

# Advanced Cryptography — Midterm Exam

## Solution

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- duration: 1h45
- 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 Threshold Implementation to Mitigate Power Cryptanalysis

*This exercise is inspired from Vaudenay, Side-Channel Attacks on Threshold Implementations using a Glitch Algebra, CANS 2016, LNCS vol. 10052, Springer.*

We consider a hardware circuit to implement a cryptographic function  $F$  mapping  $k$  secret key bits and  $p$  input bits to  $q$  output bits:

$$F : \{0, 1\}^k \times \{0, 1\}^p \longrightarrow \{0, 1\}^q$$
$$(K, x) \longmapsto y$$

We assume that the circuit is composed of AND gates (denoted by  $\wedge$ ), XOR gates (denoted by  $\oplus$ ), and wires. The circuit works following a clock signal. During each time period, the wires have constant signals and the gates propagate the computations (with a small latency). Gates normally dissipate no power. So, the total power consumption of a circuit is normally null during each time period. However, a wire could have a *glitch* which makes gates compute more during the time period, trying to follow the glitch in the signal. In that case, gates dissipate power and may reproduce the glitch with a small latency to their output. To simplify the analysis, we assume that during each time period, a wire  $w$  represents a bit  $v_w \in \{0, 1\}$  and has a number of glitches equal to  $n_w$ . Concretely, we assume the following behaviors for a gate  $g : (a, b) \rightarrow c$  with input wires  $a$  and  $b$  and output wire  $c$ :

- for a  $\wedge$  gate:  $v_c = v_a v_b$  and  $n_c = v_a n_b + v_b n_a$ ;
- for a  $\oplus$  gate:  $v_c = (v_a + v_b) \bmod 2$  and  $n_c = n_a + n_b$ ;

- the gate dissipates an energy equal to  $H_g = n_c h_g$ , where  $h_g$  is a constant depending on the gate  $g$  (i.e.,  $h_g = h_\wedge$  for an AND gate and  $h_g = h_\oplus$  for a XOR gate).

We consider a hardware implementation with a built-in secret  $K \in \{0, 1\}^k$  which is randomly set up at the beginning, and unknown to the adversary. The goal of the adversary is to recover  $K$ . The adversary can arbitrarily select the input  $x$ , get  $y = F(K, x)$ , and see the total amount of energy  $H = \sum_g H_g$  which is dissipated during each time period. We assume that the adversary knows the structure of the hardware circuit. We further assume that  $n_w = 0$  for all the input gates except one special wire  $w_0$  for which  $n_{w_0} = 1$ . The adversary knows  $w_0$  and  $n_{w_0}$  as well.

- Q.1** To start with a simple example, we assume that  $w_0$  is such that  $v_{w_0} = x_i$ , the  $i$ th input bit in  $x$ , and that  $w_0$  is an input wire to an AND gate  $g$  where the second input wire is  $w_1$  such that  $v_{w_1} = K_j$ , the  $j$ th bit of  $K$ . Show how the adversary can obtain  $K_j$  in this attack model.

*Let  $w_2$  be the output wire of  $g$ . We have  $n_{w_2} = v_{w_0}n_{w_1} + v_{w_1}n_{w_0} = x_i \times 0 + K_j \times 1 = K_j$ . If  $n_{w_2} = 0$ , no glitch propagate to the rest of the circuit so  $H = 0$ . Otherwise,  $H \geq n_{w_2}h_g = K_j h_\wedge$ . Hence, the adversary deduces  $K_j = 1_{H \geq h_\wedge}$ .*

- Q.2** We now consider a special way to compute an AND. Assume we want to compute the AND between a bit  $A$  and a bit  $B$ . We first represent  $A$  and  $B$  by two random pairs of bits  $(v_{a_1}, v_{a_2})$  and  $(v_{b_1}, v_{b_2})$  such that  $A = v_{a_1} \oplus v_{a_2}$  and  $B = v_{b_1} \oplus v_{b_2}$ . Then, we evaluate the following formula in a circuit:

$$c_1 = \text{random} \quad c_2 = (((c_1 \oplus (a_1 \wedge b_1)) \oplus (a_1 \wedge b_2)) \oplus (b_1 \wedge a_2)) \oplus (a_2 \wedge b_2)$$

We thus have a circuit with input wires  $a_1, a_2, b_1, b_2$  and output wires  $c_1, c_2$  and gates as defined by the above formula.

