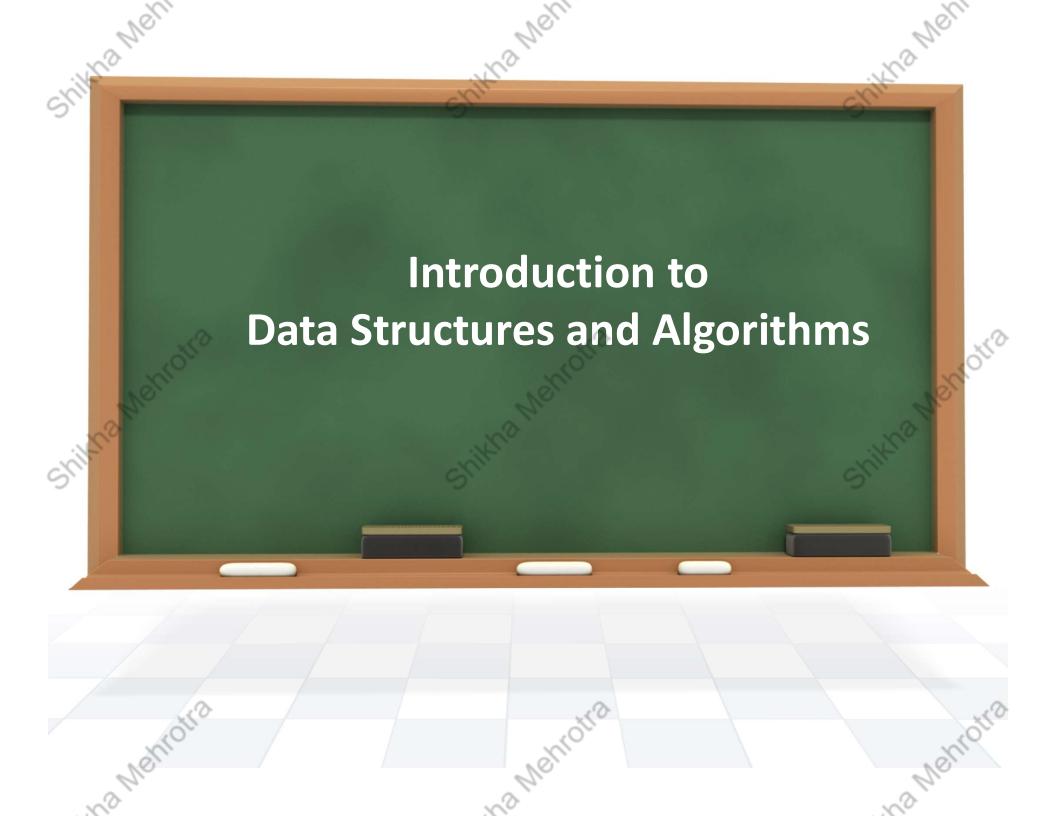
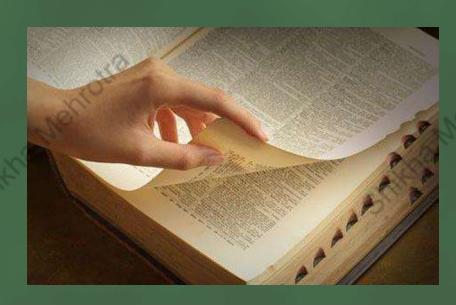


Shikha Mehrotra



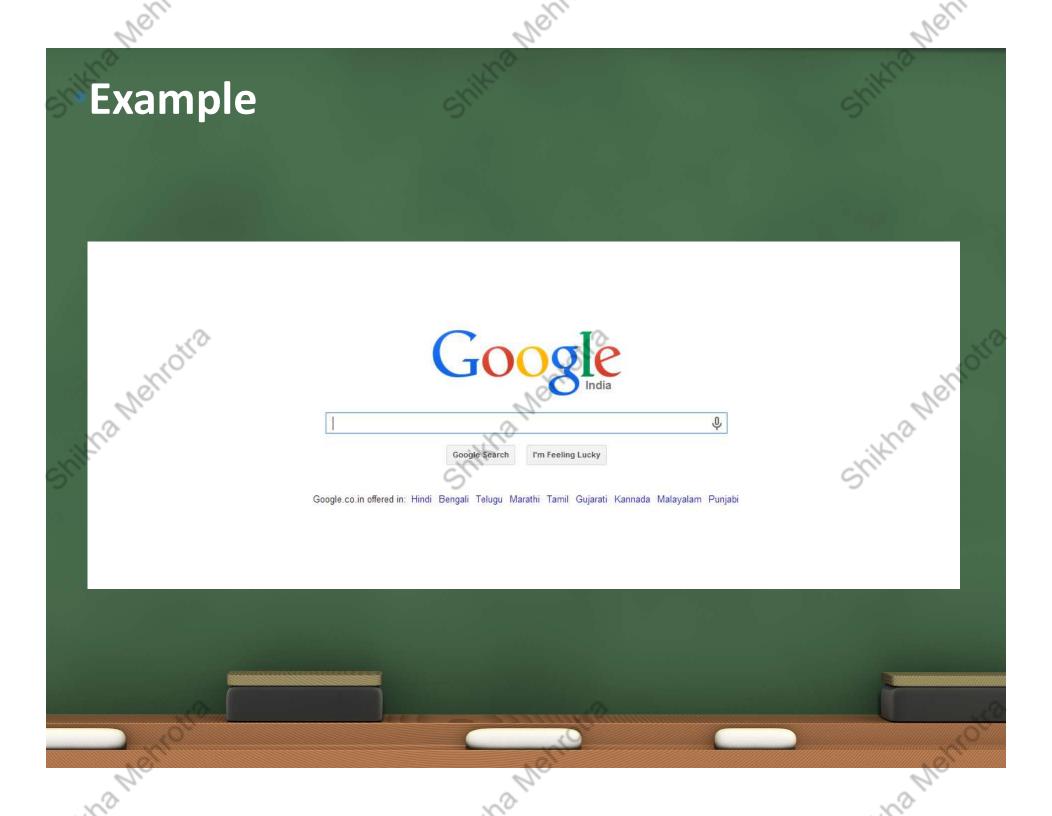
Introduction to Data Structures and Algorithms



