

# Batch-OT with Optimal Rate

Zvika Brakerski<sup>1</sup>, Pedro Branco<sup>2</sup> Nico Döttling<sup>3</sup>, and Sihang Pu<sup>3</sup>

<sup>1</sup> Weizmann Institute of Science

<sup>2</sup> IT, IST - University of Lisbon

<sup>3</sup> CISPA Helmholtz Center for Information Security

**Abstract.** We show that it is possible to perform  $n$  independent copies of 1-out-of-2 oblivious transfer in two messages, where the communication complexity of the receiver and sender (each) is  $n(1 + o(1))$  for sufficiently large  $n$ . Note that this matches the information-theoretic lower bound. Prior to this work, this was only achievable by using the heavy machinery of rate-1 fully homomorphic encryption (Rate-1 FHE, Brakerski et al., TCC 2019).

To achieve rate-1 both on the receiver’s and sender’s end, we use the LPN assumption, with slightly sub-constant noise rate  $1/m^\epsilon$  for any  $\epsilon > 0$  together with either the DDH, QR or LWE assumptions. In terms of efficiency, our protocols only rely on linear homomorphism, as opposed to the FHE-based solution which inherently requires an expensive “bootstrapping” operation. We believe that in terms of efficiency we compare favorably to existing batch-OT protocols, while achieving superior communication complexity. We show similar results for Oblivious Linear Evaluation (OLE).

For our DDH-based solution we develop a new technique that may be of independent interest. We show that it is possible to “emulate” the binary group  $\mathbb{Z}_2$  (or any other small-order group) inside a prime-order group  $\mathbb{Z}_p$  in a *function-private* manner. That is,  $\mathbb{Z}_2$  operations are mapped to  $\mathbb{Z}_p$  operations such that the outcome of the latter do not reveal additional information beyond the  $\mathbb{Z}_2$  outcome. Our encoding technique uses the discrete Gaussian distribution, which to our knowledge was not done before in the context of DDH.

## 1 Introduction

Oblivious Transfer (OT) [34,20] is one of the most basic cryptographic primitives. In the simple 1-out-of-2 OT, a receiver holds a bit  $b \in \{0, 1\}$  and a sender holds two bits  $x_0, x_1$ . In the end of the protocol, the receiver should learn  $x_b$ , but nothing about  $x_{1-b}$ , and the sender should learn nothing about the value of  $b$ . In most applications, one OT is not enough and one is required to perform many OT operations in parallel. We let  $n$  denote the number of parallel executions. Various techniques have been developed to address this task of *batch-OT* [29,6,5]. For the most part, they involve a preprocessing “offline” phase where the parties generate random OT correlations.<sup>4</sup> Given such correlations, executing the OT protocol

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<sup>4</sup> That is, a protocol in which the receiver obtains  $b, x_b$  and the sender obtains  $x_0, x_1$ , where  $b, x_0, x_1$  are all (pseudo-)randomly sampled.

in the so-called “online phase” is computationally very simple. This approach is very useful for purposes of computational efficiency, since the offline phase can be carried out even before the actual inputs of the computation are known. However, in terms of communication complexity, there is an inherent cost, even just in the online phase, of  $n$  receiver bits and  $2n$  sender bits. In contrast, the insecure implementation only requires  $n$  bits to be sent from each party in a two-message protocol: the receiver sends its input, and the sender returns all of the appropriate  $x_b$  values. As always in cryptography, we wish to understand what is the “cost of privacy”, namely can we approach the information theoretic minimum without losing privacy. Note that we can only hope to achieve this for a sufficiently large  $n$ , due to the security parameter overhead.<sup>5</sup>

In prior work, Döttling et al. [19] showed that if the same receiver bit is used for multiple OT instances, then the sender’s response can be compressed to  $n(1 + o(1))$ , achieving an optimal amortized rate. This was shown under a variety of computational assumptions: Decisional Diffie-Hellman (DDH), Quadratic Residuosity (QR), or Learning with Errors (LWE). It was also shown by Brakerski et al. [10] and by Gentry and Halevi [23] that fully homomorphic encryption (FHE) can achieve optimal communication complexity, which in particular implies that under the LWE assumption, optimal rate batch-OT is achievable. However, the FHE-based protocol inherently requires the use of a computationally exorbitant “bootstrapping” mechanism in order to compress the receiver’s message.

## 1.1 Our Contribution

We show that optimal-rate<sup>6</sup> batch-OT can be achieved from various computational assumptions, and without giving up on computational efficiency. In particular, we require the LPN assumption with a small-inverse-polynomial noise<sup>7</sup>, in addition to one of the assumptions DDH, QR or LWE. In terms of computational cost, our protocol does not require heavy operations such as bootstrapping and relies on linear homomorphism only. We believe that in terms of overall cost it compares favorably even with random-OT based methods. All of our results are in the semi-honest (honest-but-curious) setting.

We further extend our results to the task of Oblivious Linear Evaluation (OLE) [30,14,24,12], where the sender holds a linear function over a ring and the receiver holds an input for the function, and we wish for the receiver to

<sup>5</sup> In more detail, since 2-message OT implies a public-key encryption scheme, the messages must have length that relates to the security parameter of the underlying computation assumption. This is the case even for single-bit OT.

<sup>6</sup> Achieving optimal rate (or any rate above  $1/2$ ) seems to involve a “phase-transition” and should be viewed as more than a “constant factor” improvement. For example, OT beyond this threshold implies the existence of lossy trapdoor functions (see discussion in [19], Section 6.3). Therefore one could expect such a protocol to inherently be heavier on public-key operations.

<sup>7</sup> This is still a regime where LPN alone is not known to imply public-key encryption.

learn the output on its input and nothing more, and the sender learns nothing as usual. OLE has been shown to be useful in various settings [27,14].

Our techniques rely mostly on linear homomorphism, namely on the ability to evaluate linear functions on encrypted data (see Section 2 below). Notably, we require a linearly homomorphic scheme over  $\mathbb{Z}_2$  (more generally  $\mathbb{Z}_q$  for OLE) where the evaluation is *function-private*. Namely, the output ciphertext should not reveal any information about the linear function that was evaluated. This was not known to be achievable from DDH prior to this work, and we introduce a new technique that we believe may be of independent interest. The reason for this is that DDH works “natively” over the group  $\mathbb{Z}_p$  where  $p$  is a super-polynomially large prime. Furthermore, we only have access to the  $\mathbb{Z}_p$  elements in the exponent of a group generator  $g$ . Indeed, one can encode  $0 \rightarrow g^0$ ,  $1 \rightarrow g^1$ , and linear  $\mathbb{Z}_2$  homomorphism will follow in the sense that after applying a linear function in the exponent, we obtain  $g^x$ , where  $x \pmod{2}$  is the desired  $\mathbb{Z}_2$  output. This creates two obstacles: first we need to be able to efficiently map  $g^x \rightarrow x$ , which means that  $x$  must come from a polynomially-bounded domain, and second that recovering  $x$  reveals more information than just  $x \pmod{2}$ . We develop a new method to resolve this issue using *discrete Gaussian variables*. A technique that was used in the context of the LWE assumption but to the best of our knowledge not for DDH. We view this as an additional contribution of this work, which may find additional applications. In particular we show that it can be used to enhance the key-dependent-message security properties of the well-known encryption scheme [3].

For more details on all of our contributions, see the technical overview in Section 2.

## 1.2 Related Work

The communication complexity of OT has been extensively studied throughout the decades. Here we present a brief overlook of previous works.

*OT from Pseudorandom Correlations.* A recent line of research studies the feasibility of efficiently extending OTs in a *silent* manner [6,5]. In these works, a setup phase is performed to distributed some *shares* between the parties. These shares can later be expanded into random OT correlations. In the most efficient scheme [5] the setup phase can be performed in just two rounds assuming just a pseudorandom generator and an OT scheme. Using this scheme for performing the setup together with the standard transformations from random OT to chosen-input OT, [5] shows that  $n$  independent instances of OT for  $s$ -bit strings can be performed with communication complexity  $(2s + 1)n + o(n)$ . For bit OT, this yields a communication complexity  $3n + o(n)$  bits.

*Download rate-1 OT.* We say that an OT protocol has download rate 1 if the rate of the sender’s message is asymptotically close to 1. OT protocols with download rate 1 were presented in [19,21,15]. However, these protocols do not achieve upload rate 1, that is, the rate of the receiver’s message is far from being

1. Moreover, it is not clear how we can extend these protocols to achieve upload rate 1.

*Using rate-1 FHE.* As mentioned before, optimal-rate OT can be achieved using the recent scheme for rate-1 fully homomorphic encryption (FHE) of [10,23] together with (semi-honest) circuit-privacy techniques for FHE (e.g. [4]). However this can only be instantiated using LWE.

*Laconic OT.* Laconic OT [16,33,28,1] is a flavor of two-round OT where the first message sent by the receiver is sublinear (ideally polylogarithmically) in the size of its input. However, by a simple information-theoretical argument, the sender's message has size at least as large as the size of the sender's input. Note that, if this is not the case, then we would have an OT protocol with asymptotically better communication than an insecure OT protocol.

## 2 Technical Overview

### 2.1 Oblivious Transfer from Homomorphic Encryption

Our starting point is a textbook construction of oblivious transfer from simple homomorphic encryption schemes, such as ElGamal. For a cryptographic group  $\mathbb{G} = \langle g \rangle$  of prime order  $p$ , recall that an ElGamal public key is of the form  $\text{pk} = (g, h = g^x) \in \mathbb{G}^2$ , where  $x \xleftarrow{\$} \mathbb{Z}_p$  is the secret key. Ciphertexts are of the form  $c = (c_1, c_2) = (g^r, h^r \cdot g^b)$ , where  $r \xleftarrow{\$} \mathbb{Z}_p$  is uniformly random and  $b \in \{0, 1\}$  is the encrypted message. Given such a ciphertext  $c$ , the public key  $\text{pk}$  and two bits  $m_0, m_1 \in \{0, 1\}$ , anyone can homomorphically compute a new ciphertext  $c'$  which is distributed identically to a fresh encryption of  $m_b$ , by *homomorphically evaluating* the linear function  $f(x) = (1-x) \cdot m_0 + x \cdot m_1 = (m_1 - m_0) \cdot x + m_0$  on the ciphertext  $c$  and rerandomizing the resulting ciphertext. Note that if  $b \in \{0, 1\}$  is a bit, then it holds that  $f(b) = m_b$ . This homomorphic evaluation can be achieved by computing

$$\begin{aligned} c'_1 &\leftarrow g^{r^*} \cdot c_1^{m_1 - m_0} \\ c'_2 &\leftarrow h^{r^*} \cdot c_2^{m_1 - m_0} \cdot g^{m_0}, \end{aligned}$$

where  $r^* \xleftarrow{\$} \mathbb{Z}_p$  is chosen uniformly random. Note that it holds that

$$\begin{aligned} c'_1 &= g^{r^* + r \cdot (m_1 - m_0)} \\ c'_2 &= h^{r^* + r \cdot (m_1 - m_0)} \cdot g^{(m_1 - m_0) \cdot b + m_0} = h^{r^* + r \cdot (m_1 - m_0)} \cdot g^{m_b}. \end{aligned}$$

Since  $r^* \xleftarrow{\$} \mathbb{Z}_p$  is chosen uniformly random, it holds that  $r' = r^* + r \cdot (m_1 - m_0)$  is distributed uniformly random and we can conclude that  $c' = (c'_1, c'_2)$  is distributed identical to a fresh encryption of  $m_b$ . Since  $c'$  does not reveal more than the function value  $f(b) = m_b$ , we call the above homomorphic evaluation procedure function private.

This immediately implies an OT protocol: An OT-receiver holding a choice-bit  $b \in \{0, 1\}$  generates a pair  $(\text{pk}, \text{sk})$  of ElGamal public and secret keys, encrypts the bit  $b$  under  $\text{pk}$  and sends the resulting ciphertext to the OT-sender. The OT-sender, holding messages  $m_0, m_1$ , homomorphically computes a ciphertext  $c'$  encrypting  $m_b$  and sends  $c'$  back to the OT-receiver, who decrypts  $c'$  to  $m_b$ . Security against semi-honest senders follows from the IND-CPA security of ElGamal, whereas security against semi-honest receivers follows from the function privacy property established above.

## 2.2 Download-Rate Optimal String OT

While the above OT protocol is simple and efficient, it suffers from a very poor communication rate. While the receiver's message encrypts just a single bit, he needs to send 4 group elements, whereas the sender sends 2 group elements, each of size  $\text{poly}(\lambda)$ .

Döttling et al. [19] proposed a compression technique for *batched ElGamal ciphertexts* based on the share-conversion technique of [7]. A batched ElGamal ciphertext is of the form  $\mathbf{c} = (c_0, c_1, \dots, c_\ell) = (g^r, h_1^r \cdot g^{b_1}, \dots, h_\ell^r \cdot g^{b_\ell})$ , where  $\text{pk} = (g, h_1, \dots, h_\ell)$  is the corresponding public key and  $\text{sk} = (s_1, \dots, s_\ell)$  with  $h_i = g^{s_i}$  is the secret key. The compression technique of [19] keeps  $c_0$  and compresses each of the  $c_1, \dots, c_\ell$  into just a single bit. The idea is instead of sending each  $c_i \in \mathbb{G}$  (for  $i \geq 1$ ) in full, to first compute the distance  $d$  to the next pseudorandom *break-point* in  $\mathbb{G}$ , and then only send its parity  $d \bmod 2$ . The break points  $\mathcal{P} \subseteq \mathbb{G}$  are the set of all points  $h \in \mathbb{G}$  satisfying  $\text{PRF}_K(h) = 0^t$ , where  $\text{PRF} : \mathbb{G} \rightarrow \{0, 1\}^t$  is a pseudorandom function with a range of size  $2^t = \text{poly}(\lambda)$ . Thus, the distance  $d = d(c_i)$  of a group element  $c_i$  to the nearest break point is the smallest non-negative  $d$  such that  $c_i \cdot g^d \in \mathcal{P}$ . Given that neither  $c_i$  nor  $c_i \cdot g^{-1}$  is a breakpoint, we can recover the bit  $b_i$  from  $c_0 = g^r$ ,  $\beta_i = d(c_i) \bmod 2$  and the secret key component  $s_i$ . It was shown in [9] that for a given ciphertext  $c = (c_0, c_1, \dots, c_\ell)$ , the PRF-key  $K$  can be (efficiently) chosen such that all  $c_i$  are good, in the sense that neither  $c_i$  nor  $c_i \cdot g^{-1}$  is a breakpoint. This ensures that a receiver can recover the  $b_1, \dots, b_\ell$  from  $c' = (K, c_0, \beta_1, \dots, \beta_\ell)$ , where  $\beta_i = d(c_i) \bmod 2$ . Since all the  $\beta_i$  are bits, such a compressed ciphertext only has additive size-overhead consisting of  $K, c_0$ . For a sufficiently large  $\ell$ , this fixed overhead becomes insignificant and the ciphertext rate approaches 1.

