

# Advanced Cryptography — Final Exam

## Solution

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- duration: 3h
- any document allowed
- a pocket calculator is allowed
- communication devices are not allowed
- the exam invigilators will **not** answer any technical question during the exam
- readability and style of writing will be part of the grade

*The exam grade follows a linear scale in which each question has the same weight.*

### 1 $\Sigma$ Protocol in an Group of Exponent 2

Given an integer  $s$ , we consider an Abelian group  $G$  (with multiplicative notations) such that for all  $x \in G$ , we have  $x^2 = 1$ . We assume that there are deterministic polynomial time algorithms to compute the order  $n$  of  $G$ , to multiply and to compare two group elements. More precisely, given  $x$  and  $y$  we can compute  $xy$  and say whether  $x = y$ . For  $x = (x_1, \dots, x_m, y) \in G^{m+1}$  and  $w = (w_1, \dots, w_m) \in \{0, 1\}^m$ , we consider the following relation:

$$R(x; w) \iff y = x_1^{w_1} \cdots x_m^{w_m}$$

We consider the following protocol:

Prover	Verifier
$w$	$x$
pick $r = (r_1, \dots, r_m) \in_U \{0, 1\}^m$	
$a = x_1^{r_1} \cdots x_m^{r_m} \xrightarrow{a}$	
$\xleftarrow{e}$ pick $e \in_U \{0, 1\}$	
$z = r \oplus ew \xrightarrow{z}$ check $x_1^{z_1} \cdots x_m^{z_m} = ay^e$	

where  $\oplus$  denotes the XOR operation (exclusive OR) between two bits and  $\in_U$  denotes a random selection with uniform distribution and fresh coins.

**Q.1** Following the terminology of  $\Sigma$  protocols, show that the above protocol has *special soundness*.

Given  $x$ ,  $a$ , and the answers  $z_0$  and  $z_1$  to  $e = 0$  and  $e = 1$ , we compute  $w = z_0 \oplus z_1$ . We obtain  $y = x_1^{w_1} \cdots x_m^{w_m}$ . So, we have an extractor for special soundness.

**Q.2** Following the terminology of  $\Sigma$  protocols, show that the above protocol is *special HVZK* (special honest verifier zero-knowledge).

We first observe that in the proposed protocol,  $z$  is uniformly distributed when  $e = 0$  (this is just  $r = (r_1, \dots, r_m)$ ). When  $e = 1$ , this is the XOR between  $r$  and  $w$ . Since  $r$  is uniform and independent from  $w$ ,  $z$  is uniformly distributed in this case as well.

Next, we observe that we can always compute  $a$  from  $z$ ,  $e$ , and  $y$ .

Given  $x$  and  $e$ , we pick  $z \in \{0, 1\}^m$  uniformly at random and deduce  $a$  from the value  $y$  inside  $x$ . Clearly, the obtained  $(a, e, z)$  has a correct distribution with  $e$  imposed. So, we have a perfect simulator for special HVZK.

**Q.3** Show that the proposed protocol is a  $\Sigma$  protocol.

The protocol follows the structure of  $\Sigma$  protocols:

- the verifier is polynomially bounded;
- it has 3 moves initiated by the prover;
- the challenge message from the verifier is selected at random from a set independently from the first message from the prover;
- the acceptance condition only depends on  $(x, a, e, z)$ .

Furthermore, we have prover special soundness and special HVZK. So, we have a  $\Sigma$  protocol.

This exercise was quite simple. Nevertheless, we spotted two frequent mistakes.

- The extractor and the simulator for  $\Sigma$  protocols work with transcripts instead of views.
- For the extractor, we cannot assume that the answers  $z_0$  and  $z_1$  to  $e = 0$  and  $e = 1$  are  $z_0 = r$  and  $z_1 = r \oplus w$  and simply answer that  $z_0 \oplus z_1$  is the witness. Instead, we should show that  $z_0 \oplus z_1$  is a valid witness from the properties of  $z_0$  and  $z_1$ : given that  $x^{z_e} = ay^e$ , then  $x^{z_1 - z_0} = y$ .

Also: for the simulator, it is important to show that the output has the same distribution as a natural transcript.

## 2 On Generator Generation in Diffie-Hellman Problems

In the Computational Diffie-Hellman (CDH) Problem and the Decisional Diffie-Hellman (DDH) Problem, there is a security parameter (integer)  $s$  as input to a probabilistic polynomial-time (PPT) algorithm  $\text{Gen}(1^s) \rightarrow (q, \text{parms}, g)$  to generate a prime number  $q$  together with an element  $g$  and some parameters  $\text{parms}$ . The values  $q$  and  $\text{parms}$  define a group  $G = (q, \text{parms})$  of order  $q$  in which  $g$  is a generator. We denote  $G = \langle g \rangle$  and  $x \in G$  to mean that  $g$  generates  $G$  and  $x$  belongs to  $G$ . We assume multiplicative notations in the group. We assume we have two deterministic polynomial-time algorithms  $\text{MUL}$  and  $\text{EQ}$  such that for all  $x, y \in G$ , we have  $\text{MUL}(G, x, y) = xy$  and  $\text{EQ}(G, x, y) = 1_{x=y}$ .

