

# Computational Complexity

Lecture 1: P, NP and NP-completeness

Ronald de Haan

me@ronalddehaan.eu

University of Amsterdam

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- Lecturer: Ronald de Haan ([me@ronalddehaan.eu](mailto:me@ronalddehaan.eu))
- TAs: Wouter Vromen, Hannah Van Santvliet
- Course web page: <https://staff.science.uva.nl/r.dehaan/complexity2024/>
- Canvas page: <https://canvas.uva.nl/courses/42595>
- Discourse: <https://talk.computational-complexity.nl/>
- Book: *Computational Complexity: A Modern Approach* (Arora & Barak, 2009)

- We'll use an online discussion board (using the *Discourse* system):  
<https://talk.computational-complexity.nl/>
  - Questions about the material  
(Feel free to answer each other's questions)
  - Reflecting on the material
  - Summarizing the material together
  
- Feel free to start discussion topics on any of these

- During the course:
  - Please give your ideas for improvement, e.g., anonymously on Discourse
- After the course:
  - Please fill in the course evaluation questionnaire (for the OC and lecturer)

- Lectures:
  - Twice 45 minutes, with 15 minute break in between, **not recorded**
- Exercise sessions:
  - Practice with material, discuss previous homework assignments
- Homework assignments (50% of grade):
  - Three assignments, hand in via Canvas
  - You may work in pairs (but you do not have to)
- Take-home exam (50% of grade):
  - At the end, open book, one week time to complete exam
- Online discussions, question answering

## What will we do today?

- Decision problems
- The complexity class P
- Nondeterministic Turing machines
- More complexity classes: EXP, NP, coNP
- Polynomial-time reductions
- NP-hardness and NP-completeness

## Quadratic vs. Exponential

- Important difference between algorithms that run in time, say,  $n^2$  vs. algorithms that run in time, say,  $2^n$
- Illustration (time needed for  $10^{10}$  steps per second):

$n$	$n^2$ steps	$2^n$ steps
2	0.00000002 msec	0.00000002 msec
5	0.00000015 msec	0.00000019 msec
10	0.00001 msec	0.0001 msec
20	0.00004 msec	0.10 msec
50	0.00025 msec	31.3 hours
100	0.001 msec	$9.4 \times 10^{11}$ years
1000	0.100 msec	$7.9 \times 10^{282}$ years

- # of atoms in universe  $\approx 10^{80}$

- To simplify the theory, we restrict our attention to yes/no questions

### Definition (Decision problems)

A *decision problem* is a function  $f : \Sigma^* \rightarrow \{0, 1\}$  where for each input  $x \in \Sigma^*$  the correct output  $f(x)$  is either 0 or 1.

Alternatively: a formal language  $L \subseteq \Sigma^*$  where  $x \in L$  if and only if  $f(x) = 1$ .

- For decision problems, we typically look at TMs that have two halting states:  
 $q_{\text{acc}}$  (for *accept*:  $f(x) = 1$ )  
and  $q_{\text{rej}}$  (for *reject*:  $f(x) = 0$ )



### Definition (polynomial-time computability)

A function  $f : \Sigma^* \rightarrow \Sigma^*$  is *polynomial-time computable* (or *computable in polynomial time*) if there exist a TM  $M$  and a constant  $c \in \mathbb{N}$  such that:

- $M$  computes  $f$
- $M$  runs in time  $O(|x|^c)$

### Definition (the complexity class P)

P is the class (set) consisting of all decision problems  $L \subseteq \Sigma^*$  that are computable in polynomial time.

- **Tractability**: there exists a polynomial-time algorithm that solves the problem
- **Intractability**: there exists **no** polynomial-time algorithm that solves the problem  
(or sometimes phrased as: all algorithms that solve the problem take exponential time or more, in the worst case)
- How do we find out which of these two is the case?

