

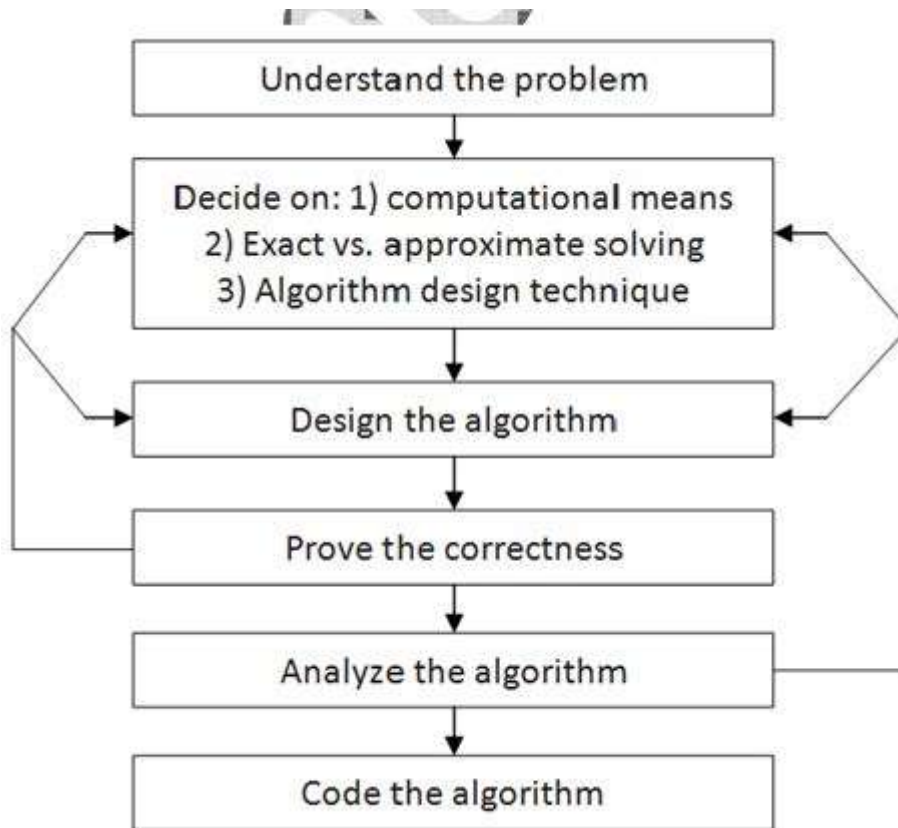
Sub:	<b>DESIGN AND ANALYSIS OF ALGORITHMS</b>						Code:	17CS43	
Date:	06/03/2019	Duration:	90 mins	Max Marks:	50	Sem:	IV	Branch:	<b>ISE</b>
Answer Any <b>FIVE FULL</b> Questions									

1 (a) What is Algorithm? What are the characteristics of algorithm? Explain the algorithm design process with flow chart?

An algorithm is a set of steps of operations to solve a problem performing calculation, data processing, and automated reasoning tasks. An algorithm is an efficient method that can be expressed within finite amount of time and space.

A sequence of steps one typically goes through in designing and analyzing an algorithm

(b)



$T1(n)=O(g1(n))$  And  $T2(n)=O(g2(n))$  Prove that  $T1(n)+T2(n)=O(\text{Max}(g1(n),g2(n)))$

Marks	OBE	
	CO	RBT
[6]	CO1	L2
[4]	CO2	L4

3) If  $t_1(n) \in O(g_1(n))$  and  $t_2(n) \in O(g_2(n))$ , then prove that  $t_1(n) + t_2(n) \in O(\max\{g_1(n), g_2(n)\})$

For any four arbitrary real numbers,  $a_1, b_1, a_2$  and  $b_2$  such that  $a_1 \leq b_1$  and  $a_2 \leq b_2$ .  
We have  $a_1 + a_2 \leq 2 \max\{b_1, b_2\}$

