

Intro. Algorithms & Models of Computation

CS/ECE 374A, Fall 2024

Turing Machines

Lecture 8

Tuesday, September 17, 2024

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8.1

In the search for thinking machines

“Most General” computer?

1. **DFA**s are simple model of computation.
2. Accept only the regular languages.
3. Is there a kind of computer that can accept any language, or compute any function?
4. Recall counting argument. Set of all languages:
 $\{L \mid L \subseteq \{0, 1\}^*\}$ is ~~countably infinite~~ / uncountably infinite
5. Set of all programs:
 $\{P \mid P \text{ is a finite length computer program}\}$:
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6. **Conclusion:** There are languages for which there are no programs.

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What can be computed?

Most General Computer:

1. If not all functions are computable, which are?
2. Is there a “most general” model of computer?
3. What languages can they recognize?

History: Formalizing mathematics

1. 19th century: Oops. Math is a mess. Oy.
Fix calculus, invented set theory (Cantor), etc.
2. David Hilbert (1862–1943)
 - 2.1 1900: The list of 23 problems.
 - 2.2 Early 1900s – crisis in math foundations
attempts to formalize resulted in paradoxes, etc.
 - 2.3 1920: Hilbert's Program: "mechanize" mathematics.
 - 2.4 Finite axioms, inference rules turn crank, determine truth needed: axioms consistent & complete
 - 2.5 Hilbert: "No one shall expel us from the paradise that Cantor has created."
3. Kurt Gödel (1906–1978)
German logician, at age 25 (1931) proved: "There are true statements that can't be proved or disproved". (i.e., "no" to Hilbert)
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More history: Turing...

Alan Turing (1912–1954):

1. British mathematician
2. cryptoanalysis during WW II (enigma project)
3. Defined a computing model/program. In 1936 (age 23) provided foundations for investigating fundamental question of what is computable, what is not computable.
4. Gay, suicide.
5. Movies, UK apology.
6. Proved the halting theorem: Deciding if a computer program stops on a given input can not be decided by a program.

Turing original paper...

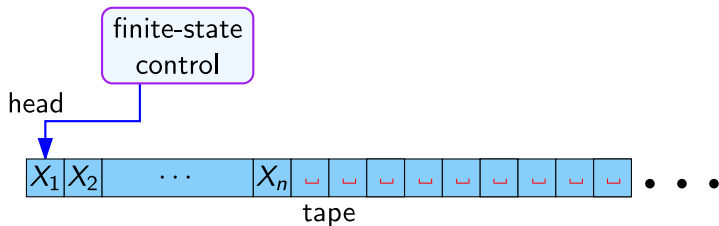
Is quite readable. Available here:

https://www.cs.virginia.edu/~robins/Turing_Paper_1936.pdf

8.2

What is a Turing machine

Turing machine



1. Input written on (infinite) one sided tape.
2. Special blank characters.
3. Finite state control (similar to **DFA**).
4. Ever step: Read character under head, write character out, move the head right or left (or stay).

High level goals

1. Church-Turing thesis: **TMs** are the most general computing devices. So far no counter example.
2. Every **TM** can be represented as a string.
3. Existence of Universal Turing Machine which is the model/inspiration for stored program computing. **UTM** can simulate any **TM**
4. Implications for what can be computed and what cannot be computed

Turing machine: Formal definition

A Turing machine is a 7-tuple

$$(Q, \Sigma, \Gamma, \delta, q_0, q_{\text{acc}}, q_{\text{rej}})$$

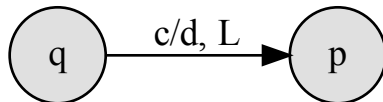
- ▶ Q : finite set of states.
- ▶ Σ : finite input alphabet.
- ▶ Γ : finite tape alphabet.
- ▶ $\delta : Q \times \Gamma \rightarrow Q \times \Gamma \times \{L, R, S\}$: Transition function.
- ▶ $q_0 \in Q$ is the initial state.
- ▶ $q_{\text{acc}} \in Q$ is the accepting/final state.
- ▶ $q_{\text{rej}} \in Q$ is the rejecting state.
- ▶ \sqcup or \sqsubset : Special blank symbol on the tape.

Turing machine: Transition function

$$\delta : Q \times \Gamma \rightarrow Q \times \Gamma \times \{L, R, S\}$$

As such, the transition

$$\delta(q, c) = (p, d, L)$$



1. **q**: current state.
2. **c**: character under tape head.
3. **p**: new state.
4. **d**: character to write under tape head
5. **L**: Move tape head left.

Missing transitions lead to hell state.

“Blue screen of death.”

