

# Solution Sheet #11

*Advanced Cryptography 2021*

## Solution 1 Decorrelation

Recall that the  $a$ -norm of the matrix  $M$  is defined as

$$\max_{x_1} \left\{ \sum_{y_1} \max_{x_2} \left\{ \sum_{y_2} |M_{(x_1, x_2), (y_1, y_2)}| \right\} \right\}.$$

We consider first the most internal maximum.

- When  $x_1 = 0$  and  $y_1 = 0$ , we have  $\max \{5 + 4, 6 + 8\} = 14$ .
- When  $x_1 = 0$  and  $y_1 = 1$ , we have  $\max \{1 + 2, 0 + 4\} = 4$ .
- When  $x_1 = 1$  and  $y_1 = 0$ , we have  $\max \{2 + 4, 10 + 0\} = 10$ .
- When  $x_1 = 1$  and  $y_1 = 1$ , we have  $\max \{4 + 5, 0 + 1\} = 9$ .

Hence,  $\|M\|_a = \max \{14 + 4, 10 + 9\} = 19$ .

## Solution 2 Decorrelation and Differential Cryptanalysis

Let  $a \neq 0$  and  $b$  be such that

$$\text{EDP}_{\max}^C = E(\text{DP}^C(a, b)).$$

As  $\text{DP}^C(a, b) \geq 0$  and as  $\text{DP}^C(a, 0) = 0$ , we can assume that  $b \neq 0$ . We consider the distinguisher described in Algorithm 1. This distinguisher is limited to two queries. Its advantage must thus be less than  $\text{BestAdv}_{\text{CI}_a^2}(C, C^*)$ .

We now look for an expression of this advantage. When the oracle implements  $C$ , the probability that the distinguisher outputs 1 is  $E(\text{DP}^C(a, b))$ . When it implements  $C^*$ , the probability that it outputs 1 is  $1/(2^m - 1)$  since  $y_1$  and  $y_2$  are different random elements and  $b \neq 0$ . Therefore, the advantage of the distinguisher is equal to

$$E(\text{DP}^C(a, b)) - \frac{1}{2^m - 1}.$$

This leads to the inequality.

Hence, we deduce that studying the order two decorrelation of  $C$  is a good way to find an upper-bound on the best differential property and, thus, can be used to prove the resistance of a cipher against differential attacks.

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**Algorithm 1** A differential distinguisher between  $C$  and  $C^*$

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**Input:** an oracle  $\mathcal{O}$  implementing either  $C$  or  $C^*$ , two masks  $a$  and  $b$  such that  $a \neq 0$  and  $b \neq 0$

**Output:** 0 (if the guess is that  $\mathcal{O}$  implements  $C^*$ ) or 1 (if the guess is that  $\mathcal{O}$  implements  $C$ )

**Processing:**

- 1: pick  $x$  uniformly at random
  - 2: submit  $x$  and  $x \oplus a$  to  $\mathcal{O}$  and get  $y_1$  and  $y_2$
  - 3: **if**  $y_1 = y_2 \oplus b$  **then**
  - 4:   output 1
  - 5: **else**
  - 6:   output 0
  - 7: **end if**
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### Solution 3 Decorrelation (2)

1. By definition,

$$||| [C]^d - [C^*]^d |||_\infty = 2 \cdot \text{Adv}_{\text{C}_{\text{na}}^d}.$$

As this “measure” represents the advantage of the best non-adaptive distinguisher using  $d$  queries, it is rather clear that

$$||| [C]^{d-1} - [C^*]^{d-1} |||_\infty \leq ||| [C]^d - [C^*]^d |||_\infty$$

since the best non-adaptative distinguisher using  $d-1$  queries can be considered as a non-adaptative distinguisher using  $d$  queries, including one which is not taken into account.

