

# Computational complexity

lecture 8

## Ladner's theorem

Theorem (Ladner, 1975) – existence of NP-intermediate problems:  
If  $\mathbf{P} \neq \mathbf{NP}$ , then there is a problem, which is in  $\mathbf{NP} \setminus \mathbf{P}$ , but is not  $\mathbf{NP}$ -hard with respect to polynomial-time reductions (so even more with respect to logarithmic-space reductions).

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Proof:

Supposing that  $\text{SAT} \notin \mathbf{P}$  we will give a language  $L \in \mathbf{NP}$  such that:

- $L$  is not in  $\mathbf{P}$ , and
- $\text{SAT}$  does not reduce to  $L$  in polynomial time

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Supposing that  $\text{SAT} \notin \mathbf{P}$  we will give a language  $L \in \mathbf{NP}$  such that:

- $L$  is not in  $\mathbf{P}$ , and
- SAT does not reduce to  $L$  in polynomial time

We create  $L$  as a variant of SAT with an appropriate amount of padding. In general, with padding we can change a problem into a simpler one. We want to add enough padding so that the SAT problem stops to be  $\mathbf{NP}$ -complete, but not too much, so that still it is not in  $\mathbf{P}$ .

The definition will be:

$$L = \{w01^{f(|w|)} : w \in \text{SAT}\}$$

for an appropriate function  $f$

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We now define  $f$

- Fix a computable enumeration  $M_1, M_2, M_3, \dots$  of Turing machines, such that  $M_i$  works in time  $O(n^i)$ , and every language in **P** is recognized by some  $M_i$
- To this end, we take a list  $M'_1, M'_2, M'_3, \dots$  on which every Turing machine appears infinitely often. To  $M'_i$  we add a counter, which stops the machine after  $n^i$  steps – this results in  $M_i$

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The function  $f$  is defined by the following algorithm:

- (a) take  $i=1, n=1$
- (b) put  $f(n)=n^i$
- (c) if there is a word  $v$  of length  $\leq \log(n)$  such that  $M_i$  incorrectly recognizes whether  $v$  belongs to  $L$ , then increase  $i$  by 1
- (d) increase  $n$  by 1, go back to (b)

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Fact 1: It can be checked in polynomial time whether a word is of the proper form (i.e., if the number of ones is appropriate).

- In order to compute  $f(n)$  we repeat the loop  $n$  times, in every repetition we check polynomially many words  $v$  (of logarithmic length)
- On every word  $v$  we run  $M_i$ , which works in time  $O(\log^i n)$
- We can spend this time, as the input should have length  $\geq f(n) \geq n^i$  (we interrupt the loop as soon as there are not enough ones)
- Remark:  $i$  is not a constant (time  $O(\log^i n)$  by itself is not polynomial)
- Remark 2: the simulation time depends on  $|M_i|$ , but  $|M_i| = |i| = \log(i) \leq \log(n)$ , so this is OK

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- We can spend this time, as the input should have length  $\geq f(n) \geq n^i$  (we interrupt the loop as soon as there are not enough ones)
- We also need to check whether  $v \in L$  (where  $|v| \leq \log n$ )
  - we check the number of ones in  $v$  by the induction assumption
  - we check whether prefix  $\in \text{SAT}$  in time exponential in  $\log(n)$



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Corollary:  $L \in \mathbf{NP}$

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Fact 2: if  $\text{SAT} \notin \mathbf{P}$  then  $L \notin \mathbf{P}$

- If  $L \in \mathbf{P}$ , then some  $M_i$  recognizes  $L$ , so from some moment on (i.e. for  $n \geq n_0$  for some  $n_0$ ) we have that  $f(n)=n^i$
- Then it is easy to solve SAT in  $\mathbf{P}$  (a contradiction):
  - if  $|w| \geq n_0$  we append  $|w|^i$  ones at the end, and we start  $M_i$
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Corollary: Because  $L \notin \mathbf{P}$ , the function  $f$  grows faster than every polynomial

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Fact 3: if  $\text{SAT} \notin \mathbf{P}$  then  $L$  is not **NP**-hard

- Suppose that SAT reduces to  $L$  through a function  $g$  computable in time  $n^k$ . We will show a polynomial algorithm for SAT.

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- We know that there is  $n_0$  such that for  $n \geq n_0$  it holds that  $f(n) > n^k$
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- For formulas  $w$  shorter than  $n_0$  the results can be hardcoded
- For  $|w| \geq n_0$  we consider the word  $g(w)$ ; it has length  $\leq |w|^k$ .

