

Lecture 7

Lecturer: Dan Spielman

Scribe: Abhinav Kumar

Today we will talk about Randomized Complexity classes.

Last time we showed $BPP \subset P/\text{poly}$. Today we will show $BPP \subset \Sigma_2 P \cap \Pi_2 P$

BPP error amplification

Error amplification means decreasing the probability of error. The key tool in this regard is the use of *Chernoff bounds*.

Theorem 1 Let X_1, X_2, \dots, X_n be independent, identically distributed random variables taking values in $[0,1]$. Let $Y = \sum_{i=0}^n X_i$, $\mu = E[Y]$.

Then $\Pr[|Y - \mu| > \epsilon\mu] < e^{-\mu\epsilon^2/3}$, for $\epsilon \in [0, 1]$.

Proof: See handout on probability.

Amplification Lemma: Let L be a language such that there is a randomized polynomial-time TM M such that,

$$\begin{aligned} x \in L \Rightarrow \Pr[M(x) \text{ accepts}] &\geq C(|x|) \\ x \notin L \Rightarrow \Pr[M(x) \text{ accepts}] &\leq S(|x|) \\ \text{where } C(n) - S(n) &\geq \frac{1}{p(n)} \text{ for some poly } p() \end{aligned}$$

Then $\forall b > 0, \exists$ a random ptme TM M' such that

$$\begin{aligned} x \in L \Rightarrow \Pr[M'(x) \text{ accepts}] &\geq 1 - 2^{-|x|^b} \\ x \notin L \Rightarrow \Pr[M'(x) \text{ accepts}] &\leq 2^{-|x|^b} \end{aligned}$$

(Remark : This is reasonably tight)

Proof: M' runs M $k = 12(p(n)^3 + n^b)$ times, $n = |x|$. Accepts if M accepted more than $k \cdot \frac{s(n)+c(n)}{2}$ times. Applying Chernoff's bound, we get the desired behaviour of M' .

Theorem 2 (Sipser) $BPP \subset \Sigma_2 P$.

Note: This implies $\text{co-BPP} \subset \Pi_2 P$, but $\text{co-BPP} = \text{BPP}$ since definition of BPP is symmetric.

Proof:

Let L be a language in BPP. We know $\exists A \in P$ and a function $f(n) = n^{O(1)}$ such that

If $w \in L$ then $\Pr_{r:|r|=f(n)}[(r, w) \in A] > 1 - 2^{-n}$, and

if $w \notin L$ then $\Pr_{r:|r|=f(n)}[(r, w) \in A] < 2^{-n}$.

where $n = |w|$.

Def. Define $R_w = \{r : |r| = f(n) \text{ such that } (r, w) \in A\}$. Correspondingly,

If $w \in L$ then $|R_w| > (1 - 2^{-n})2^{f(n)}$ and

if $w \notin L$ then $|R_w| < 2^{-n}2^{f(n)}$.

The idea here is to take a number of translations of R_w , and see if they cover the entire space $\{0, 1\}^{f(n)}$. Each translation of R_w has the same size as R_w , and if R_w is most of the space (ie, $w \in L$) then this collection of translations would be likely to cover the space. However, if R_w is very small (ie, $w \notin L$) then this collection could never cover the space. More formally,

Def. (translation) Let $S \subset \{0, 1\}^{f(n)}$. For $t \in \{0, 1\}^{f(n)}$ let the translation $S \oplus t$ be defined as

$$\{x : x \oplus t \in S\}$$

where $x \oplus t$ is defined as the XOR of the two strings (or, the bitwise sum modulo 2).

Claim. (1) If $|S| > (1 - 2^{-n})2^{f(n)}$ then $\exists \tau = \{t_1, \dots, t_{f(n)}\}$ such that

$$\bigcup_{i=1}^{f(n)} (S \oplus t_i) = \{0, 1\}^{f(n)}$$

(2) If $|S| < 2^{-n}2^{f(n)}$ then $\forall \tau = \{t_1, \dots, t_{f(n)}\}$,

$$\bigcup_{i=1}^{f(n)} (S \oplus t_i) \neq \{0, 1\}^{f(n)}$$

First, we show that the claim proves the theorem. If the claim is true, we can design a $\Sigma_2 P$ machine M to solve L as follows:

1. Use \exists states to generate τ .
2. Use \forall states to generate $r \in \{0, 1\}^{f(n)}$.
3. Check if $r \in \bigcup_i R_x \oplus t_i$ and accept if so, otherwise, reject.