## Showing intractability: without any theory



**“I can’t find an efficient algorithm, I guess I’m just too dumb.”**

## Showing intractability: the ideal case



**“I can’t find an efficient algorithm, because no such algorithm is possible!”**

## Showing intractability: using NP-completeness



**“I can’t find an efficient algorithm, but neither can all these famous people.”**

### Definition (DTIME)

Let  $T : \mathbb{N} \rightarrow \mathbb{N}$  be a function. A language  $L \subseteq \Sigma^*$  is in  $\text{DTIME}(T(n))$  if there exists a Turing machine that decides  $L$  and that runs in time  $O(T(n))$ .

### Definition (the complexity classes P and EXP)

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

$$\text{EXP} = \bigcup_{c \geq 1} \text{DTIME}(2^{n^c})$$

### Definition (the complexity class NP)

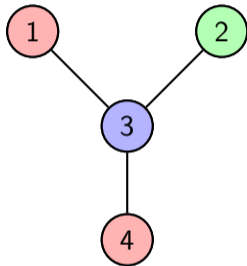
A problem  $L \subseteq \Sigma^*$  is in the complexity class NP if there is a polynomial  $p : \mathbb{N} \rightarrow \mathbb{N}$  and a polynomial-time Turing machine  $\mathbb{M}$  (the *verifier*) such that for every  $x \in \Sigma^*$ :

$x \in L$  if and only if there exists some  $u \in \{0, 1\}^{p(|x|)}$  such that  $\mathbb{M}(x, u) = 1$ .

The string  $u \in \{0, 1\}^{p(|x|)}$  is called a *certificate* for  $x$  if  $\mathbb{M}(x, u) = 1$ .

## Example problem: 3-coloring

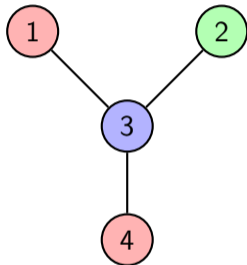
- You are given an undirected graph
- The task is to color each node with one of 3 colors so that the coloring is **proper**: **no two connected nodes have the same color**
- *Example application*: nodes are regions with their own radio station, colors are radio frequencies, and two nodes are connected if the regions border each other; assign radio frequencies without conflict





## Example problem: 3-coloring (ct'd)

- Let's see why the (decision) problem of 3-coloring is in NP.
- Let  $G = (V, E)$  be a graph with  $m$  nodes.
- Consider as witness a binary string  $u$  of length  $2m$ , where the coloring of each node  $i$  is given by the  $i$ 'th pair of bits—say, 01 for red, 10 for green, and 11 for blue.
- Given  $G$  and  $u$ , we can check in polynomial time if the coloring given by  $u$  is *proper*.



$s = 01\ 10\ 11\ 01$

## Definition

A *nondeterministic Turing machine (NTM)*  $\mathbb{M}$  is a variant of a (deterministic) Turing machine, where some things are modified.

- Instead of a single transition function  $\delta$ , there are two transition functions  $\delta_1, \delta_2$ .
  - At each step, one of  $\delta_1, \delta_2$  is chosen nondeterministically to determine the next configuration.
  - (As halting states, it has an accept state  $q_{acc}$  and a reject state  $q_{rej}$ .)
- 
- We write  $\mathbb{M}(x) = 1$  if there is some sequence of nondeterministic choices such that  $\mathbb{M}$  reaches the state  $q_{acc}$  on input  $x$ .
  - The machine  $\mathbb{M}$  runs in time  $T(n)$  if for every input  $x$  and every sequence of nondeterministic choices,  $\mathbb{M}$  halts within  $T(|x|)$  steps.

## Nondeterministic polynomial time (NP)

### Definition (NTIME)

Let  $T : \mathbb{N} \rightarrow \mathbb{N}$  be a function. A problem  $L \subseteq \Sigma^*$  is in  $\text{NTIME}(T(n))$  if there exists a nondeterministic Turing machine that decides  $L$  and that runs in time  $O(T(n))$ .

### Proposition (characterization of NP)

$$\text{NP} = \bigcup_{c \geq 1} \text{NTIME}(n^c)$$

### Definition (the complexity class coNP)

A problem  $L \subseteq \Sigma^*$  is in coNP if  $\bar{L} \in \text{NP}$ , where  $\bar{L} = \{ x \in \Sigma^* \mid x \notin L \}$ .

