# Lecture 26

# Computational Complexity Theory

#### **Abhishek Shetty**

Undergraduate Department Indian Institute of Science

#### Raghav Malhotra

Undergraduate Department Indian Institute of Science

Instructor

### Chandan Saha

Computer Science and Automation Indian Institute of Science

November 5 2015

# 1 Leftover Hash Lemma

Let  $\mathcal{H}_{n,m}$  be a family of pairwise independent hash functions from  $\{0,1\}^n$  to  $\{0,1\}^m$  and  $\epsilon \in (0,1)$  be a constant.

**Theorem 1.1.** Let  $S \subseteq \{0,1\}^n$  be such that  $|S| \ge \frac{2^{m+2}}{\epsilon^2}$ . Let  $A = \{a \in S : h(a) = 0\}$  and  $n_A = |A|$ 

$$\Pr_{h \in \mathcal{H}_{n,m}} \{ |n_A - \frac{|S|}{2^m}| \ge \epsilon \frac{|S|}{2^m} \} \le \frac{1}{4}$$

*Proof.* Let  $\chi_i$  be the indicator variable defined as follows

$$\chi_i = \begin{cases} 1 & h(a_i) = 0 \\ 0 & \text{otherwise} \end{cases}$$

We note that  $n_A = \sum_{i=1}^k \chi_i$ . Let X denote the sum of random variables  $\chi_i$  and  $\mu$  denote  $\mathbb{E}[X] = \frac{|S|}{2^m}$ . From Chebyshev's Inequality, we have,

$$Pr\{|X - \mu| \ge \epsilon \mu\} \le \frac{Var(X)}{\epsilon^2 \mu^2}$$
 (1)

As the family  $\mathcal{H}$  is pairwise independent,

$$Var\left(\sum_{i=1}^{k} \chi_i\right) = \sum_{i=1}^{k} Var(\chi_i)$$

By definition,  $Var(\chi_i) = E[\chi_i^2] - (E[\chi_i])^2$  but as  $\chi_i^2 = \chi_i$ , we get  $Var(\chi_i) = E[\chi_i] - (E[\chi_i])^2$ . Thus, we get

$$Var(\chi_i) \le E[\chi_i] = Pr\{\chi_i = 1\} = \frac{1}{2^m}$$

Plugging this in (1), we get

$$Pr\{|X - \mu| \ge \epsilon \mu\} \le \frac{|S|/2^m}{\epsilon^2(|S|/2^m)^2} = \frac{2^m}{\epsilon^2|S|} \le \frac{2^m \epsilon^2}{2^{m+2}\epsilon^2} = \frac{1}{4}$$

as required.  $\Box$ 

## 2 Toda's theorem

**Theorem 2.1.**  $NP \cup co - NP \subseteq P^{\#SAT}$ 

*Proof.* This is clear since SAT and co-SAT can be decided by querying the oracle for the number of satisfying assignments and then checking whether the returned answer is zero.  $\Box$ 

The following theorem strengthens the above theorem considerably stating the whole of the polynomial hierarchy can be decided by a polynomial machine with access to an #SAT oracle.

Theorem 2.2 ([Tod91], [AB09]). 
$$PH \subseteq P^{\#SAT}$$

We prove this theorem in two steps. We show that any problem in the polynomial hierarchy can be randomly reduced to a problem in  $\bigoplus P$  (defined in section 2.1), that is  $PH \subseteq BPP^{\bigoplus P}$ . This is captured in the following theorem.

**Theorem 2.3.** There exists a probabilistic poly-time algorithm  $\mathbf{A}$ , which on input  $\psi$ , a quantified boolean formula with c quantifiers and a parameter m, outputs a boolean formula  $\phi$  such that

$$\psi$$
 is true  $\implies Pr\{\#\phi \text{ is odd }\} \ge 1 - 2^{-m}$   
 $\psi$  is false  $\implies Pr\{\#\phi \text{ is odd }\} \le 2^{-m}$ 

**A** runs in time  $poly(m, |\psi|)$ .

Then we derandomize this by using a more powerful #P oracle, proving that  $BPP^{\bigoplus P} \subseteq P^{\#P}$ . Towards proving the theorems we first look at the properties of the class  $\bigoplus P$ .

# 2.1 Class $\bigoplus P$

**Definition 2.1.** A language  $L \subseteq \{0,1\}^*$  is in  $\bigoplus P$  if there is a poly-time DTM M and a poly-time computable function q such that

$$x \in L \iff |\{u \in \{0,1\}^{q(|x|)} : \mathbf{M}(x,u) = 1\}| \text{ is odd.}$$

**Definition 2.2.**  $\bigoplus$  SAT := { $\phi : \phi$  is a boolean formula and  $\#\phi$  is odd}

**Theorem 2.4.**  $\bigoplus$  SAT is complete for  $\bigoplus$  P under polynomial time Karp reductions.

*Proof.* The theorem follows from the fact that  $\forall \mathtt{L} \in \mathsf{NP}, \mathtt{L}$  reduces to SAT parsimoniously, due to the Cook-Levin Theorem.

**Definition 2.3.** The  $\bigoplus$  quantifier is defined to be such that  $\bigoplus_x \phi(x)$  is true if the number of satisfying assignments of  $\phi$  is odd.

*Remark.* We note the following properties of the parity quantifier.

