# Languages and Algorithms for Artificial Intelligence (Module 3)

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# 1 Introduction

**Computational task** Description of a problem.

Computational task

**Computational process** Algorithm to solve a task.

Computational process

**Algorithm (informal)** A finite description of elementary and deterministic computation steps.

### 1.1 Notations

**Set of the first** *n* **natural numbers** Given  $n \in \mathbb{N}$ , we have that  $[n] = \{1, \dots, n\}$ .

### 1.1.1 Strings

**Alphabet** Finite set of symbols.

Alphabet

**String** Finite, ordered, and possibly empty tuple of elements of an alphabet.

String

The empty string is denoted as  $\varepsilon$ .

**Strings of given length** Given an alphabet S and  $n \in \mathbb{N}$ , we denote with  $S^n$  the set of all the strings over S of length n.

**Kleene star** Given an alphabet S, we denote with  $S^* = \bigcup_{n=0}^{\infty} S^n$  the set of all the strings over S.

Kleene star

**Language** Given an alphabet S, a language  $\mathcal{L}$  is a subset of  $S^*$ .

Language

### 1.1.2 Tasks encoding

**Encoding** Given a set A, any element  $x \in A$  can be encoded into a string of the language  $\{0,1\}^*$ . The encoding of x is denoted as  $\lfloor x \rfloor$  or simply x.

Encoding

**Task function** Given two countable sets A and B representing the domain, a task can be represented as a function  $f: A \to B$ .

Task

When not stated, A and B are implicitly encoded into  $\{0,1\}^*$ .

**Characteristic function** Boolean function of form  $f: \{0,1\}^* \to \{0,1\}$ .

Characteristic function

Given a characteristic function f, the language  $\mathcal{L}_f = \{x \in \{0,1\}^* \mid f(x) = 1\}$  can

be defined.

**Decision problem** Given a language  $\mathcal{M}$ , a decision problem is the task of computing a boolean function f able to determine if a string belongs to  $\mathcal{M}$  (i.e.  $\mathcal{L}_f = \mathcal{M}$ ).

Decision problem

## 1.1.3 Asymptotic notation

**Big O** A function  $f: \mathbb{N} \to \mathbb{N}$  is O(g) if g is an upper bound of f.

$$f \in O(g) \iff \exists \bar{n} \in \mathbb{N} \text{ such that } \forall n > \bar{n}, \exists c \in \mathbb{R}^+ : f(n) \leq c \cdot g(n)$$

**Big Omega** A function  $f: \mathbb{N} \to \mathbb{N}$  is  $\Omega(g)$  if g is a lower bound of f.

Big O

$$f \in \Omega(g) \iff \exists \bar{n} \in \mathbb{N} \text{ such that } \forall n > \bar{n}, \exists c \in \mathbb{R}^+ : f(n) \ge c \cdot g(n)$$

**Big Theta** A function  $f: \mathbb{N} \to \mathbb{N}$  is  $\Theta(g)$  if g is both an upper and lower bound of f.

$$f \in \Theta(g) \iff f \in O(g) \text{ and } f \in \Omega(g)$$

# 2 Turing Machine

## 2.1 k-tape Turing Machine

**Tape** Infinite one-directional line of cells. Each cell can hold a symbol from a finite  $\Gamma$  alphabet  $\Gamma$ .

**Tape head** A tape head reads or writes one symbol at a time and can move left or right on the tape.

**Input tape** Read-only tape where the input will be loaded.

Work tape Read-write auxiliary tape used during computation.

**Output tape** Read-write tape that will contain the output of the computation.

**Remark.** Sometimes the output tape is not necessary and the final state of the computation can be used to determine a boolean outcome.

**Instructions** Given a finite set of states Q, at each step, a machine can:

Instructions

**Read** from the k tape heads.

**Replace** the symbols under the writable tape heads, or leave them unchanged.

Change state.

