

# Computability and Complexity

Lecture 12, Thursday September 7, 2023

Ari Feiglin

## Definition 12.1:

Given a probabilistic algorithm  $M(x)$  whose time complexity is  $t(n)$ , we define a deterministic algorithm  $M_{\text{off}}(x, r)$  which gets a second input  $r$  whose length is  $t(|x|)$ .  $M_{\text{off}}(x, r)$  runs  $M(x)$  and it uses  $r$  to make the decisions for  $M(x)$ . For example, for its first decision the simulation of  $M(x)$  will use the value of  $r_1$  (or  $r[1]$ ) to determine which choice to make.

$M_{\text{off}}$  is called  $M$ 's offline algorithm.

Without loss of generality, we can assume that all choices are binary and so for every  $x$ ,

$$\mathbb{P}(M_{\text{off}}(x, r) \text{ is correct} \mid r \in \{0, 1\}^{t(|x|)}) = \mathbb{P}(M(x) \text{ is correct})$$

This is pretty immediate (keep in mind that in the case that  $M(x)$  makes fewer than  $t(|x|)$  decisions, the remaining bits of  $r$  will not affect  $M(x, r)$ 's running and thus will not affect the probability). Notice that the run time of  $M_{\text{off}}(x, r)$  is also  $O(t(n))$ .

## Theorem 12.2:

$$\mathbf{BPP} \subseteq \mathbf{P}_{/\text{poly}}$$

### Proof:

Let  $S$  be a problem in **BPP**, so there exists a probabilistic polynomial-time algorithm  $M$  which always has a non-zero probability of being correct. By a previous theorem, we can assume that  $M$ 's probability of being correct is greater than  $1 - 2^{-p(n)}$  for a polynomial  $p$ . Let us take  $p(n) = n + 1$  (the reasoning for this is that there are  $2^n$  possible inputs, so this works).

So we know that  $M_{\text{off}}$  also satisfies this probability:

$$\mathbb{P}(M_{\text{off}}(x, r) \text{ is correct} \mid r \in \{0, 1\}^t) \geq 1 - \frac{1}{2^{n+1}}$$

We need to show that there exists a sequence of advice (previously called commands),  $\{a_n\}_{n=0}^{\infty}$  such that  $M_{\text{off}}(x, a_{|x|}) = 1$  where  $a_n$ 's length is bound polynomially. We can't just take an  $r$  which makes  $M_{\text{off}}(x, r)$  correct, as this  $r$  may differ for every  $x$ , and the advice must be the same for all  $x$  of the same length.

We say that a sequence of choices  $r_n$  is *accurate* if for every input  $x$  of length  $n$ ,  $M_{\text{off}}(x, r_n)$  returns the correct answer. So to define our advice, we just take accurate sequences of choices. Therefore we need to show that for every  $n > 0$ , there exists an accurate sequence of choices. We will do this by showing that the probability a sequence of choices is accurate is non-zero, which necessitates the existence of an accurate sequence of choices. So let  $r_n$  be a random (uniformly chosen) sequence of choices of length  $n$ , we will compute the probability that it is accurate.

$$\begin{aligned} \mathbb{P}(r_n \text{ is an accurate sequence of choices}) &= \mathbb{P}(\forall |x| = n: M_{\text{off}}(x, r_n) \text{ is correct}) \\ &= 1 - \mathbb{P}(\exists |x| = n: M_{\text{off}}(x, r_n) \text{ is incorrect}) \geq 1 - \sum_{|x|=n} \mathbb{P}(M_{\text{off}}(x, r_n) \text{ is incorrect}) \end{aligned}$$

Now, the probability  $M_{\text{off}}(x, r_n)$  is incorrect is equal to the probability  $M(x)$  is incorrect, which is less than  $\frac{1}{2^{n+1}}$  and since there are  $2^n$  strings of length  $n$ , we get that this is greater than

$$\geq 1 - 2^n \cdot \frac{1}{2^{n+1}} = 1 - \frac{1}{2} = \frac{1}{2}$$

And so the probability that  $r_n$  is accurate is non-zero, meaning there must exist an accurate sequence of choices of length  $n$ .