This is polynomial time, since we can check whether $r \in R_x$ in polynomial time, and $f(n)$ is polynomial in n . By the claim, if $x \in L$ then on any correct τ we accept. If $x \notin L$ we reject, since there is no such τ . Therefore, we only have to prove the claim.

First, we prove part (2) of the claim. If $|S| < 2^{-n}2^{f(n)}$ then

$$\left| \bigcup_i (S \oplus t_i) \right| \leq f(n)2^{-n}2^{f(n)}.$$

Since $f(n) = n^{O(1)}$, $f(n)2^n < 1$ for sufficiently large n . Therefore, this union doesn't cover $\{0, 1\}^{f(n)}$.

Note that the "sufficiently large n " clause here doesn't cause a problem. If we take this into account, we need only hard-code the correct answer for all words smaller than this bound into our $\Sigma_2 P$ machine.

Next, we prove part (1). Let

$$\begin{aligned} p &= \Pr_{\tau}[\bigvee_{|r|=f(n)} r \in \bigcup_{i=1}^{f(n)} (t_i \oplus S)] \\ &= 1 - \Pr_{\tau}[\bigwedge_{|r|=f(n)} r \notin \bigcup_{i=1}^{f(n)} (t_i \oplus S)] \\ &\geq 1 - \sum_{|r|=f(n)} \Pr_{\tau}[r \notin \bigcup_{i=1}^{f(n)} (t_i \oplus S)]. \end{aligned}$$

We choose the t_i 's independently, so we can consider them independently. Therefore,

$$\begin{aligned} p &\geq 1 - \sum_{|r|=f(n)} \prod_{i=1}^{f(n)} \Pr_{t_i}[r \notin t_i \oplus S] \\ &= 1 - \sum_{|r|=f(n)} \prod_{i=1}^{f(n)} 2^{-n} \end{aligned}$$

since $r \in t_i \oplus S$ if and only if $t_i \in r \oplus S$,

$$= 1 - 2^{f(n)}(2^{-n})^{f(n)} = 1 - 2^{-f(n)(n-1)} > 0.$$

Since this probability is nonzero, there must be at least some τ for which the union of the translations determined by τ covers $\{0, 1\}^{f(n)}$. This completes the proof. ■

Verifying Polynomial identities

Let p be a given polynomial in k variables, q_1, \dots, q_k given polynomials in m variable. The equation

$$p(q_1(y_1, \dots, y_m), \dots, q_k(y_1, \dots, y_m)) = 0$$

can be difficult to check deterministically. Expanding the polynomial out would give us an exponential number of terms. However, if we use randomization there is an easy test. Choose x_1, \dots, x_m at random and see if we get zero. If we choose x'_i 's in a large enough range, the probability that the test is passed but p is not identically zero becomes exponentially small. So we can check polynomial identities in co-RP.

Lemma 3 (Schwartz's Lemma) *Let $P(x_1, x_2, \dots, x_n)$ be a polynomial of degree d . Then if $P \not\equiv 0$, then*

$$\Pr_{x_1, x_2, \dots, x_n \in S}[P(x_1, x_2, \dots, x_n) = 0] \leq \frac{dn}{|S|}.$$

Proof: We use induction on n .

Base case ($n = 1$): If $P \not\equiv 0$, then there are at most d zeroes in p . At most $d = d \cdot 1$ of them are in S .

Inductive step: Write

$$P(x_1, x_2, \dots, x_n) = \sum_{i=0}^{d_1} x_1^i P_i(x_2, \dots, x_n).$$

By hypothesis,

$$\Pr_{x_2, \dots, x_n}[P_{d_1}(x_2, \dots, x_n) = 0] \leq \frac{(n-1)d}{|S|}.$$

If $P_{d_1}(x_2, \dots, x_n) \neq 0$, then $\Pr_{x_1}[P(x_1, \dots, x_n)] \leq \frac{d_1}{|S|}$. ■