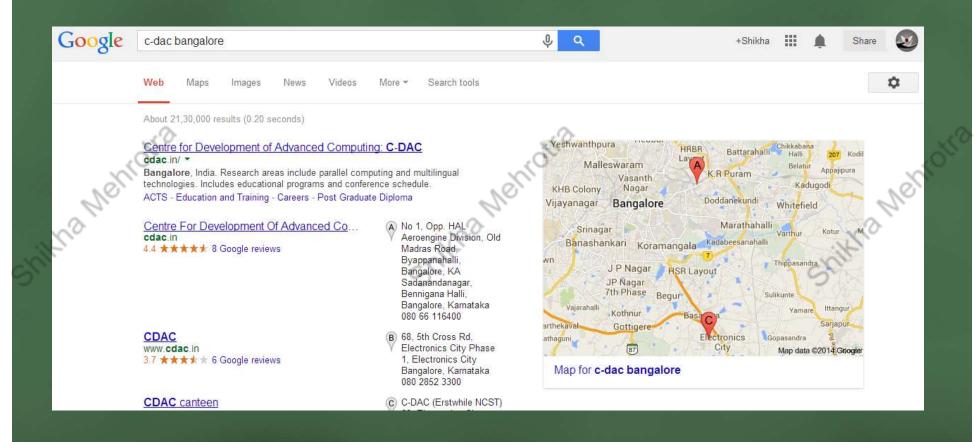
Dictionary

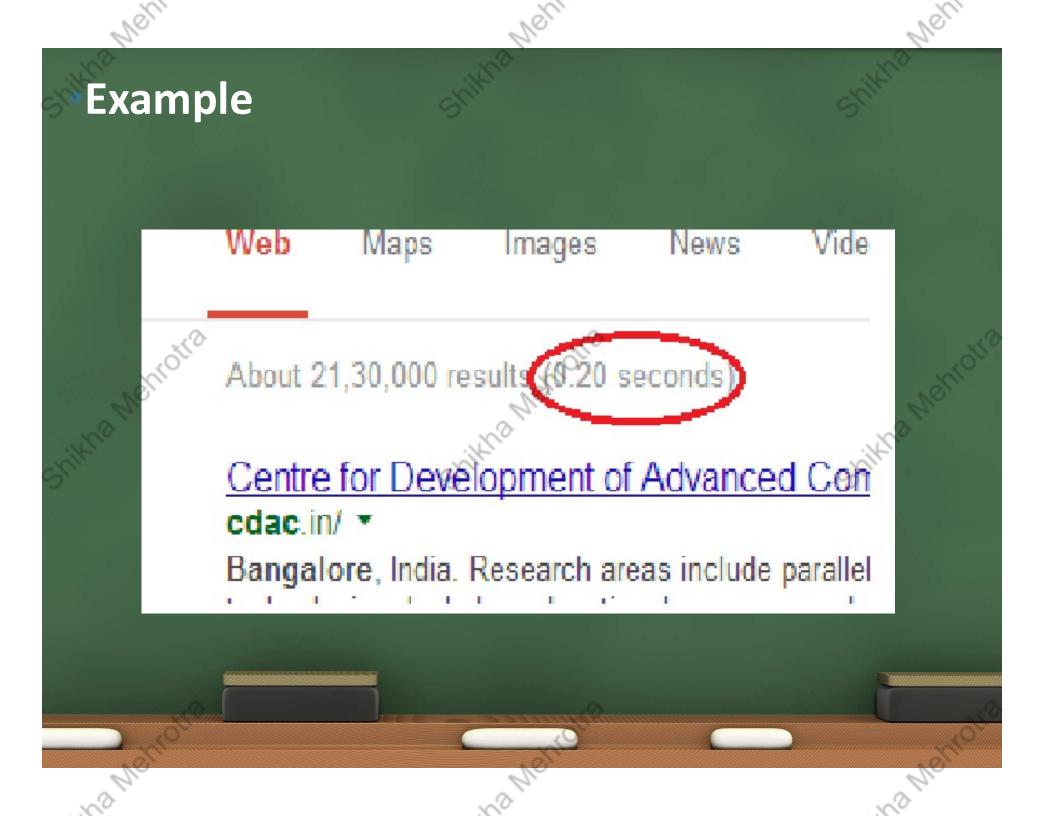
ABC Hardware Cash Book - 03/01/2013 to 03/31/2013

S. no.	Date	Particulars	Debit	Credit
1	03/01/2013	Opening balance		50000
2	03/02/2013	Transport bill	2000	~ (°
3	03/07/2013	Goods sales		1500
4	03/08/2013	Bank Loan	. X	5000
5	03/15/2013	Goods sales	- H	1000
6	03/17/2013	Electiricty bill	1200	
7	03/21/2013	Good sales	7	1200
8	03/25/2013	Hardware purchase	500	
9	03/29/2013	Employee salary	20000	
10	03/31/2013	Closing Balance	35000	
		Total	58,700	58,700



Example





For Developing Software

Media Player



For Developing Software Games

For Developing Software Multimedia

For Developing Software

Stock Analysis









For Developing Software Browser

For Developing Software



Algorithm

Data Structures



Media Player



Games



Multimedia



Stock Analysis



Browser

Data Structures

A data structure is a way to store and organize data in a computer, so it can be used efficiently

We talk about data structures as:

- Mathematical / Logical models (ADT)
- 2. Implementation

ADT (Abstract Data Type)

- <u>Car</u>
- Engine,
- Gear.
- Petrol tank,
- Tires, etc.
- **Operations:**
- StartEngine()
- Drive()
- ApplyBreaks()
- IsPetrolAllow()



Array • Create() • Get() • Set() • Remove() ADT Operations program Data Structure

