# OPRF from Input-Homomorphic Publicly Evaluable PRF

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#### Abstract

Oblivious pseudorandom function (OPRF), a counterpart of PRF in distributed setting, has become a ubiquitous primitive used in secure multi-party computation. In this work, we provide a generic construction of OPRF from publicly evaluable PRF with input-homomorphic property. By plugging PEPRF from efficient group action (EGA) to the generic construction, we obtain a OPRF with plausible post-quantum security.

Keywords: OPRF, homomorphic, PEPRF

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# 1 Introduction

Pseudorandom functions (PRFs) is a core cryptographic primitive with numerous applications. Two decades later, Freedman et al. [FIPR05] introduce its counterpart in the distributed setting, namely oblivious pseudorandom functions (OPRF). In a nutshell, OPRF is an interactive 2-party protocol between a client with input  $\{x_i\}_{i\in[n]}$ , and a server with secret key k. It allows the client learn  $\{F_k(x_i)\}_{i\in[n]}$  without anything else, and the server learns nothing. Formally, from the perspective of 2-party secure computation, an OPRF is a protocol securely fulfills the functionality  $\mathcal{F}_{\text{oprf}} = (\bot, \{F_k(x_i)\}_{i\in[n]})$ . OPRFs have shown to be a central primitive for building oblivious keyword search [FIPR05], private set intersection [KKRT16, HL10, CM20], secure data de-duplication [CCGS19], password-authenticated key exchange (PAKE) [JKK14, JKX18], private information retrieval [FIPR05], secure pattern matching [FHV13], cloud key management [JKR19].

So far, except the OPRFs based on symmetric primitives, all known efficient OPRF constructions rely on number-theoretical assumptions, which are not post-quantum secure. Recently, Boneh et al. [BKW20] propose an isogency-based OPRF. Albrecht et al. [ADDS21] present a lattice-based OPRF with *verifiable* property. These two constructions are post-quantum secure, but both them only serves as nice feasibility results, rather than practical results with good performance.

### 1.1 Motivation

Motivated by the state of affairs, we are intrigued to know:

Is there a generic construction of OPRF which admits for efficient instantiation?

#### 1.2 Our Contribution

In this work, we make positive answers to the aforementioned questions. We present a generic construction of OPRF from PEPRF with input-homomorphic property, then give an efficient post-quantum secure instantiation from EGA.

### 2 Preliminaries

#### 2.1 Notation

We use  $\kappa$  and  $\lambda$  to denote the computational and statistical parameter respectively. Let  $\mathbb{Z}_n$  be the set  $\{0,1,\ldots,n-1\}$ ,  $\mathbb{Z}_n^* = \{x \in \mathbb{Z}_n \mid \gcd(x,n)=1\}$ . We use [n] to denote the set  $\{1,\ldots,n\}$ , and use  $\mathsf{Perm}[n]$  to denote all the permutation over the set  $\{1,\ldots,n\}$ . We assume that every set X has a default order (e.g. lexicographical order), and write it as  $X = \{x_1,\ldots,x_n\}$ . For a set X, we use |X| to denote its size and use  $x \stackrel{\mathbb{R}}{\leftarrow} X$  to denote sampling x uniformly at random from X. We use  $(x_1,\ldots,x_n)$  to denote a vector, whose ith element is  $x_i$ . A function is negligible in  $\kappa$ , written  $\mathsf{negl}(\kappa)$ , if it vanishes faster than the inverse of any polynomial in  $\kappa$ . A probabilistic polynomial time (PPT) algorithm is a randomized algorithm that runs in polynomial time.

#### 2.2 Weak Pseudorandom EGA

We begin by recalling the definition of a group action.

**Definition 2.1** (Group Actions). A group  $\mathbb{G}$  is said to *act on* a set X if there is a map  $\star : \mathbb{G} \times X \to X$  that satisfies the following two properties:

- 1. Identity: if e is the identity element of  $\mathbb{G}$ , then for any  $x \in X$ , we have  $e \star x = x$ .
- 2. Compatibility: for any  $g, h \in \mathbb{G}$  and any  $x \in X$ , we have  $(gh) \star x = g \star (h \star x)$ .

From now on, we use the abbreviated notation  $(\mathbb{G}, X, \star)$  to denote a group action. If  $(\mathbb{G}, X, \star)$  is a group action, for any  $g \in \mathbb{G}$  the map  $\phi_g : x \mapsto g \star x$  defines a permutation of X.

We then define an effective group action (EGA) [AFMP20] as follows.