**Move** each of the k tape heads to the left or right, or leave unchanged.

k-tape Turing Machine (TM) A Turing Machine working on k tapes (one of which is the input tape) is a triple  $(\Gamma, Q, \delta)$ :

k-tape Turing Machine (TM)

- $\Gamma$  is a finite set of tape symbols. We assume that it contains a blank symbol  $(\Box)$ , a start symbol  $(\triangleright)$ , and the digits 0, 1.
- Q is a finite set of states. The initial state is  $q_{\text{init}}$  and the final state is  $q_{\text{halt}}$ .
- $\delta$  is the transition function that describes the instructions allowed at each step. It is defined as:

$$\delta: Q \times \Gamma^k \to Q \times \Gamma^{k-1} \times \{\mathtt{L},\mathtt{S},\mathtt{R}\}^k$$

By convention, when the state is  $q_{\text{halt}}$ , the machine is stuck (i.e. it cannot change state or operate on the tapes):

$$\delta(q_{\text{halt}}, \{\sigma_1, \dots, \sigma_k\}) = (q_{\text{halt}}, \{\sigma_1, \dots, \sigma_k\}, (S, \dots, S))$$

**Theorem 2.1.1** (Turing Machine equivalence). The following computational models have, with at most a polynomial overhead, the same expressive power: 1-tape TMs, k-tape TMs, non-deterministic TMs, random access machines,  $\lambda$ -calculus, unlimited register machines, programming languages (Böhm-Jacopini theorem), ...

## 2.2 Computation

**Configuration** Given a TM  $\mathcal{M} = (\Gamma, Q, \delta)$ , a configuration C is described by:

Configuration

- The current state q.
- The content of the tapes.
- The position of the tape heads.

**Initial configuration** Given the input  $x \in \{0,1\}^*$ , the initial configuration  $\mathcal{I}_x$  is described as follows:

- The current state is  $q_{\text{init}}$ .
- The first (input) tape contains  $\triangleright x \square \dots$  The other tapes contain  $\triangleright \square \dots$
- The tape heads are positioned on the first symbol of each tape.

**Final configuration** Given an output  $y \in \{0, 1\}^*$ , the final configuration is described as follows:

- The current state is  $q_{\text{halt}}$ .
- The output tape contains  $\triangleright y \square \dots$

**Computation (string)** Given a TM  $\mathcal{M} = (\Gamma, Q, \delta)$ ,  $\mathcal{M}$  returns  $y \in \{0, 1\}^*$  on input  $x \in \{0, 1\}^*$  (i.e.  $\mathcal{M}(x) = y$ ) in t steps if:

$$\mathcal{I}_x \xrightarrow{\delta} C_1 \xrightarrow{\delta} \dots \xrightarrow{\delta} C_t$$

where  $C_t$  is a final configuration for y.

**Computation (function)** Given a TM  $\mathcal{M} = (\Gamma, Q, \delta)$  and a function  $f : \{0, 1\}^* \to \{0, 1\}^*$ , Computation  $\mathcal{M}$  computes f iff:

$$\forall x \in \{0,1\}^* : \mathcal{M}(x) = f(x)$$

If this holds, f is a computable function.

**Computation in time T** Given a TM  $\mathcal{M}$  and the functions  $f:\{0,1\}^* \to \{0,1\}^*$  and Computation in time  $T:\mathbb{N}\to\mathbb{N}, \mathcal{M}$  computes f in time T iff:

$$\forall x \in \{0,1\}^* : \mathcal{M}(x) \text{ returns } f(x) \text{ in at most } T(|x|) \text{ steps}$$

**Decidability in time T** Given a function  $f: \{0,1\}^* \to \{0,1\}$ , the language  $\mathcal{L}_f$  is decidable Decidability in time in time T iff f is computable in time T.

# 2.3 Universal Turing Machine

**Turing Machine encoding** Given a TM  $\mathcal{M} = (\Gamma, Q, \delta)$ , the entire machine can be described by  $\delta$  through tuples of form:

$$Q\times \Gamma^k\times Q\times \Gamma^{k-1}\times \{\mathtt{L},\mathtt{S},\mathtt{R}\}^k$$

It is therefore possible to encode  $\delta$  into a binary string and consequently create an encoding  $\bot \mathcal{M} \bot$  of  $\mathcal{M}$ .

The encoding should satisfy the following conditions:

1. For every  $x \in \{0,1\}^*$ , there exists a TM  $\mathcal{M}$  such that  $x = \bot \mathcal{M} \bot$ .

2. Every TM is represented by an infinite number of strings. One of them is the canonical representation.

**Theorem 2.3.1** (Universal Turing Machine (UTM)). There exists a TM  $\mathcal{U}$  such that, for every binary strings x and  $\alpha$ , it emulates the TM defined by  $\alpha$  on input x:

Universal Turing Machine (UTM)

$$\mathcal{U}(x,\alpha) = \mathcal{M}_{\alpha}(x)$$

where  $\mathcal{M}_{\alpha}$  is the TM defined by  $\alpha$ .