Data Structures



- Turned On/Off
- Receive Signals Play Audio / Video

Abstract View

ADT

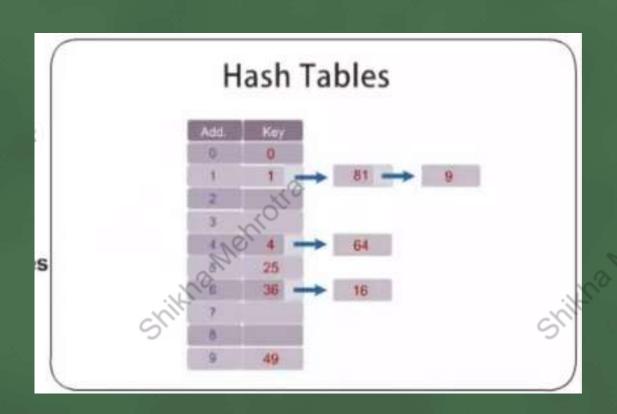
- Modify elements by position

Concrete Implementation

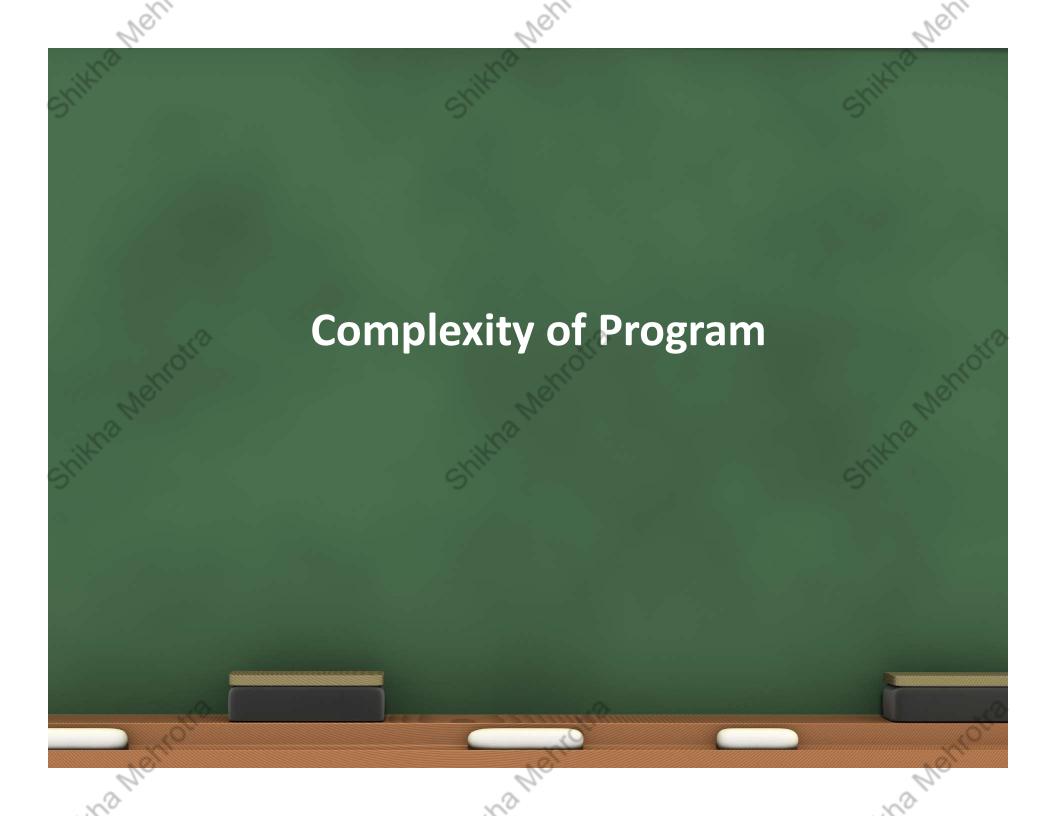
Data Structure Organize the Data Linearly Circularly Hierarchically

Different types of Data Structures

- Arrays
- Stacks
- Queues
- Linked List
- Graphs
- Trees
- Hash Tables



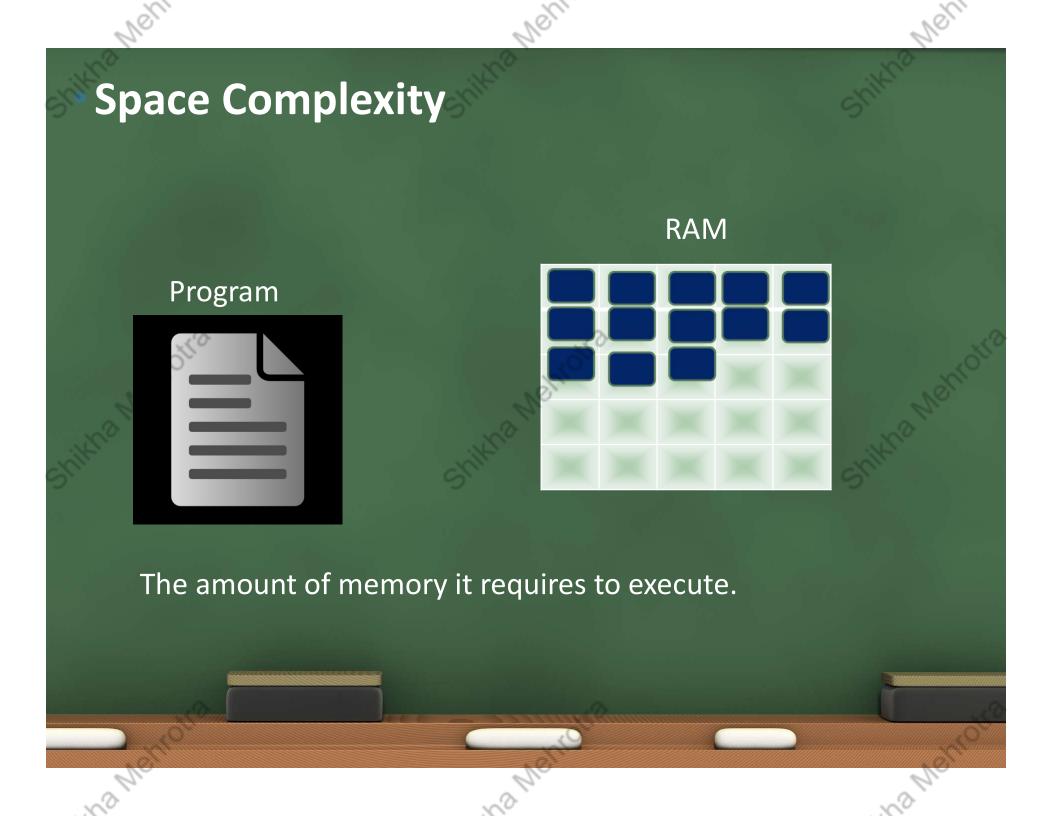
Different types of Data Structures Arrays Algorithms Stacks Queues **Traversal** Linked List Searching Graphs Sorting Trees Hash Tables