2. By definition, an advantage is given by

$$| \Pr[\mathcal{A}^C \rightarrow 1] - \Pr[\mathcal{A}^{C^*} \rightarrow 1] |.$$

As a probability measure returns always a result in the interval  $[0, 1]$ , we have

$$| \Pr[\mathcal{A}^C \rightarrow 1] - \Pr[\mathcal{A}^{C^*} \rightarrow 1] | \leq 1$$

which implies that

$$||| [C]^d - [C^*]^d |||_\infty \leq 2.$$

Furthermore, as  $||| \cdot |||_\infty$  is a norm, we have

$$||| [C]^d - [C^*]^d |||_\infty \geq 0.$$

3. The property  $\text{Dec}^d(C) = 0$  means that the distance between  $[C]^d$  and  $[C^*]^d$  is zero. By definition of a distance, this happens if and only if  $[C]^d = [C^*]^d$ . Obviously this does not depend on the choice of the distance.
4. The above property with  $d = 1$  means that  $[C]^1 = [C^*]^1$ . The coefficient of these matrices are the probabilities  $\Pr[C(x) = y]$ . Therefore, this property means that for any  $x$  and  $y$ , we have

$$\Pr[C(x) = y] = \Pr[C^*(x) = y].$$

Since  $\Pr[C^*(x) = y] = 2^{-m}$ , the property means that for any  $x$  and  $y$  we have  $\Pr[C(x) = y] = 2^{-m}$ . In this case we can prove that we have perfect secrecy.

For any  $x$  and  $y$ , we have

$$\Pr[X = x|C(X) = y] = \frac{\Pr[X = x]}{\Pr[C(X) = y]} \Pr[C(x) = y].$$

The probability  $\Pr[C(X) = y]$  can be computed as follows

$$\Pr[C(X) = y] = \sum_{x'} \Pr[C(x') = y|X = x'] \Pr[X = x'].$$

Since  $C$  and  $X$  are independent, we have

$$\Pr[C(x') = y|X = x'] = \Pr[C(x') = y] = 2^{-m}.$$

Thus  $\Pr[C(X) = y] = 2^{-m}$ . Therefore we obtain that

$$\Pr[X = x|C(X) = y] = \Pr[X = x]$$

for any distribution of  $X$ .

5.  $\text{Dec}^d(f_K) = 0$  means that for any pairwise different  $x_1, \dots, x_d$  and any  $y_1, \dots, y_d$ , we have  $\Pr[f_K(x_i) = y_i \text{ for } i = 1, \dots, d] = 2^{-md}$ .

Let us pick random pairwise different  $x_1, \dots, x_d$ . We obtain that for any  $y_1, \dots, y_d$ , the above probability is non-zero. This implies that there exists at least one key  $k$  such that  $f_k(x_i) = y_i$  for all  $i = 1, \dots, d$ . Therefore we must have at least  $2^{md}$  keys, i.e.,  $K$  must at least have a bit length of  $md$ . The purpose of the exercise is to show how to achieve this minimal key size.

6. For any  $x, y \in \{0, 1\}^m$  we have

$$[f_K]_{x,y}^1 = \Pr[f_K(x) = y] = \Pr[K = x \oplus y] = 2^{-m} = [F^*]_{x,y}^1.$$

Therefore  $[f_K]^1 = [F^*]^1$  which clearly implies that  $f_K$  is at distance 0 from  $F^*$ , i.e.,

$$\text{Dec}^1(f_K) = 0.$$

We notice that we achieve the minimal length for the key here.

7. We take  $K = (K_1, \dots, K_d) \in (\text{GF}(2^m))^d$  (which achieves the minimal length). We define  $f_K(x) = K_1 + K_2x + K_3x^2 + \dots + K_dx^{d-1}$  in the sense of  $\text{GF}(2^m)$  operations. For pairwise different  $x_1, \dots, x_d$  and any  $y_1, \dots, y_d$ , we can find a unique polynomial  $P$  such that  $P(x_i) = y_i$  by interpolation. The coefficients of this polynomial define a unique key  $K$  such that the polynomial is actually  $f_K$ . This proves that  $\Pr[f_K(x_i) = y_i \text{ for } i = 1, \dots, d] = 2^{-md}$ .