



COMPLEXITY THEORY

Lecture 1: Introduction and Motivation

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TU Dresden, 13 October 2025

World recent versions of mins slide deck might be available. For the most current version of this course, see https://iccl.inf.tu-dresden.de/web/Complexity_Theory/

Course Tutors



Markus Krötzsch Lectures



Stephan Mennicke Lectures/Exercises



Sergei Obiedkov Lectures/Exercises

Organisation

Lectures

Monday, DS 4 (13:00–14:30), APB E005 Tuesday, DS 3 (11:10–12:40), MER 0E23

Exercise Sessions (starting 21 October)

Tuesday, DS 5 (14:50-16:20), APB E005

Web Page

https://iccl.inf.tu-dresden.de/web/Complexity_Theory_(WS2025)

Lecture Notes

Slides of current and past lectures will be online.

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Goals and Prerequisites

Goals

- Introduce basic notions of computational complexity theory
- Introduce **commonly known complexity classes** (P, NP, PSpace, ...) and discuss relationships between them
- Develop tools to classify problems into their corresponding complexity classes
- Introduce some advanced topics of complexity theory (e.g., circuits, probabilistic computation, quantum computing)

(Non-)Prerequisites

- No particular prior courses needed
- Prior acquaintance with Turing Machines and basic topics in formal languages and complexity is helpful
- General mathematical and theoretical computer science skills necessary

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Reading List

- Michael Sipser: Introduction to the Theory of Computation, International Edition; 3rd Edition; Cengage Learning 2013
- Sanjeev Arora and Boaz Barak: Computational Complexity: A Modern Approach; Cambridge University Press 2009
- Michael R. Garey and David S. Johnson: Computers and Intractability; Bell Telephone Laboratories, Inc. 1979
- Erich Grädel: Complexity Theory; Lecture Notes, Winter Term 2009/10
- John E. Hopcroft and Jeffrey D. Ullman: Introduction to Automata Theory,
 Languages, and Computation; Addison Wesley Publishing Company 1979
- Christos H. Papadimitriou: Computational Complexity; 1995 Addison-Wesley Publishing Company, Inc

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Computational Problems are Everywhere

Example 1.1:

- What are the factors of 54,623?
- What is the shortest route by car from Berlin to Hamburg?
- My program now runs for two weeks. Will it ever stop?
- Is this C++ program syntactically correct?

Clear

Computational Problems are ubiquitous in our everyday life! And, depending on what we want to do, those problems might be either **easily solvable** or **hardly solvable**.

Approach to problems:

[T]he way is to avoid what is strong, and strike at what is weak.

(Sun Tzu: The Art of War, Chapter 6: Weak Points and Strong)

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Examples

Example 1.2 (Shortest Path Problem): Given a weighted graph and two vertices s, t, find the shortest path between s and t.

Easily solvable using, e.g., Dijkstra's Algorithm.

Example 1.3 (Longest Path Problem): Given a weighted graph and two vertices s, t, find the **longest** path between s and t.

No efficient algorithm known, and believed to not exist (this problem is **NP-hard**)

Observation

Difficulty of a problem is hard to assess

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Measuring the Difficulty of Problems

Question

How can we measure the complexity of a problem?

Approach

Estimate the resource requirements of the "best" algorithm that solves this problem.

Typical Resources:

- Running Time
- Memory Used

Note

To assess the complexity of a problem, we need to consider **all possible algorithms** that solve this problem.

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Problems

What actually is ... a Problem?

(Decision) Problems are word problems of particular languages.

Example 1.4: "Problem: Is a given graph connected?" will be modelled as the word problem of the language

GCONN :=
$$\{\langle G \rangle \mid G \text{ is a connected graph } \}$$
.

Then for a graph G we have

$$G$$
 is connected $\iff \langle G \rangle \in \mathsf{GCONN}$.

Note

The notation $\langle G \rangle$ denotes a **suitable encoding** of the graph G over some fixed alphabet (e.g., $\{0,1\}$).

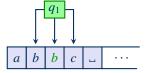
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Algorithms

What actually is ... an Algorithm?

Different approaches to formalise the notion of an "algorithm"

- Turing Machines
- Lambda Calculus
- μ-Recursion
- ...



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Avoid What is Strong

Suppose we are given a language \mathcal{L} and a word w.

Question

Does there need to exist **any** algorithm that decides whether $w \in \mathcal{L}$?

Answer

No. Some problems are undecidable.

Example 1.5:

- The Halting Problem of Turing machines
- The Entscheidungsproblem (Is a first-order logical statement true?)
- Finding the lowest air fare between two cities (→ Reference)
- Deciding syntactic validity of C++ programs (→ Reference)

Avoid: We will focus mostly on decidable problems in this course.

Time and Space

Difficulty

Measuring running time and memory requirements depends highly on the **machine**, and not so much on the **problem**.

Resort

Measure time and space only **asymptotically** using **Big-***O*-Notation:

$$f(n) = O(g(n)) \iff f(n)$$
 "asymptotically bounded by" $g(n)$

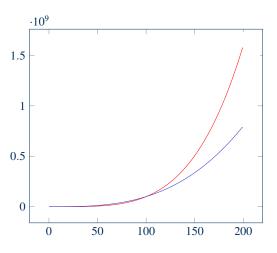
More formally:

$$f(n) = O(g(n)) \iff \exists c > 0 \, \exists n_0 \in \mathbb{N} \, \forall n > n_0 \colon f(n) \le c \cdot g(n).$$

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Big-O-Notation: Example

$$100n^3 + 1729n = O(n^4)$$
:



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Complexity of Problems

Approach

The **time (space) complexity** of a problem is the asymptotic running time of a fastest (least memory consumptive) algorithm that solves the problem.