If  $g(w)$  is not of the form  $w'01^{f(|w'|)}$ , then it is not in  $L$ , we reject (by fact 1, this can be checked in  $\mathbf{P}$ ). Otherwise  $w \in \text{SAT} \Leftrightarrow w' \in \text{SAT}$

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Moreover, either  $|w'| < n_0$ , or we have that  $|w|^k \geq |g(w)| > f(|w'|) > |w'|^k$ , thus the new formula is shorter at least by 1.

- We repeat this in a loop; after a linear number of steps the input length decreases below  $n_0$ , and we obtain a result.

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We have thus proved:

Theorem (Ladner 1975)

If  $\mathbf{P} \neq \mathbf{NP}$ , then there is a problem, which is in  $\mathbf{NP} \setminus \mathbf{P}$ , but is not  $\mathbf{NP}$ -hard with respect to polynomial-time reductions (so even more with respect to logarithmic-space reductions).



# CSP problems and the dichotomy theorem

The CSP problem

Input: variables  $x_1, \dots, x_n$ , domains  $D_1, \dots, D_n$ , constraints  $C_1, \dots, C_m$  of the form  $(t, R)$ , where  $t$  is a tuple of  $k$  variables, and  $R$  is a  $k$ -ary relation

Question: are there  $x_1 \in D_1, \dots, x_n \in D_n$  satisfying  $C_1, \dots, C_m$ ?

(a constraint  $(t, R)$  is satisfied if the tuple of variables  $t$  belong to the relation  $R$ )

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Most natural **NP**-complete problems can be easily reduced to CSP (written as CSP).

E.g. 3-coloring:

- $x_1, \dots, x_n$  – represent colors of nodes  $1, \dots, n$
- $D_1, \dots, D_n = \{1, 2, 3\}$
- for every edge  $k, l$  we have a constraint  $x_k \neq x_l$   
(i.e.,  $R$  is the binary relation  $\{(1, 2), (2, 1), (1, 3), (3, 1), (2, 3), (3, 2)\}$ )

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Most natural **NP**-complete problems can be easily reduced to CSP (written as CSP).

Problem  $\text{CSP}(\Gamma)$  – like CSP, but only relations from a set  $\Gamma$  can be used

Theorem (2017): for every set  $\Gamma$  we either have  $\text{CSP}(\Gamma) \in \mathbf{P}$ , or  $\text{CSP}(\Gamma)$  is **NP**-complete

## Berman's theorem

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Intuitively, **NP**-hard means hardest in **NP**, or even harder (so problems harder than **NP** should be **NP**-hard).

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But the definition is:  $L$  is **NP**-hard if we can reduce every problem from **NP** to  $L$ .

So: can we reduce every problem from **NP**, to every (more difficult) problem not in **NP**?

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So: can we reduce every problem from **NP**, to every (more difficult) problem not in **NP**?

The answer is **no** – we have the following theorem:

Theorem (Berman 1978)

If  $P \neq NP$ , then no language over a single-letter alphabet is **NP**-hard wrt. polynomial-time reductions (so even more wrt. logarithmic-space reductions).

## Berman's theorem

Is it the case that every problem not in **NP** is **NP**-hard?

**No** – we have the following theorem:

Theorem (Berman 1978)

If **P**≠**NP**, then no language over a single-letter alphabet is **NP**-hard.

Notice that there is a language over a single-letter alphabet that requires doubly-exponential running time (i.e., surely is not in **NP**): take any language  $L$  over  $\{0,1\}$  requiring triple-exponential running time, and take  $\{1^{|1w|_2} : w \in L\}$ , where  $|1w|_2$  is the number encoded in binary as  $1w$ .

There is also an undecidable language over a single-letter alphabet:  $\{1^k : M_k \text{ halts on empty input}\}$

These languages are not **NP**-hard, and not in **NP** (assuming **P**≠**NP**).

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Theorem (Berman 1978)

If  $\mathbf{P} \neq \mathbf{NP}$ , then no language over a single-letter alphabet is  $\mathbf{NP}$ -hard.

Proof

Let  $L$  be an  $\mathbf{NP}$ -hard language over a single-letter alphabet. We will give a polynomial-time algorithm for SAT, contradicting  $\mathbf{P} \neq \mathbf{NP}$ .



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By assumption there is a reduction  $g$  from SAT to  $L$ .