- Q.2a** Prove that  $v_{c_1} \oplus v_{c_2} = A \wedge B$ .

*We simplify*

$$\begin{aligned} v_{c_1} \oplus v_{c_2} &= v_{c_1} \oplus (((v_{c_1} \oplus (v_{a_1} \wedge v_{b_1})) \oplus (v_{a_1} \wedge v_{b_2})) \oplus (v_{b_1} \wedge v_{a_2})) \oplus (v_{a_2} \wedge v_{b_2}) \\ &= (v_{a_1} v_{b_1} + v_{a_1} v_{b_2} + v_{b_1} v_{a_2} + v_{a_2} v_{b_2}) \bmod 2 \\ &= (v_{a_1} + v_{a_2})(v_{b_1} + v_{b_2}) \bmod 2 \\ &= AB \end{aligned}$$

- Q.2b** Assume that  $w_0 = a_1$  in the above circuit. Compute  $H$  and prove that the adversary can recover  $B$  from  $H$ .

Following the above circuit, we split into the following gates:

$$\begin{aligned} g_1 &= a_1 \wedge b_1 & g_2 &= a_1 \wedge b_2 & g_3 &= b_1 \wedge a_2 & g_4 &= a_2 \wedge b_2 \\ g_5 &= c_1 \oplus g_1 & g_6 &= g_5 \oplus g_2 & g_7 &= g_6 \oplus g_3 & c_2 &= g_7 \oplus g_4 \end{aligned}$$

and we compute  $n_{g_1} = v_{b_1}$ ,  $n_{g_2} = v_{b_2}$ ,  $n_{g_3} = n_{g_4} = 0$ ,  $n_{g_5} = v_{b_1}$ , and  $n_{g_6} = n_{g_7} = n_{c_2} = v_{b_1} + v_{b_2}$ . So,

$$H = (v_{b_1} + v_{b_2})h_{\wedge} + (4v_{b_1} + 3v_{b_2})h_{\oplus}$$

If  $B = 0$ , we have  $v_{b_1} = v_{b_2}$  which are random, so  $H = v_{b_1}(2h_{\wedge} + 7h_{\oplus})$ . If  $B = 1$ , we have  $v_{b_2} = 1 - v_{b_1}$  and  $v_{b_1}$  random, so  $H = h_{\wedge} + (3 + v_{b_1})h_{\oplus}$ . So,  $H$  can only take 4 different values. The two extreme ones indicate  $B = 0$  and the two others indicate  $B = 1$ .

**Q.3** We now represent  $A = v_{a_1} \oplus v_{a_2} \oplus v_{a_3}$  and  $B = v_{b_1} \oplus v_{b_2} \oplus v_{b_3}$ , and take the following circuit

$$\begin{aligned} c_1 &= (a_2 \wedge b_2) \oplus ((a_2 \wedge b_3) \oplus (a_3 \wedge b_2)) \\ c_2 &= (a_3 \wedge b_3) \oplus ((a_1 \wedge b_3) \oplus (a_3 \wedge b_1)) \\ c_3 &= (a_1 \wedge b_1) \oplus ((a_1 \wedge b_2) \oplus (a_2 \wedge b_1)) \end{aligned}$$

**Q.3a** Prove that  $v_{c_1} \oplus v_{c_2} \oplus v_{c_3} = A \wedge B$ .

Clearly,

$$\begin{aligned} &v_{c_1} + v_{c_2} + v_{c_3} \\ &\equiv v_{a_2}v_{b_2} + v_{a_2}v_{b_3} + v_{a_3}v_{b_2} + v_{a_3}v_{b_3} + v_{a_1}v_{b_3} + v_{a_3}v_{b_1} + v_{a_1}v_{b_1} + v_{a_1}v_{b_2} + v_{a_2}v_{b_1} \\ &= (v_{a_1} + v_{a_2} + v_{a_3})(v_{b_1} + v_{b_2} + v_{b_3}) \pmod{2} \\ &= AB \end{aligned}$$

**Q.3b** Assume that  $w_0 = a_1$  in the above circuit. Prove that  $H = (v_{b_1} + v_{b_2} + v_{b_3})h_{\wedge} + (v_{b_1} + 2v_{b_2} + 2v_{b_3})h_{\oplus}$ .

