# Advanced Computational Complexity

### George Voutsadakis<sup>1</sup>

<sup>1</sup>Mathematics and Computer Science Lake Superior State University

LSSU Math 600



#### Diagonalization

- Time Hierarchy Theorem
- Nondeterministic Time Hierarchy Theorem
- Ladner's Theorem: Existence of NP-Intermediate Problems
- Oracle Machines and the Limits of Diagonalization

# Introducing Diagonalization

- We would like to be able to prove that certain complexity classes (e.g., P and NP) are not the same.
- This requires exhibiting a machine in one class that differs from every machine in the other class.
- Being different means that their answers are different on at least one input.
- The only general technique known for constructing such a machine is **diagonalization**.

# Introducing Diagonalization (Cont'd)

- The one common tool used in all diagonalization proofs is the representation of TMs by strings.
- It is effective, in the sense that there is a universal TM that, given any string x, can simulate the machine represented by x with a small - at most logarithmic - overhead.
- Every string  $x \in \{0,1\}^*$  represents some TM, denoted  $M_x$ .
- Every TM is represented by infinitely many strings.
- We use the notation  $M_i$ , where  $i \in \mathbb{N}$ , for the machine represented by the string that is the binary expansion of the number i, without the leading 1.

### Subsection 1

Time Hierarchy Theorem

## The Time Hierarchy Theorem

- The Time Hierarchy Theorem shows that allowing Turing machines more computation time strictly increases the set of languages that they can decide.
- Recall that for a function f : N → N, DTIME(f(n)) is the set of languages decided by a TM running in time O(f(n)).
- We only deal with time-constructible functions f, which means that the mapping x → f(|x|) can be computed in O(f(n)) time.

#### Theorem (Time Hierarchy Theorem)

If f, g are time-constructible functions, satisfying  $f(n) \log f(n) = o(g(n))$ , then

$$\mathsf{DTIME}(f(n)) \subsetneq \mathsf{DTIME}(g(n)).$$

# Proof of the Time Hierarchy Theorem

- We prove the simpler statement DTIME(n) ⊊ DTIME(n<sup>1.5</sup>).
  Consider the following Turing machine D.
  - Suppose it receives input *x*.
  - It uses the Universal TM  $\mathcal{U}$  to simulate the execution of  $M_x$  on x for  $|x|^{1.4}$  steps.
  - Suppose  $\mathcal{U}$  outputs some bit  $b \in \{0,1\}$  in this time.
  - Then D outputs the opposite answer, i.e., 1 b.
  - Otherwise, it outputs 0.
  - D halts within  $n^{1.4}$  steps.
  - So, the language L decided by D is in  $DTIME(n^{1.5})$ .

# Proof of the Time Hierarchy Theorem (Cont'd)

### • Claim: $L \notin \text{DTIME}(n)$ .

Assume there is some TM M and constant c, such that TM M, given any input  $x \in \{0, 1\}^*$ , halts within c|x| steps and outputs D(x).

The time to simulate M by the universal Turing machine  ${\mathcal U}$  on every input x is at most

 $c'c|x|\log|x|,$ 

where c' is a constant that:

- Depends on the alphabet size and number of tapes and states of *M*;
- Is independent of |x|.

# Proof of the Time Hierarchy Theorem (Cont'd)

• There is some number *n*<sub>0</sub>, such that

$$n^{1.4} > c' cn \log n$$
, for every  $n \ge n_0$ .

Let x be a string representing the machine M, with  $|x| \ge n_0$ . Such a string exists since M is represented by infinitely many strings. Then, D(x) will obtain the output b = M(x) within  $|x|^{1.4}$  steps. However, by definition of D, we have

$$D(x) = 1 - b \neq M(x).$$

This yields a contradiction.

• The proof for general f, g is similar and uses the observation that the slowdown in simulating a machine using U is at most logarithmic.

