

Algorithm

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- 2009 —
- GATE
- placement

Algorithm

- Design and Analysis of Algorithm

1

Mathematical

Greedy

Divide & Conquer

Dynamic

Graph

Introduction

Analysis

Aspect tool
Asymptotic Notation

Syllabus

Section 5: Algorithms

Data
structure

. Best case
. Average case (Some)
case

- Searching, sorting {hashing} Asymptotic [worst case time] and space complexity. [Algorithm design techniques: greedy, dynamic programming and divide-and-conquer.] [Graph traversals] minimum spanning trees, [shortest paths]

Application - DFS/BFS

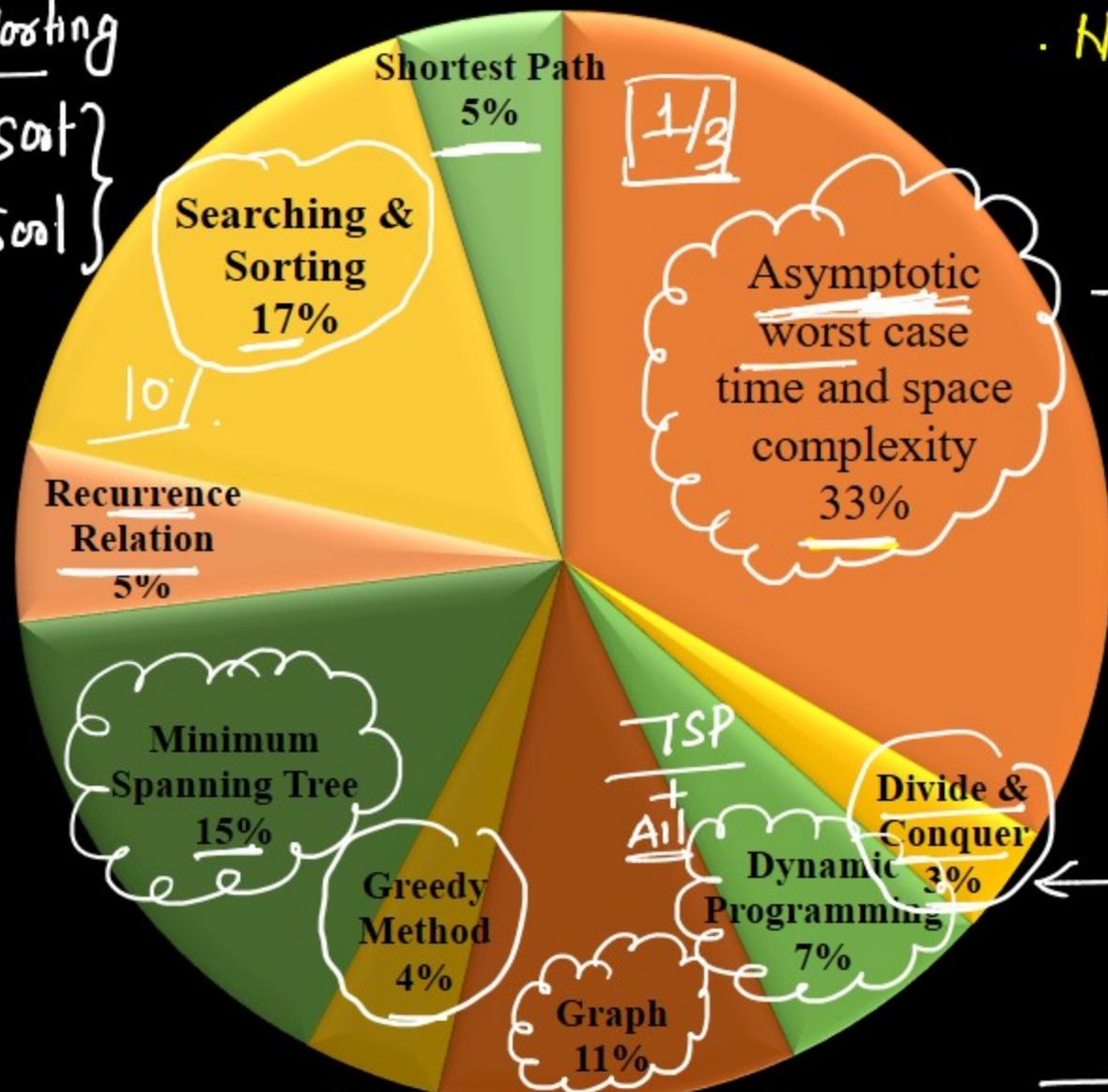
