

Fundamentals of Cryptography: Problem Set 3

Due Wed Oct 15 3PM

Collaboration is permitted (and encouraged); however, you must write up your own solutions and acknowledge your collaborators.

If a problem has **Opt**, it will not be graded.

Problem 0 Read Section 7.1, 7.2, 7.3 of “Introduction to Modern Cryptography (2nd ed)” by Katz & Lindell.

If you are curious about how to construct PRG from OWF, you may read “Pseudo-random Generators from One-Way Functions: A Simple Construction for Any Hardness” by Thomas Holenstein.

Problem 1 (Opt) Concentration Inequalities This problem recaps a few useful probability bounds. They show how random variables “concentrate” around their means. Section A of “Introduction to Modern Cryptography (2nd ed)” may help you answer this question.

Part A (Markov’s Inequality) Let X be a random variable over non-negative real numbers. Prove that, for any $a > 0$,

$$\Pr[X \geq a] \leq \frac{\mathbb{E}[X]}{a}.$$

Part B (Chernoff Bound) Let $p \in [0, 1]$ be a constant. Let X_1, \dots, X_n be random variables that are sampled independently from $\text{Bern}(p)$. That is, for each $i \in \{1, \dots, n\}$, we have $X_i \in \{0, 1\}$ and $\Pr[X_i = 1] = p$.

(1) Compute $\mathbb{E}[e^{t \sum_i X_i}]$ for any $t \in \mathbb{R}$.

(2) Prove that,

$$\Pr\left[\frac{1}{n} \sum_i X_i \geq p + \varepsilon\right] \leq \frac{\mathbb{E}[e^{t \sum_i X_i}]}{e^{tn(p+\varepsilon)}},$$

for any $t > 0$.

(3) Optimize the above bound by choosing t wisely.

The optimized bound is called *Chernoff bound*, it should look like

$$\Pr\left[\frac{1}{n} \sum_i X_i \geq p + \varepsilon\right] \leq e^{-D(p+\varepsilon\|p) \cdot n},$$

where $D(p + \varepsilon\|p)$ is the notation of KL divergence, and is defined as $D(p + \varepsilon\|p) := (p + \varepsilon) \log\left(\frac{p+\varepsilon}{p}\right) + (1 - p - \varepsilon) \log\left(\frac{1-p-\varepsilon}{1-p}\right)$. Since $D(p + \varepsilon\|p) \geq 2\varepsilon^2$, *Chernoff bound* can be relaxed to

$$\Pr\left[\frac{1}{n} \sum_i X_i \geq p + \varepsilon\right] \leq e^{-2\varepsilon^2 n}.$$

Part C (Chebyshev's Inequality) Let X be a random variable. Prove that

$$\Pr\left[|X - \mathbb{E}[X]| \geq a\right] \leq \frac{\text{Var}[X]}{a^2}$$

for any $a > 0$. Here $\text{Var}[X] := \mathbb{E}[(X - \mathbb{E}[X])^2]$ is the variance of X .

Let X_1, \dots, X_n be random variables such that $\mathbb{E}[X_i] = p$ and $\text{Var}[X_i] = \sigma^2$ for all i . We also assume that X_1, \dots, X_n are *pair-wise independent*. Prove that

$$\Pr\left[\left|\frac{1}{n} \sum_i X_i - p\right| \geq a\right] \leq \frac{\sigma^2}{na^2}$$

for any $a > 0$.

Problem 2 (14pt) Assume f is a length-preserving OWF (i.e., $|f(x)| = |x|$). In each of the following cases, prove f' is a OWF, or show a counterexample.

Part A $f'(x) := f(x) \| f(f(x))$.

Part B $f'(x) := x \oplus f(x)$.

Part C $f'(x) := f(x) \| f(\bar{x})$, where \bar{x} denote the bit-wise NOT operation.

Part D $f'(x) := f(G(x))$, where G is a PRG that $|G(s)| = |s| + 1$.

Part E $f'(x) := G(f(x))$, where G is a PRG that $|G(s)| = |s| + 1$.

Part F $f'(x) := f(x \| \underbrace{0 \dots 0}_{\log n \text{ many}})$, where $n = |x|$.

Part G $f'(x) := (f(x))_{1:(n-\log n)}$, where $n = |x|$. That is, $f'(x)$ outputs the first $n - \log(n)$ bits of $f(x)$.

Problem 3 (6pt) Hardness Amplification of Weak OWFs For simplicity, we consider length-preserving weak OWF. $f : \{0, 1\}^* \rightarrow \{0, 1\}^*$ is a length-preserving weak OWF, if $|f(x)| = |x|$ for any $x \in \{0, 1\}^*$, and there exists a polynomial q , such that for any PPT \mathcal{A} , for any sufficiently large n ,

$$\Pr_{\substack{x \xleftarrow{\$} \{0, 1\}^n \\ \hat{x} \leftarrow \mathcal{A}(f(x))}} \left[f(\hat{x}) = f(x) \right] \leq 1 - \frac{1}{q(n)}.$$

(Note the order of the quantifiers!)

Assume f is such a weak OWF. Define f' such that for $x_1, \dots, x_m \in \{0, 1\}^n$,

$$f'(x_1 \| \dots \| x_m) = f(x_1) \| \dots \| f(x_m)$$

where $m = m(n)$ is a polynomial on n . ($m(n)$ will be fixed later.)

