

# Teoria degli Algoritmi

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## Lecture 2: Turing Machines





# Our Model of Computation: Turing Machines

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  - Hilbert wondered if it exists an “effective procedure” (i.e., our informal definition of algorithm) that decides whether any mathematical statement is true or false, in a finite number of steps
  - As a special case of this decision problem, Hilbert considered the validity problem for first-order logic (a.k.a. *entscheidungsproblem*)

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## Note

The linear nature of memory tape, as opposed to random access memory, is a limitation on computation speed but not power: a TM can find any memory location, i.e., tape cell, by sequentially scanning its tape

# Turing Machines: A Formal Definition

## Definition (Turing machine)

A Turing machine  $M$  is a 6-tuple  $(Q, \Sigma, \delta_M, q_0, q_{\text{accept}}, q_{\text{reject}})$ , where:

- $Q$  is the finite set of **states**



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- $q_{\text{reject}} \in Q$  is the **reject state**, s.t.  $q_{\text{accept}} \neq q_{\text{reject}}$

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- If  $M$  ever halts, it will leave the output string on the tape, i.e.,  $\sigma_{\text{out}} \in \Sigma^*$

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- The machine can compute an **infinite** function  $f_M$  that takes as input a string  $\sigma_{\text{in}} \in \Sigma^*$  and produces another string  $\sigma_{\text{out}} \in \Sigma^*$  as output, both of arbitrary lengths



# From Turing Machines to Computable Functions

## Definition (Computable Function)

Let  $f : \Sigma^* \mapsto \Sigma^*$  be a (total) function and let  $M$  be a Turing machine. We say that  $M$  computes  $f$  if for every  $x \in \Sigma^*$ ,  $M(x) = f(x)$ .

We say that a function  $f$  is computable if there exists a Turing machine  $M$  that computes it.

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## Note

Defining a function “computable” if it can be computed by a Turing machine might seem incautious, but this is equivalent to being computable in virtually *any* reasonable model of computation.

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  - Or by any equivalent computational models proposed by Gödel (**recursive functions**) and Church ( **$\lambda$ -calculus**)
- The three formally-defined classes of computable functions coincide with the informal notion of an effectively calculable function
- Since the concept of effective calculability does not have a formal definition, the thesis, although it has near-universal acceptance, cannot be formally proven

# (Boolean) Computable Functions

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## Definition

We define by  $\mathcal{R}$  the set of **all** computable functions  $f : \Sigma^* \mapsto \Sigma$

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- This is equivalent to computing the boolean (total) function  $f : \Sigma^* \mapsto \Sigma$  defined as:

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## Definition (Turing-decidable Language)

A language  $L$  is **Turing-decidable** (or simply **decidable**) if there is a Turing machine  $M$  that decides it



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- For historical reasons, some texts also refer to computable boolean functions/decidable languages as **recursive languages**
- This is also the reason why the letter  $\mathcal{R}$  is often used
- We stick to the term *functions* rather than *languages*, although the following always holds:

$$f : \Sigma^* \mapsto \Sigma$$

$$L = \{x \in \Sigma^* \mid f(x) = 1\}$$

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## Definition

A partial function  $f : A \mapsto B$  is a function that is only defined on a subset  $A'$  of  $A$  (i.e.,  $A' \subset A$ ). We can also think of such a function as mapping from  $A$  to  $B \cup \{\perp\}$ , where  $\perp$  is a special “failure” symbol such that  $f(a) = \perp$  indicates  $f$  is not defined on input  $a$

# Turing Machines Computing Partial Functions

## Example

Consider the function  $div : \mathbb{Z}^{0+} \times \mathbb{Z}^{0+} \mapsto \mathbb{Z}^{0+}$ , defined as follows:

$$div(a, b) = \begin{cases} \lceil \frac{a}{b} \rceil, & \text{if } b > 0 \\ \perp, & \text{otherwise} \end{cases}$$

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  - If  $a > 0$  and  $b = 0$ ,  $M$  never halts but this is ok, since  $div$  is undefined on such inputs
  - If  $a = b = 0$ ,  $M$  will output 0, which is also ok, since we do not care about what the machine outputs on inputs on which  $div$  is undefined

# Computable Functions (Redefined)

## Definition

Let  $f$  be a **total** or **partial** function, such that  $f : \Sigma^* \mapsto \Sigma^*$  and let  $M$  be a Turing machine.