The compressed batched ElGamal we've outlined leads to a batch bit-oblivious transfer protocol with *download-rate 1*: The receiver generates a key-pair  $\text{pk}, \text{sk}$  for batched ElGamal, and encrypts his choice-bits  $b_1, \dots, b_\ell$  into

$$\mathbf{c}_1 = \text{Enc}_{\text{pk}}(b_1, 0, \dots, 0), \dots, \mathbf{c}_\ell = \text{Enc}_{\text{pk}}(0, \dots, 0, b_\ell),$$

i.e.  $\mathbf{c}^{(i)}$  encrypts a vector which is  $b_i$  in index  $i$  and 0 everywhere else. The OT-receiver now sends  $\text{pk}, \mathbf{c}_1, \dots, \mathbf{c}_\ell$  to the OT-sender, whose input are messages  $(m_{1,0}, m_{1,1}), \dots, (m_{\ell,0}, m_{\ell,1})$ . Using circuit private homomorphic evaluation, the sender computes ciphertexts  $\mathbf{c}'_1, \dots, \mathbf{c}'_\ell$  encrypting  $(m_{1,b_1}, 0, \dots, 0), \dots, (0, \dots, 0, m_{\ell,b_\ell})$ . Homomorphically computing the sum of the ciphertexts

$\mathbf{c}'_1, \dots, \mathbf{c}'_\ell$ , we obtain a ciphertext  $\mathbf{c}'$  encrypting  $(m_{1,b_1}, \dots, m_{\ell,b_\ell})$ . Finally, compressing  $\mathbf{c}'$  with the compression technique outlined above we obtain a compressed ciphertext  $\bar{\mathbf{c}} = (K, c_0, \beta_1, \dots, \beta_\ell)$  which the OT-sender sends back to the OT-receiver, who can decrypt  $(m_{1,b_1}, \dots, m_{\ell,b_\ell})$ .

Note that the size of the sender’s message  $\bar{\mathbf{c}}$  in this batch OT-protocol is  $\text{poly}(\lambda) + \ell$ , which means that the *amortized communication cost* per bit-OT approaches 1 bit, and is therefore asymptotically optimal. Even in terms of concrete complexity this seems hard to beat, as the only additional information sent by the sender are the PRF key  $K$  and the ciphertext header  $c_0$ .

However, in terms of the upload rate, i.e. in terms of the size of the receiver’s message, this protocol performs poorly. Specifically, to encrypt  $\ell$  bits  $b_1, \dots, b_\ell$ , the receiver needs to send ciphertexts  $\mathbf{c}_1, \dots, \mathbf{c}_\ell$  of total size  $\ell^2 \cdot \text{poly}(\lambda)$ , which has a worse dependence on  $\ell$  than just repeating the simple protocol from the last paragraph  $\ell$  times.

Clearly, we need a mechanism to compress the receiver’s message. Applying the same ElGamal compression technique as for the sender’s message quickly runs into problems: Once an ElGamal ciphertext is compressed, the scheme loses its homomorphic capabilities, i.e. we cannot perform any further homomorphic operations on compressed ciphertexts and currently we don’t know if it is possible to publicly *decompress* such ciphertexts into “regular” ElGamal ciphertexts.

### 2.3 Our Approach: Recrypting the Receiver’s Message

Instead, our approach will be to encrypt the receiver’s message under a different encryption scheme, specifically one which achieves ciphertext rate approaching 1 but at the same time can be decrypted by the homomorphic capabilities of batched ElGamal. Specifically, the decryption procedure of this encryption scheme should be a linear function in the secret key. We can get an encryption scheme which almost fulfills these requirements from the Learning Parity with Noise (LPN) assumption. The LPN assumption states that for a random  $m \times n$  matrix  $\mathbf{A} \xleftarrow{\$} \mathbb{Z}_2^{m \times n}$ , a random vector  $\mathbf{s} \xleftarrow{\$} \mathbb{Z}_2^n$  and a  $\rho$ -Bernoulli distributed <sup>8</sup>  $\mathbf{e} \in \mathbb{Z}_2^m$ , it holds that

$$(\mathbf{A}, \mathbf{A}\mathbf{s} + \mathbf{e}) \approx_c (\mathbf{A}, \mathbf{u}),$$

where  $\mathbf{u} \xleftarrow{\$} \mathbb{Z}_2^m$  is chosen uniformly at random. This gives rise to the following simple symmetric-key encryption scheme with *approximate correctness*: Assume that  $\mathbf{A}$  is a fixed public parameter, the secret key is a uniformly random  $\mathbf{s} \xleftarrow{\$} \mathbb{Z}_2^n$ . To encrypt a message  $m \in \mathbb{Z}_2^m$ , we compute a ciphertext  $\mathbf{d} \leftarrow \mathbf{A}\mathbf{s} + \mathbf{e} + \mathbf{m}$ , where  $\mathbf{e} \in \mathbb{Z}_2^m$  is chosen via a  $\rho$ -Bernoulli distribution. To decrypt such a ciphertext, we compute  $\mathbf{m}' \leftarrow \mathbf{d} - \mathbf{A} \cdot \mathbf{s}$ .

Note that this scheme is only approximately correct in the sense that it holds that  $\mathbf{m}' = \mathbf{m} + \mathbf{e}$ , i.e. in most coordinates  $\mathbf{m}'$  is identical to  $\mathbf{m}$ , but only in few

<sup>8</sup> i.e. every component of  $e_i$  of  $\mathbf{e}$  is independently 0 with probability  $1 - \rho$  and 1 with probability  $\rho$

coordinates  $\mathbf{m}'$  and  $\mathbf{m}$  differ. Furthermore, one-time security of this encryption scheme follows from the LPN assumption.

The high level strategy to use this symmetric key encryption scheme is now as follows: Assume the matrix  $\mathbf{A} \in \mathbb{Z}_2^{m \times n}$  is known to both the sender and the receiver. In the actual protocol this matrix will be chosen by the receiver, and the communication cost of sending  $\mathbf{A}$  will be amortized by reusing  $\mathbf{A}$  many times.

The OT-receiver chooses a symmetric key  $\mathbf{s} \xleftarrow{\$} \mathbb{Z}_2^n$  uniformly at random and encrypts his vector of choice bits  $\mathbf{b} = (b_1, \dots, b_\ell)$  to  $\mathbf{d} = \mathbf{A}\mathbf{s} + \mathbf{e} + \mathbf{b}$  (where again,  $\mathbf{e} \in \mathbb{Z}_2^\ell$  is  $\rho$ -Bernoulli distributed). Furthermore, the receiver will encrypt the LPN secret under ElGamal, i.e. he encrypts  $\mathbf{s}$  to  $\mathbf{c} = \text{Enc}(\text{pk}, \mathbf{s})$ . For the moment, assume that  $\mathbf{s}$  is encrypted bit-wise with standard ElGamal rather than batched ElGamal. The OT-receiver now sends the ElGamal public key  $\text{pk}$  and the ciphertexts  $\mathbf{c}$  and  $\mathbf{d}$  to the OT-sender.

Now, given these values, the sender can homomorphically decrypt the  $\mathbf{d}$  into ElGamal, effectively *key-switching* from the ciphertext  $\mathbf{d}$  into an ElGamal ciphertext. Concretely: The sender homomorphically evaluates the linear function  $f(\mathbf{x}) = \mathbf{d} - \mathbf{A}\mathbf{x}$  on the ElGamal ciphertext  $\mathbf{c} = \text{Enc}(\text{pk}, \mathbf{s})$ . This produces an ElGamal encryption  $\mathbf{c}'$  encrypting  $f(\mathbf{s}) = \mathbf{d} - \mathbf{A}\mathbf{s} = \mathbf{b} + \mathbf{e} = \mathbf{b}'$ . In other words, the OT-sender has now obtained an ElGamal encryption of a vector  $\mathbf{b}'$  which agrees with  $\mathbf{b}$  in most locations.

The high-level idea is now to let the OT-sender use this ciphertext  $\mathbf{c}'$  as the encryption of the receiver's choice bits and proceed as in the ElGamal-based OT-protocol above. If we were to naively use  $\mathbf{c}'$  in this way, the receiver would obtain the correct output  $m_{i,b_i}$  in locations where  $\mathbf{b}$  and  $\mathbf{b}'$  agree, but would get the wrong output  $m_{i,1-b_i}$  in locations where  $\mathbf{b}$  and  $\mathbf{b}'$  disagree. While there certainly are applications in which a small amount of faulty locations are tolerable, in general this leads to insecure protocols.

There is, however, another issue with this approach. In this paragraph we have implicitly assumed that ElGamal is homomorphic for linear functions modulo 2. However, since the group we implement ElGamal over is of large prime order  $p$ , when we evaluate linear functions such as  $f(\mathbf{x}) = \mathbf{d} - \mathbf{A}\mathbf{x}$  over a ciphertext encrypting a  $\mathbf{s} \in \{0, 1\}^n$ , the result of this evaluation is *not* reduced modulo 2, and the resulting ciphertext in fact encrypts  $f(\mathbf{s})$  as an integer. This does not cause major problems in terms of correctness, as this integer will still be small (at most of size  $m$ ), and hence decryption will still be efficient.

However, this does cause major problems in terms of sender-privacy, as we can only guarantee sender privacy for receiver messages that are guaranteed to encrypt a bit  $b \in \{0, 1\}$ .

For now, we will bypass this problem by relying on a homomorphic encryption scheme which is in fact homomorphic over  $\mathbb{Z}_2$  (rather than  $\mathbb{Z}_p$ ), offers function privacy for linear functions modulo 2 and is compatible with ciphertext compression. Such an encryption can in fact be constructed from the Quadratic Residuosity assumption [19].

Another small issue we haven't addressed here is that the compression mechanisms for the sender and the receiver are somewhat orthogonal, in the sense that

the sender’s message is compressed by compressing a batched ElGamal ciphertext (which generally does not allow homomorphic evaluation across different components), whereas the receiver’s compression strategy requires the homomorphic evaluation of linear functions with multiple (i.e. vector-valued) inputs. In the main body (Section 7) we will show a tradeoff which allows to reconcile these requirements, leading to a batch OT protocol with overall rate 1.

We will first discuss how to deal with the issue of errors in the key-switched ciphertext, and then return to the issue of implementing our approach with ElGamal instead of QR-based encryption.

## 2.4 Dealing with LPN Errors

To deal with the LPN errors in the key-switched ciphertext  $\mathbf{c}'$ , we will pursue the following high-level strategy: The sender will introduce an additional masking on the receiver’s output, which can only be removed in error-free locations. This masking effectively erases the receiver’s output in locations in which the receiver’s output is corrupted.

To communicate the correct outputs in the locations with errors, the parties will rely on an additional protocol which is run in parallel. Given that the number of errors is sufficiently small, the communication cost of this additional protocol will be insubstantial and not affect the overall asymptotic rate.

We will first address the problem of erasing the receiver’s output in corrupted locations. First observe that the receiver knows the locations with errors (i.e. the support of the error vector  $\mathbf{e}$ ). Assume that the LPN error vector  $\mathbf{e}$  has a fixed hamming weight  $t \approx \rho m$ , and note that hardness of fixed-weight LPN follows routinely from the hardness of Bernoulli LPN<sup>9</sup>. A  $t$ -puncturable pseudorandom function [8,6] is a pseudorandom function [25] which supports punctured keys. That is, given a PRF key  $K$  and  $t$  inputs  $x_1, \dots, x_t$ , we can efficiently compute a *punctured key*  $K'$  of size  $t \cdot \text{poly}(\lambda)$  which allows to evaluate the PRF on all inputs *except*  $x_1, \dots, x_t$ . Furthermore, the key  $K'$  does not reveal the function values at  $x_1, \dots, x_t$ , i.e.  $\text{PRF}(K, x_1), \dots, \text{PRF}(K, x_t)$  are pseudorandom given the punctured key  $K'$ .

The approach to erase the receiver’s outputs in erroneous locations is now as follows. The sender chooses a PRF key  $K$  and masks both  $m_{i,0}$  and  $m_{i,1}$  with  $\text{PRF}(K, i)$ , i.e. instead of using  $(m_{i,0}, m_{i,1})$  as OT-inputs, he uses  $m'_{i,0} = m_{i,0} \oplus \text{PRF}(K, i)$  and  $m'_{i,1} = m_{i,1} \oplus \text{PRF}(K, i)$ . Assuming that the sender can somehow communicate a punctured key  $K'$  which is punctured at the locations  $i_1, \dots, i_t$  of the errors (i.e.  $\mathbf{e}_{i_j} = 1$  and  $\mathbf{e}$  is 0 everywhere else), the receiver will be able to remove the mask from error-free locations by computing  $m_{i,b_i} = m'_{i,b_i} \oplus \text{PRF}(K', i)$ . In the erroneous locations however,  $m_{i,1-b_i}$  will be hidden from the view of the receiver as  $\text{PRF}(K, i)$  is pseudorandom even given the punctured key  $K'$ .