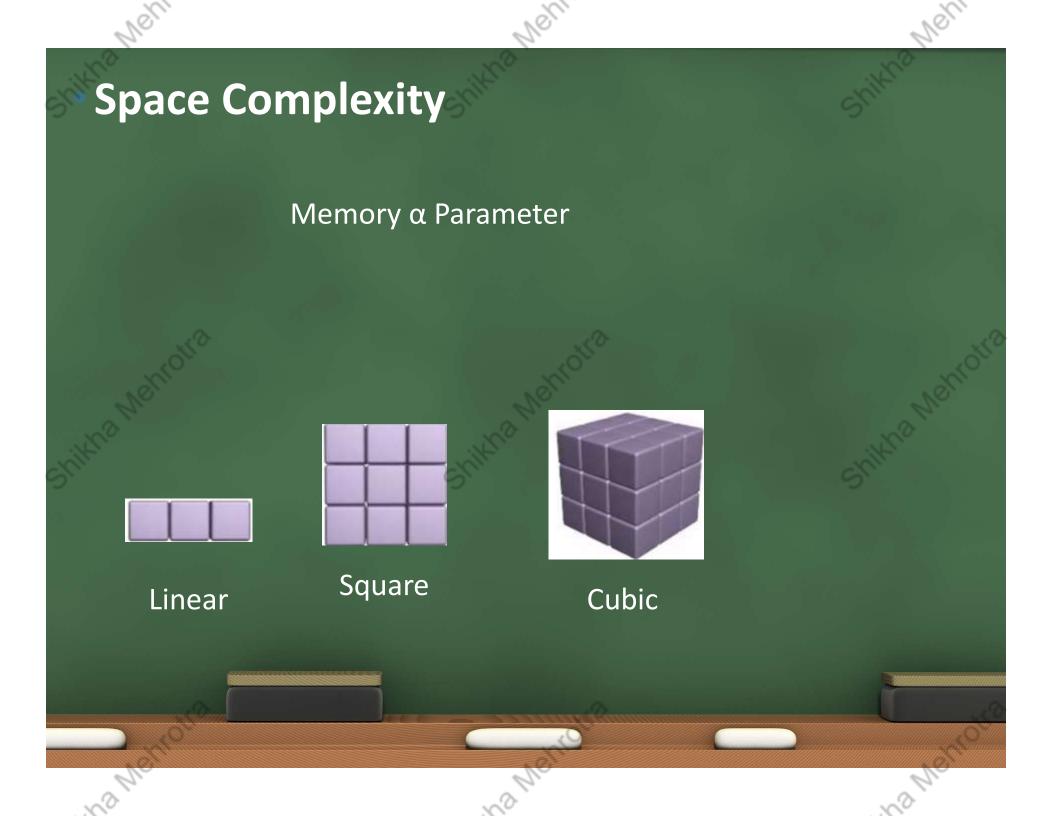


Complexity/ Efficiency of Program

When a program/algorithm is better than other?

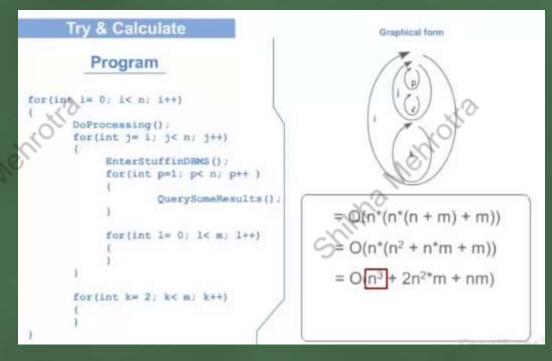
- What are the measures of comparison?
- Space Required Space Complexity
- Time Required Time Complexity



