Computational complexity

lecture 1

Introductory announcements

- Lectures will be in English
- Slides (and other materials) will be available on the webpage

Introductory announcements

Grading rules:

- homeworks (3 series) → 1.5 pt
- mid-term \rightarrow 1.5 pt
- everyone can write the June exam (no lower limit for homeworks + mid-term)
- final exam → 3 pt
 the final grade depends on: homeworks + mid-term + exam (max=6)
- final exam, second try \rightarrow 4.5 pt the final grade depends on: homeworks + exam (max = 6 pt)
- the exam consists of two parts: theory + "practical" problems

What is this course about?

- the subject of research: computational problems
 - → the basic question: how fast a given problem can be solved?
 - → what does it mean that a problem is difficult? (maybe we are not intelligent enough to solve it?)
- JAiO: there are decidable and undecidable problems
 - → here: (decidable) problems can be easy or hard (in different senses)
- algorithmics: we have a fast algorithm
 - → here: a fast algorithm does not exist

- David Hilbert: Is there an algorithm deciding any mathematical hypothesis? ("Entscheidungsproblem" - 1928)
- Alonzo Church (1936), Alan Turing (1937) NO
- The answer required a formal definition (what does it mean "an algorithm"?)
 - → a Turing machine (lambda-calculus Church)

Components of a (multitape) Turing machine:

- number of tapes k
- a finite working alphabet Γ , where $\triangleright, \perp \in \Gamma$
- an input alphabet $\Sigma \subseteq \Gamma \setminus \{ \triangleright, \bot \}$
- a finite set of states Q
- states: initial, accepting, rejecting $q_{\scriptscriptstyle I}, q_{\scriptscriptstyle A}, q_{\scriptscriptstyle R} \in Q$
- a transition function $\delta:(Q\setminus\{q_A,q_R\})\times\Gamma^k\to Q\times\Gamma^k\times\{L,R,Z\}^k$ such that ...

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A <u>configuration</u> of a Turing machine consists of:

- contents of *k tapes*; each of them is infinite to the right
- location of the *head* on every tape
- state