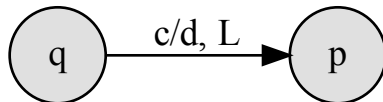
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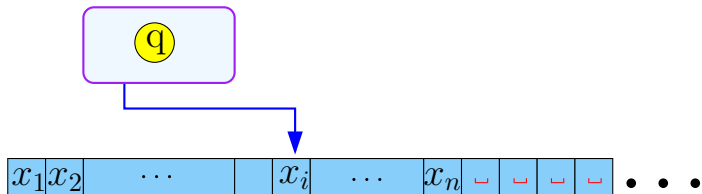
Snapshots, computation as sequence of strings

Snapshot = ID: Instantaneous Description

1. Contains all necessary information to capture “state of the computation”.
2. Includes
 - 2.1 state q of M
 - 2.2 location of read/write head
 - 2.3 contents of tape from left edge to rightmost non-blank (or to head, whichever is rightmost).

Snapshot = ID: Instantaneous Description

As a string



ID: $x_1 x_2 \dots x_{i-1} q x_i x_{i+1} \dots x_n$

$x_1, \dots, x_n \in \Gamma, q \in Q.$

A step in computation as rewriting strings

$x_1 x_2 \dots x_{i-1} q x_i x_{i+1} \dots x_n$

If transition is $\delta(q, X_i) = (p, Y, L)$, new ID is:

$$\begin{array}{l} \text{current ID :} \\ \delta(q, X_i) = (p, y, L) \implies \end{array} \quad \begin{array}{l} x_1 x_2 \dots x_{i-2} x_{i-1} q x_i x_{i+1} \dots x_n \\ x_1 x_2 \dots x_{i-2} p x_{i-1} y x_{i+1} \dots x_n \end{array}$$

If no transition defined, or illegal transition, then no next ID (crash).

Shockinglly: Computation is just a string rewriting system.

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1. Initial ID: $q_0 w$:
2. Accepting ID: $\alpha q_{\text{acc}} \alpha'$, for some $\alpha, \alpha' \in \Gamma^*$.
3. Rejecting ID: $\alpha q_{\text{rej}} \alpha'$, for some $\alpha, \alpha' \in \Gamma^*$.
4. $\mathcal{I} \rightsquigarrow \mathcal{J}$: Denotes that if we start execution of **TM** with configuration/ID encoded by \mathcal{I} , leads **TM** (after maybe several steps) to ID \mathcal{J}
5. M accepts w : If for some $\alpha, \alpha' \in \Gamma^*$, we have

$$q_0 w \rightsquigarrow \alpha q_{\text{acc}} \alpha'.$$

Acceptance happens as soon as **TM** enters accept state.

6. Language of **TM** M : $L(M) = \{w \in \Sigma^* \mid M \text{ accepts } w\}$.

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Non-accepting computation

M does not accept w if:

1. M enters q_{rej} (i.e., M rejects w)
2. M crashes (moves to left of tape, no transition available, etc).
3. M runs forever.

If the **TM** keeps running, should we wait, or is it rejection?

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Everything is a number

8.4

Languages defined by a Turing machine

Recursive vs. Recursively Enumerable

1. Recursively enumerable (aka RE) languages

$$L = \{L(M) \mid M \text{ some Turing machine}\} .$$

2. Recursive / decidable languages

$$L = \{L(M) \mid M \text{ some Turing machine that halts on all inputs}\} .$$

3. Fundamental questions:

3.1 What languages are RE?

3.2 Which are recursive?

3.3 What is the difference?

3.4 What makes a language decidable?

3.5 How much wood would a TM chuck, if a TM could chuck wood?

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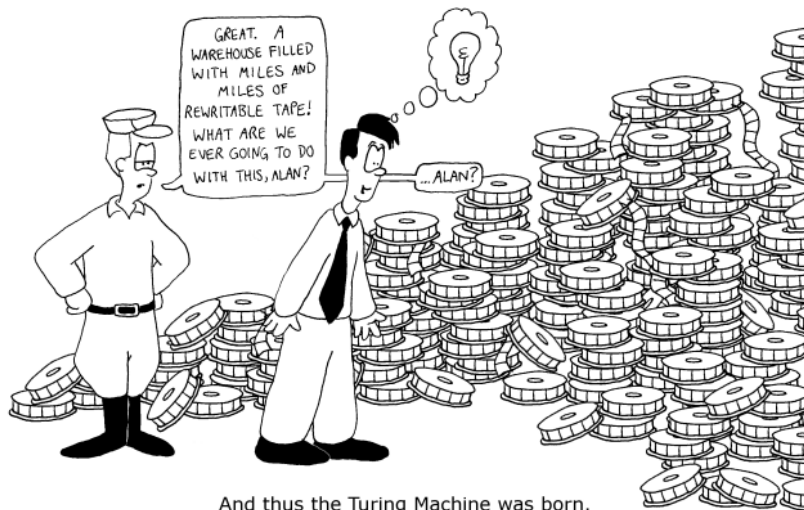
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How was the Turing Machine invented...