Moreover,  $\mathcal{U}$  simulates  $\mathcal{M}_{\alpha}$  with at most  $CT \log(T)$  time overhead, where C only depends on  $\mathcal{M}_{\alpha}$ .

## 2.4 Computability

#### 2.4.1 Undecidable functions

**Theorem 2.4.1** (Existence of uncomputable functions). There exists a function uc: Uncomputable  $\{0,1\}^* \to \{0,1\}^*$  that is not computable by any TM.

*Proof.* Consider the following function:

$$uc(\alpha) = \begin{cases} 0 & \text{if } \mathcal{M}_{\alpha}(\alpha) = 1\\ 1 & \text{if } \mathcal{M}_{\alpha}(\alpha) \neq 1 \end{cases}$$

If uc was computable, there would be a TM  $\mathcal{M}$  that computes it (i.e.  $\forall \alpha \in \{0,1\}^*$ :  $\mathcal{M}(\alpha) = uc(\alpha)$ ). This will result in a contradiction:

$$uc(\bot \mathcal{M} \bot) = 0 \iff \mathcal{M}(\bot \mathcal{M} \bot) = 1 \iff uc(\bot \mathcal{M} \bot) = 1$$

Therefore, uc cannot be computed.

**Halting problem** Given an encoded TM  $\alpha$  and a string x, the halting problem aims to Halting problem determine if  $\mathcal{M}_{\alpha}$  terminates on input x. In other words:

$$\mathtt{halt}(\llcorner(\alpha,x)\lrcorner) = \begin{cases} 1 & \text{if } \mathcal{M}_\alpha \text{ stops on input } x \\ 0 & \text{otherwise} \end{cases}$$

**Theorem 2.4.2.** The halting problem is undecidable.

*Proof.* Note: this proof is slightly different from the traditional proof of the halting problem.

Assume that halt is decidable. Therefore, there exists a TM  $\mathcal{M}_{halt}$  that decides it. We can define a new TM  $\mathcal{M}_{uc}$  that uses  $\mathcal{M}_{halt}$  such that:

$$\mathcal{M}_{uc}(\alpha) = \begin{cases} 1 & \text{if } \mathcal{M}_{\texttt{halt}}(\alpha, \alpha) = 0 \text{ (i.e. } \mathcal{M}_{\alpha}(\alpha) \text{ diverges)} \\ 0 & \text{if } \mathcal{M}_{\alpha}(\alpha) = 1 \\ 1 & \text{if } \mathcal{M}_{\alpha}(\alpha) \neq 1 \end{cases}$$
 if  $\mathcal{M}_{\texttt{halt}}(\alpha, \alpha) = 1$  (i.e.  $\mathcal{M}_{\alpha}(\alpha)$  converges)

This results in a contradiction:

- $\mathcal{M}_{uc}(\bot \mathcal{M}_{uc} \bot) = 1 \Leftarrow \mathcal{M}_{halt}(\bot \mathcal{M}_{uc} \bot, \bot \mathcal{M}_{uc} \bot) = 0 \iff \mathcal{M}_{uc}(\bot \mathcal{M}_{uc} \bot) \text{ diverges}$
- $\mathcal{M}_{halt}(\sqcup \mathcal{M}_{uc} \sqcup, \sqcup \mathcal{M}_{uc} \sqcup) = 1 \Rightarrow \mathcal{M}_{uc}$  is not computable by Theorem 2.4.1.

**Diophantine equation** Polynomial equality with integer coefficients and a finite number of unknowns.

Diophantine equation

**Theorem 2.4.3** (MDPR). Determining if an arbitrary diophantine equation has a solution is undecidable.

### 2.4.2 Rice's theorem

Semantic language Given a language  $\mathcal{L} \subseteq \{0,1\}^*,\, \mathcal{L}$  is semantic if:

Semantic language

- Any string in  $\mathcal{L}$  is an encoding of a TM.
- If  $\bot \mathcal{M} \bot \in \mathcal{L}$  and the TM  $\mathcal{N}$  computes the same function of  $\mathcal{M}$ , then  $\bot \mathcal{N} \bot \in \mathcal{L}$ .

A semantic language can be seen as a set of TMs that have the same property.

**Trivial language** A language  $\mathcal{L}$  is trivial iff  $\mathcal{L} = \emptyset$  or  $\mathcal{L} = \{0, 1\}^*$ 

**Theorem 2.4.4** (Rice's theorem). If a semantic language is non-trivial, then it is undecidable (i.e. any decidable semantic language is trivial).

