Practical session 9

Exercise 1 Present a two-sided error monte-carlo random algorithm, that given 3 polynomials, outputs 1 if and only if exactly two of them are identical.

Solution

Let PIT denote the Schwartz-Zippel Algorithm for verifying polynomial identities that was presented in Lecture 7 (except that the algorithm outputs 1 if it finds the two polynomials it compares to be identical, and 0 otherwise). We now present our algorithm for the problem at hand.

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Algorithm 0.1. Input: 3 polynomials P, Q, and R.

- 1. Let $b_1 = PIT(P, Q)$, $b_2 = PIT(Q, R)$, and $b_3 = PIT(P, R)$.
- 2. Output 1 if exactly one of the b_i's is 1, and output 0 otherwise.

Let S be the set from which the PIT algorithm samples numbers. Let us recall the failure probability of PIT, on input P and Q. If $P \equiv Q$, then $\Pr(\text{PIT}(P,Q) = 1) = 1$. If $P \not\equiv Q$, then $\Pr(\text{PIT}(P,Q)) \le \deg(P-Q)/|S|$. We next analyze Algorithm 0.1. Let OUT denote the output of Algorithm 0.1. There are 3 cases to consider. For the first case, let us assume that exactly two of the polynomials are identical. By symmetry we may assume that $P \equiv Q \not\equiv R$. Then b_1 always equals 1. Hence Algorithm 0.1 outputs the wrong answer if and only if $b_2 = 1$ or $b_3 = 1$. Thus, the probability of failure is

$$\Pr\left(OUT=0\right) = \Pr\left(\operatorname{PIT}\left(Q,R\right) = 1 \vee \operatorname{PIT}\left(P,R\right) = 1\right) \leq \frac{\deg\left(Q-R\right) + \deg\left(P-R\right)}{|S|},$$

where the above inequality holds by a union bound. Next, we assume that none of the three pairs of polynomials are identical, i.e., that $P \not\equiv Q \not\equiv R \not\equiv P$. Failure occurs if and only if OUT = 1. Thus, failure occurs if and only exactly one of b_1 , b_2 , b_3 is equal to 1. For $i \in \{1, 2, 3\}$ let E_i be the event that only b_i equals 1. Then

$$\begin{aligned} \Pr\left(OUT = 1\right) &= \Pr\left(E_1 \vee E_2 \vee E_3\right) \\ &\leq \Pr\left(E_1\right) + \Pr\left(E_2\right) + \Pr\left(E_3\right) \\ &\leq \Pr\left(\operatorname{PIT}\left(P,Q\right) = 1\right) + \Pr\left(\operatorname{PIT}\left(Q,R\right) = 1\right) + \Pr\left(\operatorname{PIT}\left(P,R\right) = 1\right) \\ &\leq \frac{\deg\left(P - Q\right) + \deg\left(Q - R\right) + \deg\left(P - R\right)}{|S|}, \end{aligned}$$

where the first inequality holds by a union bound and the second inequality holds since $E_1 \subseteq \{PIT(P,Q)=1\}$, $E_2 \subseteq \{PIT(Q,R)=1\}$, and $E_3 \subseteq \{PIT(P,R)=1\}$. For the final case, assume

that $P \equiv Q \equiv R$. Then it always holds that $b_1 = b_2 = b_3 = 1$. Therefore Algorithm 0.1 always outputs 0 implying that the probability of failure is 0.

To summarize, if the algorithm outputs 0, then it is wrong with probability at most

$$\max\left\{\frac{\deg\left(P-Q\right)+\deg\left(Q-R\right)}{|S|},\frac{\deg\left(P-Q\right)+\deg\left(P-R\right)}{|S|},\frac{\deg\left(Q-R\right)+\deg\left(P-R\right)}{|S|}\right\}$$

and if the algorithm outputs 1, then it is wrong with probability at most

$$\frac{\deg(P-Q) + \deg(Q-R) + \deg(P-R)}{|S|}.$$

Exercise 2 Let $L: \{0,1\}^* \to \{0,1\}$ and let M be a randomized algorithm such that

$$\forall x \in \{0,1\}^* \ M(x) \in \{0,1\} \ \text{and} \ \Pr(M(x) = L(x)) \ge \frac{1}{2} + \varepsilon,$$

for some $\varepsilon > 0$. Show that for any t > 0, there exists a randomized algorithm M_t , such that

$$\forall x \in \{0,1\}^* \ \Pr(M_t(x) = L(x)) \ge 1 - 2^{-t}.$$

Solution

Fix t > 0, and define M_t as follows:

Algorithm 0.2. *Input:* $x \in \{0, 1\}^*$.

- 1. Execute M(x) k times, for some k to be determined later, where each execution is independent of all other executions.
- 2. Let $b_i \in \{0,1\}$ be the output of the ith execution.
- 3. Output maj (b_1, \ldots, b_k) .

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We now analyze M_t . Let $x \in \{0,1\}^*$ and let X be the number of b_i 's that are equal to L(x). Then $X \sim \text{Bin}(k,p)$, for some $p \geq \frac{1}{2} + \varepsilon$, implying that $\mathbb{E}(X) \geq (\frac{1}{2} + \varepsilon)k$. Hence

$$\Pr(M_t(x) \neq L(x)) = \Pr\left(X < \frac{1}{2}k\right)$$

$$\leq \Pr\left(X - \mathbb{E}(X) < \frac{1}{2}k - \left(\frac{1}{2} + \varepsilon\right)k\right)$$

$$\leq \Pr\left(X \leq \mathbb{E}(X) - \varepsilon k\right)$$

$$\leq e^{-\frac{2\varepsilon^2 k^2}{k}}$$

$$= e^{-2\varepsilon^2 k},$$

where the third inequality is by Chernoff's inequality for a binomial random variable (see Lecture 1). Taking $k = \left\lceil \frac{t \ln 2}{2\varepsilon^2} \right\rceil$ then implies

$$\Pr\left(M_t(x) = L(x)\right) \ge 1 - e^{-2\varepsilon^2 k} \ge 1 - 2^{-t}.$$