### Problem 1.

**Part A** For any p.p.t. distinguisher  $\mathcal{D}$  that tries to distinguish  $F_k^{\$}$  and  $f^{\$}$ , we can construct  $\mathcal{D}'$  who emulates  $\mathcal{D}$ . Given input  $1^{\lambda}$  and oracle access to  $\mathcal{O} \in \{F_k, f\}$ , the distinguisher  $\mathcal{D}'$  emulates the execution of  $\mathcal{D}(1^{\lambda})$ , upon each query from  $\mathcal{D}$ , samples  $r \leftarrow \{0,1\}^{\lambda}$  and feed  $(r,\mathcal{O}(r))$  to  $\mathcal{D}$ .

$$\begin{aligned} & \left| \Pr[\mathcal{D}^{F_k^{\$}}(1^{\lambda}) = 1] - \Pr[\mathcal{D}^{f^{\$}}(1^{\lambda}) = 1] \right| \\ & = \left| \Pr[\mathcal{D}'^{F_k}(1^{\lambda}) = 1] - \Pr[\mathcal{D}'^f(1^{\lambda}) = 1] \right| \le \operatorname{negl}(\lambda) \end{aligned}$$

Hence F is also a weak PRF.

**Part B** Let  $f': \{0,1\}^{\lambda} \to \{0,1\}^{\lambda}$  be a random function and define

$$f(x) := \begin{cases} f'(x), & \text{if } x \text{ is even} \\ f'(x+1), & \text{if } x \text{ is odd} \end{cases}$$

By the same argument as **Part A**, for any p.p.t. distinguisher  $\mathcal{D}$ 

$$\left|\Pr[\mathcal{D}^{F_k^{\$}}(1^{\lambda}) = 1] - \Pr[\mathcal{D}^{f^{\$}}(1^{\lambda}) = 1]\right| \le \operatorname{negl}(\lambda).$$

Let BAD denote the event that in the execution of  $\mathcal{D}$ ,  $|r_i - r_j| = 1$  for some two random inputs  $r_i, r_j$ . Conditioning on BAD doesn't happen, f will act identically to a random function.

$$\begin{split} & \left| \Pr[\mathcal{D}^{f^{\$}}(1^{\lambda}) = 1] - \Pr[\mathcal{D}^{f'^{\$}}(1^{\lambda}) = 1] \right| \\ & = \left| \left( \Pr[\mathcal{D}^{f^{\$}}(1^{\lambda}) = 1 \mid \mathsf{BAD}] - \Pr[\mathcal{D}^{f'^{\$}}(1^{\lambda}) = 1 \mid \mathsf{BAD}] \right) \Pr[\mathsf{BAD}] \\ & - \left( \Pr[\mathcal{D}^{f^{\$}}(1^{\lambda}) = 1 \mid \neg \mathsf{BAD}] - \Pr[\mathcal{D}^{f'^{\$}}(1^{\lambda}) = 1 \mid \neg \mathsf{BAD}] \right) \Pr[\neg \mathsf{BAD}] \right| \\ & = \left| \left( \Pr[\mathcal{D}^{f^{\$}}(1^{\lambda}) = 1 \mid \mathsf{BAD}] - \Pr[\mathcal{D}^{f'^{\$}}(1^{\lambda}) = 1 \mid \mathsf{BAD}] \right) \Pr[\mathsf{BAD}] \right| \\ & < \Pr[\mathsf{BAD}] < \operatorname{negl}(\lambda). \end{split}$$

Then by the triangular inequality,  $\mathcal{D}$  can not distinguish between  $F_k^{\$}$  and  $f'^{\$}$ , hence F is a weak PRF.

However, F is not a PRF since  $F_k(2x+1) = F_k(2x+2)$  holds for all x.

Part C The scheme is not secure even in the presence of an eavesdropper.

Assume the weak PRF we use is constructed as in **Part B**, choose  $m_0 = x||y||x, m_1 = x||x||x$  where  $x \neq y$  and output 1 if any two adjacent blocks of ciphertext are identical.

For the ciphertext of  $m_1$  always has two identical adjacent blocks. While for  $m_0$ , such event happens with probability  $\Pr[F'_k(r) \oplus F'_k(r+1) = x \oplus y]$ , which is negligible.

**Part D** Recall how we prove the CPA security of  $\Pi$  when the function F is a PRF. For any adversary  $\mathcal{A}$  targeting the CPA security of  $\Pi$ , we construct a distinguisher  $\mathcal{D}$ , which is essentially the CPA security game  $\mathsf{PrivK}_{\Pi,\mathcal{A}}^{\mathsf{CPA}}$ . The only difference is that, in  $\mathcal{D}$ 's

emulation, the challenger does not sample k, the computation of  $F_k$  is delegated to the oracle  $\mathcal{O}$ .