**Definition 2.2** (Effective Group Actions). A group action  $(\mathbb{G}, X, \star)$  is *effective* if the following properties are satisfied:

- 1. The group  $\mathbb{G}$  is finite and there exist PPT algorithms for:
  - (a) Membership testing, i.e., to decide if a given bit string represents a valid group element in G.
  - (b) Equality testing, i.e., to decide if two bit strings represent the same group element in G.
  - (c) Sampling, i.e., to sample an element g from a uniform (or statistically close to) distribution on  $\mathbb{G}$ .
  - (d) Operation, i.e., to compute gh for any  $g, h \in \mathbb{G}$ .
  - (e) Inversion, i.e., to compute  $g^{-1}$  for any  $g \in \mathbb{G}$ .
- 2. The set X is finite and there exist PPT algorithms for:
  - (a) Membership testing, i.e., to decide if a bit string represents a valid set element.
  - (b) Unique representation, i.e., given any set element  $x \in X$ , compute a string  $\hat{x}$  that canonically represents x.
- 3. There exists a distinguished element  $x_0 \in X$ , called the origin, such that its bit-string representation is known.
- 4. There exists an efficient algorithm that given (some bit-string representations of) any  $g \in \mathbb{G}$  and any  $x \in X$ , outputs  $g \star x$ .

**Definition 2.3** (Weak Pseudorandom EGA). A group action  $(G, X, \star)$  is weakly pseudorandom if the family of efficiently commutable permutation  $\{\phi_g : X \to X\}_{g \in G}$  is weakly pseudorandom, i.e., there is no PPT adversary that can distinguish tuples of the form  $(x_i, g \star x_i)$  from  $(x_i, u_i)$  where  $g \overset{\mathbb{R}}{\leftarrow} \mathbb{G}$  and each  $x_i, u_i \overset{\mathbb{R}}{\leftarrow} X$ .

### 2.3 MPC in the Semi-honest Model

We use the standard notion of security in the presence of semi-honest adversaries. Let  $\Pi$  be a two-party protocol for computing the function  $f(x_1, x_2)$ , where party  $P_i$  has input  $x_i$ . We define security in the following way. For each party P, let  $\text{View}_P(x_1, x_2)$  denote the view of party P during an honest execution of  $\Pi$  on inputs  $x_1$  and  $x_2$ . The view consists of P's input, random tape, and all messages exchanged as part of the  $\Pi$  protocol.

**Definition 2.4.** 2-party protocol  $\Pi$  securely realizes f in the presence of semi-honest adversaries if there exists a simulator Sim such that for all inputs  $x_1, x_2$  and all  $i \in \{1, 2\}$ :

$$Sim(i, x_i, f(x_1, x_2)) \approx_c View_{P_i}(x_1, x_2)$$

Roughly speaking, a protocol is secure if the party with  $x_i$  learns no more information other than  $f(x_1, x_2)$  and  $x_i$ .

## 3 Review of Pseudorandom Function

In this section, we first recall the notion of standard pseudorandom functions (PRFs) [GGM86], then recap its counterpart in the public-key setting, namely publicly evaluable pseudorandom functions [CZ14].

### 3.1 Standard PRF

**Definition 3.1** (PRF). A family of PRFs consists of three polynomial-time algorithms as follows:

• Setup(1<sup> $\kappa$ </sup>): on input a security parameter  $\kappa$ , outputs public parameter pp specifying (F, K, X, Y), where  $F: K \times X \to Y$  is a family of keyed functions, K is the key space, X is domain, and Y is range.

- KeyGen(pp): on input pp, outputs a secret key  $k \stackrel{\text{R}}{\leftarrow} K$ .
- Eval(k,x): on input  $k \in K$  and  $x \in X$ , outputs  $y \leftarrow F(k,x)$ . For notation convenience, we will write F(k,x) as  $F_k(x)$  interchangeably.

Correctness: For any  $\kappa \in \mathbb{N}$ , any  $pp \leftarrow \mathsf{Setup}(1^{\kappa})$  and any  $k \leftarrow \mathsf{KeyGen}(pp)$ , we have  $F_{sk}(x) = \mathsf{Eval}(k, x)$ . **Pseudorandomness.** Let  $\mathcal{A}$  be an adversary against PRFs and define its advantage as:

$$\mathsf{Adv}_{\mathcal{A}}(\kappa) = \Pr \left[ \begin{matrix} & pp \leftarrow \mathsf{Setup}(1^\kappa); \\ \beta' = \beta: & k \leftarrow \mathsf{KeyGen}(pp); \\ \beta \leftarrow \{0,1\}; \\ \beta' \leftarrow \mathcal{A}^{\mathcal{O}_{\mathsf{ror}}(\beta,\cdot)}(\kappa); \end{matrix} \right] - \frac{1}{2},$$