So if we take our sequence of advice to be  $\{r_n\}_{n=0}^\infty$ , then we have that firstly,  $|r_n| \leq t(n)$  and so the length of the advice is polynomially bound. And for every  $x$ ,

$$M_{\text{off}}(x, r_{|x|}) = 1 \iff x \in S$$

since  $r_{|x|}$  is accurate. This is precisely the definition of a problem being in  $\mathbf{P}/_{\text{poly}}$ , meaning  $S \in \mathbf{P}/_{\text{poly}}$ . So we have shown that **BPP** is contained within  $\mathbf{P}/_{\text{poly}}$ , as required. ■

**Theorem 12.3:**

$$\mathbf{BPP} \subseteq \Sigma_2$$

**Proof:**

Let  $S$  be a problem in **BPP**, so there exists a deterministic polynomial-time algorithm  $M$  such that

$$\mathbb{P}\left(M(x, r) \text{ is correct} \mid r \in \{0, 1\}^t\right) \geq \frac{2}{3}$$

where  $t(n)$  is the polynomial runtime bound of  $M$  (this is the offline equivalent definition of **BPP**).

We showed last lecture that given a probabilistic algorithm  $M$  which solves a problem in **BPP**, we can do an amplification of  $M$  to get  $M'$  which runs  $M$   $k$  times and satisfies

$$\mathbb{P}\left(M'(x, r) \text{ is correct} \mid r \in \{0, 1\}^{t(n) \cdot k(n)}\right) \geq 1 - e^{-k(n)/18}$$

(We are viewing these algorithms as their offline equivalents.) The reason we must choose  $r \in \{0, 1\}^{t \cdot k}$  is since we are running  $M$   $k$  times, so each time we run it we need a new sequence of choices. Each sequence of choices must be of length  $t$ , and so in total we need a length of  $t \cdot k$ . So if we define  $k(n) = 18 \log(2t^2(n))$ , then eventually  $t \geq k$  and so we get that

$$\mathbb{P}\left(M'(x, r) \text{ is correct} \mid r \in \{0, 1\}^{18t \log(2t^2(n))}\right) \geq 1 - e^{-\log(2t^2(n))} = 1 - \frac{1}{2t^2(n)} \geq 1 - \frac{1}{2t(n)k(n)}$$

Let us define  $q(n) = t(n)k(n)$ , so we have that  $M'(x, r)$  is correct with a probability greater than  $1 - \frac{1}{2q}$ .

So  $M'$  utilizes  $q$  bits for  $r$  and returns a correct answer with a probability greater than  $1 - \frac{1}{2q}$ . Let us define  $M^*(x, r, \bar{s})$  where  $\bar{s}$  is a sequence of *masks*:  $s_1, \dots, s_q$  where for every  $i$ ,  $s_i \in \{0, 1\}^q$ .  $M^*$  will run  $M'$   $q$  times, and on the  $i$ th iteration it will run  $M'(x, r \otimes s_i)$  and it returns one if and only if at any point  $M'$  returns one. ( $\otimes$  means XOR: exclusive-or).

