

# A Note on the Communication Complexity of Multiparty Computation in the Correlated Randomness Model

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**Abstract.** Secure multiparty computation (MPC) addresses the challenge of evaluating functions on secret inputs without compromising their privacy. A central question in multiparty communication is to understand the amount of communication needed to securely evaluate a circuit of size  $s$ . In this work, we revisit this fundamental question, in the setting of information-theoretically secure MPC in the correlated randomness model, where a trusted dealer distributes correlated random coins, independent of the inputs, to all parties before the start of the protocol. This setting is of strong theoretical interest, and has led to the most practically efficient MPC protocols known to date.

While it is known that protocols with optimal communication (proportional to input plus output size) can be obtained from the LWE assumption, and that protocols with sublinear communication  $o(s)$  can be obtained from the DDH assumption, the question of constructing protocols with  $o(s)$  communication remains wide open for the important case of information-theoretic MPC in the correlated randomness model. All known protocols in this model require  $O(s)$  communication in the online phase; improving this state of affairs is a long standing open question.

In this work, we exhibit the first generic multiparty computation protocol in the correlated randomness model with communication sublinear in the circuit size, for a large class of circuits. More precisely, we show the following: any size- $s$  *layered* circuit (whose nodes can be partitioned into layers so that any edge connects adjacent layers) can be evaluated with  $O(s/\log \log s)$  communication. Our result holds for both boolean and arithmetic circuits, and security holds with respect to malicious corruption, without honest majority.

**Keywords.** multiparty computation, correlated randomness model, information-theoretic security, sublinear communication

## 1 Introduction

Secure multiparty computation (MPC) allows  $n$  players with inputs  $(x_1, \dots, x_n)$  to jointly evaluate a function  $f$ , while leaking no information on their own input

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\*\* Part of the work was done when the author was at ENS Paris, France.

beyond the output of the function. It is a fundamental problem in cryptography, which has received a considerable attention since its introduction in the seminal works of Yao [43], and Goldwasser, Micali, and Wigderson [28, 29] (GMW). One of the core questions in secure multiparty computation is to understand the amount of communication needed to securely compute a function. For almost three decades after the protocols of Yao and GMW, all known constructions of secure computation protocols required a communication proportional to the circuit size of the function, and understanding whether this was inherent was a major open problem.

**Secure Computation with Sublinear Communication.** In 2009, this situation changed with the introduction by Gentry of the first fully-homomorphic encryption scheme [27] (FHE), which led to secure computation protocols with communication independent of the size of the function (proportional only to its input size and its output size), under the LWE assumption. This resolved the long standing open problem of designing MPC protocols with optimal (asymptotic) communication, although only under a specific assumption. More recently, the circuit-size barrier was broken again under the DDH assumption in [13], for a large class of structured circuits<sup>1</sup> and in the two-party case. However, while these results are of strong theoretical interest, they require expensive computations.

**Secure Computation in the Correlated Randomness Model.** While secure computation (with no honest majority) is known to require computational assumptions, it was observed in several works (e.g. [21, 31]) that executing a pre-computation phase independent of the inputs to the protocol, during which correlated random coins are distributed to the parties, allows to make the online phase both information-theoretically secure and significantly more efficient, by removing any expensive cryptographic operation from the online computation phase. These observations led to the development of increasingly efficient secure computation protocols in the correlated randomness model, e.g. [19, 35], which are currently considered the most practical secure computation protocols. Yet, unlike computationally secure protocols, all known unconditionally secure protocols in the correlated randomness model require communication proportional to the circuit size of the function. Therefore, the major question of understanding the communication required for multiparty computation remains wide open for the important case of MPC in the correlated randomness model, which captures the best candidates for practical secure computation. This is the question we address in this work: must MPC protocols in the correlated randomness model inherently use a communication linear in the size of the circuit? Or, in other words, can we get the best of both worlds: unconditional security with high practical efficiency, and sublinear communication?

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<sup>1</sup> The work of [13] considered, as we will do in this work, boolean circuits which can be divided into layers such as any edge connects adjacent layers. Such circuits are called *layered boolean circuits*.

**On the Communication of Secure Computation in the Correlated Randomness Model.** A partial answer to this question was given in [30], where the authors designed a *one-time truth-table* protocol, which allows to evaluate any function  $f : \{0, 1\}^n \mapsto \{0, 1\}^m$  with unconditional security in the correlated randomness model, with optimal communication  $O(n + m)$ . However, this protocol requires storing an exponential number (in  $n$ ) of correlated random coins (polynomial in the size of the entire truth-table of  $f$ ), which makes it practical only for boolean functions with very small inputs. Furthermore, it was argued in [30] that reducing the amount of correlated random coins from exponential to polynomial for any function  $f$  is unlikely to be feasible, as it would imply an unexpected breakthrough for long-standing open problems related to private information retrieval.

This negative result, as well as the fact that all known protocols (with polynomial storage) have communication proportional to the circuit size  $s$  of the function, have been seen as indications that breaking the circuit-size barrier for multiparty computation in the correlated randomness model might be a hard problem. For instance, it was mentioned in [23] that “the results and evidence we know suggest that getting constant overhead [over the circuit size of the function] is the goal we can realistically hope to achieve”. More recently in [20], it was observed that “while information-theoretic secure protocols are very efficient in terms of computational work, they (seem to) require more communication and more rounds than computationally secure protocols. Whether this is inherent is an open and probably very hard problem”. Later in the same paper, the authors explicitly consider the issue of getting sublinear communication in the circuit size: “whether we can have constant round protocols and/or communication complexity much smaller than the size of the circuit and still be efficient (polynomial-time) in the circuit size of the function is a long-standing open problem”.

In [20], the authors made progress toward understanding why existing protocols have been stuck at the circuit-size barrier, by identifying a property shared by all known efficient protocols in the correlated randomness model, which states (informally) that they evaluate the function in a “gate-by-gate” fashion, and require communication for every multiplication gate. They demonstrated that all protocols following this approach (with passive security and dishonest majority) must inherently have communication proportional to the circuit size of the function. They concluded that improving the communication complexity of secure computation in the correlated randomness model requires a fundamentally new approach, and mentioned that the main question left open in their work is to find out whether their bound does hold for any protocol which is efficient in the circuit size of the function. This is the problem we address in this work.

## 1.1 Our Contribution

In this paper, we construct for the first time protocols with polynomial storage and communication sublinear in the circuit size, for a large class of circuit. Perhaps surprisingly, our results turn out to be relatively simple to obtain.

**Sublinear Protocol for Structured Circuits.** We exhibit a generic secure computation protocol in the correlated randomness model, with communication sublinear in the circuit size. More specifically, we consider *layered boolean circuits* (LBC), whose nodes can be arranged into layers so that any edge connects adjacent layers. We prove the following: for any  $N$ , there is an unconditionally secure  $N$ -party protocol that evaluates an arbitrary LBC of size  $s$  with  $n$  inputs and  $m$  outputs, with total communication

$$O\left(n + N \cdot \left(m + \frac{s}{\log \log s}\right)\right),$$

sublinear in the size of the circuit, and polynomial storage  $O(s^2/\log \log s)$ , in the correlated randomness model against semi-honest adversaries, with dishonest majority. While this requires an arguably large storage, it can be reduced to being only slightly superlinear in  $s$ , namely

$$O\left(s \cdot \frac{2^{(\log s)^{1/c}}}{\log \log s}\right),$$

at the cost of increasing the communication to  $O(n + N \cdot (m + c \cdot s/\log \log s))$  (for an arbitrary  $c = o(\log \log n)$ ). Our protocol enjoys perfect security, computational complexity  $O(s \log s/\log \log s + n + m)$ , and round complexity  $d/\log \log s$ , where  $d$  is the depth of the circuit. All the constants involved are very small (in fact equal to one, up to low order terms), and the computation involves solely searching lookup tables.

**Extensions.** We generalize our result to secure evaluation of arbitrary *layered arithmetic circuits* (LAC) over any (possibly exponentially large) field  $\mathbb{F}$ , by relying on a connection between MPC with correlated randomness and the classical notion of private simultaneous message protocols [25]. The resulting protocol for arithmetic circuits has costs comparable to the boolean version. Furthermore, we show that all our results can be extended to the stronger function-independent preprocessing model, where only a bound on the size of the circuit is known in the preprocessing phase, and that the communication can be improved for “tall and narrow” circuits. Eventually, using the techniques of [19, 34], our protocols directly extend to the malicious setting, at an additive cost of  $N \cdot \kappa$  bits of communication (for some statistical security parameter  $\kappa$ ), and a  $O(\kappa)$  overhead in computation and correlated randomness (more advanced techniques from [19] can be used to make this overhead constant).

**Static vs Adaptive Setting.** While we focus for simplicity on the static setting in this work, where the adversary decides before the protocol which parties to corrupt, our protocols can be proven to also satisfy adaptive security in a relatively straightforward way. Indeed, when it must reveal the input of a party which is being corrupted by the adversary, the simulator of our main protocol (and its

variants) can easily explain the view of the adversary as being consistent with any input of its choice, by choosing the preprocessing material in an appropriate way. As the view of the adversary will always consist of values perfectly masked by random coins generated in the preprocessing phase, there will always be a choice of preprocessing material which “unmask” the values known to the adversary to any value chosen by the simulator.

