

Cryptography Foundations

Solution Exercise 1

1.1 Variant of the IND-CPA Bit-Guessing Problem

- a) The idea is to construct a distinguisher D' for the bit-guessing problem $\llbracket S_t^{\text{ind}}; B' \rrbracket$. D' internally runs the assumed distinguisher D and emulates a view towards D that is identical to the interaction that D would have with the bit-guessing problem $\llbracket S_t^{\text{rrc}}; B \rrbracket$. This way, the decision of D will help D' to guess the bit B' . For this, D' first forwards every message queried by D during phase 2 to S_t^{ind} and returns the obtained encryptions back to D . Once D reaches phase 3 and provides a challenge message m , D' samples a uniformly random message \tilde{m} of length $|m|$ and provides as challenge the two messages m and \tilde{m} to S_t^{ind} . Upon receiving the encryption c (of either m or \tilde{m}), D' forwards this to D . Should D query any other message during phase 4, those will be again forwarded by D' , and the encryptions thereof given back to D . Finally, D' issues as its guess Z' the same bit Z that D issues in the emulated interaction. Since this emulation by D' perfectly mimics $\llbracket S_t^{\text{rrc}}; B \rrbracket$ towards D , we have $\Pr^{D'(S_t^{\text{ind}}, B')}[Z' = B'] = \Pr^{D(S_t^{\text{rrc}}, B)}[Z = B]$, and thus

$$\begin{aligned} \Lambda^{D'}(\llbracket S_t^{\text{ind}}; B' \rrbracket) &= 2 \cdot \Pr^{D'(S_t^{\text{ind}}, B')}[Z' = B'] - 1 \\ &= 2 \cdot \Pr^{D(S_t^{\text{rrc}}, B)}[Z = B] - 1 \\ &= \Lambda^D(\llbracket S_t^{\text{rrc}}; B \rrbracket). \end{aligned}$$

- b) Again, the idea is to construct a distinguisher D' for $\llbracket S_t^{\text{rrc}}; B' \rrbracket$ which internally runs the distinguisher D . Messages from D during the phases 2 and 4 are again simply forwarded. In the third phase, however, D provides two messages m_0 and m_1 , but D' can only input a single message to the bit-guessing problem. As a consequence, D' chooses a uniform random bit \tilde{B} , with the goal that D will guess \tilde{B} , and provides the message $m_{\tilde{B}}$ as challenge to S_t^{rrc} . Now, we consider two cases:

- If $B' = 0$, then challenge message $m_{\tilde{B}}$ is encrypted by S_t^{rrc} . Therefore, D essentially interacts with the bit-guessing problem $\llbracket S_t^{\text{ind}}; \tilde{B} \rrbracket$ and hence,

$$\Pr^{D'(S_t^{\text{rrc}}, B')}[Z = \tilde{B} \mid B' = 0] = \Pr^{D(S_t^{\text{ind}}, \tilde{B})}[Z = \tilde{B}].$$

- If $B' = 1$, then the challenge message is ignored by S_t^{rrc} and a uniformly random message is encrypted instead. Therefore, the interaction that D sees is independent of \tilde{B} and hence,

$$\Pr^{D'(S_t^{\text{rrc}}, B')}[Z = \tilde{B} \mid B' = 1] = \Pr^{D'(S_t^{\text{rrc}}, B')}[Z \neq \tilde{B} \mid B' = 1] = \frac{1}{2}.$$

So in short, if $B' = 0$, then D should be able to guess \tilde{B} , and if $B' = 1$ then not. Turned around, if the guess Z from D matches \tilde{B} , then D' can assume that $B' = 0$ and otherwise that $B' = 1$.