We prove f' is a OWF by contradiction. Assume f' is not a OWF, then there exists PPT \mathcal{A}' , and polynomial p such that

$$\Pr_{\substack{x_1, \dots, x_m \xleftarrow{\$} \{0, 1\}^n \\ \hat{x}_1, \dots, \hat{x}_m \leftarrow \mathcal{A}'(f(x_1) \| \dots \| f(x_m))}} \left[f(\hat{x}_1) = f(x_1), \dots, f(\hat{x}_m) = f(x_m) \right] > \frac{1}{p(n)}$$

for infinitely many integer n .

Define \mathcal{A} as

$\mathcal{A}(y)$ let $n = |y|$, let $m = m(n)$
 sample $i \xleftarrow{\$} \{1, \dots, m\}$
 for all $j \neq i$, sample $x_j \xleftarrow{\$} \{0, 1\}^n$ and let $y_j = f(x_j)$
 let $y_i = y$
 call $\hat{x}_1 \| \dots \| \hat{x}_m \leftarrow \mathcal{A}'(y_1 \| \dots \| y_m)$
 if $f'(\hat{x}_1 \| \dots \| \hat{x}_m) = y_1 \| \dots \| y_m$,
 output \hat{x}_i

We say $x \in \{0, 1\}^n$ is “good” if \mathcal{A} inverts $f(x)$ with a good probability. Concretely, we define $x \in \{0, 1\}^n$ is “good” if and only if

$$\Pr_{\hat{x} \leftarrow \mathcal{A}(f(x))} [f(\hat{x}) = f(x)] \geq \frac{1}{r(n)}$$

for a polynomial $r(n)$. ($r(n)$ will be fixed later.) If x is not “good”, we say x is “bad”.

Part A Prove that

$$\begin{aligned} \Pr_{\substack{x_1, \dots, x_m \xleftarrow{\$} \{0, 1\}^n \\ \hat{x}_1, \dots, \hat{x}_m \leftarrow \mathcal{A}'(f(x_1) \| \dots \| f(x_m))}} [f(\hat{x}_1) = f(x_1), \dots, f(\hat{x}_m) = f(x_m)] \\ \leq \frac{m^2}{r(n)} + \left(\Pr_{x \leftarrow \{0, 1\}^n} [x \text{ is “good”}] \right)^m, \end{aligned}$$

for any sufficiently large n .

Part B Choose polynomials $m(n)$, $r(n)$ properly, so that

$$\Pr_{x \leftarrow \{0, 1\}^n} [x \text{ is “bad”}] \leq \frac{1}{2q(n)}$$

for infinitely many n . (Note that, you can let $r(n)$ depend on both $p(n)$ and $q(n)$; while $m(n)$ can depend on $q(n)$ and cannot depend on $p(n)$.)

Part C Define $\mathcal{A}_{\text{repeat}}$ as

$\mathcal{A}_{\text{repeat}}(y)$ let $n = |y|$
 repeat the following for $n \cdot r(n)$ times
 call $\hat{x} \leftarrow \mathcal{A}(y)$
 if $f(\hat{x}) = y$,
 output \hat{x}

Show that $\mathcal{A}_{\text{repeat}}$ violates our assumptions on f .

The contradiction rules out our assumption. So f' must be a OWF.

Problem 4 (6pt) A PRF $F : \{0, 1\}^\lambda \times \{0, 1\}^\lambda \rightarrow \{0, 1\}^\lambda$ is called a *puncturable PRF* if

- There is a p.p.t. algorithm **puncture**, which takes a key, an input, and outputs a “punctured key”.
- There is a p.p.t. algorithm **eval**, such that for any $x' \neq x$, we have $\text{eval}(k_{-x}, x') = F_k(x')$, where $k_{-x} \leftarrow \text{puncture}(k, x)$.
- If k, u are randomly sampled, $(k_{-x}, F_k(x))$ is indistinguishable from (k_{-x}, u) .

(More formally, consider a security game: the distinguisher \mathcal{D} chooses x ; the challenger samples random k, u , computes $k_{-x} \leftarrow \text{puncture}(k, x)$, and sends

- in case 0: $(k_{-x}, F_k(x))$, or
- in case 1: (k_{-x}, u)

to the distinguisher. We require that for any p.p.t. distinguisher \mathcal{D} , the distinguisher cannot tell which case it is with non-negligible advantage.)

Your task is to construct a puncturable PRF.

Remark: A puncturable PRF F is called a *private puncturable PRF* if k_{-x} does not reveal x . Until 2017, we didn’t know how to construct private puncturable PRF from standard assumptions.

Problem 5 (3pt, Exercise 4.8 from BS) Prove that, if $F : \{0, 1\}^n \times \{0, 1\}^n \rightarrow \{0, 1\}^n$ satisfies either of the following homomorphism properties, then F is not a PRF.

Part A $F(k, x \oplus c) = F(k, x) \oplus c$ for all $k, x, c \in \{0, 1\}^n$.

Part B $F(k \oplus c, x) = F(k, x) \oplus c$ for all $k, x, c \in \{0, 1\}^n$.

Part C $F(k_1 \oplus k_2, x) = F(k_1, x) \oplus F(k_2, x)$ for all $k_1, k_2, x \in \{0, 1\}^n$.

Remark: In contrast to Part C, under well-received assumptions, there exist PRFs satisfying $F(k_1 +_1 k_2, x) = F(k_1, x) +_2 F(k_2, x)$, where the key space and the output space are interpreted as carefully-chosen groups, and $+_1, +_2$ are the corresponding group operations.