### Proposition (verifier characterization of coNP)

A problem  $L \subseteq \Sigma^*$  is in coNP if there is a polynomial  $p : \mathbb{N} \rightarrow \mathbb{N}$  and a polynomial-time Turing machine  $\mathbb{M}$  (the *verifier*) such that for every  $x \in \Sigma^*$ :

$x \in L$  if and only if **for all**  $u \in \{0, 1\}^{p(|x|)}$  it holds that  $\mathbb{M}(x, u) = 1$ .

## Proposition

NP  $\subseteq$  EXP.

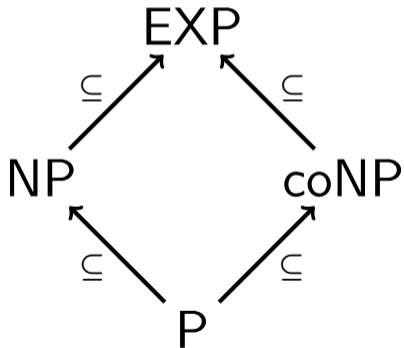
## Proof (idea).

- Iterate over all possible witnesses  $u \in \{0, 1\}^{p(|x|)}$ , and check if  $M(x, u) = 1$ .
- If for any  $u$  this is the case, return 1—otherwise, return 0.
- There are  $2^{p(|x|)}$  such strings  $u$ , and so this takes time  $2^{p(|x|)} \cdot q(|x|)$ , for some polynomial  $q$ .



# An overview of complexity classes

*(That we've seen so far..)*

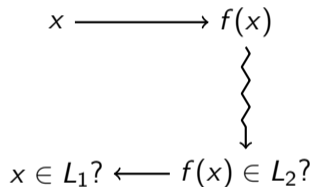


## Definition (polynomial-time reductions)

A problem  $L_1 \subseteq \Sigma^*$  is *polynomial-time reducible* to a problem  $L_2 \subseteq \Sigma^*$  if there is a polynomial-time computable function  $f : \Sigma^* \rightarrow \Sigma^*$  (the *reduction*) such that for every  $x \in \Sigma^*$  it holds that:

$$x \in L_1 \quad \text{if and only if} \quad f(x) \in L_2.$$

- We write  $L_1 \leq_p L_2$  to indicate that  $L_1$  is polynomial-time reducible to  $L_2$ .



### Definition (NP-hardness)

A problem  $L \subseteq \Sigma^*$  is *NP-hard* if every problem in NP is polynomial-time reducible to  $L$ .

### Definition (NP-completeness)

A problem  $L \subseteq \Sigma^*$  is *NP-complete* if  $L \in \text{NP}$  and  $L$  is NP-hard.



### Proposition

Polynomial-time reductions are transitive.

That is, if  $L_1 \leq_p L_2$  and  $L_2 \leq_p L_3$ , then  $L_1 \leq_p L_3$ .

### Proposition

Take two problems  $L_1, L_2 \subseteq \Sigma^*$ . If  $L_1$  is polynomial-time reducible to  $L_2$  and  $L_2 \in P$ , then  $L_1 \in P$ .

### Proposition

Take an NP-complete problem  $L \subseteq \Sigma^*$ . If  $L \in P$ , then  $P = NP$ .  
In other words, assuming that  $P \neq NP$ ,  $L \notin P$ .

### Proof.

Since deterministic TMs can be seen also as nondeterministic TMs, we get  $P \subseteq NP$ .

We show that if  $L \in P$ , then  $NP \subseteq P$ .

- (1) Take an arbitrary problem  $M \in NP$ .
- (2) Since  $L$  is NP-complete,  $M \leq_p L$ .
- (3) Since  $L \in P$ , then also  $M \in P$ .

Since  $M$  was arbitrary, we know that  $NP \subseteq P$ . □

## Showing intractability: using NP-completeness



**“I can’t find an efficient algorithm, but neither can all these famous people.”**

- Decision problems
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- Proving that NP-complete problems exist :-)