Problem

Still too difficult

Example 1.6 (Travelling Salesman Problem): Given a weighted graph, find the shortest simple path visiting every node.

- Best known algorithm runs in time O(n²2ⁿ) (Bellman-Held-Karp algorithm)
- Best known lower bound is $O(n \log n)$
- Exact complexity of TSP unknown

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Even more abstraction

Approach

Divide decision problems into the "quality" of their fastest algorithms:

- P is the class of problems solvable in polynomial time
- PSpace is the class of problems solvable in polynomial space
- ExpTime is the class of problems solvable in exponential time
- L is the class of problems **solvable in logarithmic space** (apart from the input)
- NP is the class of problems verifiable in polynomial time
- NL is the class of problems verifiable in logarithmic space

And many more!

 \oplus P, #P, AC, AC⁰, ACC0, AM, AP, APSpace, BPL, BPP, BQP, coNP, E, FP, IP, MA, MIP, NC, NExpTime, P/poly, PH, PP, RL, RP, Σ_i^p , TISP(T(n), S(n)), ZPP, ...

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Strike at What is Weak

Approach (cf. Cobham–Edmonds Thesis)

The problems in P are "tractable" or "efficiently solvable" (and those outside are not)

Example 1.7: The following problems are in P:

- Shortest Path Problem
- Satisfiability of Horn-Formulas
- Linear Programming
- Primality

Note

The Cobham-Edmonds-Thesis is only a **rule of thumb**: there are (practically) tractable problems outside of P, and (practically) intractable problems in P.

Friend or Foe?

Caveat

It is not known how big P is. In particular, it is unknown whether $P \neq NP$ or not.

Approach

Try to find out which problems in a class are at least as hard as others. **Complete** problems are then the hardest problems of a class.

Example 1.8: Satisfiability of propositional formulas is **NP-complete**: if we can efficiently decide whether a propositional formula is satisfiable, we can solve **any** problem in NP efficiently.

But: we still do not know whether we can or cannot solve satisfiability efficiently. We only know it will be difficult to find out ...

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Learning Goals

- Get an overview over the foundations of Complexity Theory
- Gain insights into advanced techniques and results in Complexity Theory
- Understand what it means to "compute" something, and what the strengths and limits of different computing approaches are
- Get a feeling of how hard certain problems are, and where this hardness comes from
- Appreciate how very little we actually know about the computational complexity of many problems

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Lecture Outline (1)

• Turing Machines (Revision)

Definition of Turing Machines; Variants; Computational Equivalence; Decidability and Recognizability; Enumeration; Oracles

Undecidability

Examples of Undecidable Problems; Mapping Reductions; Rice's Theorem; Recursion Theorem

• Time Complexity

Measuring Time Complexity; Many-One Reductions; Cook-Levin Theorem; Time Complexity Classes (P, NP, ExpTime); NP-completeness; pseudo-NP-complete problems

Space Complexity

Space Complexity Classes (PSpace, L, NL); Savitch's Theorem; PSpace-completeness; NL-completeness; NL = coNL

Lecture Outline (2)

Diagonalisation

Hierarchy Theorems (det. Time, non-det. Time, Space); Gap Theorem; Ladner's Theorem; Relativisation; Baker-Gill-Solovay Theorem

Alternation

Alternating Turing Machines; APTime = PSpace; APSpace = ExpTime; Polynomial Hierarchy

Circuit Complexity

Boolean circuits; alternative proof of Cook-Levin Theorem; parallel computation (NC); P-completeness; P/poly; (Karp-Lipton Theorem, Meyer's Theorem)

Probabilistic Computation

Randomised complexity classes (RP, PP, BPP, ZPP); Sipser-Gács-Lautemann Theorem

Quantum Computing

Quantum mechanics for computer scientists, entanglement, quantum circuits, BQP

Interactive Proofs

Prover and verifier; deterministic proof systems; probabilistic verifiers; the class IP

Looking for Project or Thesis Topics?

On **Thursday**, **16 Oct 2025 at 2.50pm in APB 3027** we will present possible topics to conduct in the **Knowledge-Based Systems** group as a **study project** (many suitable modules) or **final thesis** (BSc, MSc, Diploma).

Not only theoretical topics but also implementation work.

We also have student job opportunities (SHK/WHK).

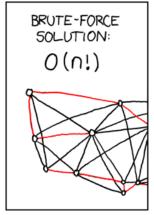
You are especially welcome if you are eager to work with Rust or LEAN:)

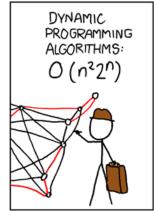
See also: https://iccl.inf.tu-dresden.de/web/Projekte_und_ Studienarbeiten_Wissensbasierte_Systeme/en

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Avoid what is Strong, and Strike at what is Weak

Sometimes the best way to solve a problem is to avoid it ...







https://xkcd.com/399/ 2.5 Randall Munroe,