# Notes for Lecture 7

In this lecture we introduce the polynomial hierarchy and prove the Karp-Lipton-Sipser theorem.

## 1   Alternating Quantifiers

One way to look at the difference between **NP** and **coNP** is that a decision problem in **NP** is asking a of "does there exist" question, where the existence of the answer can by definition be efficiently proved. On the other hand, **coNP** asks "is it true for all" questions, which do not seem to have simple, efficient proofs.

Formally, a decision problem $A$ is in **NP** if and only if there is a polynomial time procedure $V(\cdot, \cdot)$ and a polynomial time bound $p()$ such that

$$x \in A \text{ if and only if } \exists y.|y| \leq p(|x|) \wedge V(x,y) = 1$$

and a problem $A$ is in **coNP** if and only if there is a polynomial time procedure $V(\cdot, \cdot)$ and a polynomial bound $p()$ such that

$$x \in A \text{ if and only if } \forall y : |y| \leq p(|x|), V(x,y) = 1$$

Now suppose you had a decision problem $A$ defined in the following form:

$$x \in A \Leftrightarrow \exists \, y_1 \text{ s.t. } |y_1| \leq p(|x|) \; \forall \, y_2 \text{ s.t. } |y_2| \leq p(|x|) \; V(x, y_1, y_2)$$

(where $p()$ is a polynomial time bound and $V(\cdot, \cdot, \cdot)$ is a polynomial time procedure.)

In other words, an algorithm solving problem $A$ should return YES on an input $x$ if an only if there exists some string $y_1$ such that for all strings $y_2$ (both of polynomial length), the predicate $V(x, y_1, y_2)$ holds. An example of such a problem is this: given a Boolean formula $\phi$ over variables $x_1, \ldots, x_n$, is there a formula $\phi'$ which is equivalent to $\phi$ and is of size at most $k$? In this case, $y_1$ is the formula $\phi'$, $y_2$ is an arbitrary assignment to the variables $x_1, \ldots, x_n$, and $V(x, y_1, y_2)$ is the predicate which is true if and only if $x[y_2]$ and $y_1[y_2]$ are both true or both false, meaning that under the variable assignment $y_2$, $\phi$ and $\phi'$ agree. Notice that $\phi'$ is equivalent to $\phi$ if and only if it agrees with $\phi$ under all assignments of Boolean values to the variables.

As we will see, the problem $A$ is a member of the class $\Sigma_2$ in the second level of the polynomial hierarchy.

## 2   The hierarchy

The polynomial hierarchy starts with familiar classes on level one: $\Sigma_1 = $ **NP** and $\Pi_1 = $ **coNP**. For all $i \geq 1$, it includes two classes, $\Sigma_i$ and $\Pi_i$, which are defined as follows:

$$A \in \Sigma_i \Leftrightarrow \exists y_1. \, \forall y_2. \; \ldots \; .Q y_i. \; V_A(x, y_1, \ldots, y_i)$$

and
$$B \in \Pi_i \Leftrightarrow \forall y_1. \exists y_2. \ \ldots \ .Q'y_i. \ V_B(x, y_1, \ldots, y_i)$$

where the predicates $V_A$ and $V_B$ depend on the problems $A$ and $B$, and $Q$ and $Q'$ represent the appropriate quantifiers, which depend on whether $i$ is even or odd (for example, if $i = 10$ then the quantifier $Q$ for $\Sigma_{10}$ is $\forall$, and the quantifier $Q'$ for $\Pi_{10}$ is $\exists$). For clarity, we have also omitted the conditions that each string $y_i$ must be of polynomial length, but such conditions must be added for a completely formal definition of $\Sigma_i$ and $\Pi_i$.

One thing that is easy to see is that $\Pi_k = \mathrm{co}\Sigma_k$. Also, note that, for all $i \leq k - 1$, $\Pi_i \subseteq \Sigma_k$ and $\Sigma_i \subseteq \Sigma_k$. These subset relations hold for $\Pi_k$ as well. This can be seen by noticing that the predicates $V$ do not need to "pay attention to" all of their arguments, and so can represent classes lower on the hierarchy which have a smaller number of them.

## 3   An Alternate Characterization

The polynomial hierarchy can also be characterized in terms of "oracle machines." The idea here is that, instead of a standard Turing machine, we consider one which is augmented with an oracle of a certain power which can be consulted as many times as desired, and using only one computational step each time. Syntactically, this can be written as follows.

