

Probabilistic TM (PTM)

- Defn: • We call M a PTM if it has two transition fns, δ_0, δ_1 and in each transition step M randomly follows δ_i with prob. = $1/2$.

• We say M decides L if $x \in L$ iff $\Pr_{\text{steps}} [M \text{ accepts } x] \geq 2/3$.

- Naturally, we can now talk about "efficient" PTMs.

Defn: • For a $T: \mathbb{N} \rightarrow \mathbb{N}$ a PTM M decides L in $T(n)$ time if M halts on every $x \in \{0,1\}^*$ in $\leq T(|x|)$ steps, regardless of its random choices, and decides $x \in L$.

• $\underline{\text{BPTIME}}(T(n)) := \{L \subseteq \{0,1\}^* \mid \text{a PTM } M \text{ decides } L \text{ in time } O(T(n))\}$.

• $\underline{\text{BPP}} := \bigcup_{c \in \mathbb{N}} \text{BPTIME}(n^c)$.

(bounded prob. poly-time)

(unlike, prob. poly-time PP !)

Proposition: (i) $P \subseteq \text{BPP} \subseteq \text{PP} \subseteq \text{Pspace} \subseteq \text{EXP}$.

(ii) Alternatively, $L \in \underline{\text{BPP}}$ if \exists det. poly-time TM M & $c > 0$ st. $\forall x \in \{0,1\}^*$,
 $x \in L$ iff $\Pr_{z \in \{0,1\}^{|x|^c}} [M(x,z) = 1] \geq \frac{2}{3}$.

- This is closer to our notion of a "randomized poly-time" algorithm M solving a problem L .

Examples of PTMs

- Primality: Given an $n \in \mathbb{N}$. Check whether it is prime.

• Solovay-Strassen (1970s) gave the first rand. poly-time algorithm.

• It was the first formal PTM!

Algo: (1) Pick a random $a \in (\mathbb{Z}/n\mathbb{Z})^*$.

(2) Output Yes if $a^{\frac{n-1}{2}} \equiv \left(\frac{a}{n}\right) \pmod{n}$.

Jacobi symbol

Exercise: Prove the correctness & the $\tilde{O}(\lg^2 n)$ time-complexity.

Polynomial Identity testing: Given a polynomial $f \in \mathbb{F}[x_1, \dots, x_n]$ in some "compact" way, check whether $f \stackrel{?}{=} 0$.
arithmetic circuit

Exercise: Prove that a random evaluation works.

- Exs. $\sqrt{\cdot} \bmod p$ & undirected connectivity in \mathbb{L} .

Open: (1) $P = BPP$?

(theoretical evidence for "yes"!))

(2) $BPP \neq NP$?

(later we will show a PH collapse!)

- BPP captures prob. algo, with two-sided error, i.e. if a PTM M decides L then it may make an error on x regardless of $x \in L$ or $x \notin L$.

One-sided error : RP & coRP

- Defn: • $L \in \text{Rtime}(T(n))$ if \exists PTM running in time $O(T(n))$ s.t.

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

$$x \notin L \Rightarrow \Pr [M \text{ accepts } x] = 0.$$

$$\bullet \text{ RP} := \bigcup_{c \in \mathbb{N}} \text{Rtime}(n^c)$$

↑ (randomized poly-time)

- Proposition: (i) Primes \in coRP. * Primes \in RP & required different ideas.
- (ii) PIT \in coRP.
- (iii) $RP \cup \text{coRP} \subseteq \text{BPP}$.
- (iv) $RP \subseteq NP$ & $\text{coRP} \subseteq \text{coNP}$.

Zero-sided error probabilistic: ZPP (Las Vegas algo.)

- Defn: • Consider a PTM M and the random variable $\text{time}_M(x)$, on any input x . We say that M has an expected running-time $T(n)$ if $\forall x, \text{Exp}[\text{time}_M(x)] \leq T(|x|)$.

• $L \in \underline{\text{Ztime}}(T(n))$ if \exists PTM that correctly decides L in expected time $O(T(n))$.