• Identifying true with 1 and false with 0 we have

$$\bigoplus_{x \in \{0,1\}^n} \phi(x) \equiv \sum_{x \in \{0,1\}^n} \phi(x) \pmod{2}$$

• Define  $(\phi \cdot \psi)(x,y) = \phi(x) \wedge \psi(y)$ . Note that,  $\#(\phi \cdot \psi) = \#\phi \cdot \#\psi$ . This gives us,

$$\bigoplus_x \phi(x) \bigwedge \bigoplus_y \psi(y) = \bigoplus_{x,y} (\phi \cdot \psi)(x,y)$$

• Let  $\#\phi(x) = m$  and  $\#\psi(y) = n$ . We would like to construct  $\gamma$  such that  $\#\gamma = m + n$ . Consider

$$\gamma(x,y,z) = ((z=0) \bigwedge \phi(x) \bigwedge (y=0)) \bigvee ((z=1) \bigwedge \psi(y) \bigwedge (x=0))$$

Note that  $\gamma$  satisfies the requirement. We denote  $\gamma = \phi + \psi$ .

• Let  $1(y) = y_1 \wedge \cdots \wedge y_n$ . Define  $(\phi + 1)(x, y) = \phi(x) + 1(y)$ . Note that  $\#(\phi + 1)(x, y) = \#\phi + 1$ . Thus, we have

$$\neg \bigoplus_{z} \phi(x) = \bigoplus_{x,y} (\phi + 1)(x,y)$$

Theorem 2.5. co- $\bigoplus P = \bigoplus P$ 

*Proof.* This is clear from the fact that  $\phi \in \overline{\bigoplus SAT} \implies (\phi + 1) \in \bigoplus SAT$ .

#### 2.2 Valiant-Vazirani Theorem

**Definition 2.4.** USAT :=  $\{\psi : \psi \text{ is a CNF formula and } \#\psi = 1\}$ 

**Theorem 2.6** ([VV85]). There is a poly-time PTM M such that on input  $\phi$  (boolean formula in n variables), M outputs another formula  $\psi$  such that

$$\phi \in \mathtt{SAT} \implies Pr\{\psi \in \mathtt{USAT}\} \geq \frac{1}{8n}$$
 
$$\phi \notin \mathtt{SAT} \implies Pr\{\psi \notin \mathtt{SAT}\} = 1$$

Proof. Let S be a the set of all satisfying assignments of input  $\phi$ . Note that  $0 \le |S| \le 2^n$ . Say that  $2^{k-2} \le |S| \le 2^{k-1}$  where  $k \in [2, n+1]$ . M picks a k randomly from the set  $\{1, ..., n+1\}$ . With probability  $n^{-1}$ , M will pick k for which  $2^{k-2} \le |S| \le 2^{k-1}$ . M then picks a hash function h uniformly at random from a pairwise independant family  $\mathcal{H}_{n,k}$ . Let  $X = |\{a \in S : h(a) = 0^k\}|$  and note that  $E(X) = \frac{|S|}{2^k}$ . Thus, we have  $\frac{1}{4} \le E[X] \le \frac{1}{2}$ . From the inclusion-exclusion principle we have,

$$Pr[X = 1] = Pr[X \ge 1] - Pr[X \ge 2]$$

Denoting S as  $\{a_1, \ldots a_{|S|}\}$ , we defined the following sets as  $\epsilon_i = \{h : h(a_i) = 0\}$ . We bound the above probabilities as follows.

$$\begin{split} Pr[X \geq 1] &= Pr\left[\bigcup_{i=1}^{|S|} \epsilon_i\right] \\ &\geq \sum_{i=1}^{|S|} Pr[\epsilon_i] - \sum_{i \neq j} Pr[\epsilon_i \cap \epsilon_j] \\ &= \frac{|S|}{2^k} - \binom{|S|}{2} \frac{1}{2^{2k}} \end{split}$$

Continuing

$$Pr[X \ge 2] \le \sum_{i \ne j} Pr[\epsilon_i \cap \epsilon_j]$$
$$= {|S| \choose 2} \frac{1}{2^{2k}}$$

Thus, we get

$$\begin{split} Pr[X = 1] &\geq \frac{|S|}{2^k} - 2\binom{|S|}{2} \frac{1}{2^{2k}} \\ &\geq \frac{|S|}{2^k} - \frac{|S|^2}{2^{2k}} \\ &= \frac{|S|}{2^k} \left(1 - \frac{|S|}{2^k}\right) \\ &\geq \frac{1}{8} \end{split}$$

Consider  $\psi(x) = \phi(x) \wedge (h(x) = 0)$ . We can find a CNF formula corresponding to h(x) = 0, say  $\tau(x)$ . Thus, we have a CNF formula  $\psi(x)$  such that

$$\phi \in \mathtt{SAT} \implies Pr\{\psi \in \mathtt{USAT}\} \geq \frac{1}{8n}$$
 
$$\phi \notin \mathtt{SAT} \implies Pr\{\psi \notin \mathtt{SAT}\} = 1$$

as required.  $\Box$ 

# References

- [AB09] Sanjeev Arora and Boaz Barak. Computational complexity: a modern approach. Cambridge University Press, 2009.
- [Tod91] Seinosuke Toda. PP is as hard as the polynomial-time hierarchy. SIAM Journal on Computing, 20(5):865-877, 1991.
- [VV85] Leslie G Valiant and Vijay V Vazirani. NP is as easy as detecting unique solutions. In *Proceedings* of the seventeenth annual ACM symposium on Theory of computing, pages 458–463. ACM, 1985.