

# Lecture 16: Randomized Computation

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# Outline

- 1 Introduction
- 2 Probabilistic Turing Machine and the class BPP
- 3 One-Sided and Zero-Sided Error
- 4 Error Reduction for BPP
- 5 Relation of BPP with other classes

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# Introduction to Randomized Algorithms and Probabilistic Turing Machines

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- A randomized algorithm is an algorithm that is allowed access to a source of independent, unbiased, random bits. The algorithm is then permitted to use these random bits to influence its computation.
- We want to study TMs that has the power to toss random coins.

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# Probabilistic Turing Machine

## Definition: Probabilistic Turing Machine

A Probabilistic Turing Machine (PTM) is a Turing machine with two transition functions  $\delta_0, \delta_1$ . To execute a PTM  $M$  on an input  $x$ , we choose in each step with probability  $1/2$  to apply  $\delta_0$  and with probability  $1/2$  to apply  $\delta_1$ . This choice is made independently of all previous choices.

The machine outputs ACCEPT (1) or REJECT (0).  $M(x)$  denotes the output of  $M$  on  $x$  and surely this is a random variable.

For a function  $T : \mathbb{N} \rightarrow \mathbb{N}$ , we say that  $M$  runs in  $T(n)$ -time if for any input  $x$ ,  $M$  halts on  $x$  within  $T(|x|)$  steps regardless of the random choices  $M$  makes.

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- So how do we interpret  $Pr[M(x) = 1]$ ?
- It is simply the fraction of branches that end with  $M$ 's output of 1.
- An NDTM accepts if  $\exists$  one accepting branch; for a PTM, we consider the fraction of branches that leads to a 1.

# A New Class: BPP

- For a language  $L \subseteq \{0, 1\}^*$  and an input  $x \in \{0, 1\}^*$ , we define  $L(x) = 1$ , if  $x \in L$  and  $L(x) = 0$ , otherwise.

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**Definition: Class BPP (Bounded Error Probabilistic Polynomial Time)**

For  $T : \mathbb{N} \rightarrow \mathbb{N}$  and  $L \subseteq \{0,1\}^*$  we say that a PTM  $M$  decides  $L$  in time  $T(n)$  if for every  $x \in \{0,1\}^*$ ,  $M$  halts in  $T(|x|)$  steps irrespective of its random choices, and  $\Pr[M(x) = L(x)] \geq \frac{2}{3}$ , i.e.

$$\forall x \in L, \Pr[M \text{ accepts } x] \geq 2/3 \text{ and}$$

$$\forall x \notin L, \Pr[M \text{ rejects } x] \geq 2/3.$$

We let  $\text{BPTIME}(T(n))$  be the class of languages decided by PTMs in  $O(T(n))$  time and define  $\text{BPP} = \bigcup_c \text{BPTIME}(n^c)$ .

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- The above PTM satisfies the **excluded middle property**. That is, the PTM either accepts or rejects every input with a prob. at least  $2/3$ .

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- The class BPP has **two-sided error** (what it is?).

$x \notin L$ $M(x) = 0$	$x \in L$ $M(x) = 0$
$x \notin L$ $M(x) = 1$	$x \in L$ $M(x) = 1$

# An Alternate Definition

## Definition: Class BPP

A language  $L \in \text{BPP}$  if there exists a poly-time TM  $M$  and a polynomial  $p : \mathbb{N} \rightarrow \mathbb{N}$  such that for every  $x \in \{0, 1\}^*$ ,  $\Pr_{r \in_R \{0, 1\}^{p(|x|)}} [M(x, r) = L(x)] \geq \frac{2}{3}$  where  $r \in_R X$  denotes that  $r$  was chosen from the sample space  $X$ .

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### Relations between Classes P, EXP and BPP

$P \subseteq \text{BPP} \subseteq \text{EXP}$ .

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### Relations between Classes P, EXP and BPP

$P \subseteq \text{BPP} \subseteq \text{EXP}$ .

### Proof

Obvious.