Greedy Method

Data structure:
Explain :-

- Backtracking
- Branch & Bound

Weightage Analysis

Basic Sorting
. quick sort
. Merge Sort



No. of questions asked till 2021

— question from data structure

BST + Heap

work sheet

{ work + walk book
+ text book
+ GQB }

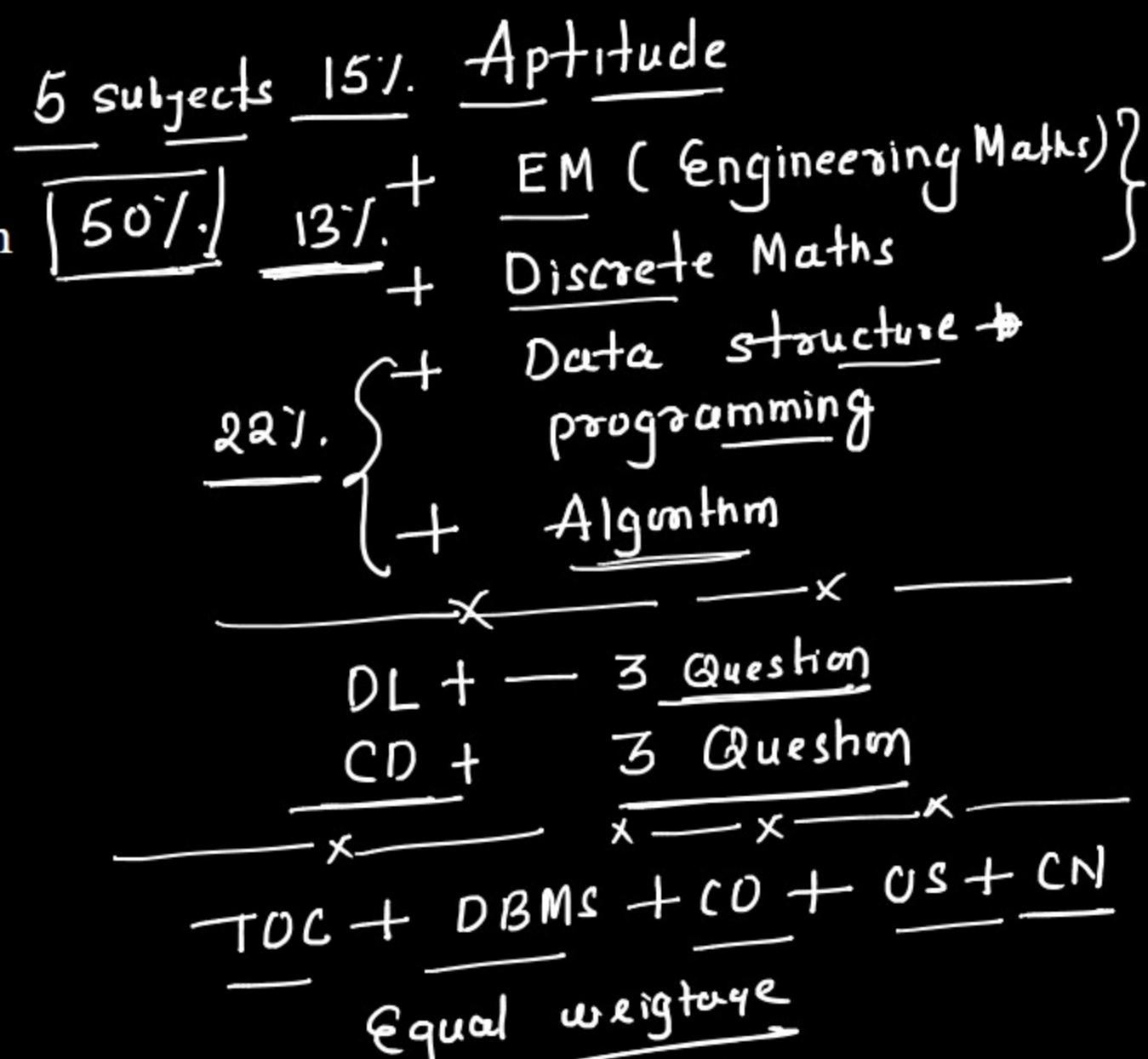
Master Method

Recurrence Relation

→ 31% + quick/Merge + Recurrence Relation
→ 18%
→ 41% + MST + shortest path

Prerequisite

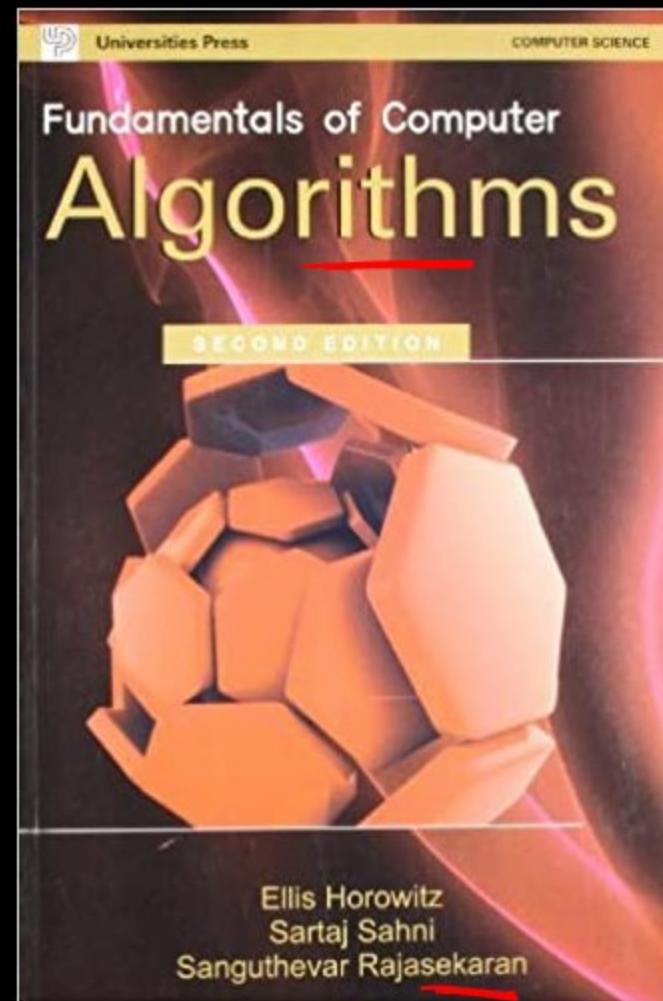
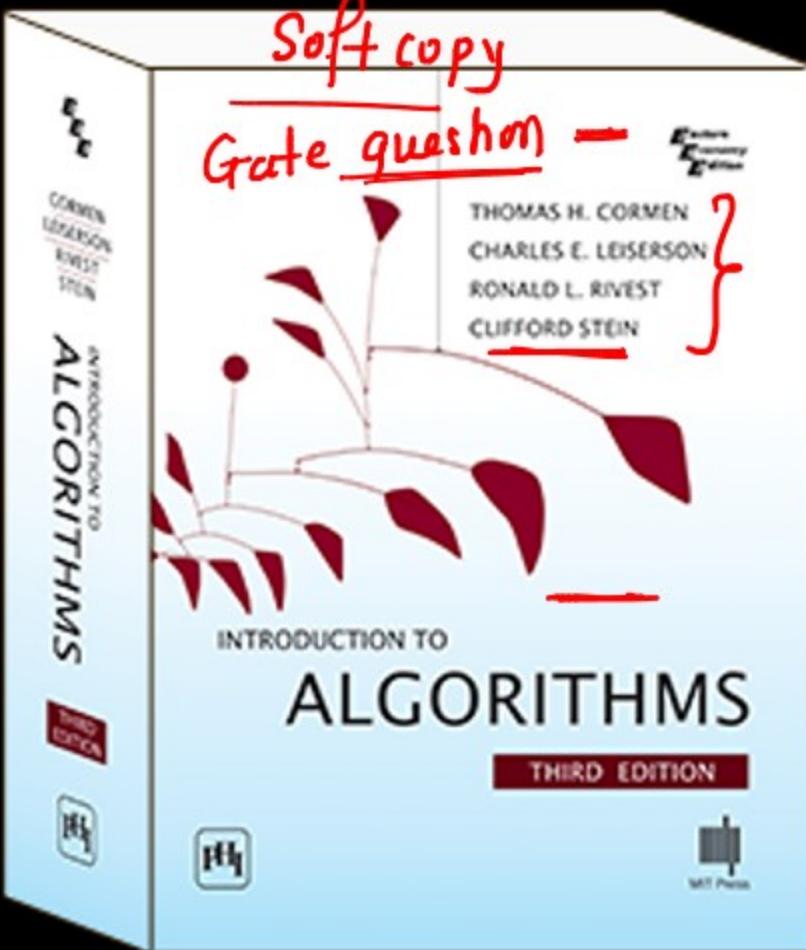
- I will cover DS - aspect
- Summation Series, Logarithms, Induction
- Probability
- Discrete Mathematics -
 - Function
 - Graph