## 1.2 Our Method

Perhaps surprisingly, our method does not depart significantly from existing techniques in secure computation. Our starting point is the one-time truth-table (OTTT) protocol of [30], which has optimal communication but requires an exponential amount of data. It has been observed in several works that using OTTT as an internal component in secure protocols can be used to reduce their communication. For example, it was suggested to use OTTT to securely compute S-boxes in AES in [19, 34], as they can be efficiently represented as small lookup-tables. More recently, the work of [24] developed methods to automatically create tradeoffs between communication and computation in secure protocols, by relying on a compiler that transforms high-level descriptions of a function into a lookup-table-based representation of the function. All these works rely on the fact that, for functions that can be broken into small interconnected lookup-tables, the protocol of [30] can be used to save some communication.

**Dividing Layered Boolean Circuits into Local Functions.** In this work, we show that this intuition can in fact be extended to *arbitrary* layered boolean circuit of size  $s$ , and that the savings obtained this way lead to a protocol with  $o(s)$  communication. Our protocol builds upon a variant of the result of [30], which states that every function can be securely evaluated in the correlated randomness model with perfect security, optimal communication, and exponential storage. Our variant relies on the observation that when evaluating *local functions*, where each output bit depends on a number  $c$  of input bits, we can reduce the storage cost of the protocol of [30] from being exponential in the input size to being only exponential in the locality parameter  $c$ . Indeed, consider the task of securely evaluating a function with  $n$  input bits, and  $m$  output bits. The protocol of [30] (called OTTT, for one-time truth table) requires the parties to store shares of (a shifted version of) the truth table of the function, which has size  $m \cdot 2^n$ , exponential in the input size. When the function is  $c$ -local, however, there is a better solution: the parties can store shares of (shifted variants of) truth tables corresponding to each function mapping  $c$  input bits to a given output bit, for a total storage cost of  $m \cdot 2^c$ . Some care must be taken, as doing straightforward parallel repetitions of the OTTT protocol for each subfunction would increase the communication from  $O(n)$  to  $O(c \cdot m)$ ; we show that carefully avoiding redundancies in the secret-shared representation of the input allows to bring this cost back to  $O(n)$ . We formally state this result in a lemma, which we call *core lemma*.

Given the core lemma, our result is obtained by breaking an arbitrary layered circuit into chunks, each chunk containing some number  $k$  of consecutive layers. We observe that, as the underlying directed graph of the circuit has indegree 2, each value associated to the last layer of a chunk can be computed as a function of at most  $2^k$  values on the last layer of the previous chunk. Therefore, computing all the values on the last layer of a chunk can be reduced to evaluating a  $2^k$ -local function of the values on the last layer of the previous chunk. Using the core lemma, this can be done using  $O(w \cdot 2^{2^k})$  bits of preprocessing material, where  $w$  is the width of the input layer, with a communication proportional to  $w$  only. If the circuit has size  $s$ , width  $w$ , and depth  $d$ , this means that the circuit can be securely evaluated in a chunk-by-chunk fashion, with total communication  $O((d/k) \cdot w) = O(s/k)$ , using  $O((d/k) \cdot 2^{2^k})$  bits of correlated randomness; setting  $k \leftarrow \log \log s$  gives the claimed result.

**Extending the Result to Arithmetic Circuit.** The above method breaks down in the case of arithmetic circuits over large order fields. While we can decompose an arbitrary LBC into polynomial-size truth tables (by breaking it into interconnected functions operating on logarithmically many inputs), this is not true anymore for arithmetic circuit over fields of exponential size, where even a very simple function with a single input will have in general an exponential-size truth table. We nevertheless obtain a comparable result for arithmetic circuit, building upon a relation with the notion of private simultaneous message (PSM) protocols [25], which establishes that PSM protocols with some additional decomposability property can be used to build two-party secure computation protocols in the correlated randomness model. This link was indirectly established in [7], where a connection was drawn both between PSM and PIR, and between MPC in the correlated randomness and PIR. Building upon a recent PSM protocol of [39] for multivariate polynomial evaluation, we get an arithmetic analogue of the protocol of [30], which relies on the representation of arithmetic functions as multivariate polynomials. From this protocol, we derive a new version of our core lemma, tailored to the arithmetic setting, which directly leads to a secure computation protocol with communication  $O(s/\log \log s)$  for layered arithmetic circuits over arbitrary fields. We note that, while lookup-table-based secure computation protocols for boolean circuits have been investigated, the extension of this approach to the arithmetic setting was (to our knowledge) never observed before.

**On the Lower Bounds of [20, 30].** It should be noted that our protocols do not follow the standard gate-by-gate design of unconditionally secure protocols in the correlated randomness model, hence our result does not contradict the lower bound of [20]. Moreover, our results apply only to circuits, while the implausibility result of [30] holds for low-storage protocols for evaluating any function. Therefore, they do not lead to unexpected breakthroughs for information-theoretic private information retrieval.

### 1.3 Related Work

The possibility of securely computing functions given access to a source of correlated random coins was first studied in the work of Beaver for the (MPC-complete) oblivious-transfer functionality in [4], and later generalized to the *commodity-based* model, where multiple servers generate correlated random coins in a honest majority setting in [5]. The study of multiparty computation in the *preprocessing model*, where the correlated-randomness coin-generation phase is implemented with a computationally secure MPC protocol, was initiated in [3, 31, 36]. These works started a rich line of work on increasingly efficient MPC protocols in the preprocessing model [9, 15, 18, 19, 21–23, 26, 32, 35, 38, 41].

The quest for secure multiparty computation protocols with low-communication was initiated in [6], which gave a protocol with optimal communication, albeit with exponential computation and only for a number of party linear in the input size. An optimal communication protocol with exponential complexity was also given in [40]. The work of [2] gives a low-communication protocol for constant-depth circuit, for a number of parties polylogarithmic in the circuit size. The breakthrough result of Gentry [27] led to optimal communication protocols in the computational setting [1, 17] under the LWE assumption. More recently, computationally secure MPC protocols with sublinear communication were achieved from the DDH assumption in [13].

The study of low-communication protocols in the correlated randomness model was initiated in [30], where a protocol with optimal communication and exponential storage complexity was presented. The same paper showed that improving the storage requirement for all functions would imply a breakthrough in information-theoretic PIR. The work of [7] reduce the storage requirement for functions with  $n$  inputs to  $2^{O(\sqrt{n})}$ , at the cost of increasing the communication complexity to  $2^{O(\sqrt{n})}$ . The work of [8] leads to low-communication protocols in the correlated randomness model for the special case of depth-2 circuits with a layer of OR gates and a layer of gates computing a sum modulo  $m$ , for composite  $m$ . All known protocols for evaluating arbitrary circuits in the correlated randomness model (with polynomial computation and storage) use communication linear in the circuit size. This limitation was formally studied recently in [20], where it was shown that it is inherent in the setting of gate-by-gate protocols.

The idea of using truth-table representation to reduce the communication of secure computation protocols first arose in [16], and was developped in [30]. It was later used implicitly in [37], to construct one-out-of-two oblivious transfer for short string from one-out-of- $N$  oblivious transfer, and in the works of [19, 24, 34] to evaluate circuits with an appropriate structure.

**On the Relation to [19].** At a late stage of our work on this paper, it was brought to our attention that the main techniques underlying the proof of our core lemma – informally, breaking a function into interconnected truth-tables, representing each outgoing wires from a table with secret-shared values, and carefully avoiding all redundancies for wires which are used by several tables – are already implicitly present in [19]. Indeed, [19] already explored the possibility of

breaking a circuit into small interconnected truth table, avoiding redundancies in the secret-shared representation of the values associated to each wire, and envisioned the possibility of generalizing this to larger tables. However, it appears that the authors of [19] have overlooked the surprising potential consequences of these techniques, which we explore in this paper. Therefore, our work can be seen as indentifying and abstracting out the technical ideas underlying our main result (as well as providing additional contributions, such as the extension to the arithmetic setting), but while the core lemma is new to our work, we cannot (and do not) claim the novelty of the techniques used in its proof, which should be credited to [19]. Still, we believe that our result remains interesting and surprising, and that it deserves to be explicitly presented.

#### 1.4 On the Practical Efficiency of our Protocols

In spite of its theoretical nature, our result can in fact lead to concrete efficiency improvements for secure multiparty computation. We focus for simplicity on the case of two-party computation. The state-of-the-art protocol for secure two-party computation in the correlated randomness model is, to our knowledge, the protocol of [19] (in both the passive setting and the active setting), which also relies on an OTTT-based evaluation of a boolean circuit. In the online phase, the protocol of [19] communicates 2 bits per AND gate (one from each player), and no bit at all for XOR and NOT gates (we note that our protocols can be readily adapted to allow for free XOR and NOT gates as well).

Using our protocol with  $k = 2$ , we get a two-party protocol which communicates on average a *single bit* per AND gate, improving over the protocol of [19] by 50% in both the passive and the active setting, for arbitrary layered circuits. This comes at the cost of storing 8 times more preprocessed data (a factor  $2^{2^k} = 8$  for  $k = 2$ ), and a factor 2 in computation (which comes from the need to search four-times larger lookup-tables). As noted in [20], the limiting factor in a concrete implementation of TinyTable is the bandwidth, hence we expect that an implementation of our protocol would result in concrete improvements over [19] in the speed of the online phase.

#### 1.5 On Implementing the Correlated Randomness Model

It is well known that the distribution of correlated random coins in the preprocessing phase can be implemented by any generic MPC protocol. However, in our setting, generic approaches would require a communication superlinear in the circuit size. We note that, for the specific case of generating random shares of correlated strings, there are better (theoretical) solutions: under the learning with error assumption, or under (variants of) the decisional Diffie-Hellman assumption in the two-party case, the preprocessing phase of our protocols can be implemented with *constant* communication  $\text{poly}(\lambda)$  (where  $\lambda$  is a security parameter), independent of the size of the circuit, resulting in protocols with sublinear total communication  $O(s/\log \log s + \text{poly}(\lambda))$ , and information-theoretically secure online phase.