We compute  $n_{a_i \wedge b_j} = v_{b_j} n_{a_i}$  for all  $i, j$ .

$$n_{c_1} = v_{b_2} n_{a_2} + v_{b_3} n_{a_2} + v_{b_2} n_{a_3}$$

$$n_{c_2} = v_{b_3} n_{a_3} + v_{b_3} n_{a_1} + v_{b_1} n_{a_3}$$

$$n_{c_3} = v_{b_1} n_{a_1} + v_{b_2} n_{a_1} + v_{b_1} n_{a_2}$$

We have also three internal XOR. Overall, with  $n_{a_1} = 1$  and  $n_{a_2} = n_{a_3} = 0$ , we obtain

$$n_{(a_2 \wedge b_3) \oplus (a_3 \wedge b_2)} = 0$$

$$n_{(a_1 \wedge b_3) \oplus (a_3 \wedge b_1)} = v_{b_3}$$

$$n_{(a_1 \wedge b_2) \oplus (a_2 \wedge b_1)} = v_{b_2}$$

$$n_{c_1} = 0$$

$$n_{c_2} = v_{b_3}$$

$$n_{c_3} = v_{b_1} + v_{b_2}$$

hence

$$H = (v_{b_1} + v_{b_2} + v_{b_3})h_{\wedge} + (v_{b_1} + 2v_{b_2} + 2v_{b_3})h_{\oplus}$$

**Q.3c** Show that  $E(H|B=0) = E(H|B=1)$  so, the expected value of  $H$  does not depend on  $B$ .

Since  $H = (v_{b_1} + v_{b_2} + v_{b_3})h_{\wedge} + (v_{b_1} + 2v_{b_2} + 2v_{b_3})h_{\oplus}$ , by linearity,  $E(H|B) = (E(v_{b_1}|B) + E(v_{b_2}|B) + E(v_{b_3}|B))h_{\wedge} + (E(v_{b_1}|B) + 2E(v_{b_2}|B) + 2E(v_{b_3}|B))h_{\oplus}$ . But  $E(v_{b_i}|B) = \frac{1}{2}$  for every  $i$  and  $B$ . So,  $E(H|B) = \frac{3}{2}h_{\wedge} + \frac{5}{2}h_{\oplus}$ . This does not depend on  $B$ .

**Q.3d** We assume that  $h_{\oplus} = 4h_{\wedge}$ . Study the probability distribution of  $H$  when  $B=0$  and when  $B=1$  and prove that the adversary can recover  $B$  from  $H$ .

For  $B = 0$ , we have the following equiprobable cases:

$v_{b_1}$	$v_{b_2}$	$v_{b_3}$	$H$	$H/h_{\wedge}$ with $h_{\oplus} = 4h_{\wedge}$
0	0	0	0	0
0	1	1	$2h_{\wedge} + 4h_{\oplus}$	18
1	0	1	$2h_{\wedge} + 3h_{\oplus}$	14
1	1	0	$2h_{\wedge} + 3h_{\oplus}$	14

For  $B = 1$ , we have the following equiprobable cases:

$v_{b_1}$	$v_{b_2}$	$v_{b_3}$	$H$	$H/h_{\wedge}$ with $h_{\oplus} = 4h_{\wedge}$
0	0	1	$1h_{\wedge} + 2h_{\oplus}$	5
0	1	0	$1h_{\wedge} + 2h_{\oplus}$	5
1	0	0	$1h_{\wedge} + 1h_{\oplus}$	4
1	1	1	$3h_{\wedge} + 5h_{\oplus}$	13

Clearly, from the value of  $H$ , we can see if we are in one case or the other.