We say that  $M$  **computes**  $f$  if for every  $x \in \Sigma^*$  on which  $f$  is defined,  $M(x) = f(x)$ .

We say that a (partial or total) function  $f$  is **computable** if there is a Turing machine that computes it.

# A Clarification on the Role of $\perp$

- We used  $\perp$  as our special “failure symbol”; if a Turing machine  $M$  fails to halt on some input  $x \in \Sigma^*$  then we denote this by  $M(x) = \perp$

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- This **does not** mean that  $M$  outputs some encoding of the symbol  $\perp$  but rather that  $M$  enters into an infinite loop when given  $x$  as input
- As such, one might be tempted to think that  $M$  halts on  $x$  if and only if  $f$  is defined on  $x$

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- This **does not** mean that  $M$  outputs some encoding of the symbol  $\perp$  but rather that  $M$  enters into an infinite loop when given  $x$  as input
- As such, one might be tempted to think that  $M$  halts on  $x$  if and only if  $f$  is defined on  $x$
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- As such, one might be tempted to think that  $M$  halts on  $x$  if and only if  $f$  is defined on  $x$
- However, for a Turing machine  $M$  to compute a partial function  $f$  it is **not** necessary to enter an infinite loop on inputs  $x$  outside the domain of  $f$
- All that is needed is for  $M$  to output  $f(x)$  on  $x \in \text{domain}(f)$ : on any other input it is OK for  $M$  to output an arbitrary value or not to halt at all

# Functions vs. Languages

- A Turing machine  $M$  **recognizes** a language  $L$  if for **every** input  $x \in \Sigma^*$ ,  $M(x)$  outputs 1 if and only if  $x \in L$

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## Definition (Turing-recognizable Language)

A language  $L$  is **Turing-recognizable** (or simply **recognizable** or **semi-decidable**) if there is a Turing machine  $M$  that recognizes it

# A Note on the Terminology

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- This is also the reason why the letter  $\mathcal{RE}$  is often used
- We stick to the term *functions* rather than *languages*, although the following always holds:

$$f : \Sigma^* \mapsto \Sigma$$

$$L = \{x \in \Sigma^* \mid f(x) = 1\}$$



# Variants of Turing Machines

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- Interestingly enough, the original computational model and its variants have all the same power
- They all compute the same functions/recognize the same set of languages

# Multi-tape Turing Machines

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- The transition function  $\delta_M$  is changed to allow for reading, writing, and moving the heads on some or all of the tapes, simultaneously
- Formally, the transition function of a  $k$ -tape Turing machine is defined as follows:

$$\delta_M : Q \times (\Sigma \cup \{\emptyset\})^k \mapsto Q \times (\Sigma \cup \{\emptyset\})^k \times \{-1, 0, +1\}^k$$

# Multi-tape Turing Machines: Example

Consider a  $k$ -tape Turing Machine, then the expression

$$\delta_M(q_i, \sigma_1, \sigma_2, \dots, \sigma_k) = (q_j, \sigma'_1, \sigma'_2, \dots, \sigma'_k, +1, 0, \dots, -1)$$

means that, if the machine is in state  $q_i$  and heads 1 through  $k$  are reading symbols  $\sigma_1$  through  $\sigma_k$ , then it goes to state  $q_j$ , writes symbols  $\sigma'_1$  through  $\sigma'_k$  and moves each head to the left (-1) or to the right (+1) of the current position, or leaves it where it is (0)

