

Syllabus - 3 credits

1 - Basic concepts of Algorithms

- 1.1 - Fundamentals of Algorithmic problem solving
- 1.2 - Important Problem types
- 1.3 - Fundamentals of the analysis of algorithm efficiency
- 1.4 - Analysis Framework
- 1.5 - Asymptotic Notations
- 1.6 - Asymptotic Notations & Basic Efficiency classes
- 1.7 - Recurrence relations
- 1.8 - Methods for solving recurrence relations

2 - Mathematical Analysis of Algorithms.

2.1 - Mathematical Analysis of Non-recursive Algorithms

- 2.2 - Non-recursive Algorithms and examples
- 2.3 - Mathematical Analysis of Recursive Algorithms
- 2.4 - Fibonacci numbers
- 2.5 - Empirical Analysis of Algorithms.

3 - Brute Force & Divide & conquer techniques

- 3.1 - Selection sort
- 3.2 - Bubble sort
- 3.3 - Brute-force string matching
- 3.4 - Merge sort
- 3.5 - Multiplication of two- n -bit Numbers
- 3.6 - Quick sort

3.7 - Binary search

3.8 - Binary tree traversal

4 - Algorithm Design paradigm

4.1 - Decrease and conquer technique :

Insertion Sort

4.2 - Depth first search & Breadth first search

4.3 - Transform & conquer Technique ; Presorting

4.4 - Dynamic programming : computing
a binomial coefficient.

4.5 - Warshall's & Floyd's Algorithm

4.6 - The knapsack problem and Memory function

4.7 - Optimal Binary Search trees

4.8 - Greedy Technique : Huffman trees

5 - NP Hard and NP- complete problems

5.1 - P and NP problems

5.2 - NP complete problems

5.3 - Backtracking : N - Queen's problem

5.4 - Hamiltonian circuit problem

5.5 - Branch and Bound Techniques

5.6 - Traveling salesman problem .

19/2/24

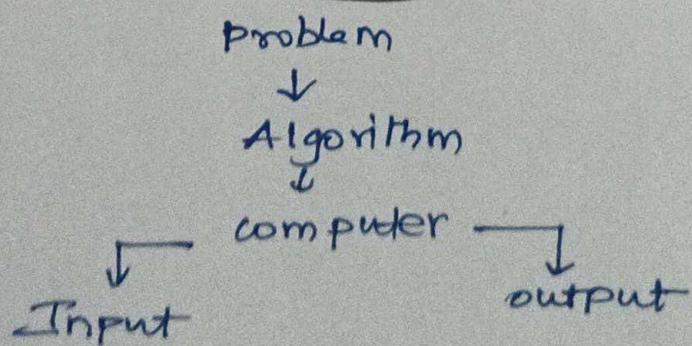
Design & Analysis of Algorithms

Algorithm

An algorithm is a sequence of unambiguous instructions for solving a problem for obtaining a required output for any legitimate input in a finite amount of time.

- * The nonambiguity requirement for each step of an algorithm cannot be compromised.
- * The range of inputs for which an algorithm works as to be specified carefully.
- * The same algorithm can be represented in several different ways.
- * Several algorithms for solving the same problem may exist.
- * Algorithm for the same problem can be based on very different ideas and can solve the problem with dramatically different speeds.