where  $\mathcal{O}_{\mathsf{ror}}(\beta,\cdot)$  denotes the real-or-random oracle controlled by  $\beta$ , i.e.,  $\mathcal{O}_{\mathsf{ror}}(0,x) = F_k(x)$ ,  $\mathcal{O}_{\mathsf{ror}}(1,x) = \mathsf{H}(x)$  (here  $\mathsf{H}$  is chosen uniformly at random from all the functions from D to  $R^1$ ).  $\mathcal{A}$  can adaptively access the oracle  $\mathcal{O}_{\mathsf{ror}}(\beta,\cdot)$  polynomial many times. We say that F is pseudorandom if for any PPT adversary  $\mathsf{Adv}_{\mathcal{A}}(\kappa)$  is negligible in  $\kappa$ . We refer to such security as full PRF security.

Sometimes the full PRF security is not needed and it is sufficient if the function cannot be distinguished from a uniform random one when challenged on random inputs. The formalization of such relaxed requirement is weak pseudorandomness, which is defined the same way as pseudorandomness except that the inputs of oracle  $\mathcal{O}_{ror}(\beta,\cdot)$  are uniformly chosen from X by the challenger instead of adversarially chosen by  $\mathcal{A}$ . PRFs that satisfy weak pseudorandomness are referred to as weak PRFs.

Remark 3.1. In a standard PRF, it is always harmless to explicitly introduce a public key, which includes the information related to secret key that can be made public. For example, in the Naor-Reingold PRF [NR04] based on the DDH assumption:  $\mathsf{F}_{\vec{a}}(x) = (g^{a_0})^{\prod_{x_i=1}a_i}$ , where  $\vec{a} = (a_0, a_1, \ldots, a_n) \in \mathbb{Z}_p^n$  is the secret key,  $g^{a_i}$  for  $1 \leq i \leq n$  can be safely published as the public key. If no information can be made public, one can simply set  $pk = \bot$ . Looking ahead, such treatment allows us get a unified syntax of PRF in both minicrypt and cryptmania.

### 3.2 Publicly Evaluable PRF

**Definition 3.2** (Publicly Evaluable PRFs). A family of PEPRFs consists of four polynomial-time algorithms as follows:

- Setup(1<sup> $\kappa$ </sup>): on input a security parameter  $\kappa$ , output public parameter pp specifying (F, PK, SK, X, Y, L, W), where  $F: SK \times X \to Y \cup \bot$  is family of keyed functions, SK is the secret key space, PK is the corresponding public key space, X is domain, Y is range,  $L \subseteq X$  is a NP language defined by some hard relation  $R_L$ , and W is the space of associated witnesses. Here, we assume that  $R_L$  is efficiently sampleable, that is, there exists a PPT algorithm SampRel which on input random coins r, output a random tuple  $(x, w) \in R_L$ .
- KeyGen(pp): on input pp, output a secret key sk and an associated public key pk.
- PrivEval(sk, x): on input sk and  $x \in X$ , output  $y \in Y \cup \bot$ .
- PubEval(pk, x, w): on input pk and  $x \in L$  together with a witness  $w \in W$ , output  $y \in Y$ .

**Correctness:** For any  $\lambda \in \mathbb{N}$ , any  $pp \leftarrow \mathsf{Setup}(1^{\lambda})$  and any  $(pk, sk) \leftarrow \mathsf{KeyGen}(pp)$ , we have:

$$\forall x \in X: \qquad F_{sk}(x) = \mathsf{PrivEval}(sk,x)$$
 
$$\forall x \in L \text{ with a witness } w: \quad F_{sk}(x) = \mathsf{PubEval}(pk,x,w)$$

<sup>&</sup>lt;sup>1</sup>To efficiently simulate access to a uniformly random function H from D to R, one may think of a process in which the adversary's queries to  $\mathcal{O}_{ror}(1,\cdot)$  are "lazily" answered with independently and randomly chosen elements in R, while keeping track of the answers so that queries made repeatedly are answered consistently.

<sup>&</sup>lt;sup>2</sup>In a standard PRF, it is also harmless to explicitly introduce a public key, which includes the information related to secret key that can be made public. For example, in the Naor-Reingold PRF [NR04] based on the DDH assumption:  $\mathsf{F}_{\vec{a}}(x) = (g^{a_0})^{\prod_{x_i=1}a_i}$ , where  $\vec{a} = (a_0, a_1, \dots, a_n) \in \mathbb{Z}_p^n$  is the secret key,  $g^{a_i}$  for  $1 \le i \le n$  can be safely published as the public key. If no information can be made public, one can always assume  $pk = \{\bot\}$ .