We briefly sketch how the preprocessing phase can be implemented with constant communication. The main technical tool is a primitive known as *homomorphic secret sharing* [13] (HSS); the idea of using HSS to implement the preprocessing phase of MPC protocols was suggested in [10, 14]. Informally, an HSS scheme for a class of functions  $F$  allows to secretly share an input  $x$  between several parties, such that given its share, each party can *locally* compute an additive share of  $f(x)$ , for any  $f \in F$ . Given an HSS scheme for all circuits, the preprocessing phase can be implemented as follows: let  $f$  be a function that takes as input a large string of random coins  $x$ , such that the correlated random coins distributed by the trusted dealer with randomness  $x$  are uniformly random shares of  $f(x)$  (e.g. in our protocol,  $f$  would output  $\approx s/\log \log s$  shifted truth-tables). The parties jointly and securely construct, using a general purpose MPC protocol, an homomorphic secret sharing of a random PRF key  $K$ . Then, all parties locally evaluate the function  $f'$  that takes some counter  $c$ , generates pseudorandom coins  $x$  from this counter using the PRF with key  $K$  (e.g. by computing  $\text{PRF}(K, c)$ ,  $\text{PRF}(K, c + 1)$ , and so on), and returns  $f(x)$ . This way, with no further communication except for a one-time generation of the sharing of  $K$  (which takes communication  $\text{poly}(\lambda)$ , independently of  $s$ ), the parties obtain correlated (pseudo) random coins. An HSS scheme for all functions (and a PRF) can be constructed from the LWE assumption [12, 33]. With a more involved construction, a protocol can also be obtained from DDH: under the DDH assumption, there exists an approximately-correct HSS scheme for  $\text{NC}_1$  [13], in the two-party setting. Noting that the preprocessing function is parallelizable (in  $\text{NC}_0$ ) and that there exists PRFs in  $\text{NC}_1$  under the DDH assumption, we can implement the previous strategy from DDH. The correlated random coins obtained this way are not all correct, but the approximately-correct HSS scheme of [13] allows the parties to make the error probability arbitrarily small, and to detect when an output is erroneous. By setting the error parameter so that, with overwhelming probability, a small (constant) number of correlated random coins will be erroneous, and by introducing some redundancy in the coins generated this way, the parties can simply reveal to each other which correlated coins are susceptible to be erroneous (indicating the position of erroneous bits only requires  $O(\log s)$  communication), and locally delete them. To prove security in spite of this small leakage, we need to rely on slightly leakage-resilient PRF and HSS, which can both be constructed from DDH-based primitives using standard approaches. We refer the reader to the full version [11] of [10] for a detailed overview of this approach.

## 1.6 Organization

Section 2 introduces our notations, and recalls standard preliminaries on circuits. In section 3, we summarize the contributions of this paper in the form of a list of theorems, formally state the core lemma on which these theorems are based, and prove it. Section 4 builds upon the core lemma; it introduces our main protocol and several variants, and proves its security. In Section 5, we discuss the extension of our protocols to the malicious setting. Eventually, Section 6 lists

some questions left open by our work, that we believe to be of interest for future works.

## 2 Preliminaries

**Notations.** Let  $k$  be an integer. We let  $\{0, 1\}^k$  denote the set of bitstrings of length  $k$ . For two strings  $(x, y)$  in  $\{0, 1\}^k$ , we denote by  $x \oplus y$  their bitwise xor. Given a subset  $S$  of  $[k]$ ,  $x[S]$  denotes the subsequence of the bits of  $x$  with indices from  $S$ . We use bold letters to denote vector; for a vector  $\mathbf{x} = (x_1, \dots, x_N)$ ,  $\mathbf{x}[S]$  denotes the vector  $(x_1[S], \dots, x_N[S])$ . For a matrix  $M$ , we denote  $M|_{i,j}$  its entry  $(i, j)$ .

### 2.1 Circuits

**Boolean Circuits.** A boolean circuit  $C$  with  $n$  inputs and  $m$  outputs is a directed acyclic graph with two types of nodes:

- The *input nodes* are labelled according to variables  $\{x_1, \dots, x_n\}$ ;
- The *gates* are labelled according to a base  $B$  of boolean functions.

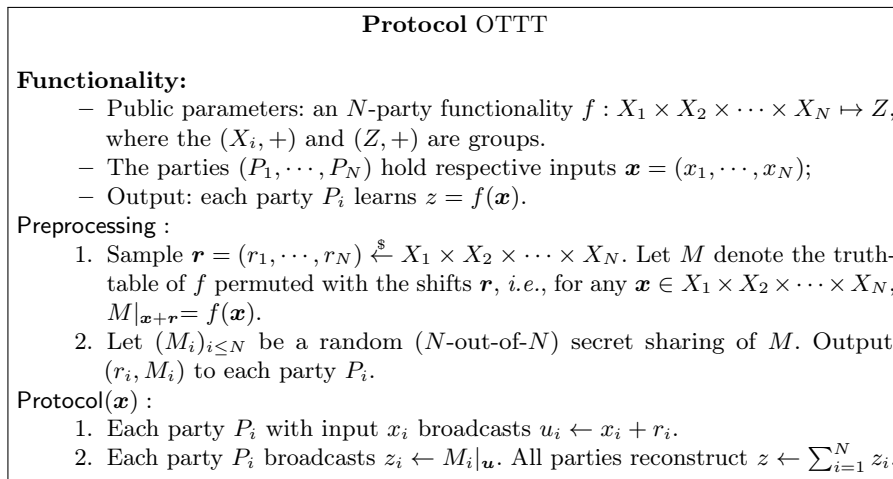
In this work, we will focus on boolean circuits with indegree two (hence,  $B$  contains boolean functions with domain  $\{0, 1\}$  or  $\{0, 1\}^2$ ).  $C$  contains  $m$  gates with no children, which are called *output gates*. If there is a path between two nodes  $(v, v')$ , we say that  $v$  is an *ancestor* of  $v'$ . The *size*  $\text{size}(C)$  of  $C$  is the number of its nodes; its *depth*  $\text{depth}(C)$  is the length of the longest path from an input node to an output gate. The *width* of a circuit  $C = (V, E)$  is defined as  $\text{width}(C) = \max_{1 \leq i \leq \text{depth}(C)} \#\{v \in V \mid (0 \leq \text{depth}(v) \leq i) \wedge (\exists w, (v, w) \in E \wedge \text{depth}(w) > i)\}$ .

**Layered Boolean Circuits.** In this work, we will consider a special type of boolean circuits, called *layered boolean circuits* (LBC). An LBC is a boolean circuit  $C$  whose nodes can be partitioned into  $d = \text{depth}(C)$  layers  $(L_1, \dots, L_d)$ , such that any edge  $(u, v)$  of  $C$  satisfies  $u \in L_i$  and  $v \in L_{i+1}$  for some  $i \leq d - 1$ . Note that the width of a layered boolean circuit is also the maximal number of non-output gates contained in any single layer. Evaluating a circuit  $C$  on input  $x \in \{0, 1\}^n$  is done by assigning the bits of  $x$  to the variables  $\{x_1, \dots, x_n\}$ , and then associating to each gate  $g$  of  $C$  (seen as a boolean function) the bit obtained by evaluating  $g$  on the values associated to its parent nodes. The output of  $C$  on input  $x$ , denoted  $C(x)$ , is the bit-string associated to the output gates.

**Arithmetic Circuits.** We define arithmetic circuits over a field  $\mathbb{F}$  comparably to boolean circuits, as directed acyclic graphs with input nodes and arithmetic gates. Input nodes are labeled with variables  $\{x_1, \dots, x_n\}$  over  $\mathbb{F}$ , and the gates compute negation, addition, or multiplication over  $\mathbb{F}$ . Note that boolean circuits correspond to the special case of arithmetic circuits over the field  $\mathbb{F}_2$ ; we extend layered boolean circuits to layered arithmetic circuits (LAC) in a similar way.

## 2.2 One-Time Truth Tables

We recall the one-time truth-table protocol of [30], which is at the heart of our protocols. It allows multiple parties to jointly evaluate a function  $f : X_1 \times X_2 \times \dots \times X_N \mapsto Z$ , by sharing between all parties a scrambled version of the truth table of  $f$ . We focus for simplicity on a scenario where all parties receive the same output, but the protocol can be trivially generalized to a setting where the parties receive different outputs. The protocol is represented on Figure 1; it has optimal communication  $\sum_i \log|X_i| + N \cdot \log|Z|$ , and exponential storage complexity  $|Z| \cdot \prod_i |X_i|$  per party.



**Fig. 1.** Protocol OTTT for evaluating an arbitrary  $N$ -party functionality  $f$  in the correlated randomness model, against a passively corrupted majority

## 3 Theorems and Core Lemma

In this section, we formally introduce the theorems which we will prove in this work, state the core lemma from which we will derive them, and prove it.

**Network Model.** We consider protocols involving  $N$  parties communicating over synchronous and authenticated broadcast channel. Note that broadcast channels can be unconditionally implemented from (insecure) point-to-point channels in the correlated randomness model.