Rice's theorem

*Proof idea.* Assuming that there exists a non-trivial decidable semantic language  $\mathcal{L}$ , it is possible to prove that the halting problem is decidable. Therefore,  $\mathcal{L}$  is undecidable.  $\square$ 

# 3 Complexity

Complexity class Set of tasks that can be computed within some fixed resource bounds. Complexity class

## 3.1 Polynomial time

**Deterministic time (DTIME)** Let  $T : \mathbb{N} \to \mathbb{N}$  and  $\mathcal{L}$  be a language.  $\mathcal{L}$  is in  $\mathbf{DTIME}(T(n))$  Deterministic time iff there exists a TM that decides  $\mathcal{L}$  in time O(T(n)).

Polynomial time (P) The class P contains all the tasks computable in polynomial time: Polynomial time (P)

$$\mathbf{P} = \bigcup_{c \geq 1} \mathbf{DTIME}(n^c)$$

**Remark.** P is closed to various operations on programs (e.g. composition of programs)

**Remark.** In practice, the exponent is often small.

**Remark.** P considers the worst case and is not always realistic. Other alternative computational models exist.

**Church-Turing thesis** Any physically realizable computer can be simulated by a TM with an arbitrary time overhead.

Church-Turing thesis

**Strong Church-Turing thesis** Any physically realizable computer can be simulated by a TM with a polynomial time overhead.

Strong Church-Turing thesis

**Remark.** If this thesis holds, the class **P** is robust (i.e. does not depend on the computational device) and is therefore the smallest class of bounds.

**Deterministic time for functions (FDTIME)** Let  $T : \mathbb{N} \to \mathbb{N}$  and  $f : \{0,1\}^* \to \{0,1\}^*$ . f is in **FDTIME**(T(n)) iff there exists a TM that computes it in time O(T(n)).

Deterministic time for functions (FDTIME) Polynomial time for functions (FP)

Polynomial time for functions (FP) The class  $\mathbf{FP}$  is defined as:

$$\mathbf{FP} = \bigcup_{c \geq 1} \mathbf{FDTIME}(n^c)$$

**Remark.** It holds that  $\forall \mathcal{L} \in \mathbf{P} \Rightarrow f_{\mathcal{L}} \in \mathbf{FP}$ , where  $f_{\mathcal{L}}$  is the characteristic function of  $\mathcal{L}$ . Generally, the contrary does not hold.

# 3.2 Exponential time

**Exponential time (EXP/FEXP)** The **EXP** and **FEXP** classes are defined as:

Exponential time (EXP/FEXP)

$$\mathbf{EXP} = \bigcup_{c \geq 1} \mathbf{DTIME} \big( 2^{n^c} \big) \qquad \quad \mathbf{FEXP} = \bigcup_{c \geq 1} \mathbf{FDTIME} \big( 2^{n^c} \big)$$

**Theorem 3.2.1.** The following hold:

$$P \subset EXP$$
  $FP \subset FEXP$ 

## 3.3 NP class

**Certificate** Given a set of pairs  $\mathcal{C}_{\mathcal{L}}$  and a polynomial  $p : \mathbb{N} \to \mathbb{N}$ , we can define the language  $\mathcal{L}$  such that:

$$\mathcal{L} = \{ x \in \{0, 1\}^* \mid \exists y \in \{0, 1\}^{p(|x|)} : (x, y) \in \mathcal{C}_{\mathcal{L}} \}$$

Given a string w and a certificate y, we can exploit  $\mathcal{C}_{\mathcal{L}}$  as a test to check whether y is a certificate for w:

$$w \in \mathcal{L} \iff (w, y) \in \mathcal{C}_{\mathcal{L}}$$

**Nondeterministic TM (NDTM)** TM that has two transition functions  $\delta_0$ ,  $\delta_1$  and, at each step, non-deterministically chooses which one to follow. A state  $q_{\text{accept}}$  is always present:

Nondeterministic TM (NDTM)

- A NDTM accepts a string iff one of the possible computations reaches  $q_{\text{accept}}$ .
- A NDTM rejects a string iff none of the possible computations reach  $q_{\text{accept}}$ .
- **Nondeterministic time (NDTIME)** Let  $T: \mathbb{N} \to \mathbb{N}$  and  $\mathcal{L}$  be a language.  $\mathcal{L}$  is in **NDTIME**(T(n)) iff there exists a NDTM that decides  $\mathcal{L}$  in time O(T(n)).

