Computational Complexity Theory

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2 Introduction

Computability and Complexity are the two facets of this topic. Although both are interconnected in some ways, my focus would be in the complexity part of the theory, to understand how efficient certain algorithms are in solving a problem. Major part of my study is primarily based on classifying the problems within certain Complexity Classes. For example, one of the very famous open questions in Theoretical Computer Science, **P** vs **NP** , which tries to differentiate between the complexity classes **P** (Polynomial Time) and **NP** (Non-deterministic Polynomial Time) is whether they are same or different.

Next, I'll discuss about Turing Machines , its working and how influencial it was in modern computing. There are different properties associated with Turing Machines, different terminologies and different types as well.Then finally I'll cover Cryptography and usefulness of Complexity Theory in cryptography.

3 Notations

3.1 Big-O Notation

Big-O Notations are used in mathematics to characterize functions according to their growth rate. In Complexity Theory, efficiency of an algorithm is measured in terms of the input length n as $n \to \infty$.

Formal definition would be

If $f : N \to N$ and $g : N \to N$ are two functions, then $f = O(g)$ if and only if $f(n) < c \cdot g(n)$ for a constant c as $n \to \infty$.

3.2 Other Notations

There are a few more notations which complement Big-O Notation. I will give a brief information about these.

For functions f & g from N to N

$$f = \Omega(g) \quad \text{if } g = O(f) \tag{1}$$
$$f = \Theta(g) \quad \text{if } f = O(g) \text{ \& } g = O(f) \tag{2}$$
$$f = o(g) \quad \text{if there exists } \varepsilon \text{ such that } f(n) < \varepsilon \cdot g(n) \tag{3}$$
$$f = \omega(g) \quad \text{if } g = o(f) \tag{4}$$

4 Turing Machine

4.1 Computational Model

Computational model is a set of operations which are used to measure the resources(time and space) that are needed for a certain problem ot be solved. In layman's terms, using this model , we can determine if the given algorithm is efficient or not.

One such model is *Turing Machine* . It was invented by Alan Turing in 1936.

4.2 Turing Machine

Turing Machine is an abstract machine which is made up of infinite tape(s) divided into discreet boxes called *cells* and an imaginary head which can read the information in the boxes and can change the values of the cells. It also has a *state register* which stores the information about the state of the head.

The basic *Transition function* of a Turing Machine is defined as follows

$$\delta : Q \times \Gamma^k \longrightarrow Q \times \Gamma^{k-1} \times \{L, S, R\}^k \tag{5}$$

The above equation describes the functioning of a simple Turing Machine consisting of k tapes where $1st$ tape is the input tape, the rest $k-1$ tapes are called *work tapes*. Last one of them is designated as output tape.

Here, the input tape takes the input, the work tapes decide what to do with the input and in which direction to move the head. Then finally, changes(if any) are done in the output tape.

4.2.1 Properties of Turing Machine

1. If the time taken by a k tape Turing Machine to compute a function is $T(n)$, where $T : N \to N$ is a Time-Constructible function. Then the time required to compute the same function by a single tape Turing Machine is $5kT(n)^2$.

2. If a bi-directional Turing Machine computes a function in time $T(n)$, where $T : N \rightarrow N$ is a Time-Constructible function. Then the time required to compute the same function by a standard uni-directional Turing Machine is $4T(n)$.

4.2.2 Universal Turing Machine

A Universal Turing Machine is a machine which can run an arbitrary Turing Machine for a given input. It is the most general form of Turing Machine, with the general form of Time complexity of UTM being $O(TlogT)$ where T is for a normal TM . The reason I think it is because of the fact that a UTM has to deal with more bits of input than a normal one, which results in it being comparatively slower by a log factor.

Another important result associated with a Universal Turing Machine is *The Halting Problem.*The Halting Problem is a very interesting problem in context of *Uncomputability*, which proves that we cannot have a UTM which can tell with certainty whether a Turing Machine will compute a program for a given input in a finite time. In simpler words, no machine can solve the Halting Problem. This is a very important result in Computer Science because most of the problems that we generally encounter are Halting Problems, that is, a solution to them would solve The Halting Problem.

   Now since Halting Problem is unsolvable , it gives us another important result, that *most problems are uncomputable*. For example, consider the *Decision Problems* which can be considered to have input as an infinite set of either 0 or 1 ,i.e $\{0, 1\}^*$. Any combination of 0's and 1's is possible, so its impossible to compute it in finite time. Similarly, many such problems exist.