### Subsection 2

#### Nondeterministic Time Hierarchy Theorem

# The Nondeterministic Time Hierarchy Theorem

Theorem (Nondeterministic Time Hierarchy Theorem)

If f, g are time constructible functions, satisfying f(n+1) = o(g(n)), then

 $\mathsf{NTIME}(f(n)) \subsetneq \mathsf{NTIME}(g(n)).$ 

• We only prove  $NTIME(n) \subsetneq NTIME(n^{1.5})$ .

The first instinct is to mimic the deterministic case, since there is a universal TM for nondeterministic computation as well.

This is not sufficient because the definition of the new machine D requires the ability to "flip the answer".

That is, to efficiently compute, given the description of a NDTM M and an input x, the value 1 - M(x).

However, as we have seen, the complement of an NTIME(n) language is not expected to be in  $NTIME(n^{1.5})$ .

# Exploiting Exponential Time

- On the other hand, the complement of every NTIME(*n*) language is trivially decidable in exponential time.
  - This is achieved by examining all the possibilities for the machine's nondeterministic choices.
  - This trivial exponential simulation of a nondeterministic machine does suffice to establish a hierarchy theorem.

## Lazy Diagonalization

- The key idea is the so-called lazy diagonalization.
  - The machine *D* flips the answer of each linear time NDTM  $M_i$  in only one string out of a sufficiently (exponentially) large set of strings. Define the function  $f : \mathbb{N} \to \mathbb{N}$  by

$$f(1) = 2;$$
  
 
$$f(i+1) = 2^{f(i)^{1.2}}.$$

Given *n*, we can find in O  $(n^{1.5})$  time the number *i*, such that *n* is sandwiched between f(i) and f(i+1).

The machine *D* will try to flip the answer of  $M_i$  on some input in the set  $\{1^n : f(i) < n \le f(i+1)\}$ .

# Lazy Diagonalization (Cont'd)

• The machine D operates as follows.

- Suppose it receives input *x*.
- If  $x \notin 1^*$ , reject.
- If  $x = 1^n$ , then compute *i*, such that  $f(i) < n \le f(i+1)$ .
- 1. If f(i) < n < f(i+1), then simulate  $M_i$  on input  $1^{n+1}$ non-deterministically in  $n^{1,1}$  time and output its answer. If  $M_i$  has not halted in this time, then halt and accept.
- 2. If n = f(i+1), accept  $1^n$  iff  $M_i$  rejects  $1^{f(i)+1}$  in  $(f(i)+1)^{1.1}$  time.

Part 2 requires going through all possible  $2^{(f(i)+1)^{1.1}}$  branches of  $M_i$  on input  $1^{f(i)+1}$ .

That is fine, since the input size f(i + 1) is  $2^{f(i)^{1.2}}$ . Hence, the NDTM *D* runs in O  $(n^{1.5})$  time.

## Correctness

• Let L be the language decided by D.

```
Claim: L \notin \text{NTIME}(n).
```

Indeed, suppose that L is decided by an NDTM M running in cn steps, for some constant c.

Since each NDTM is represented by infinitely many strings, we can find i large enough, such that:

- $M = M_i$ ;
- On inputs of length n ≥ f(i), M<sub>i</sub> can be simulated in less than n<sup>1.1</sup> steps.

This means that the two steps in the description of D ensure that:

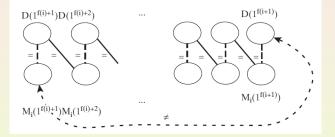
• If 
$$f(i) < n < f(i+1)$$
, then  $D(1^n) = M_i(1^{n+1})$ ;  
•  $D(1^{f(i+1)}) \neq M_i(1^{f(i)+1})$ .

# Finishing Correctness

• We obtained the conditions

• If 
$$f(i) < n < f(i+1)$$
, then  $D(1^n) = M_i(1^{n+1})$ ;

• 
$$D(1^{r(i+1)}) \neq M_i(1^{r(i)+1}).$$



By our assumption,  $M_i$  and D agree on all inputs  $1^n$  for n in the semi-open interval (f(i), f(i+1)]. By the first statement above,  $D(1^{f(i+1)}) = M_i(1^{f(i)+1})$ .

This contradicts the second statement.