**Input homomorphic:** Suppose  $L \subseteq X$  and Y are finite abelian groups. F is input homomorphic if for any  $x_1, x_2 \in L$  we have  $F_{sk}(x_1 + x_2) = F_{sk}(x_1) + F_{sk}(x_2)$ .

Weak pseudorandomness: Let  $\mathcal{A}$  be an adversary against PEPRFs and define its advantage as:

$$\mathsf{Adv}_{\mathcal{A}}(\lambda) = \Pr \left[ \begin{array}{c} pp \leftarrow \mathsf{Setup}(1^{\lambda}); \\ (pk, sk) \leftarrow \mathsf{KeyGen}(pp); \\ \beta = \beta': \quad \left\{ r_i^* \xleftarrow{\mathbb{R}} R, (x_i^*, w_i^*) \leftarrow \mathsf{SampRel}(r_i^*) \right\}_{i=1}^{p(\lambda)}; \\ \beta \leftarrow \left\{ 0, 1 \right\}; \\ \beta' \leftarrow \mathcal{A}(pp, pk, \{x_i^*, \mathcal{O}_{\mathsf{ror}}(\beta, x_i^*) \}_{i=1}^{p(\lambda)}); \end{array} \right] - \frac{1}{2},$$

where  $p(\kappa)$  is any polynomial,  $\mathcal{O}_{ror}(\beta,\cdot)$  is the real-or-random oracle. We say that F is weak pseudorandom if for any PPT adversary  $\mathcal{A}$  its advantage function  $Adv_{\mathcal{A}}(\kappa)$  is negligible in  $\kappa$ .

Looking ahead, to simplify the security proof we will use a one-time flavor definition that  $\mathcal{A}$  only queries  $\mathcal{O}_{ror}$  once when arguing the security of PEPRF constructions. A standard hybrid argument shows that such seemingly restricted definition is actually equivalent to the standard multi-query definition.

Remark 3.2. Different from standard PRFs, PEPRFs require the existence of a language  $L \subseteq X$  as well as a public evaluation algorithm. Due to this strengthening on functionality, we cannot hope to achieve standard pseudorandomness, and hence weak pseudorandomness is the best we can achieve.

# 4 OPRF from Input-Homomorphic PEPRF

#### 4.1 Definition of OPRF

In Figure 1, we adapt the definion of multi-point OPRF [CM20] under unified PRF syntax.

**Parameters:** A PRF  $F: SK \times X \to Y$ , and two parties: server  $P_1$  and client  $P_2$  Functionality:

- 1. Generate a key pair (pk, sk) for PRF. Give (pk, sk) to  $P_1$  and give pk to  $P_2$ .
- 2. Wait for input  $(x_1, \ldots, x_n)$  from  $P_2$ . Give  $(F_{sk}(x_1), \ldots, F_{sk}(x_n))$  to  $P_2$

Figure 1: Ideal functionality  $\mathcal{F}_{mqRPMT}$  for multi-point OPRF

## 4.2 Construction from Input-Homomorphic PEPRF

In Figure 2, we show how to build OPRF  $F: SK \times \{0,1\}^{\ell} \to Y$  from PEPRF  $G: SK \times X \to Y$  and cryptographic hash function  $\mathsf{H}: \{0,1\}^{\ell} \to L$ .

Remark 4.1. Looking ahead,  $\mathsf{H}:\{0,1\}^\ell\to L$  will be modeled as a random oracle. This implicitly requires L is dense.

**Correctness.** The correctness of the above OPRF construction follows from that of input-homomorphic PEPRF.

**Theorem 4.1.** The OPRF protocol described in Figure 2 is secure in the semi-honest model assuming H is a random oracle and F is a family of PEPRFs.

*Proof.* In the above construction  $F_{sk}(x): SK \times \{0,1\}^{\ell} \to Y$  is defined as  $G_{sk}(\mathsf{H}(x))$ . By assuming  $\mathsf{H}$  is a random oracle, we can reduce the pseudorandomness of F to the weak pseudorandomness of G. In other words,  $\mathsf{H}$  amplifies weak pseudorandomness to standard pseudorandomness by leveraging programmability.

We then prove the above construction is indeed a secure OPRF protocol. We exhibit simulators  $\mathsf{Sim}_{P_1}$  and  $\mathsf{Sim}_{P_2}$  for simulating corrupt  $P_1$  and  $P_2$  respectively, and argue the indistinguishability of the simulated transcript from the real execution.