# Equivalence Between Single- and Multi-Tape TMs

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- In fact, it can be proven that those two models of computations are indeed equivalent (i.e., they both recognize the same languages)
- To sketch the idea of the proof, consider two Turing machines:  $S$ ,  $M$ 
  - The former is a single-tape machine, whilst the latter is multi-tape
  - The key idea is to simulate  $M$  using  $S$
  - We can lay down the content of the  $k$  tapes of  $M$  on the single tape of  $S$ , using a special symbol as delimiter (e.g.,  $\#$ )
  - Add another extra symbol (e.g.,  $\bullet$ ) on top of the current symbol to mimic the head position on each tape

# Non-deterministic Turing Machines (NTMs)

- At any time during the computation a non-deterministic TM proceeds according to several possibilities



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- The transition function for a NTM  $\delta_M$  is defined as follows:

$$\delta_M : Q \times \Sigma \cup \{\emptyset\} \mapsto \mathcal{P}(Q \times \Sigma \cup \{\emptyset\} \times \{-1, 0, +1\})$$

where  $\mathcal{P}(A)$  stands for the **power set** of  $A$ , i.e., the set of all subsets of  $A$

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- The computation of an NTM is a **tree**, whose branches correspond to different computational paths for the machine
- If some branch leads to the accept state ( $q_{\text{accept}}$ ), the machine accepts its input

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- In fact, it can be proven that those two models of computations are indeed equivalent (i.e., they both recognize the same languages)
- To sketch the idea of the proof, consider two Turing machines:  $D$ ,  $N$ 
  - The former is a deterministic machine, whilst the latter is non-deterministic
  - The key idea is to simulate  $N$  using  $D$  by letting  $D$  try **all** the possible branches of  $N$ 's non-deterministic computation
  - If  $D$  ever reaches the accept state on one of these branches,  $D$  accepts; otherwise  $D$ 's simulation may run forever

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- **breadth first search** explores all branches at the same depth of the tree before moving to the next level
- This guarantees that  $D$  will visit every node in the tree until it encounters an accepting configuration



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- We have already seen that we can use the same binary string encoding to represent virtually **any** object
- As a special case, we can therefore encode **any** Turing machine  $M$  together with **any** of its input  $x$

# Universal Turing Machine

## Definition (Universal Turing Machine)

There exists a Turing machine  $U$ , such that on every string  $M$  which encodes a Turing machine, and  $x \in \Sigma^*$ :

$$U(M, x) = M(x)$$

If the machine  $M$  halts on  $x$  and outputs some  $y \in \Sigma^*$  (i.e.,  $M(x) = y$ ), then:

$$U(M, x) = M(x) = y$$

If  $M$  does **not** halt on  $x$  (i.e.,  $M(x) = \perp$ ) then:

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- Intuitively, the existence of  $U$  implies the existence of a “universal” algorithm that can evaluate arbitrary algorithms ( $M$ ) on arbitrary inputs ( $x$ )

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- The desired program  $U$  is an **interpreter** for Turing machines
- $U$  gets a representation of the machine  $M$  (e.g., source code), and some input  $x$ , and simulates the execution of  $M$  on  $x$

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- The **interpreter** will continue the simulation until the machine eventually halts
- Translating the interpreter above into the corresponding Turing machine is “easy”

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# Universal Turing Machine: Implications

- There is more than one Turing machine  $U$  that works as indicated above
- The existence of even a *single* such machine is already fundamental to computer science
- The idea of a “universal program” is of course not limited to theory
- The most famous practical example is represented by **compilers** (for programming languages), which are often used to compile themselves!



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- Computable functions (total/partial) are those which can be computed by a TM
- There exists few variants of standard TM like multi-tape or non-deterministic TMs yet they all have the same power
- The existence of a special Universal Turing Machine (UTM) allows us to design an algorithm that can run any other algorithm