George Voutsadakis (LSSU)

### Subsection 3

#### Ladner's Theorem: Existence of NP-Intermediate Problems

### Introducing Ladner's Theorem

- Is every problem in NP either in P or NP-complete?
  - If P = NP, then the conjecture is trivially true.
  - We show that if P ≠ NP, then this conjecture is false
    I.e., there is a language L ∈ NP\P that is not NP-complete.
- A feature of the proof is an interesting definition à la Gödel of a language SAT<sub>H</sub> which "encodes" the difficulty of solving itself.

## Ladner's Theorem

Ladner's Theorem ("NP-Intermediate" Languages)

Suppose that  $P \neq NP$ . Then there exists a language

 $L\in\mathsf{NP}\backslash\mathsf{P}$ 

that is not NP-complete.

• Consider a function  $H : \mathbb{N} \to \mathbb{N}$ .

Define the language  $SAT_H$  to contain all length-*n* satisfiable formulae that are padded with  $n^{H(n)}$  1's:

$$SAT_{H} = \{\psi 01^{n^{H(n)}} : \psi \in SAT \text{ and } n = |\psi|\}.$$

## The Function *H*

• Based on SAT<sub>H</sub>, define a function  $H : \mathbb{N} \to \mathbb{N}$ . H(n) is the smallest number  $i < \log \log n$ , such that, for every  $x \in \{0, 1\}^*$ , with  $|x| \le \log n$ ,

 $M_i$  outputs  $\text{SAT}_H(x)$  within  $i|x|^i$  steps.

If no such number *i* exists, then  $H(n) = \log \log n$ .

H(n) determines membership in SAT<sub>H</sub> of strings whose length is greater than n.

Moreover, the definition of H(n) only relies upon checking the status of strings of length at most log n.

So *H* is well defined.

# Complexity of H

#### Claim

The function H is computable in time O  $(n^3)$ .

- To compute H(n) we must:
  - (1) Compute H(i) on every  $i \leq \log n$ ;
  - Simulate at most log log n machines on inputs of lengths at most log n for less than log log n(log n)<sup>log log n</sup> = o (n) steps;
  - (3) Compute SAT on inputs of size at most log n.

H(n) can be computed in time  $T(n) \le \log nT(\log n) + O(n^2)$ . Therefore,  $T(n) = O(n^3)$ .

# Relations Between $SAT_H$ and H

#### Claim

 $SAT_H \in P$  if and only if H(n) = O(1). If  $SAT_H \notin P$ , then

 $H(n) \stackrel{n \to \infty}{\longrightarrow} \infty.$ 

We first show that, if SAT<sub>H</sub> ∈ P, then H(n) = O(1). Suppose there is a machine M solving SAT<sub>H</sub> in at most cn<sup>c</sup> steps. Recall that M is represented by infinitely many strings. So there is a number i > c, such that M = M<sub>i</sub>. The definition of H(n) implies that, for n > 2<sup>2<sup>i</sup></sup>, H(n) ≤ i. Thus, H(n) = O(1).

## Relations Between $SAT_H$ and H (Cont'd)

- We now show that, if H(n) = O(1), then  $SAT_H \in P$ .
  - If H(n) = O(1), then H can take only one of finitely many values. Hence, there exists an i, such that H(n) = i, for infinitely many n's. This implies that the TM  $M_i$  solves  $\text{SAT}_H$  in  $in^i$ -time.

Otherwise, by definition, there is an input x on which  $M_i$  fails to output the right answer within this bound.

Then, for all  $n > 2^{|x|}$ , we have  $H(n) \neq i$ .

Note that the last statement holds even if we only assume that there is some constant C, such that  $H(n) \leq C$ , for infinitely many n's.

This proves the second statement in the claim.

# NP-Completeness of $SAT_H$ Implies Polynomiality of SAT

#### Claim

Suppose  $H: \mathbb{N} \to \mathbb{N}$  is a polynomially computable function, such that

$$\lim_{n\to\infty}H(n)=\infty.$$

If  $SAT_H$  is NP-complete, then SAT is in P.