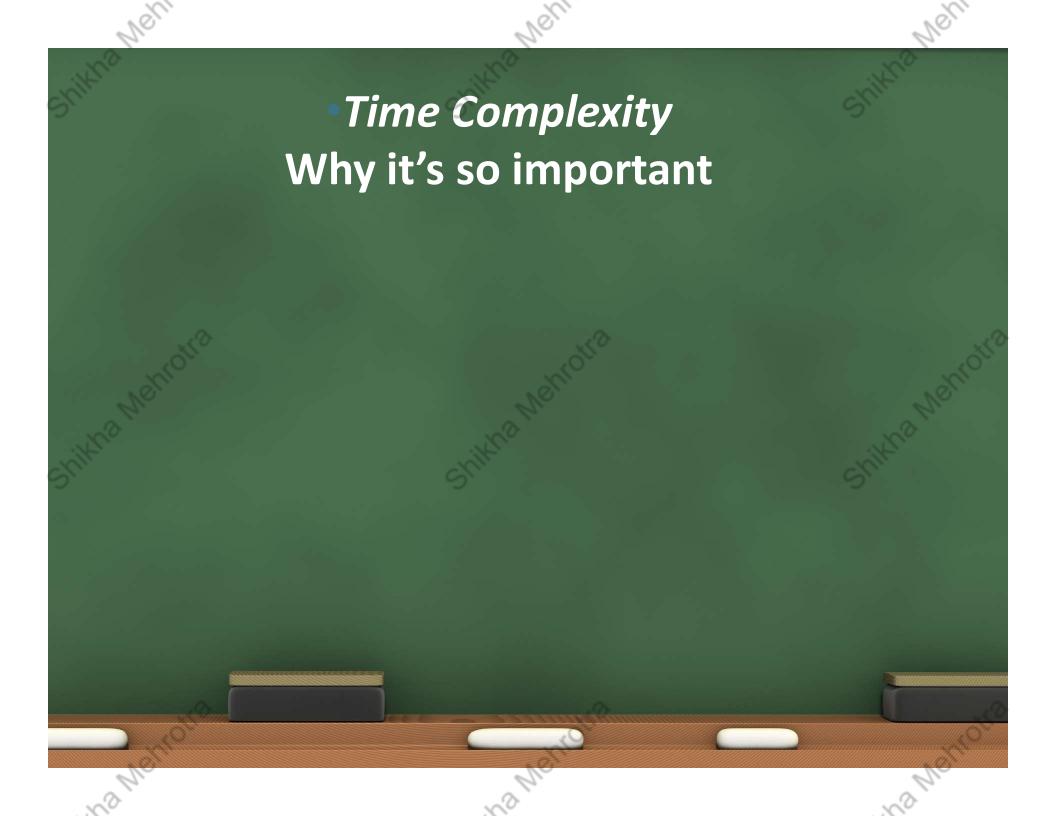


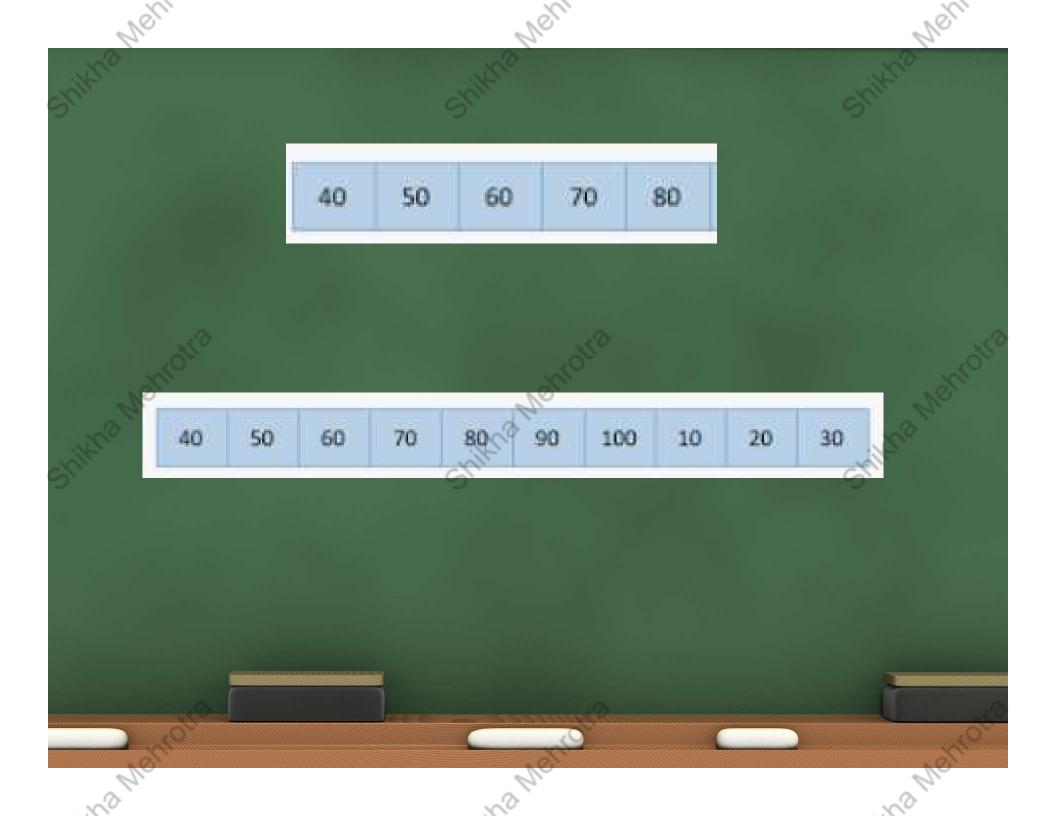
Time Complexity

The amount of computer time it requires to execute the program.