**Functionalities.** An  $N$ -party functionality  $F : X_1 \times X_2 \times \dots \times X_n \mapsto Z_1 \times Z_2 \times \dots \times Z_N$  specifies a mapping from the  $N$  input of each party to  $N$  outputs (one

for each party). Such functionalities capture arbitrary non-reactive computation tasks. A useful special case of (randomized)  $N$ -party functionalities are *secret sharing functionalities* for functions over an abelian group  $(\mathbb{G}, +)$ : a protocol computes secret shares of a function  $g : \mathbb{G} \mapsto \mathbb{G}$  if it computes the (randomized)  $N$ -party functionality which, on input  $(x_1, \dots, x_N) \in \mathbb{G}^N$ , outputs  $N$  uniformly random group elements  $(z_1, \dots, z_N) \in \mathbb{G}^N$  subject to  $\sum_{i=1}^N z_i = g(\sum_{i=1}^N x_i)$ . This captures the situation where the parties hold secret shares of an input to a (deterministic) function, and want to receive secret shares of the output of the function.

### 3.1 Theorems

Following is a summary of the results that we obtain in the subsequent sections.

**Theorem 1.** *For any  $N$ -party functionality  $f$  represented by a layered (boolean or arithmetic) circuit  $C$  of size  $s$  with  $n$  inputs and  $m$  outputs, and for any integer  $k$ , there is a perfectly secure protocol which realizes  $f$  in the preprocessing model against semi-honest parties, without honest majority, with communication  $n + N \cdot (m + \lceil s/k \rceil)$  and storage  $n/N + (m + \lceil s/k \rceil) \cdot (2^{2^k} + 1)$ .*

In the above theorem, “storage” refers to the number of correlated random coins stored by each party at the end of the preprocessing phase (counted as a number of bits in the boolean case, and as a number of field elements in the arithmetic case). This gives, setting  $k = \log \log s$ ,

**Corollary 2.** *There is a protocol that perfectly realizes any functionality  $f$  (in the function-independent preprocessing model and against semi-honest parties, without honest majority) represented by a layered (boolean or arithmetic) circuit  $C$  of size  $s$  with  $n$  inputs and  $m$  outputs, with communication  $O(n + N \cdot (m + s/\log \log s))$  and polynomial storage.*

Building on the same techniques, we can also obtain a comparable result in the stronger *function-independent* correlated randomness model, where the correlated randomness is not allowed to depend on the target functionality (but is only given a bound on its size):

**Theorem 3.** *For any  $N$ -party functionality  $f$  represented by a layered (boolean or arithmetic) circuit  $C$  of size  $s$  with  $n$  inputs and  $m$  outputs, and for any integer  $k$ , there is a perfectly secure protocol which realizes  $f$  in the function-independent preprocessing model against semi-honest parties, without honest majority, with communication  $n + N \cdot (m + \lceil s/k \rceil)$  and storage  $n/N + (m + \lceil s/k \rceil) \cdot (2^{k+2^{2^k}} + 1)$ .*

Setting  $k = \log \log \log s$  gives us

**Corollary 4.** *There is a protocol that perfectly realizes any functionality  $f$  (in the function-independent preprocessing model and against semi-honest parties, without honest majority) represented by a layered (boolean or arithmetic) circuit  $C$  of size  $s$  with  $n$  inputs and  $m$  outputs, with communication  $O(n + N \cdot (m + s/\log \log \log s))$  and polynomial storage.*

Finally, we can obtain a stronger sublinearity guarantee for “tall and narrow” layered circuits:

**Theorem 5.** *For any  $N$ -party functionality  $f$  represented by a layered (boolean or arithmetic) circuit  $C$  of size  $s$  and width  $w$  with  $n$  inputs and  $m$  outputs, and for any integer  $k$ , there is a perfectly secure protocol which realizes  $f$  in the preprocessing model against semi-honest parties, without honest majority, with communication  $n + N \cdot (m + \lceil s/k \rceil)$  and storage  $n/N + (m + \lceil s/k \rceil) \cdot (2^{w/k} + 1)$ .*

For example, setting  $k = \sqrt{\log s}$  gives us

**Corollary 6.** *There is a protocol that perfectly realizes any functionality  $f$  (in the preprocessing model and against semi-honest parties, without honest majority) represented by a “tall and narrow” layered (boolean or arithmetic) circuit  $C$  of size  $s$  and width  $w = O(\sqrt{\log s})$  with  $n$  inputs and  $m$  outputs, with communication  $O(n + N \cdot (m + s/\sqrt{\log s}))$  and polynomial storage.*

Alternatively, we get a protocol with communication  $O(s/\log s)$  for constant-width circuit (which corresponds to the complexity class  $\text{SC}_0$ ). This can again be generalized to the stronger function-independent correlated randomness model. In the next section, we proceed with the description of our protocol. We first focus on the case of layered boolean circuits, and then discuss our extension to the case of arithmetic circuits.

### 3.2 Core Lemma

In this section, we state and prove the core lemma which underlies our results.

**Definition 7 (Local Function).** *A Function  $g : \mathbb{F}_2^n \mapsto \mathbb{F}_2^m$  is  $c$ -local (for some integer  $c \leq n$ ) if on any input  $x \in \mathbb{F}_2^n$ , any output bit of  $g(x)$  depends on at most  $c$  bits from  $x$ .*

**Lemma 8 (Core Lemma).** *For any  $c$ -local function  $g : \mathbb{F}_2^n \mapsto \mathbb{F}_2^m$ , there is an information-theoretic semi-honest  $N$ -party secure computation protocol (with dishonest majority) in the correlated randomness model for computing secret shares of  $g$  with total online communication  $N \cdot n$  bits, and correlated randomness  $m \cdot 2^c + n$  bits per party.*

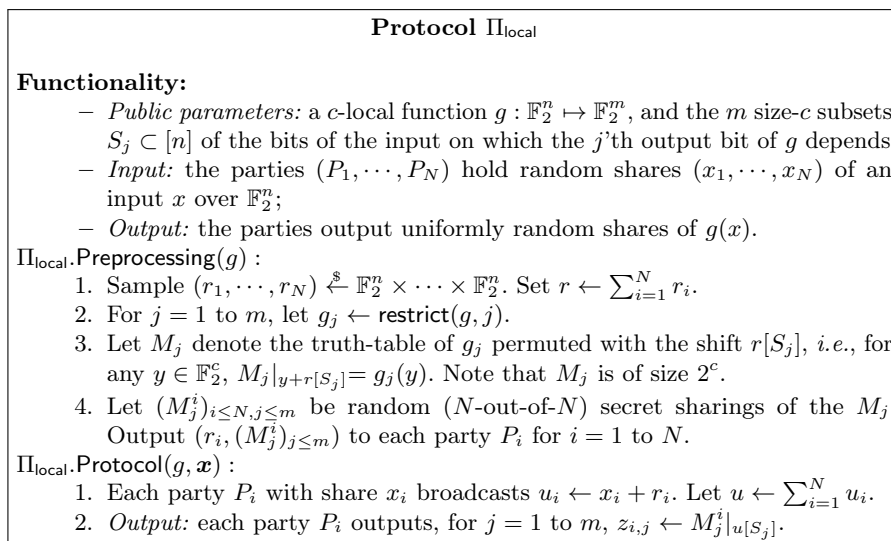
Before proving Lemma 8, it is instructive to compare its guarantees to the protocol obtained by applying directly the one-time truth-table protocol of [30] to the  $N$ -party functionality computing secret shares of  $g$ . Applying the OTTT protocol to the  $N$ -party functionality which sums its entries (over  $\mathbb{F}_2$ ) before evaluating  $g$ , we get a protocol with total communication  $N \cdot n$  and correlated randomness  $m \cdot 2^{N \cdot n}$ . However, it is straightforward to improve this protocol, by applying the OTTT protocol to the 1-party functionality  $g$ , and letting the trusted dealer distribute random shares of the shift  $r$  to all parties in the preprocessing phase: in the online phase, each party broadcasts his share of the input  $x$ , masked with his share of the shift  $r$ ; this allows all parties to reconstruct  $x + r$ . With

this modification, the parties need only to store a share of the one-dimensional truth-table of  $g$ , of size  $m \cdot 2^n$ .

Therefore, Lemma 8 can be seen as an generalization of the result of [30], which shifts the exponential cost of the correlated randomness from the input size to the locality parameter of the function. In the most general case, when  $c = n$ , we recover the result of [30] (for the special case of the secret sharing functionalities, and up to an additive factor  $n$ ); when  $c < n$ , however, this leads to a protocol which uses a smaller amount of correlated randomness.

*Proof.* Let  $g : \mathbb{F}_2^n \mapsto \mathbb{F}_2^m$  be a  $c$ -local function. Without loss of generality, we assume that each output bit of  $g$  depends on exactly  $c$  input bits. For  $j = 1$  to  $m$ , we denote by  $S_j \subset [n]$  the size- $c$  subset of the bits of the input on which the  $j$ 'th output bit depends. We denote by  $g_j \leftarrow \text{restrict}(g, j)$  the following function:  $g_j : \mathbb{F}_2^c \mapsto \mathbb{F}_2$  is the function which, for any  $x \in \mathbb{F}_2^c$ , computes the  $j$ 'th output bit of  $g(x)$  when given the appropriate subset  $x[S_j]$  of the bits of  $x$  as input.

We describe on Figure 2 the protocol  $\Pi_{\text{local}}$ , which allows  $N$  parties holding shares of an input  $x$  to securely compute (in the semi-honest model, with correlated randomness) shares of  $g(x)$ , for some  $c$ -local function  $g$ . Below, we prove that  $\Pi_{\text{local}}$  satisfies all the properties of Lemma 8. It follows immediately by inspection that the total communication of  $\Pi_{\text{local}}$  is  $N \cdot n$  bits, and that the amount of preprocessing material stored by each party is  $m \cdot 2^c + n$ . We now turn our attention to correctness and security.



**Fig. 2.** Protocol  $\Pi_{\text{local}}$  for securely computing secret shares of a function  $g$  between  $N$ -party, with semi-honest and information-theoretic security in the correlated randomness model.