## 2 The Gap Diffie-Hellman Problem

We define three problems: CDH, DDH, and GDH. They are all relative to a public parameters setup scheme  $\text{Gen}(1^\lambda) \rightarrow \text{pp}$ . We assume that  $\text{pp}$  defines a cyclic group  $G_{\text{pp}}$  with generator  $g_{\text{pp}}$  (we assume multiplicative notations) of prime order  $p_{\text{pp}}$ , and an algorithm to multiply in  $G_{\text{pp}}$ .

We define three games below. We say the CDH problem is hard if for every PPT algorithm  $\mathcal{A}$ ,  $\Pr[\text{CDH}_{\mathcal{A}}(1^\lambda) \text{ wins}]$  is negligible in the CDH game. We say the DDH problem is hard if for every PPT algorithm  $\mathcal{A}$ , the advantage

$$\text{Adv}_{\mathcal{A}}^{\text{DDH}}(\lambda) = \Pr[\text{DDH}_{\mathcal{A}}(1^\lambda, 1) \rightarrow 1] - \Pr[\text{DDH}_{\mathcal{A}}(1^\lambda, 0) \rightarrow 1]$$

is negligible in the DDH game. We say the  $\text{GDH}_{\mathcal{A}}$  problem is hard if for every PPT algorithm  $\mathcal{A}$ ,  $\Pr[\text{GDH}_{\mathcal{A}}(1^\lambda) \text{ wins}]$  is negligible in the GDH game. Essentially, the GDH problem is the CDH problem with access to an oracle  $\mathcal{O}$  who can tell if a triplet  $(g^x, g^y, g^z)$  satisfies  $z \equiv xy \pmod{p_{\text{pp}}}$ . Namely,  $\mathcal{O}(\text{pp}, g^x, g^y, g^z) = 1_{z \equiv xy \pmod{p_{\text{pp}}}}$ . We call such  $\mathcal{O}$  a *perfect DDH oracle*.

$\text{CDH}_{\mathcal{A}}(1^\lambda)$ : 1: $\text{Gen}(1^\lambda) \rightarrow \text{pp}$ 2: pick $x, y \in \mathbf{Z}_{p_{\text{pp}}}$ uniformly 3: $X \leftarrow g_{\text{pp}}^x$ 4: $Y \leftarrow g_{\text{pp}}^y$ 5: $Z \leftarrow \mathcal{A}(\text{pp}, X, Y)$ 6: win if and only if $Z = g_{\text{pp}}^{xy}$	$\text{DDH}_{\mathcal{A}}(1^\lambda, b)$ : 1: $\text{Gen}(1^\lambda) \rightarrow \text{pp}$ 2: pick $x, y, z \in \mathbf{Z}_{p_{\text{pp}}}$ uniformly 3: if $b = 1$ , overwrite $z \leftarrow xy$ 4: $X \leftarrow g_{\text{pp}}^x$ 5: $Y \leftarrow g_{\text{pp}}^y$ 6: $Z \leftarrow g_{\text{pp}}^z$ 7: $b' \leftarrow \mathcal{A}(\text{pp}, X, Y, Z)$ 8: output $b'$	$\text{GDH}_{\mathcal{A}}(1^\lambda)$ : 1: $\text{Gen}(1^\lambda) \rightarrow \text{pp}$ 2: pick $x, y \in \mathbf{Z}_{p_{\text{pp}}}$ uniformly 3: $X \leftarrow g_{\text{pp}}^x$ 4: $Y \leftarrow g_{\text{pp}}^y$ 5: $Z \leftarrow \mathcal{A}^{\mathcal{O}}(\text{pp}, X, Y)$ 6: win if and only if $Z = g_{\text{pp}}^{xy}$  oracle $\mathcal{O}(\text{pp}, A, B, C)$ : 7: compute the discrete logarithm $a \in \mathbf{Z}_{p_{\text{pp}}}$ such that $A = g_{\text{pp}}^a$ 8: $C' \leftarrow B^a$ 9: return $1_{C=C'}$
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**Q.1** Give an example of a generator  $\text{Gen}$  with which the DDH problem is easy but the CDH problem is believed to be hard.