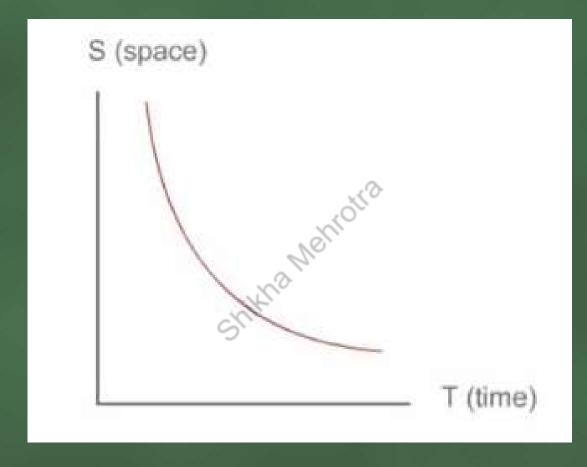
 $O(n^3)$



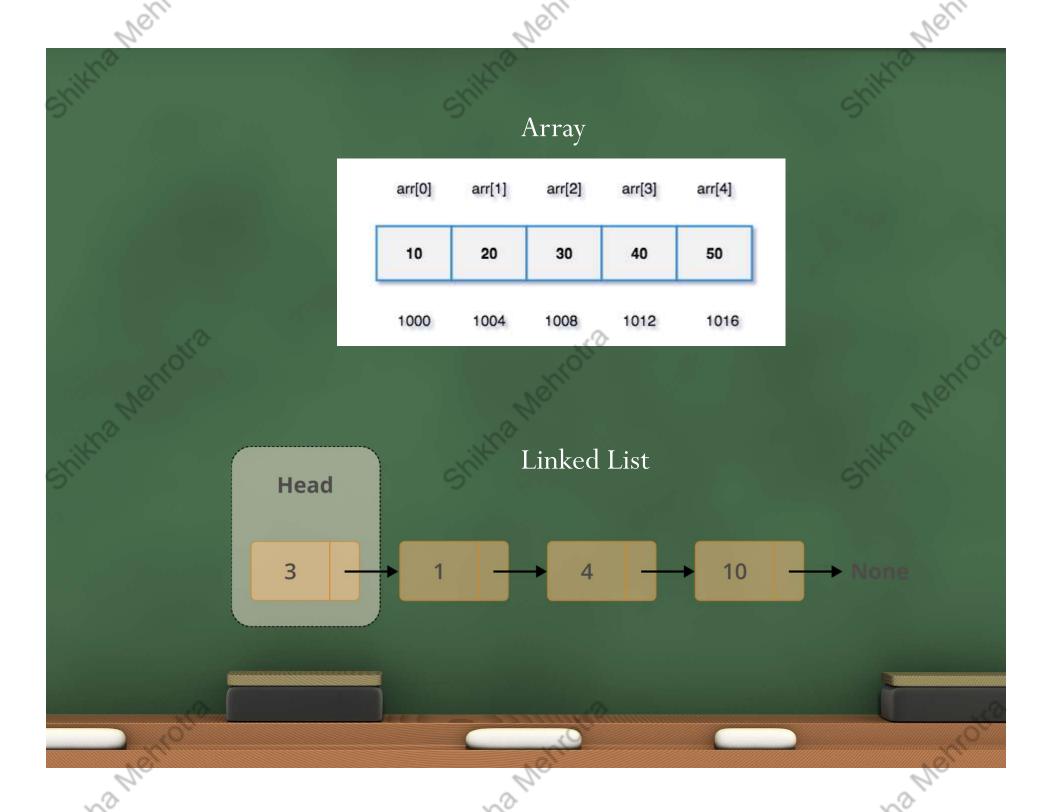




Less time to execute the program requires more space and vice versa



So there is a trade off between time and space complexity



How to analyze Time Complexity

Running time depends upon

- χ > Single vs. Multi Processor
- Read/Write Speed to Memory
- x > 32 bit vs. 64 bit

How to analyze Time Complexity

Running time depends upon

- Single vs Multi Processor
- Read/Write Speed to Memory
- > 32 bit vs 64 bit
- > Input





Complexity Analysis

1. Worst Case analysis

The maximum time needed over all inputs

1. Best Case analysis

The minimum time needed

1. Average Case analysis

average time needed

Usually only worst-case information is given since average case is much harder to estimate

 Complexity Notations (The Big O) (Omega) (Theta)

The Big O Notation

☐ Is used for worst cast analysis

An algorithm is O(f (N)) if there are constants c and N₀, such that for N \geq N₀ the time to perform the algorithm for an input size N is bounded by t(N) < c f(N)

- Consequences
 - O(f(N)) is identically the same as O(a f(N))
 - \rightarrow O(aNx + bNy) is identically the same as O(N max(x,y)
 - > O(Nx) implies O(Ny) for all y ≥ x

and Θ Notations

- \square Ω is used for best case analysis:
- An algorithm is $\Omega(f(N))$ if there are constants c and N0, such that for N \geq N0 the time to perform the algorithm for an input size N is bounded by t(N) > c f(N)
- Θ is used if worst and best case scale the same
 - \triangleright An algorithm is $\Theta(f(N))$ if it is $\Theta(f(N))$ and O(f(N))

Ω

We analyse time complexity for

- Very large input size
- worst case scenario

Algo 1:
$$T(n) = 5n^2 + 7$$

Algo 2:
$$T(n) = 17n^2 + 6n + 8$$

Quardratic rate of growth

_Useful rules

- □ simple statements (read, write, assign)
 - O(1) (constant)
- □ simple operations (+ * / == > >= < <=)</p>
 - O(1)
- sequence of simple statements/operations
 - rule of sums
- □ for, do, while loops
 - rules of products

Two important rules

- □ Rule of sums
 - if you do a number of operations in sequence, the runtime is dominated by the most expensive operation

i.e.

- □ Rule of products
 - if you repeat an operation a number of times, the total runtime is the runtime of the operation multiplied by the iteration count

Function () int a; a++; forli = 0; i < n; i++)

Notif
// simple statements notified for (i = 0; i < m; i ++) { for (i=0; i < m; i++) 11 Simple Statements

T(n) = O(1) + O(n) + O(n2)= O(n2)

Function () if (some Condition) for (i = 0; i < n'; i + +)

d 11simple statements (0(n))

state

for (i = 0; i < n'; i + +)

for (i = 0; i < n'; i + +)

for (j = 0; j < n'; i + +)

for (j = 0; j < n'; i + +)

for (j = 0; j < n'; i + +)

for (j = 0; j < n'; i + +)

for (j = 0; j < n'; i + +)

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for (j = 0; j < n'; i + +)

for (j = 0; j < n'; i + +)

for (j = 0; j < n'; i + +)

for (j = 0; j < n'; i T(n) keep one drop coef n^2 $3n^2 + 4n + 1$ $3 n^2$ $101 \text{ n}^2 + 102$ 101 n^2 $15 \text{ n}^2 + 6\text{n}$ 15 n^2 a n²+bn+c a n²

How to analyze Time Complexity-Example-1

```
Sum (a,b)
{
     return a+b
}
1
1
```

$$Tsum = 2$$

Constant time

How to analyze Time Complexity-Example-2

$$T_{sumOfList} = 1+2(n+1) + 2n +2$$

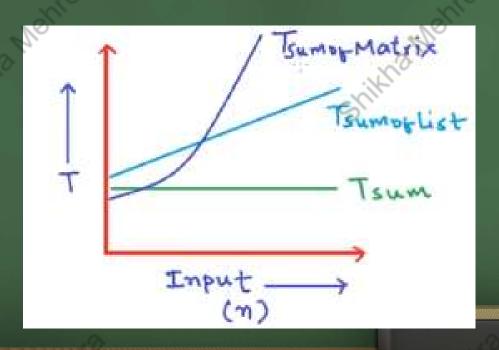
= 4n +4
 $Cn + C$

How to analyze Time Complexity

 T_{sum} O(1)

 $T_{\text{sumOflist}} = Cn+C$ O(n)

 $T_{\text{sumofMatrix}} = an^2 + bn + c O(n^2)$



Time Complexity Analysis – Fibonacci Sequence (recursion)

