Experimenting with Collaborative zk-SNARKs: Zero-Knowledge Proofs for Distributed Secrets

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https://www.usenix.org/conference/usenixsecurity22/presentation/ozdemir
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Abstract
A zk-SNARK is a powerful cryptographic primitive that provides a succinct and efficiently checkable argument that the prover has a witness to a public NP statement, without revealing the witness. However, in their native form, zk-SNARKs only apply to a secret witness held by a single party. In practice, a collection of parties often need to prove a statement where the secret witness is distributed or shared among them.

We implement and experiment with collaborative zk-SNARKs: proofs over the secrets of multiple, mutually distrusting parties. We construct these by lifting conventional zk-SNARKs into secure protocols among N provers to jointly produce a single proof over the distributed witness. We optimize the proof generation algorithm in pairing-based zk-SNARKs so that algebraic techniques for multiparty computation (MPC) yield efficient proof generation protocols. For some zk-SNARKs, optimization is more challenging. This suggests MPC “friendliness” as an additional criterion for evaluating zk-SNARKs.

We implement three collaborative proofs and evaluate the concrete cost of proof generation. We find that over a 3Gb/s link, security against a malicious minority of provers can be achieved with approximately the same runtime as a single prover. Security against N − 1 malicious provers requires only a 2× slowdown. This efficiency is unusual since most computations slow down by orders of magnitude when securely distributed. This efficiency means that most applications that can tolerate the cost of a single-prover proof should also be able to tolerate the cost of a collaborative proof.

1 Introduction
Zero-knowledge succinct, non-interactive arguments of knowledge (zk-SNARKs) [23] are publicly verifiable proofs that the prover has secret data (a witness) that satisfies a certain public NP relation. The proof reveals nothing about the secret data other than its validity. zk-SNARKs have two key strengths. First, they are very general: there are zk-SNARKs that can prove any relation expressible as a bounded size arithmetic circuit. Second, they can be easily verified: proof size and verification time are sublinear in the amount of secret data (typically kilobytes and tens of milliseconds), and verification can be performed by anyone. However, there are two key limitations. First, generating the proof is expensive: typically thousands of times slower than checking the relation directly [91, 98, 106, 108]. Second, the secret data must be

\[
\begin{align*}
\mathcal{P}(w) &\xrightarrow{\pi_w} \mathcal{V} \xleftarrow{\pi_{\overline{w}}} \mathcal{V}' \\
\mathcal{P}_1(w_1) &\xrightarrow{} \mathcal{P}_2(w_2) \xleftrightarrow{} \mathcal{P}_3(w_3)
\end{align*}
\]

(a) A conventional zk-SNARK  (b) A collaborative zk-SNARK with one party’s secret w with multiple parties’ secrets.

Figure 1: Conventional and collaborative zk-SNARKs

held by a single party: generating a proof when the secret data is distributed among multiple parties is not natively supported.

Being limited to secrets that are held by a single party bars some applications. We describe some examples in Section 2.

Collaborative zk-SNARKs In this work we generalize zk-SNARKs to collaborative zk-SNARKs. That is, we generalize from public proofs about a secret w held by a single party, to public proofs about a secret \( \overline{w} = (w_1, \ldots, w_N) \) distributed among N parties, where party \( i \) has \( w_i \), for \( i \in [N] \), as in Fig. 1. The proof generation process should reveal nothing new about \( \overline{w} \) to a coalition of parties, other than the validity of \( \overline{w} \).

A natural approach for constructing a collaborative zk-SNARK is as follows: start from a standard single-prover zk-SNARK, and run its proof generation algorithm as a secure multi-party computation (MPC) [49] among the N provers.

Recall that, informally, MPC allows N parties to compute a public function \( f \) : \( X^N \rightarrow Y \) over secret inputs \( x_1, \ldots, x_N \in X \), where party \( i \) has \( x_i \) for all \( i \in [N] \). At the end of the protocol, all parties learn the output \( y \leftarrow f(x_1, \ldots, x_N) \), but learn nothing else about each others’ inputs. MPC protocols work for any function \( f \) that can be expressed as a reasonable size arithmetic circuit.

By taking the MPC function \( f \) to be the circuit representation of a zk-SNARK proof generator, the provers can use a generic MPC protocol to jointly generate the desired proof without revealing anything else about their secret inputs. One technicality is that proof generation is a randomized algorithm, but this is easily supported by MPC protocols.

Another issue is that any prover can choose not to participate in the MPC, thereby preventing the proof from being generated. Hence, a collaborative zk-SNARK only makes sense in a setting where all N provers want to jointly generate
a valid proof. This lets us use a slightly simpler kind of MPC, called **MPC-with-aborts** [49], that does not guarantee output delivery. Regardless, using an MPC protocol to generate a proof ensures that no matter what a malicious set of provers does, it learns nothing about the secret data of the honest provers, other than the validity of the witness tuple \( \vec{w} \).

As we will see, a direct application of this approach leads to very poor performance. Generating a standalone zk-SNARK proof is already computationally expensive. Running this entire process naively through a general purpose MPC protocol would likely result in unacceptable proving time. In particular, since MPC protocols and zk-SNARK provers are each thousands of times slower than their underlying functionality, their composition is likely to be millions of times slower.

Instead, we show that for some zk-SNARKs, proof generation lends itself to a very efficient MPC protocol. For these zk-SNARKs it is quite efficient to generate the proof distributively, without the provers leaking undesired information about their secrets to the public or to each other. This suggests MPC “friendliness” of the prover as a new criterion for evaluating zk-SNARKs. As we will see, some zk-SNARKs are MPC-friendly, while others are less so.

