

# Scalable Mixed-Mode MPC

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## Abstract

Protocols for secure multi-party computation (MPC) supporting mixed-mode computation have found a lot of applications in recent years due to their flexibility in representing the function to be evaluated. However, existing mixed-mode MPC protocols are only practical for a small number of parties: they are either tailored to the case of two/three parties, or scale poorly for a large number of parties.

In this paper, we design and implement a new system for highly efficient and scalable mixed-mode MPC tolerating an arbitrary number of semi-honest corruptions. Our protocols allow secret data to be represented in Encrypted, Boolean, Arithmetic, or Yao form, and support efficient conversions between these representations.

1. We design a multi-party table-lookup protocol, where both the index and the table can be kept private. The protocol is scalable even with hundreds of parties.
2. Using the above protocol, we design efficient conversions between additive arithmetic secret sharings and Boolean secret sharings for a large number of parties. For 32 parties, our conversion protocols require  $1184\times$  to  $8141\times$  less communication compared to the state-of-the-art protocols MOTION and MP-SPDZ; this leads to up to  $1275\times$  improvement in running time under 1 Gbps network. The improvements are even larger with more parties.
3. We also use new protocols to design an efficient multi-party distributed garbling protocol. The protocol could achieve asymptotically constant communication per party.

Our implementation is available at [\[GYKW23\]](#).

## 1 Introduction

Protocols for secure multi-party computation (MPC) allow a set of parties to jointly compute on their private data while revealing nothing beyond the output. In principle, general-purpose MPC protocols can evaluate an arbitrary program by first representing that program as a Boolean or arithmetic circuit; this will typically not be very efficient. *Mixed-mode* MPC protocols, on the other hand, allow different parts of a program to be represented (and securely computed) using different models of computation, e.g., part of the computation can be represented as a Boolean circuit and another part is represented using an arithmetic circuit. These protocols are of particular interest because they allow different parts of the program to be represented in the most suitable form and, therefore can achieve greater efficiency than utilizing a monolithic representation. As a result, they have found many applications, e.g., privacy-preserving machine learning and private biometric matching.

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The first general-purpose MPC protocol supporting mixed-mode computation is TASTY [HKS<sup>+</sup>10], which supports conversions between garbled circuits and computation using additive homomorphic encryption in the two-party setting. It was later improved by the ABY protocol [DSZ15] that supports Boolean circuits (via garbled circuits or the GMW protocol) and arithmetic circuits (via Beaver triples). Follow up works further improve the efficiency in the two-party setting by moving some operations to offline [PSSY21, BCD<sup>+</sup>20]. Other works have looked at decreasing the corruption threshold (e.g., [MR18, PS20, DEK21]), and have shown efficiency improvements in the three-party and four-party settings with one corruption (thus honest majority). In the multi-party case, tolerating any number of corruptions, Rotaru and Wood [RW19] proposed mixed-mode MPC protocols supporting Boolean and arithmetic circuits for both the semi-honest and malicious settings. This was further improved in subsequent work [DEF<sup>+</sup>19, EGK<sup>+</sup>20, BDST22].

Although there has been huge progress in bringing mixed-mode MPC to practical use, state-of-the-art protocols are still far from satisfactory in the following aspects:

- **Supporting MPC with massive participants.** Most existing mixed-mode MPC protocols are specifically tailored for 2–4 parties with a single corrupted party, which is useful but not sufficient. Protocols that can support an arbitrary number of parties [RW19, DEF<sup>+</sup>19, EGK<sup>+</sup>20, BDST22], require at least quadratic total communication complexity, rendering them inefficient for a massive number of participants.
- **Supporting high corruption thresholds.** Most protocols supporting mixed-mode computation (including all the aforementioned protocols for 2–4 parties) only allow one party to be corrupted. Exceptions are the work of Rotaru and Wood [RW19] and followup works [DEF<sup>+</sup>19, EGK<sup>+</sup>20, BDST22] that tolerate an arbitrary number of corruptions, and MPClan [KPPS23] that assumes an honest majority.
- **Constant round complexity.** There has been a long line of work in bringing garbled circuits to the multi-party setting to reduce round complexity. However, all existing solutions require the total communication quadratic in the number of parties [DI05, LPSY15, BLO17, WRK17, HOSS18, BCO<sup>+</sup>21]. The only exception is [BGH<sup>+</sup>23], with total communication independent of the number of parties but in the honest-majority setting.

Note that although there exists MPC protocols for Boolean or arithmetic circuits tolerating an arbitrary number of corruptions and with communication linear in the number of parties, all existing conversion protocols (as required by mixed-mode MPC) still require quadratic communication.

## 1.1 Our Contribution

In this work, we design and implement a scalable MPC protocol for mixed-mode computation. We focus on the semi-honest setting with all-but-one corruptions. Our system is designed to run with a large number of parties and, crucially use encrypted representations to reduce the communication complexity in computing using each format and in converting between them. In details:

1. **Multi-party private lookup table.** We design a multi-party table-lookup protocol that takes as input a public/secret-shared table and a secret-shared index, and outputs the table value at the given index in an encrypted representation. Encrypted representation can be further switched to arithmetic secret sharings with a small cost. The protocol requires communication linear in the number of parties and is thus highly scalable.
2. **Multi-party secret-sharing conversions.** Based on the lookup-table protocol, we design efficient conversions between Boolean and arithmetic additive secret sharings. Conversion from

a Boolean sharing to an arithmetic sharing is viewed as a private-index lookup in a size-2 table. Thus, the protocol has similar communication complexity as the lookup protocol and is highly scalable.

3. **Linear-complexity multi-party garbled circuits.** Based on a different variant of our table-lookup protocol, we design the first multi-party garbled-circuit protocol, tolerating an arbitrary number of semi-honest corruptions, with total communication linear in the number of parties. The protocol requires lattice-based additively homomorphic encryption in the private-key setting, and thus is not competitive with existing approaches for a small number of parties. However, we estimate that the inbound communication per party is better than quadratic-cost protocols [BLO16, WRK17] for more than 128 parties.
4. **Implementation and comparison.** We propose optimizations to fully utilize the features of our protocols, and implement the protocols in a project to be open-sourced. Compared to the state-of-the-art work MOTION [BDST22] in the same setting, for 32 parties, our system reports up to  $1184\times$  improvement in communication for arithmetic-to-Boolean (A2B) conversion and  $20\times$  improvement in communication for Boolean-to-arithmetic (B2A) conversion. For 64 parties, the running time of our protocols has  $369\times$  improvement for A2B and  $2247\times$  improvement for B2A, compared to another state-of-the-art work MP-SPDZ [Kel20]. Note that compared to MOTION, MP-SPDZ is less efficient but supports more parties such as 64 parties under the same hardware configuration. Our protocols improve the communication cost of MP-SPDZ for 64 parties by a factor of  $8819\times$  for A2B and  $15384\times$  for B2A.

## 2 Technical Overview

**Notation.** We use  $\kappa$  and  $\rho$  to denote the computational and statistical security parameters, respectively. For a finite set  $S$ , we use  $x \leftarrow S$  to denote that  $x$  is sampled uniformly from  $S$ . For a distribution  $\mathcal{D}$ , we denote by  $x \leftarrow \mathcal{D}$  sampling  $x$  according to the distribution  $\mathcal{D}$ . For two integers  $a, b$  with  $a \leq b$ , we use  $[a, b]$  to denote the set  $\{a, \dots, b\}$ . We use upper-case letters like  $T$  (or bold lower-case letters like  $\mathbf{x}$ ) to denote a column vector. For a vector (or bit-string)  $\mathbf{x}$ ,  $\mathbf{x}[j]$  denotes the  $j$ -th component of  $\mathbf{x}$ , where  $\mathbf{x}[0]$  is the first component of  $\mathbf{x}$ . All arithmetic operations are computed over a finite field  $\mathbb{Z}_p$ , where  $p$  is a prime and  $\ell = \lceil \log p \rceil$  is the length of a field element. We use  $\llbracket x \rrbracket$  to denote a homomorphic encryption (HE) ciphertext on a message  $x$ ,  $\langle x \rangle^a$  to denote an arithmetic additive sharing over  $\mathbb{Z}_p$ , and  $\langle x \rangle^b$  to denote a Boolean additive sharing with  $x \in \{0, 1\}$ . Let  $P_1, \dots, P_n$  be  $n$  parties. We use  $\langle x \rangle_i^a$  or  $\langle x \rangle_i^b$  to denote the share held by  $P_i$ .

### 2.1 Mixed-Mode MPC: Prior Solutions

Rotaru and Wood [RW19], who proposed doubly authenticated bits (daBits), is the first work for mixed-mode MPC in the multi-party setting tolerating any number of corruptions. Note that in the semi-honest setting, secret sharing without authentication is sufficient, but we still use daBit to refer to the underlying semi-honest construction. A daBit (in the semi-honest setting) refers to a secret bit  $r$  that is secret shared both in Boolean domain (namely  $\langle r \rangle^b$ ) and in arithmetic domain (namely  $\langle r \rangle^a$ ) over  $\mathbb{Z}_m$  for some  $m$ . Suppose that  $n$  parties hold the secret sharing of a bit  $r$ , i.e.,  $\langle r \rangle^b = (r^1, \dots, r^n)$ . Party  $P_i$ , with share  $r^i$ , further secret shares the bit  $r^i$  in  $\mathbb{Z}_m$  so that all parties hold  $\langle r^i \rangle^a$ . Now all parties need to compute  $\langle r \rangle^a = \langle r^1 \rangle^a \oplus \dots \oplus \langle r^n \rangle^a$ . Note that because arithmetic sharings do not support XOR operations directly, they need to be simulated using multiplication based on the fact that  $x \oplus y = x + y - 2 \cdot x \cdot y$ . Because there are a total

$n - 1$  number of XOR operations to compute, this protocol requires  $O(n)$  multiplications over  $\mathbb{Z}_m$  for each bit in the Boolean-to-arithmetic conversion. Even using multiplication triples with linear communication, the total communication for one conversion, which requires  $\ell$  daBits, would be  $O(n^2\ell^3)$  bits where  $\ell$  is the bit length of the number to be converted.

An alternative approach by Escudero et al. [EGK<sup>+</sup>20] is to generate extended daBit (edaBit) in the form of  $(\langle r \rangle^a, \langle r_0 \rangle^b, \dots, \langle r_{\ell-1} \rangle^b)$ , where  $r = \sum_{j \in [0, \ell-1]} r_j \cdot 2^j \in \mathbb{Z}_m$ . Their protocol works as follows: each party  $P_i$  picks a random  $r^i \in \mathbb{Z}_m$  and then secretly shares  $r^i$  to all parties in both Boolean and arithmetic sharings, i.e.,  $\langle r^i \rangle^a$  and  $\langle r^i \rangle^b$ . The arithmetic sharing  $\langle r \rangle^a = \sum_{i \in [1, n]} \langle r^i \rangle^a$ , which can be computed for free, and Boolean sharings  $(\langle r_0 \rangle^b, \dots, \langle r_{\ell-1} \rangle^b) = \sum_{i \in [1, n]} \langle r^i \rangle^b$ , which requires computing  $O(n\ell)$  multi-party AND triples. Using silent OT protocols [BCG<sup>+</sup>19, YWL<sup>+</sup>20], this requires  $O(n^3\ell)$  bits of communication with a small underlying constant; or one can use threshold FHE to get  $O(n^2\ell)$  where the underlying constant is at least 64, the ciphertext expansion to encrypt bits [CGGI17].

Once these (extended) daBit correlations are generated, the actual conversion can be performed easily by securely evaluating a circuit with size linear in the bit length, which is very cheap compared to the cost of generating the triples in the offline phase.

**Conclusion.** The above methods can be viewed as a trade-off between a larger number of parties ( $n$ ) and high bit-length ( $\ell$ ). Based on these methods, follow-up works [DEF<sup>+</sup>19, BDST22] further optimized the concrete efficiency when  $m = 2^\ell$ , but their best variations still have a complexity of either  $O(n^2\ell^3)$  or  $O(n^3\ell)$ .

## 2.2 Mixed-Mode MPC: Our Protocols

Our high-level idea is that since homomorphic encryption (HE) is a crucial tool to obtain linear communication-complexity monolithic-circuit MPC, we could also get linear communication conversions based on it. In particular, it is already known how to convert in linear complexity between additive secret sharings and the corresponding ciphertexts, with a secret key secretly shared among all parties.

**Secure table lookup in the multi-party setting.** Building towards scalable conversions, we first propose an efficient multi-party table lookup protocol. Suppose that we have a table  $T$  of size  $m = 2^\ell$  containing elements in  $\mathbb{Z}_p$ . First, we run a cheap arithmetic sharing to encryption protocol so that  $P_1$  holds the ciphertexts of all table entries, namely  $\llbracket T \rrbracket = (\llbracket T[0] \rrbracket, \dots, \llbracket T[m-1] \rrbracket)$  where  $T[i]$  is the  $i$ -th table entry. If  $T$  has multiple outputs, i.e.,  $T[i]$  has multiple elements, then all the entries corresponding to  $i$  can be packed in a single ciphertext  $\llbracket T[i] \rrbracket$ . Then,  $P_1$  picks a random string  $r^1 \leftarrow \{0, 1\}^\ell$  and locally permutes all ciphertexts to obtain  $\llbracket T_1 \rrbracket$  such that  $T_1[j] = T[j \oplus r^1]$  for each  $j \in [0, m-1]$ .  $P_1$  re-randomizes the permuted ciphertexts, and sends the resulting ciphertexts to  $P_2$ , who picks a random  $r^2 \leftarrow \{0, 1\}^\ell$  and performs a permutation to obtain  $\llbracket T_2 \rrbracket$  such that  $T_2[j] = T_1[j \oplus r^2]$  for all  $j \in [0, m-1]$ . Now  $P_2$  sends all ciphertexts  $\llbracket T_2 \rrbracket$  to the next party after re-randomization. Finally,  $P_n$  obtains the ciphertexts that encrypt a table permuted by all parties; in other words,  $P_i$  holds  $r^i$  as the share of  $r$ , and the ciphertexts on the permuted table  $\llbracket T_n \rrbracket$  such that for each  $j \in [0, m-1]$ ,  $T_n[j] = T[j \oplus r^1 \oplus \dots \oplus r^n] = T[j \oplus r]$ . Now with this setup, a private lookup to this table on index  $j$  can be performed efficiently given a Boolean sharing  $\langle j \rangle^b$ : the parties compute locally and reconstruct  $\langle j \oplus r \rangle^b = \langle j \rangle^b \oplus \langle r \rangle^b$  to  $P_n$ , who fetches  $\llbracket T_n[j \oplus r] \rrbracket = \llbracket T[j] \rrbracket$ . Then all parties convert it to an arithmetic sharing of the underlying plaintext  $T[j]$ .

**Boolean-to-arithmetic (B2A) conversion.** Our main idea for efficient conversion from Boolean to arithmetic secret sharing is to view this conversion as a lookup of a public table using a private

$$\begin{array}{|c|c|} \hline a & 1 \\ \hline b & 1 \\ \hline c & 0 \\ \hline d & 0 \\ \hline \end{array} * \begin{array}{|c|c|} \hline b & 0 \\ \hline c & 0 \\ \hline d & 1 \\ \hline a & 0 \\ \hline \end{array} + \begin{array}{|c|c|} \hline c & 0 \\ \hline d & 1 \\ \hline a & 0 \\ \hline b & 0 \\ \hline \end{array} * \begin{array}{|c|c|} \hline d & 0 \\ \hline a & 0 \\ \hline b & 0 \\ \hline c & 1 \\ \hline \end{array} = \begin{array}{|c|c|} \hline a & b \\ \hline b & d \\ \hline c & d \\ \hline d & c \\ \hline \end{array}$$

Figure 1: Example for permutation of packed table  $\llbracket(a, b, c, d)\rrbracket$  using  $r_1 = 0$  and  $r_2 = 1$ .

index. In more detail, we use a public table of size 2 with 0 and 1 in  $\mathbb{Z}_p$ . To convert the Boolean sharing  $\langle x \rangle^b$  of an integer  $x$  to its arithmetic sharing, we essentially just want to perform a table lookup for each XOR-shared bit in  $\langle x \rangle^b$ . This produces an arithmetic sharing of each bit in  $x$  just like daBit, which can further be locally combined to an arithmetic sharing of  $x$ .

The idea is simple, but to make it highly efficient, extensive protocol optimization is required to incorporate state-of-the-art optimization on HE schemes. In particular, the description above does not assume packing, which is important in reducing ciphertext expansion. To make it compatible with packing, we design a customized protocol for size-2 table lookup. The main challenge is to independently permute the encrypted entries within each table that are all packed into the same ciphertext as efficiently as possible. For illustration, suppose that we have two size-2 tables, namely  $(a, b)$  and  $(c, d)$ . To fully utilize packing, they will be packed in one ciphertext as  $e = \llbracket(a, b, c, d)\rrbracket$ . The key observation is that for a pair-wise swap, any slot after the swap can only come from its immediate neighbors, and thus shifting by one slot is sufficient. In more detail, we locally left shift and right shift  $e$  so that  $e_1 = \llbracket(b, c, d, a)\rrbracket$  and  $e_{-1} = \llbracket(d, a, b, c)\rrbracket$ . Suppose that we use bits  $r_1$  and  $r_2$  to indicate whether we should swap the table entries. Then the final result is

$$(\overline{r_1}, \overline{r_1}, \overline{r_2}, \overline{r_2}) * e + (r_1, 0, r_2, 0) * e_1 + (0, r_1, 0, r_2) * e_2,$$

which can be computed with three scalar multiplications, all in one layer. We illustrate an example in Figure 1. To finish up one party's computation, it needs to re-randomize the ciphertext using the circuit-privacy technique such as noise flooding [Gen09, AJL<sup>+</sup>12] to ensure that  $r_1$  and  $r_2$  cannot be inferred from the resulting ciphertext.

**Arithmetic-to-Boolean (A2B) conversion.** Our protocol for arithmetic-to-Boolean conversion follows similar ideas as above but with some extra complications. Our end goal is to generate edaBit correlations over  $\mathbb{Z}_p$ , but the above protocol only generates edaBit correlations over  $\mathbb{Z}_{2^\ell}$ , which can be higher than  $p$  for some probability. Thus, we need a protocol to perform secure rejection sampling efficiently. The simplest way is to perform a secure comparison, but the cost would be high. Furthermore, to be compatible with the mainstream lattice-based scheme,  $p$  must be NTT-friendly, imposing more restriction. In Section 5.2, we discuss how to pick  $p$  so that comparing a private integer and  $p$  takes only 2 multiplication operations.

With the above described optimizations, our conversion protocols have a running time linear in the number of parties. We observe a significant improvement in running time and communication compared to the prior state-of-the-art work. Furthermore, we estimate the cost of several end-to-end applications using mixed-mode circuits. We observe an improvement of about  $1490\times$  in the monetary cost for running biometric matching with 64 parties.

### 2.3 Scalable Multi-Party Garbled Circuits

As mentioned in the introduction, all existing multi-party garbled circuit (GC) protocols in the all-but-one corruption setting require total communication quadratic in the number of parties. The closest is [BLO17], which achieves linear online communication for the GCs (i.e., the function-dependent phase) by using a key and message homomorphic PRF, which can be built from, e.g.,

lattice-based assumptions. However, to distributedly generate the garbled circuits in the preprocessing phase (i.e., producing the additive sharings of keys), it still requires communication quadratic in the number of parties.