It remains to show that this strategy actually yields the correct advantage. To this end, let $Z' := Z \oplus \tilde{B}$ be the guess submitted by D' . We then have,

$$\begin{aligned}
\Lambda^{D'}(\llbracket S_t^{\text{rrc}}; B' \rrbracket) &= 2 \cdot \Pr^{D'(S_t^{\text{rrc}}, B')}[Z' = B'] - 1 \\
&= 2 \cdot \Pr^{D'(S_t^{\text{rrc}}, B')}[Z \oplus \tilde{B} = B'] - 1 \\
&= 2 \cdot \left(\Pr^{D'(S_t^{\text{rrc}}, B')}[Z = \tilde{B} \mid B' = 0] \cdot \frac{1}{2} \right. \\
&\quad \left. + \Pr^{D'(S_t^{\text{rrc}}, B')}[Z \neq \tilde{B} \mid B' = 1] \cdot \frac{1}{2} \right) - 1 \\
&= 2 \cdot \Pr^{D(S_t^{\text{ind}}, \tilde{B})}[Z = \tilde{B}] \cdot \frac{1}{2} + 2 \cdot \frac{1}{2} \cdot \frac{1}{2} - 1 \\
&= \Pr^{D(S_t^{\text{ind}}, B)}[Z = B] - \frac{1}{2} \\
&= \frac{1}{2} \cdot \Lambda^D(\llbracket S_t^{\text{ind}}; B \rrbracket).
\end{aligned}$$

1.2 On the Security of the One-Time Pad

First note that for any $g, h \in \mathbb{G}$ we have

$$\begin{aligned}
\Pr[U + X = g, X = h] &= \Pr[U = g + (-h), X = h] \\
&= \Pr[U = g + (-h)] \cdot \Pr[X = h] \\
&= \frac{1}{|\mathbb{G}|} \Pr[X = h],
\end{aligned} \tag{1}$$

where in the second step we used that U and X are independent, and in the last step that U is uniformly distributed. Then by the law of total probability, for any $g \in \mathbb{G}$ we have

$$\Pr[U + X = g] = \sum_{h \in \mathbb{G}} \Pr[U + X = g, X = h] \stackrel{(1)}{=} \frac{1}{|\mathbb{G}|} \sum_{h \in \mathbb{G}} \Pr[X = h] = \frac{1}{|\mathbb{G}|}. \tag{2}$$

Finally, putting the two equalities together, for any $g, h \in \mathbb{G}$ we have

$$\Pr[U + X = g, X = h] \stackrel{(1)}{=} \frac{1}{|\mathbb{G}|} \Pr[X = h] \stackrel{(2)}{=} \Pr[U + X = g] \cdot \Pr[X = h],$$

which is exactly the definition of $U + X$ and X being independent.

Note that the proof of security for the one-time pad over bitstrings of length n as introduced in the lecture notes is simply the instantiation of the above where U is the key, X is the message, \mathbb{G} is the set $\{0, 1\}^n$, the operation $+$ is the bit-wise XOR, the inverse operation is the identity function on $\{0, 1\}^n$, and the neutral element is 0^n (the bitstring consisting of n zeros).

1.3 Properties of the Distinguishing Advantage

- a) We divide the proof in two parts: first, we design a concrete distinguisher D^* , and show that it is indeed optimal; then we show that its advantage in distinguishing random variables X and Y is indeed equal to their statistical distance $\delta(X, Y)$.

In order to find the optimal distinguisher D^* , first note that it must be *deterministic*, since probabilistically mixing several deterministic strategies will result in a (weighted) average of the corresponding advantages. We therefore restrict our attention to a distinguisher D , that on input a realization of a random variable $X \in \mathcal{X}$ produces a binary guess $Z := D(X)$ according to a deterministic function $\mathcal{X} \rightarrow \{0, 1\}$. Now, consider the so-called *maximum-likelihood* distinguisher, which for $\mathcal{X}^* := \{x \in \mathcal{X} \mid \Pr_Y(x) \geq \Pr_X(x)\}$ is defined as

$$D^*(x) := \begin{cases} 1, & x \in \mathcal{X}^*, \\ 0, & x \notin \mathcal{X}^*. \end{cases}$$