Let $A$ be some decision problem and $\mathcal{M}$ be a class of Turing machines. Then $\mathcal{M}^A$ is defined to be the class of machines obtained from $\mathcal{M}$ by allowing instances of $A$ to be solved in one step. Similarly, if $\mathcal{M}$ is a class of Turing machines and $\mathcal{C}$ is a complexity class, then $\mathcal{M}^{\mathcal{C}} = \bigcup_{A \in \mathcal{C}} \mathcal{M}^A$. If $L$ is a complete problem for $\mathcal{C}$, and the machines in $\mathcal{M}$ are powerful enough to compute polynomial-time computations, then $\mathcal{M}^{\mathcal{C}} = \mathcal{M}^L$.

**Theorem 1** $\Sigma_2 = \mathbf{NP}^{3SAT}$.

PROOF: Let $A \in \Sigma_2$, then for some polynomial $p()$ and polynomial-time computable $V()$ we have

$$x \in A \text{ if and only if } \exists y_1 \text{ s.t. } |y_1| \leq p(|x|). \forall y_2 \text{ s.t. } |y_2| \leq p(|x|). V(x, y_1, y_2) = 1$$

Then we define a non-deterministic machine with an **NP**-oracle as follows: on input $x$, the machine guesses a string $y_1$ of length at most $p(|x|)$, and then asks the oracle whether $\exists y_2.|y_2| \leq p(|x|) \wedge V(x, y_1, y_2) = 0$. The above question is an existential question about a polynomial-time computation, so, by Cook's theorem, it is possible to construct in polynomial time a 3SAT instance that is satisfiable if and only if the answer to the above question is YES. The machine accepts if and only if the answer from the oracle is NO. It is immediate that the machine has an accepting computation if and only if

$$\exists y_1.|y_1| \leq p(|x|).(\neg \exists y_2.|y_2| \leq p(|x|) \wedge V(x, y_1, y_2) = 0)$$

that is, the machine accepts if and only if $x \in A$.

Notice that, in the above computation, only one oracle query is made, even though the definition of $\mathbf{NP}^{3SAT}$ allows us to make an arbitrary number of oracle queries.

Let now $A \in \mathbf{NP}^{3SAT}$, and let $M$ be the oracle machine that solves $A$. We first show that there is a machine $M'$ that also solves $A$, only makes one oracle query, and accepts

if and only if the answer to the oracle query is NO. On input $x$, $M'$ guesses an accepting computation of $M(x)$, that is, $M'$ guesses all the non-deterministic choices of $M(x)$, all the oracle questions, and all the answers. Then, for each question that was answered with a YES, $M'$ guesses a satisfying assignment to verify that the guess was correct. Finally, $M'$ is left with a certain set of oracle questions, say, the formulae $\phi_1, \ldots, \phi_k$, for which it has guessed that the correct oracle answer is NO. Then $M'$ asks its oracle whether (a formula equivalent to) $\phi_1 \vee \cdots \vee \phi_k$ is satisfiable, and it accepts if and only if the answer is NO.

Consider the computation of $M'(x)$ when $x \in A$: there is a valid accepting computation of $M(x)$, and $M'(x)$ can guess that computation along with the valid oracle answers; it can also guess valid assignments for all the queries for which the answer is YES; finally, it is left with unsatisfiable formulae $\phi_1, \ldots, \phi_k$, the answer to the single oracle query of $M'$ is NO, and $M'$ accepts.

Conversely, if $M'(x)$ has an accepting computation, then there must be a valid accepting computation of $M(x)$, and so $x \in A$. $\square$

In fact, a more general result is known, whose proof works along similar lines.

**Theorem 2** *For every $i \geq 2$, $\Sigma_i = \mathbf{NP}^{\Sigma_{i-1}}$.*

# 4  Additional Properties

Here are some more facts about the polynomial hierarchy, which we will not prove:

1. $\Pi_i$ and $\Sigma_i$ have complete problems for all $i$.

2. A $\Sigma_i$-complete problem is not in $\Pi_j$, $j \leq i-1$, unless $\Pi_j = \Sigma_i$, and it is not in $\Sigma_j$ unless $\Sigma_j = \Sigma_i$.