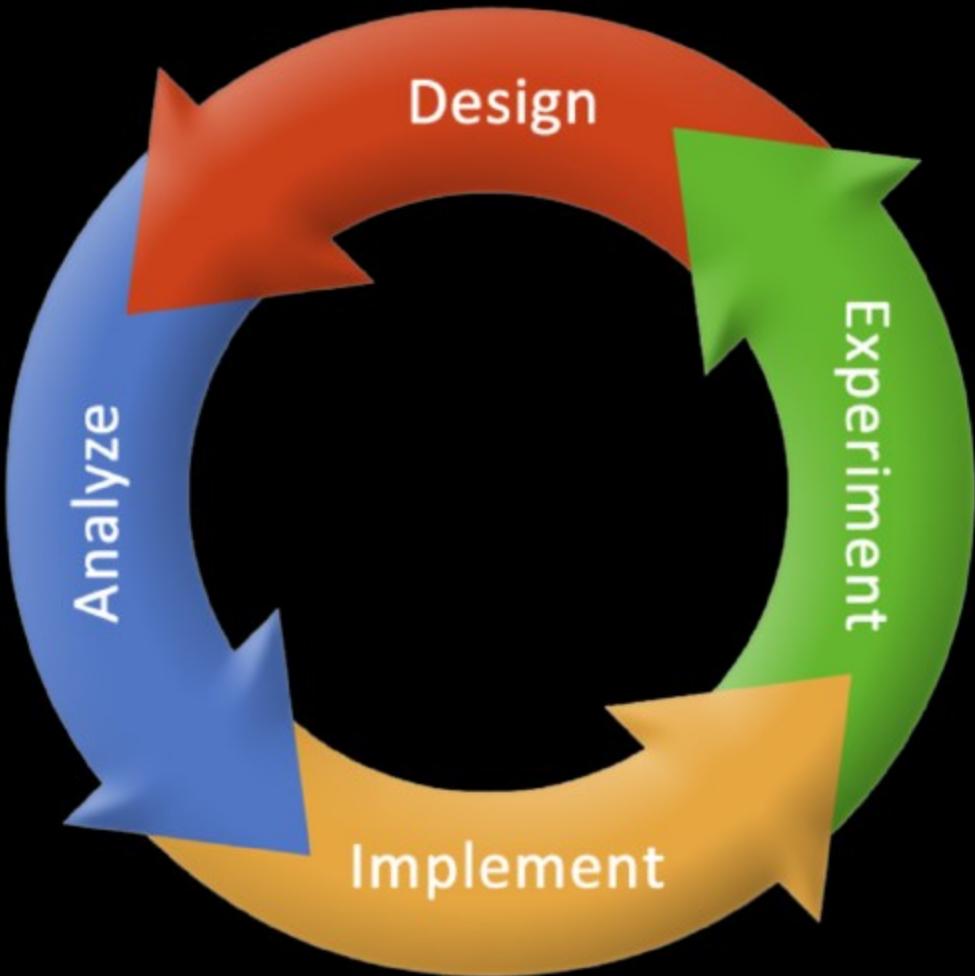


Topics to be Covered

- Design and Analysis of Algorithm

Books :- Soft





Algorithmic Analysis

Analysis - Result Analysis

Paper Analysis

~~At~~ Syllabus Analysis

Expected No. of marks : 8-10

33 %

marks

Introduction

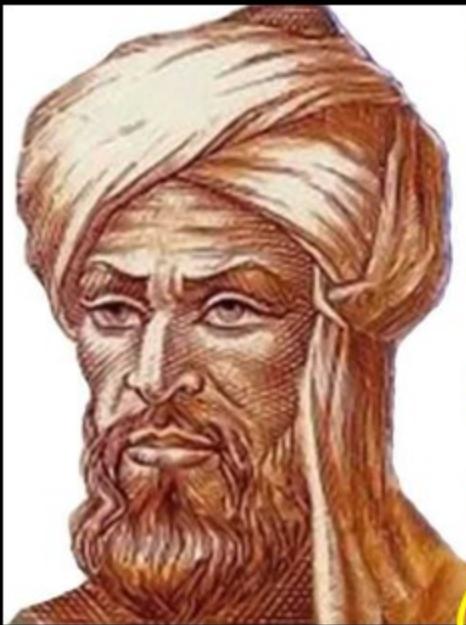
- Definition, Properties + ~~TOE~~ — theory of computation
- Why Analysis is Required
- How to Analysis
- Types of Analysis
- Framework of Analysis + Start An

(7.30)

+
Asymptotic Notation +
Comparison of function

Recurrence Relation + [Summation & Series]
+ Solving
+ Mathematical foundation

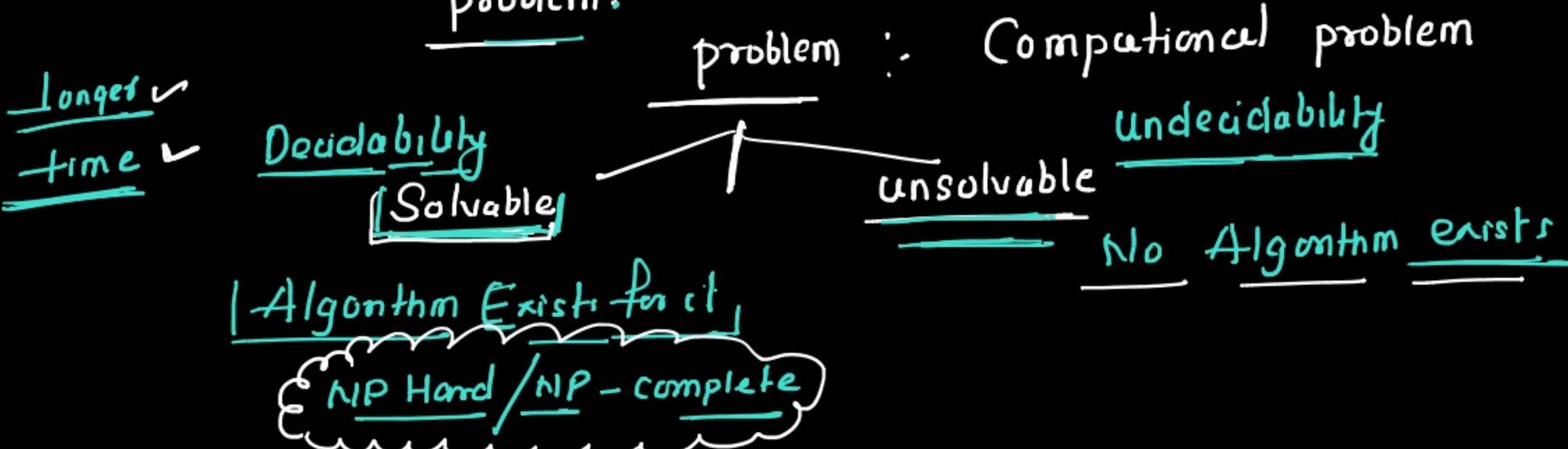
Historical Background



One of Khwārezm's most famous residents was Muhammad ibn Mūsa (al-Khwarizmī), an influential 9th century Persian scholar, astronomer, geographer, and mathematician known especially for his contributions to the study of algebra. Indeed, the (latinization) of his name, which meant ‘the native of Khwārezm’ in Persian, gave English the word (algorithm.)

Definition of Algorithm

- Solving a problem
- finite sequence of instruction / steps
when followed ~~it~~ Helps to solve a
problem.



Definition of Algorithm

- An *algorithm* is a finite set of instructions that, if followed, *solves* particular Problem.

But we are talking about solving Computational problem!!

Properties of Algorithm

- ✓ • Input : zero or more Input
- ✓ • output ; at least one output / side effect .