We then have

$$\Delta^{D^*}(X, Y) = \Pr[D^*(Y) = 1] - \Pr[D^*(X) = 1] = \Pr[Y \in \mathcal{X}^*] - \Pr[X \in \mathcal{X}^*].$$

We next argue that D^* is indeed optimal. To this end, consider an arbitrary deterministic distinguisher D' , which is characterized by a set $\mathcal{X}' \subseteq \mathcal{X}$, i.e.,

$$D'(x) := \begin{cases} 1, & x \in \mathcal{X}', \\ 0, & x \notin \mathcal{X}'. \end{cases}$$

Observe, that by definition of \mathcal{X}^* , for any $x \in \mathcal{X}^*$ we have $P_Y(x) \geq P_X(x)$, and thus

$$\Pr[Y \in (\mathcal{X}^* \setminus \mathcal{X}')] \geq \Pr[X \in (\mathcal{X}^* \setminus \mathcal{X}')],$$

and analogously, since $P_Y(x) \leq P_X(x)$ for any $x \notin \mathcal{X}^*$,

$$\Pr[Y \in (\mathcal{X}' \setminus \mathcal{X}^*)] \leq \Pr[X \in (\mathcal{X}' \setminus \mathcal{X}^*)].$$

Hence, we obtain

$$\begin{aligned} \Delta^{D'}(X, Y) &= \Pr[Y \in \mathcal{X}'] - \Pr[X \in \mathcal{X}'] \\ &\leq \Pr[Y \in \mathcal{X}'] - \Pr[X \in \mathcal{X}'] + \underbrace{\Pr[Y \in (\mathcal{X}^* \setminus \mathcal{X}')] - \Pr[X \in (\mathcal{X}^* \setminus \mathcal{X}')]_{\geq 0}}_{\geq 0} \\ &\quad + \underbrace{\Pr[X \in (\mathcal{X}' \setminus \mathcal{X}^*)] - \Pr[Y \in (\mathcal{X}' \setminus \mathcal{X}^*)]}_{\geq 0} \\ &= (\Pr[Y \in \mathcal{X}'] + \Pr[Y \in (\mathcal{X}^* \setminus \mathcal{X}')] - \Pr[Y \in (\mathcal{X}' \setminus \mathcal{X}^*)]) \\ &\quad - (\Pr[X \in \mathcal{X}'] + \Pr[X \in (\mathcal{X}^* \setminus \mathcal{X}')] - \Pr[X \in (\mathcal{X}' \setminus \mathcal{X}^*)]) \\ &= \Pr[Y \in \mathcal{X}^*] - \Pr[X \in \mathcal{X}^*] \\ &= \Delta^{D^*}(X, Y). \end{aligned}$$

Overall, we have shown so far that

$$\Delta(X, Y) = \sup_{D: \mathcal{X} \rightarrow \{0,1\}} \Delta^D(X, Y) = \Delta^{D^*}(X, Y).$$

In the second part of the proof, we show that indeed $\Delta^{D^*}(X, Y) = \delta(X, Y)$. We have

$$\begin{aligned} \delta(X, Y) &= \frac{1}{2} \sum_{x \in \mathcal{X}} |P_X(x) - P_Y(x)| \\ &= \frac{1}{2} \sum_{x \in \mathcal{X}^*} (P_Y(x) - P_X(x)) + \frac{1}{2} \sum_{x \in \mathcal{X} \setminus \mathcal{X}^*} (P_X(x) - P_Y(x)) \\ &= \frac{1}{2} (\Pr[Y \in \mathcal{X}^*] - \Pr[X \in \mathcal{X}^*]) + \frac{1}{2} (1 - \Pr[X \in \mathcal{X}^*] - (1 - \Pr[Y \in \mathcal{X}^*])) \\ &= \frac{1}{2} (\Pr[Y \in \mathcal{X}^*] - \Pr[X \in \mathcal{X}^*]) + \frac{1}{2} (\Pr[Y \in \mathcal{X}^*] - \Pr[X \in \mathcal{X}^*]) \\ &= \Pr[Y \in \mathcal{X}^*] - \Pr[X \in \mathcal{X}^*] \\ &= \Delta^{D^*}(X, Y). \end{aligned}$$

Therefore, as claimed, we have shown that $\Delta(X, Y) = \delta(X, Y)$.