Since  $t_1(n) \in O(g_1(n))$ , then there exists some constant  $c_1$  and non-negative integer  $n_1$  such that

$$t_1(n) \leq c_1 g_1(n) \text{ for all } n \geq n_1$$

Since  $t_2(n) \in O(g_2(n))$ , then there exists some constant  $c_2$  and non-negative integer  $n_2$  such that

$$t_2(n) \leq c_2 g_2(n) \text{ for all } n \geq n_2$$

Let  $c_3 = \max\{c_1, c_2\}$  and  $n_0 = \max\{n_1, n_2\}$

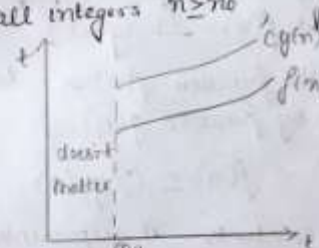
$$\begin{aligned} t_1(n) + t_2(n) &\leq c_1 g_1(n) + c_2 g_2(n) \\ &\leq c_3 g_1(n) + c_3 g_2(n) \\ &= c_3 (g_1(n) + g_2(n)) \\ &\leq 2 c_3 \max\{g_1(n), g_2(n)\} \end{aligned}$$

Hence,  $t_1(n) + t_2(n) \in O(\max\{g_1(n), g_2(n)\})$ , with the constants  $c$  and  $n_0$  required by the  $O$  definition being  $2 c_3 = 2 \max\{c_1, c_2\}$  and  $\max\{n_1, n_2\}$  respectively.

2 (a) Explain the three asymptotic notations, Big Oh, Big Omega and Big Theta notations with example

2(a) Asymptotic Notation:  
The idea of the asymptotic analysis is have the measure of the efficiency of the algorithm, that does not depend on the machine specific characteristics. Asymptotic notation is the mathematical tools that are provided to determine the time complexity of an algorithm. They are three types of Asymptotic Notations:

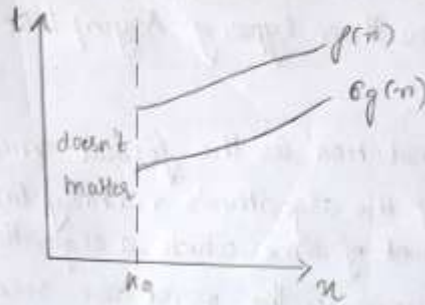
1) Big Oh Notation:  
The Big Oh ( $O$ ) notation is the formal method of expressing the upper bound of the algorithm's running time. It is the measure of the longest amount of time which a algorithm could possibly take to complete. It represents the worst case scenario of an Algorithm.  
For the two non-negative functions  $f(n)$  and  $g(n)$ ,  $f(n) = O(g(n))$  only when there exists a integer  $n_0$  and a constant  $c, c > 0$  such that for all integers  $n \geq n_0$ ,  $f(n) \leq c g(n)$ .



2) Big Omega Notation:  
The Big Omega ( $\Omega$ ) notation is the formal method of expressing the lower bound of the algorithm's running time. It is a measure of the smallest amount of time which a algorithm could possibly take to complete. It represents the best case scenario of an Algorithm.

[10]	CO1	L3

For two non-negative functions  $f(n)$  and  $g(n)$ ,  $f(n) = \Omega(g(n))$  only when there exists an integer  $n_0$  and a constant  $c, c > 0$  such that for all integers  $n \geq n_0$ ,  $f(n) \geq cg(n)$ .



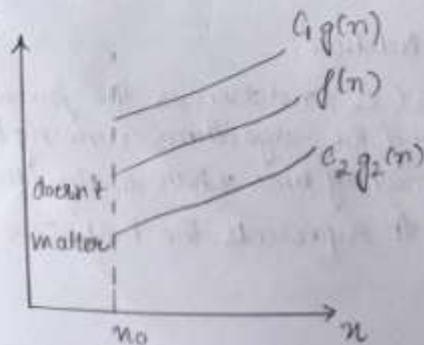
3) Big Theta notation;

For any two non-negative functions  $f(n)$  and  $g(n)$ ,  $f(n)$  is theta of  $g(n)$  i.e.,  $f(n) = \Theta(g(n))$  only when  $f(n) = O(g(n))$  and  $f(n) = \Omega(g(n))$ .

This clearly states that the function  $f(n)$  is bounded at both the top and the bottom by function  $g(n)$ .

$$0 \leq C_1(g(n)) \leq f(n) \leq C_2 g_2(n)$$

where  $C_1$  &  $C_2$  are two constants. It represents the average case scenario of an algorithm.