The algorithm is as follows:

- We are given a formula  $\phi$
- We will keep a list of formulas  $\psi_1, \dots, \psi_k$  such that:  $\phi$  is satisfiable iff some of  $\psi_1, \dots, \psi_k$  is satisfiable. Initially the list contains  $\phi$ .

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- We alternately repeat two kinds of steps:
  - 1) Replace every  $\psi_i$  by two formulas:  $\psi_i[true/x]$  and  $\psi_i[false/x]$ , obtained by substituting true/false for one of variables.  
(clearly  $\psi_i$  is satisfiable iff some of  $\psi_i[true/x]$ ,  $\psi_i[false/x]$  is satisfiable)

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The algorithm is correct. Why does it work in polynomial time?

- Recall that  $g$  is a polynomial-time reduction to a single-letter language. Thus  $|g(\psi_i)| \leq p(|\psi_i|)$  for some polynomial  $p$ .  
Since there is only one single-letter word of every length, there are only  $p(|\psi_i|) \leq p(|\phi|)$  possibilities for  $g(\psi_i)$ .
- In effect, the list has length  $\leq p(|\phi|)$  after every execution of step 2, and  $\leq 2 \cdot p(|\phi|)$  after every execution of step 1.
- Moreover, every step can be performed in polynomial time.

This finishes the proof.

## Relativisation

Many proofs in the complexity theory uses Turing machines as “black-boxes” – the proofs are of the form:

- assume that there is a machine  $M$  working in time ... recognizing ...
- Out of it, we create  $M'$ , which executes  $M$  many times in a loop...
- ... then it negates the results, executes itself on every machine ...
- at the end we obtain a machine  $M''''''$ , about which we know that it cannot exist, thus  $M$  could not exist.

Such proofs relativize, i.e., they work also when every machine in the world has access to some fixed oracle (that is, it can ask whether a word belongs to a language  $L$ , and immediately obtain an answer)

# Relativisation

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- ...

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Examples of relativizing proofs: Turing theorem about undecidability, hierarchy theorems, gap theorems, Ladner's theorem, Immerman-Szelepcseny theorem, Savitch theorem, ...

On the other hand, proofs based on circuits do not relativize (it is not at all clear what is an oracle for a circuit)

The next theorem shows that using relativizing arguments we cannot solve the **P** vs. **NP** problem.

# Baker-Gill-Solovay theorem

Theorem (Baker-Gill-Solovay, 1975)

There exist languages  $A$  and  $B$  such that  $\mathbf{P}^A = \mathbf{NP}^A$  and  $\mathbf{P}^B \neq \mathbf{NP}^B$

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Proof

As  $A$  we can take QBF – we have:

$$\mathbf{NP}^{\text{QBF}} \subseteq \mathbf{NPSPACE} = \mathbf{PSPACE} = \mathbf{P}^{\text{QBF}}$$

Steps from the left:

- instead of asking the QBF oracle about a word, a machine can itself compute the answer (questions are of polynomial length, and QBF can be solved in polynomial space)
- Savitch theorem
- **PSPACE**-completeness of the QBF problem



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- Savitch theorem
- **PSPACE**-completeness of the QBF problem

Does  $A = \text{SAT}$  work as well? –  $\mathbf{NP}^{\text{SAT}} \subseteq \mathbf{NP} \subseteq \mathbf{P}^{\text{SAT}}$

# Baker-Gill-Solovay theorem

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There exist languages  $A$  and  $B$  such that  $\mathbf{P}^A = \mathbf{NP}^A$  and  $\mathbf{P}^B \neq \mathbf{NP}^B$

Proof

As  $A$  we can take QBF – we have:

$$\mathbf{NP}^{\text{QBF}} \subseteq \mathbf{NPSPACE} = \mathbf{PSPACE} = \mathbf{P}^{\text{QBF}}$$

Steps from the left:

- instead of asking the QBF oracle about a word, a machine can itself compute the answer (questions are of polynomial length, and QBF can be solved in polynomial space)
- Savitch theorem
- **PSPACE**-completeness of the QBF problem

Does  $A = \text{SAT}$  work as well? –  ~~$\mathbf{NP}^{\text{SAT}} \subseteq \mathbf{NP} \subseteq \mathbf{P}^{\text{SAT}}$~~

NO – an **NP** algorithm for SAT doesn't give the inclusion  $\mathbf{NP}^{\text{SAT}} \subseteq \mathbf{NP}$  (maybe the external algorithm „prefers” to obtain that a formula is not satisfiable, and it will incorrectly compute its satisfiability)

It is important that QBF can be solved in deterministic **PSPACE**

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Proof

We now construct an oracle  $B$ , and we consider the language

$L = \{1^n : \text{some word } w \text{ of length } n \text{ belongs to } B\}$

- Clearly  $L \in \mathbf{NP}^B$  – nondeterministic machine can guess some  $w \in B$
- A deterministic machine recognizing  $L$  has a problem: it can only ask the oracle for consecutive words, but it has not enough time to check all of them. We only need to choose  $B$  so that indeed it is impossible to do anything better.