- Let f be a reduction from SAT to SAT<sub>H</sub> that runs in time O (n<sup>i</sup>).
  Let N be such that H(n) > i, for n > N.
  The following recursive algorithm A solves SAT in polynomial time.
  - Suppose A receives as input formula  $\varphi$ .
  - If  $|\varphi| \leq N$ , then compute the output using brute force.
  - Otherwise compute  $x = f(\varphi)$ .
    - If x is not of the form  $\psi 01^{n^{H(|\psi|)}}$ , then output FALSE.
    - Otherwise, output  $A(\psi)$ .

## Proof of Ladner's Theorem

Ladner's Theorem ("NP-Intermediate" Languages)

```
If P \neq NP, there exists a language
```

 $L \in \mathsf{NP} \backslash \mathsf{P}$ 

that is not NP-complete.

• Suppose that  $SAT_H \in P$ .

By the claim,  $H(n) \leq C$ , for a constant C.

Thus,  $SAT_H$  is SAT padded with at most  $n^C$  1's.

But then a polynomial-time algorithm for  $SAT_H$  can be used to solve SAT in polynomial time.

```
Since SAT is NP-complete, P = NP.
```

# Proof of Ladner's Theorem (Cont'd)

### • Suppose that SAT<sub>H</sub> is NP-complete.

Then, there is a reduction f from SAT to SAT<sub>H</sub> that runs in time O  $(n^i)$  for some constant i.

Since  $\operatorname{SAT}_H \notin \mathsf{P}$ , by the claim,  $H(n) \to \infty$ .

The reduction works in time  $O(n^i)$ , for large *n*.

So it must map SAT instances of size n to SAT<sub>H</sub> instances of size smaller than  $n^{H(n)}$ .

Thus, for large enough  $\varphi$ , f must map  $\varphi$  to a string  $\psi 01^{H(|\psi|)}$ , where  $\psi$  is smaller by some fixed polynomial factor, say, smaller than  $\sqrt[3]{n}$ . The last claim yields a polynomial-time recursive algorithm A for SAT. This contradicts  $P \neq NP$ .

### Remarks on Ladner's Theorem

- The theorem shows the existence of some non-NP-complete language in NP\P if NP  $\neq$  P.
- The language L seems somewhat artificial.
- Moreover, the proof has not been strengthened to yield a more natural language.
- In fact, the status of most natural languages has been resolved thanks to clever algorithms or reductions.
- Two interesting exceptions are Factoring and Graph Isomorphism.
- For these two languages:
  - No polynomial-time algorithm is currently known;
  - There is strong evidence that they are not NP-complete.

### Subsection 4

### Oracle Machines and the Limits of Diagonalization

# The Essence of Diagonalization

- We would like to qualify the limits of diagonalization.
- Towards this goal, we call "diagonalization" any technique that relies solely upon the following properties of Turing machines:
  - The existence of an effective representation of Turing machines by strings;
  - I The ability of one TM to simulate any other without much overhead in running time or space.
- Any argument that only uses these facts is treating machines as black boxes, in the sense that the machine's internal workings do not matter.

## Oracle Turing Machine and the P vs. NP

- We use a technique involving variants of Turing machines that still satisfy Properties I and II.
- A general way to define such variants of Turing machines leads to oracle Turing machines.
- We do so in two different ways.
  - One way of defining the variant results in TMs for which P = NP;
  - Another way results in TMs for which  $P \neq NP$ .
- This discrepancy shows that assuming only Properties I and II is not sufficient in order to resolve P versus NP.
- Some additional property must be used.

# Oracle Turing Machines Informally

- Oracle machines are TMs that are given access to a black box or "oracle" that can magically solve the decision problem for some language  $O \in \{0, 1\}^*$ .
  - The machine has a special oracle tape on which it can write a string  $q \in \{0,1\}^*.$
  - In one step it gets an answer to a query of the form

#### "Is q in O?"

- This can be repeated arbitrarily often with different queries.
- If *O* is a difficult language, e.g., one that cannot be decided in polynomial time or that is not even decidable, then the oracle gives an added power to the TM.