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# One-Sided Error

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$x \notin L$ $M(x) = 0$	$x \in L$ $M(x) = 0$
$x \notin L$ $M(x) = 1$	$x \in L$ $M(x) = 1$

This error is not allowed.

So,  $\forall x \notin L, \Pr[M(x) = 0] = 1$

# One-Sided Error

## Definition: Class RP

A language  $L \subseteq \{0, 1\}^*$  is said to be in  $\text{RTIME}(T(n))$  if there exists a PTM running in time  $T(n)$  s.t.

$$\forall x \in L, \Pr[M(x) = 1] \geq \frac{2}{3}$$

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The class  $\text{RP} = \bigcup_{c>0} \text{RTIME}(T(n))$ .

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## Observation

$\text{RP} \subseteq \text{NP}$  but we do not know whether  $\text{BPP} \subseteq \text{NP}$ .

# One-Sided Error: The Complement Class

Definition: Class coRP

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A language  $L \subseteq \{0, 1\}^*$  is said to be in coRP if there exists a PTM running in polynomial time s.t.

$$\forall x \in L, \Pr[M(x) = 1] = 1$$

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# Error Reduction

## Theorem

Let  $L \subseteq \{0, 1\}^*$  be a language and suppose there exists a poly-time PTM  $M$  such that for every  $x \in \{0, 1\}^*$ ,  
 $\Pr[M(x) = L(x)] \geq \frac{1}{2} + |x|^{-c}$ . Then, for every constant  $d > 0$ ,  $\exists$  a poly-time PTM  $M'$  such that for every  $x \in \{0, 1\}^*$ ,  
 $\Pr[M'(x) = L(x)] \geq 1 - 2^{-|x|^d}$ .

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## Proof

- $M'$  does the following. For every input  $x \in \{0, 1\}^*$ , run  $M(x)$  for  $k = \text{poly}(|x|)$  times obtaining  $k$  outputs  $y_1, \dots, y_k \in \{0, 1\}$ . If the majority of these outputs is 1, then  $M'$  outputs 1, else  $M'$  outputs 0.

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- Now, use Chernoff bounds to fix the error probability as mentioned in the theorem.

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# BPP and $P_{/poly}$

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- Suppose  $L \in BPP$ .
- $\exists$  a TM  $M$  that on inputs of size  $n$  uses  $m$  random bits such that for every  $x \in \{0, 1\}^*$ ,  $\Pr_r[M(x, r) \neq L(x)] \leq 2^{-n-1}$ .

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- A random bit string  $r \in \{0, 1\}^m$  is **good** for an input  $x$  if it leads to  $M(x, r) = L(x)$ ; else  $r$  is **bad**.  $m = \text{poly}(n)$ .

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- For every  $x$ , at most  $\frac{2^m}{2^{n+1}}$  strings are bad for  $x$ .
- Add over all  $x \in \{0, 1\}^n$  to have at most  $2^{m-1}$  strings  $r$  that are bad for some  $x$ .

# BPP and $P_{/poly}$

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- The bound on bad strings imply,  $\exists$  a string  $r_0 \in \{0, 1\}^m$  that is good for every  $x \in \{0, 1\}^n$ .

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- The bound on bad strings imply,  $\exists$  a string  $r_0 \in \{0, 1\}^m$  that is good for every  $x \in \{0, 1\}^n$ .
- Now,  $M(x, r)$  is computable in poly-time. So,  $\exists$  a poly-sized circuit family  $\{C_n\}_{n \in \mathbb{N}}$  where  $C_n$  simulates  $M(x, r)$  correctly on inputs  $\langle x, r \rangle$  of length  $i$ .

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- To obtain the circuits, hard-code  $r_0$  obtained as above.

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- To obtain the circuits, hard-code  $r_0$  obtained as above.
- So, we obtain circuits  $C'_1, C'_2, \dots$  where  $C'_n = C_{n+\text{poly}(n)}(x, r_0)$  recognizing the language  $L$ .

# Another Relation: Is BPP in PH?

## Theorem

$$\text{BPP} \subseteq \Sigma_2^P \cap \Pi_2^P$$