- Q.1** Show that we can design deterministic polynomial-time algorithms  $\text{UN}$ ,  $\text{INV}$ , and  $\text{POW}$  such that for all  $x \in G$  and  $e \in \mathbf{Z}$ , we have  $\text{UN}(G, x) = 1$ ,  $\text{INV}(G, x) = x^{-1}$ , and  $\text{POW}(G, x, e) = x^e$ .

CAUTION: be careful with the  $e = 0$  and  $e < 0$  cases.

*We define  $\text{POW}(G, x, e)$  for  $e$  positive by using the square-and-multiply algorithm using  $\text{MUL}$ .*

*Then, we can define  $\text{UN}(G, x) = \text{POW}(G, x, q)$  as  $x^q = 1$ . Note that it is necessary to have one element group element  $x$  to compute 1. This is the case when we have, for instance, a generator.*

*We can also define  $\text{INV}(G, x) = \text{POW}(G, x, q - 1)$  as  $x^{q-1} \times x = 1$ .*

*Finally, we can define  $\text{POW}(G, x, 0) = \text{UN}(G, x)$  and  $\text{POW}(G, x, e) = \text{POW}(G, x, e \bmod q)$  for any  $e \in \mathbf{Z}$ .*

*COMMENT after correction: we cannot just say that  $\text{UN}(G, x)$  just answer 1 as this "1" is not necessarily the integer 1 that we know. Here, we want to output the neutral element from the group and we need to reconstruct it from the given parameters.*

*In the correction, we have seen several times some non-polynomial solutions. For instance, compute  $\text{POW}(G, x, e)$  using  $e - 1$  multiplications or finding  $\text{INV}(G, x)$  by exhaustive search. These answers are not acceptable.*

In this exercise, we look at the influence on the  $g$  generation by  $\text{Gen}$  in the  $\text{Gen-CDH}$  and  $\text{Gen-DDH}$  problems. We assume a first PPT algorithm  $\text{Setup}(1^s) \rightarrow (q, \text{parms})$  to generate the group  $G = (q, \text{parms})$  and we assume that from  $G$  we can extract a generator  $g = \text{Generator}(G)$  using a deterministic polynomial-time algorithm  $\text{Generator}$ . We define two generating algorithms.

$\text{GenFixed}(1^s; \rho)$ :

- 1: run  $\text{Setup}(1^s; \rho) \rightarrow G = (q, \text{parms})$
- 2: run  $g = \text{Generator}(G)$
- 3: **output**  $(q, \text{parms}, g)$

We call  $\text{GenFixed}$  the setup with fixed generator  $g$ .

$\text{GenRand}(1^s; \rho)$ :

- 1: split  $\rho$  into two independent sequences  $\rho_1$  and  $\rho_2$
- 2: run  $\text{Setup}(1^s; \rho_1) \rightarrow G = (q, \text{parms})$
- 3: run  $g = \text{Generator}(G)$
- 4: generate  $a \in \mathbf{Z}_q^*$  with uniform distribution from  $\rho_2$
- 5: set  $h = \text{POW}(G, g, a)$
- 6: **output**  $(q, \text{parms}, h)$

We call  $\text{GenRand}$  *the setup with random generator*  $h$ .

The DDH problem specifies two distributions with parameter  $s$ :

Source  $S_0(\text{Gen})$ :

- 1: pick a large enough sequence of independent fair coin flips  $\rho$
- 2: run  $\text{Gen}(1^s; \rho) \rightarrow (q, \text{parms}, g)$
- 3: pick  $x, y \in \mathbf{Z}_q$  with uniform distribution ( $x, y$ , and  $\rho$  are independent)
- 4: set  $X = g^x, Y = g^y, Z = g^{xy}$
- 5: **output**  $(q, \text{parms}, g, X, Y, Z)$

Source  $S_1(\text{Gen})$ :

- 1: pick a large enough sequence of independent fair coin flips  $\rho$
- 2: run  $\text{Gen}(1^s; \rho) \rightarrow (q, \text{parms}, g)$
- 3: pick  $x, y, z \in \mathbf{Z}_q$  with uniform distribution ( $x, y, z$ , and  $\rho$  are independent)
- 4: set  $X = g^x, Y = g^y, Z = g^z$
- 5: **output**  $(q, \text{parms}, g, X, Y, Z)$

The Gen-DDH problem consists of distinguishing  $S_0(\text{Gen})$  and  $S_1(\text{Gen})$ . We stress that the DDH problem is relative to  $\text{Gen}$ ; this is why we call it *the Gen-DDH problem*. We define the advantage

$$\text{Adv}^{\text{Gen-DDH}}(\mathcal{A}) = \Pr[\mathcal{A}(S_0(\text{Gen})) = 1] - \Pr[\mathcal{A}(S_1(\text{Gen})) = 1]$$

The Gen-DDH is hard if and only if for all PPT distinguisher  $\mathcal{A}$ ,  $\text{Adv}^{\text{Gen-DDH}}(\mathcal{A})$  is negligible.