# Oracle Turing Machines

#### Definition (Oracle Turing Machines)

An **oracle Turing machine** is a TM M that has a special read-write tape, called M's **oracle tape**, and three special states  $q_{query}$ ,  $q_{yes}$ ,  $q_{no}$ . To execute M, we specify, in addition to the input, a language  $O \in \{0, 1\}^*$  that is used as the **oracle** for M.

- Whenever, during the execution, M enters the state  $q_{query}$ , with q being the contents of its oracle tape, the machine, then, moves into the state  $q_{yes}$ , if  $q \in O$ , and  $q_{no}$ , if  $q \notin O$ .
- Regardless of the choice of *O*, a membership query to *O* counts only as a single computational step.

If *M* is an oracle machine,  $O \in \{0,1\}^*$  a language and  $x \in \{0,1\}^*$ , then we denote the output of *M* on input *x* and with oracle *O* by  $M^O(x)$ . **Nondeterministic oracle TM**s are defined similarly.

# The Classes $\mathsf{P}^O$ and $\mathsf{N}\mathsf{P}^O$

### Definition (The Classes $P^O$ and $NP^O$ )

Let  $O \in \{0,1\}^*$ .

- P<sup>O</sup> is the set containing every language that can be decided by a polynomial-time deterministic TMs with oracle access to O.
- NP<sup>O</sup> is the set of all languages that can be decided by a polynomial-time nondeterministic TM with oracle access to O.
- Example: Let  $\overline{SAT}$  denote the language of unsatisfiable formulae. Then  $\overline{SAT} \in P^{SAT}$ .

Suppose, we are given oracle access to  $\ensuremath{\operatorname{SAT}}.$ 

To decide whether a formula  $\varphi$  is in  $\overline{\mathrm{SAT}},$  a polynomial-time oracle TM:

- Asks its oracle if  $\varphi \in SAT$ ;
- Then outputs the opposite answer.

## Example

• Let  $O \in P$ . Then

$$\mathsf{P}^{O} = \mathsf{P}.$$

Allowing an oracle can only help compute more languages. So  $P \subseteq P^{O}$ .

If  $O \in P$ , then it is redundant as an oracle.

Let  $M^O$  be any polynomial-time oracle TM using O.

We can transform  $M^O$  into a standard TM (no oracle) M by simply replacing each oracle call with the computation of O. Thus,  $P^O \subseteq P$ .

## Example

#### Let

EXPCOM = { $\langle M, x, 1^n \rangle$  : M outputs 1 on x within  $2^n$  steps}.

Then  $P^{\text{ExpCom}} = NP^{\text{ExpCom}} = EXP$ .

Clearly, an oracle to  ${\rm ExpCOM}$  allows one to perform an exponential-time computation at the cost of one call.

So 
$$\mathsf{EXP} \subseteq \mathsf{P}^{\mathsf{ExpCom}}$$

Conversely, let  $M^O$  be a nondeterministic polynomial-time oracle TM. We can simulate its execution with an EXPCOM oracle in exp time.

Such time suffices both to:

- Enumerate all of M's nondeterministic choices;
- Answer the EXPCOM oracle queries.

Thus,  $\mathsf{EXP} \subseteq \mathsf{P}^{\mathrm{ExpCom}} \subseteq \mathsf{NP}^{\mathrm{ExpCom}} \subseteq \mathsf{EXP}$ .

# **Relativizing Results**

- Regardless of what the oracle *O* is, the set of all TM's with access to *O* satisfy Properties I and II:
  - We can represent TMs with oracle O as strings;
  - We can use this representation to simulate such TMs using a universal TM, which, itself also, has access to *O*.
- It follows that any result about TMs or complexity classes that uses only I and II also holds for the set of all TMs with oracle *O*.
- Such results are called relativizing results.
- The Baker, Gill, Solovay Theorem, which follows, implies that neither P = NP nor  $P \neq NP$  can be a relativizing result.

# The Baker, Gill, Solovay Theorem

#### Theorem (Baker, Gill, Solovay)

There exist oracles A, B, such that  $P^A = NP^A$  and  $P^B \neq NP^B$ .