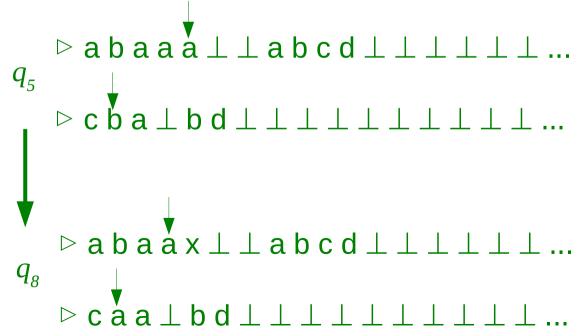
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Example (2 tapes): q_5 \triangleright a b a a a \bot \bot a b c d \bot \bot \bot \bot ... \bot c b a \bot b d \bot \bot \bot \bot \bot \bot \bot \bot ...
```

A computation: a function \rightarrow_{M} between configurations

Example:

$$\delta(q_5,a,b)=(q_8,x,a,L,Z)$$
 usual notation: $q_5,a,b \rightarrow q_8,x,a,L,Z$

- if the state is q_5 , letters under the heads are a (on tape 1), b (on tape 2)
- then change the state to $q_{_{\it 8}}$
- write x on tape 1, a on tape 2
- move head 1 to the left, do not move head 2



Additional assumptions about a configuration:

- from some moment on every tape there are only \perp symbols
- the first symbol on every tape is ▷
- the symbol ▷ never appears later

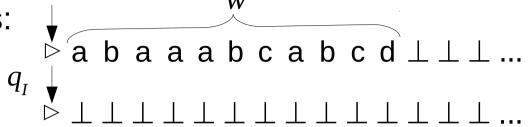
Additional assumptions about a transition function:

- the machine never replaces > by any other symbol
- the machine never writes > when it was not there
- the machine never wants to go left when it sees ▷

(in particular, this ensures that every configuration has a successor, unless the state is $q_{\scriptscriptstyle A}$ or $q_{\scriptscriptstyle B}$)

A <u>computation</u> on an input word $w \in \Sigma^*$:

• the initial configuration is:



- the machine accepts w, if it reaches a configuration with state $q_{\scriptscriptstyle A}$
- the machine *rejects* w, if it reaches a configuration with state $q_{\scriptscriptstyle R}$
- otherwise the computation is infinite (the machine *loops*)

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- otherwise the computation is infinite (the machine *loops*)
- → notation: $L(M) = \{w : M \text{ accepts } w\}$
- \rightarrow *M* has the *halting property*, if it halts on every input
- \Rightarrow a language $L \subseteq \Sigma^*$ is *semidecidable* (or *recursively enumerable*) if there exists a machine that accepts exactly words from L (i.e., L(M)=L)
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Computing functions:

- \Rightarrow a partial function $f: \Sigma^* \to \Sigma^*$ is computable, if there exists a machine M
- → that accepts every word w∈dom(f), ending in a configuration with $\triangleright f(w)\bot^{\infty}$ on the last tape,
- → and rejects every word w∉dom(f)

Variants of Turing machines

- a tape that is infinite in both directions
- multiple accepting / rejecting states
- a single tape only
- never writes ⊥
- nondeterministic machines, alternating machines (the machines defined above were deterministic)

• ...

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Fact: All variants recognize the same class of languages. [i.e., the notion of a Turing machine is *robust*]

Remark: it is enough to prove that for every machine of type X there exists an equivalent machine of type Y. In practice these constructions are computable, but to obtain the above fact we do not need to know this.

Such a distinction often appears on this lecture: when it is enough that something exists, and when we have to know how to (quickly) compute it?

Time complexity

A machine M works in time T(n) (for a function $T:\mathbb{N}\to\mathbb{N}$) if for every word $w\in\Sigma^*$ it halts after at most T(|w|) steps. (in particular it has the halting property)

A language $L \subseteq \Sigma^*$ is decidable in time T(n) if there exists a (**multitape**) machine that recognizes this language and works in time T(n). We usually talk about the asymptotic behavior of the complexity, i.e., that the time is O(T(n)).

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Theorem (linear speed-up):

If a language L is decidable in time T(n), then for every constant c>0 it is also decidable in time $c \cdot T(n) + O(n)$.

<u>Proof</u>: on tutorials (the idea: one counts the number of steps, so it is enough to simulate multiple steps while performing a single step).

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Solution – allow only *off-line machines*:

- the input tape is read-only (when I see \bot , I cannot move right)
- working tapes





In space complexity we do not include the length of the input (important when space complexity is smaller than linear)

Formally, we allow only *off-line machines*:

- the input tape is read-only (when I see \bot , I cannot move right)
- working tapes
- while computing functions: output tape, where the head cannot move left (i.e., write-only)

A machine M works in space S(n) (for a function $S:\mathbb{N}\to\mathbb{N}$) if for every word $w\in\Sigma^*$ visits at most S(|w|) cells on its working tapes.

A language $L \subseteq \Sigma^*$ is *recognizable in space* S(n) if there exists a multitape machine that <u>halts on every input</u>, accepts L, and works in space S(n).

Usually we talk about space O(S(n)) (asymptotic behavior).

It is easy to reduce space usage "times a constant" - we remember a few cells in one.

It is possible to convert a multitape machine into a machine with one working tape, which works in the same space.

Machines vs. languages

- Sometimes we talk about time / space complexity of a language (there exists a machine such that ...)
- Sometimes we talk about working time / space of a particular machine (particular algorithm)

Languages vs. decision problems

Example: reachability in a graph

- Input: a set of nodes, a set of edges, two distinguished nodes.
 - → The input is not a word, it is a more complicated object.
 - → A Turing machine reads words.
- But a graph can be written as a word:
 number_of_nodes,
 number_of_edges,
 a list of pair of nodes connected by edges (where we assume that
 nodes are numbered by consecutive natural number);
 particular numbers are separated by a special \$ sign.
 - → Multiple possible representations of a graph
 - → It is easy to convert from one representation to another.
- Usually, we talk about <u>complexity of a problem</u> which means "complexity of the corresponding language under some natural representation of inputs as words" (typically the complexity does not depend on the choice of the representation)
- Sometimes it depends, and then we should be more precise (we should say which representation of inputs is considered)

Languages vs. decision problems

- While considering a concrete problem we think about an algorithm understood in an abstract way, and usually we do not refer to a particular representation – but we are aware that it is possible to implement basic programming concepts (variables, loops, etc.) on a Turing machine
- While proving general theorems we consider Turing machines (a model that is simple, but strong enough).

Church-Turing thesis

Church-Turing thesis: every physically realizable computation device can be simulated by a Turing machine.

(this is not a mathematical theorem - it is not sure what can be physically realizable)

A stronger thesis: problems "easy" for other devices are also "easy" for Turing machines – every physically realizable computation device can be simulated by a Turing machine <u>with polynomial overhead</u>.

Random Access Machine (RAM)

[This is a side remark – RAM machines will not appear more during the lecture]

A model close to computers than Turing machines:

- Cells contain arbitrarily large numbers (instead of letters from a finite alphabet)
- A program amounts to a list of instructions. Available instructions: $X[i] \leftarrow k$ (where i,j,k,m constants written in a program) $X[i] \leftarrow X[j] + X[k]$