The proof of Part D is very similar. For any adversary  $\mathcal{A}$  targeting the CPA security of  $\Pi$ , we construct a distinguisher  $\mathcal{D}_{\text{new}}$ , which is essentially the CPA security game  $\mathsf{PrivK}_{\Pi,\mathcal{A}}^{\mathsf{CPA}}$ . The only difference is that, in  $\mathcal{D}_{\text{new}}$ 's emulation, the challenger does not sample k and whenever the challenger need to sample a random r and computes  $F_k(r)$ , the task is delegated to the probabilistic oracle. Since F is a weak PRF,

$$\left|\Pr[\mathcal{D}^{F_k^\$}_{\mathrm{new}}(1^\lambda) \to 1] - \Pr[\mathcal{D}^{f^\$}_{\mathrm{new}}(1^\lambda) \to 1]\right| \leq \mathrm{negl}(\lambda).$$

Since the challenger in  $\mathsf{PrivK}_{\Pi,\mathcal{A}}^{\mathsf{CPA}}$  only evaluates  $F_k$  on fresh random points, the behavior of  $\mathcal{D}^f$  and  $\mathcal{D}_{\mathsf{new}}^{f\$}$  are identical for any f,

$$\Pr[\mathcal{D}_{\text{new}}^{F_k^{\$}}(1^{\lambda}) \to 1] = \Pr[\mathcal{D}^{F_k}(1^{\lambda}) \to 1] = \Pr[\mathsf{PrivK}_{\Pi,\mathcal{A}}^{\text{CPA}}(\lambda) \to 1],$$

$$\Pr[\mathcal{D}_{\text{new}}^{f^{\$}}(1^{\lambda}) \to 1] = \Pr[\mathcal{D}^f(1^{\lambda}) \to 1] = \frac{1}{2} \pm \operatorname{negl}(\lambda).$$

(Both can be directly verified. But relying the equivalence between  $\mathcal{D}^f$  and  $\mathcal{D}^{f^\$}_{\text{new}}$  simplifies the proof.) Thus  $\Pr[\mathsf{PrivK}^{\mathsf{CPA}}_{\Pi,\mathcal{A}}(\lambda) \to 1] = \frac{1}{2} \pm \mathrm{negl}(\lambda)$ , the scheme  $\Pi$  is CPA-secure.

# Problem 2.

Part A P' is not a PRP.

Given oracle  $\mathcal{O}$ , the distinguisher picks  $x_L^0 \neq x_L^1, x_H^0 \neq x_H^1$  and checks if the lower parts of  $\mathcal{O}(x_L^0, x_H^0) + \mathcal{O}(x_L^1, x_H^1)$  and  $\mathcal{O}(x_L^0, x_H^1) + \mathcal{O}(x_L^1, x_H^0)$  are equal.

If  $\mathcal{O}$  is P', their parts are the same, which equals to

$$M\begin{pmatrix} y_H^0 + y_H^1 \\ y_L^0 + y_L^1 \end{pmatrix}$$
.

If  $\mathcal{O}$  is a random permutation, such probability is negligible.

**Part B** Let  $x_L^i, x_H^i, y_H^i, z_L^i, z_H^i, w_L^i, w_H^i$  denote the input, output and intermediate values of the *i*-th query. W.l.o.g., we assume the queries  $(x_L^i, x_H^i)$ r are distinct.

Consider World 1, where PRP  $F_{k_1}$ ,  $F_{k_2}$ ,  $F_{k_3}$  are replaced by random functions  $f_1$ ,  $f_2$ ,  $f_3$  respectively. Due to the security of PRF, the distinguisher cannot distinguish the real world from World 1 with a non-negligible margin.

Consider World 2, where  $f_2$ ,  $f_3$  are further replaced by "random boxes". Upon a query, a random box ignores the input and samples a fresh random value. The distinguisher cannot distinguish the ideal world from World 2 with a non-negligible margin.

It remains to show that the distinguisher cannot distinguish World 1 and World 2.

Define event Repeat =  $\{\exists i < j \text{ s.t. } z_L^i = z_L^j \lor z_H^i = z_H^j\}$ . When Repeat does not happen, World 1 and World 2 perform identically. So Pr[Repeat] in World 1 equals Pr[Repeat] in World 2, and is an upper bound of distinguishing margin.