3 (a) Discuss the 'Tower of Hanoi' problem and write a recursive algorithm to solve it. Mathematically analyze the Tower of Hanoi problem and find its complexity.

[10]

CO2

L3

### 3a) Tower of Hanoi

It is a very famous Mathematical puzzle, wherein there will be 'n' no. of disks which vary in size and 3 pegs. The problem is that 'n' no. of disks should be transferred from 1st peg to 3rd peg by making use of the 2nd peg, such that

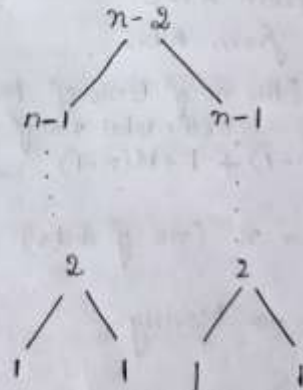
- 1) No two disks can be moved together, i.e., can move only one disk at a time.
- 2) The smaller disk should always be above the larger disks

Recursive Algorithm:

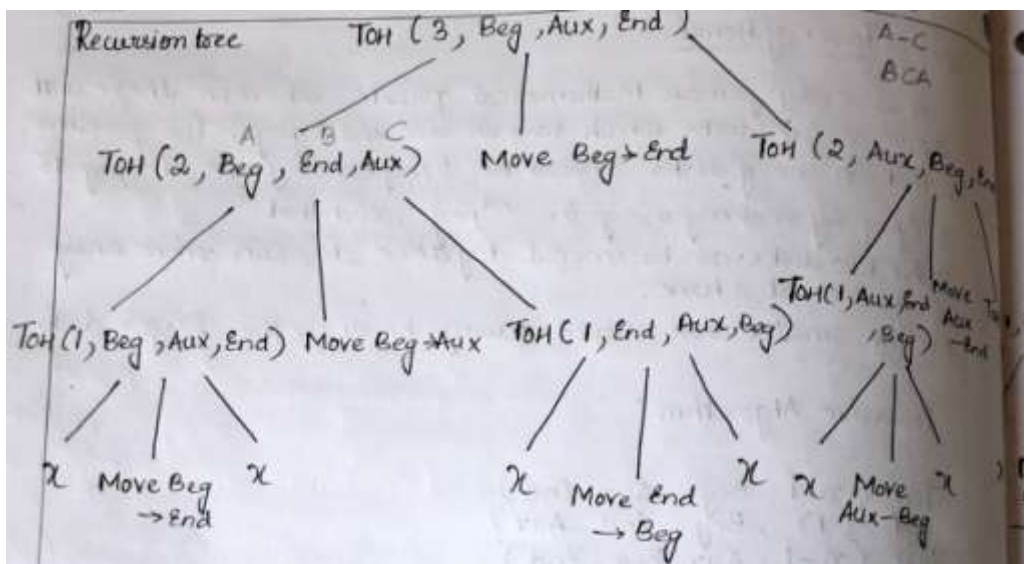
TOH (n-1, Beg, Aux, End)  
TOH (1, Beg, End, Aux),  
TOH (n-1, Aux, Beg, End)

The above functions are recursively called, till the disks reach 1.

Recursive tree



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### Mathematical Analysis of TOH

Move (n-1) disks from A to B

Move 1 disk from A to C

Move (n-1) disks from B to C.

We have  $M(n) \rightarrow$  (the no. of times of transferring the disks) or the total no. of Moves in moving n disks

$$M(n) = M(n-1) + 1 + M(n-1) \rightarrow (1)$$

① Input parameter  $\rightarrow$  'n' (no. of disks)

② Basic operation  $\rightarrow$  Moving

③  $M(n) = M(n-1) + 1 + M(n-1)$  from (1)



$\Rightarrow M(n) = 2M(n-1) + 1$ 
 $M(n-1) = 2M(n-2) + 1$   
 $M(n-2) = 2M(n-3) + 1$

$= 2 [2M(n-2) + 1] + 1$

$= 2^2 M(n-2) + 2 + 1$

$= 2^2 [2M(n-3) + 1] + 2 + 1$

$= 2^3 M(n-3) + 2^2 + 2 + 1$

$M(n) = 2^i M(n-i) + 2^{i-1} + 2^{i-2} + \dots + 2^2 + 2 + 1$

We know that GP =  $1 + 2 + 2^2 + \dots$

$a = 1$     $r = 2$

$S = \frac{a(r^n - 1)}{r - 1} = 1 \left( \frac{2^i - 1}{2 - 1} \right) = 2^i - 1 \rightarrow 2^i - 1$