We build on the prior work [BLO17], and achieve the total communication *linear* in the number of parties. In their protocol, each wire  $w$  is associated with two keys  $\mathbf{k}_{w,0}$  and  $\mathbf{k}_{w,1}$  and a random mask  $\lambda_w$ . These keys and masks are additively shared among all parties. Due to the key-homomorphic property, the parties can evaluate their shares of the PRF values locally and later combine them together. Ignoring some details, the protocol works as follows. For each gate  $g$  with input wires  $u, v$  and output wire  $w$ , for each  $\alpha, \beta \in \{0, 1\}$ , each party  $P_i$  computes  $\text{PRF}_{\mathbf{k}_{u,\alpha}^i + \mathbf{k}_{v,\beta}^i}(g \parallel \alpha \parallel \beta) + \langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle_i^a$ , where  $e_{w,\alpha,\beta} = g((\lambda_u \oplus \alpha), (\lambda_v \oplus \beta)) \oplus \lambda_w$ ,  $\mathbf{k}_{u,\alpha}^i, \mathbf{k}_{v,\beta}^i$  are the shares of  $P_i$  for two keys  $\mathbf{k}_{u,\alpha}, \mathbf{k}_{v,\beta}$  and  $\langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle_i^a$  is the arithmetic share of  $P_i$  on the key  $\mathbf{k}_{w,e_{w,\alpha,\beta}}$ . The main communication cost is to compute  $\langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle^a$ . In [BLO17], this was accomplished by an OT-based protocol, which requires the  $O(n^2)$ -communication for  $n$  parties.

Our key observation is that the computation of  $\langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle^a$  can be viewed as a secure table lookup. In particular, computing an arithmetic sharing  $\langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle^a$  boils down to computing a Boolean sharing  $\langle e_{w,\alpha,\beta} \rangle^b$ , which can be generated in a small communication using one random Beaver triple over binary field. We use  $\langle e_{w,\alpha,\beta} \rangle^b$  to perform a table lookup, which has an efficient instantiation with  $O(n)$ -communication in previous discussions. For every AND gate, we need to compute 4 private table lookups, each corresponding to one arithmetic sharing  $\langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle^a$  with  $\alpha, \beta \in \{0, 1\}$ . For each XOR gate, we need to compute only one private table lookup. In particular, for each XOR gate with input wires  $u, v$  and output wire  $w$ , we observe that  $e_{w,\alpha,\beta} = \alpha \oplus \lambda_u \oplus \beta \oplus \lambda_v \oplus \lambda_w$ , and thus one table value is  $\mathbf{k}_{w,e_{w,0,0}} = \mathbf{k}_{w,e_{w,1,1}}$  and the other value in the table is  $\mathbf{k}_{w,e_{w,1,0}} = \mathbf{k}_{w,e_{w,0,1}}$ .

### 3 Preliminaries

We use the standard ideal/real paradigm [Can00] to prove the security of our protocols in the presence of a *semi-honest, static* adversary.

#### 3.1 Additive Secret Sharings

We use  $\langle x \rangle^t$  to denote an additive sharing over a finite field  $\mathbb{F}$  in the multi-party setting, where the superscript  $t \in \{a, b\}$  indicates the type of sharings. In particular,  $\langle x \rangle^a$  denotes an arithmetic sharing over a field  $\mathbb{F} = \mathbb{Z}_p$  where  $p$  is a prime;  $\langle x \rangle^b$  represents a Boolean sharing over a field  $\mathbb{F} = \mathbb{F}_2$ . Then, we define the following algorithms for two types of additive sharings.

- $\langle x \rangle^t \leftarrow \text{Share}(x)$  : The party  $P_j$ , who holds the secret  $x$ , runs this algorithm to generate an additive sharing  $\langle x \rangle^t$ . Specifically, this algorithm samples  $\langle x \rangle_i^t \leftarrow \mathbb{F}$  for  $i \in [1, n-1]$  and computes  $\langle x \rangle_n^t := x - \sum_{i \in [1, n-1]} \langle x \rangle_i^t \in \mathbb{F}$ .
- $x \leftarrow \text{Rec}(\langle x \rangle^t, i)$  : Given all shares  $\langle x \rangle_1^t, \dots, \langle x \rangle_n^t$ ,  $P_i$  can run this algorithm to reconstruct the secret  $x$ . Specifically, this algorithm outputs  $x := \sum_{j \in [1, n]} \langle x \rangle_j^t \in \mathbb{F}$ .
- $x \leftarrow \text{Open}(\langle x \rangle^t)$  : The open procedure is run as follows:
  1. All parties run  $\text{Rec}(\langle x \rangle^t, 1)$  such that  $P_1$  obtains  $x$ .
  2.  $P_1$  sends  $x$  to all other parties.

It is well-known that additive secret sharings satisfy the linear property. That is, for any constants  $c_0, c_1, \dots, c_\ell$ , given additive sharings  $\langle x_1 \rangle^t, \dots, \langle x_\ell \rangle^t$ , the parties  $P_1, \dots, P_n$  can *locally* compute  $\langle y \rangle^t := \sum_{i=1}^{\ell} c_i \cdot \langle x_i \rangle^t + c_0$  such that  $y = \sum_{i=1}^{\ell} c_i \cdot x_i + c_0 \in \mathbb{F}$ . If  $t = b$ , the addition operation of two elements in  $\mathbb{F}_2$  corresponds to the XOR operation of two bits. In this case, we can just write  $z = x \oplus y$  and  $\langle z \rangle^b = \langle x \rangle^b \oplus \langle y \rangle^b$  instead of  $z = x + y \in \mathbb{F}_2$  and  $\langle z \rangle^b = \langle x \rangle^b + \langle y \rangle^b$ . For a vector  $\mathbf{x} \in \mathbb{F}^\ell$ , we use  $\langle \mathbf{x} \rangle^t$  to denote  $(\langle \mathbf{x}[0] \rangle^t, \dots, \langle \mathbf{x}[\ell - 1] \rangle^t)$ . By  $\langle \mathbf{x} \rangle_i^t[j]$ , we denote the share of  $\langle \mathbf{x}[j] \rangle^t$  held by the party  $P_i$ .

### 3.2 Threshold Homomorphic Encryption

We use threshold homomorphic encryption (THE) to encrypt messages and perform operations over ciphertexts. In most cases, we only need THE to support linear combination (including addition and scalar multiplication) and rotation operations over ciphertexts. In a few special cases (e.g., producing Beaver triples as shown in Appendix 3.3), we require THE to additionally support one multiplication operation over two ciphertexts (i.e., depth-1 THE). Let  $\mathcal{P} = \{P_1, \dots, P_n\}$  be the set of  $n$  parties. Our protocols work in the full-threshold setting, i.e., the secret key is shared by all parties using additive secret sharing, and no party (even if  $n - 1$  parties collude) can recover the secret key. Let  $\mathcal{M}$  be the plaintext space. Following the previous work [AJL<sup>+</sup>12, BGG<sup>+</sup>18, MTBH21], a THE scheme over  $\mathcal{M}$  consists of the following algorithms and protocols:

- **Setup:**  $pp \leftarrow \text{Setup}(1^\kappa)$ . On input  $\kappa$ , the setup algorithm outputs a set of public parameters  $pp$ , which is an implicit input to the following algorithms and protocols.
- **Key Generation:** Every party  $P_i$  generates a share of a secret key by running  $\text{sk}_i \leftarrow \text{SecKeyGen}(pp)$ . The secret key  $\text{sk}$  is identical to  $\sum_{i \in [1, n]} \text{sk}_i$ . All parties jointly produce a public key  $\text{pk}$  by executing a multi-party key-generation protocol  $\text{pk} \leftarrow \prod_{\text{PubKeyGen}}(\text{sk}_1, \dots, \text{sk}_n)$ .
- **Encryption:**  $\llbracket m \rrbracket \leftarrow \text{Enc}_{\text{pk}}(m)$ . On input a public key  $\text{pk}$  and a plaintext  $m \in \mathcal{M}$ , the encryption algorithm outputs a ciphertext  $\llbracket m \rrbracket$ .
- **Evaluation:** We consider the following operations:
  - *Linear combination* : Given ciphertexts  $\llbracket m_1 \rrbracket, \dots, \llbracket m_\ell \rrbracket$  and public coefficients  $c_0, c_1, \dots, c_\ell$ , one can compute a ciphertext  $\llbracket m \rrbracket = \sum_{i=1}^{\ell} c_i \cdot \llbracket m_i \rrbracket + c_0$  such that  $m = \sum_{i=1}^{\ell} c_i \cdot m_i + c_0$ .
  - *Multiplication* : Given two ciphertexts  $\llbracket m_1 \rrbracket, \llbracket m_2 \rrbracket$ , any party can compute the ciphertext  $\llbracket m_3 \rrbracket = \llbracket m_1 \rrbracket \cdot \llbracket m_2 \rrbracket$  such that  $m_3 = m_1 \cdot m_2$ .

In the above definition, we abuse the notation for the sake of simplicity and use the same notation  $+$ ,  $\cdot$  to denote the addition and multiplication operations over both ciphertexts and plaintexts. From the context, it is clear that these operations over ciphertexts are actually different from that over plaintexts.

- **Decryption:** Given  $\text{sk} = \sum_{i \in [1, n]} \text{sk}_i$  and a ciphertext  $\llbracket m \rrbracket$ , one party can run  $\text{Dec}_{\text{sk}}(\llbracket m \rrbracket)$  to obtain a plaintext  $m$ . Given the secret key's shares  $\text{sk}_1, \dots, \text{sk}_n$  and a ciphertext  $\llbracket m \rrbracket$ , all parties jointly execute the decryption protocol  $\prod_{\text{Dec}}(\text{sk}_1, \dots, \text{sk}_n, \llbracket m \rrbracket)$  to let some party  $P_i$  obtain  $m$ .

Given a ciphertext  $ct = f(\llbracket m_1 \rrbracket, \dots, \llbracket m_\ell \rrbracket)$  for some function  $f$ , we require that the probability that  $\text{Dec}_{\text{sk}}(ct) \neq f(m_1, \dots, m_\ell)$  is negligible in  $\kappa$ . We also require that the THE scheme satisfies the standard CPA security. Informally, for any probabilistic polynomial time (PPT) adversary  $\mathcal{A}$  who corrupts at most  $n - 1$  parties, for plaintexts  $m_0, m_1$  chosen by  $\mathcal{A}$ , the probability that  $\mathcal{A}$  distinguishes  $\text{Enc}_{\text{pk}}(m_0)$  from  $\text{Enc}_{\text{pk}}(m_1)$  is negligible in  $\kappa$ .

**Circuit privacy.** For a ciphertext  $\llbracket m \rrbracket = f(\llbracket m_1 \rrbracket, \dots, \llbracket m_\ell \rrbracket)$  that will be decrypted, it is desirable that no parties (except for the party evaluating  $\llbracket m \rrbracket$  with  $f$ ) could learn the secret function  $f$ , even if they hold secret key  $\text{sk}$ . This is modeled as *circuit privacy*, whose formal definition can be found in [BdMW16]. As in prior work [dCJV20], we define an algorithm  $\text{CP}(ct, \text{pk})$ , which takes as input an evaluated ciphertext  $ct$  and public key  $\text{pk}$ , and outputs a ciphertext  $ct'$  with circuit privacy. We adopt the noise-flooding technique [AJL<sup>+</sup>12] to achieve circuit privacy (see Appendix A.1 for details). More efficient technique called “divide-and-round” [dCJV20] can also be applied to our protocols.

**Instantiation and packing.** In the implementation, we adopt a threshold version of the BGV-HE scheme [BGV12], which is outlined in Appendix A.1. Other full-threshold HE schemes, such as the BFV-THE scheme [Bra12, FV12, MTBH21] can also be applied in our protocols. For the BGV-THE scheme, every ciphertext is defined over a ring  $R_q = R/qR$ , and any plaintext lies in a ring  $R_p = R/pR$ , where  $R = \mathbb{Z}[X]/(X^N + 1)$  is a polynomial ring with integer coefficients modulo  $X^N + 1$ ,  $N$  is a power-of-two integer, and  $p, q \in \mathbb{N}$  are co-prime. Based on the packing technique, we can pack  $N$  plaintexts in a single ciphertext where every plaintext in  $\mathbb{Z}_p$  is placed in a different slot and support parallel evaluation of plaintexts using the single instruction multiple data (SIMD) operations. Suppose that a prime  $p = 1 \pmod{2N}$  is used. We can view a ring element  $a \in R_p$  as a vector in  $(\mathbb{Z}_p)^N$ . When using the packing technique, we often use  $\mathbf{m} \in (\mathbb{Z}_p)^N$  and  $\llbracket \mathbf{m} \rrbracket$  to denote a vector of plaintexts and its ciphertext, unless otherwise specified. Note that it is natural to generalize the evaluation of ciphertexts without packing into that with packing. Due to the usage of packing, we need to perform the following operations to rotate or permute the plaintext slots in a single ciphertext.

- **Rotation:** Any party can run  $\llbracket \mathbf{m}' \rrbracket \leftarrow \text{Rotate}_{\text{pk}}(\llbracket \mathbf{m} \rrbracket, r)$  such that  $\mathbf{m}'$  is a vector obtained by cyclically left-shifting (resp., right-shifting) the components of  $\mathbf{m}$  by  $r$  if  $r > 0$  (resp.,  $r < 0$ ).
- **Permutation:** Any party can run  $\llbracket \mathbf{u} \rrbracket \leftarrow \text{Perm}_{\text{pk}}(\llbracket \mathbf{m} \rrbracket, \pi)$  such that  $\mathbf{u} = (\mathbf{m}[\pi(0)], \mathbf{m}[\pi(1)], \dots, \mathbf{m}[\pi(N-1)])$ , where  $\pi$  is a permutation. The permutation operation can be realized by a linear combination of multiple rotations.

It is well known that the BGV-THE scheme is CPA secure under the ring-LWE assumption [LPR10].

### 3.3 Arithmetic Black Box and Conversions

We model MPC via the arithmetic black-box (ABB) model [KOS16], which is an ideal functionality  $\mathcal{F}_{\text{ABB}}$  defined in Figure 2. This functionality allows a set of  $n$  parties to input/output secret-shared values and evaluate arbitrary circuits performing addition and multiplication operations. As in [EGK<sup>+</sup>20], we define an extended version of the ABB model, which handles values in both arithmetic and Boolean domains and thus can evaluate any arithmetic/boolean circuits. Furthermore, this functionality is also extended to allow the parties to encrypt values and evaluate the addition of two ciphertexts. Without loss of generality, suppose that the plaintext space for encryption is  $\mathbb{Z}_p$ . Here, we do not allow the parties to evaluate the multiplication of two ciphertexts, as our protocols do not require it when invoking functionality  $\mathcal{F}_{\text{ABB}}$ . This functionality abstracts away the underlying details of secret sharings, encryption, and MPC.

**Instantiation for  $\mathcal{F}_{\text{ABB}}$ .** We can use a THE scheme to encrypt values and perform a linear combination of encrypted values. We adopt additive secret sharings to securely compute the linear combination and multiplication of secret values. Due to the linear property of additive secret sharings, the linear combination of multiple sharings can be locally computed. For multiplication of two secret sharings, we consider two cases:



### Functionality $\mathcal{F}_{\text{ABB}}$

This functionality operates over a finite field  $\mathbb{Z}_p$  (resp.,  $\mathbb{F}_2$ ) for arithmetic secret-shared values (resp., Boolean secret-shared values), and interacts with parties  $P_1, \dots, P_n$ .

**Input:** Upon receiving  $(\text{Input}, P_i, \text{type}, \text{id}, x)$  from a party  $P_i$  and  $(\text{Input}, P_i, \text{type}, \text{id})$  from all other parties, where  $\text{type} \in \{\text{arith}, \text{bool}\}$ ,  $\text{id}$  is a fresh identifier, and either  $x \in \mathbb{Z}_p$  or  $x \in \{0, 1\}$  depending on  $\text{type}$ , store  $(\text{id}, \text{type}, x)$ .

**Random:** Upon receiving  $(\text{Random}, \text{type}, \text{id})$  from all parties where  $\text{type} \in \{\text{arith}, \text{bool}\}$  and  $\text{id}$  is a fresh identifier, sample  $r \leftarrow \mathbb{Z}_p$  or  $r \leftarrow \{0, 1\}$  relying on  $\text{type}$ , store  $(\text{id}, \text{type}, r)$ .

**Encrypt:** Upon receiving  $(\text{Enc}, \text{id}, \text{id}')$  from all parties where  $\text{id}$  is present in memory, retrieve  $(\text{id}, \text{type}, x)$  and store  $(\text{id}', \text{enc}, x)$ .

**Linear combination:** Upon receiving  $(\text{LinComb}, \text{type}, \text{id}, \text{id}', c_0, c_1, \dots, c_\ell)$  from all parties, where  $(\text{id}[j], \text{type})$  for  $j \in [1, \ell]$  are present in memory, and  $c_j \in \mathbb{Z}_p$  (resp.,  $c_j \in \{0, 1\}$ ) for  $j \in [0, \ell]$  if  $\text{type} \in \{\text{arith}, \text{enc}\}$  (resp.,  $\text{type} = \text{bool}$ ), retrieve  $(\text{id}[j], \text{type}, x_j)$  for  $j \in [1, \ell]$ , then compute  $y := \sum_{j=1}^{\ell} c_j \cdot x_j + c_0$  modulo  $p$  if  $\text{type} \in \{\text{arith}, \text{enc}\}$  and modulo 2 if  $\text{type} = \text{bool}$ . Store  $(\text{id}', \text{type}, y)$ .

**Multiply:** Upon receiving  $(\text{Mult}, \text{type}, \text{id}_1, \text{id}_2, \text{id}_3)$  from all parties where  $(\text{id}_1, \text{type})$  and  $(\text{id}_2, \text{type})$  are present in memory and  $\text{type} \in \{\text{arith}, \text{bool}\}$ , retrieve  $(\text{id}_1, \text{type}, x)$  and  $(\text{id}_2, \text{type}, y)$ , compute  $z := x \cdot y$  modulo  $p$  if  $\text{type} = \text{arith}$  and modulo 2 if  $\text{type} = \text{bool}$ , and store  $(\text{id}_3, \text{type}, z)$ .

**Output:** Upon receiving  $(\text{Output}, P_i, \text{type}, \text{id})$  from all parties, where  $(\text{id}, \text{type})$  is present in memory, retrieve  $(\text{id}, \text{type}, x)$  and then output it to  $P_i$ .

Figure 2: Functionality for the MPC black box.

### Functionality $\mathcal{F}_{\text{Conv}}$

This functionality has all of the same features as  $\mathcal{F}_{\text{ABB}}$  with the following additional commands.

**From Boolean to Arithmetic:** Upon receiving  $(\text{B2A}, \text{id}, \text{id}')$  from all parties where  $(\text{id}, \text{bool})$  is present in memory, retrieve  $(\text{id}, \text{bool}, x)$  and store  $(\text{id}', \text{arith}, x)$ .

**From Arithmetic to Boolean:** Upon receiving  $(\text{A2B}, \text{id}, \text{id}_0, \dots, \text{id}_{\ell-1})$  from all parties where  $(\text{id}, \text{arith})$  is present in memory and  $\ell = \lceil \log p \rceil$ , retrieve  $(\text{id}, \text{arith}, x)$  and store  $(\text{id}_i, \text{bool}, x_i)$  for  $i \in [0, \ell - 1]$  where  $x = \sum_{i=0}^{\ell-1} x_i \cdot 2^i \pmod p$ .

**From Encryption to Arithmetic:** Upon receiving  $(\text{E2A}, \text{id}, \text{id}')$  from all parties where  $(\text{id}, \text{enc})$  is present in memory, retrieve  $(\text{id}, \text{enc}, x)$  and store  $(\text{id}', \text{arith}, x)$ .