How can we communicate the punctured key  $K'$  to the receiver with small communication cost in such a way that the sender does not learn the error-

<sup>9</sup> See e.g. [18,6]



locations  $i_1, \dots, i_t$ ? This could be achieved generically by relying on the punctured PRF construction of [8] and transferring keys using a sublinear private information retrieval (PIR) scheme [17,19]. However, recently [6] provided a protocol to achieve this task very efficiently via a two round protocol communicating only  $t\text{poly}(\lambda)$  bits. In the main body (Section 6), we will refer to this primitive as *co-PIR*, since effectively it allows to communicate a large pseudorandom database to a receiver except in a few locations chosen by the receiver.

Finally, to communicate the correct outputs to the receiver in the locations with errors, we will in fact rely on a two-message PIR scheme with polylogarithmic communication. Such schemes are known e.g. from LWE [11] and were recently constructed from a wide variety of assumptions [19], such as DDH and QR. The idea is as follows: For each error location  $i_j$  the receiver sends an additional OT message  $OT_1(b_{i_j})$  using an off-the-shelf low-rate OT protocol (e.g. the basic ElGamal based protocol sketched above), as well as a PIR message  $PIR_1(i_j)$ . The sender speculatively completes this OT protocol for each index  $i$  (since the index  $i_j$  is not known to the sender), collects his OT responses in a database of size  $\ell$ , runs the PIR sender algorithm on this database, and sends the response back to the receiver. The receiver will now be able to recover the correct  $OT_2$  message via PIR, complete the OT and recover  $m_{i_j, b_{i_j}}$ . We remark that for this protocol to be secure against semi-honest senders, we need a PIR protocol with sender privacy. However, e.g. the protocols provided in [19] readily have this feature.

Carefully putting all these components together, we obtain a batch bit-OT protocol with rate-1, for both the sender and the receiver.

## 2.5 Emulating Small Subgroups

We now return to the issue that ElGamal does not provide function privacy for linear functions modulo 2. Recall that the issue essentially boils down to the fact that the plaintext space of ElGamal is *natively*  $\mathbb{Z}_p$ , and when we encode messages in the least significant bits, i.e. encoding a bit  $b$  as  $g^b$ , then for all practical purposes homomorphic evaluations of linear functions with  $\{0,1\}$  coefficients are over  $\mathbb{Z}_2$ , i.e. the resulting ciphertext encodes the result of the function evaluation *without* reduction modulo 2.

From an algebraic perspective, this problem is rooted in the fact that since  $p$  is prime,  $\mathbb{Z}_p$  has no non-trivial subgroup, i.e. it just does not support modular reductions with respect to anything else than  $p$ .

To approach this problem, we will take inspiration from the domain of lattice cryptography [35]. There, messages are typically encoded in the high order bits of group elements, i.e. to encode  $b$  in  $\mathbb{Z}_p$ , we would like to encode it as  $b \cdot \frac{p}{2}$ . However, since  $p$  is odd, first have to round  $\frac{p}{2}$  to the nearest integer in order to get a proper  $\mathbb{Z}_p$  element, i.e. we encode  $b$  via  $b \cdot \lceil \frac{p}{2} \rceil$ . If we could encode  $b$  with respect to  $\frac{p}{2} \notin \mathbb{Z}_p$ , we would get a subgroup of order 2, i.e. for bits  $b, b' \in \{0,1\}$  it holds that

$$\left(b \cdot \frac{p}{2} + b' \cdot \frac{p}{2}\right) \bmod p = (b + b' \bmod 2) \cdot \frac{p}{2}.$$

However, once we round  $\frac{p}{2}$  to the next integer, we get essentially the same problem as before: If we perform group operations on  $b \lceil \frac{p}{2} \rceil$  and  $b' \lceil \frac{p}{2} \rceil$ , then the rounding errors start to accumulate information about  $b$  and  $b'$  which is cannot be obtained from  $b + b' \bmod 2$ . Specifically

$$\begin{aligned} b \lceil \frac{p}{2} \rceil + b' \lceil \frac{p}{2} \rceil \bmod p &= b \left( \frac{p}{2} + \frac{1}{2} \right) + b' \left( \frac{p}{2} + \frac{1}{2} \right) \bmod p \\ &= (b + b' \bmod 2) \frac{p}{2} + (b + b') \frac{1}{2} \bmod p. \end{aligned}$$

Thus, now the least significant bit of  $b \lceil \frac{p}{2} \rceil + b' \lceil \frac{p}{2} \rceil \bmod p$  e.g. leaks if  $b = b' = 1$ , something which cannot be learned from  $b + b' \bmod 2$ .

Consequently, at first glance the idea of encoding a bit  $b$  in the “high-order” bits of a  $\mathbb{Z}_p$  element seems ineffective. However, the lattice toolkit still has more to offer. In particular, in the context of sampling discrete gaussians from lattices, Peikert [32] considered a technique called *randomized rounding*. The basic idea is, given a real number  $r \in \mathbb{R}$  to not always round to the same value e.g.  $\lceil r \rceil$ , but to sample an integer  $z$  close to  $r$ . In [32], this distribution is a discrete gaussian  $Z$  on  $\mathbb{Z}$  centered at  $r$ , i.e. the expectation of  $Z$  is  $r$ . Such a discrete gaussian is parametrized by a gaussian parameter  $\sigma$ , which essentially controls the standard deviation of the discrete gaussian. We denote  $Z$  by  $\lceil r \rceil_\sigma$ .

Now, given any two  $r, r' \in \mathbb{R}$  and  $\sigma_1, \sigma_2 > \omega(\sqrt{\log(\lambda)})$  (more generally the *smoothing parameter* of  $\mathbb{Z}$ ), Peikert [32] shows that

$$\lceil r \rceil_{\sigma_1} + \lceil r' \rceil_{\sigma_2} \approx_s \lceil r + r' \rceil_{\sqrt{\sigma_1^2 + \sigma_2^2}}.$$

In other words, while  $\lceil r \rceil_{\sigma_1} + \lceil r' \rceil_{\sigma_2}$  and  $\lceil r + r' \rceil_{\sqrt{\sigma_1^2 + \sigma_2^2}}$  are not the same, they are statistically close. This means that anything that can be learned from  $\lceil r \rceil_{\sigma_1} + \lceil r' \rceil_{\sigma_2}$  could have as well been learned from  $\lceil r + r' \rceil_{\sqrt{\sigma_1^2 + \sigma_2^2}}$ . While this comes at the expense of an increase “error” term with parameter  $\sqrt{\sigma_1^2 + \sigma_2^2}$ , this *additive error* is very small (of size approx  $\sigma$ ) controlling the growth of this error term can be handled by standard techniques.

Returning to our goal of emulating small subgroups in  $\mathbb{Z}_p$ , our approach follows almost instantly: Instead of encoding a bit  $b \in \mathbb{Z}_2$  as  $b \cdot \lceil \frac{p}{2} \rceil$ , we will encode it as  $\lceil b \cdot \frac{p}{2} \rceil_\sigma$  (for a  $\sigma > \omega(\sqrt{\log(\lambda)})$ ). For  $b, b' \in \{0, 1\}$  this ensures that

$$\lceil b \cdot \frac{p}{2} \rceil_\sigma + \lceil b' \cdot \frac{p}{2} \rceil_\sigma \bmod p \approx_s \lceil (b + b' \bmod 2) \cdot \frac{p}{2} \rceil_{\sqrt{2}\sigma} \bmod p.$$

Thus, we have ensured that  $\lceil b \cdot \frac{p}{2} \rceil_\sigma + \lceil b' \cdot \frac{p}{2} \rceil_\sigma \bmod p$  does not leak more information than  $b + b' \bmod 2$ .

*Function-Private Evaluation for ElGamal* We will now briefly discuss how this idea leads to a modulo 2 function private homomorphic evaluation procedure for ElGamal. Say we have two ElGamal ciphertexts  $c_1 = (g^{r_1}, h^{r_1} \cdot g^{b_1})$  and  $c_2 = (g^{r_2}, h^{r_2} \cdot g^{b_2})$  for a public key  $\text{pk} = (g, h)$  and we want to homomorphically

evaluate the function  $f(x_1, x_2) = a_1x_1 + a_2x_2 \pmod 2$  (for  $a_1, a_2 \in \{0, 1\}$ ) on this pair of ciphertexts. In the first step, we *randomly encode* the function  $f$  as

$$f'(x_1, x_2) = x_1 \cdot \left\lceil a_1 \frac{p}{2} \right\rceil_\sigma + (1 - x_1) \cdot \lceil 0 \rceil_\sigma + x_2 \cdot \left\lceil a_2 \frac{p}{2} \right\rceil_\sigma + (1 - x_2) \cdot \lceil 0 \rceil_\sigma,$$

noting that this is still a linear function (chosen from a distribution). Homomorphically evaluating  $f'$  on the ciphertexts  $c, c_2$  we obtain a ciphertext  $c'$  encrypting

$$\begin{aligned} f'(b_1, b_1) &= b_1 \cdot \left\lceil a_1 \frac{p}{2} \right\rceil_\sigma + (1 - b_1) \cdot \lceil 0 \rceil_\sigma + b_1 \cdot \left\lceil a_2 \frac{p}{2} \right\rceil_\sigma + (1 - b_1) \cdot \lceil 0 \rceil_\sigma \\ &= \left\lceil b_1 a_1 \frac{p}{2} \right\rceil_\sigma + \left\lceil b_1 a_2 \frac{p}{2} \right\rceil_\sigma \\ &\approx_s \left\lceil (b_1 a_1 + b_1 a_2 \pmod 2) \frac{p}{2} \right\rceil_{\sqrt{2}\sigma}. \end{aligned}$$

In other words, this ciphertext could have been simulated knowing only the function result  $f(b_1, b_1) = b_1 a_1 + b_1 a_2 \pmod 2$ , establishing that this homomorphic evaluation procedure is function private.

One aspect to note is that while the messages  $b_1, b_1$  are encoded in  $c_1, c_2$  in the “low-order-bits” via  $g^{b_1}$  and  $g^{b_2}$ , the function result  $f(b_1, b_2)$  encrypted in  $c'$  is encoded in the high order bits, i.e. it is encoded as  $\approx g^{f(b_1, b_2) \frac{p}{2}}$ . This makes it necessary to change the decryption procedure: Let  $c' = (c'_1, c'_2)$  and  $s$  be the secret key. To decrypt  $c'$  we compute  $f = c'_2 \cdot (c'_1)^{-s} \approx_s g^{\lceil f(s_1, s_2) \cdot \frac{p}{2} \rceil}$ , we test if  $f$  is close to  $g^0 = 1$  or  $g^{\lceil p/2 \rceil}$ . This recovers  $f(s_1, s_2)$ , as the error introduced by the rounding operation is of size at most  $\text{poly}(\lambda)$  via standard gaussian tail bounds.

Finally, we remark this “high-order-bit” encoding is still compatible with ElGamal ciphertext compression, i.e. we can still compress homomorphically evaluated batch ElGamal ciphertexts down asymptotically optimal size, using a slightly different compression mechanism. This mechanism is discussed in Section 5. We expect this technique to have additional applications. As one immediate application, it allows to upgrade the key-dependent message secure encryption scheme of Boneh et al. [3] to support arbitrary linear functions modulo 2.

### 3 Preliminaries

The acronym PPT denotes “probabilistic polynomial time”. Throughout this work,  $\lambda$  denotes the security parameter. By  $\text{negl}(\lambda)$ , we denote a negligible function in  $\lambda$ , that is, a function that vanishes faster than any inverse polynomial in  $\lambda$ . Let  $n \in \mathbb{N}$ . Then,  $[n]$  denotes the set  $\{1, \dots, n\}$ . If  $\mathcal{A}$  is an algorithm, we denote by  $y \leftarrow \mathcal{A}(x)$  the output  $y$  after running  $\mathcal{A}$  on input  $x$ . If  $S$  is a (finite) set, we denote by  $x \leftarrow \$S$  the experiment of sampling uniformly at random an element  $x$  from  $S$ . If  $D$  is a distribution over  $S$ , we denote by  $x \leftarrow \$D$  the element  $x$  sampled from  $S$  according to  $D$ . We denote by  $S[i]$  the  $i$ -th element of  $S$  (where the elements are ordered by ascending order except when explicitly stated otherwise)

For two probability distributions  $X, Y$ , we use the notation  $X \approx_s Y$  to state that the distributions are statistically indistinguishable.

For two vectors  $\mathbf{u}, \mathbf{v} \in \mathbb{F}^n$  over a finite field  $\mathbb{F}$ , we denote by  $\mathbf{u} \odot \mathbf{v}$  their component-wise multiplication. We denote by  $\text{Supp}(\mathbf{u})$  the support of  $\mathbf{u}$ , that is, the set of indices where  $\mathbf{u}$  is different from 0.<sup>10</sup> For  $S \subseteq [n]$ ,  $\mathbf{u}_S$  denotes the vector  $(\mathbf{u}_i)_{i \in S}$ . Finally,  $\mathbf{u}^T$  denotes the transpose of  $\mathbf{u}$  and  $\text{hw}(\mathbf{u})$  denotes the hamming weight of  $\mathbf{u}$  (that is, the number of coordinates of  $\mathbf{u}$  different from 0).

### 3.1 Lattices and Gaussians

We now review some basic notions of lattices and gaussian distributions.

Let  $\mathbf{B} \in \mathbb{R}^{k \times n}$  be a matrix. We denote the lattice generated by  $\mathbf{B}$  by  $\Lambda = \Lambda(\mathbf{B}) = \{\mathbf{x}\mathbf{B} : \mathbf{x} \in \mathbb{Z}^k\}$ .<sup>11</sup> The dual lattice  $\Lambda^*$  of a lattice  $\Lambda$  is defined by  $\Lambda^* = \{\mathbf{x} \in \mathbb{R}^n : \forall \mathbf{y} \in \Lambda, \mathbf{x} \cdot \mathbf{y} \in \mathbb{Z}\}$ . It holds that  $(\Lambda^*)^* = \Lambda$ . The orthogonal lattice  $\Lambda_q^\perp$  is defined by  $\{\mathbf{y} \in \mathbb{Z}_q^n : \mathbf{A}\mathbf{y}^T = 0 \pmod{q}\}$ .