*Claim.* The protocol  $\Pi_{\text{local}}$  is correct.

Proof: for any  $j \in [m]$ ,

$$\begin{aligned} \sum_{i=1}^N z_{i,j} &= \sum_{i=1}^N M_j^i|_{u[S_j]} \\ &= M_j|_{u[S_j]} \text{ by definition of the } M_j^i \\ &= M_j|_{\sum_i x_i[S_j] + r_i[S_j]} \text{ by definition of } u \\ &= M_j|_{x[S_j] + r[S_j]} \\ &= g_j(x[S_j]) \text{ by definition of } M_j \\ &= g(x)[j] \text{ by definition of } g_j. \end{aligned}$$

We now turn our attention to security. We represent on Figure 3 the ideal secret-sharing functionality for  $g$ . Note that the functionality explicitly allows the adversary to choose the output of the corrupted parties; this is a standard (and minor) technicality of protocols whose output is secret shared between the parties. An alternative is to let the functionality pick the output of all parties at random; however, to realize this functionality, we would need to add a (simple) resharing step at the end of the protocol  $\Pi_{\text{local}}$ , which would add unnecessary communication to the protocol.

**Ideal Functionality  $\mathcal{F}_{\text{local}}$**

The functionality is parametrized with the description of a function  $g : \mathbb{F}_2^n \mapsto \mathbb{F}_2^m$ , and the identities of an adversary  $\mathcal{A}$  and  $N$  parties  $P_1, \dots, P_N$ . The functionality aborts if it receives any incorrectly formatted message.

1. On input a message (**corrupt**,  $C$ ) with  $C \subsetneq [N]$  from  $\mathcal{A}$ , set  $H \leftarrow [N] \setminus C$  and store  $(H, C)$ .
2. On input a message (**input**,  $x_i$ ) from each party  $P_i$  for  $i \in [N]$ , store  $z \leftarrow g(\sum_{i \in [N]} x_i)$  and send **ready** to  $\mathcal{A}$ .
3. On input a message (**set-output**,  $(z_i)_{i \in C}$ ) from  $\mathcal{A}$ , pick  $|H|$  uniformly random values  $(z_i)_{i \in H} \in (\mathbb{F}_2^m)^{|H|}$  under the constraint  $\sum_{i \in H} z_i = z - \sum_{i \in C} z_i$ .
4. On input a message (**send**,  $R$ ) from  $\mathcal{A}$  with  $R \in [N]$ , send  $z_i$  to each party  $P_i$  with  $i \in R$ , and  $\perp$  to all other parties, then terminate.

**Fig. 3.** Ideal Functionality  $\mathcal{F}_{\text{local}}$  for the secure computation of secret shares of  $g(x)$  on an input  $x \in \mathbb{F}_2^n$  shared between  $N$  parties.

*Claim.* The protocol  $\Pi_{\text{local}}$  implements the ideal functionality  $\mathcal{F}_{\text{local}}$  with perfect security against a semi-honest corruption of a majority of the parties.

Proof: let  $H \subset [N]$  denote the subset of honest parties, and let  $C \leftarrow [N] \setminus H$  denote the subset of (passively) corrupted parties; the simulator Sim first

sends  $(\text{corrupt}, C)$  to  $\mathcal{F}_{\text{local}}$  on behalf of the ideal adversary  $\mathcal{A}$ . **Sim** simulates the preprocessing phase by distributing uniformly random coins  $(r_i, (M_j^i)_{j \leq m})_{i \in C}$  to all corrupted parties. In the online phase, the simulator picks random  $u_i$  in  $\mathbb{F}_2^n$  for every  $i \in H$ , and broadcasts them on behalf of the honest parties. When he receives  $(u_i)_{i \in C}$ , he computes for each  $i \in C$   $x_i \leftarrow u_i - r_i$ , and  $z_i \leftarrow (M_j^i|_{u[S_j]})_{j \leq m} \in \mathbb{F}_2^m$ . He sends  $(\text{input}, x_i)$  on behalf of each corrupted party  $P_i$  to the ideal functionality  $\mathcal{F}_{\text{local}}$ , and wait until he receives **ready** from  $\mathcal{F}_{\text{local}}$ . Then, he sends  $(\text{set-output}, (z_i)_{i \in C})$  and  $(\text{send}, R)$  on behalf of  $\mathcal{A}$  to  $\mathcal{F}_{\text{local}}$ , where  $R$  is the set of parties that can obtain the output (which **Sim** can obtain by observing which corrupted parties aborted early). It is immediate to see that the view of the environment (which consists of the preprocessing material, the  $u_i$ , and the outputs of the parties) in the ideal world with **Sim** is perfectly distributed as its view in the real world. This concludes the proof of the core lemma.

## 4 A Sublinear Protocol for Layered Circuits

In this section, we prove Theorem 1, by exhibiting a generic secure multiparty computation protocol in the correlated randomness model against passive corruption of a majority of the parties, for any layered boolean circuit, with sublinear communication in the circuit size  $s$ . Informally, the construction proceeds by breaking the layered circuit into chunks, each chunk containing  $k = k(s)$  consecutive layers, for some function  $k$ . The parties will evaluate the circuit by computing shares of the values carried by the wires leaving a chunk, given as input shares of the values carried by the wires entering the chunk. As a chunk contains  $k$  layers and the directed graph of the circuit has indegree 2, this task corresponds to the secure evaluation of (shares of) a  $2^k$ -local function, with (approximately)  $w$  inputs and  $w$  outputs (where  $w$  is the width of the circuit). By the core lemma (Lemma 8), this can be done with communication  $O(w)$  and using  $O(w \cdot 2^{2^k})$  bits of correlated randomness per party. After  $d/k$  chunk evaluations ( $d$  is the depth of the circuit), the parties end up with shares of the values the output wires, which they can broadcast to reconstruct the output. The total communication involved is  $O(dw/k) = O(s/k)$ , with  $O(2^{2^k} \cdot s/k)$  bits of correlated randomness per party.

### 4.1 Construction

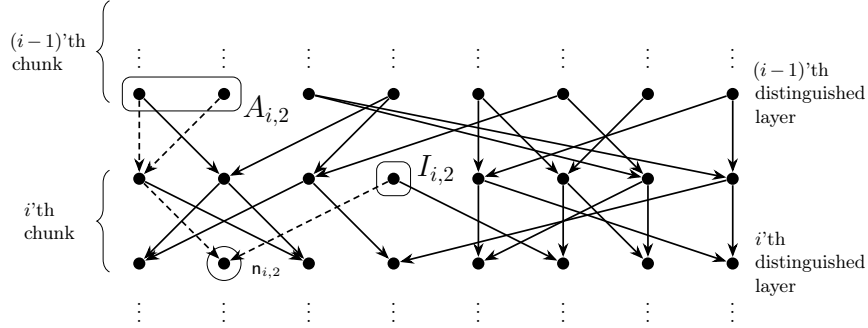
Let  $C$  be a layered boolean circuit with  $n$  inputs and  $m$  outputs, of size  $s$  and depth  $d = d(n)$ , with layers  $(L_1, \dots, L_d)$ . For  $i = 1$  to  $d$ , we let  $w_i$  denote the width of the layer  $L_i$ . We fix an arbitrary ordering of the nodes.

Let  $k$  be an integer. We divide  $C$  into  $d' = \lceil d/k \rceil$  chunks  $(\text{ch}_i)_{i \leq d'}$ , each chunk containing  $k$  consecutive layers (the last chunk contains less layers if  $k$  does not divide  $d$ ). Let  $t \in [k]$  be chosen so that the sum of the widths of the  $t$ 'th layer of each chunk is bounded by  $\lceil s/k \rceil$  (such a  $t$  necessarily exists, otherwise, we would get a contradiction:  $s = \sum_{i=1}^d |L_i| = \sum_{i=1}^k (\sum_{j=1}^{\lceil d/k \rceil} |L_{jk+i}|) > \sum_{i=1}^k \lceil s/k \rceil \geq s$ ).



For  $i = 1$  to  $d'$ , we denote  $t_i$  the index of the  $t$ 'th layer in  $\text{ch}_i$ ; it holds that  $\sum_{i=1}^{d'} w_{t_i} \leq \lceil s/k \rceil$ .