**From Arithmetic to Encryption:** Upon receiving  $(\text{A2E}, \text{id}, \text{id}')$  from all parties where  $(\text{id}, \text{arith})$  is present in memory, retrieve  $(\text{id}, \text{arith}, x)$  and store  $(\text{id}', \text{enc}, x)$ .

Figure 3: Functionality for the black box of conversions.

- *Arithmetic sharings* : We can compute the multiplication of two arithmetic sharings  $\langle x \rangle^a$  and  $\langle y \rangle^a$  using threshold HE that supports 1-depth multiplications. Specifically, every party  $P_i$  with  $i \neq 1$  runs  $\llbracket \langle x \rangle_i^a \rrbracket \leftarrow \text{Enc}_{\text{pk}}(\langle x \rangle_i^a)$  and  $\llbracket \langle y \rangle_i^a \rrbracket \leftarrow \text{Enc}_{\text{pk}}(\langle y \rangle_i^a)$ , and then sends  $(\llbracket \langle x \rangle_i^a \rrbracket, \llbracket \langle y \rangle_i^a \rrbracket)$  to  $P_1$  who computes  $\llbracket x \rrbracket := \sum_{i=1}^n \llbracket \langle x \rangle_i^a \rrbracket$  and  $\llbracket y \rrbracket := \sum_{i=1}^n \llbracket \langle y \rangle_i^a \rrbracket$ , where  $\llbracket x \rrbracket_1, \llbracket y \rrbracket_1$  are computed by  $P_1$  by running  $\text{Enc}_{\text{pk}}(\cdot)$ . Then,  $P_1$  locally computes  $\llbracket z \rrbracket := \llbracket x \rrbracket \cdot \llbracket y \rrbracket$  and sends  $\llbracket z \rrbracket$  to all other parties. Finally, all parties call the E2A command of  $\mathcal{F}_{\text{Conv}}$  to convert  $\llbracket z \rrbracket$  into  $\langle z \rangle^a$ . This needs to send at most  $3.5(n - 1)$  HE ciphertexts in total, when the E2A command of  $\mathcal{F}_{\text{Conv}}$  is instantiated by the protocol shown in Figure 15 of Section A.3.

**Functionality  $\mathcal{F}_{\text{Prep-LUT}}$**

Let  $M = 2^m$  be the length of a public/private table. This functionality has all of the same features as  $\mathcal{F}_{\text{ABB}}$  shown in Figure 2, with the following additional commands.

**Public masked table:** Upon receiving  $(\text{MaskedPubTab}, T, \mathbf{id}_1, \mathbf{id}_2)$  from all parties, where  $T \in (\mathbb{Z}_p)^M$  is a vector defining a public table, and  $\mathbf{id}_1, \mathbf{id}_2$  are two vectors of fresh identifiers with respective length  $m$  and  $M$ , sample  $r \leftarrow \{0, 1\}^m$ , write  $r = (r_0, \dots, r_{m-1})$  with  $r_j \in \{0, 1\}$  for  $j \in [0, m-1]$ , store  $(\mathbf{id}_1[j], \text{bool}, r_j)$  for  $j \in [0, m-1]$ , and store  $(\mathbf{id}_2[j], \text{arith}, T[r \oplus j])$  for  $j \in [0, M-1]$ .

**Private masked table:** Upon receiving  $(\text{MaskedPriTab}, \mathbf{id}_1, \mathbf{id}_2, \mathbf{id}_3)$  from all parties, where  $(\mathbf{id}_1[j], \text{arith})$  for all  $j \in [0, M-1]$  are present in memory, and  $\mathbf{id}_2, \mathbf{id}_3$  are two vectors of fresh identifiers with respective length  $m$  and  $M$ , retrieve  $(\mathbf{id}_1[j], \text{arith}, T[j])$  for  $j \in [0, M-1]$ , set the vector  $T$  accordingly, sample  $r \leftarrow \{0, 1\}^m$  with  $r = (r_0, \dots, r_{m-1})$ , store  $(\mathbf{id}_2[j], \text{bool}, r_j)$  for  $j \in [0, m-1]$ , and store  $(\mathbf{id}_3[j], \text{arith}, T[r \oplus j])$  for  $j \in [0, M-1]$ .

Figure 4: Functionality for masked lookup tables.

- *Boolean sharings*: We can use the above approach based on threshold HE to multiply two Boolean sharings  $\langle x \rangle^b, \langle y \rangle^b$ , where XOR is simulated by multiplication and addition operations. In this way, the communication complexity of  $O(n)$  can be achieved, but the computation complexity is high. Alternatively, we can adopt the standard protocol based on correlated oblivious transfer (COT) to perform pairwise bit multiplications, and then locally combine the shares of these bit multiplications to obtain Boolean sharing  $\langle z \rangle^b$  with  $z = x \wedge y$ . We can use the recent PCG-like COT protocols (e.g., [BCG<sup>+</sup>19, YWL<sup>+</sup>20]) to generate COT correlations. Although the COT approach has the communication complexity of  $O(n^2)$ , it allows us to obtain much faster computation.

We can adopt the Beaver’s multiplication technique to improve the online performance. In this case, the online communication per multiplication is  $4(n-1) \log p$  bits (resp.,  $4(n-1)$  bits) in total for  $t = a$  (resp.,  $t = b$ ). In the offline phase, the parties first generate a random Beaver triple  $(\langle x \rangle^t, \langle y \rangle^t, \langle z \rangle^t)$  with  $z = x \cdot y$  and  $t \in \{a, b\}$ , where  $\langle x \rangle^t, \langle y \rangle^t$  are random additive sharings jointly produced by the parties, and  $\langle z \rangle^t$  is computed as described above. In the online phase, given additive sharings  $\langle u \rangle^t, \langle v \rangle^t$  associated with actual inputs, the parties can consume the random Beaver triple  $(\langle x \rangle^t, \langle y \rangle^t, \langle z \rangle^t)$  to generate an additive sharing  $\langle w \rangle^t$  with  $w = u \cdot v$  using the standard Beaver approach [Bea92].

**Functionality for arithmetic, Boolean and encryption conversions.** Our protocol would securely realize functionality  $\mathcal{F}_{\text{Conv}}$  shown in Figure 3. This functionality allows the parties to convert between arithmetic secret-shared values and Boolean secret-shared values and also allows them to convert between arithmetic secret-shared values and encrypted values. We omit the conversions between Boolean secret-shared values and encrypted values, as they can be realized by performing Boolean-to-arithmetic and arithmetic-to-encryption conversions. As for the conversions between arithmetic secret-shared values and encrypted values, we w.l.o.g. assume that the space of secret values for arithmetic sharings is the same as that of HE plaintexts. We show how to perform efficient conversions between arithmetic sharings and encrypted values in Appendix A.3 and will present our conversion protocols between arithmetic sharings and Boolean sharings in Section 5.

### Functionality $\mathcal{F}_{\text{LUT}}$

Let  $M = 2^m$  be the length of a public/private table. This functionality has all of the same features as  $\mathcal{F}_{\text{ABB}}$  shown in Figure 2, with the following additional commands.

**Public table lookup:** Upon receiving  $(\text{PubTabLookup}, T, \text{id}_1, \text{id}_2)$  from all parties, where  $T \in (\mathbb{Z}_p)^M$  is a vector defining a public table, and  $(\text{id}_1[j], \text{bool})$  for  $j \in [0, m - 1]$  are present in memory, retrieve  $(\text{id}_1[j], \text{bool}, x[j])$  for  $j \in [0, m - 1]$ , set  $x \in \{0, 1\}^m$  and store  $(\text{id}_2, \text{arith}, T[x])$ .

**Private table lookup:** Upon receiving  $(\text{PriTabLookup}, \text{id}_1, \text{id}_2, \text{id}_3)$  from all parties, where  $(\text{id}_1[j], \text{arith})$  for all  $j \in [0, M - 1]$  and  $(\text{id}_2[j], \text{bool})$  for  $j \in [0, m - 1]$  are present in memory, retrieve  $(\text{id}_1[j], \text{arith}, T[j])$  for  $j \in [0, M - 1]$  and  $(\text{id}_2[j], \text{bool}, x[j])$  for  $j \in [0, m - 1]$ , set  $T \in (\mathbb{Z}_p)^m$  and  $x \in \{0, 1\}^m$  accordingly, and store  $(\text{id}_3, \text{arith}, T[x])$ .

Figure 5: Functionality for MPC using look-up tables.

## 4 Multi-Party Lookup-Table Protocol

In this section, we present a multi-party lookup-table protocol with linear communication complexity, where either the table is public, or a private table is secretly shared. We separate the lookup-table protocol into two sub-protocols, where the preprocessing sub-protocol generates a masked table and the online sub-protocol realizes the lookup table using the masked table. We model the preprocessing of public/private masked tables in functionality  $\mathcal{F}_{\text{Prep-LUT}}$  shown in Figure 4. Functionality  $\mathcal{F}_{\text{Prep-LUT}}$  samples a random string  $r$  to permute the table  $T$ , which is equivalent to generating the additive sharings of masked table  $T'$  and  $r$ , where  $T'[j] = T[r \oplus j]$  for  $j \in [0, M - 1]$ . By invoking  $\mathcal{F}_{\text{Prep-LUT}}$ , our online protocol securely realizes functionality  $\mathcal{F}_{\text{LUT}}$  shown in Figure 5. Both functionalities  $\mathcal{F}_{\text{Prep-LUT}}$  and  $\mathcal{F}_{\text{LUT}}$  are an extension of the lookup-table functionality [KOR<sup>+</sup>17] to additionally support private tables. In both  $\mathcal{F}_{\text{Prep-LUT}}$  and  $\mathcal{F}_{\text{LUT}}$ , we w.l.o.g. assume that the table size is power-of-two. As in [KOR<sup>+</sup>17], we also let  $\mathcal{F}_{\text{Prep-LUT}}$  and  $\mathcal{F}_{\text{LUT}}$  involve the commands defined in functionality  $\mathcal{F}_{\text{ABB}}$  shown in Figure 2. In Section 4.1 and Section 4.2, we describe two preprocessing protocols for generating masked tables. Then, we present the online protocol in Section 4.3.

Our conversion and multi-party garbling protocols shown in the next sections only need a lookup-table protocol for *size-2* tables. Therefore, we first describe the multi-party lookup-table protocol for size-2 tables. Then, we show how to construct a multi-party lookup-table protocol for any polynomial-sized tables, which may be of independent interest for other applications, e.g., secure AES evaluation.

### 4.1 Preprocessing for Masked Two-Sized Tables

In Figure 6, we show a multi-party preprocessing protocol  $\Pi_{\text{prepLUT}}^{\text{size2}}$  for masking the tables that have only two entries. This protocol works in the  $\mathcal{F}_{\text{E2A}}$ -hybrid model, where  $\mathcal{F}_{\text{E2A}}$  consists of the commands defined in  $\mathcal{F}_{\text{ABB}}$  (shown in Figure 2) and the E2A command defined in  $\mathcal{F}_{\text{Conv}}$  (shown in Figure 3). Besides, this protocol adopts a THE scheme supporting packing and SIMD, where the THE scheme adopts the plaintext space  $(\mathbb{Z}_p)^N$  for a prime  $p$ , and the number of plaintext slots is  $N$  (power of two).

**Theorem 1.** *Protocol  $\Pi_{\text{prepLUT}}^{\text{size2}}$  (shown in Figure 6) securely realizes functionality  $\mathcal{F}_{\text{Prep-LUT}}$  with size-2 tables (shown in Figure 4) against semi-honest adversaries in the  $\mathcal{F}_{\text{E2A}}$ -hybrid model, assuming that the THE scheme is CPA secure and satisfies circuit privacy.*

The proof of Theorem 1 is provided in Appendix B.1.

**Protocol  $\Pi_{\text{prepLUT}}^{\text{size2}}$**

**Inputs:** Parties  $P_1, \dots, P_n$  have the following inputs:

- *Case 1* : Let  $T \in (\mathbb{Z}_p)^2$  be a public vector corresponding to a size-2 public table. *Case 2* : Let  $\langle T \rangle^a$  be an arithmetic sharing w.r.t. a private vector  $T \in (\mathbb{Z}_p)^2$  that is related to a size-2 private table.
- The THE scheme with Enc and Rotate. Suppose that the public parameters and public key  $\text{pk}$  have been established.

**Preprocessing of a masked size-2 table:**

1. All parties call the (Random) command of functionality  $\mathcal{F}_{\text{E2A}}$  to sample a vector of random Boolean sharings  $\langle r \rangle^b$  with  $r = (r_0, r_1, \dots, r_{N/2-1}) \in \{0, 1\}^{N/2}$ .
2.  $P_1$  obtains a ciphertext  $\llbracket \mathbf{m} \rrbracket$  where  $\mathbf{m}[2j] = T[0]$  and  $\mathbf{m}[2j+1] = T[1]$  for  $j \in [0, N/2 - 1]$ , by executing the following depending on whether  $T$  is public or not.
  - In Case 1,  $P_1$  sets  $\mathbf{m}$  as above and runs  $\llbracket \mathbf{m} \rrbracket \leftarrow \text{Enc}_{\text{pk}}(\mathbf{m})$ .
  - In Case 2, every party  $P_i$  sets  $\mathbf{m}_i \in (\mathbb{Z}_p)^N$  as  $\mathbf{m}[2j] = \langle T \rangle_i^a[0]$  and  $\mathbf{m}[2j+1] = \langle T \rangle_i^a[1]$  for  $j \in [0, N/2 - 1]$ , and then runs  $\llbracket \mathbf{m}_i \rrbracket \leftarrow \text{Enc}_{\text{pk}}(\mathbf{m}_i)$ . For each  $i \neq 1$ ,  $P_i$  sends  $\llbracket \mathbf{m}_i \rrbracket$  to  $P_1$ , who computes  $\llbracket \mathbf{m} \rrbracket := \sum_{i \in [1, n]} \llbracket \mathbf{m}_i \rrbracket$ .
3. From  $i = 1$  to  $n$ , the parties execute the following steps:
  - (a)  $P_1$  sets  $c_0 := \llbracket \mathbf{m} \rrbracket$ . If  $i \neq 1$ ,  $P_i$  gets  $c_{i-1}$  from  $P_{i-1}$ .
  - (b)  $P_i$  computes two ciphertexts  $t_1 \leftarrow \text{Rotate}_{\text{pk}}(c_{i-1}, 1)$  and  $t_2 \leftarrow \text{Rotate}_{\text{pk}}(c_{i-1}, -1)$ .
  - (c)  $P_i$  initializes three zero-vectors  $\mathbf{h} = \mathbf{h}_1 = \mathbf{h}_2 = 0^N$ , and for each  $j \in [0, N/2 - 1]$ , does the following:
    - i. If  $\langle r_j \rangle_i^b = 0$ , then set  $\mathbf{h}[2j] = \mathbf{h}[2j+1] = 1$ .
    - ii. If  $\langle r_j \rangle_i^b = 1$ , then set  $\mathbf{h}_1[2j] = \mathbf{h}_2[2j+1] = 1$ .
  - (d)  $P_i$  computes  $c_i := \mathbf{h} \cdot c_{i-1} + \mathbf{h}_1 \cdot t_1 + \mathbf{h}_2 \cdot t_2$ , and then update  $c_i$  as a circuit-private ciphertext  $\text{CP}(c_i, \text{pk})$ .
  - (e) If  $i \neq n$ ,  $P_i$  sends  $c_i$  to  $P_{i+1}$ . If  $i = n$ ,  $P_n$  sends  $c_n$  to all parties.
4. The parties call functionality  $\mathcal{F}_{\text{E2A}}$  to convert ciphertext  $c_n$  into arithmetic sharings  $\langle T'_i \rangle^a$  for  $i \in [0, N/2 - 1]$ , where  $T'_i[j] = T[r_i \oplus j]$  for  $i \in [0, N/2 - 1]$ ,  $j \in \{0, 1\}$ .
5. The parties output additive sharings  $\langle r_i \rangle^b$  and  $\langle T'_i \rangle^a$  for  $i \in [0, N/2 - 1]$ .

Figure 6: Protocol for the preprocessing of masked size-2 tables in the  $\mathcal{F}_{\text{E2A}}$ -hybrid model.

**Communication complexity.** In a standard way, a polynomial number of random Boolean sharings can be generated using random seeds and a pseudo-random generator (PRG), and thus the communication to generate the sharings can be amortized to negligible. Let  $|ct|$  be the size of a THE packed ciphertext. When the table is private, the generation of ciphertext  $\llbracket \mathbf{m} \rrbracket$  takes the communication cost of  $(n-1)|ct|$  bits. The step (3) needs  $2(n-1)|ct|$  bits of communication in total. The protocol (shown in Figure 15 of Section A.3) instantiating the (E2A) command of functionality  $\mathcal{F}_{\text{E2A}}$  takes communication of  $(n-1)|ct|/2$  bits. Overall, protocol  $\Pi_{\text{prepLUT}}^{\text{size2}}$  (Figure 6) has the total communication cost of at most  $3.5(n-1)|ct|$  bits, and thus achieves the communication complexity linear in the number of parties.

**Reducing noise growth.** Every ciphertext  $c_i$  produced by  $P_i$  for  $i \geq 2$  is computed by performing rotation and scalar-multiplication operations over ciphertext  $c_{i-1}$ . In this case, the noise will grow

with the number of parties, which will lead to very large parameters. To solve the issue, we reduce the noise growth by refreshing every ciphertext  $c_i$  with the Bootstrapping operation. This optimization is described in Section 7.1, and keeps the ciphertext size almost constant. Instead of bootstrapping every ciphertext, one only needs to bootstrap the ciphertexts  $\{c_j\}$  where  $j = i \cdot k$  for  $i \in [1, n/k]$  and some integer  $k \geq 2$ , by tuning the parameters.

**Special case that table entries are in  $(\mathbb{Z}_p)^N$ .** When each table entry is taken from the message space of the packed THE scheme (i.e.,  $T[0], T[1] \in (\mathbb{Z}_p)^N$ ), we can use a simpler approach to generate a masked lookup table. The special case occurs in the application of our multi-party garbling protocol shown in Section 6. The preprocessing protocol for the special case is the same as the protocol  $\Pi_{\text{prepLUT}}^{\text{size}2}$  shown in Figure 6, except for the following differences:

1. A random Boolean sharing  $\langle r \rangle^b$  with  $r \in \{0, 1\}$  (instead of  $r \in \{0, 1\}^{N/2}$ ) is generated.
2. A THE ciphertext on a public/private table  $T$  is computed as  $(\llbracket \mathbf{x} \rrbracket, \llbracket \mathbf{y} \rrbracket)$ , where  $\mathbf{x} = T[0]$  and  $\mathbf{y} = T[1]$ .
3. From  $i = 1$  to  $n$ ,  $P_i$  does the following:
  - (a) If  $i = 1$ , set  $c_0 = (\llbracket \mathbf{x} \rrbracket, \llbracket \mathbf{y} \rrbracket)$ . If  $i \neq 1$ , receive  $c_{i-1}$  from  $P_{i-1}$ .
  - (b) Parse ciphertext  $c_{i-1} = (\llbracket \mathbf{x}_{i-1} \rrbracket, \llbracket \mathbf{y}_{i-1} \rrbracket)$ . If  $\langle r \rangle_i^b = 1$ , then *swap*  $(\llbracket \mathbf{x}_{i-1} \rrbracket, \llbracket \mathbf{y}_{i-1} \rrbracket)$  and update  $c_{i-1}$  accordingly. Otherwise, keep  $c_{i-1}$  unchanged.
  - (c) Compute THE ciphertexts with circuit privacy  $c_i := (\text{CP}(c_{i-1}[0], \text{pk}), \text{CP}(c_{i-1}[1], \text{pk}))$ .

