

## Lecture 11

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## 1 Recap

We defined **RP** as the class of languages accepted by PPT machine with one-sided error bounded below  $1/3$ , **BPP** with two-sided error with gap  $1/3$ . **RP** was shown to be robust in the following sense.

Define  $\mathbf{RP}_e$  such that  $L \in \mathbf{RP}_e$  if for some poly-time TM  $M$  and random bits  $y$ ,

$$x \in L \Rightarrow \Pr[M(x, y) \text{ rejects}] \leq e(|x|)$$

$$x \notin L \Rightarrow \Pr[M(x, y) \text{ accepts}] = 0$$

Then  $\mathbf{RP}_{1-1/\text{poly}(n)} = \mathbf{RP} = \mathbf{RP}_{1/2^{\text{poly}(n)}}$  (the two poly's may be different polynomials), yet  $\mathbf{RP}_{1-1/2^n} = \mathbf{NP}$ .

We will see that **BPP** is robust in the similar sense. Define  $\mathbf{BPP}_{c,s}$  such that  $L \in \mathbf{BPP}_{c,s}$  if for some poly-time TM  $M$  and random bits  $y$ ,

$$x \in L \Rightarrow \Pr[M(x, y) \text{ accepts}] \geq c(|x|)$$

$$x \notin L \Rightarrow \Pr[M(x, y) \text{ accepts}] \leq s(|x|)$$

Let us assume that, as often necessary, that  $s$  is “nice”, ie fully time constructible.

(Quick note: If  $c \leq s$  then  $\mathbf{BPP}_{c,s}$  would contain every language. While it is not required that  $c(n) \geq 0.5$  and  $s(n) \leq 0.5$ , one can shift the probability by proper amount so that  $c, s$  do straddle 0.5.)

## 2 Amplification for BPP

Using Chernoff bound we will see that  $\mathbf{BPP}_{f(n)+1/\text{poly}(n), f(n)-1/\text{poly}(n)} = \mathbf{BPP} = \mathbf{BPP}_{1-2^{-\text{poly}(n)}, 2^{-\text{poly}(n)}}$ .

**Theorem 1 (Chernoff bound)** Let  $X_1, \dots, X_k \in [0, 1]$  be independent random variables and  $X = \sum_i X_i/t$ . Then  $\Pr[|X - E[X]| \geq \epsilon] \leq e^{-(k\epsilon^2/2)}$ .

Suppose some poly-time TM  $M$  places  $L$  in  $\mathbf{BPP}_{f(n)+1/p(n), f(n)-1/p(n)}$  where  $p$  is a polynomial, and  $f$  a “nice” function. Intuitively if one runs  $M$  for  $k$  times (with different random bits) and output according to whether the average of  $k$  answers exceeds  $f(n)$ , the error probability should decrease.

By how much? Let random variable  $X_i$  denote the output of  $i$ th run. For  $x \in L$ , error occurs if  $\sum_i X_i/k < f(n)$  ie at least  $p(n)$  off expectation, thus with probability  $O(e^{-k/2p(n)^2})$  by Chernoff bound. With  $k$  polynomial in  $n$ , this can be as small as  $2^{-q(n)}$  for any polynomial  $q$ . Likewise for  $x \notin L$ .

Here we have used polynomially many more random bits to reduce error. Can we do with fewer? The state-of-art, using ideas from pseudorandomness (ie expanders), is that  $O(k)$  extra random bits *can* reduce error from  $1/3$  to  $2^{-k}$ .

## 3 $\mathbf{BPP} \subseteq \mathbf{P}/\text{poly}$

In advice (ie non-uniform) classes, one piece of (short) advice is expected to help *all*  $2^n$  computations on length  $n$  input. This might seem weak at first, but often times randomization is not more powerful than non-uniformity. In particular Adleman showed that  $\mathbf{BPP} \subseteq \mathbf{P}/\text{poly}$ .