### Parameters:

- Common input: PEPRF  $G: SK \times X \to Y$ , hash function  $\mathsf{H}: \{0,1\}^\ell \to L$ .
- Input of server  $P_1$ : security parameter  $\kappa$ .
- Input of client  $P_2$ :  $X = \{x_1, \dots, x_n\} \subseteq \{0, 1\}^{\ell}$

### **Protocol:**

- 1.  $P_1$  runs  $pp \leftarrow \text{PEPRF.Setup}(1^{\kappa})$  and  $(pk, sk) \leftarrow \text{PEPRF.KeyGen}(pp)$ , then sends pp and pk to  $P_2$ .
- 2.  $P_2$  runs  $\mathsf{SampRel}(pp)$  independently n times and obtains n tuples  $(z_i, w_i)$ , then sends  $\{\mathsf{H}(x_i) + z_i\}_{i \in [n]}$  to  $P_1$ .
- 3.  $P_1$  computes  $v_i \leftarrow \{F_{sk}(\mathsf{H}(x_i) + z_i)\}_{i \in [n]}$  and sends them to  $P_2$ .  $P_2$  computes  $y_i \leftarrow v_i \mathsf{PEPRF}.\mathsf{PubEval}(pk, z_i, w_i)$ .

$$G: SK \times X \to Y, \ \mathsf{H}: \{0,1\}^{\ell} \to X$$

$$P_1 \ (\mathsf{server}) \\ \kappa \\ pp \leftarrow \mathsf{PEPRF.Setup}(1^\kappa) \\ (pk,sk) \leftarrow \mathsf{PEPRF.KeyGen}(pp)$$

$$\xrightarrow{\left\{\mathsf{H}(x_i) + z_i\right\}_{i \in [n]}} \qquad (z_i,w_i) \leftarrow \mathsf{SampRel}(pp)$$

$$\xrightarrow{\left\{v_i \leftarrow F_{sk}(\mathsf{H}(x_i) + z_i)\right\}_{i \in [n]}} \quad \alpha_i \leftarrow \mathsf{PEPRF.PubEval}(pk,z_i,w_i)$$

$$y_i \leftarrow v_i - \alpha_i$$

Figure 2: OPRF from input-homomorphic PEPRF

Security against corrupt server.  $\mathsf{Sim}_{P_1}$  simulates the view of corrupt server  $P_1$ , which consists of  $P_1$ 's randomness, input, output and received messages. We formally show  $\mathsf{Sim}_{P_1}$ 's simulation is indistinguishable from the real execution via a sequence of hybrid transcripts.

Hybrid<sub>0</sub>:  $P_1$ 's view in the real protocol.

Hybrid<sub>1</sub>: Given  $P_1$ 's input X,  $\mathsf{Sim}_{P_1}$  chooses the randomness for  $P_1$  (i.e., picks  $r_i \overset{\mathbb{R}}{\leftarrow} R$  to sample  $(z_i, w_i) \in \mathsf{R}_L$ ), and simulates with the knowledge of X.  $\mathsf{Sim}_{P_1}$  outputs  $(\mathsf{H}(x_i) + z_i, \dots, \mathsf{H}(x_n) + z_n)$ . Clearly,  $\mathsf{Sim}_{P_1}$ 's simulated view in Hybrid<sub>1</sub> is identical to  $P_1$ 's real view.

Hybrid<sub>2</sub>:  $\mathsf{Sim}_{P_1}$  simulates without the knowledge of X.  $\mathsf{Sim}_{P_1}$  picks  $r_i \overset{\mathbb{R}}{\leftarrow} R$  to sample  $(z_i, w_i) \in \mathsf{R}_L$ , then outputs  $(z_1, \ldots, z_n)$ .

Clearly, the simulated view in Hybrid<sub>1</sub> and Hybrid<sub>2</sub> are identical due to  $z_i$  distributes uniformly at random over L.

Security against corrupt client.  $\mathsf{Sim}_{P_2}$  simulates the view of corrupt client  $P_2$ , which consists of  $P_2$ 's randomness, input, output and received messages. We formally show  $\mathsf{Sim}_{P_2}$ 's simulation is indistinguishable from the real execution via a sequence of hybrid transcripts.

Hybrid<sub>0</sub>:  $P_2$ 's view in the real protocol.

Hybrid<sub>1</sub>: Given  $P_2$ 's input  $x_i$ , randomness  $r_i$ , and output pk and  $y_i$ ,  $Sim_{P_2}$  samples  $(z_i, w_i) \in \mathsf{R}_L$ , then simulates the received messages  $v_i := y_i + \mathsf{PEPRF}.\mathsf{PubEval}(pk, z_i, w_i)$ .