Through the above approach, only one table (instead of  $N/2$  tables) is masked for each protocol execution. In the special case, it is unnecessary to bootstrap ciphertexts, as only circuit-privacy operations are involved and noise growth is slower. Therefore, for the special case, this protocol is more computation-efficient than the protocol  $\Pi_{\text{prepLUT}}^{\text{size}2}$ . The security of the improved protocol for the special case is easy to be proved in the presence of semi-honest adversaries following the proof of Theorem 1.

## 4.2 Preprocessing for Masked Poly-Sized Tables

Now, we describe the multi-party preprocessing protocol for masking any polynomial-sized tables. This protocol still works in the  $\mathcal{F}_{\text{E}2\text{A}}$ -hybrid model, and adopts the threshold HE scheme to encrypt the public/private table. This protocol makes the parties sequentially permute the encrypted table, and requires more rotation operations compared to the protocol  $\Pi_{\text{prepLUT}}^{\text{size}2}$  shown in Figure 6. The details of the protocol is shown in Figure 7. Similarly, to control the noise growth, we would adopt the bootstrapping technique to refresh evaluated ciphertexts. For the sake of simplicity, we do not involve the packing technique for the THE scheme. Below, we will give the overview how to adopt the packing technique to optimize the protocol for moderate-sized tables.

**Theorem 2.** Protocol  $\Pi_{\text{prepLUT}}^{\text{polysize}}$  (shown in Figure 7) securely realizes functionality  $\mathcal{F}_{\text{Prep-LUT}}$  with poly-sized tables (shown in Figure 4) against semi-honest adversaries in the  $\mathcal{F}_{\text{E}2\text{A}}$ -hybrid model, assuming that the THE scheme is CPA secure and satisfies circuit privacy.

The proof of Theorem 2 can be found in Appendix B.2.

**Optimization with packing.** When a packed THE scheme is adopted, we can further optimize the protocol if the table size  $M$  is two times smaller than the number of slots  $N$ . Let  $L = \lfloor N/M \rfloor$ . In this case, we can encrypt and permute  $L$  tables packed in a single ciphertext. In particular,

**Protocol  $\Pi_{\text{prepLUT}}^{\text{polysize}}$**

**Inputs:** Parties  $P_1, \dots, P_n$  have the following inputs:

- *Case 1* : Let  $T \in (\mathbb{Z}_p)^M$  be a vector corresponding to a size- $M$  public table. *Case 2* : Let  $\langle T \rangle^a$  be an arithmetic sharing w.r.t. a vector  $T \in (\mathbb{Z}_p)^M$  related to a size- $M$  private table. Let  $m$  be the length of indices, i.e.,  $M = 2^m$ .
- The THE scheme with Enc and Perm. Suppose that the public parameters and public key  $\text{pk}$  have been established.

**Preprocessing of a masked size- $M$  table:**

1. All parties call the (Random) command of  $\mathcal{F}_{\text{E2A}}$  to sample random Boolean sharings  $\langle r \rangle^b$  with  $r \in \{0, 1\}^m$ .
2.  $P_1$  computes a ciphertext  $\llbracket T \rrbracket$  by executing the following depending on if  $T$  is public.
  - In Case 1,  $P_1$  runs  $\llbracket T \rrbracket \leftarrow \text{Enc}_{\text{pk}}(T)$ .
  - In Case 2, every party  $P_i$  runs  $\llbracket T^i \rrbracket \leftarrow \text{Enc}_{\text{pk}}(\langle T \rangle_i^a)$ . For  $i \neq 1$ ,  $P_i$  sends  $\llbracket T^i \rrbracket$  to  $P_1$ , who computes  $\llbracket T \rrbracket := \sum_{i \in [1, n]} \llbracket T^i \rrbracket$ .
3. From  $i = 1$  to  $n$ , the parties execute the following steps:
  - (a) If  $i = 1$ ,  $P_1$  sets  $\llbracket T_0 \rrbracket := \llbracket T \rrbracket$ . If  $i \neq 1$ ,  $P_i$  receives  $\llbracket T_{i-1} \rrbracket$  from  $P_{i-1}$ .
  - (b)  $P_i$  defines  $\pi_i$  as  $\pi_i(j) = j \oplus \langle r \rangle_i^b \in \{0, 1\}^m$  for  $j \in [0, M - 1]$ , and then runs  $\llbracket T_i \rrbracket \leftarrow \text{Perm}_{\text{pk}}(\llbracket T_{i-1} \rrbracket, \pi_i)$ . Then  $P_i$  updates  $\llbracket T_i \rrbracket$  as a circuit-private ciphertext  $\text{CP}(\llbracket T_i \rrbracket, \text{pk})$ .
  - (c) If  $i \neq n$ ,  $P_i$  sends  $\llbracket T_i \rrbracket$  to  $P_{i+1}$ . If  $i = n$ ,  $P_n$  sends  $\llbracket T_n \rrbracket$  to all other parties.
4. The parties call  $\mathcal{F}_{\text{E2A}}$  to convert ciphertext  $\llbracket T_n \rrbracket$  into a vector of arithmetic sharings  $\langle T' \rangle^a$  with  $T'[j] = T[r \oplus j]$  for  $j \in [0, m - 1]$ .
5. The parties output additive sharings  $\langle r \rangle^b$  and  $\langle T' \rangle^a$ .

Figure 7: Protocol for the preprocessing of masked poly-sized tables in the  $\mathcal{F}_{\text{E2A}}$ -hybrid model.

every party can permute each encrypted table independently and randomly. For a moderate table size  $M$ , we can select a suitable parameter  $N$  to obtain a better efficiency.

**Communication complexity.** The analysis of communication complexity of protocol  $\Pi_{\text{prepLUT}}^{\text{polysize}}$  (shown in Figure 7) is totally similar to that of protocol  $\Pi_{\text{prepLUT}}^{\text{size2}}$  (shown in Figure 6). In particular,  $\Pi_{\text{prepLUT}}^{\text{polysize}}$  has the total communication cost of at most  $3.5(n - 1)|ct|$  bits and thus a linear communication complexity. When the packing optimization as described above is used, a single protocol execution can be used to generate  $L = \lfloor N/M \rfloor$  masked lookup tables.

### 4.3 Online Protocol for Lookup Table

The online lookup-table protocol follows the known approach [IKM<sup>+</sup>13, DNNR17, KOR<sup>+</sup>17, DKS<sup>+</sup>17, BHS<sup>+</sup>23], and allows the table to be public or private. The detailed protocol is shown in Figure 8, and works in the  $\mathcal{F}_{\text{Prep-LUT}}$ -hybrid model. This protocol takes an input a vector of Boolean sharings  $\langle x \rangle^b$  with  $x \in \{0, 1\}^m$  and outputs an arithmetic sharing  $\langle T[x] \rangle^a$ . In the  $\mathcal{F}_{\text{Prep-LUT}}$ -hybrid model, protocol  $\Pi_{\text{Lookup}}$  shown in Figure 8 only needs the communication of  $2(n - 1)m$  bits, where  $m$  is the length of a table entry.

**Protocol  $\Pi_{\text{Lookup}}$**

**Input:** Parties  $P_1, \dots, P_n$  have the following inputs:

- Let  $M = 2^m$  is the table length.
- *Case 1* : Let  $T \in (\mathbb{Z}_p)^M$  be a public vector corresponding to a public-table map  $f : \{0, 1\}^m \rightarrow \mathbb{Z}_p$  such that  $T[j] = f(j)$  for  $j \in \{0, 1\}^m$ . *Case 2* : Let  $\langle T \rangle^a$  be the arithmetic sharing of a private vector  $T$  related to a private-table map.
- $\langle x \rangle^b$  is the Boolean sharings of a private index  $x \in \{0, 1\}^m$ .

**Preprocessing of masked table:** In Case 1 (resp., Case 2), all parties call the (MaskedPubTab,  $T$ ) (resp., (MaskedPriTab)) command of functionality  $\mathcal{F}_{\text{Prep-LUT}}$  to generate a masked table  $(\langle r \rangle^b, \langle T' \rangle^a)$  with  $r \in \{0, 1\}^m$  and  $T'[j] = T[r \oplus j]$  for  $j \in [0, M - 1]$ .

**Online table lookup:** Given  $(\langle r \rangle^b, \langle T' \rangle^a)$  and  $\langle x \rangle^b$ , the parties generate  $\langle T[x] \rangle^a$  as follows:

1. All parties locally compute  $\langle u \rangle^b := \langle x \rangle^b \oplus \langle r \rangle^b$ .
2. The parties run the `Open`( $\langle u \rangle^b$ ) procedure such that they obtain  $u = x \oplus r \in \{0, 1\}^m$ .
3. The parties locally compute  $\langle T[x] \rangle^a := \langle T' \rangle^a[u]$ .

Figure 8: Online lookup-table protocol.

**Protocol  $\Pi_{\text{B2A}}$**

**Inputs:**  $P_1, \dots, P_n$  hold Boolean sharings  $\langle \mathbf{x} \rangle^b$  where  $\mathbf{x} \in (\mathbb{F}_2)^\ell$  and  $\ell = \lceil \log p \rceil$  is the length of an element in  $\mathbb{Z}_p$ .

**Conversion from Boolean to arithmetic sharings:**

1. All parties set a public vector  $T = (0, 1)$  corresponding to a public lookup-table  $f_T(j) = j$  for  $j \in \{0, 1\}$ .
2. For  $j \in [0, \ell - 1]$ , the parties call the (PubTabLookup) command of functionality  $\mathcal{F}_{\text{LUT}}$  on input  $(T, \langle \mathbf{x}[j] \rangle^b)$  to obtain  $\langle T[\mathbf{x}[j]] \rangle^a = \langle \mathbf{x}[j] \rangle^a$ .
3. The parties compute and output  $\langle x \rangle^a := \sum_{j=0}^{\ell-1} 2^j \cdot \langle \mathbf{x}[j] \rangle^a$ , where  $x = \sum_{j=0}^{\ell-1} 2^j \cdot \mathbf{x}[j] \pmod p$ .

Figure 9: Protocol for converting Boolean sharings to arithmetic sharings in the  $\mathcal{F}_{\text{LUT}}$ -hybrid model.

**Theorem 3.** *Protocol  $\Pi_{\text{Lookup}}$  (shown in Figure 8) securely realizes functionality  $\mathcal{F}_{\text{LUT}}$  in the presence of semi-honest adversaries in the  $\mathcal{F}_{\text{Prep-LUT}}$ -hybrid model.*

The proof of Theorem 3 is postponed to Appendix B.3.

## 5 Conversions of Sharings from LUT

We first show how to convert Boolean sharings into arithmetic sharings in the  $\mathcal{F}_{\text{LUT}}$ -hybrid model. Then, we describe the protocol to convert arithmetic sharings into Boolean sharings in the  $\mathcal{F}_{\text{LUT}}$ -hybrid model.

**Protocol  $\Pi_{A2B}$**

**Inputs:** Parties  $P_1, \dots, P_n$  hold an arithmetic sharing  $\langle x \rangle^a$  with  $x \in \mathbb{Z}_p$ . Let  $\ell = \lceil \log p \rceil$ .

**Conversion from arithmetic to Boolean sharings:**

1. All parties call the (Random) command of  $\mathcal{F}_{LUT}$  to sample a vector of Boolean sharings  $\langle \mathbf{r} \rangle^b$  with  $\mathbf{r} \in (\mathbb{F}_2)^\ell$ .
2. The parties execute  $\Pi_{B2A}$  shown in Figure 9 to obtain an arithmetic sharing  $\langle r \rangle^a$  with  $r = \sum_{j=0}^{\ell-1} 2^j \cdot \mathbf{r}[j]$ .
3. If  $(2^\ell - p)/2^\ell > 1/2^\rho$ , all parties call the LinComb and Mult commands of  $\mathcal{F}_{LUT}$  on input  $\langle \mathbf{r} \rangle^b$  to check if  $r = \sum_{j=0}^{\ell-1} 2^j \cdot \mathbf{r}[j] < p$ . If  $r \geq p$ , the parties go back to step 1.
4. All parties locally compute  $\langle u \rangle^a := \langle x \rangle^a - \langle r \rangle^a$ . Then, the parties run  $\text{Rec}(\langle u \rangle^a, 1)$  to make  $P_1$  reconstruct  $u = (x - r) \bmod p$ , and locally define a vector of Boolean sharings  $\langle \mathbf{u} \rangle^b$  via letting  $P_1$  set  $\langle \mathbf{u} \rangle_1^b[j]$  as the  $j$ -th bit of the bit-decomposition of  $u \in \mathbb{Z}_p$  and letting  $P_i$  for  $i \neq 1$  set  $\langle \mathbf{u} \rangle_i^b[j] = 0$  for  $j \in [0, \ell - 1]$ .
5. The parties call the Input, LinComb and Mult commands of  $\mathcal{F}_{LUT}$  on input  $(\langle \mathbf{u} \rangle^b, \langle \mathbf{r} \rangle^b)$  to compute a modulo-addition circuit, which takes as input  $\mathbf{u}, \mathbf{r} \in \{0, 1\}^\ell$  and outputs  $u + r \bmod p$ . Functionality  $\mathcal{F}_{LUT}$  returns  $\langle \mathbf{x} \rangle^b$  to the parties, where  $x = \sum_{j=0}^{\ell-1} 2^j \cdot \mathbf{x}[j] \bmod p$ .
6. The parties output  $\langle \mathbf{x} \rangle^b$  with  $\mathbf{x} \in (\mathbb{F}_2)^\ell$ .

Figure 10: Protocol for converting arithmetic sharings to Boolean sharings.

## 5.1 Boolean to Arithmetic Conversion

In Figure 9, we show a LUT-based protocol that converts a vector of Boolean sharings  $\langle \mathbf{x} \rangle^b$  into an arithmetic sharing  $\langle x \rangle^a$  with  $x = \sum_{j=0}^{\ell-1} 2^j \cdot \mathbf{x}[j] \bmod p$ . Specifically, all parties compute an arithmetic sharing for each bit  $\mathbf{x}[j]$  by defining a public table  $T = (0, 1)$  and calling  $\mathcal{F}_{LUT}$ . Then, the parties locally sum the arithmetic sharings on all bits to get  $\langle x \rangle^a$ . The communication cost of protocol  $\Pi_{B2A}$  shown in Figure 9 totally depends on  $\ell$  executions of the protocol that securely realizes  $\mathcal{F}_{LUT}$ . Thus,  $\Pi_{B2A}$  has the communication complexity of  $O(n\ell|ct|/N)$  bits.

**Theorem 4.** *Protocol  $\Pi_{B2A}$  (shown in Figure 9) securely realizes the B2A command of functionality  $\mathcal{F}_{\text{Conv}}$  against semi-honest adversaries in the  $\mathcal{F}_{LUT}$ -hybrid model.*

The proof of Theorem 4 is given in Appendix B.4.

## 5.2 Arithmetic to Boolean Conversion

In Figure 10, we describe a protocol to convert an arithmetic sharing  $\langle x \rangle^a$  to Boolean sharings  $\langle \mathbf{x} \rangle^b$  with  $x = \sum_{j=0}^{\ell-1} 2^j \cdot \mathbf{x}[j] \bmod p$ . This protocol also works in the  $\mathcal{F}_{LUT}$ -hybrid model where  $\mathcal{F}_{LUT}$  involves the commands defined in  $\mathcal{F}_{ABB}$  (shown in Figure 2), and invokes  $\Pi_{B2A}$  (shown in Figure 9) as a sub-protocol. Specifically, the parties sample a vector of random Boolean sharings  $\langle \mathbf{r} \rangle^b$  and run  $\Pi_{B2A}$  to get  $\langle r \rangle^a$ , where  $\mathbf{r} \in (\mathbb{F}_2)^\ell$  and  $r = \sum_{j=0}^{\ell-1} 2^j \cdot \mathbf{r}[j] \bmod p$ . They open  $u = x - r$ , and jointly compute a modulo-addition circuit with input  $\langle \mathbf{u} \rangle^b$  and  $\langle \mathbf{r} \rangle^b$  to obtain  $\langle \mathbf{x} \rangle^b$ , where  $\langle \mathbf{u} \rangle^b$  can be locally computed by the parties given  $u$ .

If  $(2^\ell - p)/2^\ell \leq 1/2^\rho$ , then  $r = \sum_{j=0}^{\ell-1} 2^j \cdot \mathbf{r}[j] \bmod p$  is indistinguishable from a uniform element in  $\mathbb{Z}_p$  except with probability at most  $1/2^\rho$ , where  $\mathbf{r}$  is a random vector in  $(\mathbb{F}_2)^\ell$ . Otherwise, we need a check if  $r < p$  to assure that  $r$  is random in  $\mathbb{Z}_p$  and re-sample  $\langle \mathbf{r} \rangle^b$  if  $r \geq p$ . In general, the



parties compute a comparison circuit with input  $\langle \mathbf{r} \rangle^b$  to decide if  $r < p$ . In the special case that  $p = 2^{32} - 2^{30} + 1$  used in our implementation, we provide a more efficient approach to determine if  $r < p$ . Particularly, the parties do the following:

1. During executing sub-protocol  $\Pi_{\text{B2A}}$ , all parties store the arithmetic sharings  $\langle \mathbf{r}[j] \rangle^a$  for  $j \in [0, \ell - 1]$ .
2. The parties set  $\langle a \rangle^a := \langle \mathbf{r}[\ell - 1] \rangle^a$  and  $\langle b \rangle^a := \langle \mathbf{r}[\ell - 2] \rangle^a$ , and compute  $\langle c \rangle^a := \sum_{j=0}^{\ell-3} \langle \mathbf{r}[j] \rangle^a$ .
3. The parties call the (Mult) command of  $\mathcal{F}_{\text{LUT}}$  on input  $(\langle a \rangle^a, \langle b \rangle^a, \langle c \rangle^a)$  to obtain  $\langle d \rangle^a$  with  $d = a \cdot b \cdot c \in \mathbb{Z}_p$ .
4. The parties run  $d \leftarrow \text{Open}(\langle d \rangle^a)$ , and output the bit indicating  $r < p$  if  $d = 0$  or  $r \geq p$  if  $d \neq 0$ .

**Theorem 5.** *Protocol  $\Pi_{\text{A2B}}$  (shown in Figure 10) securely realizes the A2B command of functionality  $\mathcal{F}_{\text{Conv}}$  against semi-honest adversaries in the  $\mathcal{F}_{\text{LUT}}$ -hybrid model.*

The proof of Theorem 5 is deferred to Appendix B.5.

**Communication complexity.** We analyze the communication complexity of protocol  $\Pi_{\text{A2B}}$  when instantiating the functionality  $\mathcal{F}_{\text{LUT}}$ . In step 1, random Boolean sharings can be computed with the PRG seeds that are established in the setup phase. As before, we ignore the communication cost to generate these seeds, as it can be amortized to negligible. In step 2,  $\Pi_{\text{B2A}}$  has the communication complexity of  $O(n\ell|ct|/N)$  bits where  $|ct|$  is the size of a THE ciphertext. In step 3, if  $p = 2^{32} - 2^{30} + 1$ , we need the communication complexity of  $O(n|ct|/N)$  bits using THE to multiply two arithmetic sharings; otherwise, we require the  $O(n^2\ell)$  communication complexity using the OT-based GMW protocol [GMW87] to compute the comparison circuit. In step 4, the open procedure needs the communication of  $O(n\ell)$  bits. In step 5, we use the GMW protocol to compute the modulo-addition circuit, which requires  $O(n^2\ell)$ -bit communication. Overall, the communication complexity of protocol  $\Pi_{\text{A2B}}$  is  $O(n\ell|ct|/N + n^2\ell)$  bits. In addition, the term  $O(n^2\ell)$  can be reduced to  $O(n\ell)$  when using THE to multiply two Boolean sharings in the GMW protocol. Therefore, the protocol  $\Pi_{\text{A2B}}$  is able to achieve the communication complexity linear in the number of parties.