• ZPP := $\bigcup_{C \in \mathbb{N}} \text{Ztime}(n^C)$.

Las Vegas
↓

Monte Carlo (BPP)
↓ ↓

Proposition: (i) $ZPP \subseteq RP \cap coRP$.

(ii) $RP \cap coRP \subseteq ZPP$.

(iii) $ZPP = RP \cap coRP \subseteq NP \cap coNP$.

Proof:

(i) Let $L \in ZPP$ be decided by a PTM M with expected running-time $T(n)$.

• On an input x :

1) Run $M(x)$ for $3 \cdot T(|x|)$ steps.

2) If x is not accepted, output NO.

• If $x \notin L$, we made no error.

• If $x \in L$, we err with prob.

$$\leq \frac{T(|x|)}{3 \cdot T(|x|)} = \frac{1}{3}.$$

Markov's inequality \Rightarrow

$\Rightarrow L \in RP$.

($\because L \in ZPP$)

• Similarly, we can prove $L \in coRP$.

(Instead of NO, output Yes.)

$\Rightarrow L \in RP \cap coRP$.

□

(ii) Let $L \in RP \cap coRP$ be decidable by PTMs M_1 resp. M_2 in time $\leq n^c$, for a constant $c > 0$.

• On input x :

1) Pick a random r .

2) Run $M_1(x, r)$ & $M_2(x, r)$.

3) If $M_1(x, r) = M_2(x, r)$ then output the common decision. else repeat (1).

• Suppose $x \in L$. Thus, $M_2(x, r) = \text{Yes}$.

$$\Pr_r [M_1(x, r) \neq \text{Yes}] \leq 1/3.$$

$$\Rightarrow \Pr_{r_1, \dots, r_t} [\forall i \in [t], M_1(x, r_i) \neq \text{Yes}] \leq 3^{-t}.$$

$$\Rightarrow \text{Exp} [\# \text{ iterations}] \leq \sum_{t=1}^{\infty} (t+1) \cdot \frac{1}{3^t} = O(1).$$

$$\Rightarrow \text{Expected time complexity} = O(n^c).$$

- The case of $x \notin L$ is similar.
 $\Rightarrow L \in ZPP. \quad \square$

Why 2/3? Prob. amplification

- The (2/3)-rd in the defn. of prob. classes is arbitrary. In fact, we can use any fraction that is inverse-polynomial away from $1/2$.

Theorem: Let a PTM M be deciding L s.t.
 $\forall x, x \in L \text{ iff } \Pr[M \text{ accepts } x] \geq$
 $(\frac{1}{2} + |x|^{-c}).$

Then, $\forall d, \exists$ PTM M' s.t. $\forall x,$
 $x \in L \text{ iff } \Pr[M' \text{ accepts } x] \geq (1 - 2^{-|x|^d}).$

Pf sketch:

- Idea: Run M k times on x , and output the majority value. Apply the Chernoff bound on error prob.

• The PTM M' is: (Fix $k = 8 \cdot |x|^{d+2c}$)

On input x , run $M(x)$ k times.
Let the outputs be $y_1, \dots, y_k \in \{0, 1\}$.
Output Majority(y_1, \dots, y_k).

• For $i \in [k]$, let X_i be the random variable $\begin{cases} 0, & \text{if } y_i \text{ is wrong} \\ 1, & \text{otherwise.} \end{cases}$

Chernoff's bound: Let X_1, \dots, X_k be independent identically distributed (i.i.d.) boolean random variables. Let $\Pr[X_i = 1] =: p$ for $i \in [k]$ & $\delta \in (0, 1)$. Then,

$$\Pr \left[\left| \frac{\sum_{i \in [k]} X_i}{k} - p \right| > \delta \right] < e^{-\delta^2 pk/4}.$$

$$\Rightarrow \Pr[M' \text{ is wrong}] = \Pr \left[\sum_{i=1}^k X_i < k/2 \right]$$