Let  $\rho_s(\mathbf{x})$  be probability distribution of the Gaussian distribution over  $\mathbb{R}^n$  with parameter  $s$  and centered in 0. We define the discrete Gaussian distribution  $D_{S,s}$  over  $S$  and with parameter  $s$  by the probability distribution  $\rho_s(\mathbf{x})/\rho(S)$  for all  $\mathbf{x} \in S$  (where  $\rho_s(S) = \sum_{\mathbf{x} \in S} \rho_s(\mathbf{x})$ ).

For  $\varepsilon > 0$ , the smoothing parameter  $\eta_\varepsilon(\Lambda)$  of a lattice  $\Lambda$  is the least real  $\sigma > 0$  such that  $\rho_{1/\sigma}(\Lambda^* \setminus \{0\}) \leq \varepsilon$  [31].

**Lemma 1 ([2])** For all  $\alpha \in \mathbb{R}$ ,  $\|\mathbf{x}\| \leq \alpha\sqrt{n}$  for  $\mathbf{x} \leftarrow D_{\mathbb{Z}, \alpha}^n$ , except with negligible probability in  $n$ .

We will make use of the following convolution property of discrete gaussians.

**Lemma 2 ([22], Corollary 4.8)** Let  $\Lambda_1, \Lambda_2 \subseteq \mathbb{R}^n$  be lattices, let  $\sigma_1, \sigma_2 > 0$  be such that  $1/\sqrt{1/\sigma_1^2 + 1/\sigma_2^2} > \eta_\varepsilon(\Lambda_1 \cap \Lambda_2)$  for some  $\varepsilon = \text{negl}(\lambda)$ . Then it holds for all  $\mathbf{a}, \mathbf{b} \in \mathbb{R}^n$  that  $D_{\Lambda_1 + \mathbf{a}, \sigma_1} + D_{\Lambda_2 + \mathbf{b}, \sigma_2}$  is statistically close to  $D_{\Lambda_1 + \Lambda_2 + \mathbf{a} + \mathbf{b}, \sqrt{\sigma_1^2 + \sigma_2^2}}$ .

We just need the following simple corollary of Lemma 2, which can be obtained by setting  $\Lambda_1 = \Lambda_2 = \mathbb{Z}$ .

**Corollary 1.** Let  $\sigma_1, \sigma_2, \sigma_3 = \sqrt{\sigma_1^2 + \sigma_2^2}$  be such that  $\sigma_1\sigma_2/\sigma_3 > \eta_\varepsilon(\mathbb{Z})$  for a negligible  $\varepsilon$  and let  $a, b \in \mathbb{Z}$ . Then  $D_{\mathbb{Z} + a, \sigma_1} + D_{\mathbb{Z} + b, \sigma_2}$  and  $D_{\mathbb{Z} + a + b, \sigma_3}$  are statistically close.

### 3.2 Distributed GGM-PPRF Correlation

Let  $\text{PPRF}_{\text{GGM}} = (\text{KeyGen}, \text{Eval}, \text{Puncture}, \text{EvalPunct})$  be the GGM-PPRF scheme based on [26]. The distributed GGM-PPRF correlation functionality [5] considers two parties: A receiver with input  $\alpha \in \{0, 1\}^\ell$  and a sender with input  $\beta \in \mathbb{F}_{p^r}$  and a GGM-PPRF key  $\mathbf{K}$ . The functionality outputs a punctured key  $\mathbf{K}_\alpha$  and a hardwired value  $\beta - \text{PPRF.Eval}(\mathbf{K}, \alpha)$  to the receiver. We now present the formal definition of the functionality.

<sup>10</sup> If there is only one index different from zero,  $\text{Supp}(\mathbf{u})$  denotes this index.

<sup>11</sup> The matrix  $\mathbf{B}$  is called a basis of  $\Lambda(\mathbf{B})$ .

*Distributed GGM-PPRF correlation functionality.* The functionality  $\mathcal{F}_{\text{PPRF-GGM}}$  is parametrized by integers  $\ell, p, r \in \mathbb{N}$ . Moreover, let  $\text{PPRF}_{\text{GGM}} = (\text{KeyGen}, \text{Eval}, \text{Puncture}, \text{EvalPunct})$  be the GGM PPRF scheme with input space  $\{0, 1\}^\ell$  and output space  $\mathbb{F}_{p^r}$ . The functionality works as follows:

- **Receiver phase.** R sends  $\alpha$  to  $\mathcal{F}_{\text{PPRF-GGM}}$  where  $\alpha \in \{0, 1\}^\ell$ .
- **Sender phase.** S sends  $(\beta, K)$  to  $\mathcal{F}_{\text{PPRF-GGM}}$  where  $\beta \in \mathbb{F}_{p^r}$  and  $K \leftarrow \text{PPRF.KeyGen}(1^\lambda)$ .  $\mathcal{F}_{\text{PPRF-GGM}}$  sends  $K_\alpha \leftarrow \text{PPRF.Puncture}(K, \alpha)$  and  $\gamma \leftarrow \beta - \text{PPRF.Eval}(K, \alpha)$  to R.

## 4 Compression-friendly Subgroup Emulation via Gaussian Rounding

We will now provide our new subgroup emulation technique. We first define the gaussian rounding functionality.

**Definition 1** *Let  $\sigma > 0$ . For any  $x \in \mathbb{R}$ , the gaussian rounding  $\lceil x \rceil_\sigma$  is a random variable supported on  $\mathbb{Z}$  defined by*

$$\lceil x \rceil_\sigma = x + D_{\mathbb{Z}-x, \sigma}.$$

In other words,  $\lceil x \rceil_\sigma$  is a discrete gaussian centered on  $x \in \mathbb{R}$  but supported on  $\mathbb{Z}$ .

We will use the following convolution lemma which provides a *simulation property* for gaussian rounding.

**Lemma 3** *Let  $\epsilon > 0$  be bounded by a sufficiently small constant and let  $\sigma_1, \sigma_2 \geq \eta_\epsilon(\mathbb{Z})$ . Then it holds for all  $x, y \in \mathbb{R}$  that*

$$\lceil x \rceil_{\sigma_1} + \lceil y \rceil_{\sigma_2} \approx_s \lceil x + y \rceil_{\sqrt{\sigma_1^2 + \sigma_2^2}}.$$

It immediately follows from Lemma 3 that it holds for every integer  $p \geq 2$  that

$$\lceil x \rceil_{\sigma_1} + \lceil y \rceil_{\sigma_2} \pmod{p} \approx_s \lceil x + y \rceil_{\sqrt{\sigma_1^2 + \sigma_2^2}} \pmod{p}.$$

Please refer to Appendix B of the full version of this paper for the proofs of lemmas in this section.

**Lemma 4** *Let  $p > q \geq 2$  be integers with  $q \leq 2^k$ , and let  $\sigma > \eta_\epsilon(\mathbb{Z})$  for a negligible  $\epsilon$ . Let  $f : \mathbb{Z}_q^n \rightarrow \mathbb{Z}_q$  be given by  $f(x_1, \dots, x_n) = \sum_{i=1}^n a_i x_i + c$  for  $a_1, \dots, a_n, c \in \mathbb{Z}_q$ . Define the randomized function  $\hat{f} : \{0, 1\}^{nk} \rightarrow \mathbb{Z}_p^n$  via*

$$\hat{f}(x_{1,1}, \dots, x_{n,k}) = \sum_{i=1}^n \sum_{j=1}^k \left( x_{i,j} \cdot \left\lceil 2^j \cdot \frac{p}{q} a_i \right\rceil_\sigma + (1 - x_{i,j}) \lceil 0 \rceil_\sigma \right) + \left\lceil \frac{p}{q} c \right\rceil_\sigma.$$

*Then it holds for all  $x_{1,1}, \dots, x_{n,k} \in \{0, 1\}$  that*

$$\hat{f}(x_{1,1}, \dots, x_{n,k}) \approx_s \left\lceil \frac{p}{q} \cdot f \left( \sum_{j=1}^k x_{1,j} 2^j, \dots, \sum_{j=1}^k x_{n,j} 2^j \right) \right\rceil_{\sqrt{2nk+1}\sigma}.$$

## 5 Rate-1 Circuit-Private Linearly Homomorphic Encryption

In this section we define circuit-private LHE and present constructions based on LWE, DDH or QR<sup>12</sup>. All constructions achieve rate 1.

**Definition 2** A (packed) linearly homomorphic encryption (LHE) scheme LHE over a finite group  $\mathbb{G}$  is composed by a tuple of algorithms (KeyGen, Enc, Eval, Shrink, DecShrink) such that:

- KeyGen( $1^\lambda, k$ ) takes as input a security parameter  $\lambda$  and  $k \in \mathbb{N}$ . It outputs a pair of public and secret keys  $(\mathbf{pk}, \mathbf{sk})$ .
- Enc( $\mathbf{pk}, \mathbf{m} = (m_1, \dots, m_k)$ ) takes as input a public key  $\mathbf{pk}$  and a message  $\mathbf{m} = (m_1, \dots, m_k) \in \mathbb{G}^k$ . It outputs a ciphertext  $\mathbf{ct}$ .
- Eval( $\mathbf{pk}, f, (\mathbf{ct}_1, \dots, \mathbf{ct}_\ell)$ ) takes as input a public key  $\mathbf{pk}$ , a linear function  $f : (\mathbb{G}^k)^\ell \rightarrow \mathbb{G}^k$  and  $\ell$  ciphertexts  $(\mathbf{ct}_1, \dots, \mathbf{ct}_\ell)$ . It outputs a new ciphertext  $\tilde{\mathbf{ct}}$ .
- Shrink( $\mathbf{pk}, \mathbf{ct}$ ) takes as input a public key  $\mathbf{pk}$  and a ciphertext  $\mathbf{ct}$ . It outputs a new shrunken ciphertext  $\mathbf{ct}'$ .
- DecShrink( $\mathbf{sk}, \mathbf{ct}$ ) takes as input a secret key  $\mathbf{sk}$  and a shrunken ciphertext  $\mathbf{ct}$ . It outputs a message  $\mathbf{m}$ .

For simplicity, we define the algorithm Eval&Shrink( $\mathbf{pk}, f, (\mathbf{ct}_1 \dots, \mathbf{ct}_\ell)$ ) which outputs a ciphertext  $\tilde{\mathbf{ct}}$  and is defined as

$$\text{Eval\&Shrink}(\mathbf{pk}, f, (\mathbf{ct}_1 \dots, \mathbf{ct}_\ell)) = \text{Shrink}(\mathbf{pk}, \text{Eval}(\mathbf{pk}, f, (\mathbf{ct}_1, \dots, \mathbf{ct}_\ell)))$$

for any linear function  $f$ .

We require the following properties from a (circuit-private) packed LHE: Correctness, semantic security, compactness and circuit-privacy.

**Definition 3 (Correctness)** A packed LHE scheme LHE is said to be correct if for any  $\ell \in \mathbb{N}$ , any messages  $\mathbf{m}_1, \dots, \mathbf{m}_\ell$  and any linear function  $f : (\mathbb{G}^k)^\ell \rightarrow \mathbb{G}^k$  we have that

$$\Pr \left[ \tilde{\mathbf{m}} \leftarrow \text{DecShrink}(\mathbf{sk}, \tilde{\mathbf{ct}}) : \begin{array}{l} (\mathbf{pk}, \mathbf{sk}) \leftarrow \text{KeyGen}(1^\lambda, k) \\ \mathbf{ct}_i \leftarrow \text{Enc}(\mathbf{pk}, \mathbf{m}_i) \text{ for } i \in [\ell] \\ \tilde{\mathbf{ct}} \leftarrow \text{Eval\&Shrink}(\mathbf{pk}, f, (\mathbf{ct}_1 \dots, \mathbf{ct}_\ell)) \end{array} \right] = 1$$

where  $\tilde{\mathbf{m}} \leftarrow f(\mathbf{m}_1, \dots, \mathbf{m}_\ell)$ .

**Definition 4 (Semantic Security)** A packed LHE scheme LHE is said to be semantically secure if for all  $\lambda \in \mathbb{N}$ , all  $k = \text{poly}(\lambda)$  and all adversaries  $\mathcal{A} = (\mathcal{A}_0, \mathcal{A}_1)$  we have that

$$\left| \Pr \left[ b \leftarrow \mathcal{A}_1(\text{st}, \mathbf{ct}) : \begin{array}{l} (\mathbf{pk}, \mathbf{sk}) \leftarrow \text{KeyGen}(1^\lambda, k) \\ (\mathbf{m}_0, \mathbf{m}_1, \text{st}) \leftarrow \mathcal{A}_0(\mathbf{pk}) \\ b \leftarrow_{\mathcal{S}} \{0, 1\} \\ \mathbf{ct} \leftarrow \text{Enc}(\mathbf{pk}, \mathbf{m}_b) \end{array} \right] - \frac{1}{2} \right| \leq \text{negl}(\lambda).$$

<sup>12</sup> Please refer to Appendix D.1 and D.2 of the full version paper for the construction of LWE and QR.

**Definition 5 (Compactness)** We require that a packed LHE scheme LHE has the following compactness properties:

- For  $(\text{pk}, \text{sk}) \leftarrow \text{KeyGen}(1^\lambda, k)$ , the size of the public key  $|\text{pk}|$  is bounded by  $k \cdot \text{poly}(n)$ .
- For any linear function  $f : (\mathbb{G}^k)^\ell \rightarrow \mathbb{G}^k$  and any  $(\mathbf{m}_1, \dots, \mathbf{m}_\ell) \in (\mathbb{G}^k)^\ell$  we have that

$$\liminf_{\lambda \rightarrow \infty} \frac{|f(\mathbf{m}_1, \dots, \mathbf{m}_\ell)|}{|\text{Eval\&Shrink}(\text{pk}, f, (\text{ct}_1 \dots, \text{ct}_\ell))|} \rightarrow 1$$

for sufficiently large  $k$ , where  $(\text{pk}, \text{sk}) \leftarrow \text{KeyGen}(1^\lambda, k)$  and  $\text{ct}_i \leftarrow \text{Enc}(\text{pk}, \mathbf{m}_i)$  for  $i \in [\ell]$ . In this case, we say that the scheme has rate 1.