- ✓ • Definiteness : Each statement must be clear
(vacuous + unambiguous)
- ✓ • Finiteness : The steps must be completed
in finite amount of time
- Termination .

Software Development Life Cycle

Why Analysis is Required

Solution :-

✓ 1. Performance - word
Rotates Image

1/2 min —

How much time Required to perform the task

✓ 2. Resource Utilization - App. Space (Memory)
— . Battery space (Physical)

✓ 3. Comparison of algorithm :-
user - Sol₁ user - Sol₂
· Data (Mobile) · Power

Why Analysis is Required

Analyse - Solution/Algorithm

✓ Performance

- Time take to perform spell Correction, rotating an image

✓ Utilization of Resources

- Power
- Comparison of Algorithms

How to Analyze

Solution / Algorithm

Software (Compiler)

• Posterior Analysis :-

Algorithm → Implement the Algorithm
(programming Language)

→ M/C (Hardware) ISA

Run the program. → Estimate
the time

• Prion Analysis :- Algorithm → [Mathematical tool]

Theoretical Analysis

(Asymptotic Notation)
performance of Algorithm]

How to Analyze

The complexity of an algorithm can be calculated by using two methods:

- Posteriori Analysis
- Priori Analysis

Difference between Posteriori analysis and A Priori Analysis:

Posteriori Analysis

Priori Analysis

Difference between Posteriori analysis and A Priori Analysis:

Posteriori Analysis

- Posteriori analysis is a relative analysis.
- It is dependent on language of compiler and type of hardware.
- It will give exact answer.

