

Chapter 8 - Probabilistic complexity classes

- ▶ Define probabilistic complexity classes
- ▶ Including **BPP**, **RP**, and **ZPP**
- ▶ Show how **BPP** relates to the polynomial time hierarchy, i.e.
BPP $\subseteq \Sigma_2 \cap \Pi_2$

Probabilistic algorithms - an example

- ▶ $MM := \{(A, B, C) \in \mathbb{Z}_2^{n \times n} \times \mathbb{Z}_2^{n \times n} \times \mathbb{Z}_2^{n \times n} \mid n \in \mathbb{N}, A \cdot B = C\}$
- ▶ $\overline{MM} := \{(A, B, C) \in \mathbb{Z}_2^{n \times n} \times \mathbb{Z}_2^{n \times n} \times \mathbb{Z}_2^{n \times n} \mid n \in \mathbb{N}, A \cdot B \neq C\}$

$M_{\overline{MM}}$ = "On input $A, B, C \in \mathbb{Z}_2^{n \times n}$:

1. Choose $x \in \mathbb{Z}_2^n$ uniformly at random.
2. Compute $y := B \cdot x, z := A \cdot y, w := C \cdot x$.
3. *Accept*, if $z \neq w$, otherwise *reject*."

Lemma 8.1

For all $A, B, C \in \mathbb{Z}_2^{n \times n}$:

1. if $(A, B, C) \notin \overline{MM}$, then $M_{\overline{MM}}$ rejects the triple (A, B, C) with probability 1,
2. if $(A, B, C) \in \overline{MM}$, then $M_{\overline{MM}}$ accepts the triple (A, B, C) with probability at least $1/2$.

In both cases, the probability is over the choice of x .

Balanced Turing machines

Definition 8.2

We call an NTM $N = (Q, \Sigma, \Gamma, \delta, q_0, q_{\text{accept}}, q_{\text{reject}})$ balanced, if there is a function $f : \mathbb{N} \rightarrow \mathbb{N}$ such that for all $x \in \Sigma^$ all computation branches of N on input x have length $f(|x|)$ and have for every nondeterministic step exactly two possible choices. We identify computation branches with elements in $\{0, 1\}^{f(|x|)}$.*

Properties

- ▶ Every polynomial time NTM can be simulated by a balanced polynomial time NTM.
- ▶ If N is a balanced NTM such that computation branches on input x have length $p(|x|)$ for a polynomial $p : \mathbb{N} \rightarrow \mathbb{N}$, then we call N *p-balanced*.

The classes **RP**, **co-RP**, and **ZPP**

Definition 8.3

The class **RP** consists of all languages L for which there is a polynomial $p : \mathbb{N} \rightarrow \mathbb{N}$ and a p -balanced NTM N with the following properties:

1. If $w \notin L$, then all computation branches of N on input w reject.
2. If $w \in L$, then at least half of the computation branches of N on input w accept, i.e. on input w NTM N has at least $2^{p(|w|)-1}$ accepting computation branches.

Definition 8.4

ZPP := **RP** \cap **co-RP**.

The class **BPP**

Definition 8.5

The class **BPP** consists of all languages L for which there is a polynomial $p : \mathbb{N} \rightarrow \mathbb{N}$ and a p -balanced NTM N with the following properties:

1. If $w \notin L$, then at most $1/4$ of the computation branches of N on input w accept, i.e. for $w \notin L$ the NTM has at most $2^{p(|w|)-2}$ accepting computation branches.
2. If $w \in L$, then at least $3/4$ of the computation branches of N on input w accept, i.e. on input w NTM N has at least $3 \cdot 2^{p(|w|)-2}$ accepting computation branches.

Amplifying the probabilities

Theorem 8.6

For $L \in \mathbf{BPP}$ there is a polynomial $p : \mathbb{N} \rightarrow \mathbb{N}$ and a p -balanced NTM N with the following properties:

1. If $w \notin L$, then N has at most $2^{-|w|} \cdot 2^{p(|w|)}$ accepting computation branches on input w .
2. If $w \in L$, then N has at least $(1 - 2^{-|w|}) \cdot 2^{p(|w|)}$ accepting computation branches on input w .

BPP and the polynomial time hierarchy

Theorem 8.7

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

Corollary 8.8

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

Proof of Theorem 8.7

- ▶ Language $L \in \mathbf{BPP}$
- ▶ $p : \mathbb{N} \rightarrow \mathbb{N}$ polynomial and N p -balanced NTM with $L(N) = L$
- ▶ for $x \in \{0, 1\}^*$ identify elements in $\{0, 1\}^{p(|x|)}$ with computation branches of N on input x
- ▶ set

$$A(x) := \{w \in \{0, 1\}^{p(|x|)} \mid w \text{ describes an } \textit{accepting} \\ \textit{computation branch of } N \textit{ on input } x\}$$