## 6 Scalable Multi-Party Garbling

We present how to generate multi-party garbled circuits (MPGCs) using a key-homomorphic additive homomorphic encryption (AHE) scheme in the private-key setting and private table lookup. Building upon this, we describe a scalable MPC protocol with linear communication complexity and  $O(n)$  rounds. In Appendix A.2, we show that the BGV-AHE scheme [BGV12] in the private-key setting is key-homomorphic.

### 6.1 Private-Key AHE with Key Homomorphism

We provide the definition of key-homomorphic AHE schemes in the private-key setting. Let  $\mathcal{K}$  and  $\mathcal{M}$  denote the secret-key space and message space, respectively. We always assume that  $\mathcal{K} \subseteq \mathcal{M} \subseteq \mathbb{F}^N$  where  $\mathbb{F}$  is some finite field (e.g.,  $\mathbb{F} = \mathbb{Z}_p$ ) and  $N$  is a parameter determining the length of vectors. The private-key AHE scheme with the key-homomorphic property involves the following algorithms:

- **Setup:**  $pp \leftarrow \text{Setup}(1^\kappa)$ . The setup algorithm is defined as in the threshold HE scheme shown in Section 3.2.

### Protocol $\Pi_{\text{MPGC}}$

**Inputs:** Parties  $P_1, \dots, P_n$  have 1) a Boolean circuit  $\mathcal{C}$  with the set of circuit-input wires  $\mathcal{I}$ , the set of circuit-output wires  $\mathcal{O}$  and the set of all AND and XOR gates  $\mathcal{G}$ ; 2) a Boolean sharing  $\langle x_w \rangle^b$  for each input bit  $x_w \in \{0, 1\}$ ; 3) a key-homomorphic private-key AHE scheme equipped with  $(\text{SecKeyGen}, \text{Enc}, \text{Dec})$ . Suppose that the set of public parameters  $pp$  has been established.

#### Preprocessing phase for generating multi-party garbled circuit:

1. All parties call functionality  $\mathcal{F}_{\text{LUT}}$  to sample a random Boolean sharing  $\langle \lambda_w \rangle^b$  for every wire  $w$ . Then for each wire  $w$ , every party  $P_i$  samples two secret keys  $\mathbf{k}_{w,0}^i, \mathbf{k}_{w,1}^i$  by running  $\text{SecKeyGen}(pp)$ , such that  $\mathbf{k}_{w,0}^i[0] = 0$ , and  $\mathbf{k}_{w,1}^i[0] = 0$  if  $i \neq 1$  or  $\mathbf{k}_{w,1}^i[0] = 1$  if  $i = 1$ . These keys constitute two vectors of arithmetic sharings  $\langle \mathbf{k}_{w,0} \rangle^a$  and  $\langle \mathbf{k}_{w,1} \rangle^a$  where  $\mathbf{k}_{w,0} = \sum_{i \in [1,n]} \mathbf{k}_{w,0}^i$ ,  $\mathbf{k}_{w,1} = \sum_{i \in [1,n]} \mathbf{k}_{w,1}^i$ ,  $\mathbf{k}_{w,0}[0] = 0$  and  $\mathbf{k}_{w,1}[0] = 1$ .
2. For every gate  $g \in \mathcal{G}$  with input wires  $u, v$  and output wire  $w$ , the parties generate a Boolean sharing  $\langle e_{w,\alpha,\beta} \rangle^b$  for all  $\alpha, \beta \in \{0, 1\}$ , where  $e_{w,\alpha,\beta} \stackrel{\text{def}}{=} g((\lambda_u \oplus \alpha), (\lambda_v \oplus \beta)) \oplus \lambda_w \in \{0, 1\}$ .
  - (a) If  $g$  is an AND gate, then  $e_{w,\alpha,\beta} = \lambda_u \lambda_v \oplus \beta \lambda_u \oplus \alpha \lambda_v \oplus \alpha \beta \oplus \lambda_w$ . In this case, all parties call the **(Mult)** command of functionality  $\mathcal{F}_{\text{LUT}}$  on input  $(\langle \lambda_u \rangle^b, \langle \lambda_v \rangle^b)$  to compute a Boolean sharing  $\langle \lambda_u \lambda_v \rangle^b$ . Then, the parties locally compute  $\langle e_{w,\alpha,\beta} \rangle^b = \langle \lambda_u \lambda_v \rangle^b \oplus \beta \langle \lambda_u \rangle^b \oplus \alpha \langle \lambda_v \rangle^b \oplus \langle \lambda_w \rangle^b \oplus \alpha \beta$  for each  $\alpha, \beta \in \{0, 1\}$ .
  - (b) If  $g$  is a XOR gate, then  $e_{w,\alpha,\beta} = \lambda_u \oplus \lambda_v \oplus (\alpha \oplus \beta) \oplus \lambda_w$ . The parties locally compute  $\langle e_{w,\alpha,\beta} \rangle^b = \langle \lambda_u \rangle^b \oplus \langle \lambda_v \rangle^b \oplus \langle \lambda_w \rangle^b \oplus (\alpha \oplus \beta)$ .
3. For the output wire  $w$  of each gate  $g \in \mathcal{G}$ , all parties generate  $\langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle^a$  for all  $\alpha, \beta \in \{0, 1\}$  as follows:
  - (a) The parties define a vector of arithmetic sharings  $\langle T \rangle^a$  such that  $\langle T[j] \rangle^a = \langle \mathbf{k}_{w,j} \rangle^a$  for  $j \in \{0, 1\}$ .
  - (b) If  $g$  is an AND gate, then for each  $\alpha, \beta \in \{0, 1\}$ , all parties call the **(PriTabLookup)** command of functionality  $\mathcal{F}_{\text{LUT}}$  on input  $(\langle T \rangle^a, \langle e_{w,\alpha,\beta} \rangle^b)$  to obtain  $\langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle^a$ .
  - (c) If  $g$  is a XOR gate, the parties perform the following steps:
    - i. Call the **(MaskedPriTab)** command of functionality  $\mathcal{F}_{\text{Prep-LUT}}$  on input  $\langle T \rangle^a$  to generate  $\langle T' \rangle^a$  and  $\langle r \rangle^b$  such that  $r \in \{0, 1\}$  is a random bit and  $T'[j] = T[r \oplus j]$  for  $j \in \{0, 1\}$ .
    - ii. Locally compute  $\langle u \rangle^b := \langle e_{w,0,0} \rangle^b \oplus \langle r \rangle^b$ , and then execute **Open**( $\langle u \rangle^b$ ) to obtain  $u = e_{w,0,0} \oplus r \in \{0, 1\}$ .
    - iii. Set  $\langle \mathbf{k}_{w,e_{w,0,0}} \rangle^a = \langle \mathbf{k}_{w,e_{w,1,1}} \rangle^a = \langle T' \rangle^a[u]$  and  $\langle \mathbf{k}_{w,e_{w,0,1}} \rangle^a = \langle \mathbf{k}_{w,e_{w,1,0}} \rangle^a = \langle T' \rangle^a[u \oplus 1]$ .
4. For the output wire  $w$  of each gate  $g \in \mathcal{G}$ , for each  $\alpha, \beta \in \{0, 1\}$ , all parties compute a garbled row  $gg_{w,\alpha,\beta}$  as follows:
  - (a) Every party  $P_i$  sets a secret key  $\text{sk}_i := \mathbf{k}_{u,\alpha}^i + \mathbf{k}_{v,\beta}^i$ , and runs  $\llbracket \langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle_i^a \rrbracket \leftarrow \text{Enc}_{\text{sk}_i}((g, \alpha, \beta), \langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle_i^a)$ .
  - (b) For each  $i \neq 1$ ,  $P_i$  sends  $\llbracket \langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle_i^a \rrbracket$  to  $P_1$ , who computes  $\llbracket \mathbf{k}_{w,e_{w,\alpha,\beta}} \rrbracket := \sum_{i \in [1,n]} \llbracket \langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle_i^a \rrbracket$ , where  $\llbracket \mathbf{k}_{w,e_{w,\alpha,\beta}} \rrbracket = \sum_{i \in [1,n]} \text{Enc}_{\text{sk}_i}((g, \alpha, \beta), \langle \mathbf{k}_{w,e_{w,\alpha,\beta}} \rangle_i^a) = \text{Enc}_{\text{sk}}((g, \alpha, \beta), \mathbf{k}_{w,e_{w,\alpha,\beta}})$ , where  $\text{sk} = \sum_{i \in [1,n]} \text{sk}_i$ .
5. Now,  $P_1$  obtains a garbled circuit  $\mathcal{GC} = \{(gg_{w,0,0}, gg_{w,0,1}, gg_{w,1,0}, gg_{w,1,1})\}_{w \in \mathcal{W}}$ , where for each  $\alpha, \beta \in \{0, 1\}$ ,  $gg_{w,\alpha,\beta} = \llbracket \mathbf{k}_{w,e_{w,\alpha,\beta}} \rrbracket$  and  $\mathcal{W}$  is the set of output wires of all gates in  $\mathcal{G}$ .

Figure 11: Protocol for multi-party garbling in the  $(\mathcal{F}_{\text{Prep-LUT}}, \mathcal{F}_{\text{LUT}})$ -hybrid model.

- **Key Generation:**  $\text{sk} \leftarrow \text{SecKeyGen}(pp)$ . On input  $pp$ , the key-generation algorithm outputs a secret key  $\text{sk} \in \mathcal{K}$ .

Protocol  $\Pi_{\text{MPGC}}$ , continued

**Online phase for evaluating multi-party garbled circuit:** When  $\langle x_w \rangle^b$  for all  $w \in \mathcal{I}$  are known, the parties execute the following:

6. For each  $w \in \mathcal{I}$ , all parties locally compute  $\langle e_w \rangle^b = \langle x_w \rangle^b \oplus \langle \lambda_w \rangle^b$ , and then the parties run  $\text{Open}(\langle e_w \rangle^b)$  to obtain a masked bit  $e_w = x_w \oplus \lambda_w$ .
7. For each  $w \in \mathcal{I}$ , every party  $P_i$  with  $i \neq 1$  sends  $\mathbf{k}_{w,e_w}^i$  to  $P_1$ , who computes  $\mathbf{k}_{w,e_w} := \sum_{i \in [1,n]} \mathbf{k}_{w,e_w}^i$ .
8. In a topological order, for each gate  $g \in \mathcal{G}$  with input wires  $u, v$  and output wire  $w$ ,  $P_1$  holds  $(\mathbf{k}_{u,e_u}, e_u)$  and  $(\mathbf{k}_{v,e_v}, e_v)$ , and then computes  $\mathbf{k}_{w,e_w} := \text{Dec}_{\mathbf{k}_{u,e_u} + \mathbf{k}_{v,e_v}}((g, e_u, e_v), gg_{w,e_u,e_v})$  and sets  $e_w := \mathbf{k}_{w,e_w}[0]$ .
9. For each  $w \in \mathcal{O}$ ,  $P_1$  sets  $\langle y_w \rangle_1^b := e_w \oplus \langle \lambda_w \rangle_1^b$  and  $P_i$  for each  $i \neq 1$  sets  $\langle y_w \rangle_i^b := \langle \lambda_w \rangle_i^b$ , and the parties output  $\langle y_w \rangle^b$ .

Figure 12: Protocol for multi-party garbling in the  $(\mathcal{F}_{\text{Prep-LUT}}, \mathcal{F}_{\text{LUT}})$ -hybrid model, continued.

- **Encryption:**  $[\mathbf{m}]_{\text{sk},t} \leftarrow \text{Enc}_{\text{sk}}(t, \mathbf{m})$ . On input  $\text{sk}$ , a label  $t$  and a vector of messages  $\mathbf{m} \in \mathcal{M}$ , the encryption algorithm outputs a ciphertext  $[\mathbf{m}]_{\text{sk},t}$ . Suppose that the scheme is instantiated by a lattice-based AHE such as private-key BGV [BGV12] shown in Appendix A.2. In this case,  $t$  is used to retrieve/derive a vector  $\mathbf{a}$  and then  $\mathbf{a}$  is used to encrypt  $\mathbf{m}$ , which has been used in [BLO17]. When the context is clear, we simply write  $[\mathbf{m}]_{\text{sk},t}$  as  $[\mathbf{m}]$ .
- **Decryption:**  $\mathbf{m} \leftarrow \text{Dec}_{\text{sk}}(t, [\mathbf{m}])$ . On input the secret key  $\text{sk}$ , a label  $t$  and a ciphertext  $[\mathbf{m}]$ , the decryption algorithm outputs a vector of messages  $\mathbf{m}$ .
- **Key-homomorphic [AJL<sup>+</sup>12]:** Given two ciphertexts  $[\mathbf{m}_1]_{\text{sk}_1,t}$  and  $[\mathbf{m}_2]_{\text{sk}_2,t}$  under different secret keys  $\text{sk}_1, \text{sk}_2$  and the same label  $t$ , any party can locally compute a ciphertext  $[\mathbf{m}_3]_{\text{sk}_3,t} = [\mathbf{m}_1]_{\text{sk}_1,t} + [\mathbf{m}_2]_{\text{sk}_2,t}$  such that  $\text{sk}_3 = \text{sk}_1 + \text{sk}_2$  and  $\mathbf{m}_3 = \mathbf{m}_1 + \mathbf{m}_2$ .

A ciphertext  $[\mathbf{m}]$ , which is obtained by running the encryption algorithm or performing key-homomorphic addition operations, should be decrypted to the correct message  $\mathbf{m}$  with overwhelming probability. We need that the AHE scheme satisfies a simple variant of the CPA security. Specifically, any PPT adversary  $\mathcal{A}$  can make a query  $(\mathbf{m}_i, t_i)$  to the encryption oracle which returns  $[\mathbf{m}_i]_{\text{sk},t_i}$  to  $\mathcal{A}$  for each  $i \in [1, \ell]$  where  $\ell$  is the number of oracle queries, and then  $\mathcal{A}$  can choose two message-label pairs  $(\mathbf{m}_0^*, t_0^*)$  and  $(\mathbf{m}_1^*, t_1^*)$  with  $t_0^*, t_1^* \notin \{t_1, \dots, t_\ell\}$ . Then the probability that  $\mathcal{A}$  distinguishes  $[\mathbf{m}_0^*]_{\text{sk},t_0^*}$  from  $[\mathbf{m}_1^*]_{\text{sk},t_1^*}$  is negligible in  $\kappa$ . Furthermore, the CPA security holds for a polynomial number of secret keys, which is guaranteed using a standard hybrid argument. When a lattice-based AHE scheme is adopted,  $t$  uniquely determines  $\mathbf{a}$  used in encryption, and the security notion is naturally equivalent to the standard CPA security. In the private-key setting, we show that the BGV scheme with a single level [BGV12] is a key-homomorphic AHE scheme, which is described in Appendix A.2. Alternatively, the BFV-AHE scheme [Bra12, FV12] is another candidate.

## 6.2 Multi-Party Garbled Circuits

In Figure 11 and Figure 12, we give the details of the MPC protocol based on multi-party garbled circuits. Without loss of generality, we assume that only one party  $P_1$  can evaluate the garbled circuit, which is easy to be extended to support that all parties are able to evaluate the garbled circuit. To be compatible with the conversion protocols between arithmetic sharings and Boolean sharings shown in Section 5, we consider that the inputs and outputs of the parties are Boolean

sharings. In a general case that the inputs are secret bits, the parties can run the `Share` algorithm to generate corresponding Boolean sharings. To make  $P_1$  get the output, the parties can send the shares of the Boolean sharings on circuit-output wires to  $P_1$  who reconstructs the output bits by running the `Rec` algorithm. We can use a standard approach to support that all parties obtain different outputs.

We divide the MPC protocol into two phases: the preprocessing phase and online phase. In the preprocessing phase, for each wire  $w$ , the parties generate the Boolean sharing of a random mask  $\langle \lambda_w \rangle^b$ , and we refer to  $e_w = z_w \oplus \lambda_w$  as a masked value where  $z_w \in \{0, 1\}$  is an actual value for  $w$ . For each wire  $w$ , every party  $P_i$  samples two random keys  $\mathbf{k}_{w,0}^i, \mathbf{k}_{w,1}^i$  for the private-key AHE scheme, where the sums of these keys are the keys  $\mathbf{k}_{w,0}, \mathbf{k}_{w,1}$  encrypted in the garbled circuit. As in prior work [BLO17], we set the first components of vectors  $\mathbf{k}_{w,0}, \mathbf{k}_{w,1}$  are 0 and 1 respectively, which allows the evaluator to extract the masked value  $e_w$  from  $\mathbf{k}_{w,e_w}$ . At first glance, this loses one dimension of the secret key, which slightly reduces security. On the one hand, when using BGV as the AHE scheme, the secret keys can be sampled uniformly from  $R_3$ , and have already sufficiently high entropy to guarantee the security. On the other hand, as we need only private-key AHE, the secret keys can actually be sampled uniformly from  $R_q$  (instead of  $R_3$ ) where  $q \gg 3$ . For each gate with output wire  $w$ , the parties also compute the Boolean sharings of four masked values  $\langle e_{w,\alpha,\beta} \rangle^b$  for all  $\alpha, \beta \in \{0, 1\}$ , where  $\alpha, \beta$  enumerate all four possible values of masked values on input wires of the gate. These Boolean sharings  $\langle e_{w,\alpha,\beta} \rangle^b$  for each  $\alpha, \beta \in \{0, 1\}$  can be used to generate arithmetic sharings  $\langle \mathbf{k}_{w,e_w,\alpha,\beta} \rangle^a$  using our table-lookup approach. Through the key-homomorphic addition operations of the private-key AHE scheme, for each gate with input wires  $u, v$  and output wire  $w$ , the parties jointly compute the garbled rows like the classical Yao's GC.