$(2) \rightarrow M(n) = 2^i M(n-i) + 2^i - 1 \rightarrow (3)$

$n - i = 1$   
 $\rightarrow i = n - 1$   
 replace  $i$  by  $n - 1$  in (3)

$M(n) = 2^{n-1} M(n - (n-1)) + 2^{n-1} - 1$

$= 2^{n-1} M(1) + 2^{n-1} - 1$       ( $M(1) = 1$ )

$= 2^{n-1} + 2^{n-1} - 1$

$= 2^{n-1} + 2^{n-1} + 1 = 2^n (2^{-1} + 2^{-1}) + 1$

$= 2^n + 1$

$\therefore \underline{O(2^n)}$

4 (a) Explain the general plan for mathematical analysis of non-recursive function with example?

[10]	C01	L2
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#### 4(a) Analysis of non-Recursive function:

- ① Identify the input parameter.
- ② Find the basic operation
- ③ Check whether the no. of basic operation varies for diff value of the input of the same size. Find the best, worst and avg case.
- ④ Express the total number of basic operation in summation form.
- ⑤ By using the standard Mathematical formula's, find the closed formula for the algorithm and find the / reestablish the order of growth.

Example: Uniqueness of an Array.

This algorithm it checks for the uniqueness of an array, that is there are no duplicate elements in the array.

The algorithm returns true, if array elements are distinct otherwise false if duplicates exist.

```
for (i = 0 to n-2)
{
  for (j = i+1 to n-1)
  {
    if (a[i] == a[j])
  }
  return false,
else
return true is the algorithm
```

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① Input parameter - 'n' (The no. of array elements)

② Basic Operation - Comparison.

③ Find the Worst Case Complexity.

→ where there are no equal elements otherwise when the first two array elements are only pair of elements equal.

$$C_{\text{worst}} = \sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1 \quad \left( \begin{array}{l} 1, \text{ since there} \\ \text{only one basic} \\ \text{operation is,} \\ \text{compare} \end{array} \right)$$

$$= \sum_{i=0}^{n-2} (UB - LB + 1)$$

$$= \sum_{i=0}^{n-2} (n-1 - i - i + 1)$$

$$= \sum_{i=0}^{n-2} (n-1-i)$$

$$= \sum_{i=0}^{n-2} (n-1) - \sum_{i=0}^{n-2} i$$

$$= (n-1) \sum_{i=0}^{n-2} 1 - \sum_{i=0}^{n-2} i \quad \leq i = \frac{n(n+1)}{2}$$

$$= (n-1) [n-2-0+1] - \sum_{i=0}^{n-2} i$$

$$= (n-1)(n-1) - \left[ \frac{n(n+1)}{2} \text{ replace } n \text{ by } n-2 \right]$$

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

$$= \frac{2n^2 + 2 - 4n - n^2 - n}{2}$$



$$= \frac{n^2 - 5n + 2}{2} \cdot \frac{n-1}{2} \left[ \frac{2n-2}{2} - \frac{n-2}{2} \right] = \frac{n(n-1)}{2}$$

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

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

$\therefore O(n^2)$  (order of growth)

8a)

	SE	frequency		Time	
		$n \leq 1$	$n > 1$	$n \leq 1$	$n > 1$
void fibonacci (int n)	-	-	-	-	-
{	-	-	-	-	-
if (n ≤ 1)	1	1	1	1	1
cout << n	1	1	0	1	0
else {	-	-	-	-	-
int f1=0, f2=1;	1	0	1	0	1
int fn;	1	0	1	0	1
for (i=2; i ≤ n; i++) {	1	0	n	0	n
fn = f1 + f2	1	0	n-1	0	n-1
f1 = f2	1	0	n-1	0	n-1
f2 = fn	1	0	n-1	0	n-1
}	-	-	-	-	-
cout << fn	1	0	1	0	1
}	-	-	-	-	-
				2	
				$O(1)$	
					$2 + n + 3(n-1)$ $2 + n + 3n - 3$ $4n - 1$

5 (a) Explain Masters theorem. Prove that

- $T(n) = 16T(n/4) + n = \theta(n^2)$
- $T(n) = 3T(n/2) + n^2 = \theta(n^2)$
- $T(n) = 3T(n/3) + \sqrt{n} = \theta(n)$
- $T(n) = 4T(n/2) + n^2 = \theta(n^2 \log n)$

[2+4] CO2 L3

(b) Write the algorithm for sequential search, obtain the time complexity of this algorithm for Successful and unsuccessful search in the worst case and best case.