- ▶ from Theorem 8.6 we obtain

$$x \in L \Rightarrow |A(x)| \geq (1 - 2^{-|x|})2^{p(|x|)}$$

$$x \notin L \Rightarrow |A(x)| \leq 2^{-|x|}2^{p(|x|)}$$

Proof of Theorem 8.7

- ▶ for $S \subseteq \{0, 1\}^{\rho(|x|)}$ and $t \in \{0, 1\}^{\rho(|x|)}$ set

$$t \oplus S := \{t \oplus z \mid z \in S\}$$

- ▶ we show

$$x \in L \Rightarrow \exists t_1 \dots, t_{\rho(|x|)} \in \{0, 1\}^{\rho(|x|)} : \bigcup_{i=1}^{\rho(|x|)} t_i \oplus A(x) = \{0, 1\}^{\rho(|x|)}$$

$$x \notin L \Rightarrow \forall t_1 \dots, t_{\rho(|x|)} \in \{0, 1\}^{\rho(|x|)} : \bigcup_{i=1}^{\rho(|x|)} t_i \oplus A(x) \neq \{0, 1\}^{\rho(|x|)}$$

Proof of Theorem 8.7 - the case $x \notin L$

▶ $x \notin L \Rightarrow |A(x)| \leq 2^{-|x|} 2^{p(|x|)}$

\Rightarrow

$$\left| \bigcup_{i=1}^{p(|x|)} t_i \oplus A(x) \right| \leq p(|x|) 2^{-|x|} 2^{p(|x|)}$$

for all $t_1, \dots, t_{p(|x|)}$

▶ wlog. assume $p(|x|) 2^{-|x|} < 1$ (p is a polynomial)

\Rightarrow

$$x \notin L \Rightarrow \forall t_1, \dots, t_{p(|x|)} \in \{0, 1\}^{p(|x|)} : \bigcup_{i=1}^{p(|x|)} t_i \oplus A(x) \neq \{0, 1\}^{p(|x|)}$$

Proof of Theorem 8.7 - the case $x \in L$

Proof strategy

- ▶ Use the *probabilistic method*, i.e.
- ▶ show that for $t_1, \dots, t_{p(|x|)}$ chosen uniformly, independently at random

$$\Pr \left(\bigcup_{i=1}^{p(|x|)} t_i \oplus A(x) = \{0, 1\}^{p(|x|)} \right) > 0$$

$\Rightarrow t_1, \dots, t_{p(|x|)}$ with $\bigcup_{i=1}^{p(|x|)} t_i \oplus A(x) = \{0, 1\}^{p(|x|)}$ exist

Proof of Theorem 8.7 - the probabilistic argument

- ▶ fix $x \in L$ and choose $t_1, \dots, t_{p(|x|)}$ uniformly, independently at random, probabilities over this choice
- ▶ for all i

$$\Pr(y \notin t_i \oplus A(x)) = \Pr(t_i \notin y \oplus A(x)) \leq 2^{-|x|},$$

since $|A(x)| = |y \oplus A(x)| \geq (1 - 2^{-|x|})2^{p(|x|)}$

- ▶ the t_i 's are chosen independently, hence

$$\Pr\left(y \notin \bigcup_{i=1}^{p(|x|)} t_i \oplus A(x)\right) \leq 2^{-|x|p(|x|)},$$

Proof of Theorem 8.7 - the probabilistic argument

- ▶ by union bound

$$\Pr \left(\exists y \in \{0, 1\}^{\rho(|x|)} : y \notin \bigcup_{i=1}^{\rho(|x|)} t_i \oplus A(x) \right) \leq 2^{\rho(|x|)} \cdot 2^{-|x|\rho(|x|)}$$

- ▶ assuming $|x| \geq 2$, we have $2^{\rho(|x|)} \cdot 2^{-|x|\rho(|x|)} < 1$

⇒

$$\Pr \left(\bigcup_{i=1}^{\rho(|x|)} t_i \oplus A(x) = \{0, 1\}^{\rho(|x|)} \right) > 0$$

Proof of Theorem 8.7 - combining both cases

- ▶ overall

$$x \in L \Leftrightarrow \exists t = (t_1, \dots, t_{p(|x|)}) \in \{0, 1\}^{p(|x|)^2} \forall y \in \{0, 1\}^{p(|x|)} :$$
$$y \in \bigcup_{i=1}^{p(|x|)} t_i \oplus A(x).$$

- ▶ define language

$$A := \{(x, y, t) \in \{0, 1\}^{|x| \times p(|x|) \times p(|x|)^2} \mid t = (t_1, \dots, t_{p(|x|)}),$$
$$y \in \bigcup_{i=1}^{p(|x|)} t_i \oplus A(x)\}$$

Proof of Theorem 8.7 - combining both cases

- ▶ using A obtain

$$x \in L \Leftrightarrow \exists t = (t_1, \dots, t_{p(|x|)}) \in \{0, 1\}^{p(|x|)^2} \forall y \in \{0, 1\}^{p(|x|)} : \\ (x, y, t) \in A$$

- ▶ using Corollary 7.6 obtain $L \in \Sigma_2$