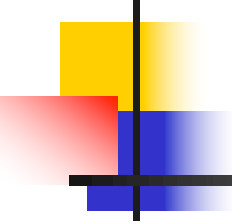


Turing Machines, part IV

The Definition of Algorithm





In last lecture, we saw several variants of TMs

- Multi-Tape TMs
- Nondeterministic TMs
- Enumerators

And established that they were equivalent to standard TM



Equivalence with yet other models

- Many other models of general purpose computation have been proposed
- Some are very much like TMs, but others are quite different
- All share the essential feature of TMs – namely, unrestricted access to unlimited memory – distinguishing them from weaker models such as FA
- Remarkably, all models with that feature turn out to be equivalent in power, so long as they satisfy reasonable requirements (e.g. the ability to perform only a finite amount of work in a single step)
- This phenomenon is analogous to “equivalence of programming languages”
- This analogy has profound implication – **definition of algorithm** -- the subject of today's lecture

What is an algorithm?

- Intuitively, an algorithm is a collection of simple instructions for carrying out some task
- The notion is much much older than Computer Science
 - Gaussian method (18th century)
 - Euclid's method for greatest common divisor (c. 300 BC)
 - Finding prime numbers
- The intuitive notion is **insufficient for gaining a deeper understating of algorithms** – precise definition had to wait until the 20th century



Carl Fredrich Gauss
(1777—1855)



Eukleides of Alexandrai
(mid-4th century BC –
mid-3rd century BC)

What is an algorithm?

- Hilbert's problems

- In 1900, Hilbert posed 23 mathematical problems as challenges for the coming century
- His 10th problem was to devise a procedure (an algorithm) that tests whether a polynomial has an integral root
- Example of polynomial:

$$6x^3yz^2 + 3xy^2 - x^3 - 10$$

- We know today *no algorithm exists for this task*
- The intuitive understanding of algorithms (that the 19th century mathematicians had at the time) was insufficient for showing such impossibility result
- Proving impossibility (negative) result requires precise definition of algorithm



Portrait of David Hilbert
in the 1900s

What is an algorithm?

- **Hilbert's problems**

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- Proving impossibility (negative) result requires precise definition of algorithm

- That definition came in 1936 – **Church-Turing thesis**, which basically states that

Intuitive notion of algorithms

equals

Turing machine algorithms



Portrait of David Hilbert
in the 1900s



Alan Turing
(1912--1954)

Turing Machines



Alonzo Church
(1903--1995)

λ -calculus



Digging deeper into Hilbert's 10th problem

- Let us phrase the problem in our terminology
Let $D = \{p \mid p \text{ is a polynomial with an integral root}\}$
- Hilbert in essence asked whether D is decidable
- The answer is negative
- In contrast, we show that it is Turning-recognizable
- Before doing so, let us first consider a simpler problem
Let $D_1 = \{p \mid p \text{ is a polynomial over a single variable } x \text{ with an integral root}\}$



Digging deeper into Hilbert's 10th problem

- Here is a TM M_1 that recognizes D_1

M_1 = “on input $\langle p \rangle$: where p is a polynomial over the variable x .

- Evaluate p with x set successively to the values $0, 1, -1, 2, -2, 3, -3, \dots$.
If at any point the polynomial evaluates to zero, *accept*.”
- If p has an integral root, M_1 eventually will find it and accept. If p does not have an integral root, M_1 will run forever.
- For the multivariable case, we can present a similar TM M that recognizes D .
- Both M_1 and M are recognizers but not deciders.
- We can convert M_1 to be decider for D_1 because we can calculate *bounds* within which the roots of a single variable polynomial must lie and restrict the search to these bounds.
- There is a result (Matijasevic, 1970) that shows that *calculating such bounds for multivariable polynomials is impossible*.

$$\pm k \frac{c_{\max}}{c_1}$$



Turning point

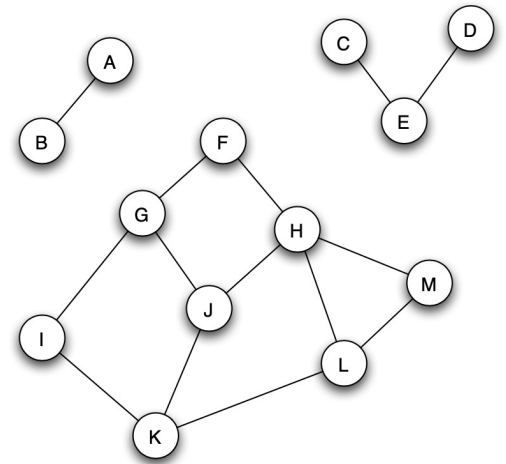
- We have come to a turning point in our study of theory of computation
- We will continue to speak of Turing machines, but our real focus from now on is on **algorithms**
- TM merely serves as a **precise model** for the definition of algorithms
- We will rely on **high-level description of TM algorithms** and follow a certain **format and notation**
 - The input to a TM is always **a string**
 - If we want to provide an input other than a string as input, we first represent – **encode** -- that object as a string. This is always possible.
 - Our notation for encoding an object O into its representation as a string is $\langle O \rangle$
 - If we have several objects O_1, O_2, \dots, O_k , we denote their encodings into a single string $\langle O_1, O_2, O_3, \dots, O_k \rangle$
 - The encoding can be done in **many** reasonable ways, and it doesn't matter which one we pick.
 - TM algorithms will be described as **indented segments of text within quotes**.
 - The algorithms are broken into **stages**, each potentially involving many steps.

Example

- Let A be the language consisting of all strings representing undirected graphs that are connected. That is,

$$A = \{ \langle G \rangle \mid G \text{ is a connected undirected graph} \}$$

- The following (next slide) is a high-level description of a TM M that decides A



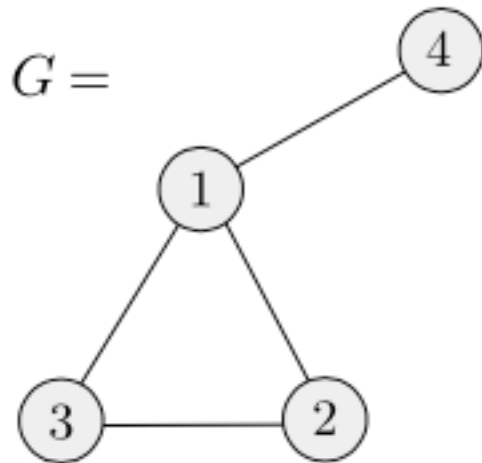


TM that decides A

$M =$ “On input $\langle G \rangle$, the encoding of a graph G :

1. Select the first node of G and mark it.
2. Repeat the following stage until no new nodes are marked:
3. For each node in G , mark it if it is attached by an edge to a node that is already marked.
4. Scan all the nodes of G to determine whether they all are marked. If they are, *accept*; otherwise, *reject*.”

A graph G and its encoding $\langle G \rangle$



$$\langle G \rangle = (1, 2, 3, 4) ((1, 2), (2, 3), (3, 1), (1, 4))$$