Clearly, the simulated view in  $\operatorname{Hybrid}_1$  and  $\operatorname{Hybrid}_2$  are identical due to the publicly evaluable property of the underlying PEPRF.

This proves the theorem.

# 5 Instantiations of Input-Homomorphic PEPRF

It remains to give efficient instantiations of input-homomorphic PEPRF.

# 5.1 Input-Homomorphic PEPRF from the DDH Assumption

- Setup(1<sup> $\kappa$ </sup>): on input a security parameter  $\kappa$ , runs ( $\mathbb{G}, g, p$ )  $\leftarrow$  GroupGen(1<sup> $\kappa$ </sup>), output public parameter pp specifying (F, PK, SK, X, Y, L, W), where  $PK = \mathbb{G}, SK = \mathbb{Z}_p, X = L = Y = \mathbb{G}, W = \mathbb{Z}_p, F : SK \times X \to Y$  is defined as  $F_{sk}(x) := x^{sk}, (x, w) \in \mathsf{R}_L$  iff  $g^w = x$ .
- KeyGen(pp): on input pp, picks secret key  $sk \stackrel{\mathbb{R}}{\leftarrow} \mathbb{Z}_p$  and computes public key  $pk \leftarrow g^{sk}$ .
- PrivEval(sk, x): on input sk and  $x \in X$ , output  $y \leftarrow x^{sk}$ .
- PubEval(pk, x, w): on input pk and  $x \in L$  together with a witness  $w \in W$ , output  $y \leftarrow pk^w$ .

Input homomorphic.  $\forall x_1, x_2 \in X$ , we have  $(x_1 \cdot x_2)^{sk} = x_1^{sk} \cdot x_2^{sk}$ .

**Theorem 5.1.** The above construction is weakly pseudorandom if the DDH assumption holds.

*Proof.* The weak pseudorandomness immediately follows from the DDH assumption. We omit the security proof here due to its straightforwardness.  $\Box$ 

The above construction and security proof directly carry to the weak-pusudorandom EGA setting.

### 5.2 Input-Homomorphic PEPRF from the LWE Assumption

LWE is parameterized by positive integers n and q, and an error distribution  $\chi$  over  $\mathbb{Z}$ . For concreteness, n and q can be thought of as roughly the same as in SIS, and  $\chi$  is usually taken to be a discrete Gaussian of width  $\alpha q$  for some  $\alpha < 1$ , which is often called the relative "error rate".

**Definition 5.1** (LWE distribution). For a vector  $s \in \mathbb{Z}_q^n$  called the secret, the LWE distribution  $A_{s,\chi}$  over  $\mathbb{Z}_q^n \times \mathbb{Z}_q$  is sampled by choosing  $a \in \mathbb{Z}_q^n$  uniformly at random, choosing  $e \stackrel{\mathbb{R}}{\leftarrow} \chi$ , and outputting  $(a, b = \langle s, a \rangle + e \mod q)$ .

**Definition 5.2** (Decision-LWE<sub> $n,q,\chi,m$ </sub>). Given m independent samples  $(a_i,b_i) \in \mathbb{Z}_q^n \times \mathbb{Z}_q$  where every sample is distributed according to either: (1)  $A_{s,\chi}$  for a uniformly random  $s \in \mathbb{Z}_q$  (fixed for all samples), or (2) the uniform distribution, distinguish which is the case (with non-negligible advantage).

We give a LWE-based PEPRF as below.

• Setup(1<sup> $\kappa$ </sup>): on input a security parameter  $\kappa$ , first chooses a uniformly random  $\mathbf{A} \in \mathbb{Z}_q^{n \times m}$ , output public parameter pp specifying (F, PK, SK, X, Y, L, W), where  $PK = \mathbb{Z}_q^m$ ,  $SK = \mathbb{Z}_q^n$ ,  $X = L = \mathbb{Z}_q^{n+1}$ ,  $W = \{0,1\}^{m+1}$ . Here the collection of  $\mathcal{NP}$  languages indexed by  $pk = \mathbf{u} \in \mathbb{Z}_q^m$  is defined as:

$$L_{\mathbf{u}} = \{\mathbf{x} = (\mathbf{c}, v) \in \mathbb{Z}_q^{n+1} | \exists \mathbf{w} = (\mathbf{a} \in \{0, 1\}^m, b \in \{0, 1\}) \text{ s.t. } \mathbf{c} = \mathbf{A} \cdot \mathbf{a}, v = \mathbf{u}^t \cdot \mathbf{a} + b \cdot \lfloor q/2 \rceil \}$$

 $F: SK \times X \to Y$  is defined as  $F_{\mathbf{s}}(\mathbf{x} = (\mathbf{c}, v)) := \mathsf{Decode}(v - \mathbf{s}^t \cdot \mathbf{c})$ , where  $\mathsf{Decode}(z)$  outputs "0" if z is closer to 0 or "1" if z is closer to  $\lfloor q/2 \rfloor$  modulo q.