[4] CO3 L2

6 (a) Prove if following equalities are correct or incorrect

i)  $6n^2 - 5 = \theta(n^2)$   $c_1g(n) < f(n) < c_2g(n)$

ii)  $2n^2 + 8 = O(n)$  fail

iii)  $120n + 5 = \Omega(n^3)$  fail

(b) Explain the concept of divide and conquer strategy along with its general recurrence equation.

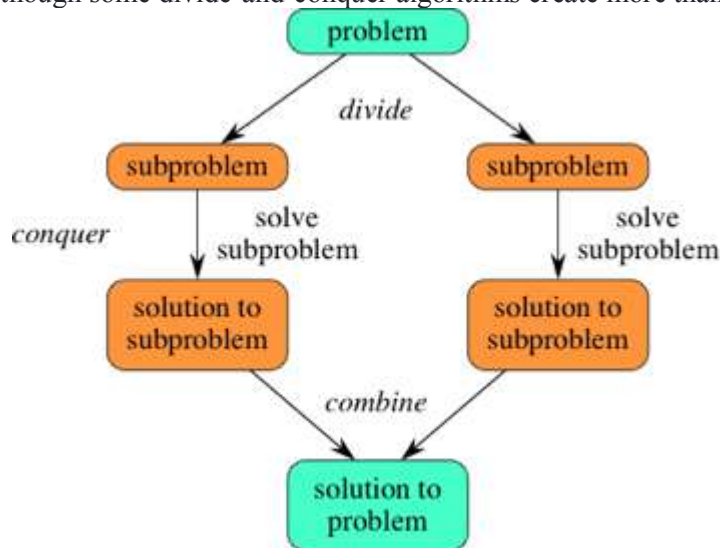
This paradigm, **divide-and-conquer**, breaks a problem into subproblems that are similar to the original problem, recursively solves the subproblems, and finally combines the solutions to the subproblems to solve the original problem. Because divide-and-conquer solves subproblems recursively, each subproblem must be smaller than the original problem, and there must be a base case for subproblems. You should think of a divide-and-conquer algorithm as having three parts:

**Divide** the problem into a number of subproblems that are smaller instances of the same problem.

**Conquer** the subproblems by solving them recursively. If they are small enough, solve the subproblems as base cases.

**Combine** the solutions to the subproblems into the solution for the original problem.

You can easily remember the steps of a divide-and-conquer algorithm as *divide, conquer, combine*. Here's how to view one step, assuming that each divide step creates two subproblems (though some divide-and-conquer algorithms create more than two):



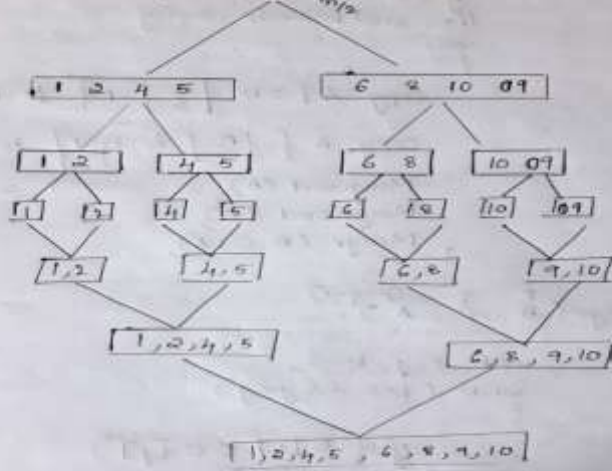
7 (a) Write the merge sort algorithm and explain with example. Also give the recurrence relation of merge sort algorithm and find its time complexity.

