

## Lecture 8. Randomized Complexity Classes

### 8.1 A randomized computation model

For studying randomized algorithms, we introduce a randomized Turing machine, which is a standard Turing machine equipped with a *random source tape*. The tape head of this tape can move only to the right reading one bit from the tape. Before the execution, we assume that a random binary sequence (which we call a *random source*) is given on this tape. For each such Turing machine  $M$ , we use  $r_M(\ell)$  to denote an upper bound of the length of a random source consumed by the machine on length  $\ell$  inputs, which we call a *random source length bound*. Clearly, we have  $r_M(\ell) \leq \text{time}_M(\ell)$ . For any  $\ell$ , any input  $x \in \{0, 1\}^\ell$ , and any  $u \in \{0, 1\}^{r_M(\ell)}$ , let  $M(x; u)$  denote the output of  $M$  when it is executed with  $x$  on its input tape and  $u$  on its random source tape. We will sometimes use  $M(x; u)$  to mean “the execution of  $M$  on input  $x$  and random source  $u$ .”

We define the probability that the randomized Turing machine  $M$  outputs  $y$  on input  $x$  (denoted as the left hand side, or, more simply (or more explicitly,  $\Pr_M[M(x) = y]$ ) as follows:

$$\Pr_u[M(x; u) = y] = \frac{\left\| \left\{ u \in \{0, 1\}^{r_M(\ell)} \mid M(x; u) = y \right\} \right\|}{2^{r_M(\ell)}}.$$

Note that a random binary string  $u$  is the source of the randomness used by  $M$ ; for any event on the execution of  $M$  on a given input  $x$ , we define its probability in the same way. For example, we can define the *average* running time of  $M$  on a given input  $x$  as follows.

$$E_u[\text{time}_M(x)] = \sum_{t \geq 1} t \cdot \Pr_u[M(x; u) \text{ terminates after the } t\text{th move}].$$

### 8.2 Randomized complexity classes

There are several ways to interpret the output of a randomized Turing machine. For any problem  $L$ , we consider here the following ways to solve  $L$ . (Note that  $L$  is a subset of  $\{0, 1\}^*$ ; recall that we identify a decision problem with a set of ‘yes’ instances.)

( $M$  barely solves  $L$ )

$$\begin{aligned} x \in L &\Rightarrow \Pr_M\{M(x) = 1\} > 1/2, \\ x \notin L &\Rightarrow \Pr_M\{M(x) = 0\} > 1/2. \end{aligned}$$

Bounded error: ( $\mathbf{M}$  solves  $L$  in BP-style)

$$x \in L \Rightarrow \Pr_{\mathbf{M}}\{\mathbf{M}(x) = 1\} \geq 2/3,$$

$$x \notin L \Rightarrow \Pr_{\mathbf{M}}\{\mathbf{M}(x) = 0\} \geq 2/3.$$

One sided error: ( $\mathbf{M}$  solves  $L$  in R-style)

$$x \in L \Rightarrow \Pr_{\mathbf{M}}\{\mathbf{M}(x) = 1\} \geq 2/3,$$

$$x \notin L \Rightarrow \Pr_{\mathbf{M}}\{\mathbf{M}(x) = 0\} = 1.$$

One sided error: ( $\mathbf{M}$  solves  $L$  in coR-style)

$$x \in L \Rightarrow \Pr_{\mathbf{M}}\{\mathbf{M}(x) = 1\} = 1,$$

$$x \notin L \Rightarrow \Pr_{\mathbf{M}}\{\mathbf{M}(x) = 0\} \geq 2/3.$$

Zero error: ( $\mathbf{M}$  solves  $L$  in ZP-style)

$$x \in L \Rightarrow \Pr_{\mathbf{M}}\{\mathbf{M}(x) = 1\} = 1,$$

$$x \notin L \Rightarrow \Pr_{\mathbf{M}}\{\mathbf{M}(x) = 0\} = 1.$$

Then define the following “standard” randomized complexity classes<sup>1</sup>.

$$\text{BPP} = \{L \mid \exists \text{ poly. time } \mathbf{M} \text{ solves } L \text{ in BP-style}\},$$

$$\text{RP} = \{L \mid \exists \text{ poly. time } \mathbf{M} \text{ solves } L \text{ in RP-style}\},$$

$$\text{coRP} = \{L \mid \exists \text{ poly. time } \mathbf{M} \text{ solves } L \text{ in coRP-style}\},$$

$$\text{ZPP} = \{L \mid \exists \text{ average poly. time } \mathbf{M} \text{ solves } L \text{ in ZPP-style}\}.$$

The following class is a bit different from the above classes because problems in this class may be still much more difficult than P. That is, the PP-style solvability is too weak to guarantee that the polynomial-time (or almost polynomial-time) tractability.

$$\text{PP} = \{L \mid \exists \text{ poly. time } \mathbf{M} \text{ solves } L \text{ in PP-style}\}.$$

Note here that the choice of the threshold  $2/3$  is not so essential w.r.t. the polynomial-time computability. We can increasing correct probability quite easily stated as follows.