$p := \Pr[M(x) \text{ is correct}]$

$$= \Pr \left[p - \frac{\sum X_i}{k} > p - \frac{1}{2} \right] = \Pr \left[\left| p - \frac{\sum X_i}{k} \right| > n^{-c} \right]$$

$$< \exp\left(-\frac{1}{4} \cdot n^{-2c} \cdot \left(\frac{1}{2} + n^{-c}\right) \cdot 8n^{d+2c}\right)$$

$$= \exp\left(-n^d \cdot (1 + 2n^{-c})\right)$$

$$< e^{-n^d} < 2^{-n^d} \quad \square$$

Exercise: The Chernoff bound has a neat proof using $\mathbb{E}\left[e^{t \cdot \sum X_i}\right]$ & the

Markov's bound.

BPP & the PH

- $BPP \subseteq? NP$ is not known, but $BPP \subseteq \Sigma_2 \cap \Pi_2$ is!

Theorem (Sipser-Gács 1983): $BPP \subseteq \Sigma_2 \cap \Pi_2$.

Pf:

- It suffices to show $BPP \subseteq \Sigma_2$.

• Let $L \in \text{BPP}$, and M be a poly-time TM ($m := n^c$) s.t. $\forall x \in \{0,1\}^n$,
 $x \in L \Rightarrow \Pr_{r \in \{0,1\}^m} [M(x,r) = 1] \geq (1 - 2^{-n})$

$x \notin L \Rightarrow \Pr_r [M(x,r) = 1] \leq 2^{-n}$.

• Denote $S := \{r \in \{0,1\}^m \mid M(x,r) = 1\}$.

Then, as before,

$|S| \geq (1 - 2^{-n}) 2^m$ if $x \in L$,

$|S| \leq 2^{m-n}$ if $x \notin L$.

• The idea is to check the "largeness" of this S in Σ_2 . (Use "expansion" in a graph.)

• For $u = \{u_1, \dots, u_k\} \subseteq \{0,1\}^m$, define an undirected graph G_u with:

$\{0,1\}^m$ as vertices, and

edges (s, s') , where $s \oplus s' = u_i$ for some i .

• Note that G_u is regular with $\text{deg} = k$.

• Fix $k := \lfloor \frac{m}{n} \rfloor + 1$.

• For any $S \subseteq \{0,1\}^m$, define $\Gamma_u(S)$ to be the neighbours of S in G_u .

Claim 1: $|S| \leq 2^{m-n} \Rightarrow \forall u, |u|=k, |\Gamma_u(S)| < 2^m$.

Pf:

• $|\Gamma_u(S)| \leq k \cdot |S| \leq \frac{k \cdot 2^m}{2^n} < 2^m$. \square

Claim 2: $|S| \geq (1-2^{-n})2^m \Rightarrow \exists u, |u|=k, \Gamma_u(S) = \{0,1\}^m$.

Pf:

• We construct a u probabilistically!

• Choose $u_1, \dots, u_k \in \{0,1\}^m$ randomly.

• Let E_r be the event that $r \notin \Gamma_u(S)$
& $E_{r,i}$ " " " " $r \notin S \oplus u_i$.

• Clearly,

$$\Pr_u[E_{r,i}] = 1 - \frac{|r \oplus s|}{2^m} \leq 2^{-n}.$$

• So,

$$\Pr_u[E_r] \leq \prod_{i=1}^k \Pr_u[E_{r,i}] \leq 2^{-nk} < 2^{-m}.$$

$$\Rightarrow \Pr_u[\exists r, E_r] < 1.$$

$$\Rightarrow \Pr_u[\forall r, \neg E_r] > 0.$$

$$\Rightarrow \exists u, \tau_u(s) = \{0,1\}^m. \quad \square$$

• Claims 1 & 2 imply: $\forall x \in \{0,1\}^n$,
 $x \in L$ iff $\exists u_1, \dots, u_k, \forall r, \forall_{i \in [k]} M(x, r \oplus u_i) = 1$.
 $\overset{R}{\in \{0,1\}^m} \rightarrow$

$$\Rightarrow L \in \Sigma_2. \quad \square$$