Suppose machine  $M$  places  $L$  in **BPP** with error probability below  $2^{-p(n)}$ ,  $p(n) > n$  (okay due to amplification). Is there a random string  $y$  good for all  $2^n$  inputs of length  $n$ , ie  $M(x, y) = L(x)$  for each  $x \in \{0, 1\}^n$ ? Indeed, for each  $x$  only a  $2^{-p(n)}$  fraction of all random strings are bad; summing over all  $2^n$  possible  $x$  this fraction is still below 1! Thus some advice works for *all* inputs as random tape.

## 4 $\mathbf{BPP} \subseteq \Sigma_2^p \cap \Pi_2^p$

How about some uniform class upper bounding  $\mathbf{BPP}$ ? It is clear that  $\mathbf{BPP} \subseteq \mathbf{PSPACE}$ ; it is unclear how  $\mathbf{BPP}$  is related to  $\mathbf{NP}$ . Nevertheless we can show something intermediate:  $\mathbf{BPP} \subseteq \Sigma_2^p$ . (Which implies  $\mathbf{BPP} \subseteq \Sigma_2^p \cap \Pi_2^p$  as  $\mathbf{BPP}$  is closed under complementation.)

As before, suppose some machine  $M$  places  $L$  in  $\mathbf{BPP}$  with error probability below  $2^{-n}$ . Let  $x$  be a length  $n$  input, and  $M$  uses  $m$  random bits on  $x$ .

(Note that letting  $\exists$ -player show a set of polynomially many strings good for  $x$ , as evidence, is not enough. To decide  $L$  by a 2-round debate one must ensure some kind of “fairness”, eg say one might let  $\exists$ -player to produce first half bits of  $y$ ,  $\forall$ -player second half, and see if  $M(x, y)$  accepts. This is still too crude to work, but illustrates the point.)

The idea is we do let  $\exists$ -player show a set of polynomially many strings good for  $x$ , and the  $\forall$ -player tries to find some bijection mapping *all* of them to strings *bad* for  $x$ . The bijections allowed are  $\oplus y'$  for any  $y'$ . Intuitively, for  $x \in L$  it is hard for  $\forall$ -player to come up with such bijection that works on *all* good strings, and for  $x \notin L$  it is easy (and “easy” in a stronger sense than it is hard in the  $x \in L$  case).

Formally, one claims

$$L = \{x : \exists y_1, \dots, y_m \forall y' [ \bigvee_{1 \leq i \leq m} M(x, y_i \oplus y') = 1 ]\}$$

Proof. Suppose  $x \in L$ . Imagine one picks  $y_1, \dots, y_m$  at random. Probability that  $\bigwedge_i M(x, y_i \oplus y') = 0$ , for each  $y'$ , is below  $2^{-mn}$ ; union bound over all possible  $y'$  the probability is still below 1, ie *some*  $y_1, \dots, y_m$  make this false for all  $y'$ .

Now suppose  $x \notin L$ . Imagine one picks  $y'$  at random. Probability that  $\bigvee_i M(x, y_i \oplus y') = 1$ , for each  $y_1, \dots, y_m$ , is at most  $m2^{-n} < 1$ , ie *some*  $y'$  makes this false for all  $y_1, \dots, y_m$ .

This very idea can also be used to show  $\mathbf{promiseBPP} \subseteq \mathbf{promiseRP}^{\mathbf{promiseRP}}$  (ie if  $\mathbf{P} = \mathbf{promiseRP}$  then  $\mathbf{P} = \mathbf{promiseBPP}$ ).

## 5 Next time

We will talk about promise problems, which arise naturally eg when  $\mathbf{BPP}$  has no known complete problem (as we don't know how to enumerate error-bounded PPTs, ie to verify error-bounded-ness) yet  $\mathbf{promiseBPP}$  has complete problems (eg given input  $(M, x)$ , promised that  $M$  is indeed error bounded, does  $M(x) = 1$ ?).

We will talk about the complexity of UNIQUE – SAT, ie SAT with the promise that the satisfying assignment is either unique or non-existing. Is UNIQUE – SAT hard? We shall see that  $\mathbf{NP} \neq \mathbf{RP} \Rightarrow$  UNIQUE – SAT hard. (This problem arises in cryptography, where we want a mapping easy to compute one-way, but hard to revert. The mapping certainly should be one-to-one, so UNIQUE – SAT may be a good candidate.)