 $\begin{array}{c} {\rm Nondeterministic} \\ {\rm time} \ ({\bf NDTIME}) \end{array}$ 

**Remark.** A NDTM  $\mathcal{M}$  runs in time  $T: \mathbb{N} \to \mathbb{N}$  iff for every input, any possible computation terminates in time O(T(n)).

### Complexity class NP

Complexity class  $\mathbf{NP}$ 

**NDTM formulation** The class **NP** contains all the tasks computable in polynomial time by a nondeterministic TM:

$$\mathbf{NP} = \bigcup_{c \geq 1} \mathbf{NDTIME}(n^c)$$

**Verifier formulation** Let  $\mathcal{L} \in \{0,1\}^*$  be a language.  $\mathcal{L}$  is in **NP** iff there exists a polynomial  $p: \mathbb{N} \to \mathbb{N}$  and a polynomial TM  $\mathcal{M}$  (verifier) such that:

$$\mathcal{L} = \{x \in \{0,1\}^* \mid \exists y \in \{0,1\}^{p(|x|)} : \mathcal{M}(\mathsf{L}(x,y)\mathsf{J}) = 1\}$$

In other words,  $\mathcal{L}$  is the language of the strings that can be verified by  $\mathcal{M}$  in polynomial time using a certificate y of polynomial length.

Theorem 3.3.1.  $P \subseteq NP \subseteq EXP$ .

*Proof.* We have to prove that  $P \subseteq NP$  and  $NP \subseteq EXP$ :

 $P \subseteq NP$ ) Given a language  $\mathcal{L} \in P$ , we want to prove that  $\mathcal{L} \in NP$ .

By hypothesis, there is a polynomial time TM  $\mathcal{N}$  that decides  $\mathcal{L}$ . To prove that  $\mathcal{L}$  is in  $\mathbf{NP}$ , we show that there is a polynomial verifier  $\mathcal{M}$  that certifies  $\mathcal{L}$  with a polynomial certificate. We can use any constant certificate (e.g. of length 1) and use  $\mathcal{N}$  as the verifier  $\mathcal{M}$ :

$$\mathcal{M}(x,y) = \begin{cases} 1 & \text{if } \mathcal{N}(x) = 1\\ 0 & \text{otherwise} \end{cases}$$

 $\mathcal{M}$  can ignore the polynomial certificate and it "verifies" a string in polynomial time through  $\mathcal{N}$ .

 $NP \subseteq EXP$ ) Given a language  $\mathcal{L} \in NP$ , we want to prove that  $\mathcal{L} \in EXP$ .

By hypothesis, there is a polynomial time TM  $\mathcal{N}$  that is able to certify any string in  $\mathcal{L}$  with a polynomial certificate. Given a polynomial p, can define the following algorithm:

```
\begin{array}{ll} \texttt{def} & \texttt{np\_to\_exp}\,(x \in \{0,1\}^*): \\ & \texttt{foreach} \ y \in \{0,1\}^{p(|x|)}: \\ & \texttt{if} \ \mathcal{M}(x,y) == 1: \\ & \texttt{return} \ 1 \\ & \texttt{return} \ 0 \end{array}
```

The algorithm has complexity  $O(2^{p(|x|)}) \cdot O(q(|x|+|y|)) = O(2^{p(|x|)+\log(q(|x|+|y|))})$ , where q is a polynomial. Therefore, the complexity is exponential.

**Polynomial-time reducibility** A language  $\mathcal{L}$  is poly-time reducible to  $\mathcal{H}$  ( $\mathcal{L} \leq_p \mathcal{H}$ ) iff:

Polynomial-time reducibility

$$\exists f: \{0,1\}^* \to \{0,1\}^* \text{ such that } (x \in \mathcal{L} \iff f(x) \in \mathcal{H}) \text{ and}$$

$$f \text{ is computable in poly-time}$$

f can be seen as a mapping function.

**Remark.** Intuitively, when  $\mathcal{L} \leq_p \mathcal{H}$ ,  $\mathcal{H}$  is at least as difficult as  $\mathcal{L}$ .