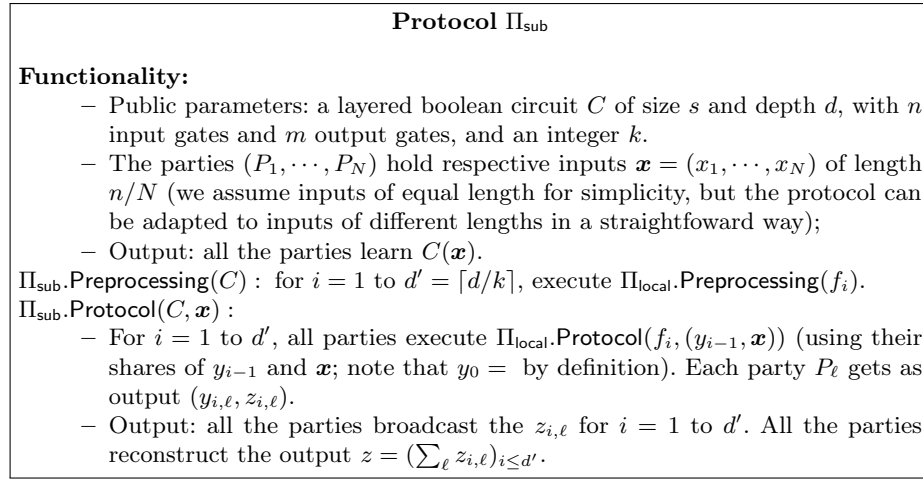
For  $i = 1$  to  $d'$ , we let  $m_i$  denote the number of output nodes between the layers  $L_{t_{i-1}}$  and  $L_{t_i}$  ( $\sum_i m_i = m$ ). For any  $i \leq d'$ , and  $j \leq w_{t_i} + m_i$ , we denote  $n_{i,j}$  the  $j$ 'th node of the layer  $L_{t_i} \in \text{ch}_i$  if  $j \leq w$ , and the  $(j - w)$ 'th output node between the layers  $L_{t_{i-1}}$  and  $L_{t_i}$  otherwise. We associate two sets to each  $n_{i,j}$ : we let  $A_{i,j}$  denote the set of ancestors of  $n_{i,j}$  which belong to  $L_{t_{i-1}}$  ( $A_{1,j}$  is empty for all  $j \leq w_{t_1} + m_1$ ), and we let  $I_{i,j}$  denote the set of input nodes between the layers  $L_{t_{i-1}}$  and  $L_{t_i}$  which are ancestors of  $n_{i,j}$ . We let  $\alpha_{i,j}$  (resp.  $\iota_{i,j}$ ) denote the size of the set  $A_{i,j}$  (resp.  $I_{i,j}$ ). We illustrate this construction on Figure 4. Observe that  $C$  has indegree 2, which implies that any node  $n_{i,j}$  of the  $t$ 'th layer of a chunk can have at most  $2^k$  ancestors in the  $t$ 'th layer of the previous chunk, hence  $\alpha_{i,j} + \iota_{i,j} \leq 2^k$ .



**Fig. 4.** Illustration of the construction of the sets  $(A_{i,j}, I_{i,j})$  for a node  $n_{i,j}$  on a layered directed acyclic graph. The index  $j$  is taken equal to 2 on this figure. The dashed edges denote the paths of the graph that end at  $n_{i,2}$ .

Our protocol proceeds by evaluating the circuit  $C$  on an input  $\mathbf{x}$  (seen as a size- $N$  vector  $(x_1, \dots, x_N)$  over  $\{0, 1\}^{n/N}$ , where  $x_j$  is the input of the party  $P_j$ ) in a chunk-by-chunk fashion. We say that the parties *evaluate a chunk*  $i$  when they compute (shares of) all the values associated to the nodes of the layer  $L_{t_i}$ , as well as (shares of) all the values associated to the output nodes between the layers  $L_{t_{i-1}}$  and  $L_{t_i}$ . Each chunk will be evaluated during a round. We will denote by  $y_{i,\ell}$  the bitstring of the shares of the values on  $L_{t_i}$  computed by the party  $P_\ell$  in the  $i$ 'th round, and  $y_i = \bigoplus_{\ell=1}^N y_{i,\ell}$  the reconstructed value. Similarly, we denote by  $z_{i,\ell}$  the bitstring of the shares of the values on the output wires between  $L_{t_{i-1}}$  and  $L_{t_i}$  computed by the party  $P_\ell$  in the  $i$ 'th round, and  $z_i = \bigoplus_{\ell=1}^N z_{i,\ell}$  the reconstructed output string. For simplicity, for any  $\ell \leq N$ , we denote by  $y_{0,\ell}$  an arbitrary dummy string (this is just to simplify the description of the protocol; as the  $A_{1,j}$  are empty, these strings will not have any effect on the protocol anyway).

For any  $i \leq d'$  and  $j \leq w_{t_i} + m_i$ , we let  $f_{i,j}$  denote the following function: on input the substring  $\mathbf{x}[I_{i,j}]$  of the input string  $\mathbf{x}$ , and the bitstring  $y_{i-1}[A_{i,j}]$  (whose bits form a substring of the values in  $L_{t_{i-1}}$ ),  $f_{i,j}$  outputs the value associated to the node  $n_{i,j}$ . We let  $\delta_i \leftarrow w_{t_i} + m_i$  denote the number of functions  $f_{i,j}$  for a fixed  $i$ . Finally, we denote by  $f_i : \mathbb{F}_2^{w_{t_i} + m_i} \mapsto \mathbb{F}_2^{\delta_i}$  the following function: on input the string  $y_{i-1}$  associated to the distinguished layer of the  $(i-1)$ 'th chunk and the input string  $\mathbf{x}$ ,  $f_i$  outputs  $(f_{i,j}(\mathbf{x}[I_{i,j}], y_{i-1}[A_{i,j}]))_{j \leq \delta_i} = (y_i, z_i)$ . Observe that, by construction,  $f_i$  is a  $2^k$ -local function (the  $j$ 'th output bit of  $f_i$  depends on  $\alpha_{i,j} + \iota_{i,j} \leq 2^k$  input bits). The full protocol is represented on Figure 5.



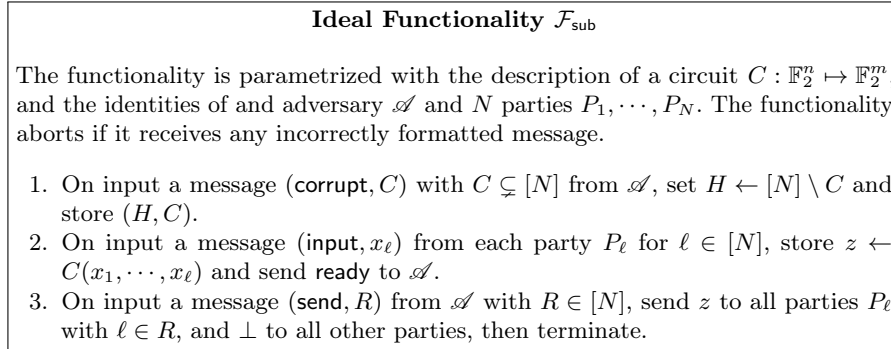
**Fig. 5.** Protocol  $\Pi_{\text{sub}}$  for evaluating a layered boolean circuit  $C$  of size  $s$  and depth  $d$ , with  $n$  input gates and  $m$  output gates, in the correlated randomness model against passive corruption of up to  $N-1$  parties.

## 4.2 Proof of Theorem 1

We now argue that the protocol  $\Pi_{\text{sub}}$  satisfies all the properties outlined in Theorem 1.

**Correctness.** It follows immediately by inspection: by the correctness of  $\Pi_{\text{local}}$ , the values  $y_{i,\ell}$  computed by the parties form shares of the outputs of the functions  $f_{i,j}$  evaluated on the ancestors (in  $L_{t_{i-1}}$ ) of the nodes of layer  $L_{t_i}$  (and the ancestors in  $L_{t_{i-1}}$  of the output nodes between the layers  $L_{t_{i-1}}$  and  $L_{t_i}$ ), as well as on the input nodes between the layers  $L_{t_{i-1}}$  and  $L_{t_i}$ . By definitions, those values are exactly the values associated to the output nodes between the layers  $L_{t_{i-1}}$  and  $L_{t_i}$  and the nodes in the layer  $L_{t_i}$ . From there, it immediately follows that the reconstructed outputs  $(z_1, \dots, z_m)$  are correct.

**Security.** We prove that the protocol  $\Pi_{\text{sub}}$  is perfectly secure against an adversary passively corrupting a majority of the parties. The ideal functionality  $\mathcal{F}_{\text{sub}}$  that  $\Pi_{\text{sub}}$  must realize is straightforward; it is represented on Figure 6. The simulator  $\text{Sim}$  simply simulates  $\Pi_{\text{sub}}$  in the  $\mathcal{F}_{\text{local}}$  hybrid model, relying on the simulator for  $\Pi_{\text{local}}$  to interface with the real protocol. As  $\Pi_{\text{sub}}$  is a simple sequential composition of executions of  $\Pi_{\text{local}}$ , security follows immediately.



**Fig. 6.** Ideal Functionality  $\mathcal{F}_{\text{sub}}$  for the secure computation of secret shares of  $g(x)$  on an input  $x \in \mathbb{F}_2^n$  shared between  $N$  parties.

**Complexity.** We now analyze the communication, storage, computation, and interaction of the protocol  $\Pi_{\text{sub}}$ . We first outline a straightforward optimization: observe that each execution of  $\Pi_{\text{local}}$  to evaluate (shares of) the output of one of the  $f_i$  operates, in particular, on the input  $\mathbf{x}$  (whose length is  $n$ ). Instead of using independent executions of  $\Pi_{\text{local}}$ , where the input vector  $\mathbf{x}$  ends up being re-shared between the parties for each execution, the parties can share it only once in an “input sharing step”, before the execution of the first instance of  $\Pi_{\text{local}}$ , and reuse these shares in each execution. With this optimization, the parties exchange  $n$  bits in the input sharing step, and  $N \cdot (\delta_i)$  bits during the  $i$ 'th round of the circuit evaluation step, for  $i = 1$  to  $d' = \lceil d/k \rceil$ . Therefore, the total number of bits exchanged is

$$n + N \cdot \sum_{i=1}^{d'} w_{t_{i-1}} + m_{i-1} \leq n + N \cdot (m + \lceil s/k \rceil)$$

(note that the additive factor  $n$  would be  $n \cdot d'$  without the simple optimization outlined above). The amount of correlated randomness stored by each party can be upper bounded by

$$n/N + \sum_{i=1}^{d'} \sum_{j=1}^{\delta_i} 2^{\alpha_{i,j} + \iota_{i,j}} \leq n/N + \sum_{i=1}^{d'} \sum_{j=1}^{\delta_i} 2^{2^k} \leq n/N + (m + \lceil s/k \rceil) \cdot 2^{2^k},$$

where the first inequality comes from the fact that any node  $n_{i,j}$  of the  $t$ 'th layer of a chunk can have at most  $2^k$  ancestors in the  $t$ 'th layer of the previous chunk, which leads to the claimed total storage. Eventually, the round complexity of the protocol is  $d' + 1 = O(s/k)$ , and the computation performed by each party essentially boils down to performing  $m + \lceil s/k \rceil$  searches in lookup tables of size bounded by  $2^{2^k}$ , which takes time  $(m + \lceil s/k \rceil) \cdot 2^k$ .