We also need that the packed LHE scheme fulfills circuit privacy (in the semi-honest case).

**Definition 6 (Circuit Privacy)** A packed LHE scheme LHE is said to be circuit-private if for all messages  $(\mathbf{m}_1, \dots, \mathbf{m}_\ell) \in (\mathbb{G}^k)^\ell$  and all linear functions  $f : (\mathbb{G}^k)^\ell \rightarrow \mathbb{G}^k$ , there exists a simulator  $\text{Sim}$  such that for all adversaries  $\mathcal{A}$  we have that

$$\left| \Pr \left[ 1 \leftarrow \mathcal{A}(\text{pk}, \text{sk}, \tilde{\text{ct}}) : \begin{array}{l} (\text{pk}, \text{sk}) \leftarrow \text{KeyGen}(1^\lambda, k) \\ \text{ct}_i \leftarrow \text{Enc}(\text{pk}, \mathbf{m}_i) \text{ for } i \in [\ell] \\ \tilde{\text{ct}} \leftarrow \text{Eval\&Shrink}(\text{pk}, f, (\text{ct}_1 \dots, \text{ct}_\ell)) \end{array} \right] - \Pr \left[ 1 \leftarrow \mathcal{A}(\text{pk}, \text{sk}, \tilde{\text{ct}}) : \begin{array}{l} (\text{pk}, \text{sk}) \leftarrow \text{KeyGen}(1^\lambda, k) \\ \tilde{\text{ct}} \leftarrow \text{Sim}(\text{pk}, \tilde{\mathbf{m}}) \end{array} \right] \right| \leq \text{negl}(\lambda)$$

where  $\tilde{\mathbf{m}} \leftarrow f(\mathbf{m}_1, \dots, \mathbf{m}_\ell)$ .

In other words, since  $\text{Sim}$  does not use  $f$  to compute  $\tilde{\text{ct}}$ , no information about it is leaked from  $\tilde{\text{ct}}$  (apart from what is trivially leaked by  $f$ ).

*Encryption of matrices.* Above, we defined LHE that supports encryption of vectors  $\mathbf{m} \in \mathbb{G}^k$ . We can easily extend the definition to support encryption of matrices  $\mathbf{M} \in \mathbb{G}^{k \times \alpha}$  for any  $\alpha = \text{poly}(\lambda)$ : Given a public key  $\text{pk}$ , an encryption  $\text{Enc}(\text{pk}, \mathbf{M})$  of  $\mathbf{M}$  is defined as

$$\text{Enc}(\text{pk}, \mathbf{M}) = \left( \begin{array}{c|c|c} \text{Enc}(\text{pk}, \mathbf{m}^{(1)}) & \dots & \text{Enc}(\text{pk}, \mathbf{m}^{(\alpha)}) \\ \hline \end{array} \right)$$

where  $\mathbf{m}^{(i)}$  is the  $i$ -th column of  $\mathbf{M}$ .

## 5.1 Construction from DDH

In the following, let  $\mathcal{G}$  be a (prime-order) *group generator*, that is,  $\mathcal{G}$  is an algorithm that takes as an input a security parameter  $1^\lambda$  and outputs  $(\mathbb{G}, p, g)$ , where  $\mathbb{G}$  is the description of a multiplicative cyclic group,  $p$  is the order of the group which is always a prime number unless differently specified, and  $g$  is a generator of the group. In the following we state the decisional version of the Diffie-Hellman (DDH) assumption.

**Definition 7 (Decisional Diffie-Hellman Assumption)** Let  $(\mathbb{G}, p, g) \leftarrow \mathcal{G}(1^\lambda)$ . We say that the DDH assumption holds (with respect to  $\mathcal{G}$ ) if for any PPT adversary  $\mathcal{A}$

$$|\Pr[1 \leftarrow \mathcal{A}((\mathbb{G}, p, g), (g^a, g^b, g^{ab}))] - \Pr[1 \leftarrow \mathcal{A}((\mathbb{G}, p, g), (g^a, g^b, g^c))]| \leq \text{negl}(\lambda)$$

where  $a, b, c \leftarrow \mathbb{Z}_p$ .

**Shrinking ciphertexts.** We first present how we can shrink DDH-based ciphertexts to achieve rate 1. The shrinking mechanism presented below is a modification of the one presented in [9] (which is itself based on previous works [7,19]).

Let  $(\mathbb{G}, p, g) \leftarrow \mathcal{G}(1^\lambda)$  and  $k \in \mathbb{Z}$ . Consider an El Gamal public key of the form  $\text{pk} = (g, (h_1, \dots, h_k) = (g, (g^{x_1}, \dots, g^{x_k})) \in \mathbb{G}^{k+1}$  for  $x_1, \dots, x_k \leftarrow \mathbb{Z}_p$  (here,  $\mathbf{x} = (x_1, \dots, x_k)$  is the secret key). Consider the following modified El Gamal encryption algorithm where a ciphertext for  $\mathbf{m} = (m_1, \dots, m_k) \in \{0, 1\}^k$  is of the form  $\text{ct} = (c_1, (c_{2,1}, \dots, c_{2,k})) \in \mathbb{G}^{k+1}$  where  $c_1 = g^r$  and  $c_{2,i} = h_i^{r_i} g^{\lceil m_i(p/2) \rceil_\sigma}$ .<sup>13</sup> We now show how to compress ciphertexts of this form.

We will need the following ingredients: Let  $B, T \in \text{poly}(\lambda)$  and  $\text{PRF} = (\text{KeyGen}, \text{Eval})$  be a PRF that maps  $g \in \mathbb{G}$  to  $\{0, 1\}^\tau$  for some  $\tau \in \mathbb{Z}$ . We also define the function  $\text{LEq}_< : \mathbb{G}^2 \rightarrow \{0, 1\}$  which receives two group elements  $g_0, g_1$  and outputs 1 if  $g_0 < g_1$  and 0 otherwise, for some order relation  $<$  (e.g. the lexicographic order).

$\text{Shrink}_{\text{DDH}}(\text{pk}, \text{ct}) :$

- Parse  $\text{pk} = (g, (h_1, \dots, h_k))$  and  $\text{ct} = (c_1, (c_{2,1}, \dots, c_{2,k}))$ . Let  $w = g^{\lfloor p/2 \rfloor}$ .
- Sample a PRF key  $K \leftarrow \text{PRF.KeyGen}(1^\lambda)$  such that the following conditions are simultaneously satisfied:
  1. For every  $i \in [k]$  and  $j \in \{-B, \dots, B\}$  we have that

$$\text{PRF.Eval}(K, c_{2,i} \cdot g^j) \neq 0 \text{ and } \text{PRF.Eval}(K, c_{2,i} \cdot w \cdot g^j) \neq 0.$$

2. For all  $i \in [k]$  there exists  $\ell \in \{B+1, \dots, T\}$  such that

$$\text{PRF.Eval}(K, c_{2,i} \cdot g^\ell) = 0 \text{ and } \text{PRF.Eval}(K, c_{2,i} \cdot w \cdot g^\ell) = 0.$$

- For every  $i \in [k]$ , let  $\delta_{0,i}, \delta_{1,i} > 0$  be the smallest integer such that

$$\text{PRF.Eval}(K, c_{2,i} \cdot g^{\delta_{0,i}}) = 0 \text{ and } \text{PRF.Eval}(K, c_{2,i} \cdot w \cdot g^{\delta_{1,i}}) = 0.$$

Let  $\alpha_{0,i} = c_{2,i} \cdot g^{\delta_{0,i}}$  and  $\alpha_{1,i} = c_{2,i} \cdot w \cdot g^{\delta_{1,i}}$ . If  $\text{LEq}_<(\alpha_{0,i}, \alpha_{1,i}) = 0$ , then set  $b_i = 0$ . Else, set  $b_i = 1$ .

- Output  $\bar{\text{ct}} = (c_1, K, (b_1, \dots, b_k))$ .

$\text{DecShrink}_{\text{DDH}}(\text{sk}, \bar{\text{ct}}) :$

- Parse  $\text{sk} = \mathbf{x} = (x_1, \dots, x_k)$  and  $\bar{\text{ct}} = (c_1, K, (b_1, \dots, b_k))$ . Let  $w = g^{\lfloor p/2 \rfloor}$ .
- For every  $i \in [k]$ , compute  $\beta_{0,i} = c_1^{x_i}$  and  $\beta_{1,i} = c_1^{x_i} \cdot w$ .

<sup>13</sup> Note that  $\lceil \cdot \rceil_\sigma$  is defined in section 4.



- For every  $i \in [k]$ , find the smallest integers  $\gamma_{0,i}, \gamma_{1,i} > 0$  such that

$$\text{PRF.Eval}(\mathbf{K}, \beta_{0,i} \cdot g^{\gamma_{0,i}}) = 0 \text{ and } \text{PRF.Eval}(\mathbf{K}, \beta_{1,i} \cdot g^{\gamma_{1,i}}) = 0.$$

Let  $\bar{\alpha}_{0,i} = \beta_{0,i} \cdot g^{\gamma_{0,i}}$  and  $\bar{\alpha}_{1,i} = \beta_{1,i} \cdot g^{\gamma_{1,i}}$ . If  $\text{LEq}_<(\bar{\alpha}_{0,i}, \bar{\alpha}_{1,i}) = b_i$ , set  $m_i = 0$ . Else, set  $m_i = 1$ .

- Output  $\mathbf{m} = (m_1, \dots, m_k)$ .

**Lemma 5 (Correctness)** *Let  $B = \text{poly}(\lambda)$  be such that  $B > \lambda\sigma + 1$ . Then the shrinking procedure presented above is correct.*

Please refer to Appendix C.1 of the full version paper for the proof.

**Lemma 6 (Runtime)** *Let PRF be a PRF,  $\tau = \log(8Bk)$  and  $T = 2^\tau \lambda \log_e(k) + B(1 + 4k)$ . Then, the shrinking algorithm  $\text{Shrink}_{\text{DDH}}$  described above terminates in polynomial time, except with negligible probability.*

Please refer to Appendix C.2 of the full version paper for the proof.

*Ciphertext rate.* After applying  $\text{Shrink}_{\text{DDH}}$  we obtain a ciphertext composed by  $\tilde{\text{ct}} = (c_1, \mathbf{K}, (b_1, \dots, b_k)) \in \mathbb{G} \times \mathcal{K} \times \{0, 1\}^k$ . Hence,

$$\frac{|\tilde{\text{ct}}|}{|\mathbf{m}|} = \frac{|c_1| + |\mathbf{K}| + |(b_1, \dots, b_k)|}{k} = \frac{2\lambda + k}{k} = 1 + \frac{2\lambda}{k}$$

which tends to 1 for large enough  $k$ .

**Function-private LHE from DDH.** We now present our circuit-private LHE over  $\mathbb{Z}_2$  based on DDH.

$\text{KeyGen}(1^\lambda, k)$  :

- $(\mathbb{G}, p, g) \leftarrow \mathcal{G}(1^\lambda)$
- Sample  $x_1, \dots, x_k \leftarrow \mathbb{Z}_p$ . Compute  $h_i = g^{x_i}$ .
- Output  $\text{pk} = (\mathbb{G}, p, g, h_1, \dots, h_k)$  and  $\text{sk} = \mathbf{x} = (x_1, \dots, x_k)$ .

$\text{Enc}(\text{pk}, \mathbf{m} = (m_1, \dots, m_k))$  :

- Parse  $\text{pk}$  as  $(\mathbb{G}, p, g, h_1, \dots, h_k)$ .
- Sample  $r \leftarrow \mathbb{Z}_p$ . Compute  $c_1 = g^r$  and  $c_{2,i} = h_i^r g^{m_i}$  for  $i \in [k]$ .
- Output  $\text{ct} = (c_1, (c_{2,1}, \dots, c_{2,k}))$ .

$\text{Eval}(\text{pk}, f, (\text{ct}_1, \dots, \text{ct}_\ell))$

- Parse  $\text{pk}$  as  $(\mathbb{G}, p, g, h_1, \dots, h_k)$ ,  $f$  as  $f(\mathbf{x}_1, \dots, \mathbf{x}_\ell) = \sum_{i=1}^\ell a_i \mathbf{x}_i + \mathbf{b}$  for  $\mathbf{a} = (a_1, \dots, a_\ell) \in \mathbb{Z}_2^\ell$  and  $\mathbf{b} \in \mathbb{Z}_2^k$  and  $\text{ct}_i$  as  $(c_{1,i}, \mathbf{c}_{2,i})$  where  $\mathbf{c}_{2,i} = (c_{2,1,i}, \dots, c_{2,k,i})$  for  $i \in [\ell]$ .
- Compute  $\bar{\text{ct}} = (\bar{c}_1, (\bar{c}_{2,1}, \dots, \bar{c}_{2,1}))$  where

$$\bar{c}_1 = \prod_{i=1}^\ell \left( c_{1,i}^{\lceil a_i \frac{\sigma}{2} \rceil} \cdot (g \cdot c_{1,i}^{-1})^{\lfloor 0 \rfloor_\sigma} \right) \cdot g^t$$

and

$$\bar{\mathbf{c}}_2 = \bigodot_{i=1}^{\ell} \left( \mathbf{c}_{2,i}^{\lceil a_i \frac{p}{2} \rceil_\sigma} \odot (g \cdot \mathbf{c}_{2,i}^{-1})^{\lceil 0 \rceil_\sigma} \right) \odot \left( g^{\lceil b_1 \frac{p}{2} \rceil_\sigma}, \dots, g^{\lceil b_k \frac{p}{2} \rceil_\sigma} \right) \odot (h_1^t, \dots, h_k^t).$$

for  $t \leftarrow_{\$} \mathbb{Z}_p$  and where  $\odot$  denotes the component-wise multiplication.