**Theorem 3.3.2.** The relation  $\leq_p$  is a pre-order (i.e. reflexive and transitive).

*Proof.* We want to prove that  $\leq_p$  is reflexive and transitive:

**Reflexive)** Given a language  $\mathcal{L}$ , we want to prove that  $\mathcal{L} \leq_{p} \mathcal{L}$ .

We have to find a poly-time function  $f: \{0,1\}^* \to \{0,1\}^*$  such that:

$$x \in \mathcal{L} \iff f(x) \in \mathcal{L}$$

We can choose f as the identity function.

**Transitive)** Given the languages  $\mathcal{L}, \mathcal{H}, \mathcal{J}$ , we want to prove that:

$$(\mathcal{L} \leq_n \mathcal{H}) \wedge (\mathcal{H} \leq_n \mathcal{J}) \Rightarrow (\mathcal{L} \leq_n \mathcal{J})$$

By hypothesis, it holds that  $\mathcal{L} \leq_p \mathcal{H}$  and  $\mathcal{H} \leq_p \mathcal{J}$ . Therefore, there are two poly-time functions  $f, g : \{0, 1\}^* \to \{0, 1\}^*$  such that:

$$x \in \mathcal{L} \iff f(x) \in \mathcal{H} \text{ and } y \in \mathcal{H} \iff f(y) \in \mathcal{J}$$

We want to find a poly-time mapping from  $\mathcal{L}$  to  $\mathcal{J}$ . This function can be the composition  $(g \circ f)(z) = g(f(z))$ .  $(g \circ f)$  is poly-time as f and g are poly-time.

**NP-hard** Given a language  $\mathcal{H} \in \{0,1\}^*,\,\mathcal{H}$  is **NP-**hard iff:

NP-hard

$$\forall \mathcal{L} \in \mathbf{NP} : \mathcal{L} \leq_p \mathcal{H}$$

**NP-complete** Given a language  $\mathcal{H} \in \{0,1\}^*$ ,  $\mathcal{H}$  is **NP-complete** iff:

NP-complete

 $\mathcal{H} \in \mathbf{NP}$  and  $\mathcal{H}$  is  $\mathbf{NP}$ -hard

#### Theorem 3.3.3.

- 1. If  $\mathcal{L}$  is **NP**-hard and  $\mathcal{L} \in \mathbf{P}$ , then  $\mathbf{P} = \mathbf{NP}$ .
- 2. If  $\mathcal{L}$  is **NP**-complete, then  $\mathcal{L} \in \mathbf{P} \iff \mathbf{P} = \mathbf{NP}$ .

Proof.

- 1. Let  $\mathcal{L}$  be **NP**-hard and  $\mathcal{L} \in \mathbf{P}$ . We want to prove that  $\mathbf{P} = \mathbf{NP}$ :
  - $P \subseteq NP$ ) Proved in Theorem 3.3.1.
  - $NP \subseteq P$ ) Let  $\mathcal{H}$  be a language in NP. As  $\mathcal{L}$  is NP-hard, by definition it holds that  $\mathcal{H} \leq_p \mathcal{L}$ . Moreover, by hypothesis, it holds that  $\mathcal{L} \in P$ . Therefore, we can conclude that  $\mathcal{H} \in P$  as it can be reduced to a language in P.
- 2. Let  $\mathcal{L}$  be **NP**-complete. We want to prove that  $\mathcal{L} \in \mathbf{P} \iff \mathbf{P} = \mathbf{NP}$ :

$$(\mathcal{L} \in \mathbf{P}) \Rightarrow (\mathbf{P} = \mathbf{NP})$$
) Trivial for Point 1 as  $\mathcal{L}$  is also  $\mathbf{NP}$ -hard.

$$(\mathcal{L} \in P) \Leftarrow (P = NP)$$
) Let  $P = NP$ . As  $\mathcal{L}$  is  $NP$ -complete, it holds that  $\mathcal{L} \in NP = P$ .

Theorem 3.3.4. The problem TMSAT of simulating any TM is NP-complete:

$$\texttt{TMSAT} = \{(\alpha, x, 1^n, 1^t) \mid \exists u \in \{0, 1\}^n : \mathcal{M}_{\alpha}(x, u) = 1 \text{ within } t \text{ steps}\}$$

**Theorem 3.3.5** (Cook-Levin). The following languages are **NP**-complete:

Cook-Levin theorem

$$\mathtt{SAT} = \{ \bot F \ | \ F \text{ is a satisfiable CNF} \}$$
 
$$\mathtt{3SAT} = \{ \bot F \ | \ F \text{ is a satisfiable 3CNF} \}$$