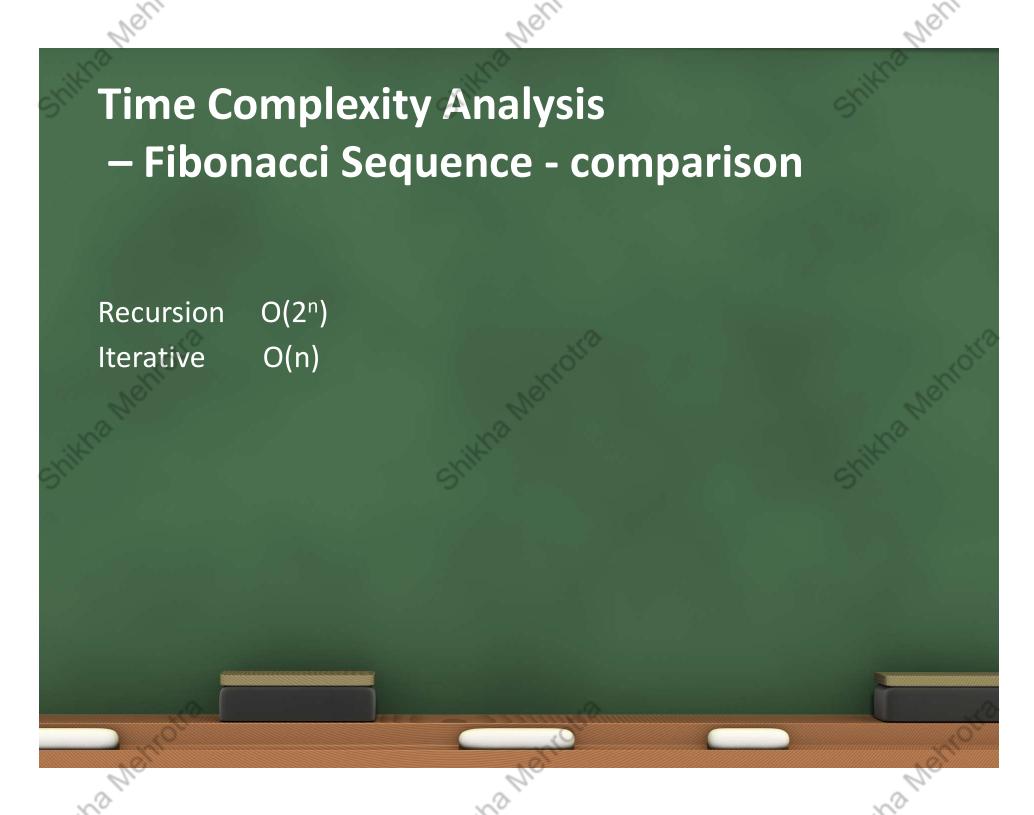
```
0 1 1 2 3 5 8 .....
Fib(n)
{
   if n<=1
      return n
   else
      return Fib(n-1) + Fib(n-2)
}</pre>
```

```
T(n)=T(n-1) + T(n-2) + 4
For n<=1
T(0)=T(1)= 2 \text{ (constant)}
T(n-2) = T(n-1) \text{ (<)}
T(n)= 2T(n-1) + C \text{ (c=4)}
= 2 \{ 2T(n-2) + C \} + C
= 4T(n-2) + 3C
= 8T(n-2) + 7C
= 2^{k} T(n-k) + (2^{k}-1) C
n-k=0 ==> k=n
T(n)=2^{n} T(0)+(2^{n}-1)C
T(n)=(1+C) 2^{n} - C
```

Time Complexity Analysis - Fibonacci Sequence

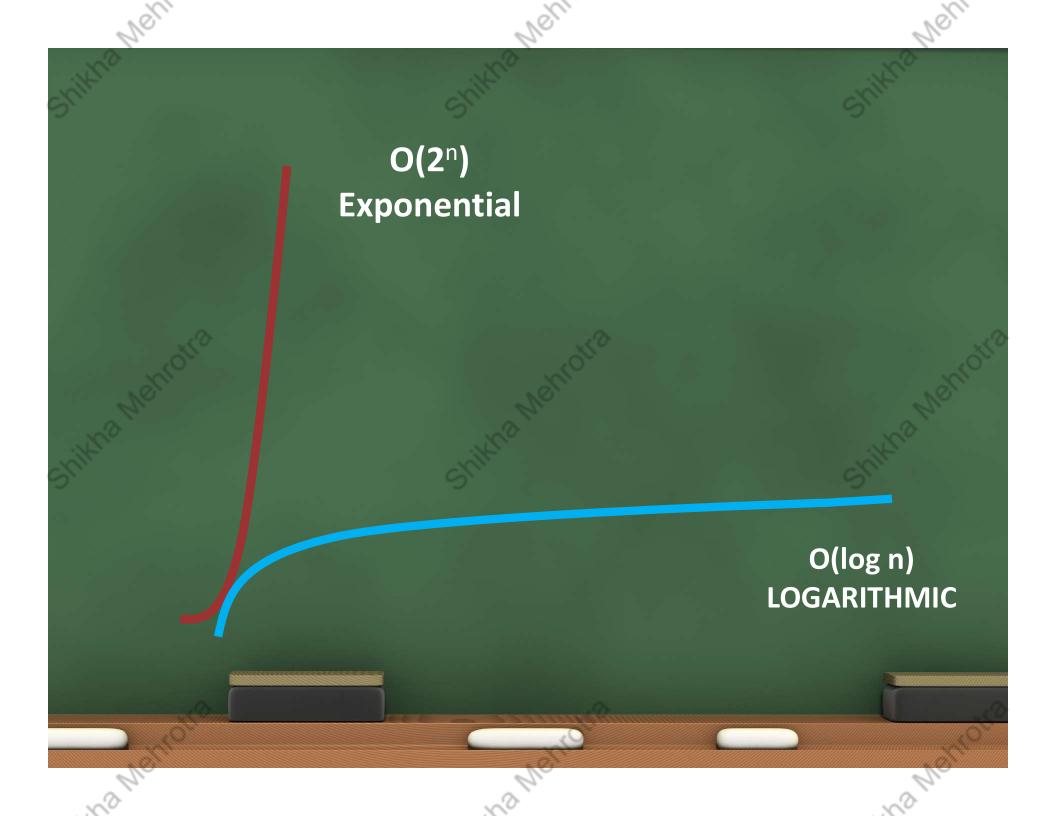
```
F(n)
  if n=0 then
          return 0
if n=1 then
           return 1
 else
 for(int i=0; i<n; i++)
      sum=a[n-1] + a[n-2];
       a[n-2]=a[n-1];
       a[n-1]=sum;
  return sum;
```

O(n)