$(\alpha, \beta)$	garbled row $gg_{w,\alpha,\beta}$
(0, 0)	$\text{Enc}_{\text{sk}=\mathbf{k}_{u,0}+\mathbf{k}_{v,0}}^{(g,0,0)}(\mathbf{k}_{w,e_w,0,0})$
(0, 1)	$\text{Enc}_{\text{sk}=\mathbf{k}_{u,0}+\mathbf{k}_{v,1}}^{(g,0,1)}(\mathbf{k}_{w,e_w,0,1})$
(1, 0)	$\text{Enc}_{\text{sk}=\mathbf{k}_{u,1}+\mathbf{k}_{v,0}}^{(g,1,0)}(\mathbf{k}_{w,e_w,1,0})$
(1, 1)	$\text{Enc}_{\text{sk}=\mathbf{k}_{u,1}+\mathbf{k}_{v,1}}^{(g,1,1)}(\mathbf{k}_{w,e_w,1,1})$

Table 1: **Garbled table for each gate  $g$  with wires  $u, v, w$ .**

In the online phase, for each circuit-input wire  $w$ , all parties open  $e_w$ , and then every party sends  $\mathbf{k}_{w,e_w}^i$  to  $P_1$  who reconstructs the key  $\mathbf{k}_{w,e_w}$ . Then,  $P_1$  can evaluate the garbled circuit by decrypting the corresponding garbled rows. Finally, for each circuit-output wire  $w$ , the parties can locally compute a Boolean sharing  $\langle y_w \rangle^b$ . In the following theorem, we prove that protocol  $\Pi_{\text{MPGC}}$  securely realizes the standard MPC functionality  $\mathcal{F}_{\text{MPC}}$ . In the following theorem, we prove that protocol  $\Pi_{\text{MPGC}}$  securely realizes functionality  $\mathcal{F}_{\text{MPC}}$ , which is defined as follows:

- Upon receiving (Input,  $\mathbf{id}, \mathbf{x}^i$ ) from every party  $P_i$ , where  $\mathbf{id}$  are vectors of fresh identifiers and  $\mathbf{x}^i \in \{0, 1\}^m$ , compute  $\mathbf{x} := \bigoplus_{i \in [1, n]} \mathbf{x}^i$  and store  $(\mathbf{id}[j], \mathbf{x}[j])$  for each  $j \in [0, \dots, m - 1]$ .
- Upon receiving (Eval,  $\mathbf{id}, \mathbf{id}', \mathcal{C}$ ) from all parties, where  $(\mathbf{id}[j], \mathbf{x}[j])$  for each  $j \in [0, \dots, m - 1]$  is present in memory and  $\mathcal{C} : \{0, 1\}^m \rightarrow \{0, 1\}^\ell$  is a Boolean circuit, compute  $\mathbf{y} := \mathcal{C}(\mathbf{x})$  and then store  $(\mathbf{id}'[j], \mathbf{y}[j])$  for each  $j \in [0, \dots, \ell - 1]$ .

**Theorem 6.** *Protocol  $\Pi_{\text{MPGC}}$  (shown in Figures 11 and 12) securely realizes functionality  $\mathcal{F}_{\text{MPC}}$  against semi-honest adversaries in the  $(\mathcal{F}_{\text{Prep-LUT}}, \mathcal{F}_{\text{LUT}})$ -hybrid model, assuming that the key-homomorphic AHE scheme is CPA secure.*

The proof of Theorem 6 can be found in Appendix B.6.

**Complexity analysis.** Below, we first analyze the communication complexity of the protocol  $\Pi_{\text{MPC}}$ . Let  $A$  and  $R$  denote the number of AND gates and the number of XOR gates respectively. In the step (1), random Boolean sharings are generated with the PRG seeds that are established in the setup phase. We ignore the communication cost generating the PRG seeds as it can be amortized. In the step (2), all parties compute one binary Beaver triple for each AND gate, which requires the communication complexity of at most  $O(n|ct|A)$  bits if  $|ct| \gg n$ , where  $|ct|$  is the AHE ciphertext size. In the step (3), the parties take the communication of  $O(n|ct|(A + R))$  bits to generate the arithmetic sharings of all secret keys using the table-lookup approach. In the step (4), for each gate, every party  $P_i$  sends  $4|ct|$ -bit ciphertexts to  $P_1$ ; and thus this step takes  $O(n|ct|(A + R))$ -bit communication. Overall, the communication complexity in the preprocessing phase is  $O(n \cdot |ct| \cdot |\mathcal{C}|)$  and thus is linear in the number of parties, where  $|\mathcal{C}| = A + R$  denotes the circuit size. As all gates can be garbled in parallel, the preprocessing phase takes  $O(n)$  rounds.

In the online phase, in the step (6),  $O(n|\mathcal{I}|)$  bits are taken to open the masked bits on all circuit-input wires; in the step (7),  $O(n \cdot |\text{sk}| \cdot |\mathcal{I}|)$  bits are needed to transfer the secret keys on all circuit-input wires, where  $|\mathcal{I}|$  is the number of circuit-input wires and  $|\text{sk}|$  is the size of secret keys for private-key AHE. The total communication in the online phase is  $O(n \cdot |\text{sk}| \cdot |\mathcal{I}|)$  bits. The computation cost for evaluating a garbled circuit is  $|\mathcal{C}|$  decryption operations and independent of the number of parties. The rounds of the online phase is three rounds and can be reduced to two rounds when the Open procedure is realized by letting every party send its share to all other parties rather than  $P_1$ .

## 7 Implementation Optimizations

We discuss two optimizations for the preprocessing protocols of masked lookup tables  $\Pi_{\text{prepLUT}}^{\text{size}2}$  (Figure 6) and  $\Pi_{\text{prepLUT}}^{\text{polysize}}$  (Figure 7), along with one optimization for the scalable MPC protocol  $\Pi_{\text{MPC}}$  (Figure 11).

### 7.1 Optimizations for Lookup Table Protocol

**Bootstrapping.** In protocol  $\Pi_{\text{prepLUT}}^{\text{size}2}$  (shown in Figure 6), we use the packing technique to reduce the communication cost incurred by ciphertext expansion, which in turn requires a customized private pair-wise swapping. To obtain the communication complexity linear in the number of parties, the underlying THE scheme requires a budget to support the number of scale-multiplication and rotation operations that is linear in the number of all parties. However, for a large number of parties, it would require a very large size of parameters. We use bootstrapping to refrain from having a large budget. We consider two approaches to realize bootstrapping for threshold HE.

1. *Non-interactive bootstrapping:* All parties run an MPC protocol to generate a bootstrapping key, and then use the key to bootstrap HE ciphertexts and thus reduce the noise size.
2. *Interactive bootstrapping:* The parties can convert an evaluated HE ciphertext to a vector of arithmetic sharings, and then convert it back to a fresh HE ciphertext, following the previous work [MTBH21].

The non-interactive bootstrapping allows us to achieve communication complexity linear in the number of parties. Our implementation adopts interactive bootstrapping as it is more efficient for most of the reasonable network configurations. However, the communication complexity is

now quadratic in the number of parties with a very small constant ( $\approx 0.1$ ). Both bootstrapping approaches as described above can also be applied in protocol  $\Pi_{\text{prepLUT}}^{\text{polysize}}$  (Figure 7) in the same way.

**Pipelining.** Our protocols (i.e.,  $\Pi_{\text{prepLUT}}^{\text{size2}}$  and  $\Pi_{\text{prepLUT}}^{\text{polysize}}$ ) do not require a party to remain active after it sends a HE ciphertext to the next party. This enables us to pipeline the computation, such that the computation complexity is linear in the number of parties rather than quadratic. Instead of processing all the ciphertexts at once and sitting idle, each party processes one ciphertext, then sends it to the next party and starts processing the next ciphertext. For a small number of parties where bootstrapping is not required, this method of pipelining works.

When interactive bootstrapping is involved, the parties are idle but waiting for a bootstrapping request to respond. To handle such case, we use the system called `poll`. On one thread, every party executes the protocol regularly as if the ring structure works (i.e., receiving a ciphertext from the previous party, processing it, and sending it to the next party), and on the other thread, it listens for any ready bootstrapping request and responds to it. We allow the maximum multiplicative depth for the underlying THE scheme to be 10, and choose the minimum budget that requires the same number of bootstrapping requests as with a budget of 10.

We evaluate the performance of the protocol  $\Pi_{\text{prepLUT}}^{\text{size2}}$  with and without pipelining. The performance evaluation is reported in Figure 13a. As shown in the figure, with pipelining the running time scales linearly with the number of parties rather than quadratically as without pipelining.

## 7.2 Optimization for Multi-Party Garbling

In the MPGC protocol (shown in Figures 11 and 12), every garbler  $P_i$  with  $i \neq 1$  sends four AHE ciphertexts for each gate to the evaluator  $P_1$ , who combines all the ciphertexts into a garbled circuit. When the circuit is large or the number of parties  $n$  is large,  $P_1$  needs a very large bandwidth to receive these ciphertexts. When the bandwidth of  $P_1$  is not sufficient, this would form an efficiency bottleneck. To solve the efficiency issue, we adopt a different communication pattern. That is, we let  $P_n$  send the AHE ciphertexts to  $P_{n-1}$ , who combines these ciphertexts with the AHE ciphertexts computed by itself; and then  $P_{n-1}$  sends the resulting ciphertexts to  $P_{n-2}$  and so on. This amortizes the  $O(n)$  communication of  $P_1$  to constant communication for every party. Nevertheless, this increases the round complexity from one round to  $n - 1$  rounds. To reduce the rounds, we can adopt a binary-tree architecture to transfer AHE ciphertexts. In particular, all the parties are arranged in a binary tree such that each node only interacts with its children and parent nodes. Two children nodes  $P_i$  and  $P_{i+1}$  send the AHE ciphertexts to their parent node  $P_{i+2}$ , who aggregates the AHE ciphertexts from three parties and then sends the resulting ciphertexts to the parent node of  $P_{i+2}$ . The communication bandwidth of every party is at most  $2 \times$  larger than the first approach, and the rounds are reduced from  $n - 1$  to  $\log n$ . In addition, the AHE ciphertexts generated by every party  $P_i$  can be sent in a pipelined way.

# 8 Performance Evaluation

## 8.1 Summary of Evaluation

We summarize the key findings from our performance evaluation below.

1. We show that pipelining protocol  $\Pi_{\text{prepLUT}}^{\text{size2}}$  (Figure 6) improves its execution time from quadratic in the number of parties to almost linear in the number of parties.
2. We compare our conversion protocols with the state-of-the-art works MOTION [BDST22] and MP-SPDZ [Kel20].

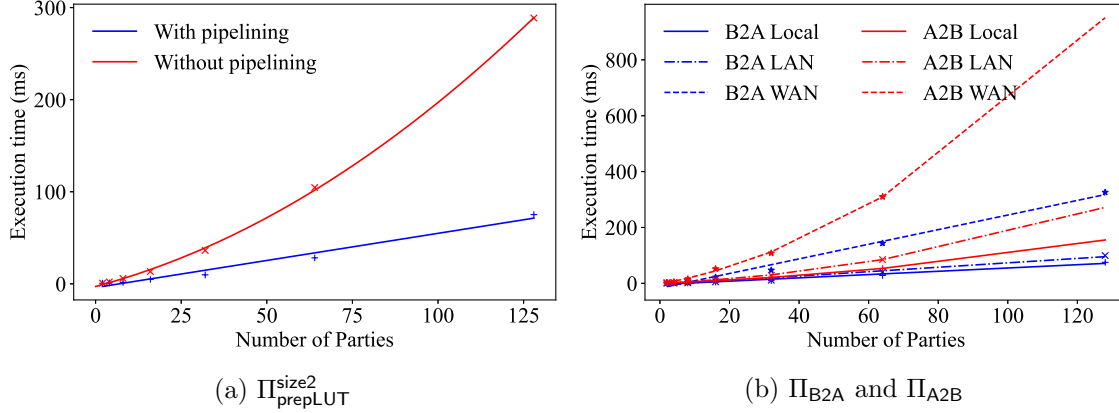


Figure 13: **Microbenchmarks of our protocols.** (a) The running time of protocol  $\Pi_{\text{prepLUT}}^{\text{size2}}$  with and without pipelining optimization. (b) Running time of the offline phase of our Boolean-to-arithmetic and arithmetic-to-Boolean conversion protocols.

- (a) In terms of running time, for 64 parties, our protocols improve MP-SPDZ by a factor of  $2247\times$  for B2A conversion and  $369\times$  for A2B conversion.
  - (b) In terms of communication cost, for 64 parties, our protocols improve MP-SPDZ by a factor of  $15384\times$  for B2A conversion and  $8819\times$  for A2B conversions; for 32 parties, we improve MOTION by factor of  $20\times$  for B2A and  $1184\times$  for A2B.
3. When applying our protocol on end-to-end applications, we achieve  $8242\times$  improvements in communication and up to  $1490\times$  improvements in monetary cost.
  4. Protocol  $\Pi_{\text{MPGC}}$  (Figure 11) reduces the inbound communication per party of optimized BMR [BLO16] by about 56 GB for evaluating an AES circuit among 128 parties. Note that the inbound communication is an efficiency bottleneck of multi-party distributed garbling, as a central party receives garbled circuits from all other parties.

## 8.2 Evaluation Setup

We implemented our protocols using EMP-toolkit [WMK16] for correlated OT and OpenFHE [BBB+22] for threshold homomorphic encryption. The implementation is open-sourced at [GYKW23]. All experiments are conducted on AWS of instance type `m5.2xlarge`. We consider three settings with up to 128 parties, which are described as follows:

1. **Local setting:** The network bandwidth is up to 10 Gbps with 0.1 ms latency.
2. **LAN setting:** The network bandwidth is up to 1 Gbps with 0.1 ms latency.
3. **WAN setting:** The network bandwidth is up to 200 Mbps with 100 ms latency.

For threshold HE in the public-key setting, we choose the parameters that achieve the 128-bit security level [ACC+19]. The plaintext prime  $p$  is equal to  $2^{32} - 2^{30} + 1$ , the length of the ciphertext prime  $q$  is more than 530 bits, and the number of slots  $N = 65536$ .

## 8.3 Performance of Conversions

We evaluate the performance of conversion protocols  $\Pi_{\text{A2B}}$  and  $\Pi_{\text{B2A}}$  in the offline phase and the online phase, respectively. We compare the performance of our conversion protocols with the

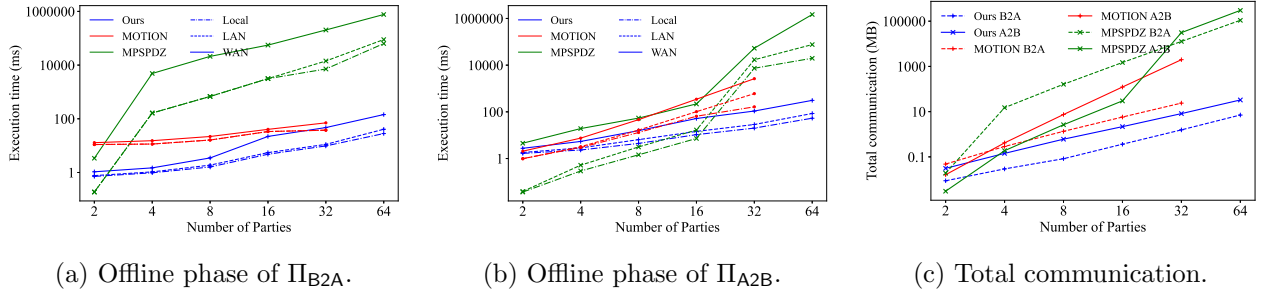


Figure 14: **Performance comparison for the conversion protocols of our framework, MOTION [BDST22] and MP-SPDZ [Kel20].** For the MOTION benchmarks for 32 parties, each party runs on a machine with double the resources than all other benchmarks.

Protocol	Setting	2	4	8	16	32	64
B2A	Local	0.038	0.046	0.05	0.066	0.11	0.494
	LAN	0.038	0.049	0.06	0.09	0.157	3.461
	WAN	0.066	0.076	0.181	0.357	0.988	4.696
A2B	Local	0.012	0.002	0.003	0.004	0.006	0.05
	LAN	0.006	0.004	0.006	0.012	0.023	0.049
	WAN	0.157	0.087	0.12	0.147	0.209	0.333

Table 2: **Performance of the online phase of our conversion protocols.** Running time is measured in milliseconds (*ms*), and the first row (2~64) is the number of parties. Running time is amortized over many conversions.

state-of-the-art protocols MOTION [BDST22] and MP-SPDZ [Kel20] for semi-honest security with all-but-one corruption. When using `m5.2xlarge`, MOTION can only execute up to 16 parties due to their high requirement of hardware resources. When we increase the instance size to `m5.4xlarge`, MOTION can be successfully executed with 32 parties; to further scale MOTION with 64 parties, one would need to use even larger machines. We note that our framework relies on the RLWE assumption, while MOTION (and MP-SPDZ, resp.) depend on LPN (and Minicrypt, resp.).

**Execution time of offline phase.** We evaluate the performance of the offline phase for our conversion protocols up to 128 parties in all three settings. The performance for  $\Pi_{B2A}$  and  $\Pi_{A2B}$  is reported in Figure 13b, which shows that the performance is linear in the number of parties. The execution time for the WAN setting grows faster than that in the LAN setting due to the communication overhead.

We compare the execution time with MOTION and MP-SPDZ for B2A and A2B conversions in Figure 14a and Figure 14b, respectively. We observe that our framework outperforms the existing conversion protocols in all three settings. Compared to MP-SPDZ, our B2A (resp., A2B) protocol improves the running time of the offline phase by a factor of  $2247\times$  (resp.,  $369\times$ ) for 64 parties. Compared to MOTION, in the offline phase, the running-time improvement of our B2A (resp., A2B) protocol is up to  $15\times$  (resp.,  $20\times$ ) for 32 parties.

**Execution time of online phase.** The online phase of our conversion protocols is interchangeable with that of MOTION and MP-SPDZ. We evaluate the online running time for our protocols in all three network settings. Benchmarks for the online time can be found in Table 2. We note that the running time of the online phase is at most 10% of the offline time.

**Communication cost of conversions.** We benchmark the total communication required by each



Triples	Setting	2	4	8	16	32	64
Boolean	Local	0.7	3.6	8.2	8.1	17	34.2
	LAN	0.7	3.5	8.5	9.3	18.6	37.2
	WAN	0.7	3.5	14.9	24.1	49.2	99.5
Arithmetic	Local	11.4	26.7	46.7	61	94.6	174.9
	LAN	12.6	27.9	63.4	109.3	191.1	418.2
	WAN	11.4	27.9	128	312.2	674	1621.5

Table 3: **Performance of Beaver triple generation.** Running time in microseconds ( $\mu s$ ) for Boolean sharings (using Ferret-COT) and arithmetic sharings (using BGV THE).

#Parties	Biomatch	Kmeans	MNIST	Gauss Dist.	Merge DB
4	1.93	6.16	25.49	0.07	0.24
16	9.7	17.79	94.44	0.27	1.18
64	64.97	76.6	387.7	1.52	8.51

Table 4: **Performance of end-to-end applications.** Performance estimation in seconds ( $s$ ) for running the end-to-end applications in the LAN setting.

conversion using our protocols, MOTION and MP-SPDZ. The comparison of the total communication cost is reported in Figure 14c. The multi-party LUT protocol used as the main building block in our conversion protocols does not require parties to communicate with every other party; thus, our system requires a significantly smaller amount of communication — up to  $20\times$  less than MOTION and  $15384\times$  less than MP-SPDZ for B2A and  $1184\times$  less than MOTION and  $8819\times$  less than MP-SPDZ for A2B.

#### 8.4 Performance of Boolean/Arithmetic Triples

For completeness, we benchmark the performance of triple generation for arithmetic and Boolean circuits in Table 3. With increasing number of parties, we observe that the cost of arithmetic triple generation increases slower than the cost of Boolean triple generation. This is because we use COT to generate the Boolean triples. This can be replaced by THE. However, using THE increases the computational overhead significantly.

#### 8.5 Performance for End-to-End Applications

We estimate the performance of several end-to-end applications using mixed-mode circuits generated by Silph [CZO+23]. The applications are listed with the number of operations in Table 6. The performance estimation to run the end-to-end applications is given in Table 4.

**Monetary cost analysis.** We analyze the monetary cost of running the applications among 64 parties using our framework and MP-SPDZ. Suppose that we run the instances in a single region,

Protocol	Biomatch	Kmeans	MNIST	Gauss Dist.	Merge DB
Ours	0.54	1.55	2.97	0.02	0.08
MP-SPDZ	803.3	399.8	102.45	21.75	10.05

Table 5: **Comparison of monetary cost for our framework and MP-SPDZ [Kel20].** The monetary cost estimates in USD for running the applications among 64 parties.

