

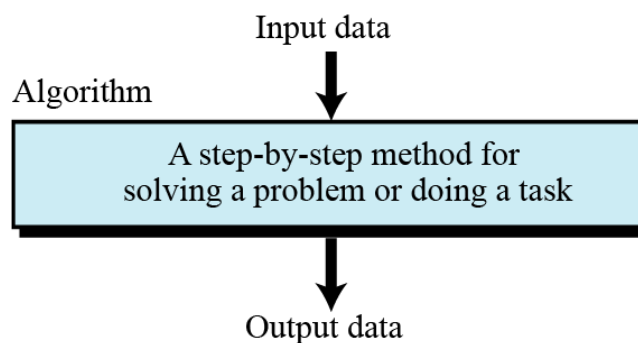
DESIGN AND ANALYSIS OF ALGORITHMS (DAA)

Module 1,2

Radhika Chapaneri

An informal definition of an algorithm is:

Algorithm: a step-by-step method for solving a problem or doing a task.



Definition

- An algorithm is a finite set of instructions that, if followed, accomplishes a particular task.
- An algorithm is any well defined computational procedure that takes some value, or set of values, as input and produces some value, or set of values, as output.
- An algorithm is thus a sequence of computational steps that transform the input into the output.

What is Analysis of algorithms

- **Analysis of Algorithms** is the area of computer science that provides tools to analyze the efficiency of different methods of solutions.



Properties of Algorithms

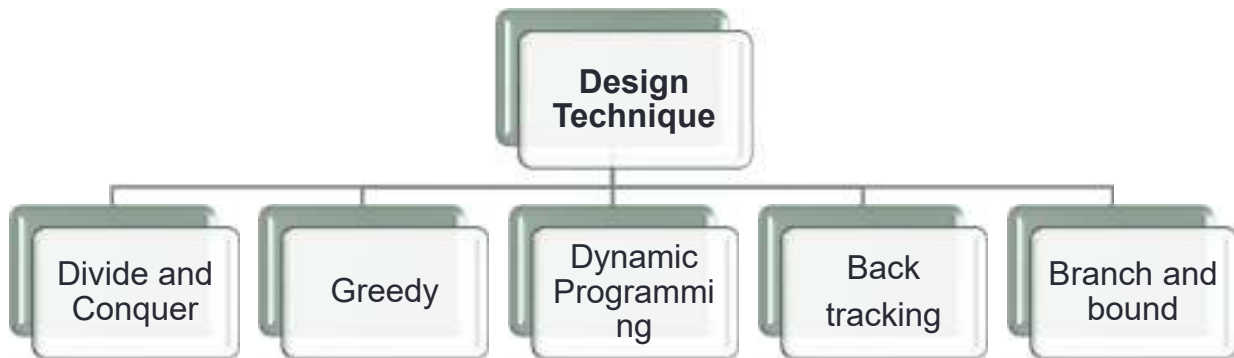
- **Input** : an algorithm accepts zero or more inputs
 - **Output** : it produces at least one output.
 - **Finiteness**: If we trace out the instructions of an algorithm, then for all cases, the terminates after a finite number of steps.
 - **Definiteness**: Each step in algorithm is unambiguous. This means that the action specified by the step cannot be in multiple ways & can be performed without any confusion.
- Effectiveness**: It consists of basic instructions that are realizable.
The operations are doable

Algorithm design techniques

- **Brute force**: For many nontrivial problems, there is a natural brute-force search algorithm that checks every possible solution.
- Typically takes 2^n time or worse for inputs of size n .
- Unacceptable in practice.