Priori Analysis

- Priori analysis is an absolute analysis.
- It is independent of language of compiler and types of hardware.
- It will give approximate answer.

Difference between Posteriori analysis and A Priori Analysis:

A Posteriori Analysis

- It doesn't use asymptotic notations to represent the time complexity of an algorithm.

A Priori Analysis

- It uses the asymptotic notations to represent how much time the algorithm will take in order to complete its execution.

Difference between Posteriori analysis and A Priori Analysis:

A Posteriori Analysis

- The time complexity of an algorithm using a posteriori analysis differ from system to system.

A Priori Analysis

- The time complexity of an algorithm using a priori analysis is same for every system.

Difference between Posteriori analysis and A Priori Analysis:

A Posteriori Analysis

- If the time taken by the algorithm is less, then the credit will go to compiler and hardware.

1

A Priori Analysis

- If the program running faster, credit goes to the programmer. ←

Asymptotic

Code optimization

Theoretical +

Small Break - 7.45 pm

- Coreman

∴ Soft copy . Insertion sort

[A Priori Analysis :-]

The RAM Model of Computation

Computing time Complexity Algorithm (Pariot Analysis) a single program
 * No. of programming steps taken by an algorithm/program. $[a = b + c]$ -①

$$a = b + c;$$

M/C dependent

✓ Mov R_1 , b $R_1 \leftarrow b$
✓ MOU R_2 , c $R_2 \leftarrow c$
✓ Add R_1 , R_2 , $R_1 \leftarrow R_1 + R_2$
✓ MOU a , R_1

} unit of E operation

* Loop : multiple steps

* Memory Access 1 programming step

The RAM Model of Computation

Theoretical M/c

- The RAM (Random Access Machine) model of computation measures the run time of an algorithm by summing up the number of steps needed to execute the algorithm on a set of data.

The RAM model operates by the following principles:

The RAM Model of Computation

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The RAM (Random Access Machine) model of computation measures the run time of an algorithm by summing up the number of steps needed to execute the algorithm on a set of data. The operates by RAM model he following principles ✓

- One programming step consists of basic logical or arithmetic operations (+, *, =, if, call can also be considered to be simple operations that take one time step).

fun() {
 for (i=1 to n) m+=1 ()
 3 }

Sum ↳ a = b + c; ←
 Single programming
n+1 ↳ for i = 1 to n
n+1+n a = b + c; ←
2n+1 ↳ for i = 1 to n
 fun() - 2
1 step ↳ fun() {
 a = b + c
 return b.
}

The RAM Model of Computation

- Loops and subroutines are complex operations composed of multiple time steps. ✓
- All memory access takes exactly one time step.

framework

Count No. of
programming steps
taken by a program

Algorithm

Framework of Analysis

Step count for given Input
↑
{ programming step }

Framework of Analysis

- Recursive and non-Recursive Program
- Step count
 - The primitive operation (The basic Operation)

Time Complexity

Time complexity is a measure of the amount of time required by a computer program to run as a function of the size of the input.

Time complexity is often expressed using Big O notation.

The time complexity of an algorithm is determined by the number of operations it performs.

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Time Complexity $\leq \underline{\text{Step count}}$

- Time complexity of the program/^{Algorithm} is calculate as number of program steps needed to complete the algorithm.

total No. of programming steps represents the time complexity

Step count

Number of Steps Taken By An Algorithm

Sum of an Array Elements	Name of the array Size	Total Steps
Algorithm Sum (<u>a</u> , n) {	<u>1</u> \leftarrow <u>Size</u>	
<u>s := 0.0;</u>		<u>1</u>
for <u>i := 1</u> to n do {		<u>n+1</u> \leftarrow Last time condition is false
s = s <u>+ a[i];</u>		<u>n</u>
}		<u>1</u>
<u>return s;</u>		<u>Sum, t</u>
}		
<u>total</u>		<u>2n+3</u> \leftarrow

n - Size of the array

for (i=1; i<=n; i++)

time condition is false

Number of Steps Taken By An Algorithm

Sum of Matrix	<u>Step count</u>	<u>if ()</u>	Total Steps
Algorithm Add(a, b, c, n, n) { for i := 1 to n do for j := 1 to n do c[i, j] := a[i, j] + b[i, j]; }	<u>Sabari</u>		Minimum step count :- $\frac{2n^2+2n}{4}$
			Maximum step count :- $\frac{2n^2+2n+1}{4}$
		<u>Exact Bound</u>	upper bound = Lower bound = Exact bound

Primitive

Number of Steps Taken By An Algorithm

Asymptotic

Steps Count

Sum of Matrix	$O(n^2)$	Total Steps
Algorithm Add (<u>a</u> , <u>b</u> , <u>c</u> , <u>n</u> , <u>n</u>) {		
for i := 1 to n do {	$\sum_{i=1}^n$	$n+1 = n+1$
for j := 1 to n do {		$n \times (n+1) = n^2+n$
c[i, j] := a[i, j] + b[i, j];		$n^2 = n^2$
}		$2n^2 + 2n + 1$ Minimum = Maximum
Total	1	Maximum

Number of Steps Taken By An Algorithm

Addition of two matrixes

```
Algorithm Add(a,b,c,n,n)
{
    for i := 1 to n do
        for j := 1 to n do
            c[i,j] := a[i,j] + b[i,j];
}
```