It is easier to bound the probably of Repeat in World 2. In World 2, the adversary receives no information of  $y_H^i, z_H^i, z_L^i$ , so it has to non-adaptively choose  $(x_L^i, x_H^i)_i$ . For each i < j, the probability  $\Pr[z_L^i = z_L^j]$  and  $\Pr[z_H^i = z_H^j]$  are bounded by  $2^{-\lambda}$ .

### Part B alternative proof P'' is a PRP.

Since F is a PRP, we can replace  $F_{k_1}, F_{k_2}, F_{k_3}$  by i.i.d. uniform  $f_1, f_2, f_3$  respectively.

$$x_{H} \rightarrow f_{1} - y_{H} \rightarrow M - z_{H} \rightarrow f_{2} \rightarrow w_{H}$$

$$x_{L} \rightarrow f_{3} \rightarrow w_{L}$$

Let  $x_L^i, x_H^i, y_H^i, z_L^i, z_H^i, w_L^i, w_H^i$  denote the input, output and intermediate values of the *i*-th query. W.l.o.g., all  $(x_L^i, x_H^i)$  are distinct.

Due to the randomness of  $f_1$ , with overwhelming probability,  $z_L^i \neq z_L^j \wedge z_H^i \neq z_H^j$  for any  $i \neq j$ . In such case, every output  $(w_L^i, w_H^i)$  is fresh random, and thus cannot be distinguished from a random permutation.

The intuition can be formalized. Define the following statements:

- $A_t$ : with overwhelming probability, for all  $i < j \le t$ ,  $z_L^i \ne z_L^j \land z_H^i \ne z_H^j$ .
- $B_t$ : the joint distribution of the first t outputs  $(w_L^i, w_H^i)_{i=1}^t$  is close to uniform.
- $C_t$ : the distribution of  $f_1$  conditioning on the first t outputs  $(w_L^i, w_H^i)_{i=1}^t$  is close to uniform

 $A_t \implies B_t$  follows directly from the randomness of  $f_2, f_3$ .

 $A_t \implies C_t$  also follows from the randomness of  $f_2, f_3$ . Due to the effect of  $f_2, f_3$ , the only leaked information of  $f_1$  is whether  $z_L^i$  equals  $z_L^j$  and whether  $z_H^i$  equals  $z_H^j$ . As claimed by  $A_t$ , such leakage is negligible.

 $C_{t-1} \Longrightarrow A_t$  follows from the randomness of  $f_1$ . For each j < t, if  $x_H^j = x_H^t$  then there is definitely no collision; if  $x_H^j \neq x_H^t$ , the randomness of  $y_H^t$  ensures  $z_L^t \neq z_L^j \land z_H^t \neq z_H^j$  with overwhelming probability.

# Problem 3.

Part A A counterexample is 3-round Feistel network.

**Part B** F' is a PRF.

When k is hidden,  $F(k,\cdot)$  is indistinguishable from a random function  $f(\cdot)$  under oracle access. Therefore, as a standard trick, it suffices to show that  $f'(x) = x \oplus f(x)$  is indistinguishable from a random function under oracle access. Note that the distribution of f' is identical to a random function, thus it is indistinguishable from a random function.

More formally, consider the following three oracles:

```
F'(k,\cdot) Given x, output F'(k,x) = x \oplus F(k,x). (k is a random key.) f' Given x, output f'(x) = x \oplus f(x). (f is a random function.) f Given x, output f(x). (f is a random function.)
```

The first two are indistinguishable because F being a PRF. The last two are indistinguishable because they are identical.

#### Part C F' is a PRP.

Consider the following three oracles:

```
F'(k,\cdot) Given x, output F'(k,x) = F(k_2, F(k_1,x)). (k = k_1 || k_2 \text{ is a random key.})

f' Given x, output f'(x) = f_2(f_1(x)). (f_1, f_2 \text{ are random permutations.})

f Given x, output f(x). (f \text{ is a random permutation.})
```

The first two are indistinguishable because F being a PRP. The last two are indistinguishable because they are identical.

### Part D F' is a PRP.

Consider the following three oracles:

```
F'(k,\cdot) Given x, output F'(k,x) = F(k_2, F(k_1,x)). (k = k_1 || k_2 \text{ is a random key.})

f' Given x, output f'(x) = f(f(x)). (f \text{ is a random function.})

f Given x, output f(x). (f \text{ is a random function.})
```

The first two are indistinguishable because F being a PRF. Thus it suffices to show the indistinguishability between the last two. (We also rely on the fact that a random function and a random permutation are indistinguishable.)