3. Suppose that $\Sigma_i = \Pi_i$ for some $i$. Then $\Sigma_j = \Pi_j = \Sigma_i = \Pi_i$ for all $j \geq i$.

4. Suppose that $\Sigma_i = \Sigma_{i+1}$ for some $i$. Then $\Sigma_j = \Pi_j = \Sigma_i$ for all $j \geq i$.

5. Suppose that $\Pi_i = \Pi_{i+1}$ for some $i$. then $\Sigma_j = \Pi_j = \Pi_i$ for all $j \geq i$.

We will just prove the following special case of part (3).

**Theorem 3** *Suppose $\mathbf{NP} = \mathbf{coNP}$. Then, for every $i \geq 2$, $\Sigma_i = \mathbf{NP}$.*

PROOF: Let us first prove that, under the assumption of the theorem, $\Sigma_2 = \mathbf{NP}$. Let $A \in \Sigma_2$ and let $M$ be the non-deterministic oracle machine that decides $A$ using oracle access to 3SAT. Let also $M'$ be the non-deterministic polynomial time Turing machine that decides the complement of the 3SAT problem. We now describe a non-deterministic polynomial time Turing machine $M''$ to decide $A$: on input $x$, $M''$ guesses an accepting computation of $M(x)$, along with oracle queries and answers; for each oracle question $\phi$ for which a YES answer has been guessed, $M''$ guesses a satisfying assignment; for each oracle question $\psi$ for which a NO answer has been guessed, $M''$ guesses an accepting computation of $M'(\psi)$. It is easy to verify that $M''(x)$ has an accepting computation if and only if $M^{3SAT}(x)$ has an accepting computation.

We can prove by induction on $i$ that $\Sigma_i = \mathbf{NP}$. We have covered the base case. Let us now suppose that $\Sigma_{i-1} = \mathbf{NP}$; then $\Sigma_i = \mathbf{NP}^{\Sigma_{i-1}} = \mathbf{NP}^{\mathbf{NP}} = \Sigma_2 = \mathbf{NP}$. $\square$

While it seems like an artificial construction right now, in future lectures we will see that the polynomial hierarchy helps us to understand other complexity classes.

## 5   The Karp-Lipton-Sipser Theorem

**Theorem 4 (Karp-Lipton-Sipser)** *If $\mathbf{NP} \subseteq \mathbf{SIZE}(n^{O(1)})$ then $\mathbf{PH} = \Sigma_2$. In other words, the polynomial hierarchy would collapse to its second level.*

Before proving the above theorem, we first show a result that contains some of the ideas in the proof of the Karp-Lipton-Sipser theorem.

**Lemma 5** *If $\mathbf{NP} \subseteq \mathbf{SIZE}(n^{O(1)})$ then there is a family of polynomial-size circuits that on input a 3CNF formula $\phi$ outputs a satisfying assignment for $\phi$ if one such assignment exists and a sequence of zeroes otherwise.*

PROOF: We define the circuits $C_n^1, \ldots, C_n^n$ as follows:

- $C_n^1$, on input a formula $\phi$ over $n$ variables outputs 1 if and only if there is a satisfying assignment for $\phi$ where $x_1 = 1$,

- $\cdots$

- $C_n^i$, on input a formula $\phi$ over $n$ variables and bits $b_1, \ldots, b_{i-1}$, outputs 1 if and only if there is a satisfying assignment for $\phi$ where $x_1 = b_1, \ldots, x_{i-1} = b_{i-1}, x_i = 1$

- $\cdots$

- $C_n^n$, on input a formula $\phi$ over $n$ variables and bits $b_1, \ldots, b_{n-1}$, outputs 1 if and only if $\phi$ is satisfied by the assignment $x_1 = b_1, \ldots, x_{n-1} = b_{n-1}, x_n = 1$.

Also, each circuit realizes an $\mathbf{NP}$ computation, and so it can be built of polynomial size. Consider now the sequence $b_1 = C_n^1(\phi)$, $b_2 = C_n^2(b_1, \phi)$, $\ldots$, $b_n C_n^n(b_1, \ldots, b_{n-1}, \phi)$. The reader should be able to convince himself that this is a satisfying assignment for $\phi$ if $\phi$ is satisfiable, and a sequence of zeroes otherwise. $\square$

We now prove the Karp-Lipton-Sipser theorem.