**Q.2** Given a GenRand-DDH distinguisher  $\mathcal{A}$ , construct a GenFixed-DDH distinguisher  $\mathcal{B}$  with the same advantage and a similar complexity.

We define  $\mathcal{B}$  as follows:

$\mathcal{B}(q, \text{params}, g, X, Y, Z)$ :

- 1: set  $G = (q, \text{params})$
- 2: pick  $a \in \mathbf{Z}_q^*$  with uniform distribution
- 3: set  $h = \text{POW}(G, g, a)$ ,  $X' = \text{POW}(G, X, a)$ ,  $Y' = \text{POW}(G, Y, a)$ ,  $Z' = \text{POW}(G, Z, a)$
- 4: run  $b = \mathcal{A}(G, h, X', Y', Z')$
- 5: **output**  $b$

Clearly, with the sources  $S_0(\text{GenFixed})$  and  $S_1(\text{GenFixed})$ , the input  $(q, \text{params}, g)$  to  $\mathcal{B}$  follows the distribution of  $\text{GenFixed}$ . So,  $(q, \text{params}, h)$  is identical to the distribution induced by  $\text{GenRand}$ . Hence, the input  $(q, \text{params}, h, X, Y, Z)$  of  $\mathcal{A}$  in Step 4 follows either source  $S_0(\text{GenRand})$  or  $S_1(\text{GenRand})$ . Therefore, the advantage of  $\mathcal{B}$  equals the advantage of  $\mathcal{A}$ .

The complexity overhead in  $\mathcal{B}$  corresponds to Step 2 and Step 3 which are polynomial.

*COMMENT after correction: some students just answered that  $\mathcal{A}(G, g, X, Y, Z)$  will work as it is a special case. We only know that  $\mathcal{A}$  works for random inputs so it is not guaranteed that it works for the special input  $g$ . This is the main problem in this exercise which requires to randomize inputs. We did not accept these answers. The same problem occurred in the next questions.*

**Q.3** Show that if the  $\text{GenFixed-DDH}$  is hard, then the  $\text{GenRand-DDH}$  problem is hard.

*Any PPT  $\text{GenRand-DDH}$  distinguisher has the same advantage of some PPT  $\text{GenFixed-DDH}$  distinguisher. If the  $\text{GenFixed-DDH}$  is hard, it must be negligible. So, the  $\text{GenRand-DDH}$  problem is hard.*

Unfortunately, we have no implication in the other direction for the DDH problem, but there is for the CDH problem.

The computational Diffie-Hellman (CDH) problem has instances defined by the following source:

Source  $S(\text{Gen})$ :

- 1: pick a large enough sequence of independent fair coin flips  $\rho$
- 2: run  $\text{Gen}(1^s; \rho) \rightarrow (q, \text{params}, g)$
- 3: pick  $x, y \in \mathbf{Z}_q^*$  with uniform distribution ( $x$ ,  $y$ , and  $\rho$  are independent)
- 4: set  $X = g^x$ ,  $Y = g^y$  {the solution to the problem is  $g^{xy}$ }
- 5: **output**  $(q, \text{params}, g, X, Y)$

Given a CDH solver  $\mathcal{A}$ , we define

$$\text{Succ}^{\text{Gen-CDH}}(\mathcal{A}) = \Pr[\mathcal{A}(S(\text{Gen})) = g^{xy}]$$

The  $\text{Gen-CDH}$  is hard if and only if for all PPT solver  $\mathcal{A}$ ,  $\text{Succ}^{\text{Gen-CDH}}(\mathcal{A})$  is negligible.

**Q.4** Given a GenRand-CDH solver  $\mathcal{A}$ , construct a GenFixed-CDH solver  $\mathcal{B}$  with similar complexity and

$$\text{Succ}^{\text{GenRand-CDH}}(\mathcal{A}) = \text{Succ}^{\text{GenFixed-CDH}}(\mathcal{B})$$

We define  $\mathcal{B}$  as follows:

$\mathcal{B}(q, \text{params}, g, X, Y)$ :

- 1: set  $G = (q, \text{params})$
- 2: pick  $a \in \mathbf{Z}_q^*$  with uniform distribution
- 3: set  $g' = \text{POW}(G, g, a)$
- 4: set  $X' = \text{POW}(G, X, a)$
- 5: set  $Y' = \text{POW}(G, Y, a)$
- 6: run  $Z' = \mathcal{A}(G, g', X', Y')$
- 7: set  $Z = \text{POW}(G, Z', 1/a \bmod q)$
- 8: **output**  $Z$

Just like for DDH, we show that  $(q, \text{params}, g', X', Y')$  follows the distribution of  $S(\text{GenRand})$  with solution  $Z^a$ , where  $Z$  is the solution to the GenFixed-CDH problem. So, the probability of success is preserved. Similarly, the complexity overhead is small.