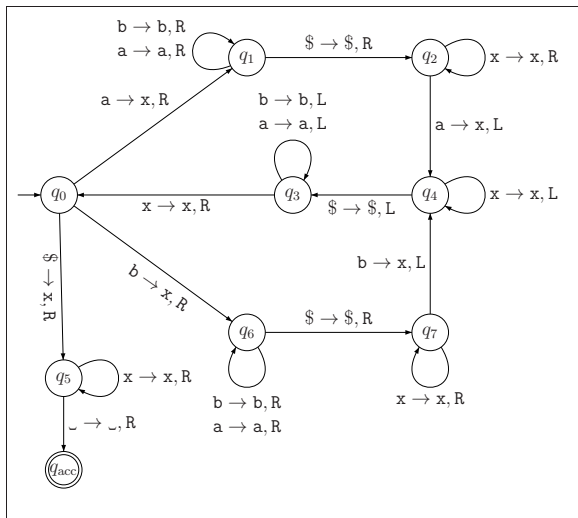
8.5

Some examples of Turing machines

8.5.1

Turing machine for $w\$w$

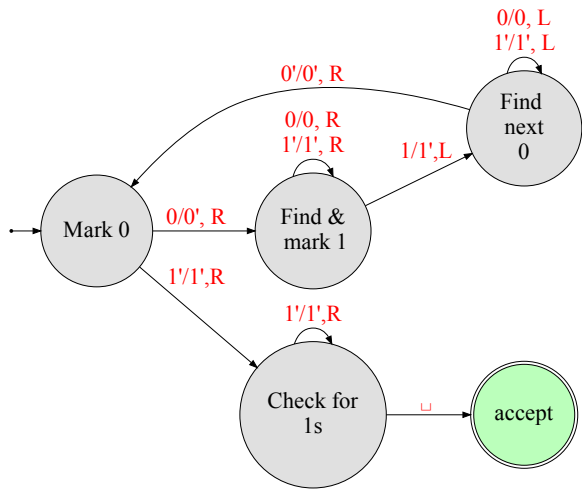
Example: Turing machine for $w\$w$



8.5.2

Turing machine for 0^n1^n

Example: Turing machine for 0^n1^n

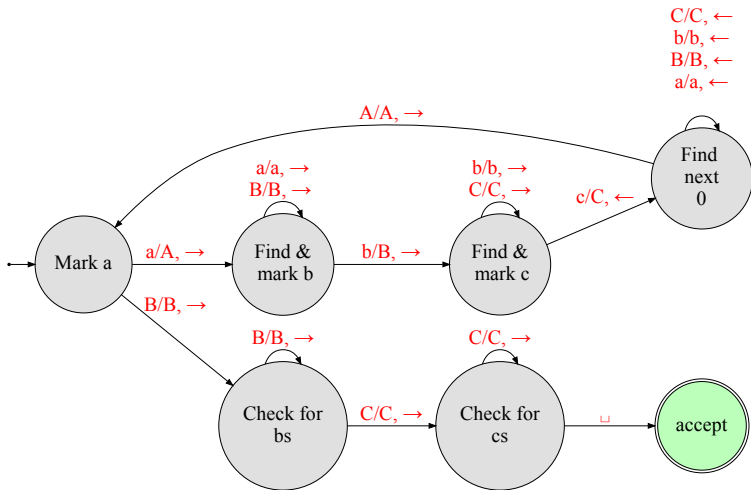


8.5.3

Turing machine for $a^n b^n c^n$

Example: Turing machine for $a^n b^n c^n$

A language that is not context free...



8.6

Why Turing Machine is a “real” computer?

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TM can compute anything that a real computer can, if very very very tediously.

1. Add/multiply two numbers in binary representation.
2. Move input tape one position to the right.
3. Simulate a TM with two tapes.
4. Simulate a TM with many tapes.
5. Stack.
6. Subroutines.
7. Compile say any C program into a **TM**.
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So what Turing Machines are good for?

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Universal Turing Machine

Turing Machine that simulates another Turing Machine

UTM: A Turing machine that can simulate another Turing machine.

1. Programs can self replicate.
2. Program can modify themselves (a big no no nowadays).
3. Program can rewrite a program.
4. Turing had created a Pandora box...
...which we will open in the next lecture.