- KeyGen(pp): on input pp, first picks a secret key  $\mathbf{s} \stackrel{\mathbb{R}}{\leftarrow} \mathbb{Z}_q^n$ , then picks  $\mathbf{e} \stackrel{\mathbb{R}}{\leftarrow} \chi^m$  and computes public key  $\mathbf{u} \leftarrow \mathbf{A}^t \cdot \mathbf{s} + \mathbf{e} \in \mathbb{Z}_q^m$ .
- PrivEval(s, x): on input s and  $\mathbf{x} = (\mathbf{c}, v)$ , output  $y \leftarrow \mathsf{Decode}(v \mathbf{s}^t \cdot \mathbf{c}) \in \{0, 1\}$ .
- PubEval( $\mathbf{u}, \mathbf{x}, \mathbf{w}$ ): on input public key  $\mathbf{u}$  and  $\mathbf{x} = (\mathbf{u}, v) \in L$  together with a witness  $\mathbf{w} = (\mathbf{a}, b) \in W$ , output  $y \leftarrow b$ .

Correctness. Note that  $\mathbf{s}^t \cdot \mathbf{c} - \mathbf{u}^t \cdot \mathbf{a} = \mathbf{s}^t \cdot \mathbf{A} \cdot \mathbf{a} - (\mathbf{A}^t \cdot \mathbf{s} + \mathbf{e})^t \cdot \mathbf{a} = \mathbf{e}^t \cdot \mathbf{a} \in \mathbb{Z}_q$ . The correctness holds as long as  $(\mathbf{e}^t \cdot \mathbf{a})/q$  is negligible in  $\kappa$ .

Input homomorphic. For any  $\mathbf{x}_1 = (\mathbf{A} \cdot \mathbf{a}_1, \mathbf{u}^t \cdot \mathbf{a}_1 + b_1 \cdot \lfloor q/2 \rceil)$  and  $\mathbf{x}_2 = (\mathbf{A} \cdot \mathbf{a}_2, \mathbf{u}^t \cdot \mathbf{a}_2 + b_2 \cdot \lfloor q/2 \rceil)$ , we have:

$$\begin{split} F_{\mathbf{s}}(\mathbf{x}_1 + \mathbf{x}_2) &= \operatorname{Decode}(v_1 - \mathbf{s}^t \cdot \mathbf{c}_1 + v_2 - \mathbf{s}^t \cdot \mathbf{c}_2) \\ &= \operatorname{Decode}(\mathbf{e}^t \cdot \mathbf{a}_1 + b_1 \lfloor q/2 \rceil + \mathbf{e}^t \cdot \mathbf{a}_2 + b_2 \lfloor q/2 \rceil) \\ &\approx \operatorname{Decode}((b_1 + b_2) \lfloor q/2 \rceil) \\ &= b_1 + b_2 \approx F_{\mathbf{s}}(\mathbf{x}_1) + F_{\mathbf{s}}(\mathbf{x}_2) \end{split}$$

Security. The weak pseudorandomness follows from the following hybrid arguments.

- Game<sub>0</sub>: the original security game.  $\mathcal{A}$  gets to see  $(pk = \mathbf{u}, \mathbf{x} \overset{\mathbb{R}}{\leftarrow} L_{\mathbf{u}}, k^*)$ , where  $k^* = b$  if  $\beta = 0$  and  $k^*$  is a random bit if  $\beta = 1$ .
- Game<sub>1</sub>: same as Game<sub>0</sub> except that  $\mathcal{CH}$  changes **u** from the LWE distribution to a uniform distribution over  $\mathbb{Z}_q^m$ . By the LWE assumption, we have  $\operatorname{Game}_0 \approx_c \operatorname{Game}_1$ .
- Game<sub>2</sub>: same as Game<sub>1</sub> except that  $\mathcal{CH}$  changes  $\mathbf{x}$  to a uniform distribution over  $\mathbb{Z}_q^{n+1}$ . By the leftover hash lemma, we have  $\mathrm{Game}_1 \approx_s \mathrm{Game}_2$ . Clearly, (even unbounded)  $\mathcal{A}$ 's advantage in  $\mathrm{Game}_2$  is zero.