COMMENT after correction: we have seen several pearls for which we gave a penalty:  $g^{xy} = g^{xya^2} g^{a^{-2}}$ ,  $(g^{axy})^{-a} = g^{xy}$ , or  $(g^x)^u = g^{u^x}$  !!!

**Q.5** Given a GenFixed-CDH solver  $\mathcal{A}$ , we denote

$$\varepsilon_\rho = \Pr[\mathcal{A}(S(\text{GenFixed})) = g^{xy} | \rho]$$

the probability of success when  $\rho$  in  $S$  is fixed. So, GenFixed always returns the same group and generator (due to  $\rho$  being fixed). Only  $x$ ,  $y$ , and possible coins used by  $\mathcal{A}$  remain random.

Given a GenFixed-CDH solver  $\mathcal{A}$ , show that we can construct an algorithm Mu such that for any integer  $x$  and  $y$  (i.e., not only for random  $x$  and  $y$ ) and any  $\rho$ , we have

$$\text{Mu}(q, \text{params}, g, g^x, g^y) = g^{xy}$$

for  $\text{GenFixed}(1^s; \rho) \rightarrow (q, \text{params}, g)$  with probability at least  $\varepsilon_\rho$  over the distribution of  $x$ ,  $y$ , and possible coins by  $\mathcal{A}$ .

We just have to randomize  $g^x$  and  $g^y$ . There is a particular case when one of the two is equal to 1. We can check it using EQ. In this case, the algorithm should answer 1 and it is correct with probability 1.

In other cases, we pick  $a$  and  $b$  in  $\mathbf{Z}_q^*$  uniformly at random and run  $\mathcal{A}(q, \text{params}, g, g^{ax}, g^{by}) = Z$  then output  $Z^{\frac{1}{ab}}$ . Now,  $g^{ax}$  and  $g^{by}$  are independent and uniformly distributed in  $G - \{1\}$ . So they are correctly distributed and  $\mathcal{A}(q, \text{params}, g, g^{ax}, g^{by}) = g^{abxy}$  with probability  $\text{Succ}^{\text{GenFixed-CDH}}(\mathcal{A})$ . So,  $Z^{\frac{1}{ab}} = g^{xy}$  with this probability.

**Mu**( $q, \text{params}, g, X, Y$ ): {say  $X = g^x$  and  $Y = g^y$ }

1: set  $G = (q, \text{params})$  and  $u = \text{UN}(G, g)$

2: **if** EQ( $G, X, u$ ) = 1 or EQ( $G, Y, u$ ) = 1 **then**

3:     **output**  $u$

4: **end if**

5: pick  $a, b \in \mathbf{Z}_q^*$  with uniform distribution

6: compute  $X^a = \text{POW}(G, X, a)$  and  $Y^b = \text{POW}(G, Y, b)$

7: run  $Z = \mathcal{A}(q, \text{params}, g, X^a, Y^b)$  {we should have  $Z = g^{abxy}$ }

8: **output**  $Z^{\frac{1}{ab}}$

**Q.6** Show that we can construct an algorithm **In** such that for any integer  $x$  and any  $\rho$ , we have

$$\text{In}(q, \text{params}, g, g^x) = g^{\frac{1}{x}}$$

for  $\text{GenFixed}(1^s; \rho) \rightarrow (q, \text{params}, g)$  with probability at least  $\varepsilon_\rho^w$  for  $w = \mathcal{O}(\log q)$ .

We have  $g^{\frac{1}{x}} = g^{x^{q-2}}$ .

Using **Mu** from the previous question we compute  $g^{x^{q-2}}$  with a square-and-multiply algorithm. If  $w$  denotes the number of squares or multiplications to perform, we have  $w = \mathcal{O}(\log q)$ .

Once  $\rho$  is fixed, all  $w$  operations are independent. Since they each succeed with probability  $\varepsilon_\rho$ , the overall process succeeds with probability at least  $\varepsilon_\rho^w$ .

COMMENT after correction: several times we say incorrect answers such as  $\text{In}(g, g^x) = \text{Mu}(g^x, g^x, g)$  of other answers where the generator given as input to  $g$  was not the fixed one  $g$ . However, we constructed  $g$  to work for this very particular fixed  $g$  and our main problem is to construct one which works on average for a random  $g$ .