Notation of Algorithm

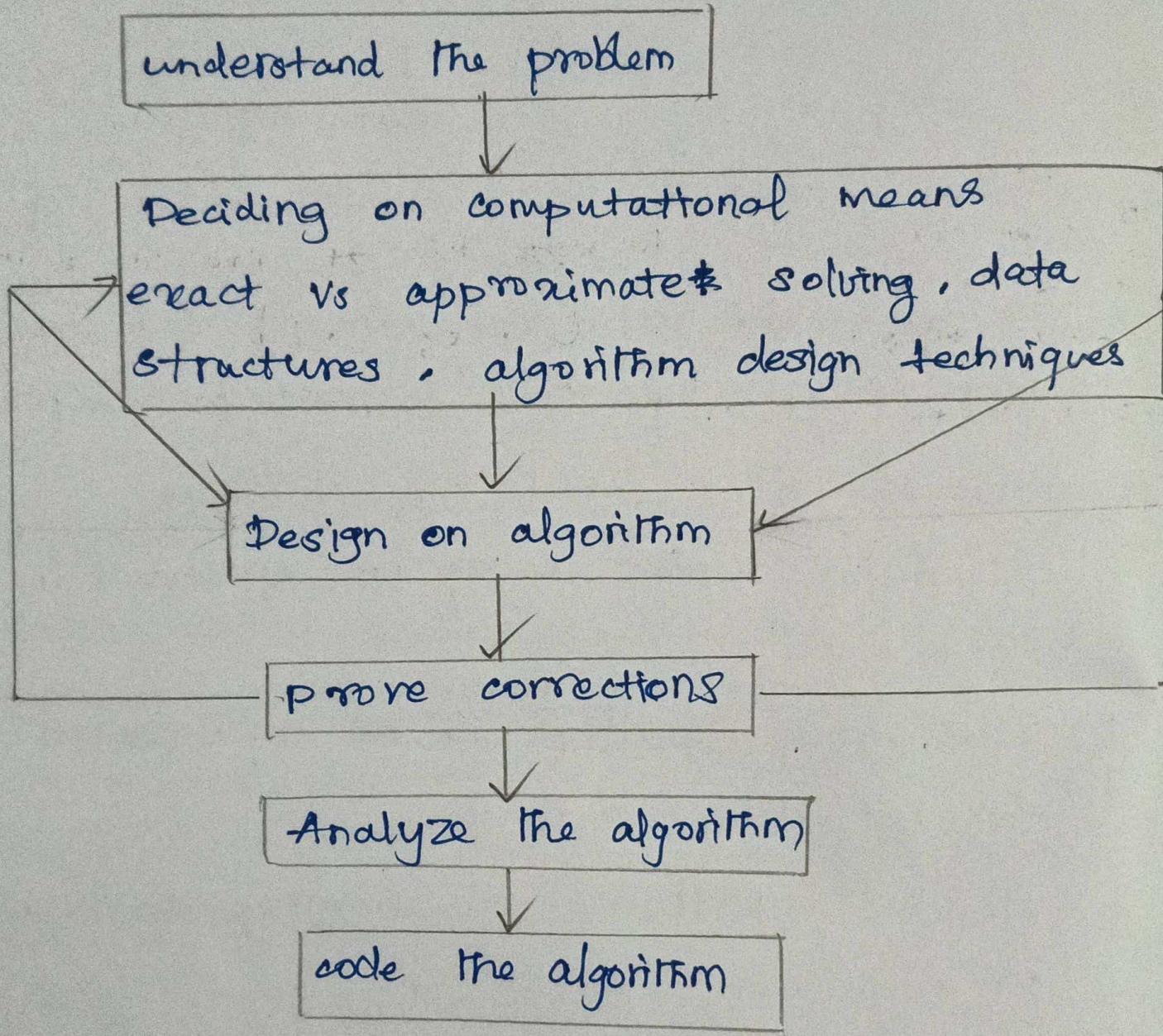


4/3/24

Fundamentals of algorithmic problem

6m

solving



understanding the problem

An input to an algorithm specifies the instance of the problem the algorithm solves.

Ascertaining the capabilities of computational devices

RAM: * Sequential algorithm

* Parallel algorithm

Exact Vs Approximate Solving

If exact ans comes it is
approximate ans comes if it is AA
exact algo is simple eg $\sqrt{2} = 1.414\ldots$

eg $\sqrt{432}$ million * square root

* nonlinear equations

* Evaluating definite integrals

Deciding on appropriate data structures

* Sorting, * Searching

Data Structures + Algorithms



programs

In the new world of object oriented programming

data structures remains important for both design and analysis of algorithm.

Algorithmic design technique

Q1 (An algorithm design technique or strategy of paradigm is a general approach to solving problems algorithmically that is

Q: D) how do you design an algorithm to solve a given problem?

& design an algorithm to compute the area and circumference of a circle
pseudocode, fc, algo, code

applicable to a variety of problems from different areas of computing.)

Methods of specifying an algorithm

- * pseudocode → It is a mixture of natural language and programming language like constructs. A pseudocode is usually more precise than a natural language and its usage of an each more succinct algorithm description.
- * flowchart → A method of expressing an algorithm by a collection of connected geometric shapes containing descriptions of the algorithm step providing an algorithm's correctness yields the
- * To prove that the algorithm is required result for every legitimate input in a finite amount of time.
- * A common technique for proving correctness is to use mathematical induction because an algorithms iteration provide a natural sequence of steps needed for such proof.

Analyzing the algorithm

3. What are the important problem types with eg.

Analyze → time efficiency, space efficiency

Time efficiency → how fast the algorithm runs

Space efficiency → how much exact memory the algorithm needs

simplicity, generality, value

b13

Important problem types

- * Sorting [key] (Shell, quick, Bubble, Selection, Insertion, Merge, Heap, Radix)
- * Searching [key] (binary, linear, heap)
- * String processing (is a sequence of char)
- * Graph problems
- * Combinational problems
- * Geometric problems
- * Numerical problems

String processing in the applications dealing with non numerical data has intensified the interest of researchers & computing practitioners in string handling algorithm.