– Output  $\bar{\mathbf{c}}_t$ .

$\text{Shrink}(\text{pk}, \text{ct})$  : Output  $\bar{\mathbf{c}}_t \leftarrow \text{Shrink}_{\text{DDH}}(\text{pk}, \text{ct})$ .

$\text{DecShrink}(\text{sk}, \text{ct})$  : Output  $\mathbf{m} \leftarrow \text{DecShrink}_{\text{DDH}}(\text{sk}, \bar{\mathbf{c}}_t)$ .

Correctness and expected polynomial runtime of the LHE described above is guaranteed by Lemma 5 and Lemma 6 by setting  $B > \lambda(\sigma(\sqrt{2\ell} + 1))$ . Semantic security of the scheme can be established by a simple reduction to the DDH assumption in a similar way as in many previous works (the reduction is similar to the one that proves that El Gamal is semantically secure). It is also easy to see that the scheme has rate-1 for large enough  $k$ .

We now show that the scheme is circuit private. Essentially, circuit privacy can be established by resorting to Lemma 4.

**Lemma 7 (Circuit-privacy)** *The scheme presented above is circuit private.*

Please refer to Appendix C.3 of the full version paper for the proof.

*Larger plaintext space.* As in the LWE case, in the construction presented above, the plaintext space is  $\mathbb{Z}_2^k$ . Both the shrinking algorithm and the function-private LHE schemes can be extended to support plaintext space  $\mathbb{Z}_q^k$  where  $q = \text{poly}(\lambda)$  and  $q = 2^\nu$  for some  $\nu \in \mathbb{Z}$  (the constrain of  $q$  being a power of 2 comes from Lemma 4)

## 6 Co-Private Information Retrieval

In this section, we present a new cryptographic primitive that we call *co-PIR*. In a co-PIR scheme, a receiver (with input a set of indices  $S$ ) and a sender (with no input) interact such that, at the end, the sender obtains a string  $\mathbf{y} \in \mathbb{Z}_q^m$  and receiver obtains  $\mathbf{y}_{-S}$  (all positions of  $y$  except for the indices in  $S$ ).

In terms of security, we require that the sender learns nothing about  $S$ , whereas the string  $\mathbf{y}_S$  looks pseudorandom to the receiver. In terms of efficiency, we require that the total communication of the protocol scales only with  $|S| \text{poly}(\lambda) \text{polylog}(m)$  (that is, it scales only poly-logarithmically with  $m$ ). We present a construction for Co-PIR from the distributed GGM-PPRF correlation (as shown in [5]) in Appendix E.1 of the full version paper; We also present another construction with black-box usage of PPRF and PIR in Appendix E.2 of the full version paper.

## 6.1 Definition

We start by defining Co-PIR and present its security properties.

**Definition 8 (Co-PIR)** A (two-round) Co-PIR scheme  $\text{CoPIR}$  over  $\mathbb{Z}_q$  is parametrized by an integer  $m$  where  $m = \text{poly}[\lambda]$ , and is composed by a tuple of algorithms (Query, Send, Retrieve) such that

- $\text{Query}(1^\lambda, S)$  takes as input a set of indices  $S \subseteq [m]$ . It outputs a message  $\text{copir}_1$  and a private state  $\text{st}$ .
- $\text{Send}(\text{copir}_1)$  takes as input a first message  $\text{copir}_1$ . It outputs a second message  $\text{copir}_2$  and a string  $\mathbf{y} \in \mathbb{Z}_q^m$ .
- $\text{Dec}(\text{copir}_2, \text{st})$  takes as input a second message  $\text{copir}_2$  and a state  $\text{st}$ . It outputs a string  $\tilde{\mathbf{y}} \in \mathbb{Z}_q^m$ .

**Definition 9 (Correctness)** A Co-PIR scheme  $\text{CoPIR}$  is said to be correct if for any  $m = \text{poly}(\lambda)$  and  $S \subseteq [m]$  we have that

$$\Pr \left[ \begin{array}{l} (\text{copir}_1, \text{st}) \leftarrow \text{Query}(1^\lambda, S) \\ \mathbf{y}_{[m] \setminus S} = \tilde{\mathbf{y}}_{[m] \setminus S} : (\text{copir}_2, \mathbf{y}) \leftarrow \text{Send}(\text{copir}_1) \\ \tilde{\mathbf{y}} \leftarrow \text{Retrieve}(\text{copir}_2, \text{st}) \end{array} \right] = 1.$$

In other words, the strings  $\mathbf{y}$  and  $\tilde{\mathbf{y}}$  match for every coordinate  $i \in [m] \setminus S$ .

In terms of security, we require two properties: receiver security and sender security.

**Definition 10 (Receiver security)** A Co-PIR scheme  $\text{CoPIR}$  is said to be receiver secure if for all  $m = \text{poly}(\lambda)$ , any subsets  $S_1, S_2 \subseteq [m]$  we have that for any adversary  $\mathcal{A}$

$$\left| \Pr [1 \leftarrow \mathcal{A}(k, \text{copir}_1) : (\text{copir}_1, \text{st}) \leftarrow \text{Query}(1^\lambda, S_1)] - \Pr [1 \leftarrow \mathcal{A}(k, \text{copir}_1) : (\text{copir}_1, \text{st}) \leftarrow \text{Query}(1^\lambda, S_2)] \right| \leq \text{negl}(\lambda).$$

**Definition 11 (Sender security)** A Co-PIR scheme  $\text{CoPIR}$  is said to be sender secure if for any  $m = \text{poly}(\lambda)$ , any subset  $S \subseteq [m]$  we have that for all adversaries  $\mathcal{A}$

$$\left| \Pr \left[ 1 \leftarrow \mathcal{A}(k, \text{st}, \text{copir}_2, \mathbf{y}_S) : \begin{array}{l} (\text{copir}_1, \text{st}) \leftarrow \text{Query}(1^\lambda, S) \\ (\text{copir}_2, \mathbf{y}) \leftarrow \text{Send}(\text{copir}_1, \mathbf{x}) \end{array} \right] - \Pr \left[ 1 \leftarrow \mathcal{A}(k, \text{st}, \text{copir}_2, \mathbf{y}'_S) : \begin{array}{l} (\text{copir}_1, \text{st}) \leftarrow \text{Query}(1^\lambda, S) \\ (\text{copir}_2, \mathbf{y}) \leftarrow \text{Send}(\text{copir}_1, \mathbf{x}) \\ \mathbf{y}'_S \leftarrow \mathbb{Z}_q^{|S|} \end{array} \right] \right| \leq \text{negl}(\lambda).$$

**Definition 12 (Compactness)** A Co-PIR scheme  $\text{CoPIR}$  is said to be compact if  $|\text{copir}_1|, |\text{copir}_2| = |S| \cdot \text{polylog}(m) \cdot \text{poly}(\lambda)$  for any  $S \subseteq [m]$  where  $(\text{copir}_1, \text{st}) \leftarrow \text{Query}(1^\lambda, S)$  and  $(\text{copir}_2, \mathbf{y}) \leftarrow \text{Send}(\text{copir}_1)$ . In other words, the communication complexity depends only poly-logarithmically in  $m$ .

## 7 Oblivious Transfer with Overall Rate 1

We will now provide our construction of an oblivious transfer protocol with overall rate 1.

*Ingredients.* We will make use of the following ingredients.

- A packed linearly homomorphic encryption scheme  $\text{LHE} = (\text{KeyGen}, \text{Enc}, \text{Eval}, \text{Shrink}, \text{DecShrink})$  with plaintext space  $\{0, 1\}^\ell$  and a post homomorphism shrinking procedure  $\text{Shrink}$  which converts ciphertexts into a rate 1 representation.<sup>14</sup>
- The binary LPN( $n, m, \rho$ ) problem with dimension  $n = \text{poly}(n)$ ,  $m = n \cdot \ell \cdot \text{poly}(n)$  samples and slightly sub-constant noise-rate  $\rho = m^{1-\epsilon}$ .
- A 2-round PIR scheme  $\text{PIR} = (\text{Query}, \text{Send}, \text{Retrieve})$  with poly-logarithmic communication complexity and sender privacy.
- A 2-round Co-PIR scheme  $\text{CoPIR} = (\text{Query}, \text{Send}, \text{Retrieve})$  over  $\mathbb{Z}_2$  parametrized by  $m$ .

*Additional Notation.* Furthermore, to declutter notation we define the following embedding functions.

$\text{RowMatrix}(\ell, n, \mathbf{v}_1, \dots, \mathbf{v}_\ell)$ : Takes row-vectors  $\mathbf{v}_1, \dots, \mathbf{v}_\ell \in \{0, 1\}^n$  and outputs a matrix

$$\mathbf{V} = \begin{pmatrix} \text{---} \mathbf{v}_1 \text{---} \\ \vdots \\ \text{---} \mathbf{v}_\ell \text{---} \end{pmatrix},$$

i.e. for every  $i \in [\ell]$  the  $i$ -th row of  $\mathbf{V}$  is the row-vector  $\mathbf{v}_i$ .

$\text{SingleRowMatrix}(\ell, n, i, \mathbf{v})$ : Takes a row-vector  $\mathbf{v} \in \{0, 1\}^n$  and outputs a matrix

$$\mathbf{V} = \begin{pmatrix} 0 \dots 0 \\ \vdots \quad \quad \vdots \\ 0 \dots 0 \\ \text{---} \mathbf{v} \text{---} \\ 0 \dots 0 \\ \vdots \quad \quad \vdots \\ 0 \dots 0 \end{pmatrix},$$

i.e. the  $i$ -th row of  $\mathbf{V}$  is  $\mathbf{v}$ , but  $\mathbf{V}$  is 0 everywhere else.

$\text{Diag}(n, \mathbf{v})$ : Takes a vector  $\mathbf{v} = (v_1, \dots, v_n) \in \{0, 1\}^n$  and outputs a matrix

$$\mathbf{D} = \begin{pmatrix} v_1 & 0 \\ & \ddots \\ 0 & v_n \end{pmatrix},$$

<sup>14</sup> Recall that we use the notation  $\text{Eval}\&\text{Shrink}$  to denote the composition of algorithms  $\text{Eval}$  and  $\text{Shrink}$ .

i.e.  $\mathbf{D} \in \{0, 1\}^{n \times n}$  is a diagonal matrix with the components of  $\mathbf{v}$  on its diagonal.

We observe the following:

- For any  $\mathbf{v}_1, \dots, \mathbf{v}_\ell \in \{0, 1\}^n$  it holds that

$$\text{RowMatrix}(\ell, n, \mathbf{v}_1, \dots, \mathbf{v}_\ell) = \sum_{i=1}^{\ell} \text{SingleRowMatrix}(\ell, n, i, \mathbf{v}_i).$$

- For  $\mathbf{x}, \mathbf{y} \in \{0, 1\}^n$  it holds that

$$\mathbf{x} \cdot \text{Diag}(n, \mathbf{y}) = \mathbf{x} \odot \mathbf{y},$$

where  $\odot$  denotes component-wise multiplication.

### 7.1 The Protocol

The protocol  $\text{OT} = (\text{OTR}, \text{OTS}, \text{OTD})$  is given as follows.

**OTR**( $\mathbf{b} \in \{0, 1\}^{m\ell}$ ) :

- Parse  $\mathbf{b} = (\mathbf{b}_1, \dots, \mathbf{b}_\ell)$ , where the  $\mathbf{b}_i \in \{0, 1\}^m$  are blocks of size  $m$ .
- Choose  $\mathbf{A} \leftarrow \mathcal{S} \{0, 1\}^{n \times m}$  uniformly at random and compute a pair of public and secret key  $(\text{pk}, \text{sk}) \leftarrow \text{LHE.KeyGen}(1^\lambda, \ell)$ .
- For all  $i \in [\ell]$ , choose  $\mathbf{s}_i \leftarrow \mathcal{S} \{0, 1\}^n$ , and  $\mathbf{e}_i \leftarrow \mathcal{S} \chi_{m,t}$ , compute  $\mathbf{c}_i \leftarrow \mathbf{s}_i \mathbf{A} + \mathbf{e}_i + \mathbf{b}_i$ , and set  $\mathbf{S}_i \leftarrow \text{SingleRowMatrix}(\ell, n, i, \mathbf{s}_i)$ . Compute a matrix-ciphertext  $\text{ct}_i \leftarrow \text{LHE.Enc}(\text{pk}, \mathbf{S}_i)$ .
- For all  $i \in [\ell]$  set  $J_i = \text{Supp}(\mathbf{e}_i)$  to be the support of  $\mathbf{e}_i$ . Compute  $(\text{copir}_{1,i}, \text{st}_i) \leftarrow \text{CoPIR.Query}(J_i)$ . Additionally, for  $j \in [t]$  compute  $(\mathbf{q}_{i,j}, \hat{\text{st}}_{i,j}) = \text{PIR.Query}(J_i[j])$ .
- Output  $\text{ot}_1 = (\text{pk}, \mathbf{A}, \{\text{ct}_i, \mathbf{c}_i, \text{copir}_{1,i}\}_{i \in [\ell]}, \{\mathbf{q}_{i,j}\}_{i \in [\ell], j \in [t]})$  and  $\text{st} = (\text{sk}, \{\text{st}_i, J_i\}_{i \in [\ell]}, \{\hat{\text{st}}_{i,j}\}_{i \in [\ell], j \in [t]})$ .