### 4.3 Extension to Layered Arithmetic Circuits

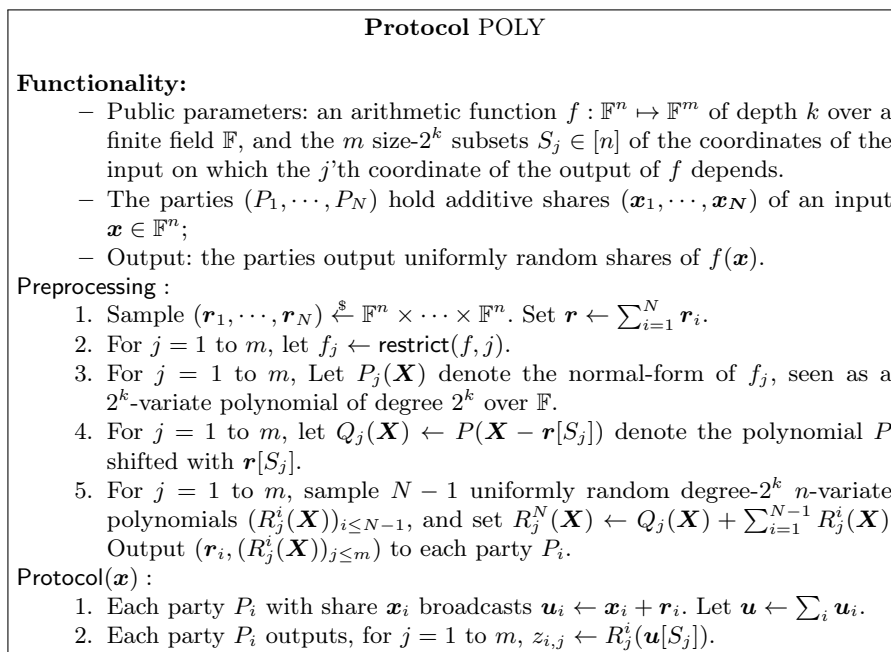
So far, our protocol does not readily extend to arithmetic circuits over (exponentially large) finite fields. The main obstacle toward getting an arithmetic analogue of the protocol  $\Pi_{\text{sub}}$  lies in the generalization of the core lemma to the arithmetic setting: our proof of Lemma 8 relies on the fact that we can use the OTTT protocol of [30] to evaluate functions with a “small enough” truth-table. While in the boolean case, any functionality with  $c$  input bits has a truth table of size  $2^c$ , this is not true anymore for arithmetic functionalities over large fields, where even single-input functions have truth table of exponential size. In addition, the standard conversion of arithmetic circuits into boolean circuits would blow up the size too much: any size- $s$  arithmetic circuit can be securely evaluated (in the correlated randomness model) with communication  $O(s)$  (counting the number of field elements), but the conversion to a boolean circuit will in general blow up the circuit size by a  $\log s$  factor, while our protocol only saves a factor  $\log \log s$ , and does therefore not lead to a sublinear communication protocol for arithmetic circuits.

Nevertheless, we show that our protocol can be extended to the arithmetic setting, by exhibiting a natural analogue of the OTTT protocol, tailored to arithmetic functions. Our starting point is the recent work of [39], on conditional disclosure of secret and private simultaneous message (PSM) protocols. The authors of [39] build an elegant PSM protocol for multivariate polynomial evaluation. The protocol has the following features: Alice holds an  $n$ -variate polynomial  $P$  of degree  $\text{deg}$ , Bob holds a vector of input  $\mathbf{x} \in \mathbb{F}^n$ , and both parties share a common random string. They send a single simultaneous message to a third player, Charlie, with optimal communication (Alice’s message has size  $O(\binom{n+\text{deg}}{\text{deg}})$ , Bob’s message has size  $O(n)$ ). This allows Charlie to learn  $P(\mathbf{x})$ , and nothing more. The protocol works as follows:

- The shared randomness is  $\mathbf{r} \in \mathbb{F}^n$  and a random  $n$ -variate degree- $\text{deg}$  polynomial  $R$ .
- Alice sends  $(\mathbf{x}', u) \leftarrow (\mathbf{x} + \mathbf{r}, R(\mathbf{x} + \mathbf{r}))$ .
- Bob sends the polynomial  $Q(\mathbf{X}) = P(\mathbf{X} - \mathbf{r}) + R(\mathbf{X})$ .
- Charlie outputs  $Q(\mathbf{x}') - u$ .

The correctness follows immediately by inspection, and security follows by the argument of [39, Section 5.1]. The above PSM can be readily converted into a 2-player arithmetic analogue of the OTTT protocol, which relies on a multivariate polynomial representation of an arithmetic circuit (instead of a truth table

representation). We represent on Figure 7 a variant of the protocol  $\Pi_{\text{local}}$ , tailored to the arithmetic setting (over an arbitrary field  $\mathbb{F}$ ). Each party sends  $n$  field elements (which is essentially optimal), and stores  $O\left(\binom{n+\text{deg}}{\text{deg}}\right)$  field elements.



**Fig. 7.** Protocol POLY for evaluating an arithmetic function  $f$  over a finite field  $\mathbb{F}$  in the correlated randomness model, against a passively corrupted majority

Using the above protocol, we immediately get a generalization of Lemma 8:

**Lemma 9.** *For any depth- $k$  arithmetic circuit  $f : \mathbb{F}^n \mapsto \mathbb{F}^m$ , there is an information-theoretic semi-honest  $N$ -party secure computation protocol (with dishonest majority) in the correlated randomness model for computing secret shares of  $f$  with total online communication  $N \cdot n$  elements of  $\mathbb{F}$ , and correlated randomness  $m \cdot \binom{2^{k+1}}{2^k} + n \approx m \cdot 2^{2^{k+1}} / \sqrt{\pi 2^k} + n$  elements of  $\mathbb{F}$  per party.*

Therefore, we get polynomial storage (in  $s$ ) by setting  $k \leftarrow \log \log s$ , as before. This leads to a protocol for arithmetic circuits of size  $s$ , with  $n$  inputs and  $m$  outputs, with polynomial storage and total communication  $O(n + N \cdot (m + s / \log \log s))$ .

#### 4.4 Further Extensions

We sketch in this section how to extend our protocol to the case of function-independent correlated randomness, and to the case of tall-and-narrow circuits.

**Function-Independent Preprocessing.** We introduce below a variant of the core lemma, tailored to function-independent preprocessing. Theorem 3 follows immediately from this variant.

**Lemma 10.** *For any  $c$ -local function  $g : \mathbb{F}_2^n \mapsto \mathbb{F}_2^m$ , there is an information-theoretic semi-honest  $N$ -party secure computation protocol (with dishonest majority) in the function-independent correlated randomness model for computing secret shares of  $g$  with total online communication  $N \cdot n$  bits, and correlated randomness  $m \cdot 2^{c+2^c} + n$  bits per party.*

*Proof.* To prove Lemma 10, we modify  $\Pi_{\text{local}}$  as follows: instead of computing shares of the truth table  $M_j$  of  $g_j$  (which is of size  $2^c$ ) permuted with the shift  $r[S_j]$ , we consider the list  $(M_{j,q})_{q \leq 2^{2^c}}$  of all possible truth tables, corresponding to a lexicographic ordering of all possible functions  $g_j$ , each table being shifted with the same  $r[S_j]$ . Each party  $P_i$  receives  $(r_i, (M_{j,q}^i)_q)$ , which amounts to  $n + 2^c \cdot 2^{2^c}$  bits of correlated randomness. In the online protocol  $\Pi_{\text{local.Protocol}}$ , when the functions  $g_j$  are revealed, the parties locally drop all unnecessary shares of shifted truth tables, keeping only the one corresponding to  $g_j$ . The security analysis immediately follows from the analysis of  $\Pi_{\text{local}}$ .

**Tall-and-Narrow Circuits.** For tall-and-narrow circuits, whose width  $w$  is small, the proof follows by observing that in this situation the bound on the size of the sets  $A_{i,j}$  and  $I_{i,j}$  can be refined to  $|A_{i,j}| + |I_{i,j}| \leq w \cdot k$ , hence the  $f_{i,j}$  have truth tables of size bounded by  $2^{w \cdot k}$ . Theorem 5 follows immediately.

## 5 Malicious Setting

In the two-party case, combining our passively secure protocol  $\Pi_{\text{sub}}$  with the techniques of [19] directly implies the existence of a (statistical) unconditionally secure two-party protocol secure against malicious adversaries, with communication  $O(n + m + \frac{s}{\log \log s} + \kappa)$  for a layered boolean circuit of size  $s$ , where  $\kappa$  is a statistical security parameter. Indeed, the protocol of [19] has a structure similar to our protocol: it decomposes the circuit into tables, and distributes scrambled version of these tables to the parties in the preprocessing phase. Each gate of the circuit is evaluated using the OTTT protocol to obliviously select the output of the gate from its corresponding scrambled truth-table.