Graph problem is a problem of finding shortest tour through n cities that visits every city exactly once

↳ salesman problem.

Graph coloring problem

It needs to assign the smallest no. of colors to the vertexes of a graph.

So that known to the adjacent vertices are same color. This problem comes in several

such as event scheduling. If the events are represented by vertices that are connected by the edge if and only

If the corresponding events can't be scheduled at the same time. The solution to the graph coloring problem yields an optimal solution.

Combinatorial problems

These are the problems that ask explicitly or implicitly to find a combinatorial object such as permutation, combinations or a subset that satisfies certain constraints.

A desire combinatorial object may also be required to have some additional property such as maximum value or minimum cost.

Fundamentals of Analysis of Algorithm Efficiency

Analysis Framework

Time efficiency, Space efficiency

Measuring the input size

e.g., It takes longer to sort larger array.

Multiply larger matrix therefore it is logical to investigate an algorithm efficiency as a function of some particular parameter to indicate the algorithm input size.

e.g.: i) Large array sort.

ii) mul of matrix

iii) Evaluating polynomial

Units of measuring runtime:

milliseconds, seconds, microseconds, nanoseconds

Drawbacks:

- * Dependencies on speed of particular computer or the quality of program implementing the algorithm.
- * Difficulty in calculating the accurate runtime of an algorithm.
Two approaches:
 - Count the no. of time each of the algorithm operation is executed but it is excessively difficult.

(8)3

To identify the most important operation of the algorithm called basic operations.

The operations contribute the most of the total running time and compute the no. of times the basic operations is executed.

e.g.: sorting algorithm.

• Matrix Multiplication

polynomial evaluation.

established framework for the analysis of an algorithm's time efficiency. Measuring it by counting the no. of times of the algorithm's basic operations is executed if the I/P of size n .

COP : execution time of an algorithm basic operation $c(n)$ - no. of times this operation need to be performed.

$T(n) \rightarrow$ RunTime of the program

$$T(n) \approx \text{cop}(n)$$

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

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

$$\approx \frac{1}{2} n^2$$

$$\frac{T(2^n)}{T(n)} = \frac{\text{cops}(2^n)}{\text{cops}(n)} = \frac{\frac{1}{2} (2^n)^2}{\frac{1}{2} n^2} = 4$$

The efficiency analysis framework ignores multiplicative constants and concentrate on order of growth.

n	$\log_2 n$	n	$n \log n$	n^2	n^3	2^n
10	3.3	10	3.3×10	10^2	10^3	10^3
10^2	6.6	10^2	6.6×10^2	10^4	10^6	1.3×10^{30}
10^3	9.9	10^3		10^6	10^9	
10^4	13.9	10^4		10^8	10^{12}	
10^5	17.9	10^5		10^{10}	10^{15}	
10^6	20.9	10^6		10^{12}		

① worst case, best case, average case
sequential search ($A[0 \dots n-1], k$)

I/P : array $A[0 \dots n-1]$ and search key - k

O/P : The index of the 1st element of A that matches k or -1 if no match $i \leftarrow 0$

while $i < n$ and $A[i] \neq k$ do

$i \leftarrow i + 1;$

 if $i < n$ return i ;

 else return -1

Eg : $A[] = \{7, 8, 9, 2, 4\}$ $n = 5$

Time complexity $T(C) = 1$

① write algorithm and give example of best, worst, average case.

Average case efficiency :

* standard assumptions

i) Probability of a successful search

ii) Probability of 1st match occurs in
im position

$$\text{cavg}(n) = \left[1 + \frac{P}{n} + 2 \frac{P}{n} + \dots + \frac{P}{n} + \dots + \frac{P}{n} \right] + n(1-p)$$
$$= \frac{P}{n} [1+2+\dots+n] + n(1-p)$$

If $i \leq n$, then $\frac{P}{n} \times \frac{n(n+1)}{2} + n(1-p)$

$$\text{cavg}(n) = \frac{P(n+1)}{2} + n(1-p)$$

Best Case = $P=1$, $\frac{n+1}{2} + 0 = \frac{n+1}{2}$, $P=0$
Worst Case

Averaged Efficiency

It applies not to a single run of an algorithm but rather to a sequence of operations performed on the same data structure.

A single operation can be expensive but the total time for a entire sequence of n search operations is always significantly better than worst case efficiency

Asymptotic Notations.

big oh (O)

big omega (Ω)

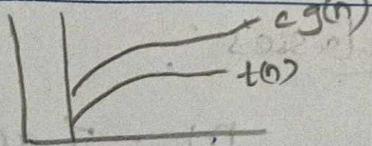
big theta (Θ)

$O(g(n))$ is a set of all functions with a smaller or same order of growth.