1. **function**  $M^*(x, r, \bar{s})$
2.     **for** ( $i$  **from** 1 **to**  $q(|x|)$ )
3.         **if** ( $M'(x, r \otimes s_i) = 1$ ) **return** 1
4.     **end for**
5.     **return** 0
6. **end function**

So we claim that  $M^*$  satisfies the requirements for  $\Sigma_2$ :

$$x \in S \iff \exists \bar{s} \forall r: M^*(x, r, \bar{s}) = 1$$

Let us first show that if  $x \notin S$  then for all  $\bar{s}$  there exist an  $r$  where  $M^*(x, r, \bar{s}) = 0$ . Let us randomly choose an  $r$ , and show that the probability  $M^*(x, r, \bar{s}) = 0$  is non-zero. Notice that since  $r$  is uniformly chosen, so is  $s_i \otimes r$  (since  $s_i \otimes r = a$  if and only if  $r = s_i \otimes a$ , which has uniform probability). Thus

$$\begin{aligned} \mathbb{P}(M^*(x, \bar{s}, r) = 0 \mid r \in \{0, 1\}^q) &= \mathbb{P}(\forall i: M'(x, s_i \otimes r) = 0 \mid r \in \{0, 1\}^q) \\ &\geq 1 - \mathbb{P}(\exists i: M'(x, s_i \otimes r) = 1 \mid r \in \{0, 1\}^q) \geq 1 - \sum_{i=1}^q \mathbb{P}(M'(x, s_i \otimes r) = 1 \mid r \in \{0, 1\}^q) = 1 - q \cdot \frac{1}{2q} = \frac{1}{2} \end{aligned}$$

Since the probability that  $M'(x, s_i \otimes r) = 1$  when  $x \notin S$  is less than  $\frac{1}{2q}$  (since as stated before,  $s_i \otimes r$  distributes uniformly). So there must exist an  $r$  such that  $M^*(x, r, \bar{s}) = 1$  for any  $\bar{s}$ , as required.

Now we will show that if  $x \in S$ , there exists a sequence of masks  $\bar{s}$  where for every sequence of choices  $r$ ,  $M^*(x, r, \bar{s}) = 1$ . Again here we will randomly choose a sequence of masks  $\bar{s} = s_1, \dots, s_q$  and show that with a non-zero probability, it satisfies the condition.

$$\begin{aligned} \mathbb{P}(\forall r: M^*(x, \bar{s}, r) = 1 \mid s_i \in \{0, 1\}^q) &= \mathbb{P}(\forall r \exists i: M'(x, r \otimes s_i) = 1 \mid s_i \in \{0, 1\}^q) \\ &= 1 - \mathbb{P}(\exists r \forall i: M'(x, r \otimes s_i) = 0 \mid s_i \in \{0, 1\}^q) \geq 1 - \sum_{r \in \{0, 1\}^q} \mathbb{P}(\forall i: M'(x, r \otimes s_i) = 0 \mid s_i \in \{0, 1\}^q) \end{aligned}$$

Since each  $s_i$  is chosen independently,  $r \otimes s_i$  is independent and so the events where  $M'(x, r \otimes s_i) = 0$  are independent. This means that

$$\mathbb{P}(\forall i: M'(x, r \otimes s_i) = 0 \mid s_i \in \{0, 1\}^q) = \prod_{i=1}^q \mathbb{P}(M'(x, r \otimes s_i) = 0 \mid s_i \in \{0, 1\}^q) \geq \frac{1}{(2q)^q}$$

So continuing our computations, we get

$$\mathbb{P}(\forall r: M^*(x, \bar{s}, r) = 1 \mid s_i \in \{0, 1\}^q) \geq 1 - \sum_{r \in \{0, 1\}^q} \frac{1}{(2q)^q} = 1 - 2^q \cdot \frac{1}{(2q)^q} = 1 - \frac{1}{q^q}$$

This is non-zero, meaning that there must exist such a sequence of masks.

So we have shown that

$$x \in S \iff \exists \bar{s} \forall r: M^*(x, r, \bar{s}) = 1$$

meaning that  $S \in \Sigma_2$ , as required. ■

Since **BPP** is closed under complements, we have that **BPP** = **coBPP**  $\subseteq$  **co** $\Sigma_2$  =  $\Pi_2$ . Thus we have shown

**Corollary 12.4:**

$$\mathbf{BPP} \subseteq \Sigma_2 \cap \Pi_2$$