Consider  $\text{Gen}(1^\lambda)$  which generates a random prime number  $p$  of  $\lambda$  bits then finds a generator  $g$  of  $\mathbf{Z}_p^*$ . We define  $G_{\text{pp}} = \mathbf{Z}_p^*$ ,  $g_{\text{pp}} = g$ , and  $p_{\text{pp}} = p$ . We know that the DDH problem is easy in this case using the following distinguisher:

$\mathcal{A}(\text{pp}, X, Y, Z)$ :

- 1: set  $x \in \{0, 1\}$  such that  $(-1)^x = \left(\frac{X}{p}\right)$
- 2: set  $y \in \{0, 1\}$  such that  $(-1)^y = \left(\frac{Y}{p}\right)$
- 3: set  $z \in \{0, 1\}$  such that  $(-1)^z = \left(\frac{Z}{p}\right)$
- 4: return  $1_{z=xy}$

However, the CDH problem is believed to be hard.

**Q.2** Prove that the GDH problem reduces to the CDH problem (i.e., solving CDH implies solving GDH).

*Assuming an oracle  $\mathcal{S}$  which solves the CDH problem, solving the GDH problem is trivial: we just define*

$$\mathcal{A}^{\mathcal{O},\mathcal{S}}(\text{pp}, X, Y) = \mathcal{S}(\text{pp}, X, Y)$$

*without using  $\mathcal{O}$ .*

**Q.3** We let  $\mathcal{O}$  be a perfect DDH oracle. We now assume there exists a PPT distinguisher  $\mathcal{D}$  such that for any PPT algorithm  $\mathcal{G}(\text{pp}) \rightarrow (X, Y, Z)$ , if we generate  $\text{Gen}(1^\lambda) \rightarrow \text{pp}$  then  $\mathcal{G}(\text{pp}) \rightarrow (X, Y, Z)$ , then  $\mathcal{D}(\text{pp}, X, Y, Z) \rightarrow b$ , then  $b = \mathcal{O}(\text{pp}, X, Y, Z)$  except with negligible probability.

**Q.3a** Prove that for any PPT algorithm  $\mathcal{A}$  with access to an oracle, then running  $\mathcal{A}$  with oracle  $\mathcal{D}$  or  $\mathcal{O}$  and the same random coins produces the same result, except with negligible probability.

*Conditioned to that  $\mathcal{D}$  returned the same result as  $\mathcal{O}$  for the first  $i - 1$  queries, the  $i$ th query is defined by a PPT algorithm  $\mathcal{G}$ . Due to the previous question, the answer will match the one of  $\mathcal{O}$  except with negligible probability. We have a polynomially bounded number of queries. So, by induction, the probability that any query does not match the one of  $\mathcal{O}$  is negligible.*

**Q.3b** Under the same assumption that  $\mathcal{D}$  exists, prove that the CDH problem is as hard as the GDH problem.

*We have proven one reduction in the previous question. We then prove that the CDH problem reduces to the GDH problem. Assume that  $\mathcal{S}$  is a GDH oracle, i.e., if  $\mathcal{O}$  is a perfect DDH oracle, then  $\mathcal{S}^{\mathcal{O}}$  is a perfect CDH solver. We define  $\mathcal{A}(\text{pp}, X, Y) = \mathcal{S}^{\mathcal{D}}(\text{pp}, X, Y)$ . Due to the previous question,  $\mathcal{A}$  returns the same as  $\mathcal{S}^{\mathcal{O}}$ , except with negligible probability. So,  $1 - \Pr[\text{CDH}_{\mathcal{A}}(1^\lambda) \text{ wins}]$  is negligible.*

### 3 Number of Samples to Distinguish Distributions

A *distribution* is a function  $P$  from a set  $\mathcal{Z}$  to  $\mathbf{R}$  such that for all  $z \in \mathcal{Z}$ , we have  $P(z) \geq 0$  and  $\sum_{z \in \mathcal{Z}} P(z) = 1$ . (We implicitly focus on discrete distributions on finite sets  $\mathcal{Z}$ .)