- b) First we define the new distinguisher D' which internally uses the distinguisher D : it simply provides D with access to S and gives D a uniformly sampled bit U . Once D outputs its decision bit Z , D' outputs its guess bit $Z' := Z \oplus U$, that is, D' outputs 0 if and only if $Z = U$ (see Figure 1). Then, using the law of total probability on the partition defined by the (disjoint) events $U = B$ and $U \neq B$ (each clearly occurring with probability $\frac{1}{2}$), we have

$$\begin{aligned}
\Lambda^{D'}(\llbracket S; B \rrbracket) &= 2 \cdot \Pr^{D'(S,B)}[Z' = B] - 1 \\
&= 2 \cdot \Pr^{D'(S,B)}[Z \oplus U = B] - 1 \\
&= 2 \cdot \Pr^{D'(S,B)}[Z = U \oplus B \mid U = B] \cdot \frac{1}{2} \\
&\quad + 2 \cdot \Pr^{D'(S,B)}[Z = U \oplus B \mid U \neq B] \cdot \frac{1}{2} - 1 \\
&= \Pr^{D[S,B]}[Z = 0] + \Pr^{D[S,\bar{B}]}[Z = 1] - 1 \\
&= \Pr^{D[S,\bar{B}]}[Z = 1] - \Pr^{D[S,B]}[Z = 1] \\
&= \Delta^D([S, B], [S, \bar{B}]).
\end{aligned}$$

Recall from the lecture that we consider here the problem of distinguishing pairs $[S, B]$ and $[S, U]$ where $[S, U]$ stands for a pair that is sampled like $[S, B]$, but where the distinguisher does not see the bit B (correlated with S), but an independent and uniformly distributed bit U . We again apply the law of total probability on the event whether B and U are identical and conclude:

$$\begin{aligned}
\Delta^D([S, B], [S, U]) &= \Pr^{D[S,U]}[Z = 1] - \Pr^{D[S,B]}[Z = 1] \\
&= \Pr^{D[S,U]}[Z = 1 \mid U = B] \cdot \frac{1}{2} \\
&\quad + \Pr^{D[S,U]}[Z = 1 \mid U \neq B] \cdot \frac{1}{2} - \Pr^{D[S,B]}[Z = 1] \\
&= \frac{1}{2} \cdot \Pr^{D[S,B]}[Z = 1] \\
&\quad + \frac{1}{2} \cdot \Pr^{D[S,\bar{B}]}[Z = 1] - \Pr^{D[S,B]}[Z = 1] \\
&= \frac{1}{2} \cdot \Pr^{D[S,\bar{B}]}[Z = 1] - \frac{1}{2} \cdot \Pr^{D[S,B]}[Z = 1] \\
&= \frac{1}{2} \cdot \Delta^D([S, B], [S, \bar{B}]).
\end{aligned}$$

Finally, putting the two equalities together, we have

$$\Delta^D([S, B], [S, U]) = \frac{1}{2} \cdot \Lambda^{D'}(\llbracket S; B \rrbracket).$$

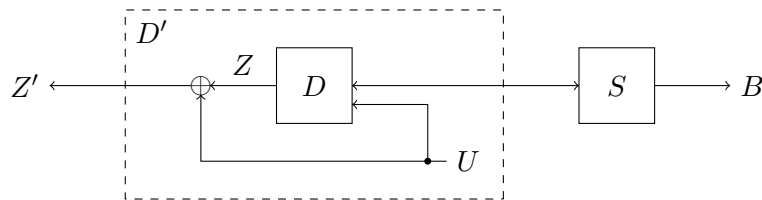


Figure 1: The mapping $D \mapsto D'$.