**Lemma 8.1** (Correct probability amplification lemma)

For any decision problem  $L$ , suppose that we have a randomized  $\mathbf{M}$  that solves  $L$  in BP-style. For any  $m \geq 1$  (assuming odd), let  $\mathbf{M}^{(m)}$  be a randomized Turing machine that, for a given input  $x$ , executes  $\mathbf{M}(x)$  for  $m$  times independently and outputs the majority of the outputs of  $\mathbf{M}(x)$ . Then for any input  $x$ , we have

$$\Pr_{\mathbf{M}^{(m)}} [\mathbf{M}^{(m)}(x) \neq L(x)] \leq 2^{-m/32}.$$

This can be proved by using the following fact.

**Fact** (Chernoff bound) Consider independent random variables  $X_1, \dots, X_n$  such that each  $X_i$  takes value 1 with probability  $p$  and value 0 with probability  $1 - p$ . Let  $X = \sum_{i=1}^n X_i$  and  $\mu = pn$ . Then we have the following probability bounds.

$$\Pr[X > (1 + \varepsilon)\mu] \leq \exp(-\mu\varepsilon^2/3),$$

$$\Pr[X < (1 - \varepsilon)\mu] \leq \exp(-\mu\varepsilon^2/2)$$

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<sup>1</sup>The class ZPP is defined in a different way in the Japanese textbook.

### 8.3 Complexity analysis of randomized complexity classes

Note that the solvability condition has the following order: ZPP-type  $\Rightarrow$  R-type, co-Rtype  $\Rightarrow$  BP-type  $\Rightarrow$  PP-type, from which the following relations are immediate.

#### Theorem 8.2

$$P \subseteq RP, \text{ coRP} \subseteq BPP \subseteq PP.$$

Though not trivial as above, the following relations are also easy.

#### Theorem 8.3

$$\text{ZPP} = RP \cap \text{coRP}.$$

Intuitively we think that BPP (and its subclasses) is close to the class P. In fact, we may even conjecture that  $P = BPP$ . We show some justification for this conjecture. Recall<sup>2</sup> that PSIZE is the class of problems solvable by polynomial-size circuits. Due to the nonuniformity of our circuit model, we have  $\text{PSIZE} - P \neq \emptyset$ , intuitively, we may think that they are very close complexity classes. By using this class, we can also show that BPP is close to P.

#### Theorem 8.4

$$BPP \subseteq \text{PSIZE}.$$

**Proof.** Consider any problem  $L$  in BPP, and let  $M$  be a polynomial-time randomized Turing machine that solves  $L$  in BP-style. By the correct probability amplification lemma, we can define  $M_1$  whose error probability is less than  $2^{-\ell}$ . We may assume that  $M_1$  is still polynomial-time and hence the random source length bound is also polynomial.

Consider any input length  $\ell$ , and let  $r = r_{M_1}(\ell)$  be the length of a random source used by  $M_1$  on any input of length  $\ell$ . (For some input,  $M$  may not use all bits of a given random source.) Then by using the fact that the error probability is less than  $2^{-\ell}$  we can show that there exists a random source  $u_\ell$  with which  $M_1$  does not make any error; that is,  $M_1(x; u_\ell) = L(x)$  for all  $x \in \{0, 1\}^\ell$ . We call this  $u_\ell$  a *universal sequence*. Then using this universal sequence, we can define a circuit  $C_\ell$  for the problem  $L$  on inputs of length  $\ell$ . A family of circuits are defined by using such  $C_\ell$ 's for all  $\ell \geq 1$ .

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### Homework exercise from Lecture 8

**Homework rule:** Choose one of the basic problems or the advanced problem, and hand your answer in at the next class (for the basic problem) and at the next<sup>2</sup> class (for the advanced problem). (If you cannot attend the next class, you can submit your answer via email before the class.) You do not have to write a long answer. Usually one page would be enough. I will decide OK or NG, and you can get one point (for a basic problem) and two points (for an advanced problem) by each OK answer.

\* For writing an answer, you may use Japanese.

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<sup>2</sup>I might have forgot to defining this class; if not, then (sorry and) take this as the definition of PSIZE.

**Basic problems**

1. Prove Lemma 8.1.
2. Prove that  $\text{ZPP} \subseteq \text{RP} \cap \text{coRP}$ .
3. Prove that  $\text{ZPP} \supseteq \text{RP} \cap \text{coRP}$ .

**An Advanced problem**

1. In the proof of Theorem 8.4, we can also define  $M_2$  such that the 99% of its random sources are in fact universal sequences. In other words, almost all random sources are universal. Explain how to define  $M_2$  and why.