PROOF: [Of Theorem 4] We will show that if $\mathbf{NP} \subseteq \mathbf{SIZE}(n^{O(1)})$ then $\Pi_2 \subseteq \Sigma_2$. By a result in a previous lecture, this implies that $\mathbf{PH} = \Sigma_2$.

Let $L \in \Pi_2$, then there is a polynomial $p()$ and a polynomial-time computable $V()$ such that

$$x \in L \text{ iff } \forall y_1.|y_1| \le p(|x|)\exists y_2.|y_2| \le p(|x|).V(x, y_1, y_2) = 1$$

By adapting the proof of Lemma 5 (see Figure 1), or by using the statement of the Lemma and Cook's theorem, we can show that, for every $n$, there is a circuit $C_n$ of size polynomial in $n$ such that for every $x$ and every $y_1$, $|y_1| \le p(|x|)$,
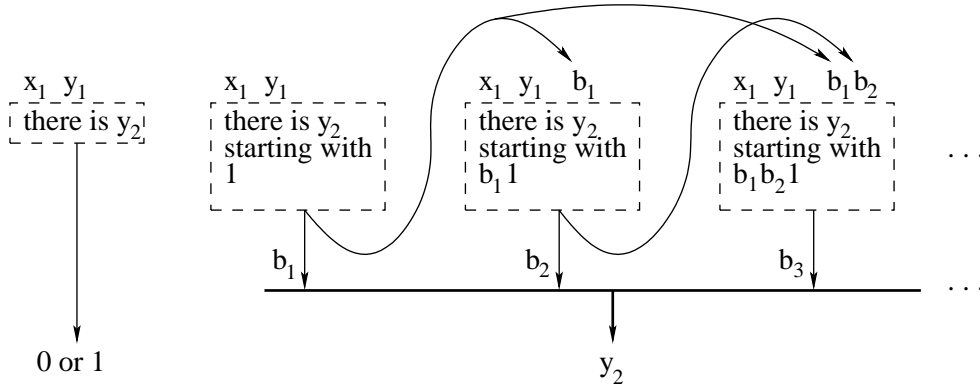
4

Figure 1: How to use decision problem solvers to find a witness to a search problem.

$$\exists y_2.|y_2| \leq p(|x|) \wedge V(x, y_1, y_2) = 1 \text{ iff } V(x, y_1, C_n(x, y_1)) = 1$$

Let $q(n)$ be a polynomial upper bound to the size of $C_n$.

So now we have that for inputs $x$ of length $n$,

$$x \in L \text{ iff } \exists C_n.|C_n| \leq q(n).\forall y_1.|y_1| \leq p(n).V(x, y_1, C_n(x, y_1)) = 1$$

which shows that $L$ is in $\Sigma_2$. □

## 6  References

The polynomial time hierarchy was defined by Stockmeyer [Sto76]. Wrathall [Wra76] shows that every class in the polynomial hierarchy has complete problems.

The Karp-Lipton-Sipser theorem appears in [KL80].

## References

[KL80]   R.M. Karp and R.J. Lipton. Some connections between nonuniform and uniform complexity classes. In *Proceedings of the 12th ACM Symposium on Theory of Computing*, pages 302–309, 1980. 5

[Sto76]  L.J. Stockmeyer. The polynomial-time hierarchy. *Theoretical Computer Science*, 3:1–22, 1976. 5

[Wra76]  C. Wrathall. Complete sets for the polynomial hierarchy. *Theoretical Computer Science*, 3:23–34, 1976. 5

# Exercises

1. In the MAX SAT problem we are given a formula $\phi$ in conjunctive normal form and we want to find the assignment of values to the variables that maximizes the number of satisfied clauses. (For example, if $\phi$ is satisfiable, the optimal solution satisfies all the clauses and the MAX SAT problem reduces to finding a satisfying assignment.) Consider the following decision problem: given a formula $\phi$ in conjunctive normal form and an integer $k$, determine if $k$ is the number of clauses of $\phi$ satisfied by an optimal assignment.

   - Prove that this problem is in **NP** if and only if **NP** = $co$**NP**.
     [Hint: prove that it is both **NP**-hard and $co$**NP**-hard.]
   - Prove that this problem is in $\Sigma_2$.

2. Define **EXP** = **DTIME**$(2^{n^{O(1)}})$. Prove that if **EXP** $\subseteq$ **SIZE**$(n^{O(1)})$ then **EXP** = $\Sigma_2$.