Given two distributions  $P$  and  $Q$ , we define

$$d(P, Q) = \frac{1}{2} \sum_{z \in \mathcal{Z}} |P(z) - Q(z)|$$

as the *statistical distance* between  $P$  and  $Q$ . We recall that  $d$  is a distance, which means that for all distributions  $P, Q$ , and  $R$ , we have  $d(P, Q) \geq 0$ ,  $d(P, Q) = 0$  is equivalent to  $P = Q$ ,  $d(P, Q) = d(Q, P)$ , and  $d(P, R) \leq d(P, Q) + d(Q, R)$ . We also define

$$F(P, Q) = \sum_{z \in \mathcal{Z}} \sqrt{P(z)Q(z)}$$

as the *fidelity* between  $P$  and  $Q$ . The fidelity  $F$  is not a distance but  $H = \sqrt{1 - F}$  is. (This is the *Hellinger distance*.) The statistical distance and the fidelity satisfy the *Fuchs – van de Graaf inequality*

$$1 - F(P, Q) \leq d(P, Q) \leq \sqrt{1 - F(P, Q)^2}$$

Given two distributions  $P$  and  $Q$  on sets  $\mathcal{A}$  and  $\mathcal{B}$  respectively, we define a distribution

$R = P \otimes Q$  on the set  $\mathcal{A} \times \mathcal{B}$  by  $R(a, b) = P(a)Q(b)$ . We define  $P^{\otimes n} = \overbrace{P \otimes \dots \otimes P}^{n \text{ times}}$ .

**Q.1** For any distributions  $P, P', Q, Q'$ , prove that  $F(P \otimes P', Q \otimes Q') = F(P, Q) \times F(P', Q')$ .

We have

$$\begin{aligned} F(P \otimes P', Q \otimes Q') &= \sum_{z, z'} \sqrt{P(z)P'(z')Q(z)Q'(z')} \\ &= \sum_{z, z'} \sqrt{P(z)Q(z)} \sqrt{P'(z')Q'(z')} \\ &= \left( \sum_z \sqrt{P(z)Q(z)} \right) \times \left( \sum_{z'} \sqrt{P'(z')Q'(z')} \right) \\ &= F(P, Q) \times F(P', Q') \end{aligned}$$

**Q.2** Given a real number  $t \in [0, 1]$ , we let  $n_t$  be the minimal number of samples  $n$  such that there exists a distinguisher  $\mathcal{A}$  using  $n$  independent and identically distributed samples to distinguish  $P$  from  $Q$  such that  $\text{Adv}(\mathcal{A}) \geq t$ . Prove that for any  $t$ , we have

$$\frac{\log(1 - t^2)}{2 \log F(P, Q)} \leq n_t < 1 + \frac{\log(1 - t)}{\log F(P, Q)}$$



By definition of  $n_t$ , we have  $d(P^{\otimes n_t}, Q^{\otimes n_t}) \geq t$  and  $d(P^{\otimes(n_t-1)}, Q^{\otimes(n_t-1)}) < t$ .

We have

$$t \leq d(P^{\otimes n_t}, Q^{\otimes n_t}) \leq \sqrt{1 - F(P^{\otimes n_t}, Q^{\otimes n_t})^2} = \sqrt{1 - F(P, Q)^{2n_t}}$$

so

$$F(P, Q)^{2n_t} \leq 1 - t^2$$

This shows the lower bound on  $n_t$ . Similarly, we have

$$t > d(P^{\otimes(n_t-1)}, Q^{\otimes(n_t-1)}) \geq 1 - F(P^{\otimes(n_t-1)}, Q^{\otimes(n_t-1)}) = 1 - F(P, Q)^{n_t-1}$$

so

$$F(P, Q)^{n_t-1} > 1 - t$$

This shows the upper bound on  $n_t$ .