Without loss of generality, we assume the distinguisher never query the same input twice. The oracle f always returns a fresh random output upon a new query.

When the distinguisher is interacting with the oracle f', let  $x_i$  denote the i-th query, let  $y_i, z_i$  denote the corresponding intermediate value and output. For each  $t, x_t \notin \{x_1, \ldots, x_{t-1}, y_1, \ldots, y_{t-1}\}$  with overwhelming probability, thus  $y_t = f(x_t)$  is a fresh random value and  $y_t \notin \{x_1, \ldots, x_t, y_1, \ldots, y_{t-1}\}$  with overwhelming probability, then  $z_t = f(y_t)$  is a fresh random value.

The intuition can be formalized.

**Formalization 1.** Define the following statements:

- $A_t$ : with overwhelming probability,  $x_t \notin \{x_1, \ldots, x_{t-1}, y_1, \ldots, y_{t-1}\}$ .
- $B_t$ : with overwhelming probability,  $y_t \notin \{x_1, \dots, x_t, y_1, \dots, y_{t-1}\}.$
- $C_t$ : the joint distribution of the first t outputs  $z_1, \ldots, z_t$  is close to uniform.
- $D_t$ : the distribution of  $y_1, \ldots, y_t$  conditioning on  $x_1, \ldots, x_t, z_1, \ldots, z_t$  is close to uniform

 $D_{t-1} \implies A_t$ :  $x_t \notin \{x_1, \dots, x_{t-1}\}$  comes from the assumption of no duplicated queries.  $y_t \notin \{x_1, \dots, x_{t-1}\}$  w.h.p. follows from  $D_{t-1}$ .

 $A_t \implies B_t$  follows directly from the randomness of f.

 $B_t + C_{t-1} \implies C_t$  also follows from the randomness of f.

 $D_{t-1} + B_t \implies D_t$  because  $x_t$  is determined by  $x_1, \ldots, x_{t-1}, z_1, \ldots, z_{t-1}$ , thus revealing no information; and  $z_t$  is just fresh randomness.

Formalization 2. The oracle f' can be implemented by the following program, if parameter threshold is set as 0.

Initialize an empty table f in the setup phase.

```
if f(x_i) is defined and i \ge \text{threshold}
let y_i \leftarrow f(x_i)
otherwise, sample y_i randomly and set f(x_i) = y_i
```

otherwise, sample  $y_i$  randomly and set  $f(x_i) = y_i$ 

if  $f(y_i)$  is defined and  $i \ge$ threshold let  $y_i \leftarrow f(z_i)$ 

Upon receiving input  $x_i$ ,

otherwise, sample  $z_i$  randomly and set  $f(y_i) = z_i$ 

output  $z_i$ 

If threshold is set to be the number of queries, then the program always return i.i.d. random outputs.

Comparing the program when threshold = t and threshold = t+1, the only difference is in the t-th query. When  $x_t$  is received,  $f(x_t)$  is undefined with overwhelming probability because  $y_1, \ldots, y_{t-1}$  are completely hidden from the distinguisher. Then as a consequence,  $y_t$  is random and  $f(y_t)$  is undefined with overwhelming probability. In short, the program parameterized by threshold = t and the program parameterized by threshold = t and the program probability. By a hybrid argument, the program parameterized by threshold = t0 and the program parameterized by threshold = t1 perform exactly the same with overwhelming probability.

TODO: update proof

# Problem 4.

#### **Part A.** It is a PRP.

The proof is almost the same as the proof for independent-key 3-round Feistel.

The first step is to replace the PRFs by random functions. No PPT adversary can distinguish

$$\text{Feistel}_{f(k_1,\cdot),f(k_2,\cdot),f(k_2,\cdot)}$$
 from  $\text{Feistel}_{F_1(\cdot),F_2(\cdot),F_2(\cdot)}$ ,

by having oracle access to them. (As we have shown in problem 10.) So it suffices to show no PPT adversary can distinguish

$$\text{Feistel}_{F_1(\cdot),F_2(\cdot),F_2(\cdot)}$$

from a random permutation over  $\{0,1\}^{2n}$ , by having oracle access.

W.l.o.g., we can assume the adversary never makes duplicate queries. Under such assumption, when the oracle is a random permutation, it gets a random 2n-bit-string whenever it queries the oracle. We need to show that the same happens when the oracle is  $\text{Feistel}_{F_1(\cdot),F_2(\cdot),F_2(\cdot)}$ .