H.W

Number of Steps Taken By An Algorithm

		Total steps	
		<u>n = 0</u>	<u>n ≥ 1</u>
Algorithm	<u>Rsum (a, n)</u> {		
[if	(n <= 0) then]	1	1
return	0.0;	1	0
else			
return	RSum(a, n - 1) +a		
}			

Recursive
Algorithm

Number of Steps Taken By An Algorithm

	Total steps	
	$n = 0$	$n > 1$
Algorithm Rsum (a, n) {	0	0
if (n <= 0) then	1	1
return 0.0;	1	0
else	0	0
return RSum(a, n - 1) +a	0	1+1
}	2	2 +x
	$x = t rsum (n-1)$	

The Time Complexity of the Program

• Summation of step count

• Function of Input size

The time complexity of the program (Number of program step taken by algorithm) can be expressed as function of input

$T(n)$, Where n is the input size

$$\text{Summation} = \overbrace{[2n+3]}^{\substack{n+1 \\ n(n+1)}}}$$

$$\text{Addition of mat'nx} = \overbrace{2n^2+2n+1}^{n^2}$$

$$\frac{1}{n} \left(\underbrace{1+2+3+4+\dots+n}_{\text{1}} \right)$$

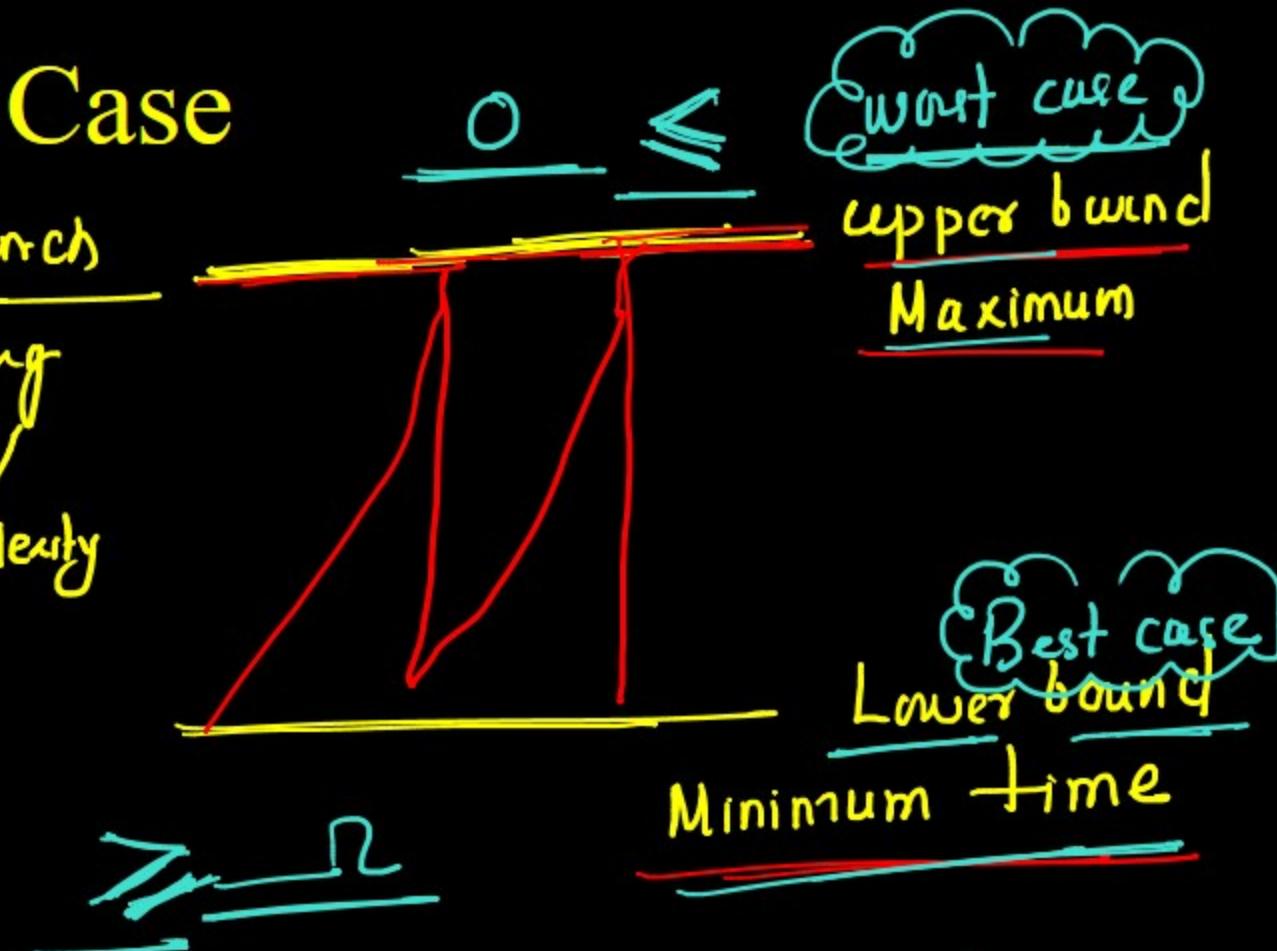
$$\frac{1}{n} * \frac{n(n+1)}{2} = \frac{(n+1)}{2}$$

Types of Analysis

- Best Case Analysis :- Minimum No. of programming step required for termination.
- Average Case Analysis $\frac{1}{n} \left[\underbrace{1+2+3+\dots+n}_{\text{1}} \right] = \frac{1}{n} * \frac{n(n+1)}{2} = \frac{n+1}{2}$
- Worst Case Analysis Maximum No. of programming step required for termination
- Amortized Analysis

Best Case

- Linear Search
R programming
Step count/
time complexity



- upper bound = Lower bound = Exact bound

Clear -

Worst Case

Average Case

What is Average Case?