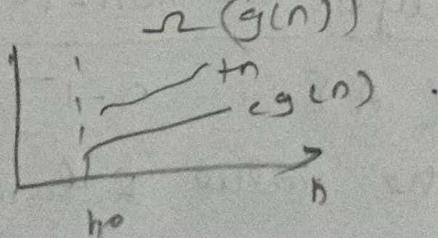
big oh (O) notation or O -notation

A function $t(n)$ is said to be of order of $g(n)$ denoted $t(n) \in O(g(n))$

If $t(n)$ is bounded above by some constant multiple of $g(n)$ for large n . i.e if there exist some positive constant c and some not negative n_0 such that

$$t(n) \leq cg(n) \text{ for all } n \geq n_0$$


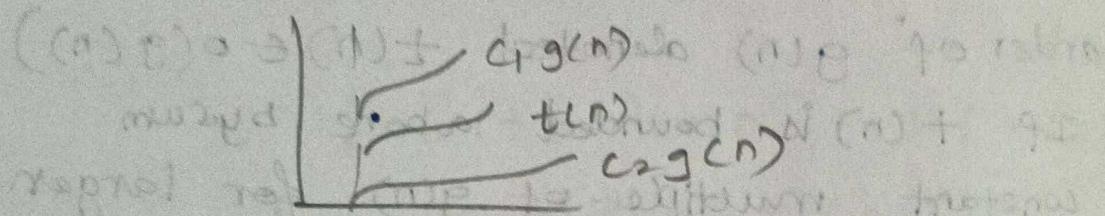
A function $t(n)$ is said to be of order $\Omega(g(n))$ if $t(n)$ is bounded below by some positive constant multiple of $g(n)$ for all large n (ie) if there exist some positive constant and some not -ve Int such that $t(n) \geq cg(n)$ (for all $n \geq n_0$)



A function $t(n)$ is said to be in the class

$$t(n) \in \Theta(g(n))$$

if $t(n)$ is bounded both above & below by some +ve constant multiples of $g(n)$,
for all large n that is if there exist
some positive constant c_1 & c_2 and some
large n_0 such that $c_2 g(n) \leq t(n) \leq c_1 g(n)$, for all $n \geq n_0$



• notation $t(n) \leq c g(n)$ means $t(n) \leq c_1 g(n)$

• " " $t(n) \geq c g(n)$ means $t(n) \geq c_2 g(n)$

• $c_2 g(n) \leq t(n) \leq c_1 g(n)$

Basic efficiency classes:

1 - constant

$n! =$ factorial

$\log n$ - logarithm

n - linear

$n \log n$ - $n \cdot \log n$

n^2 - quadratic

n^3 - cubic

2^n - exponential

Mathematical Analysis of non-recursive Algorithms

Algo MaxElement ($A[0 \dots n-1]$)

// determines the value of largest

I/P: An array $A[0 \dots n-1]$ of real nos

O/P: The value of largest element in A

for $i \leftarrow i$ to $n-1$ do
 if $A[i] > \text{maxval}$
 $\text{maxval} \leftarrow A[i]$

* return maxvalue

$$C(n) = \sum_{i=1}^{n-1} 1 = n-1 \in O(n)$$

non recursive steps

* Desired on parameters indicating the IP

- * Identify the algorithm basic operation.
- * & check whether the no. of times the basic operation is executed depends only on the size of the IP If it also depends on additional properties worst case average case & best case to the investigator.

* Setup a sum expressing no. of times the algorithm's basic operation is executed.

* Using standard formula & roots of some manipulation either find a closed form of the count or establish its order of growth.

recursive

1st 3 points are same:

- i) Setup a recurrence relation with the appropriate initial condition for the no. of time based the basic operation is executed.
- v) Solve the recurrence order of growth of its solution.
eg: Tower of Hanof,

unique elements ($A[0 \dots n-1]$)
for $i \leftarrow 0$ to $n-2$ do

 for $j \leftarrow i+1$ to $n-1$ do

 if $A[i] = A[j]$, return false

return true.

Rules : If size

basic operation

$c(n)$

notation,

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

$$= [n-1 - i] \cancel{+} \cancel{+}$$

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

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

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

$$= (n-1)(n-2-0+1) - (n-2+1)$$

$$= (n-1)(n-1) - (n-1)$$

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

$$\approx \frac{1}{2} n^2 < o(n)$$

Empirical Analysis of Algorithm

1. understand the experimental purpose
2. decide ^{on} the efficiency metric to be measured and measurement unit.
3. define the characteristics of the I/P sample
4. prepare a program implementing the pro. algorithm for the experiment.
5. generating the ^{sample of} O/P
6. run the algorithm on the Sample I/P and record the data observed
7. Analyze the data