Applications	#AND	#MULT	#B2A	#A2B
Biomatch	23,205	1024	1028	256
Kmeans	1,568,660	8800	118	116
MNIST	1,624,460	666,600	1192	1
Gauss Dist.	8275	14	15	7
Merge DB	3930	200	201	1

Table 6: **The number of operations using mixed-mode circuits for end-to-end applications.**

and the communication cost is USD 0.01 per GB. We run each party on AWS of instance type m5.2xlarge, which costs USD 0.384 per hour. A detailed comparison is given in Table 5. We observe that MP-SPDZ is up to  $1490\times$  more expensive than our framework.

## 8.6 Communication Cost of MPGC

The OpenFHE library [BBB<sup>+</sup>22] does not support private-key AHE, and so we only give a conservative estimation of the communication cost of our protocol  $\Pi_{\text{MPGC}}$  (Figure 11). For the parameters with 128-bit security level, we select the plaintext prime  $p$  to be  $2^{16} + 1$ , the length of the ciphertext prime  $q$  is more than 45 bits, and the number of slots  $N = 4096$ . We find that our protocol has constant inbound and outbound communication per party. We report the inbound and outbound communication per party when securely computing an AES-128 circuit with 128 parties using our protocol  $\Pi_{\text{MPGC}}$  and the optimized BMR protocol [BLO16]. Both inbound and outbound communication for  $\Pi_{\text{MPGC}}$  is 51.12 GB. The inbound and outbound communication for BMR is 107.37 GB and 0.33 GB, respectively. Compared to BMR, the inbound communication per party for  $\Pi_{\text{MPGC}}$  is lower, while the outbound communication is higher.

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# A BGV Homomorphic Encryption

## A.1 Public-Key BGV Full-Threshold HE Scheme

We outline a full-threshold version of the public-key BGV HE scheme. We refer the reader to [BGV12, HS15, KPR18, MTBH21] for more details (e.g., rotation and bootstrapping). The set of public parameters  $pp$  defines the following parameters:

- The number of slots  $N$ . The plaintext modulus  $p$  and the ciphertext modulus  $q$  such that  $p$  and  $q$  are co-prime and  $n \cdot p < q$  where  $n$  is the number of parties, i.e., the plaintext and ciphertext are defined in  $R_p$  and  $R_q$  respectively.
- The standard variance  $\sigma$  defines a discrete Gaussian distribution  $\chi(\sigma)$ . For circuit privacy based on noise flooding,  $\sigma_{\text{cp}}$  defines another discrete Gaussian distribution  $\chi(\sigma_{\text{cp}})$ , where  $\sigma_{\text{cp}}$  is exponentially larger than  $\sigma$ .
- The polynomial ring  $R_3$  where the coefficients are picked from  $\{-1, 0, 1\}$ . Let  $\mathcal{Z}$  be the distribution in which sampling one polynomial in  $R_3$  such that each coefficient is 1 with probability 1/4,  $-1$  with probability 1/4 and 0 with probability 1/2.
- Let  $\mathcal{H}$  be the distribution sampling one polynomial in  $R_3$  such that at least  $h$  coefficients are non-zero for some parameter  $h$ . A common random polynomial  $a \leftarrow R_q$ .

Given the above public parameters, the BGV-THE scheme has the following algorithms:

- $\text{SecKeyGen}(pp)$ : Each party  $P_i$  samples  $s_i \leftarrow \mathcal{H}$  and sets  $\text{sk}_i = s_i$ .
- $\Pi_{\text{PubKeyGen}}(\text{sk}_1, \dots, \text{sk}_n)$ : Each party  $P_i$  samples  $e_i \leftarrow \chi(\sigma)$  and computes  $b_i := -a \cdot s_i + p \cdot e_i \in R_q$ . For each  $i \neq 1$ ,  $P_i$  sends  $b_i$  to  $P_1$ . Then,  $P_1$  computes  $b := \sum_{i=1}^n b_i \in R_q$  and sends  $b$  to all other parties. The parties  $P_1, \dots, P_n$  output  $\text{pk} = (a, b)$ .
- $\text{Enc}_{\text{pk}}(m)$ : To encrypt a message  $m \in R_p \cong (\mathbb{Z}_p)^N$ , sample  $v \leftarrow \mathcal{Z}, e_0, e_1 \leftarrow \chi(\sigma)$ , and compute  $c_0 := b \cdot v + p \cdot e_0 + m \in R_q$  and  $c_1 := a \cdot v + p \cdot e_1 \in R_q$ . Output a ciphertext  $\llbracket m \rrbracket = (c_0, c_1)$ .
- $\text{Dec}_{\text{sk}}(\llbracket m \rrbracket)$ : On input a secret key  $\text{sk} = \sum_{i \in [1, n]} \text{sk}_i$  and  $\llbracket m \rrbracket = (c_0, c_1)$ , write  $s = \text{sk}$ , compute  $m' := c_0 + s \cdot c_1 \in R_q$  and set  $m := m' \bmod p$ . Protocol  $\Pi_{\text{Dec}}(\text{sk}_1, \dots, \text{sk}_n, \llbracket m \rrbracket)$  is not used in our lookup table protocols, and thus is omitted.

**Circuit privacy with noise flooding** [Gen09]. The transformation algorithm  $\text{CP}(\llbracket m \rrbracket)$  is performed as follows:

1. Sample two large noises  $e'_0, e'_1 \leftarrow \chi(\sigma_{\text{cp}})$ , and then run  $\llbracket 0 \rrbracket \leftarrow \text{Enc}_{\text{pk}}(0; (e'_0, e'_1))$ , where  $\text{Enc}$  adopts  $e'_0, e'_1$  (instead of sampling noises from  $\chi(\sigma)$ ) to encrypt zero.
2. Output a circuit-private ciphertext  $ct := \llbracket m \rrbracket + \llbracket 0 \rrbracket$ .

The above approach can transform an evaluated ciphertext to a ciphertext with circuit privacy, using exponentially large noises to flood the noises underlying the ciphertext  $\llbracket m \rrbracket$ .

## A.2 Private-Key Key-Homomorphic BGV-AHE

Below, we outline the private-key BGV-AHE scheme with the key-homomorphic property. We refer the reader to [BGV12, JVC18] for more details. Let  $\{t_1, \dots, t_\ell\}$  be the set of all possible labels. As such, the set of public parameters  $pp$  in the private-key setting also define the parameters  $p, q, N$  and the error distribution  $\chi(\sigma)$ . Differently,  $pp$  now defines a set of random polynomials  $a_1, \dots, a_\ell \leftarrow R_q$ , where each polynomial  $a_i$  corresponds to a label  $t_i$ . If the set of labels is large, then the size of  $pp$  is very large. We have the following two approaches to solve the issue:

- Let  $F : \{0, 1\}^\kappa \times \{0, 1\}^\kappa \rightarrow R_q$  be a random oracle. Sample a random key  $\text{key} \leftarrow \{0, 1\}^\kappa$ , and any party can compute  $a_i := F(\text{key}, t_i)$  for each  $i \in [1, \ell]$ . Now,  $pp$  only needs to involve  $\text{key}$ , but this approach adds a random-oracle computation for each encryption.
- For the application of the MPGC protocol (described in Section 6), each label  $t_i$  corresponds to a triple  $(g, \alpha, \beta)$  where  $g$  is a gate and  $\alpha, \beta \in \{0, 1\}$ . In this application, following the work [BLO17], we can only put random polynomials  $a_1, \dots, a_{8 \cdot f_{out}} \in R_q$  into  $pp$ , where  $f_{out}$  is the maximal fan-out of the circuit. Using the algorithm in [BLO17], one can assign a random polynomial  $a_i$  into the encryption of each garbled row, such that any two gates sharing an input wire do not share any of the random polynomials. In this way,  $pp$  includes  $8 \cdot f_{out}$  polynomials in  $R_q$  but less computation is required.

In the following, we give the construction of the private-key BGV-AHE scheme.

- **SecKeyGen( $pp$ )**: Sample  $s \leftarrow R_q$  and output  $\text{sk} = s$ .
- **Enc $_{\text{sk}}(t, m)$** : On input a message  $m \in R_p$ , a label  $t$  and a secret key  $\text{sk} = s$ , retrieve  $a$  from  $pp$  according to  $t$ , sample  $e \leftarrow \chi(\sigma)$  and compute  $c := a \cdot s + p \cdot e + m \in R_q$ . Output a ciphertext  $\llbracket m \rrbracket = c$ .
- **Dec $_{\text{sk}}(t, \llbracket m \rrbracket)$** : On input a ciphertext  $\llbracket m \rrbracket = c$ , a label  $t$  and  $\text{sk} = s$ , retrieve  $a$  from  $pp$  based on  $t$ , compute  $m' := c - a \cdot s \in R_q$ , and output  $m := m' \bmod p$ .
- **Key-homomorphic addition**: Given two ciphertexts  $c_1 = a \cdot s_1 + p \cdot e_1 + m_1$  and  $c_2 = a \cdot s_2 + p \cdot e_2 + m_2$ , any party can compute  $c_3 := c_1 + c_2 = a \cdot (s_1 + s_2) + p \cdot (e_1 + e_2) + (m_1 + m_2)$ . Let  $s_3 = s_1 + s_2$ ,  $e_3 = e_1 + e_2$  and  $m_3 = m_1 + m_2$ . Then  $c_3 = a \cdot s_3 + p \cdot e_3 + m_3$ .

Under the ring-LWE assumption, it is easy to prove that the above private-key BGV-AHE scheme satisfies the CPA security (in Section 6.1), following prior works [BGV12, AJL<sup>+</sup>12].

## A.3 Conversions between BGV-THE Encryption and Arithmetic Sharings

We show how to convert between the BGV ciphertexts and arithmetic sharings, where the public-key BGV-THE scheme (shown in Appendix A.1) is used for encryption. In Figure 15, we describe the conversion protocol from BGV public-key encryption to arithmetic sharings. This protocol follows the protocol in [MTBH21], except for replacing BFV THE with BGV THE. In Figure 15, the noise  $e_i$  is sampled from a discrete Gaussian distribution  $\chi(\sigma_{\text{cp}})$ , which is used to keep  $s_i$  private based on noise flooding. The BGV-THE scheme supports the packing technique, and thus a single ciphertext encrypts a vector in  $(\mathbb{Z}_p)^N$  and the protocol would output a vector of arithmetic sharings. It is easy to prove that protocol  $\Pi_{\text{E2A}}$  (Figure 15) securely realizes the (E2A) command of functionality  $\mathcal{F}_{\text{Conv}}$  (Figure 3) under the ring-LWE assumption following the work [MTBH21].

For conversion from arithmetic sharings to encryption of BGV-THE, the protocol is constructed as follows:



**Protocol  $\Pi_{E2A}$**

**Inputs:** Parties  $P_1, \dots, P_n$  hold the following inputs:

- The set of public parameters  $pp$  for public-key BGV-THE.
- A ciphertext  $[[\mathbf{x}]] = (c_0, c_1)$  with  $\mathbf{x} \in R_p$  and  $c_0, c_1 \in R_q$ .
- $P_i$  holds a share of the secret key  $\mathbf{sk}_i = s_i$ .

**Conversion from encryption to arithmetic sharings:**

1. For each  $i \neq 1$ ,  $P_i$  samples  $e_i \leftarrow \chi(\sigma_{cp})$  and  $\langle \mathbf{x} \rangle_i^a \leftarrow R_p$ , and then sends  $h_i := s_i \cdot c_1 + p \cdot e_i - \langle \mathbf{x} \rangle_i^a \in R_q$  to  $P_1$ .
2. The party  $P_1$  computes  $\langle \mathbf{x} \rangle_1^a := s_1 \cdot c_1 + c_0 + \sum_{i=2}^n h_i$ .
3. All parties output  $\langle \mathbf{x} \rangle^a$  with  $\mathbf{x} \in (\mathbb{Z}_p)^N$ .

Figure 15: Protocol for converting BGV HE encryption into arithmetic sharings.

1.  $P_1, \dots, P_n$  are given a vector of arithmetic sharings  $\langle \mathbf{x} \rangle^a$  with  $\mathbf{x} \in (\mathbb{Z}_p)^N$  and a public key  $\mathbf{pk}$  for BGV-THE.
2. For  $i \in [1, n]$ ,  $P_i$  encodes  $\langle \mathbf{x} \rangle_i^a$  into a polynomial in  $R_p$  and then runs  $[[\langle \mathbf{x} \rangle_i^a]] \leftarrow \text{Enc}_{\mathbf{pk}}(\langle \mathbf{x} \rangle_i^a)$ .
3. For  $i \neq 1$ ,  $P_i$  sends  $[[\langle \mathbf{x} \rangle_i^a]]$  to  $P_1$ . Then,  $P_1$  computes  $[[\mathbf{x}]] := \sum_{i \in [1, n]} [[\langle \mathbf{x} \rangle_i^a]]$  and sends it to all other parties.

It is easy to see that if the BGV-THE scheme is CPA secure, then the above protocol securely realizes the (A2E) command of functionality  $\mathcal{F}_{\text{Conv}}$  (shown in Figure 3).

## B Formal Proofs of Our Protocols

In the proofs of security for our protocols, we need to prove that the joint distribution of the outputs of the adversary and honest parties in the real-world execution is computationally indistinguishable from that of the simulator and honest parties in the ideal-world execution. In particular, the ideal functionality allows the corrupted parties to input their shares, and samples random shares for honest parties such that the sum of all shares is identical to the secret. For all our proofs, we use  $\mathcal{H}$  and  $\mathcal{M}$  to denote the set of honest parties and the set of corrupted parties, respectively. Simulator  $\mathcal{S}$  simulates the view of adversary  $\mathcal{A}$ , and always outputs whatever  $\mathcal{A}$  outputs. Below, we give the detailed proofs of all theorems one by one.

### B.1 Proof of Theorem 1

**Theorem 7** (Theorem 1, restated). *Protocol  $\Pi_{\text{prepLUT}}^{\text{size2}}$  (shown in Figure 6) securely realizes functionality  $\mathcal{F}_{\text{Prep-LUT}}$  with size-2 tables (shown in Figure 4) against semi-honest adversaries in the  $\mathcal{F}_{E2A}$ -hybrid model, assuming that the THE scheme is CPA secure and satisfies circuit privacy.*

*Proof.* We construct a PPT simulator  $\mathcal{S}$  given access to functionality  $\mathcal{F}_{\text{Prep-LUT}}$  with size-2 tables that runs the PPT adversary  $\mathcal{A}$  as a subroutine, and emulates functionality  $\mathcal{F}_{E2A}$ . Specifically, the simulation is constructed as follows:

1. For generating a vector of Boolean sharings  $\langle r \rangle^b$ ,  $\mathcal{S}$  emulates  $\mathcal{F}_{E2A}$  by recording the shares of corrupted parties on  $\langle r \rangle^b$  sent by  $\mathcal{A}$  to  $\mathcal{F}_{E2A}$ .

2. If the table is private, for  $i \in \mathcal{H}$ ,  $\mathcal{S}$  sends a fresh zero-ciphertext  $\llbracket \mathbf{0} \rrbracket$  to  $\mathcal{A}$ , and initializes  $\llbracket \mathbf{m} \rrbracket := \llbracket \mathbf{0} \rrbracket$  if  $P_1$  is honest. If the table is public and  $P_1$  is honest,  $\mathcal{S}$  initializes  $\llbracket \mathbf{m} \rrbracket$  following the protocol specification.
3. For each  $i \in \mathcal{H}$ ,  $\mathcal{S}$  generates a fresh zero-ciphertext  $\llbracket \mathbf{0} \rrbracket$  and sends it to  $\mathcal{A}$ .
4. For each arithmetic sharing  $\langle T'_i \rangle^a$  with  $i \in [0, N/2 - 1]$ ,  $\mathcal{S}$  emulates  $\mathcal{F}_{E2A}$  by recording the corrupted parties' shares sent by  $\mathcal{A}$  to  $\mathcal{F}_{E2A}$ .
5. For each  $i \in [0, N/2 - 1]$ ,  $\mathcal{S}$  sends the shares of corrupted parties about  $\langle r_i \rangle^b$  and  $\langle T'_i \rangle^a$  to functionality  $\mathcal{F}_{\text{Prep-LUT}}$ .

It is easy to see that the simulation of  $\mathcal{F}_{E2A}$  is perfect. Under the assumption that the THE scheme is CPA secure, the simulation in the above step (2) using a zero ciphertext is computationally indistinguishable from the real ciphertexts. Furthermore, under the assumption that the THE scheme satisfies circuit privacy, the simulation in the above step (3) using a zero ciphertext is also computationally indistinguishable from the real ciphertext. In both worlds, the output shares of corrupted parties and honest parties satisfy that their sum is the correct value. Therefore, the joint distributions of the adversary's view and the honest parties' outputs are computationally indistinguishable in both worlds.  $\square$

## B.2 Proof of Theorem 2

**Theorem 8** (Theorem 2, restated). *Protocol  $\Pi_{\text{prepLUT}}^{\text{polysize}}$  (shown in Figure 7) securely realizes functionality  $\mathcal{F}_{\text{Prep-LUT}}$  with poly-sized tables (shown in Figure 4) against semi-honest adversaries in the  $\mathcal{F}_{E2A}$ -hybrid model, assuming that the THE scheme is CPA secure and satisfies circuit privacy.*

*Proof.* We construct a PPT simulator  $\mathcal{S}$  given access to functionality  $\mathcal{F}_{\text{Prep-LUT}}$  with polynomial-sized tables that runs the PPT adversary  $\mathcal{A}$  as a subroutine, and emulates  $\mathcal{F}_{E2A}$ . Specifically, the simulation is constructed as follows:

1. For generating a vector of Boolean sharings  $\langle r \rangle^b$  with  $r \in \{0, 1\}^m$ ,  $\mathcal{S}$  emulates  $\mathcal{F}_{E2A}$  by recording the shares of corrupted parties about  $\langle r \rangle^b$  sent by  $\mathcal{A}$  to  $\mathcal{F}_{E2A}$ .
2. If the table is private, for  $i \in \mathcal{H}$ ,  $\mathcal{S}$  sends a fresh zero-ciphertext  $\llbracket \mathbf{0} \rrbracket$  to  $\mathcal{A}$ , and initializes  $\llbracket \mathbf{m} \rrbracket := \llbracket \mathbf{0} \rrbracket$  if  $P_1$  is honest. If the table is public and  $P_1$  is honest,  $\mathcal{S}$  initializes  $\llbracket \mathbf{m} \rrbracket$  following the protocol description.
3. For each  $i \in \mathcal{H}$ ,  $\mathcal{S}$  generates a fresh zero-ciphertext  $\llbracket \mathbf{0} \rrbracket$  and sends it to  $\mathcal{A}$ .
4. For generating arithmetic sharing  $\langle T' \rangle^a$ ,  $\mathcal{S}$  emulates  $\mathcal{F}_{E2A}$  by recording the shares of corrupted parties on  $\langle T' \rangle^a$  sent by  $\mathcal{A}$  to  $\mathcal{F}_{E2A}$ .
5.  $\mathcal{S}$  sends the shares of corrupted parties about  $\langle r \rangle^b$  and  $\langle T' \rangle^a$  to functionality  $\mathcal{F}_{\text{Prep-LUT}}$ .

