  
    for(int i = 0 ; i < n ; i++)  
        sysout(arr[i]);  
}
```

loop iterations = n

Time $\propto n$

Time complexity

$$T(n) = O(n)$$

2. print 2D array on console

```
void print2DArray(int arr[][], int m, int n){  
    for(int i = 0 ; i < m ; i++)  
        for(int j = 0 ; j < n ; j++)  
            sysout(arr[i][j]);  
}
```

Outer loop iterations = m
inner loop iterations = n

Total iterations = $m * n$

Time $\propto m * n$

Time complexity

$$T(m, n) = O(m * n)$$

$\therefore n == m$

Time $\propto n^2$

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

3. add two numbers

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

- time requirement is same for different values of $n1$ & $n2$
- constant time requirement

Time complexity = $T(n) = O(1)$

4. print table of given number

```
void printTable(int num){  
    for(int i = 1 ; i <= 10 ; i++)  
        sysout(num * i);  
}
```

- loop will iterate constant amount of time

- time requirement is constant

Time complexity = $T(n) = O(1)$

5. print binary of decimal number

2	9
	4
	2
	1

1
0
0
1

```
void printBinary(int n){
    while(n > 0){
        sysout(n % 2);
        n = n / 2;
    }
}
```

n	n > 0	n%2
9	T	1
4	T	0
2	T	0
1	T	1
0	F	

$$(9)_{10} = (1001)_2$$

$$n = n, n/2, n/4, n/8, \dots, n/2^{\text{itr}}$$

$$n = n/2^0, n/2^1, n/2^2, \dots, n/2^{\text{itr}}$$

for $n=1$, last time loop condition will be true

$$n/2^{\text{itr}} = 1$$

$$n = 2^{\text{itr}}$$

$$\log n = \log 2^{\text{itr}}$$

$$\text{itr} \log 2 = \log n$$

$$\text{itr} = \frac{\log n}{\log 2}$$

$$\text{time} \propto \text{itr}$$

$$\text{time} \propto \frac{\log n}{\log 2}$$

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

Time complexities : $O(1)$, $O(\log n)$, $O(n)$, $O(n \log n)$, $O(n^2)$, $O(n^3)$, ... $O(2^n)$, ...

modification : '+' / '-' : in terms of n

modification : '*' / '/' : in terms of $\log n$

$\text{for}(i=0; i < n; i++) \rightarrow O(n)$
 $\text{for}(i=n; i > 0; i--) \rightarrow O(n)$
 $\text{for}(i=0; i < n; i+=2) \rightarrow O(n)$
 $\text{for}(i=0; i < n; i+=20) \rightarrow O(n)$
 $\text{for}(i=n; i > 0; i/=2) \rightarrow O(\log n)$
 $\text{for}(i=1; i < n; i*=2) \rightarrow O(\log n)$
 $\text{for}(i=0; i < n; i++)$
 $\quad \text{for}(j=0; j < n; j++) \rightarrow O(n^2)$
 $\text{for}(i=0; i < n; i++);$
 $\quad \text{for}(j=0; j < n; j++); \rightarrow n$
 $\text{for}(i=0; i < n; i++)$
 $\quad \text{for}(j=n; j > 0; j/=2); \rightarrow \log n$

$\rightarrow n$
 $+ = 2n$
 $\rightarrow n$

Time $\propto 2n$
 $T(n) = O(n)$

$\rightarrow n$
 $\rightarrow \log n$
 $\times = n \log n$ Time $\propto n \log n$
 $T(n) = O(n \log n)$

Space Complexity

- finding approximate space required to execute an algorithm

Total space = **input space** + **Auxillary space**
(space of actual data) (space required to process actual data/input)

- find sum of array elements

step1 : create sum and initialize to 0

step2: traverse array from 0 to N-1 index

step3 : add each element into sum variable

step4 : return / print sum

Auxillary Space Analysis

Processing variables = sum, i, N

Auxillary space = 3 units

space $\propto 3$

$S(n) = O(1)$

Array size = n

Input variable - array
Input space = n units

Processing variables = sum, i, N

Auxillary space = 3 units

Total space = $n + 3$

space $\propto n + 3$

$\because n \gg 3$

space $\propto n$

space complexity $S(n) = O(n)$

Searching Algorithms

- finding some key(data to be searched) into collection(set) of values

1. linear search (data is random)

2. binary search (data is sorted)

1. Linear Search

//1. take key from user

//2. traverse array from 0 to N-1 index

//3. compare key with every element of array

//4. if key is found return true / i

//5. if key is not found false / -1

2. Binary Search

//1. take key from user

//2. divide array into two parts (find middle element)

//3. compare middle element with key

//3.1 if key is matching return index of it

//4. if key is less than middle element then search it in left partition

//5. if key is greater than middle element then search it in right partition

//6. repeat step 2 to 5 until key is found

//7. if key is not found return -1

Searching Algorithms Analysis

- for searching and sorting algorithms, we count number of comparisons
- time is directly proportional to number of comparison

Linear Search

Best case : key is found in first few comparison : $O(1)$
Avg case : key is found in middle positions : $O(n)$
Worst case : key is found in last few comparisons : $O(n)$
key is not found

Binary Search

Best case : key is found in first few comparison : $O(1)$
Avg case : key is found in middle positions : $O(\log n)$
Worst case : key is found in last few comparisons : $O(\log n)$
key is not found



$$2^3 = 8$$

$$2^n = n$$

$$\text{itr} = \frac{\log n}{\log 2}$$

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

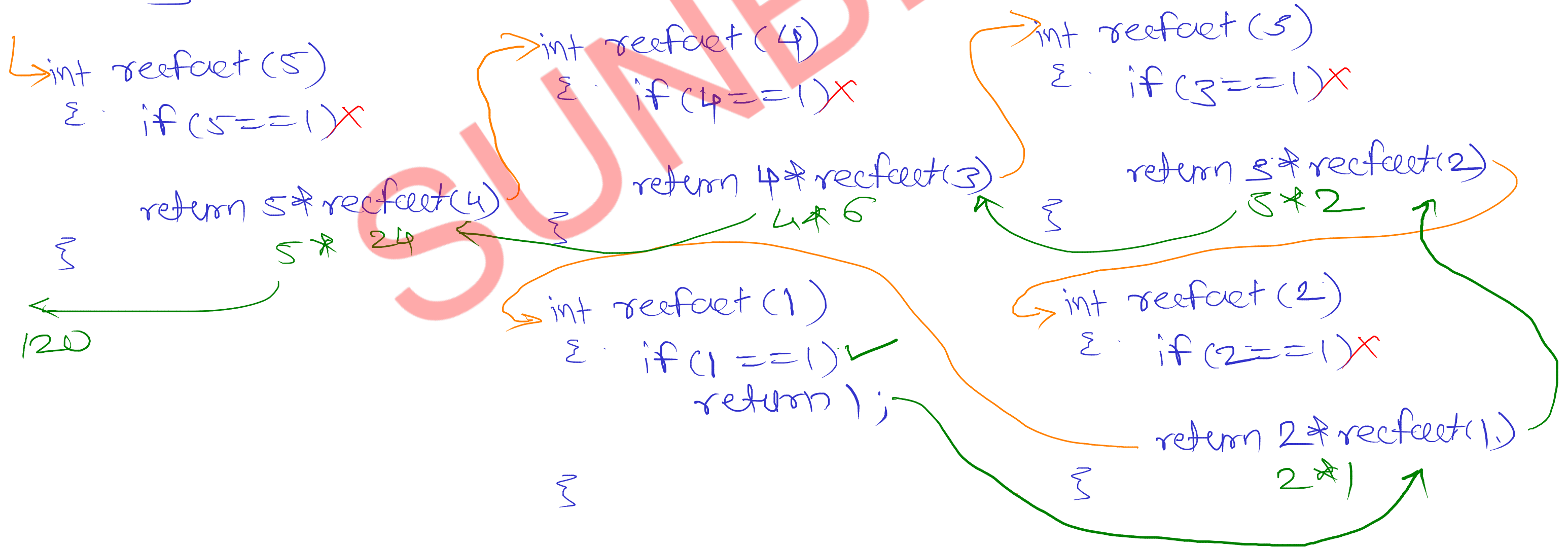
Recursion

- function calling itself
- we can use recursion if
 1. we know the process/formula in term of itself
 2. we know the terminating condition

$$n! = n * (n-1)!$$

$$1! = 1$$