**Q.7** Given a GenFixed-CDH solver  $\mathcal{A}$ , construct a GenRand-CDH solver  $\mathcal{B}$  with similar complexity and

$$\text{Succ}^{\text{GenRand-CDH}}(\mathcal{B}) \geq (\text{Succ}^{\text{GenFixed-CDH}}(\mathcal{A}))^{\mathcal{O}(\log q)}$$

Let  $\text{Mu}$  and  $\text{In}$  be the algorithms from the previous questions.

$\mathcal{B}(q, \text{params}, h, X, Y)$ : {say  $h = g^a$ ,  $X = g^{ax}$ ,  $Y = g^{ay}$ }

1: set  $G = (q, \text{params})$

2: run  $g = \text{Generator}(G)$

3: run  $h' = \text{In}(G, g, h)$  {we should have  $h' = g^{\frac{1}{a}}$ }

4: run  $A = \text{Mu}(G, g, X, Y)$  {we should have  $A = g^{a^2xy}$ }

5: run  $B = \text{Mu}(G, g, h', A)$  {we should have  $B = g^{axy}$ }

6: run  $Z = \text{Mu}(G, g, h', B)$  {we should have  $Z = g^{xy}$ }

7: **output**  $Z$

Clearly, this works with probability at least  $E_\rho(\varepsilon_\rho^{w+3})$ . By using the Jensen inequality, we obtain  $E_\rho(\varepsilon_\rho^{w+3}) \geq E_\rho(\varepsilon_\rho)^{w+3} = (\text{Succ}^{\text{GenFixed-CDH}}(\mathcal{A}))^{w+3}$ . Then, we use  $w = \mathcal{O}(\log q)$ .

We could also have computed  $h' = g^{\frac{1}{a^2}}$  directly (by  $g^{a^{q-2}}$ ) and obtained  $B = g^{xy}$  to save one  $\text{Mu}$  operation.

COMMENT after correction: in the copies given during the exam, the question was given with

$$\text{Succ}^{\text{GenFixed-CDH}}(\mathcal{B}) \geq (\text{Succ}^{\text{GenRand-CDH}}(\mathcal{A}))^{\mathcal{O}(\log q)}$$

The mistake was corrected during the exam and written on the black board.

**Q.8** Show that  $\text{GenFixed-CDH}$  is hard if and only if  $\text{GenRand-CDH}$  is hard.

If  $\text{GenFixed-CDH}$  is hard, any  $\text{GenRand-CDH}$  solver has a probability of success equal to the one of a  $\text{GenFixed-CDH}$  solver (due to Q.4). So, it must be negligible. Hence,  $\text{GenRand-CDH}$  is hard.

If  $\text{GenRand-CDH}$  is hard, any  $\text{GenFixed-CDH}$  solver has a probability of success bounded by the one of a  $\text{GenRand-CDH}$  solver raised to the power  $\frac{1}{\mathcal{O}(\log q)}$  (due to Q.7). Since  $\text{GenFixed}$  returns  $q$ ,  $\log q$  must be polynomially bounded. So, this power must be negligible. Hence,  $\text{GenFixed-CDH}$  is hard.

COMMENT after correction: We gave 2/3 of the grade if one direction was missing. We gave a penalty when nothing was said about the negligible advantage. We gave no points to answers such as “if we can easily solve  $\text{GenRand}$  then we can easily solve  $\text{GenFixed}$  as it is a particular case”.



### 3 Equivalent PRF Notions

We consider a function family  $f_s$  which depends on a security parameter  $s$ . Given  $s$ , the function  $f_s$  takes a key  $k \in \mathcal{K}_s$  and an input  $x \in \mathcal{X}_s$ . It produces an output  $y = f_s(k, x) \in \mathcal{Y}_s$ . To have lighter notations, from now on the subscript  $s$  is omitted. We further write the input  $k$  of  $f$  as a subscript to write  $f_k(x) = f(k, x)$ . We say that the function family  $f$  is a pseudorandom function (PRF) if it can be computed in polynomial time (in terms of  $s$ ) and if for every probabilistic polynomial-time (PPT) algorithm  $\mathcal{A}$ , the function  $\text{Adv}_{\mathcal{A}}^{\text{PRF}}$  (this is a function in terms of  $s$ ) is a negligible function where

$$\text{Adv}_{\mathcal{A}}^{\text{PRF}} = \Pr[\Gamma_0^{\text{PRF}}(\mathcal{A}) = 1] - \Pr[\Gamma_1^{\text{PRF}}(\mathcal{A}) = 1]$$

and  $\Gamma_b^{\text{PRF}}(\mathcal{A})$  is defined with a bit  $b$  as follows:

**Game**  $\Gamma_b^{\text{PRF}}(\mathcal{A})$ :

- 1: pick  $s \in \mathcal{K}$  at random
- 2: set  $\rho$  to a long enough sequence of random coins
- 3: set  $i = 1$
- 4:  $(q, x_i) \leftarrow \mathcal{A}(\rho)$
- 5: **while**  $q \neq \text{final}$  **do**
- 6:   **if**  $x_i \in \{x_1, \dots, x_{i-1}\}$  **then**
- 7:     abort {it is not allowed to repeat a query}
- 8:   **end if**
- 9:   **if**  $b = 0$  **then**
- 10:     set  $y_i = f_s(x_i)$
- 11:   **else**
- 12:     set  $y_i \in \mathcal{Y}$  at random
- 13:   **end if**
- 14:    $i \leftarrow i + 1$
- 15:    $(q, x_i) \leftarrow \mathcal{A}(y_1, \dots, y_{i-1}; \rho)$
- 16: **end while**
- 17: **output**  $x_i$

Here,  $\mathcal{A}$  returns a pair  $(q, x)$ . The string  $q$  is either **query** or **final**. If  $q = \text{query}$ , it means that  $\mathcal{A}$  wants to query  $f_s(x)$  and continue. If  $q = \text{final}$ , it means that  $\mathcal{A}$  is done and returning a bit  $x$  as a final output.

We recall that a function  $\text{Adv}(s)$  is negligible is for all  $c > 0$ , we have  $\text{Adv}(s) = \mathcal{O}(s^{-c})$  when  $s \rightarrow +\infty$ .

In this exercise, we consider another notion defined by the following game:

**Game**  $\Gamma_b^{\text{prePRF}}(\mathcal{A})$ :

- 1: pick  $s \in \mathcal{K}$  at random
- 2: set  $\rho$  to a long enough sequence of random coins
- 3: set  $i = 1$  and unset **flag**
- 4:  $(q, x_i) \leftarrow \mathcal{A}(\rho)$
- 5: **while**  $q \neq \text{final}$  **do**

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6:  if  $x_i \in \{x_1, \dots, x_{i-1}\}$  then
7:    abort {it is not allowed to repeat a query}
8:  end if
9:  if  $q = \text{challenge}$  and flag is set then
10:    abort {it is not allowed to make two challenges}
11:  end if
12:  if  $q = \text{challenge}$  then
13:    set flag { $\mathcal{A}$  is making a challenge}
14:  end if
15:  if  $q = \text{challenge}$  and  $b = 1$  then
16:    set  $y_i \in \mathcal{Y}$  at random
17:  else
18:    set  $y_i = f_s(x_i)$ 
19:  end if
20:   $i \leftarrow i + 1$ 
21:   $(q, x_i) \leftarrow \mathcal{A}(y_1, \dots, y_{i-1}; \rho)$ 
22: end while{we must have  $q = \text{final}$ }
23: output  $x_i$ 

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Essentially,  $\mathcal{A}$  always plays with  $f$  with  $q = \text{query}$  and at some point uses only once a special  $q = \text{challenge}$ . For this “challenge”, the response which is returned to him is  $f_s(x)$  if  $b = 0$  or something random if  $b = 1$ . An equivalent way consists of saying that  $\mathcal{A}^{f_k(\cdot)}$  works in two phases, playing with a  $f_k(\cdot)$  oracle. In between the two phases, it makes a challenge which is answer by  $f_k(\cdot)$  or at random.

We define

$$\text{Adv}_{\mathcal{A}}^{\text{prePRF}} = \Pr[\Gamma_0^{\text{prePRF}}(\mathcal{A}) = 1] - \Pr[\Gamma_1^{\text{prePRF}}(\mathcal{A}) = 1]$$

and we say that the function family  $f$  is a prePRF if it can be computed in polynomial time (in terms of  $s$ ) and if for every PPT algorithm  $\mathcal{A}$ ,  $\text{Adv}_{\mathcal{A}}^{\text{prePRF}}$  is negligible.

The objective of this exercise is to show that PRF and prePRF are equivalent security notions.

**Q.1** Given a prePRF adversary  $\mathcal{A}$  and a bit  $b$ , we construct a PRF adversary  $\mathcal{B}_b$  as follows:

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 $\mathcal{B}_b(y_1, \dots, y_{i-1}; \rho)$ :
1: if  $i = 1$  then
2:   set  $\text{seq}_x$  and  $\text{seq}_y$  to the empty sequence {first execution of  $\mathcal{B}_b$ }
3: else
4:   set  $\text{seq}_y \leftarrow (\text{seq}_y, y_{i-1})$  { $y_{i-1}$  is the answer to the previous query}
5: end if
6: run  $(q, x) = \mathcal{A}(\text{seq}_y; \rho)$ 
7: if  $x \in \text{seq}_x$  then
8:   abort {it is not allowed to repeat a query}
9: end if
10: set  $\text{seq}_x \leftarrow (\text{seq}_x, x)$  {insert  $x$  in the list of queries}

```

11: **if**  $q = \text{challenge}$  and  $b = 1$  **then**  
 12:   set  $y \in \mathcal{Y}$  at random  
 13:   set  $\text{seq}_y \leftarrow (\text{seq}_y, y)$   
 14:   run  $(q, x) = \mathcal{A}(\text{seq}_y; \rho)$   
 15:   **if**  $x \in \text{seq}_x$  **then**  
 16:     abort {it is not allowed to repeat a query}  
 17:   **end if**  
 18:   set  $\text{seq}_x \leftarrow (\text{seq}_x, x)$  {insert  $x$  in the list of queries}  
 19: **end if**  
 20: **output**  $(q, x)$

So,  $\mathcal{B}$  simulates  $\mathcal{A}$  and simulates the answer to random for the  $q = \text{challenge}$  and  $b = 1$  case.

Show that

$$\begin{aligned}
 \Pr[\Gamma_0^{\text{prePRF}}(\mathcal{A}) = 1] &= \Pr[\Gamma_0^{\text{PRF}}(\mathcal{B}_0) = 1] \\
 \Pr[\Gamma_1^{\text{prePRF}}(\mathcal{A}) = 1] &= \Pr[\Gamma_0^{\text{PRF}}(\mathcal{B}_1) = 1] \\
 \Pr[\Gamma_1^{\text{PRF}}(\mathcal{B}_0) = 1] &= \Pr[\Gamma_1^{\text{PRF}}(\mathcal{B}_1) = 1]
 \end{aligned}$$

Essentially,  $\mathcal{B}_b$  simulates  $\mathcal{A}$  who plays the prePRF game but skips a response to a challenge set to random. Clearly, the game  $\Gamma_0^{\text{prePRF}}(\mathcal{A})$  is fully simulated by  $\Gamma_0^{\text{PRF}}(\mathcal{B}_0)$ . Similarly,  $\Gamma_1^{\text{prePRF}}(\mathcal{A})$  is fully simulated by  $\Gamma_0^{\text{PRF}}(\mathcal{B}_1)$ . If we consider  $\Gamma_1^{\text{PRF}}$ , all answers to queries are random so there is no difference between  $\mathcal{B}_0$  and  $\mathcal{B}_1$ . Hence,  $\Gamma_1^{\text{PRF}}(\mathcal{B}_0)$  and  $\Gamma_1^{\text{PRF}}(\mathcal{B}_1)$  are identical.

**Q.2** Show that if  $f$  is a PRF, then  $f$  is a prePRF.

Let  $\mathcal{A}$  be a prePRF adversary. We construct an adversary  $\mathcal{B}_b$  as in the previous question. We have

$$\begin{aligned}
 \text{Adv}_{\mathcal{A}}^{\text{prePRF}} &= \Pr[\Gamma_0^{\text{prePRF}}(\mathcal{A}) = 1] - \Pr[\Gamma_1^{\text{prePRF}}(\mathcal{A}) = 1] \\
 &= \Pr[\Gamma_0^{\text{PRF}}(\mathcal{B}_0) = 1] - \Pr[\Gamma_0^{\text{PRF}}(\mathcal{B}_1) = 1] \\
 &= \Pr[\Gamma_0^{\text{PRF}}(\mathcal{B}_0) = 1] - \Pr[\Gamma_1^{\text{PRF}}(\mathcal{B}_0) = 1] + \\
 &\quad \Pr[\Gamma_1^{\text{PRF}}(\mathcal{B}_1) = 1] - \Pr[\Gamma_0^{\text{PRF}}(\mathcal{B}_1) = 1] \\
 &= \text{Adv}_{\mathcal{B}_0}^{\text{PRF}} - \text{Adv}_{\mathcal{B}_1}^{\text{PRF}}
 \end{aligned}$$

Since  $f$  is a PRF, both  $\text{Adv}_{\mathcal{B}_0}^{\text{PRF}}$  and  $\text{Adv}_{\mathcal{B}_1}^{\text{PRF}}$  are negligible. So,  $\text{Adv}_{\mathcal{A}}^{\text{prePRF}}$  is negligible. As this holds for any PPT adversary  $\mathcal{A}$ ,  $f$  is a prePRF.

**Q.3** We define the following game:

**Game**  $\Gamma^j(\mathcal{A})$ :

1: pick  $s \in \mathcal{K}$  at random

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2: set  $\rho$  to a long enough sequence of random coins
3: set  $i = 1$ 
4:  $(q, x_i) \leftarrow \mathcal{A}(\rho)$ 
5: while  $q \neq \text{final}$  do
6:   if  $x_i \in \{x_1, \dots, x_{i-1}\}$  then
7:     abort {it is not allowed to repeat a query}
8:   end if
9:   if  $i \leq j$  then
10:    set  $y_i = f_s(x_i)$  {answer using  $f_s$  to the  $j$  first queries}
11:   else
12:    set  $y_i \in \mathcal{Y}$  at random
13:   end if
14:    $i \leftarrow i + 1$ 
15:    $(q, x_i) \leftarrow \mathcal{A}(y_1, \dots, y_{i-1}; \rho)$ 
16: end while{we must have  $q = \text{final}$ }
17: output  $x_i$ 