**Q.3** Let  $T$  be a random process mapping an input  $x \in \mathcal{X}$  and some random coins  $\rho \in \{0, 1\}^*$  to an output  $T(x; \rho) \in \mathcal{Y}$ . If  $X$  follows a distribution  $P$  on  $\mathcal{X}$ , and the random coins  $\rho$  are independent and following the uniform distribution, we say that  $T(X; \rho)$  follows a distribution  $P^T$  on  $\mathcal{Y}$ . Similarly, a distribution  $Q$  on  $\mathcal{X}$  induces a distribution  $Q^T$  on  $\mathcal{Y}$ .

Prove that  $d(P^T, Q^T) \leq d(P, Q)$ .

Given  $y \in \mathcal{Y}$ , let  $I_y = \{(x, \rho) \in \mathcal{X} \times \{0, 1\}^*; T(x; \rho) = y\}$ . We have

$$\begin{aligned}
 d(P^T, Q^T) &= \frac{1}{2} \sum_{y \in \mathcal{Y}} \left| \sum_{(x, \rho) \in I_y} (P(x) \Pr[\rho] - Q(x) \Pr[\rho]) \right| \\
 &= \frac{1}{2} \sum_{y \in \mathcal{Y}} \left| \sum_{(x, \rho) \in I_y} (P(x) - Q(x)) \Pr[\rho] \right| \\
 &\leq \frac{1}{2} \sum_{y \in \mathcal{Y}} \sum_{(x, \rho) \in I_y} |P(x) - Q(x)| \Pr[\rho] \\
 &= \frac{1}{2} \sum_{x \in \mathcal{X}} \sum_{\rho \in \{0, 1\}^*} |P(x) - Q(x)| \Pr[\rho] \\
 &= \frac{1}{2} \sum_{x \in \mathcal{X}} |P(x) - Q(x)| \\
 &= d(P, Q)
 \end{aligned}$$

Another way to prove this is to use the equivalence between the statistical distance and the advantage of the best distinguisher limited to one sample. We have  $d(P^T, Q^T) = \text{Adv}(\mathcal{A})$  for some distinguisher  $\mathcal{A}(Y)$  who gets one sample  $Y$  and produce a bit. Let  $X \in \mathcal{X}$ . We define  $\mathcal{B}(X)$  as follows:

- 1: pick  $\rho$  at random
- 2: output  $\mathcal{A}(T(X; \rho))$

Clearly,  $\mathcal{B}$  is a distinguisher between  $P$  and  $Q$  and has advantage  $\text{Adv}(\mathcal{B}) = \text{Adv}(\mathcal{A})$ . Hence,

$$d(P^T, Q^T) = \text{Adv}(\mathcal{A}) = \text{Adv}(\mathcal{B}) \leq d(P, Q)$$

**Q.4** Use the previous question to prove that  $d(P \otimes P', Q \otimes Q') \leq d(P, Q) + d(P', Q')$ .

HINT: use first the triangular inequality  $d(P \otimes P', Q \otimes Q') \leq d(P \otimes P', Q \otimes P') + d(Q \otimes P', Q \otimes Q')$ .

Since  $d$  is a distance, we can use the triangular inequality  $d(P \otimes P', Q \otimes Q') \leq d(P \otimes P', Q \otimes P') + d(Q \otimes P', Q \otimes Q')$ . Next, we show that  $d(P \otimes P', Q \otimes P') \leq d(P, Q)$  and  $d(Q \otimes P', Q \otimes Q') \leq d(P', Q')$  by the same technique. We show it only for the first one.

To show  $d(P \otimes P', Q \otimes P') \leq d(P, Q)$ , we use a sampling algorithm  $G(\rho)$  which converts random coins  $\rho$  into a random variable following the distribution  $P'$ . Then, we define  $T(z; \rho) = (z, G(\rho))$ . We observe that  $P^T = P \otimes P'$  and  $Q^T = Q \otimes P'$ . So,  $d(P \otimes P', Q \otimes P') = d(P^T, Q^T) \leq d(P, Q)$ , following the previous question.

**Q.5** With the notations from Q.2, deduce that  $n_t \geq \frac{t}{d(P, Q)}$ .

We have

$$t \leq d(P^{\otimes n_t}, Q^{\otimes n_t}) \leq n_t \times d(P, Q)$$