It is the expected time complexity.

It is calculated by summing up all possible times multiplied by their probability.

It is also known as Expected Time Complexity.

It is denoted by $E(T)$.

It is calculated as follows:

$$E(T) = \sum_{i=1}^{n-1} T_i \cdot P_i$$

Where T_i is the time complexity for the i^{th} case and P_i is the probability of the i^{th} case occurring.

For example, if there are two cases with time complexities of 10 and 20 respectively, and probabilities of 0.5 and 0.5, then the average case time complexity would be:

$$E(T) = 10 \cdot 0.5 + 20 \cdot 0.5 = 15$$

So the average case time complexity is 15.

This is a useful way to analyze the performance of an algorithm, as it takes into account all possible cases and their probabilities.

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Amortized Analysis

Amortized Analysis

- This analysis is used when the occasional operation is very slow, but most of the operations which are executing very frequently are faster. Data structures we need amortized analysis for Hash Tables, Disjoint Sets etc.

Amortized Analysis

- Aggregate Analysis

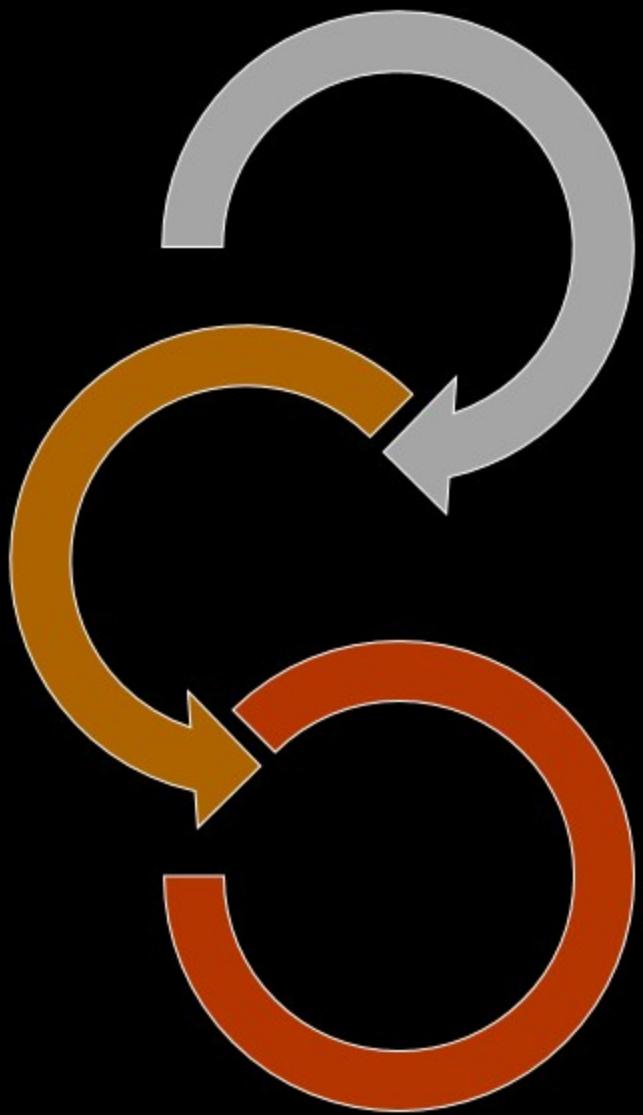
Relationship Among Best, Average and Worst Case

Best Case: Minimum number of comparisons required.

Average Case: Expected number of comparisons required.

Worst Case: Maximum number of comparisons required.

Relationship: Best Case ≤ Average Case ≤ Worst Case



Algorithmic Complexity

Linear Search

C-program - unordered Array

```
int search(int arr[], int n, int x) {  
    int i;  
    for (i=0; i<n; i++) {  
        if (arr[i] == x)  
            return i; ← Terminate  
    }  
    return -1;  
}
```

0
1
2
3
4
5
6
7
8
9

Ordered Array : Binary Search

Linear Search

$$1+11+10+10 \xrightarrow{+1} 32 \Rightarrow 33$$

```
int search(int arr[], int n, int x) {
```

```
    int i; ①
    for (i=0; i<n; i++) ②
        if (arr[i] == x) ③
            return i; ④
    return -1; ⑤
```

Step count: Smallest No. of steps required for termination of Search

0	←
1	
2	
3	
4	
5	
6	
7	
8	
9	

• Comparison -

1 Comparison

Search Element that
is been Searched

is in 1st position

• Worst case Analysis:

Maximum No. of comparison - N

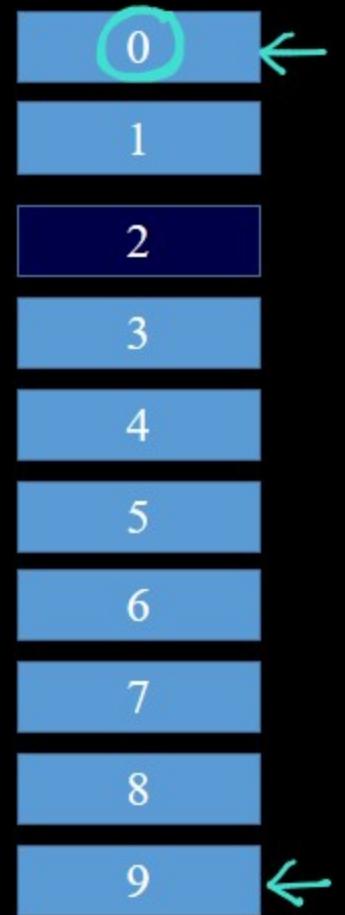
• Best case

Last position - element is not present

Linear Search

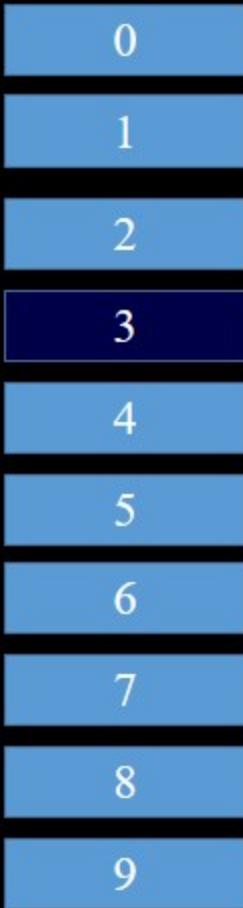
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```