Let  $(x_0^i, x_1^i)$  denotes the adversary's *i*-th query, let  $(x_3^i, x_4^i)$  denotes the corresponding output, and let  $x_2^i$  denote the corresponding intermediate value.

Consider statement  $P_t$ : with probability 1 - negl(n), all of  $(P_t.1)$ ,  $(P_t.2)$ ,  $(P_t.3)$  hold.

- (P<sub>t</sub>.1) "There is no collision on  $x_2, x_3$ ." That is,  $x_2^1, x_2^1, \dots, x_2^t, x_3^t$  are all distinct values.
- ( $P_t$ .2) "The *i*-th output is uniform." That is, conditioning on  $(x_0^i, x_1^i, x_3^i, x_4^i)_{i < t}$ , the conditional distribution of  $(x_3^t, x_4^t)$  is close to uniform.
- ( $P_t$ .3) " $F_1$  is hidden." That is, conditioning on  $(x_0^i, x_1^i, x_3^i, x_4^i)_{i \le t}$ , the conditional distribution of  $F_1$  is close to uniform.

We prove that statement  $P_t$  holds for all  $t \leq \text{poly}(n)$  inductively.

Assume  $P_{t-1}$  holds. Due to  $(P_{t-1}.3)$ ,  $x_2^t$  collides with a previous value with at most negligible probability. Since  $x_2^t$  does not collides with any previous  $x_2^i$  or  $x_3^i$ , thus  $F_2(x_2^t)$  is a fresh random value. Since  $x_3^t$  is one-time padded by  $F_2(x_2^t)$ , the value of  $x_3^t$  does not collides with any previous  $x_2^i$  or  $x_3^i$  with overwhelming probability. (So  $(P_t.1)$  holds.) Then  $F_2(x_3^t)$  is also a fresh random value.

Since  $F_2(x_2^t)$ ,  $F_2(x_3^t)$  are fresh random values, the distribution of  $(x_3^t, x_4^t)$  is uniform, even conditioning on previous information. (So  $(P_t.2)$  holds.)

In the t-th query, the only information about  $F_1$  is  $F_1(x_1^t)$ . But the information is perfectly hidden, because the output  $(x_3^t, x_4^t)$  is one-time padded by  $F_2(x_2^t), F_2(x_3^t)$ . (So  $(P_t.3)$  holds.)

**Part A alternative proof.** Here we present a more formal proof. W.l.o.g., we assume the distinguisher never makes duplicate queries. Let  $x_0^i, x_1^i, x_2^i, x_3^i, x_4^i$  denotes the input, intermediate value, output corresponding to the *i*-th query.

- Real world: The distinguisher has oracle access to Feistel<sub> $f(k_1,\cdot),f(k_2,\cdot),f(k_2,\cdot)$ </sub>.
- World 1: PRFs are replaced by random functions. The distinguisher has oracle access to  $\text{Feistel}_{F_1(\cdot),F_2(\cdot),F_2(\cdot)}$ .

- World 2:  $F_2$  is further replaced by a "random box". Upon a query, it will always sample fresh random output.
- World 3:  $x_2^i$  is computed, and is ignored.  $x_3^i, x_4^i$  are freshly uniformly sampled.
- Ideal world: The distinguisher has oracle access to a random function  $\{0,1\}^{2\lambda} \to \{0,1\}^{2\lambda}$ .

It is easy to argue that real world and World 1 are indistinguishable, World 2 and World 3 are identical, ideal world and World 3 are indistinguishable.

If  $F_2$  is never evaluated upon same input twice, it behaves exactly the same as a random box. Let Repeat denote the event that  $F_2$  is evaluated on some input twice. Then the advantage distinguishing World 1 and World 2 is no more than Pr[Repeat] (in World 1 or World 2 or World 3).

Repeat = 
$$\{x_r^i = x_s^j \text{ for some } i, j \in [T], r, s \in \{2, 3\} \text{ s.t. } (i, r) \neq (j, s)\}$$

It is easier to bound  $\Pr[\mathsf{Repeat}]$  in World 3. In World 3, fresh random  $x_3^i$  is unlikely to collide with other values. In World 3, the distinguisher learns no information about  $F_1$ , so the distinguisher can only make non-adaptive queries  $\{x_0^i, x_1^i\}_i$ . The randomness of  $F_1$  ensures  $x_2^i$  will not collides with other values with overwhelming probability.

Part B. It is not even a PRP, because

$$f_4((k_1, k_2), (x_0, x_1)) = (x_4, x_5) \implies f_4((k_1, k_2), (x_5, x_4)) = (x_1, x_0).$$