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We now choose  $B$ :

- Fix a list  $M_1, M_2, M_3, \dots$  of all Turing machines with oracle working in polynomial time
  - an oracle is not a part of the definition of the machine,
  - for every  $M_i$  there should exist a polynomial  $p_i$  such that for every oracle the machine  $M_i$  works in time  $p_i(n)$
  - if some  $M$  with oracle  $C$  recognizes a language  $L$  in polynomial time, then some  $M_i$  with oracle  $C$  also recognizes  $L$
  - such a list  $M_1, M_2, M_3, \dots$  is created as in the proof of Ladner's theo.
  - this time, we do not use the fact that the list is computable (conversely to the proof of the Ladner's theorem)
- We construct  $B$  gradually, cheating consecutive machines

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We create  $B = \bigcup_{i \in \mathbb{N}} B_i$  and a sequence  $n_i$  such that:

- $M_i^{B_i}$  incorrectly recognizes the word  $1^{n_i}$
- $M_i^B$  agrees with  $M_i^{B_i}$  on the word  $1^{n_i}$

We start with  $B_0 = \emptyset$ ; then for consecutive  $i$ :

- we take  $n_i$  so large that for all  $j < i$ , machine  $M_j$  on the word  $1^{n_j}$  produces only queries shorter than  $n_i$  (thanks to this the machines that were cheated earlier remain cheated), and such that  $M_i$  on the word  $1^{n_i}$  works in less than  $2^{n_i}$  steps

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  - run  $M_i^{B_{i-1}}$  on the word  $1^{n_i}$
  - if it accepts, take  $B_i = B_{i-1}$  – then  $1^{n_i} \notin L$ , we have cheated  $M_i$
  - if it rejects, find a word  $w$  of length  $n_i$  about which  $M_i$  haven't asked (it exists, since  $M_i$  has made  $< 2^{n_i}$  step) and define  $B_i = B_{i-1} \cup \{w\}$
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- The language  $B$  is computable, but in this theorem this is meaningless

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## Search problems

The **NP** class was defined for decision problems („yes/no”),  
e.g., does there exist a valuation satisfying a formula,  
does there exist a Hamiltonian cycle, ...

We can also consider search problems,  
e.g., find a valuation satisfying a formula,  
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- Of course search problems are not easier than decision problems. Thus if **P**≠**NP**, then search problems cannot be solved in polynomial time as well.
- And what if **P**=**NP**? Maybe it is possible to decide quickly whether there is a Hamiltonian cycle, but it is impossible to quickly find it?



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- And what if **P**=**NP**? ~~Maybe it is possible to decide quickly whether there is a Hamiltonian cycle, but it is impossible to quickly find it?~~
- Then it possible to solve also search problems in polynomial time.

# Search problems

## Theorem

If **P=NP**, then for every language  $L \in \mathbf{NP}$  there is a polynomial algorithm that reads  $v \in L$  and finds a witness for  $v$ .

We refer here to the definition of **NP** using witnesses:

**NP** contains languages of the form  $\{v : \exists w. v\$w \in R\}$ , where  $R$  is a relation recognizable in polynomial time and such that  $v\$w \in R$  implies  $|w| \leq p(|v|)$  for some polynomial  $p$ .

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## Proof

Consider first the SAT problem – we assume that there is a polynomial-time algorithm  $A$  for SAT, we want to find a valuation:

- Using  $A$  we check whether the formula is satisfiable
- If yes, we set  $x_1 = 1$  and we check whether it is still satisfiable
- Yes  $\Rightarrow$  keep  $x_1 = 1$  and continue for a smaller formula
- No  $\Rightarrow$  set  $x_1 = 0$  and continue for a smaller formula
- In this way we eliminate consecutive variables, and we obtain a whole valuation

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## Proof

- For SAT we already know, consider now an arbitrary problem from **NP**
- It is enough to see that the reduction from the Cook-Levin theorem (**NP**-hardness of SAT) is actually a Levin reduction (i.e., it allows to recover witnesses)