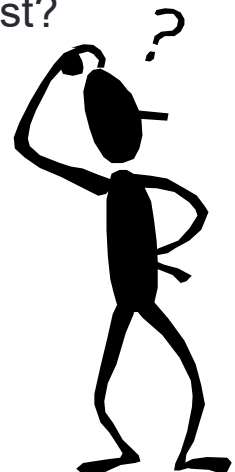


Algorithm design techniques



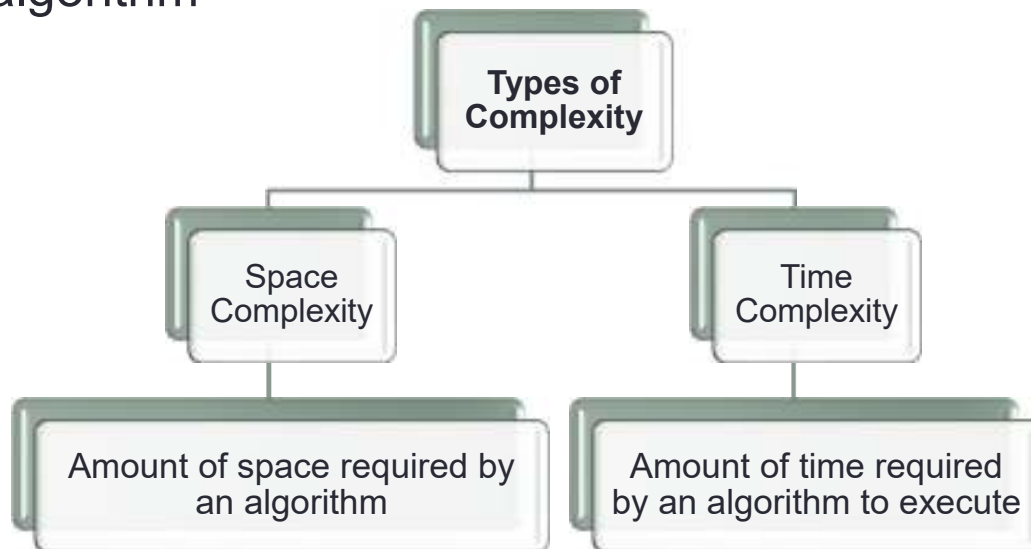
Performance analysis

- When is a program / algorithm said to be better than another?
- The algorithms are correct, but which is the best?
- What are the measures for comparison?



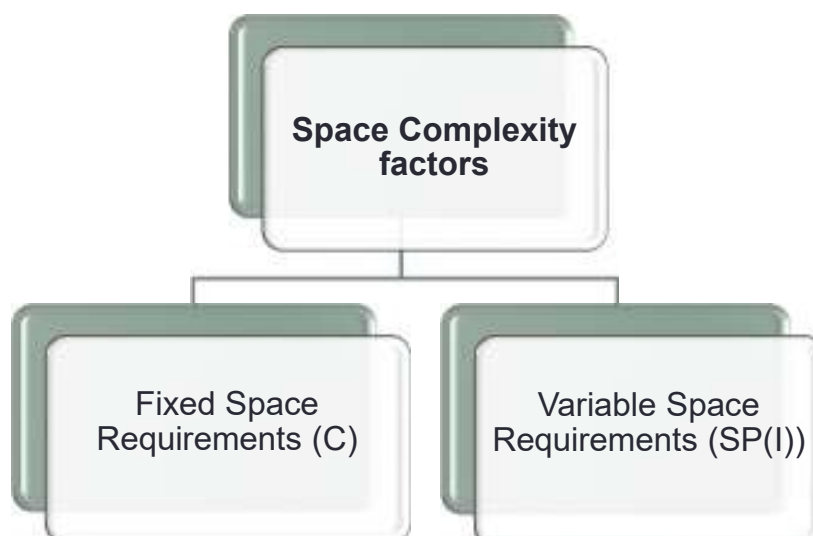
Performance analysis

- The computational complexity of an algorithm addresses the resources needed to run the algorithm



Space Complexity

- **Space complexity** = The amount of memory required by an algorithm to run to completion



Space Complexity

- **Fixed Space Requirements (C) : Independent of the characteristics of the inputs and outputs**
 - instruction space
 - space for simple variables
 - fixed-size structured variable
 - Constants
- **Variable Space Requirements (SP(I)) : depend on the instance characteristic I**
 - number, size, values of inputs and outputs associated with I
 - recursive stack space, local variables, return address

Space Complexity

Total space complexity:

$$S(P) = C + S_P(I)$$

Where

C = Fixed Space Requirements

$S_P(I)$ = Variable Space Requirements

S(P) = Total space complexity

Space Complexity : Simple Example

// This algorithm computes addition of three elements

// Input: a, b, c are of floating type

// Output: The addition is returned

Algorithm:

```
Add (a, b, c)
{
    return a + b + c
}
```

Space Complexity:

$$S(P) = C + S_P(I)$$

$$S_P(I) = 0, S(P) = C$$

If we assume that a, b and c occupy one word size then
 $S(P) = 3$

Space Complexity : Using array

// This algorithm computes addition of all the elements in an array.