```
int recfact(int n) {  
    if(n==1)  
        return 1;  
    return n * recfact(n-1)  
}
```



Algorithm Implementation Approches

Any algorithm can be implemented using two approches

1. Iterative approach

- loops are used

```
int fact(int n) {  
    int fact = 1;  
    for(i = 1; i <= n; i++)  
        fact = fact * i;  
    return fact;  
}
```

Time & no. of iterations
of loop

loop iterations = n

$$T(n) = O(n)$$

2. Recursive approach

- recursion is used

```
int recfact(int n) {  
    if(n == 1)  
        return 1;  
    return n * recfact(n-1)  
}
```

Time & no. of recursive
calls

recursive calls = n

$$T(n) = O(n)$$

Sorting Algorithms

- arrangement of data in either ascending or descending order of their values
- Basic sorting algorithms
 1. Selection sort
 2. Bubble sort
 3. Insertion sort
- Advanced sorting algorithms
 4. Merge sort
 5. Quick sort
 6. Heap sort

Selection sort

- //1. select one position (index) of array**
- //2. compare selected position element with all other elements one by one**
- //3. if selected position element is greater than other element**
 - //3.1 swap both the elements**
- //4. repeat above steps till array is not sorted**

Sorting Algorithms Analysis

no. of elements = n

no. of passes = $n-1$

$n=6$

pass1 = 5 $n-1$

pass2 = 4 $n-2$

pass3 = 3 $n-3$

pass4 = 2

pass5 = 1

$$\begin{aligned} \text{Total Comps} &= (n-1) + (n-2) + (n-3) + \dots + 1 \\ &= 1 + 2 + 3 + \dots + \frac{(n-1)}{n} \end{aligned}$$

$$= \frac{n(n+1)}{2}$$

$$= \frac{n^2 + n}{2}$$

n	n^2
1	1
10	100
100	10000
1000	1000000

Time \propto comps

$$\text{Time} \propto \frac{n^2 + n}{2}$$

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

- Mathematical polynomial
- Degree of polynomial
 - ↳ highest degree
- highest degree term is highest growing term in polynomial always.