To enhance this protocol to security against malicious adversaries, [19] uses a simple and natural information-theoretic authentication procedure. Namely, for each entry  $b \in \{0, 1\}$  of a table given to the first party (let us call it  $A$ ), the trusted dealer additionally generates two  $\kappa$ -bit string  $(x_0, x_1)$ , hands  $x_b$  to  $A$ , and  $(x_0, x_1)$  to the other player  $B$ . This way, when  $A$  must send the entry  $b$  of the table to  $B$ , she can authenticate  $b$  by sending it along with  $x_b$ .  $B$  then retrieves the corresponding value  $x_b$  and checks that  $A$  honestly opened  $b$ ; if  $A$  is dishonest, she will be caught with probability  $1 - 2^{-\kappa}$ . Note that the only local computation performed by the parties are searches through lookup table, hence

authenticating each entry this way suffices to guarantee security of the entire protocol.

However, directly applying this approach would require transmitting  $\kappa$  bits per output of a table, which would increase the total communication by a factor of  $\kappa$ . To avoid this overhead, the authors of [19] observe that it is not necessary to explicitly authenticate each entry sent by a party. Instead, each time  $A$  reveals an entry  $b$  corresponding to some authentication string  $x_b$ , she locally updates a “global MAC key”  $\Delta_A \leftarrow \Delta_A \oplus x_b$ , where  $\Delta_A$  is set to 0 at the start of the protocol. Simultaneously, when he receives an entry  $b$  from  $A$ ,  $B$  retrieves the pair  $(x_0, x_1)$  corresponding to this entry, and locally updates  $\Gamma_B \leftarrow \Gamma_B \oplus x_b$ , where  $\Gamma_B$  is set to 0 at the start of the protocol. The parties proceed symmetrically, with  $\Delta_B, \Gamma_A$ , when  $B$  sends an entry to  $A$ . At the end of the protocol,  $A$  reveals  $\Delta_A$  and  $B$  reveals  $\Delta_B$ . If  $\Delta_B \neq \Gamma_A$ ,  $A$  aborts the protocol;  $B$  does the same if  $\Delta_A \neq \Gamma_B$ . If both checks passed, the parties reconstruct the output. The analysis of [19] shows that this guarantees that no party can cause its opponent to accept an incorrect output, except with probability  $2^{-\kappa}$ . It increases the amount of preprocessed material by a factor  $\kappa$ ,<sup>2</sup> but only adds  $2\kappa$  bits to the total communication.

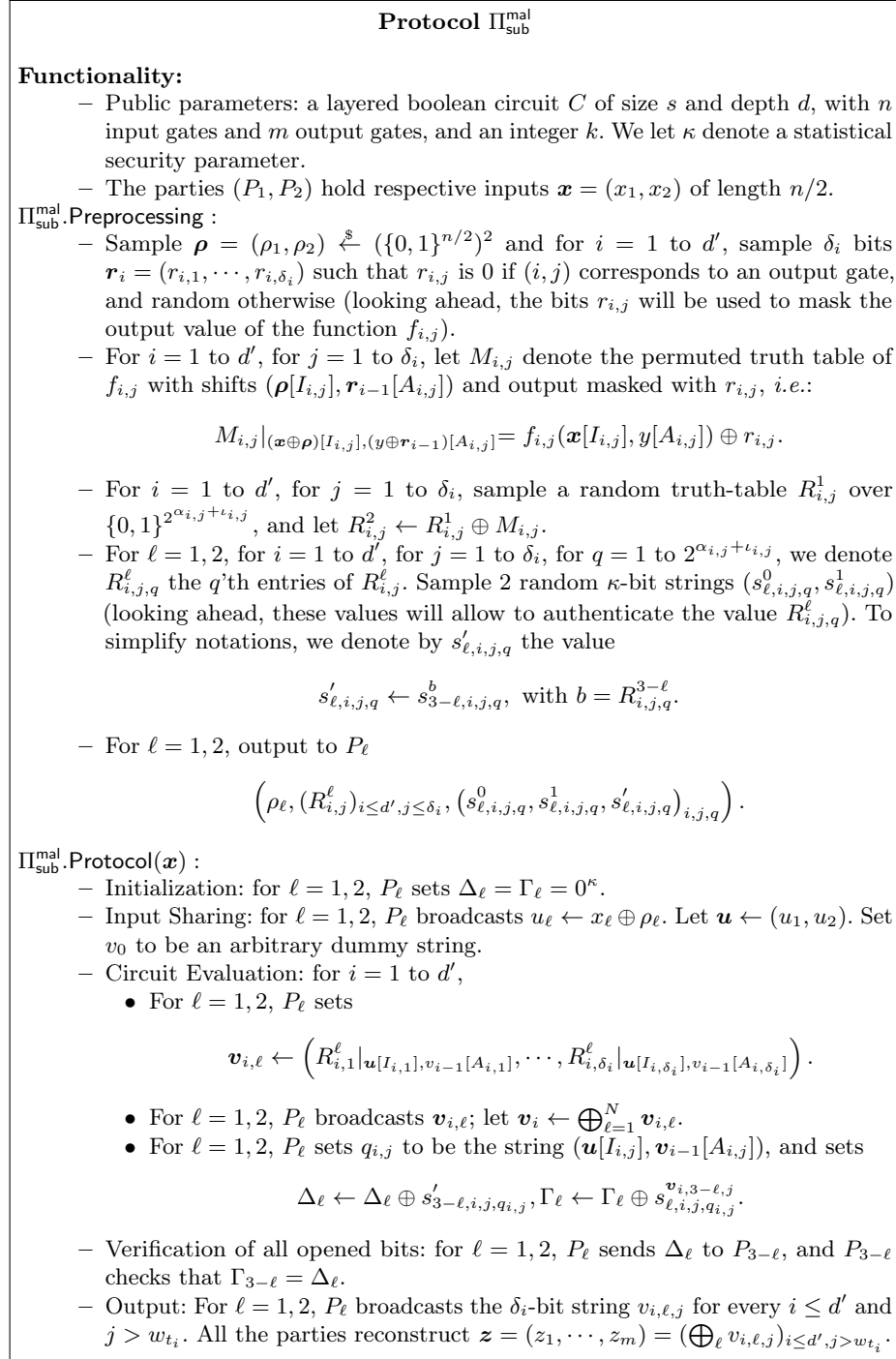
For completeness, we provide a full self-contained description of the maliciously-secure two-party version of our protocol on Figure 8. We refer the reader to Theorem 1 of [19] for a detailed proof of security against malicious adversaries; it is straightforward to adapt the proof to our protocol (we note that, while [19] focuses on small tables implementing standard two-input boolean gates, [19, Section 2.3] already observes that this mechanisms can be directly generalized to protocols evaluating larger tables).

**Extension to  $N$  Parties.** While the work of [19] focused only on (maliciously secure) two-party computation, it was subsequently observed in [34] that the techniques used in [19] can be easily generalized to the multiparty setting, for an arbitrary number  $N$  of parties. We refer the reader to [34] for more details; this directly gives:

**Theorem 11.** *For any  $N$ -party functionality  $f$  represented by a layered boolean circuit  $C$  of size  $s$  with  $n$  inputs and  $m$  outputs, and for any integer  $k$  and statistical security parameter  $\kappa$ , there is a  $\kappa$ -secure protocol which realizes  $f$  in the preprocessing model against malicious adversaries with adaptive corruption (of up to  $N - 1$  parties), with communication  $n + N \cdot (m + \lceil s/k \rceil + \kappa)$  and correlated randomness  $n/N + (3\kappa + 1) \cdot (m + \lceil s/k \rceil) \cdot (2^{2^k} + 1)$  per party.*

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<sup>2</sup> A technique to amortize this overhead, using a linear MAC scheme, is described in [19]; it applies to our setting as well, and allows to remove this factor  $\kappa$  overhead in the storage complexity, but we focus on the more naive approach in this work for simplicity.



**Fig. 8.** Two-party protocol  $\Pi_{\text{sub}}^{\text{mal}}$  for evaluating a layered boolean circuit  $C$  of size  $s$  and depth  $d$ , with  $n$  input gates and  $m$  output gates, in the correlated randomness model against active corruption of one of the parties.



## 6 Open Questions

While our work shows that a large class of circuits of size  $s$  can be securely evaluated in the correlated randomness model using  $o(s)$  communication, many questions related to the communication of MPC in the correlated randomness model remain open.

*Question 1.* Can our protocols be extended to arbitrary non-layered circuits?

It is immediate to extend our protocol to any circuit that is layered “by blocks” of depth  $c$ , in the sense that no edge crosses more than  $c$  consecutive layers, for any  $c = o(\log \log s)$ . However, generalizing our result to all circuits remains an interesting open question.

*Question 2.* Can we achieve better sublinearity for unconditional MPC in the correlated randomness model, in general or for specific circuits?

It is known that some specific functions can be evaluated in the correlated randomness model, with stronger sublinearity guarantees than those obtained in this work. In particular, *matrix multiplication* can be computed with communication linear in the size  $n^2$  of the matrices, while the best known algorithm for multiplying matrices of size  $n$  requires  $O(n^t)$  communication, with  $t \approx 2.3$ . The work of [8] also implies the existence of low-communication protocols in the correlated randomness model, for  $N \geq 3$  parties, for specific types of constant-depth circuits. It would be interesting to improve the sublinearity of our work, and to characterize the functions for which better sublinearity can be achieved.

*Question 3.* Can we achieve sublinear communication and linear storage at the same time?

By a lower bound of [42], linear storage is the best we can hope for. Our protocols only achieve slightly superlinear storage; in the regime where the  $1/\log \log s$  factor would give non-trivial communication savings, this implies that a rather large storage is required. Protocols for specific functions, such as matrix multiplication, achieve both sublinearity and linear storage, but the question remains open for more general functions.

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