// Input: Array x of floating type, n is total number of elements

// Output: returns sum which is of data type float

Algorithm:

```
{
    sum = 0.0;
    for i = 1 to n do
        sum = sum + x[i]
    return sum
}
```

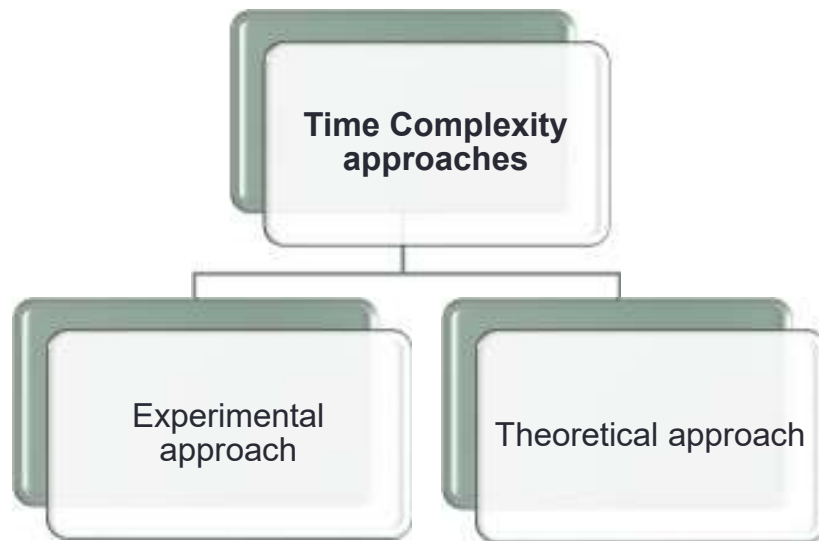
Space Complexity:

$$S(P) = C + S_P(I)$$

The 'n' space required for x[],
 one unit space for n,
 one unit for I and
 one unit for sum.

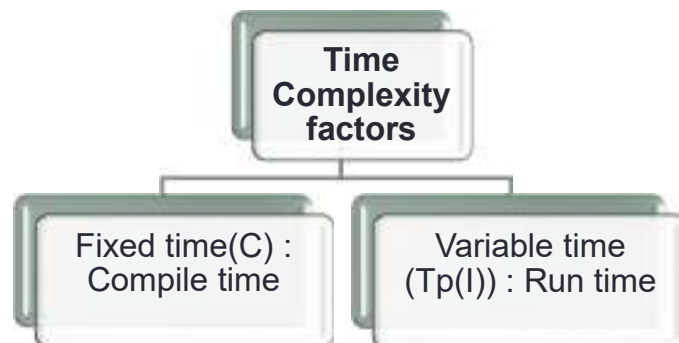
$$S(P) \geq (n+3)$$

Time Complexity



How to find time complexity?

$$T(P) = C + T_p(I)$$



- The compile time does not depend on the instance characteristics.
- We assume that a compiled program run several times without recompilation
- Hence we are concern just about the running time of the program.

Time Complexity : Using array

// Algorithm computes addition of all the elements in an array.

// Input: Array x of floating type, n is total number of elements

// Output: returns sum which is of data type float

Algorithm:

```
{
  sum = 0.0;
  for i = 1 to n do
    sum = sum + x[i];
  return sum;
}
```

Algorithm	Steps/ execution	Frequ ncy	Total steps
Algorithm Sum(a[], n)	0	-	0
{	0	-	0
sum = 0.0;	1	1	1
for i = 1 to n do	1	n + 1	n + 1
sum = sum + x[i];	1	n	n
return sum;	1	1	1
}	0	-	0
Time Complexity step per execution × frequency			2n + 3

Time Complexity : Matrix addition

// This algorithm computes addition of all the elements in matrix

// Input: two matrices a, b

// Output: returns sum in matrix

Algorithm:

Add(a,b,c,m,n)

```
{
  for i = 1 to m do
    for j = 1 to n do
      c[i, j] = a[i, j] + b[i, j]
}
```

Algorithm	s/e	f	Total steps
Add(a,b,c,m,n)	0	-	0
{	0	-	0
for i = 1 to m do	1	m + 1	m + 1
for j = 1 to n do	1	m (n+1)	mn + m
c[i, j] = a[i, j] + b[i, j]	1	mn	mn
}	0	-	0
Time Complexity			2mn + 2m + 1

Order of growth

Time complexity of algorithm A and B

Input	Algo1 = $100n$	Algo B = $10n^2$	Algo C = 2^n
n	$100n$	$10n^2$	2^n
5	500	250	32
10	1000	1000	1024
20	2000	4000	1,048,57
50	5000	25,000	1.259×10^{15}
1000	100,000	10,000,000	$> 10^{300}$

Order of growth

- If an algorithm need 2^n steps for execution, then when $n = 40$, the number of steps needed is approximately 1.1×10^{12} .
- On a computer performing one billion steps per second, this would require about 18.3 minutes for $n = 40$.
- If $n = 50$ the same algorithm would run for about 13 days on this computer.
- When $n = 100$ about 4×10^{13} years are needed.
- So utility of algorithm with exponential complexity is limited to small n ($n < 40$)