```

Show that for a PPT adversary  $\mathcal{A}$ , there exists some polynomially bounded  $Q$  such that we have

$$\Pr[\Gamma^Q(\mathcal{A}) = 1] = \Pr[\Gamma_0^{\text{PRF}}(\mathcal{A}) = 1]$$

$$\Pr[\Gamma^0(\mathcal{A}) = 1] = \Pr[\Gamma_1^{\text{PRF}}(\mathcal{A}) = 1]$$

*For  $j = 0$ , the  $\Gamma^0$  game always answers to queries at random, so just like the  $\Gamma_1^{\text{PRF}}$  game. If we set  $Q$  to at least the total number of queries by  $\mathcal{A}$ , which must be polynomially bounded, the  $\Gamma^Q$  game always answers to queries by setting  $y_i \in \mathcal{Y}$  at random, so just like the  $\Gamma_0^{\text{PRF}}$  game.*

**Q.4** Given a PPT adversary  $\mathcal{A}$  and an integer  $j$ , we construct an adversary  $\mathcal{B}_j$  as follows:

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 $\mathcal{B}_j(y_1, \dots, y_{i-1}; \rho)$ :
1: run  $(q, x_i) = \mathcal{A}(y_1, \dots, y_{i-1}; \rho)$ 
2: if  $i = j$  then
3:   set  $q$  to challenge
4: end if
5: if  $i > j$  then
6:   while  $q \neq \text{final}$  and  $x_i \notin \{x_1, \dots, x_{i-1}\}$  do
7:     set  $y_i \in \mathcal{Y}$  at random
8:      $i \leftarrow i + 1$ 
9:     run  $(q, x_i) = \mathcal{A}(y_1, \dots, y_{i-1}; \rho)$ 
10:   end while
11: end if

```

12: **output**  $(q, x_i)$

Show that

$$\begin{aligned}\Pr[\Gamma^j(\mathcal{A}) = 1] &= \Pr[\Gamma_0^{\text{prePRF}}(\mathcal{B}_j) = 1] \\ \Pr[\Gamma^{j-1}(\mathcal{A}) = 1] &= \Pr[\Gamma_1^{\text{prePRF}}(\mathcal{B}_j) = 1]\end{aligned}$$

*This means that  $\mathcal{B}_j$  will make exactly  $j - 1$  queries which fully simulate the  $\mathcal{A}$  queries. The  $j$ th one will be given as a challenge in the **prePRF** game. Then,  $\mathcal{B}$  will simulate  $\mathcal{A}$  who is always answered at random (unless a query repeats in which case it will repeat as well and let the game abort).*

*In  $\Gamma_0^{\text{prePRF}}$ , exactly  $j$  queries use  $f_s$  and others use random output, just like in  $\Gamma^j$ . In  $\Gamma_1^{\text{prePRF}}$ , exactly  $j - 1$  queries use  $f_s$  and others use random output, just like in  $\Gamma^{j-1}$ .*

**Q.5** Show that if  $f$  is a prePRF, then  $f$  is a PRF.

*Let  $\mathcal{A}$  be a PPT adversary playing the PRF game. We construct the  $\mathcal{B}_j$  adversaries. Due to the two previous questions, for  $Q$  large enough, we have*

$$\begin{aligned}\text{Adv}_{\mathcal{A}}^{\text{PRF}} &= \Pr[\Gamma_0^{\text{PRF}}(\mathcal{A}) = 1] - \Pr[\Gamma_1^{\text{PRF}}(\mathcal{A}) = 1] \\ &= \Pr[\Gamma^Q(\mathcal{A}) = 1] - \Pr[\Gamma^0(\mathcal{A}) = 1] \\ &= \sum_{j=1}^Q \Pr[\Gamma^j(\mathcal{A}) = 1] - \Pr[\Gamma^{j-1}(\mathcal{A}) = 1] \\ &= \sum_{j=1}^Q \Pr[\Gamma_0^{\text{prePRF}}(\mathcal{B}_j) = 1] - \Pr[\Gamma_1^{\text{prePRF}}(\mathcal{B}_j) = 1] \\ &= \sum_{j=1}^Q \text{Adv}_{\mathcal{B}_j}^{\text{prePRF}}\end{aligned}$$

*If  $f$  is a prePRF, then all terms in this sum are negligible. Since the number  $Q$  of terms is polynomially bounded, the sum is also negligible. So,  $\text{Adv}_{\mathcal{A}}^{\text{PRF}}$  is negligible. As this holds for any PPT adversary  $\mathcal{A}$ ,  $f$  is a PRF.*