```
X[i] \leftarrow X[j] + X[K]

X[i] \leftarrow X[j] - X[k]

X[i] \leftarrow X[X[j]]

X[X[i]] \leftarrow X[j]

if X[i] > 0 then goto m
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- Every operation is performed in constant time (by definition)
- There is no multiplication it can be realized in time linear in the number of bits

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- Every operation is performed in constant time (by definition)
- There is no multiplication it can be realized in time linear in the number of bits
- Input (and output) in cells X[1],...,X[n]; additionally X[0]=n
- The size of the input is defined as the total number of bits

Random Access Machine (RAM)

- A computation of a Turing machine using time T(n) can be simulated on RAM in time O(T(n))
- A computation of a RAM using time T(n) can be simulated on a Turing machine in time $O(T(n)^3)$

Time complexity – basic classes

[Now we come back to Turing machines]

- DTIME(T(n)) languages recognizable in time O(T(n))
- P = $\bigcup_{k \in \mathbb{N}} \mathsf{DTIME}(n^k)$ i.e., languages recognizable in time p(n) for some polynomial p
- EXPTIME = $\bigcup_{k \in \mathbb{N}} \mathsf{DTIME}(2^{n^k})$

Space complexity – basic classes

- DSPACE(S(n)) languages recognizable in space O(S(n))
- L = $\bigcup_{k \in \mathbb{N}} \mathsf{DSPACE}(\log n^k) = \mathsf{DSPACE}(\log n)$
- PSPACE = $\bigcup_{k \in \mathbb{N}}$ DSPACE (n^k) i.e., languages recognizable in space p(n) for some polynomial p
- EXPSPACE = $\bigcup_{k \in \mathbb{N}} DSPACE(2^{n^k})$

 $\mathsf{DTIME}(f(n)) \subseteq \mathsf{DSPACE}(f(n))$

Proof: In time f(n) a machine can visit at most $k \cdot f(n)$ cells (k = the number of tapes)

Conversely: DSPACE(f(n)) $\subseteq \bigcup_{c>0} DTIME(n \cdot c^{f(n)})$

if $f(n) \ge log(n)$, then simply: DSPACE $(f(n)) \subseteq \bigcup_{c>0} DTIME(c^{f(n)})$

<u>Proof</u>: Take some $L \in DSPACE(f(n))$, recognized by M. M does not loop, so (the number of visited configurations) = (the number of steps) (the number of all configurations) \geq (the number of steps)

the number of all configuration equals:

$$|Q| \cdot (n+2) \cdot (4|\Gamma|)^{df(n)}$$

contents of working tapes + a special marker for:

- the position of the head
- the last visited cell on the tape

position on the input tape

$$\mathsf{DTIME}(f(n)) \subseteq \mathsf{DSPACE}(f(n)) \subseteq \bigcup_{c>0} \mathsf{DTIME}(n \cdot c^{f(n)})$$

In particular:

 $L \subseteq P \subseteq PSPACE \subseteq EXPTIME \subseteq EXPSPACE$

$$\mathsf{DTIME}(f(n)) \subseteq \mathsf{DSPACE}(f(n)) \subseteq \bigcup_{c>0} \mathsf{DTIME}(n \cdot c^{f(n)})$$

In particular:

$$L \subseteq P \subseteq PSPACE \subseteq EXPTIME \subseteq EXPSPACE$$

Are these classes different?

it is **NOT** known whether:

- L≠P
- P≠PSPACE
- PSPACE≠EXPTIME
- EXPTIME≠EXPSPACE

It is known (and we will prove this soon), that

- L≠PSPACE≠EXPSPACE
- P≠EXPTIME

Sipser's theorem

Theorem. Consider a machine M working in space S(n), but not necessarily having the halting property.

Then there exists a machine M' such that:

- L(M')=L(M)
- M' works in space S(n)
- M' halts on every input

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Thus: in the following definition

A language $L \subseteq \Sigma^*$ is *recognizable in space* S(n) if there exists a multitape machine that <u>halts on every input</u>, accepts L, and works in space S(n).

this condition was redundant

Sipser's theorem

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- *M'* works in space *S*(*n*)
- M' halts on every input

Proof

Approach 1: (in which the resulting M' uses a lot of space) Key observation: in an accepting run no configuration repeats.

- after every move we copy the current configuration to an additional working tape,
- additionally we check whether the current configuration equals to some configuration saved earlier
- a configuration has repeated ⇒ a loop ⇒ we reject