condition check



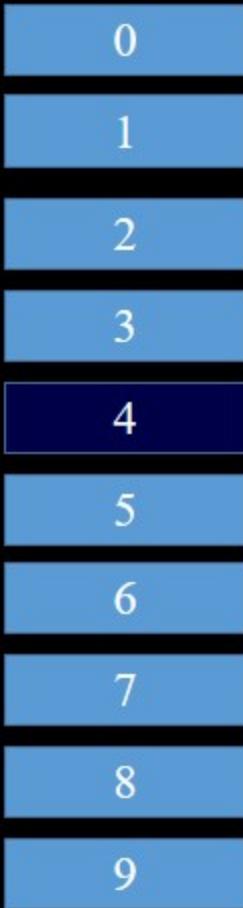
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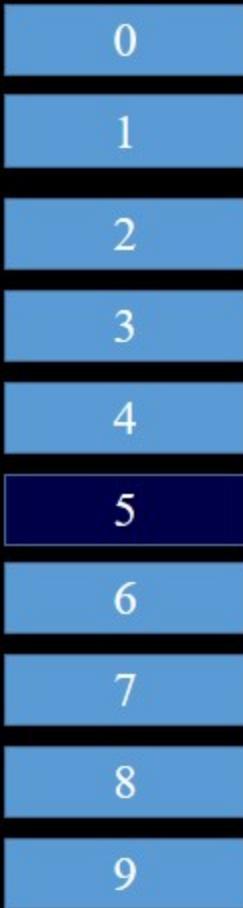
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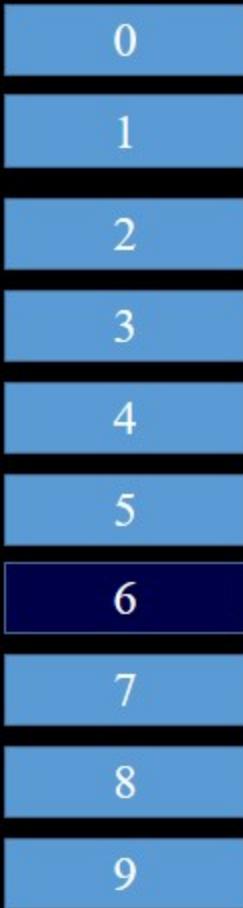
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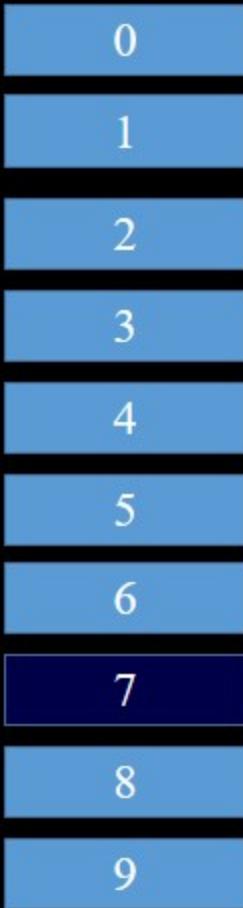
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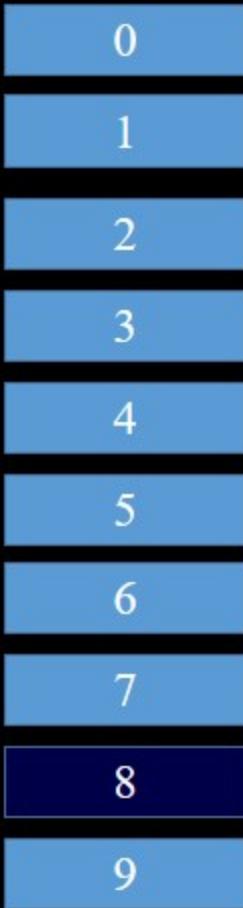
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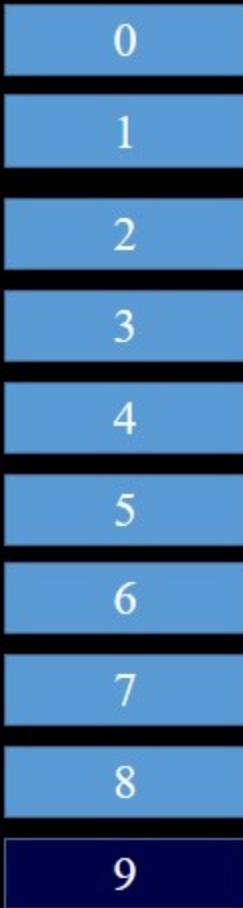
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Linear Search

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    for (i=0; i<n; i++)  
        if (arr[i] == x)  
            return i;  
    return -1;  
}
```



Bound

Step count

- Upper bound - Maximum time
- Lower Bound - Minimum
- Exact bound - Max = min

Worst case

Running
w(n)

Average case - A(n)

Best case - B(n)

w(n), A(n), B(n)

Related with each other

$$\boxed{B(n) \leq A(n) \leq w(n)}$$