• First, we take A to be the language EXPCOM. Then, by the preceding example,

$$\mathsf{P}^{\mathsf{A}} = \mathsf{N}\mathsf{P}^{\mathsf{A}} = \mathsf{E}\mathsf{X}\mathsf{P}.$$

We turn to the second construction. For any language B, let  $U_B$  be the unary language

 $U_B = \{1^n : \text{some string of length } n \text{ is in } B\}.$ 

For every oracle B, the language  $U_B$  is in NP<sup>B</sup>.

A nondeterministic TM can make a nondeterministic guess for the string  $x \in \{0, 1\}^n$ , such that  $x \in B$ .

# The Baker, Gill, Solovay Theorem (Cont'd)

- We showed that, for every oracle B, U<sub>B</sub> ∈ NP<sup>B</sup>.
  We now construct an oracle B, such that U<sub>B</sub> ∉ P<sup>B</sup>.
  Combining, we get that P<sup>B</sup> ≠ NP<sup>B</sup>.
  For every i we let M: be the oracle TM represented by
  - For every i, we let  $M_i$  be the oracle TM represented by the binary expansion of i.
  - We construct B in stages.
  - Stage *i* ensures that  $M_i^B$  does not decide  $U_B$  in  $\frac{2^n}{10}$  time.
  - Initially, we let B be empty and gradually add strings to it.
  - Each subsequent stage determines the status of a finite number of strings (i.e., whether or not these strings will ultimately be in B).

## Construction of the Oracle B

• Stage *i*: So far, we have declared whether or not a finite number of strings are in *B*.

Choose n large enough so that it exceeds the length of any such string.

Run  $M_i$  on input  $1^n$  for  $\frac{2^n}{10}$  steps.

- Whenever *M<sub>i</sub>* queries the oracle about strings whose status has been determined, we answer consistently.
- When *M<sub>i</sub>* queries strings whose status is undetermined, we declare that the string is not in *B*.

After letting  $M_i$  finish computing on  $1^n$ , we now wish to ensure that the answer of  $M_i$  on  $1^n$  (whatever it was) is incorrect.

# Construction of the Oracle *B* (Cont'd)

 Note that, during this computation, we have only decided the fate of at most <sup>2<sup>n</sup></sup>/<sub>10</sub> strings in {0,1}<sup>n</sup>.

All of them were decided to be not in B.

- Suppose M<sub>i</sub> accepts 1<sup>n</sup>. Then we declare that all remaining strings of length n are also not in B. This ensures 1<sup>n</sup> ∉ U<sub>B</sub>.
- Suppose M<sub>i</sub> rejects 1<sup>n</sup>.
  We pick any string x of length n that M<sub>i</sub> has not queried.
  Such string exists because M<sub>i</sub> made at most <sup>2<sup>n</sup></sup>/<sub>10</sub> queries.
  We declare that x is in B.
  This ensures that 1<sup>n</sup> ∈ U<sub>B</sub>.

In either case, the answer of  $M_i$  is incorrect.

# The Baker, Gill, Solovay Theorem (Conclusion)

- Every polynomial p(n) is smaller than  $\frac{2^n}{10}$ , for large enough n.
- Every TM *M* is represented by infinitely many strings.
- So the construction ensures that M does not decide  $U_B$ .
- Thus,  $U_B \notin \mathbb{P}^B$ .

## Remarks on Diagonalization

- Is it possible that diagonalization or some other simulation method might help resolve the P vs. NP?
- If so, it has to use some fact about TMs that does not hold in the presence of oracles (i.e., a nonrelativizing fact).
- Even though many results in complexity relativize, there are some notable exceptions, but we still do not know how to use these nonrelativizing techniques to answer this fundamental question.

## **Remarks on Oracles**

- In some settings in complexity an oracle provided to a TM is not merely a language (i.e., Boolean function) but a general function f : {0,1}\* → {0,1}\*.
- An oracle TM is a useful abstraction of an algorithm.
- It uses another function as a black-box subroutine, without caring about the specifics of its implementation.