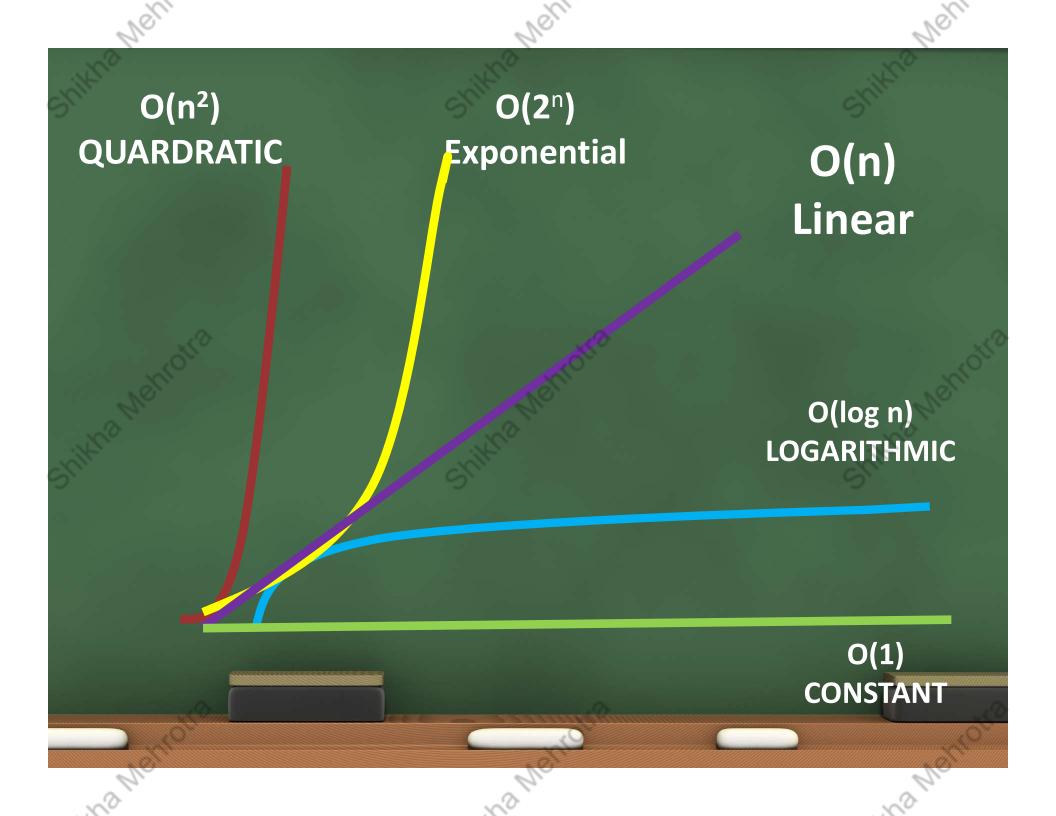


O(n²) QUARDRATIC O(n) Linear

O(n) Linear O(log n) LOGARITHMIC



O(1) Constant O(log n) Logarithmic O(n) $O(n^2)$ Quadratic O(2ⁿ) Exponential



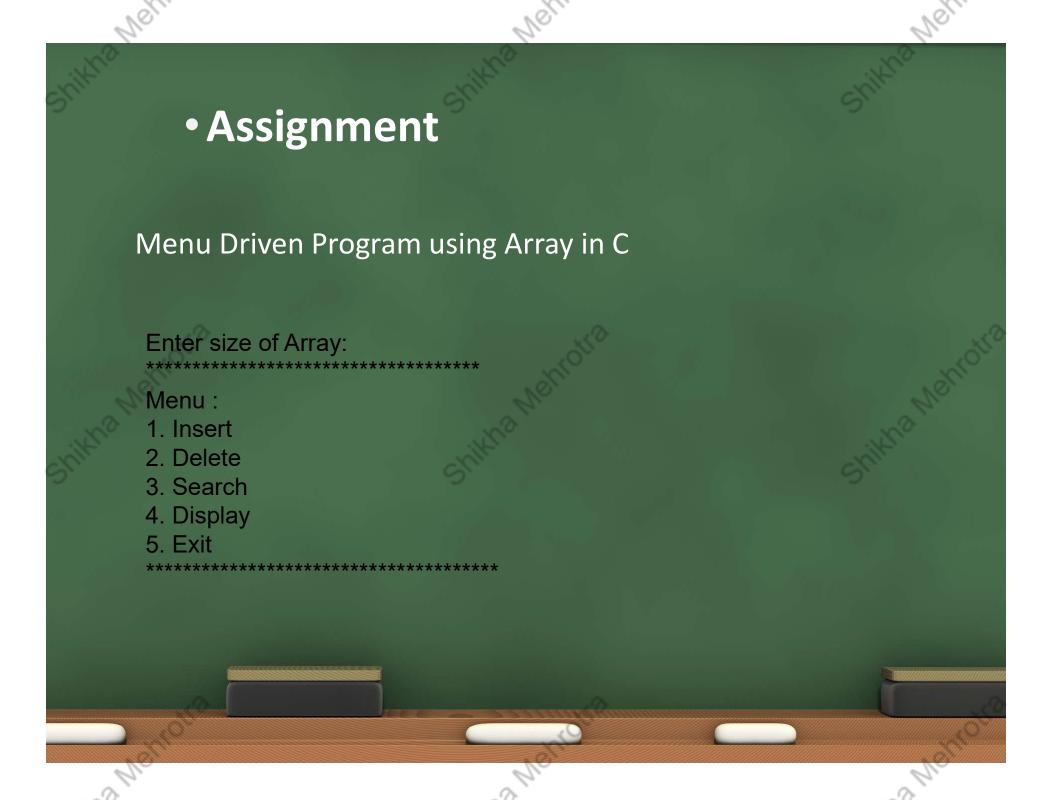
Time Complexity O(n log n) O(h!) CPU Operations O(log_n)>
O(1) Input Size

Which Algorithm do you prefer?

Algorithm A		A	Algorithm B		Which do you pick?		
	O(log N)		O(N)			А	
	O(log N)		O(Nlog N)			А	
	O(Nlog N)		O(N)	50		В	
	O(log N)		O(N ²)			Α	
	O(N!)		O(2 ^N)			В	
	O(2 ^N)		O(N ²)			В	
			31,				•

Complexity – Exercise-1

```
max=0;
for(i=0; i<n; i++)
{
    read(number)
    if ( max < number )
        max=number;
}
print(max);
}</pre>
```



Hash Tables Queues Linked List Trees Graphs Stacks Array