**OTS**( $(\mathbf{m}_0, \mathbf{m}_1) \in (\{0, 1\}^{m\ell})^2, \text{ot}_1$ ) :

- Parse  $\mathbf{m}_0 = (\mathbf{m}_{0,1}, \dots, \mathbf{m}_{0,\ell})$  and  $\mathbf{m}_1 = (\mathbf{m}_{1,1}, \dots, \mathbf{m}_{1,\ell})$ , where each  $\mathbf{m}_{b,i} = (m_{b,i,1}, \dots, m_{b,i,m}) \in \{0, 1\}^m$ . Parse  $\text{ot}_1 = (\text{pk}, \mathbf{A}, \{\text{ct}_i, \mathbf{c}_i, \text{copir}_{1,i}\}_{i \in [\ell]}, \{\mathbf{q}_{i,j}\}_{i \in [\ell], j \in [t]})$ .
- For  $i \in [\ell]$   $(\mathbf{y}_i, \text{copir}_{2,i}) \leftarrow \text{CoPIR.Send}(\text{copir}_{1,i})$  where  $\mathbf{y}_i = (y_{i,1}, \dots, y_{i,m})$ . Set  $\mathbf{z}_i = \mathbf{m}_{0,i} + \mathbf{y}_i$ .
- Set  $\mathbf{Z} = \text{RowMatrix}(\ell, m, \mathbf{z}_1, \dots, \mathbf{z}_\ell)$ .
- For all  $i \in [\ell]$  set  $\mathbf{C}_i = \text{SingleRowMatrix}(\ell, m, i, \mathbf{c}_i)$  and  $\mathbf{D}_i = \text{Diag}(m, \mathbf{m}_{1,i} - \mathbf{m}_{0,i})$ .
- Define the  $\mathbb{Z}_2$ -linear function  $f : (\{0, 1\}^{\ell \times n})^\ell \rightarrow \{0, 1\}^{\ell \times m}$  via

$$f(\mathbf{X}_1, \dots, \mathbf{X}_\ell) = \left( \sum_{i=1}^{\ell} (-\mathbf{X}_i \mathbf{A} + \mathbf{C}_i) \cdot \mathbf{D}_i \right) + \mathbf{Z}.$$

- Compute  $\tilde{\text{ct}} \leftarrow \text{LHE.Eval\&Shrink}(\text{pk}, f, \text{ct}_1, \dots, \text{ct}_\ell)$ .

- For  $i \in [\ell]$  set  $DB_i = (y_{i,1} + (m_{1,i,1} - m_{0,i,1}), \dots, y_{i,m} + (m_{1,i,m} - m_{0,i,m}))$ .
  - For all  $j \in [t]$  compute  $r_{i,j} \leftarrow \text{PIR.Send}(DB_i, \mathbf{q}_{i,j})$ .
  - Output  $\text{ot}_2 = (\tilde{\text{ct}}, \{\text{copir}_{2,i}\}_{i \in [\ell]}, \{r_{i,j}\}_{i \in [\ell], j \in [t]})$ .
- OTD( $\text{ot}_2, \text{st}$ ):
- Parse  $\text{ot}_2 = (\tilde{\text{ct}}, \{\text{copir}_{2,i}\}_{i \in [\ell]}, \{r_{i,j}\}_{i \in [\ell], j \in [t]})$  and  $\text{st} = (\text{sk}, \{\text{st}_i, J_i\}_{i \in [\ell]}, \{\hat{\text{st}}_{i,j}\}_{i \in [\ell], j \in [t]})$ .
  - For all  $i \in [\ell]$  compute  $\tilde{\mathbf{y}}_i = (\tilde{y}_{i,1}, \dots, \tilde{y}_{i,m}) \leftarrow \text{CoPIR.Retrieve}(\text{copir}_{2,i}, \text{st}_i)$ .
  - For  $i \in [\ell]$  and  $j \in [t]$  compute  $\tilde{z}_{i,j} \leftarrow \text{PIR.Retrieve}(r_{i,j}, \hat{\text{st}}_{i,j})$ .
  - For  $i \in [\ell]$  set  $\mathbf{z}_i = (z_{i,1}, \dots, z_{i,m})$  where
 
$$z_{i,l} = \begin{cases} \tilde{z}_{i,j} & \text{if } l = J_i[j] \\ \tilde{y}_{i,l} & \text{otherwise} \end{cases}.$$
  - Set  $\mathbf{Z} = \text{RowMatrix}(\ell, m, \mathbf{z}_1, \dots, \mathbf{z}_\ell)$ .
  - Compute  $\tilde{\mathbf{W}} \leftarrow \text{LHE.DecShrink}(\text{sk}, \tilde{\text{ct}})$  and  $\mathbf{W} = \tilde{\mathbf{W}} - \mathbf{Z}$ .
  - Let  $\mathbf{w}_1, \dots, \mathbf{w}_\ell$  be the rows of  $\mathbf{W}$ . Output  $\mathbf{w} = (\mathbf{w}_1 \| \dots \| \mathbf{w}_\ell) \in \{0, 1\}^{m\ell}$ .

*Correctness.* We will first show that OT is correct, given that LHE, CoPIR and PIR are correct.

**Theorem 1.** *Assume that LHE, CoPIR and PIR are correct. Then the scheme presented above is correct.*

*Proof.* First note that by linear-homomorphic correctness of LHE it holds that

$$\begin{aligned} \tilde{\mathbf{W}} &= \text{LHE.DecShrink}(\text{sk}, \text{LHE.Eval\&Shrink}(\text{pk}, f, \text{LHE.Enc}(\text{pk}, \mathbf{S}_1), \dots, \text{LHE.Enc}(\text{pk}, \mathbf{S}_\ell))) \\ &= f(\mathbf{S}_1, \dots, \mathbf{S}_\ell) \\ &= \left( \sum_{i=1}^k (-\mathbf{S}_i \mathbf{A} + \mathbf{C}_i) \cdot \mathbf{D}_i \right) + \mathbf{Z} \end{aligned}$$

Let  $\tilde{\mathbf{w}}_i$  be the  $i$ -th row of  $\tilde{\mathbf{W}}$ . It holds by definition  $\mathbf{S}_i$ ,  $\mathbf{C}_i$  and  $\mathbf{Z}_i$  that

$$\begin{aligned} \tilde{\mathbf{w}}_i &= (-\mathbf{s}_i \mathbf{A} + \mathbf{c}_i) \mathbf{D}_i + \mathbf{z}_i \\ &= (-\mathbf{s}_i \mathbf{A} + \mathbf{s}_i \mathbf{A}_i + \mathbf{e}_i + \mathbf{b}_i) \mathbf{D}_i + \mathbf{m}_{0,i} + \mathbf{y}_i \\ &= \mathbf{b}_i \odot (\mathbf{m}_{1,i} - \mathbf{m}_{0,i}) + \mathbf{m}_{0,i} + \mathbf{e}_i \odot (\mathbf{m}_{1,i} - \mathbf{m}_{0,i}) + \mathbf{y}_i. \end{aligned}$$

where  $\mathbf{y}_i = (y_{i,1}, \dots, y_{i,m})$  is part of the output of  $\text{CoPIR.Send}$ .

Let  $J_i$  be the support of  $\mathbf{e}_i$  and let  $\tilde{\mathbf{y}}_i = (\tilde{y}_{i,1}, \dots, \tilde{y}_{i,m}) \leftarrow \text{CoPIR.Retrieve}(\text{copir}_{2,i}, \text{st}_i)$ . By the correctness of the Co-PIR scheme  $\text{CoPIR}$  we have that  $\tilde{y}_{i,j} = y_{i,j}$  for all  $j \notin J_i$ . On the other hand, by the correctness of the PIR scheme  $\text{PIR}$  it holds that

$$\tilde{z}_{i,j} = y_{i,j} + (m_{1,i,j} - m_{0,i,j})$$

for all  $j \in J_i$ . Consequently, we have that

$$z_{i,j} = \begin{cases} y_{i,j} + (m_{1,i,j} - m_{0,i,j}) & \text{if } l = J_i[j] \\ y_{i,j} & \text{otherwise} \end{cases}.$$

In other words, the term  $(m_{1,i,j} - m_{0,i,j})$  only appears in the coordinates where  $\mathbf{e}_i$  is equal to one. Then, it holds that

$$\mathbf{z}_i = \mathbf{e}_i \odot (\mathbf{m}_{1,i} - \mathbf{m}_{0,i}) + \mathbf{y}_i.$$

We conclude that

$$\mathbf{w} = \tilde{\mathbf{w}}_i - \mathbf{z}_i = \mathbf{b}_i \odot (\mathbf{m}_{1,i} - \mathbf{m}_{0,i}) + \mathbf{m}_{0,i}.$$

Since  $\mathbf{w} = (\mathbf{w}_1 \parallel \dots \parallel \mathbf{w}_\ell)$  it follows that

$$\mathbf{w} = \mathbf{b} \odot (\mathbf{m}_1 - \mathbf{m}_0) + \mathbf{m}_0,$$

i.e. OT is correct.

*Communication complexity.* We will now analyze the communication complexity of OT and show which choice of parameters leads to an overall rate approaching 1.

The bit-size of the message  $\text{ot}_1 = (\text{pk}, \mathbf{A}, \{\text{ct}_i, \mathbf{c}_i, \text{copir}_{1,i}\}_{i \in [\ell]}, \{\mathbf{q}_{i,j}\}_{i \in [\ell], j \in [t]})$  can be bounded as follows.

- $|\text{pk}| = \ell \cdot \text{poly}(\lambda)$
- $|\mathbf{A}| = n \cdot m$
- $|\{\text{ct}_i\}_{i \in [\ell]}| = \ell^2 \cdot n \cdot \text{poly}(\lambda)$
- $|\{\mathbf{c}_i\}_{i \in [\ell]}| = \ell \cdot m$
- $|\{\text{copir}_{1,i}\}_{i \in [\ell]}| = \ell \cdot t \cdot \text{polylog}(m) \cdot \text{poly}(\lambda)$
- $|\{\mathbf{q}_{i,j}\}_{i \in [\ell], j \in [t]}| = \ell \cdot t \cdot \text{polylog}(m) \cdot \text{poly}(\lambda).$

Consequently, the overall upload-rate  $\rho_{\text{up}}$  can be bounded by

$$\begin{aligned} \rho_{\text{up}} &= \frac{|\text{pk}| + |\mathbf{A}| + |(\text{ct}_i)_{i \in [\ell]}| + |(\mathbf{c}_i)_{i \in [\ell]}| + |\{\text{copir}_{1,i}\}_{i \in [\ell]}| + |(\mathbf{q}_{i,j})_{i \in [\ell], j \in [t]}|}{\ell m} \\ &\leq 1 + \frac{\text{poly}(\lambda)}{m} + \frac{n}{\ell} + \frac{\ell \cdot n \cdot \text{poly}(\lambda)}{m} + \frac{t \cdot \text{polylog}(m) \cdot \text{poly}(\lambda)}{m} \\ &\leq 1 + \frac{n}{\ell} + \frac{\ell \cdot n \cdot \text{poly}(\lambda)}{m} + \frac{t \cdot \text{polylog}(m) \cdot \text{poly}(\lambda)}{m}. \end{aligned}$$

We get an overall upload rate of  $\rho_{\text{up}} = 1 + O(1/\lambda)$  by choosing  $\ell = \lambda \cdot n$  and  $m = n^2 \cdot \text{poly}(\lambda)$  for a sufficiently large  $\text{poly}(\lambda)$  depending on  $\epsilon$  (where  $t = m^{1-\epsilon}$ ).

The bit-size of the message  $\text{ot}_2 = (\tilde{\text{ct}}, \{\text{copir}_{2,i}\}_{i \in [\ell]}, \{\mathbf{r}_{i,j}\}_{i \in [\ell], j \in [t]})$  can be bounded as follows.

- $|\tilde{\text{ct}}| = \ell m (1 + \rho_{\text{LHE}})$ , where  $1 + \rho_{\text{LHE}}$  is the ciphertext rate of LHE.
- $|\{\text{copir}_{2,i}\}_{i \in [\ell]}| = \ell \cdot t \cdot \text{polylog}(m) \cdot \text{poly}(\lambda)$
- $|\{\mathbf{r}_{i,j}\}_{i \in [\ell], j \in [t]}| = \ell \cdot t \cdot \text{polylog}(m) \cdot \text{poly}(\lambda)$

Thus, the download-rate  $\rho_{\text{down}}$  can be bounded by

$$\begin{aligned} \rho_{\text{down}} &= \frac{|\tilde{\text{ct}}| + |\{\text{copir}_{2,i}\}_{i \in [\ell]}| + |\{\mathbf{r}_{i,j}\}_{i \in [\ell], j \in [t]}|}{\ell m} \\ &\leq 1 + \rho_{\text{LHE}} + \frac{\ell \cdot t \cdot \text{polylog}(m) \cdot \text{poly}(\lambda)}{m}. \end{aligned}$$

By the above choice of  $m$  this comes down to  $\rho_{\text{down}} \leq 1 + \rho_{\text{LHE}} + O(1/\lambda)$ .

## 7.2 Security

*Receiver Security* We now focus on the security of the scheme. We start by proving that the scheme is secure against semi-honest senders.

**Theorem 2.** *Assume that LHE is a semantic secure LHE scheme, PIR is a user-private PIR scheme, CoPIR is a receiver secure Co-PIR scheme and that the LPN( $n, m, \rho$ ) assumption holds for  $\rho = m^{1-\varepsilon}$  for  $\varepsilon > 0$ . Then the scheme presented in Section 7.1 is receiver secure against semi-honest adversaries.*

Recall that the receiver’s message is composed by LHE ciphertexts, LPN samples, Co-PIR and PIR first messages. In a nutshell, receiver security follows from the fact that the ciphertexts hide the LPN secret, the LPN samples hide the receiver’s input  $\mathbf{b}$  and finally the Co-PIR and PIR first messages hide the indices  $J_i$ . We prove the above theorem in Appendix G.1 of the full version paper.

*Sender Security*

**Theorem 3.** *Assume that LHE is a statistically function-private LHE scheme, PIR is a sender-private PIR scheme and CoPIR is a sender-private Co-PIR scheme. Then the scheme presented in Section 7.1 is sender secure.*

In a nutshell, we can use the sender security of PIR and Co-PIR to remove any information about the indices of  $DB_i$  that are not in  $\text{Supp}(\mathbf{e}_i)$ , and finally invoke circuit-privacy of the LHE. We prove sender security in Appendix G.2 of the full version paper.