The simulation of  $\mathcal{F}_{E2A}$  is perfect. Under the assumption that the THE scheme is CPA secure and satisfies circuit privacy, the simulation using zero ciphertexts is computationally indistinguishable from the real ciphertexts. In both the ideal-world execution and real-world execution, the output shares of corrupted parties and honest parties satisfy the correct correlation on additive sharings. Thus, the joint distributions of the adversary's view and the honest parties' outputs are computationally indistinguishable in both worlds.  $\square$

### B.3 Proof of Theorem 3

**Theorem 9** (Theorem 3, restated). *Protocol  $\Pi_{\text{Lookup}}$  (shown in Figure 8) securely realizes functionality  $\mathcal{F}_{\text{LUT}}$  in the presence of semi-honest adversaries in the  $\mathcal{F}_{\text{Prep-LUT}}$ -hybrid model.*

*Proof.* We construct a PPT simulator  $\mathcal{S}$  given access to functionality  $\mathcal{F}_{\text{LUT}}$  that runs the PPT adversary  $\mathcal{A}$  as a subroutine, and emulates functionality  $\mathcal{F}_{\text{Prep-LUT}}$ . Specifically, the simulation is constructed as follows:

1.  $\mathcal{S}$  emulates  $\mathcal{F}_{\text{Prep-LUT}}$  by recording the shares of corrupted parties w.r.t.  $\langle r \rangle^b$  and  $\langle T' \rangle^a$  received from  $\mathcal{A}$ .
2. Given the shares of corrupted parties on input sharings  $\langle x \rangle^b$ ,  $\mathcal{S}$  computes the shares of corrupted parties w.r.t.  $\langle u \rangle^b = \langle x \rangle^b \oplus \langle r \rangle^b$ .
3.  $\mathcal{S}$  samples  $u \leftarrow \{0, 1\}^m$ , and then samples the shares of honest parties uniformly such that the sum of the shares of all parties is equal to  $u$ . Then,  $\mathcal{S}$  sends the shares of honest parties to  $\mathcal{A}$  during the  $\text{Open}(\langle u \rangle^b)$  procedure.
4.  $\mathcal{S}$  sends the shares of corrupted parties about  $\langle T[x] \rangle^a$  to functionality  $\mathcal{F}_{\text{LUT}}$ .

Clearly, the simulation of  $\mathcal{F}_{\text{Prep-LUT}}$  is perfect. In the real protocol execution,  $u$  is uniform in  $\{0, 1\}^m$  due to the uniformity of  $r$ . Therefore, the simulation of string  $u$  is also perfect. In both worlds, the output shares of corrupted parties and honest parties satisfy the correct correlation about arithmetic sharings. Overall, the joint distributions of the adversary's view and the honest parties' outputs are perfectly indistinguishable in both worlds.  $\square$

### B.4 Proof of Theorem 4

**Theorem 10** (Theorem 4, restated). *Protocol  $\Pi_{\text{B2A}}$  (shown in Figure 9) securely realizes the B2A command of functionality  $\mathcal{F}_{\text{Conv}}$  against semi-honest adversaries in the  $\mathcal{F}_{\text{LUT}}$ -hybrid model.*

*Proof.* The simulation is fairly straightforward by simply emulating the functionality  $\mathcal{F}_{\text{LUT}}$ . In particular, a PPT simulator  $\mathcal{S}$  emulates  $\mathcal{F}_{\text{LUT}}$  by recording the shares of corrupted parties on  $\langle T[\mathbf{x}[j]] \rangle^a$  for  $j \in [0, \ell - 1]$  sent by a PPT adversary  $\mathcal{A}$  to  $\mathcal{F}_{\text{LUT}}$ . Then,  $\mathcal{S}$  computes the corrupted parties' shares on  $\langle x \rangle^a$  following the protocol specification, and sends them to functionality  $\mathcal{F}_{\text{Conv}}$ . It is clear that the simulation is perfect, and  $\mathcal{A}$  sees nothing from the protocol. In both worlds, the output shares of all parties satisfy the correct correlation, which completes the proof.  $\square$

### B.5 Proof of Theorem 5

**Theorem 11** (Theorem 5, restated). *Protocol  $\Pi_{\text{A2B}}$  (shown in Figure 10) securely realizes the A2B command of functionality  $\mathcal{F}_{\text{Conv}}$  against semi-honest adversaries in the  $\mathcal{F}_{\text{LUT}}$ -hybrid model.*

*Proof.* We construct a PPT simulator  $\mathcal{S}$  given access to functionality  $\mathcal{F}_{\text{Conv}}$  that runs the PPT adversary  $\mathcal{A}$  as a subroutine, and emulates functionality  $\mathcal{F}_{\text{LUT}}$ . Specifically, the simulation is constructed as follows:

1.  $\mathcal{S}$  emulates  $\mathcal{F}_{\text{LUT}}$  by recording the shares of corrupted parties on Boolean sharings  $\langle r \rangle^b$  received from  $\mathcal{A}$ .
2.  $\mathcal{S}$  invokes the simulator for  $\Pi_{\text{B2A}}$  to simulate the generation of arithmetic sharing  $\langle r \rangle^a$ . During the procedure,  $\mathcal{S}$  records the shares of corrupted parties on  $\langle r \rangle^a$ .

3. If  $(2^\ell - p)/2^\ell > 1/2^\rho$ , then  $\mathcal{S}$  emulates  $\mathcal{F}_{\text{LUT}}$  by sampling  $\mathbf{r}' \leftarrow \{0, 1\}^\ell$  and outputting the bit that indicates if  $\sum_{j \in [0, \ell-1]} 2^j \cdot \mathbf{r}'[j] < p$  to  $\mathcal{A}$ . If  $\sum_{j \in [0, \ell-1]} 2^j \cdot \mathbf{r}'[j] \geq p$ ,  $\mathcal{S}$  restarts the protocol simulation.
4. Given the shares of corrupted parties on  $\langle x \rangle^a$ ,  $\mathcal{S}$  computes the corrupted parties' shares about  $\langle u \rangle^a = \langle x \rangle^a - \langle r \rangle^a$ .  $\mathcal{S}$  samples  $u \leftarrow \mathbb{Z}_p$  and the shares of honest parties uniformly such that the shares of all parties sum to  $u$ . Then,  $\mathcal{S}$  sends  $u$  to  $\mathcal{A}$ .
5.  $\mathcal{S}$  emulates  $\mathcal{F}_{\text{LUT}}$  by recording the shares of corrupted parties about  $\langle x \rangle^b$  sent by  $\mathcal{A}$  to  $\mathcal{F}_{\text{LUT}}$ . Then,  $\mathcal{S}$  sends these shares to functionality  $\mathcal{F}_{\text{Conv}}$ .

It is easy to see that the simulation of  $\mathcal{F}_{\text{LUT}}$  is perfect. Following the proof of Theorem 4, the simulation of sub-protocol  $\Pi_{\text{B2A}}$  is also perfect. If  $(2^\ell - p)/2^\ell > 1/2^\rho$ , through the “rejection-sampling” procedure, we guarantee that  $r = \sum_{j \in [0, \ell-1]} 2^j \cdot \mathbf{r}[j]$  is uniform in  $\mathbb{Z}_p$ . Otherwise, the distribution of  $r$  is identical to the uniform distribution in  $\mathbb{Z}_p$ , except with probability at most  $1/2^\rho$ . For the simulation checking if  $r < p$ ,  $\mathcal{S}$  samples a random string  $\mathbf{r}'$  that has the same distribution as the real string  $\mathbf{r}$ , and simulates the output bit by deciding if  $\sum_{j \in [0, \ell-1]} 2^j \cdot \mathbf{r}'[j] < p$ . Clearly, this is perfectly indistinguishable from the real protocol execution. From the uniformity of  $r \in \mathbb{Z}_p$ , we have that  $u$  is uniform in  $\mathbb{Z}_p$  except with probability at most  $1/2^\rho$  in the real protocol execution. Therefore, the simulation of  $u$  is statistically indistinguishable from the real protocol execution. In both the real-world execution and ideal-world execution, the output shares of all parties satisfy the correct correlation about arithmetic sharings. In conclusion, the joint distributions of the adversary's view and the honest parties' outputs are statistically indistinguishable in both worlds.  $\square$

## B.6 Proof of Theorem 6

**Theorem 12** (Theorem 6, restated). *Protocol  $\Pi_{\text{MPGC}}$  (shown in Figures 11 and 12) securely realizes functionality  $\mathcal{F}_{\text{MPC}}$  against semi-honest adversaries in the  $(\mathcal{F}_{\text{Prep-LUT}}, \mathcal{F}_{\text{LUT}})$ -hybrid model, assuming that the key-homomorphic AHE scheme is CPA secure.*

*Proof.* We construct a PPT simulator  $\mathcal{S}$  given access to functionality  $\mathcal{F}_{\text{MPC}}$  that runs a PPT adversary  $\mathcal{A}$  as a subroutine, and emulates functionalities  $\mathcal{F}_{\text{Prep-LUT}}$  and  $\mathcal{F}_{\text{LUT}}$ . Specifically, the simulator is constructed as follows:

1. For each wire  $w$ ,  $\mathcal{S}$  emulates  $\mathcal{F}_{\text{LUT}}$  by recording the shares of corrupted parties on  $\langle \lambda_w \rangle^b$  sent by  $\mathcal{A}$  to  $\mathcal{F}_{\text{LUT}}$ .
2. For each AND gate  $g$  with input wires  $u, v$ ,  $\mathcal{S}$  emulates  $\mathcal{F}_{\text{LUT}}$  by recording the shares of corrupted parties about  $\langle \lambda_u \lambda_v \rangle^b$  received from  $\mathcal{A}$ . Following the protocol specification, for each gate  $g$  with output wire  $w$ ,  $\mathcal{S}$  computes the corrupted parties' shares on  $\langle e_{w, \alpha, \beta} \rangle^b$  for each  $\alpha, \beta \in \{0, 1\}$ .
3. For the output wire  $w$  of each gate  $g$ ,  $\mathcal{S}$  simulates the generation of  $\langle \mathbf{k}_{w, e_{w, \alpha, \beta}} \rangle^a$  for each  $\alpha, \beta \in \{0, 1\}$ .
  - If  $g$  is an AND gate,  $\mathcal{S}$  emulates  $\mathcal{F}_{\text{LUT}}$  by receiving the corrupted parties' shares on  $\langle \mathbf{k}_{w, e_{w, \alpha, \beta}} \rangle^a$  from  $\mathcal{A}$ .
  - If  $g$  is a XOR gate,  $\mathcal{S}$  emulates  $\mathcal{F}_{\text{Prep-LUT}}$  by recording the shares of corrupted parties on  $\langle T' \rangle^a$  and  $\langle r \rangle^b$  sent by  $\mathcal{A}$  to  $\mathcal{F}_{\text{Prep-LUT}}$ . Then,  $\mathcal{S}$  computes the corrupted parties' shares about  $\langle u \rangle^b = \langle e_{w, 0, 0} \rangle^b \oplus \langle r \rangle^b$ .  $\mathcal{S}$  samples  $u \leftarrow \{0, 1\}$  and the shares of honest parties such that the shares of all parties are sum to  $u$ . For  $\text{Open}(\langle u \rangle^b)$ ,  $\mathcal{S}$  sends the shares of honest parties on

$\langle u \rangle^b$  to  $\mathcal{A}$ . Following the protocol specification,  $\mathcal{S}$  computes the shares of corrupted parties on  $\langle \mathbf{k}_{w,e_w,\alpha,\beta} \rangle^a$ .

4. For generating garbled rows in the step 4,  $\mathcal{S}$  simulates as follows:
  - (a) For each wire  $w$ ,  $\mathcal{S}$  samples a masked value  $e_w \leftarrow \{0, 1\}$  at random. Thus, for each gate  $g$  with input wires  $u, v$ ,  $\mathcal{S}$  knows masked values  $e_u, e_v \in \{0, 1\}$ .
  - (b)  $\mathcal{S}$  can obtain the “active path” indicating which garbled rows  $\mathcal{A}$  can decrypt. That is, for each gate  $g$  with input wires  $u, v$ ,  $\mathcal{A}$  can only decrypt the garbled row indexed by  $(g, e_u, e_v)$ .
  - (c) For each wire  $w$ , following the protocol specification,  $\mathcal{S}$  generates  $\mathbf{k}_{w,e_w}^i \leftarrow \text{SecKeyGen}(pp)$  for  $i \in \mathcal{H}$ .
  - (d) For each gate  $g$  with input wires  $u, v$  and output wire  $w$ , for each  $\alpha, \beta \in \{0, 1\}$  and  $i \in \mathcal{H}$ ,  $\mathcal{S}$  sets  $\llbracket \mathbf{k}_{w,e_w,\alpha,\beta}^i \rrbracket = \llbracket \mathbf{0} \rrbracket$  if  $(\alpha, \beta) \neq (e_u, e_v)$  and computes  $\llbracket \mathbf{k}_{w,e_w,\alpha,\beta}^i \rrbracket \leftarrow \text{Enc}_{\mathbf{k}_{u,e_u}^i + \mathbf{k}_{v,e_v}^i}((g, e_u, e_v), \mathbf{k}_{w,e_w}^i)$ , where  $\llbracket \mathbf{0} \rrbracket$  is a fresh zero ciphertext. Then, for  $i \in \mathcal{H}$ ,  $\mathcal{S}$  sends  $\llbracket \mathbf{k}_{w,e_w,\alpha,\beta}^i \rrbracket$  for each  $\alpha, \beta \in \{0, 1\}$  to  $\mathcal{A}$ .
  - (e) If  $P_1$  is honest,  $\mathcal{S}$  receives the AHE ciphertexts from  $\mathcal{A}$  for each corrupted party  $P_i \in \mathcal{M}$ .
5. Through the above step,  $\mathcal{S}$  simulates a garbled circuit  $\mathcal{GC}$ . In the online phase,  $\mathcal{S}$  holds the corrupted parties’ shares on all input bits.
6. For each circuit-input wire  $w \in \mathcal{I}$ ,  $\mathcal{S}$  computes the shares of corrupted parties about  $\langle e_w \rangle^b = \langle x_w \rangle^b \oplus \langle \lambda_w \rangle^b$ , and samples the honest parties’ shares on  $\langle e_w \rangle^b$  such that the shares of all parties sum to  $e_w$  chosen by itself.  $\mathcal{S}$  simulates the  $\text{Open}(\langle e_w \rangle^b)$  procedure by sending the shares of honest parties to  $\mathcal{A}$ .
7. For each circuit-input wire  $w \in \mathcal{I}$ , on behalf of every honest party  $P_i$ ,  $\mathcal{S}$  sends  $\mathbf{k}_{w,e_w}^i$  to  $\mathcal{A}$ . If  $P_1$  is honest,  $\mathcal{S}$  also receives the keys of corrupted parties from  $\mathcal{A}$ .
8. For each circuit-output wire  $w \in \mathcal{O}$ , for each corrupted party  $P_i \in \mathcal{M}$ ,  $\mathcal{S}$  sets its share on  $\langle y_w \rangle^b$  as  $\langle \lambda_w \rangle_i^b$  if  $i \neq 1$ , or computes its share on  $\langle y_w \rangle^b$  by  $e_w \oplus \langle \lambda_w \rangle_1^b$  otherwise. Then,  $\mathcal{S}$  sends the shares of corrupted parties on  $\langle y_w \rangle^b$  to functionality  $\mathcal{F}_{\text{MPC}}$ .

It is clear that the simulation of  $\mathcal{F}_{\text{LUT}}$  and  $\mathcal{F}_{\text{Prep-LUT}}$  is perfect. For the output wire  $w$  of each XOR gate, a bit  $u = e_{w,0,0} \oplus r$  is opened in the real protocol execution. Since  $r$  is a uniform bit,  $u \in \{0, 1\}$  is random. In the ideal-world execution,  $\mathcal{S}$  directly samples a random bit  $u$  and opens it by sending the shares of honest parties. Therefore, the distribution of  $u$  is the same in both worlds. Furthermore, the shares of honest parties on  $\langle u \rangle^b$  are uniform under the condition that the sum of all shares is equal to  $u$  in both worlds. Hence, these shares sent by  $\mathcal{S}$  during the  $\text{Open}(\langle u \rangle^b)$  procedure have the identical distribution in both worlds.

As for the simulation of garbled circuits, we first consider the case that  $P_1$  is corrupted. The evaluator  $P_1$  knows all the keys and masked values in the active path, and thus is able to decrypt all the AHE ciphertexts received from every honest party  $P_i$  in the active path in the real protocol execution. In the ideal-world execution, these AHE ciphertexts sent by every honest party  $P_i$  are generated honestly by  $\mathcal{S}$  following the protocol specification. Therefore, the AHE ciphertexts of every honest party in the active path have the identical distribution in both worlds. Differently, the AHE ciphertexts sent by every honest party outside the active path encrypt the corresponding keys in the real protocol execution, while these ciphertexts encrypt the zero vector in the ideal-world execution. We can bound the difference in two worlds by a hybrid argument based on the assumption that the private-key AHE scheme is CPA secure. Specifically, for each honest party  $P_i$ ,

we prove that the real AHE ciphertexts are computationally indistinguishable from the ciphertexts on  $\mathbf{0}$  under the assumption that the private-key AHE scheme is CPA secure. Let  $\ell$  be the number of all gates in the circuit and  $G_0$  be the real protocol execution. For each  $j \in [1, \ell]$ , we define  $G_j$  which is the same as  $G_{j-1}$  except that for the  $j$ -th gate, the four corresponding real AHE ciphertexts sent by honest party  $P_i$  are replaced with four fresh ciphertexts on  $\mathbf{0}$ . For each  $j \in [1, \ell]$ , we bound the difference between  $G_{j-1}$  and  $G_j$  assuming the CPA security of the AHE scheme. In particular, we can replace each AHE ciphertext w.r.t. the  $j$ -th gate sent by honest party  $P_i$  in  $G_{j-1}$  with a fresh ciphertext on  $\mathbf{0}$ . Adversary  $\mathcal{A}$  cannot distinguish each replaced ciphertext from a real ciphertext, which is directly guaranteed by the CPA security of the private-key AHE scheme. Note that no two ciphertexts are created with the same secret key and label  $(g, \alpha, \beta)$ . Overall, the real AHE ciphertexts are computationally indistinguishable from the ciphertexts on  $\mathbf{0}$ . If  $P_1$  is honest, the analysis is the same, where  $\mathcal{A}$  learns less information in this case.

In the online phase of the real protocol execution, the opened bit  $e_w$  for each circuit-input wire  $w$  is the XOR of the real input bit  $x_w$  and random mask  $\lambda_w$ . We first consider the case that  $P_1$  is corrupted. From the above analysis w.r.t. AHE ciphertexts used to generate a garbled circuit, we have that  $e_{w,\alpha,\beta}$  for the output wire  $w$  of each gate and  $\alpha, \beta \in \{0, 1\}$  except for the opened bit  $e_w = e_{w,e_u,e_v}$  are kept secret under the assumption that the private-key AHE scheme is CPA secure. That is, for each wire  $w$ ,  $\mathcal{A}$  only learns  $e_w$ . Therefore, in the real protocol execution,  $\lambda_w$  for each wire  $w$  is a uniform bit, which guarantees that  $e_w$  is random in  $\{0, 1\}$ . In the ideal-world execution,  $e_w \in \{0, 1\}$  is sampled at random by  $\mathcal{S}$ . Thus, the distribution of  $e_w$  for each circuit-input wire  $w$  is computationally indistinguishable in both worlds. Furthermore, the keys sent in the online phase have the same distribution in both worlds. If  $P_1$  is honest, then  $\mathcal{A}$  learns less information, e.g.,  $e_w$  for each wire  $w$  that is not in  $\mathcal{I}$  is kept unknown for  $\mathcal{A}$ . In this case, the distributions of  $e_w$  and the keys are also computationally indistinguishable in both worlds. In both the real-world execution and ideal-world execution, the shares of all parties on circuit-output bits satisfy the correct correlation of Boolean sharings. In conclusion, the joint distributions of the adversary's view and the honest parties' outputs are computationally indistinguishable in both worlds.  $\square$