### 5.3 Input-Homomorphic PEPRF from the Ring-LWR Assumption

Define ring  $R_q = \mathbb{Z}_q[X]/(X^n+1)$ ,  $R_p = \mathbb{Z}_p[X]/(X^n+1)$ , where  $q > p \ge 2$  are integers,  $n = 2^k$  for integer k > 0, and  $X^n + 1$  is the  $2^{k+1}$ -th cyclotomic polynomial.

Define the modular function  $[\cdot]_p: \mathbb{Z} \to \mathbb{Z}_p$  as  $[x]_p = x \mod p$ , and define the rounding function  $[\cdot]_p: \mathbb{Z}_q \to \mathbb{Z}_p$  as  $[x]_p = \lfloor \frac{p}{q} \cdot x \rceil \mod p$ . The definition can be extended naturally in a coefficient-wise manner on  $R_q$  and  $R_p$ .

• Setup(1<sup> $\kappa$ </sup>): define pp = (F, PK, SK, X, Y, L, W), where  $PK = R_p$ ,  $SK = R_q$ ,  $X = Y = L = R_p$ ,  $W = R_q$ .  $F: SK \times X \to Y$  is defined as  $F_{sk}(x) = \lfloor x \cdot sk \rceil_p$ . For public parameter  $a \in R_q$ ,  $(x, w) \in R_L$  iff  $x = \lfloor aw \rceil_p$ .

- KeyGen(pp): on input pp, select the secret key  $sk \leftarrow \chi(R_q)$ , where  $\chi$  is a sparse distribution over  $R_q$ , and set the public key as  $pk \leftarrow \lfloor a \cdot sk \rceil_p$ .
- PrivEval(sk, x): on input sk and  $x \in X$ , output  $y \leftarrow \lfloor x \cdot sk \rfloor_p$ .
- PubEval(pk, x, w): on input pk and  $x \in L$  with the witness  $w \in W$ , output  $y \leftarrow \lfloor pk \cdot w \rfloor_p$ .

Properties of the above ring-LWR based IHPEPRF construction:

**Correctness.** For any  $\kappa \in \mathbb{N}$ , any  $pp \leftarrow \mathsf{Setup}(1^{\kappa})$  and any  $(pk, sk) \leftarrow \mathsf{KeyGen}(pp)$ :

- $\forall x \in X$ :  $F_{sk}(x) = \lfloor x \cdot sk \rceil_p = \mathsf{PrivEval}(sk, x)$ .
- $\forall x \in L \text{ with witness } w$ :

$$\mathsf{PubEval}(pk,x,w) = \lfloor pk \cdot w \rceil_p = \lfloor \lfloor \lfloor \frac{p}{q} \cdot a \cdot sk \rceil \rfloor_p \cdot w \rceil_p,$$

$$F_{sk}(x) = \lfloor x \cdot sk \rceil_p = \lfloor \lfloor \lfloor \frac{p}{q} \cdot a \cdot w \rfloor \rfloor_p \cdot sk \rfloor_p$$

**Input homomorphic.** A bit tricky while dealing with the rounding function: it is not linear.

$$F_{sk}(x_1 + x_2) = \left[ \left\lfloor \frac{p}{q}(x_1 + x_2) \cdot sk \right\rceil \right]_p,$$

$$F_{sk}(x_1) + F_{sk}(x_2) = \left[ \left\lfloor \frac{p}{q} \cdot x_1 \cdot sk \right\rceil \right]_p + \left[ \left\lfloor \frac{p}{q} \cdot x_2 \cdot sk \right\rceil \right]_p.$$

If 
$$[x_1] = [x_2]$$
, then  $F_{sk}(x_1 + x_2) = F_{sk}(x_1) + F_{sk}(x_2)$ .

**Security.** Guaranteed by the ring LWR assumption:  $(a, |a \cdot s|_p) \approx (a, u)$ .

The above construction is floppy and could be problematic.

# 6 Concrete Post-quantum Secure OPRF

By plugging PEPRF from weak pseudorandom EGA in Section 5.1 to our generic construction, we obtain a OPRF with plausible post-quantum security.

We encounter a technical hurdle when plugging the instantiation in Section 5.2 to our generic construction. This is because the weak pseudorandomness of F is tied to special distribution over  $L_{\mathbf{u}}$ , not a uniform distribution. The consequence is that we can not assume there exists a random oracle from  $\{0,1\}^*$  to  $L_{\mathbf{u}}$  anymore. One attempt is to remove v from  $\mathbf{x}$  and only keep  $\mathbf{c}$ . However, this somehow requires a fuzzy and homomorphic randomness extractor, which again seems quite tough.

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# A Missing Definitions