*Hardness assumptions for optimal-rate OT.* When we instantiate the LHE with one of the schemes from Section 5, the Co-PIR with the construction from Section 6 and the PIR with a known black-box construction based on LWE, DDH or QR [19], we obtain the following corollary

**Corollary 1** *Assuming the LWE, DDH or QR assumptions together with the LPN( $n, m, \rho$ ), there is a black-box construction for optimal-rate OT.*

## 8 Oblivious Matrix-Vector Product and Oblivious Linear Evaluation with Overall Rate 1

In this section we show how we can extend the techniques from the previous section to build protocols for OMV and OLE that achieve optimal rate.

We start by presenting a secure protocol for oblivious matrix-vector product (OMV). In an OMV functionality there is a sender, with input a matrix  $\mathbf{M} \in \mathbb{Z}_q^{m \times m}$  and a vector  $\mathbf{v} \in \mathbb{Z}_q^m$ , and a receiver with input  $\mathbf{b} \in \mathbb{Z}_q^m$ . In the end, the receiver gets the value  $\mathbf{bM} + \mathbf{v}$  but learns nothing about  $\mathbf{M}$  and  $\mathbf{v}$  whereas the sender learns nothing about  $\mathbf{b}$ . We start by defining the functionality:



*OMV functionality.* The functionality  $\mathcal{F}_{\text{OMV}}$  is parametrized by integers  $m = \text{poly}(\lambda)$  and  $q$  and works as follows:

- **Receiver phase.** R sends  $\mathbf{b}$  to  $\mathcal{F}_{\text{OMV}}$  where  $\mathbf{b} \in \mathbb{Z}_q^m$ .
- **Sender phase.** S sends  $(\mathbf{M}, \mathbf{v})$  to  $\mathcal{F}_{\text{OMV}}$  where  $\mathbf{M} \in \mathbb{Z}_q^{m \times m}$  and  $\mathbf{v} \in \{0, 1\}^m$ .  $\mathcal{F}_{\text{OMV}}$  sends  $\mathbf{bM} + \mathbf{v} \in \mathbb{Z}_q^m$  to R.

We describe the concrete OMV protocol in Appendix H of the full version paper.

### 8.1 OLE Protocol

An oblivious linear evaluation (OLE) is a protocol between a sender, with input an affine function  $f$ , and a receiver, with input a point  $b$ . It allows for the receiver to obliviously learn  $f(b)$ . We now show how we can obtain an OLE using the OMV protocol presented in Appendix H of the full version paper.

We start by defining the functionality:

*OLE functionality.* The functionality  $\mathcal{F}_{\text{OLE}}$  is parametrized by integers  $k = \text{poly}(\lambda)$  and  $q$  and works as follows:

- **Receiver phase.** R sends  $\mathbf{b}$  to  $\mathcal{F}_{\text{OLE}}$  where  $\mathbf{b} \in \mathbb{Z}_q^k$ .
- **Sender phase.** S sends  $(\mathbf{u}_0, \mathbf{u}_1)$  to  $\mathcal{F}_{\text{OLE}}$  where  $\mathbf{u}_0, \mathbf{u}_1 \in \mathbb{Z}_q^k$ .  $\mathcal{F}_{\text{OLE}}$  sends  $\mathbf{b} \odot \mathbf{u}_0 + \mathbf{u}_1 \in \mathbb{Z}_q^k$  to R.

**Protocol for Small Fields** We briefly sketch how we can construct an OLE scheme over  $\mathbb{Z}_q$  where  $q = \text{poly}(\lambda)$ . The protocol follows as a particular case of the protocol of Appendix H. We give a brief overview of the scheme below.

Using the notation of Appendix H, let  $\mathbf{b} = (\mathbf{b}_1, \dots, \mathbf{b}_\ell) \in \mathbb{Z}_q^{m\ell}$  be the receiver's input and let  $(\mathbf{u}_0 = (\mathbf{u}_{0,1}, \dots, \mathbf{u}_{0,\ell}), \mathbf{u}_1 = (\mathbf{u}_{1,1}, \dots, \mathbf{u}_{1,\ell})) \in (\mathbb{Z}_q^{m\ell})^2$  be the sender's input. To achieve OLE, the sender constructs the matrices  $\mathbf{D}_i = \text{Diag}(m, \mathbf{u}_{0,i})$  and sets  $\mathbf{v}_i = \mathbf{u}_{1,i}$  for all  $i \in [\ell]$ . Then they run the OMV protocol where the receiver inputs  $\mathbf{b}$  and the sender inputs  $\mathbf{D} = (\mathbf{D}_1, \dots, \mathbf{D}_\ell)$  and  $\mathbf{v} = (\mathbf{v}_1, \dots, \mathbf{v}_\ell)$ . It is easy to see that the output of the receiver is  $\mathbf{y} = (\mathbf{y}_1, \dots, \mathbf{y}_\ell)$  where

$$\mathbf{y}_i = \mathbf{b}_i \mathbf{D}_i + \mathbf{v}_i = \mathbf{b}_i \odot \mathbf{u}_{0,i} + \mathbf{u}_{1,i}$$

be the correctness of the OMV protocol.

Moreover,  $\text{hw}(\mathbf{D}_i) = 1 \leq m^{1-\zeta}$  for some  $\zeta > 0$  such that  $\zeta + \epsilon > 1$ . Thus the resulting protocol achieves overall rate 1. Finally, in terms of hardness assumptions, the OLE protocol inherits the same security.

**Extending OLE to Larger Rings** Following [19], we briefly explain how we can achieve OLE over larger rings (which can potentially have super-polynomial size in  $\lambda$ ).

*OLE over  $\mathbb{Z}_N$*   $= \mathbb{Z}_{q_1} \times \cdots \times \mathbb{Z}_{q_\delta}$ . Let  $N = \prod_{i=1}^{\delta} q_i$  be an integer (which might be superpolynomial in  $\lambda$ ) such that for all  $i \in [\delta]$   $q_i = \text{poly}(\lambda)$  are different prime numbers. Then, via the Chinese Remainder Theorem,  $\mathbb{Z}_N$  is isomorphic to  $\mathbb{Z}_{q_1} \times \cdots \times \mathbb{Z}_{q_\delta}$ . Thus, performing an OLE over  $\mathbb{Z}_N$  boils down to performing  $\delta$  OLEs over each one of the smaller fields  $\mathbb{Z}_{q_i}$ . It is easy to see that, if each OLE over  $\mathbb{Z}_{q_i}$  has overall rate 1, then the resulting OLE over  $\mathbb{Z}_N$  also achieves overall rate 1.

*OLE over extension fields.* We now show how these techniques can be adapted to perform OLE over an extension field  $\mathbb{F}_{q^k}$  of order  $q^k$  for a prime  $q$ . Here, we rely on the fact that multiplication over  $\mathbb{F}_{q^k}$  can be expressed as a linear function over the field  $\mathbb{Z}_q$ . That is, suppose that an element  $\mathbf{x} \in \mathbb{F}_{q^k}$  is of the form  $\mathbf{x} = x_1 + x_2\alpha + \cdots + x_k\alpha^{k-1}$  where each  $x_i \in \mathbb{Z}_q$  and  $\alpha$  is a symbol. Then, for elements  $\mathbf{a}, \mathbf{x} \in \mathbb{F}_{q^k}$  the product

$$\mathbf{x}\mathbf{a} = f_{1,\mathbf{a}}(\mathbf{x}) + f_{2,\mathbf{a}}(\mathbf{x})\alpha + \cdots + f_{k,\mathbf{a}}(\mathbf{x})\alpha^{k-1}$$

where each  $f_{i,\mathbf{a}}$  is a  $\mathbb{Z}_q$ -linear function which depends solely on  $\mathbf{a}$ .

Given this, we briefly describe how we can perform several OLEs over  $\mathbb{F}_{q^k}$  while preserving overall rate 1. The receiver has input  $\mathbf{b} = (\mathbf{b}_1, \dots, \mathbf{b}_t) \in \mathbb{F}_{q^k}^t$  such that  $kt = m\ell$  and  $k|m$  (using the same notation as in Appendix H). It parses each  $\mathbf{b}_i$  as a  $k$ -dimensional vectors  $\bar{\mathbf{b}}_i \in \mathbb{Z}_q^k$ . Then, it organizes all  $t$  vectors  $\mathbf{b}_i$  in blocks  $\mathbf{c}_i \in \mathbb{Z}_q^m$  of size  $m$ . It inputs  $\mathbf{c} = (\mathbf{c}_1, \dots, \mathbf{c}_\ell)$  into the OMV protocol.

The sender, with input  $\mathbf{u}, \mathbf{v} \in \mathbb{F}_{q^k}$  rearranges  $\mathbf{u}, \mathbf{v}$  in the same way as the receiver and obtains  $\mathbf{w} = (\mathbf{w}_1, \dots, \mathbf{w}_\ell), \mathbf{z} = (\mathbf{z}_1, \dots, \mathbf{z}_\ell)$  respectively. Then, for each  $\mathbf{w}_i = (\mathbf{w}_{i,1}, \dots, \mathbf{w}_{i,m/k})$ , it computes the functions  $f_{j,\mathbf{w}_{i,r}}$  for each  $j \in [k]$ ,  $i \in [\ell]$  and  $r \in [m/k]$ . Let  $\bar{\mathbf{f}}_{j,\mathbf{w}_{i,r}}$  be the vector composed by the coefficients of  $f_{j,\mathbf{w}_{i,r}}$ . The sender computes the matrices

$$\bar{\mathbf{D}}_{i,r} = \begin{pmatrix} | & & | \\ \mathbf{f}_{1,\mathbf{w}_{i,r}} & \cdots & \mathbf{f}_{k,\mathbf{w}_{i,r}} \\ | & & | \end{pmatrix}$$

and then sets

$$\mathbf{D}_i = \begin{pmatrix} \bar{\mathbf{D}}_{i,1} & & \\ & \ddots & \\ & & \bar{\mathbf{D}}_{i,m/k} \end{pmatrix}.$$

It inputs  $\mathbf{D} = (\mathbf{D}_1, \dots, \mathbf{D}_\ell)$  and  $\mathbf{z}$  into the OMV protocol.

It is easy to see that the receiver's output will be  $\mathbf{b} \odot \mathbf{u} + \mathbf{v}$  where  $\odot$  denotes component-wise multiplication over  $\mathbb{F}_{q^k}$ . Moreover,  $\text{hw}(\mathbf{D}_i) = k$ . By choosing  $k$  such that  $k \leq \mu = m^{1-\zeta}$  we achieve a protocol with overall rate 1. In particular, we can set the parameters such that  $k = \lambda$  and we achieve an OLE over the field  $\mathbb{F}_{q^\lambda}$  of exponential size.

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## A Additional Preliminaries

### A.1 UC Security

In terms of security, we work in the standard UC-framework [13]. Let  $\mathcal{F}$  be a functionality,  $\pi$  a protocol that implements  $\mathcal{F}$  and  $\mathcal{E}$  be an environment, an entity that oversees the execution of the protocol in both the real and the ideal worlds. Let  $\text{IDEAL}_{\mathcal{F}, \text{Sim}, \mathcal{E}}$  be a random variable that represents the output of  $\mathcal{E}$  after the execution of  $\mathcal{F}$  with adversary  $\text{Sim}$ . Similarly, let  $\text{REAL}_{\pi, \mathcal{A}, \mathcal{E}}$  be a random variable that represents the output of  $\mathcal{E}$  after the execution of  $\pi$  with adversary  $\mathcal{A}$ .

In this work, we only consider semi-honest adversaries.

**Definition 13** *A protocol  $\pi$  implements  $\mathcal{F}$  if for every PPT adversary  $\mathcal{A}$  there is a PPT simulator  $\text{Sim}$  such that for all PPT environments  $\mathcal{E}$ , the distributions  $\text{IDEAL}_{\mathcal{F}, \text{Sim}, \mathcal{E}}$  and  $\text{REAL}_{\pi, \mathcal{A}, \mathcal{E}}$  are computationally indistinguishable.*

### A.2 Learning Parity with Noise

The LPN assumption is closely related to the problem of decoding a random linear code. Informally, it states that it is hard to find a solution for a noisy system of linear equations over  $\mathbb{Z}_2$ .

**Definition 14 (LPN assumption)** *Let  $n, m, t \in \mathbb{N}$  such that  $n \in \text{poly}(\lambda)$  and let  $\chi_{m,t}$  be uniform distribution over the set of error vectors of size  $m$  and hamming weight  $t$ . The Learning Parity with Noise (LPN) assumption  $\text{LPN}(n, m, \rho)$  holds if for any PPT adversary  $\mathcal{A}$  we have that*

$$\left| \Pr \left[ 1 \leftarrow \mathcal{A}(\mathbf{A}, \mathbf{s}\mathbf{A} + \mathbf{e}) : \begin{array}{l} \mathbf{A} \leftarrow_{\$} \{0, 1\}^{n \times m} \\ \mathbf{s} \leftarrow_{\$} \{0, 1\}^n \\ \mathbf{e} \leftarrow_{\$} \chi_{m,t} \end{array} \right] - \Pr \left[ 1 \leftarrow \mathcal{A}(\mathbf{A}, \mathbf{y}) : \begin{array}{l} \mathbf{A} \leftarrow_{\$} \{0, 1\}^{n \times m} \\ \mathbf{y} \leftarrow_{\$} \{0, 1\}^m \end{array} \right] \right| \leq \text{negl}(\lambda)$$

where  $\rho = m/t$  ( $\rho$  is called the noise rate).

In this work, we assume that the noise rate  $\rho$  is  $m^{1-\varepsilon}$  for any constant  $\varepsilon > 0$ . The LPN assumption is believed to be hard for that noise rate (see e.g. [5] and references therein).

For other missing preliminaries please refer to the full version paper.