[6]	CO2	L4
[4]	CO4	L3
[10]	CO4	L3

70) Merge Sort Algorithm

Let us consider the example of array A having 8 elements

1 2 4 5 6 8 10 11



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Algorithm:

Merge Sort ( $A[0 \text{ to } n-1]$ )

{

if ( $n = 1$ )  
// no need to divide, display

else

{

copy  $A[0 \text{ to } \lfloor \frac{n}{2} \rfloor - 1]$  to B

copy  $A[\lfloor \frac{n}{2} \rfloor \text{ to } n-1]$  to C

Merge sort (B),

Merge sort (C),

Merge (B, C, A);

}

Merge ( $B, C, A$ )  
 $P \quad q \quad (p+q-1)$

{

int  $i, j, k = 0;$

while ( $i < p$  &  $j < q$ )

{

if ( $B[i] \leq C[j]$ )

$A[k] = B[i];$

$i++;$

}

--	--	--

```

else
{
    A[k] = C[j],
    j++
}
k++
}
if (i == -p)
copy C [j to q-1] to A [k to p+q-1]
else
copy B [i to p-1] to A [k to p+q-1]
}

```

Ex

Worst case: When there are elements coming from both B and C alternating.

Best case: When all the elements of B are the smallest and sorted and the array is copied directly to A.

**Time Complexity:**  
 It is found by using Master's Theorem

$$T(n) = aT(n/b) + f(n)$$

$$f(n) = \theta(n^d)$$

$$T(n) = \begin{cases} \theta(n^d) & , a < b^d \\ \theta(n^d \log n) & , a = b^d \\ \theta(n \log_b^a n) & , a > b^d \end{cases}$$

for Mergesort,  $f(n) = n-1$

$$T(n) = 2T(n/2) + (n-1)$$

$$n-1 = n^d \quad a=2 \quad b=2$$

$$d=1$$

$$2 \times 2^1$$

$$2 = 2$$

$$\rightarrow a = b^d$$

$$T(n) \Rightarrow \theta(n^d \log n)$$

$$= \theta(n \log n) \rightarrow \text{which is the time complexity.}$$

8 (a) How can you measure time complexity of algorithm for Fibonacci series using tabular method? Write algorithm and explain?

[5]	CO1	L3
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$$= \frac{n^2 - 5n + 2}{2} \cdot \frac{n-1}{2} \left[ 2n-2 - n-2 \right] = \frac{n(n-1)}{2}$$

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

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

$\therefore O(n^2)$  (order of growth)

8a)

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int f1=0, f2=1;	1	0	1	0	1
int fn;	1	0	1	0	1
for (i=2; i ≤ n; i++)	1	0	n	0	n
fn = f1+f2	1	0	n-1	0	n-1
f1 = f2	1	0	n-1	0	n-1
f2 = fn	1	0	n-1	0	n-1
}					
cout << fn	1	0	1	0	1
}					
				2	
				O(1)	
					$2 + n + 3(n-1)$ $2 + n + 3n - 3$ $4n - 1$

[5]	CO1	L2

(b) Explain important problem types

## 8(b) Important Problem Types

### 1) Sorting:

Sorting problem, refers to arrange the elements in a array in a non-decreasing order. Also there are many Sorting Algorithms, but there is no best solution, as few algorithms may be simple but takes long time, other's may be faster but would be complex. And also all algorithms cannot be used to sort large storage files on disk.

### 2) Searching:

Searching Problem refers to of finding a particular element (Search key) from a group of elements. Also the worst case would be, that the key is at the end of the array or at all there. There are many searching techniques such as Selection Linear search and the most important Binary Search. But there is no algorithm that could take the least time.

### 3) Graph problems:

It is one of the oldest and interesting topic in Algorithmics, where in several points are considered as nodes/ vertices and there are edges b/w them. Graphs can be used for modelling various applications such as communication, transportation. Also it is current active research topic.

### 4) Combinatorial problems.

This is one of the major problem, as there is no solution for this in a ~~am~~ for a small amount of time. The problem arises because, the Combinatorial objects grow very fast in size with increasing size of the problem, to an extent that cannot be imagined of. The solution cannot be found in limited amount of time.

### 5) String Processing problems:

It occurs, where there is conjunction in computer language and in social activities. It is majorly seen in Computer Science.

String is a sequence of characters / alphabets.