### Our techniques

We design optimized multi-prover protocols by using three ideas:

1. **Building on previous and new techniques**, we apply MPC techniques based on secret-sharing **directly** to elliptic curves, making secure, distributed curve operations cheap.
2. **By using secret sharing over curve points** we streamline key prover bottlenecks, such as multi-scalar multiplication, polynomial division, and Fourier transform.
3. **We employ an optimized MPC protocol** for computing sequences of partial products. In particular, for field elements \( x_1, \ldots, x_n \) we use an efficient MPC protocol from [4] to compute \( p_j := \prod_{i=1}^j x_i \) for all \( j \in \{1, \ldots, n\} \).

Ultimately, we adapt four zk-SNARKs to multiple provers: Groth16 [66], Plonk [53], Marlin [40], and Fractal [41].

### Our results

- First, we formally define collaborative zk-SNARKs (§4).
- Next, we design four collaborative proofs by adapting four existing zk-SNARK provers into MPC protocols that exploit the algebraic nature of these provers: (§5):
  - Groth16 and Marlin give very efficient protocols.
  - Plonk results in a protocol with more communication between the provers, due to the larger number of multiplications during proof generation.
  - Fractal can achieve MPC efficiency, but at the expense of increasing the proof size and verification time by a factor of \( N \).

This list classifies these zk-SNARKs by their MPC-friendliness, from most to least friendly. Some of these constructions depend on a trusted setup.

- We implement our Groth16-, Marlin-, and Plonk-based protocols. We do so by lifting a single-prover code base called **arkworks** [48] and making it support multiple provers. The key implementation technique is to replace **arkworks’ low level implementation of fields and curves with secret shares** that implement multi-party protocols for these types (§6).

- Finally, we demonstrate the concrete efficiency of our constructions. We find that with a 3Gb/s link, protocols secure against a malicious minority of provers run in essentially the same time as a single-party proof. Protocols secure against a malicious majority run in twice the time of a single-party proof (§7). These results are unusual in that MPC typically incurs a significant overhead. These results also show that collaborative proofs are practical.

**Publicly auditable MPC** Collaborative zk-SNARKs give an efficient construction for a cryptographic primitive called **publicly auditable MPC** (PA-MPC) [7]. A PA-MPC protocol is an MPC that also produces a proof by which the public can verify that the computation was performed correctly with respect to commitments to the inputs. Classic PA-MPC constructions [7] have linear size proofs. Collaborative zk-SNARKs yield PA-MPC with constant size proofs. We discuss this further in Sections 2 and 4.1. Concurrent work [75] builds on a similar idea.

### 2 Example applications

Collaborative zk-SNARKs come up naturally in many settings, both for real-world applications and for conceptual ones. In this section we briefly survey a number of situations where they are needed.

1. **Healthcare statistics** Healthcare providers provide services (medication, operations, etc.) to millions of patients every day. Their actions concern the intimate details of individuals’ well-being and **must not** be publicly revealed. Yet, the public benefits from understanding costs in the healthcare system, in aggregate. For example, a recent investigation revealed that many hospitals routinely charge uninsured patients the highest rates [50].

   The challenge is to compute aggregate statistics over sensitive data in a fashion that certifies the accuracy of the aggregates, despite the participants’ incentives to omit data or mis-aggregate. This problem is not unique to healthcare; it also arises for criminal justice statistics [51].

   Collaborative proofs yield a clean solution. First, every healthcare provider publishes a short Merkle commitment to the list of services that it provided. Second, regulators can request aggregate statistics over the set of services provided by all healthcare providers. The requested statistic is defined by an arithmetic circuit that computes the statistic. Third,
providers can compute the aggregate statistic via an MPC among the providers, and publish a collaborative proof that the claimed result is consistent with all of their local commitments. This protects both the privacy of patient data and ensures the integrity of the aggregate.

The Merkle commitments to the underlying data are crucial for integrity. If at a later time there is a dispute over the accuracy of the aggregate statistic, a forensics investigator can request the providers to open their local Merkle commitments. A provider that committed to incorrect data will immediately be caught and penalized, and this provides a strong incentive for providers to commit to accurate data (refusing to open a commitment could similarly carry a penalty). Patients are also incentivized to check (via a Merkle proof) that the provider accurately committed to the service they received.

Overall, the collaborative proof that the computed aggregate is consistent with the committed data ensures accuracy of the aggregate while maintaining data privacy.

(2) Computing credit scores The three US credit bureaus collect financial information from many institutions and use that data to compute a credit score for all US individuals. Credit scores are used to make credit related decisions such as who gets a loan. When a credit bureau is hacked [47], everyone’s information is exposed.

Collaborative proofs provide a very different solution for computing credit scores. Let $A_1, \ldots, A_N$ be $N$ institutions whose data is used to compute a person’s credit score. Each $A_i$ could publish a short Merkle commitment to the entirety of its local dataset. When Alice needs her credit score, the $N$ institutions engage in an MPC to compute her score based on their local data, and collaboratively construct a proof that the score is consistent with the public commitments. Alice is given her score $s$ and the short collaborative proof $\pi$. She can present $(s, \pi)$ to a lender, who can validate $\pi$ with respect to the public commitments to ensure that Alice’s score is correct.

$A_i$’s commitment to its local data ensures that if there is a dispute over Alice’s credit score, her leaves of the Merkle commitment can be opened. If an incorrect credit score is computed, there is enough information to identify the cause. This is a more satisfying guarantee than simply asking the institutions to certify the computed credit score by signing it.

An example. Consider a lender that is reviewing a loan application. The lender wants to know the sum of the applicant’s credits and debits with the banks where the applicant has an account. The applicant could request a collaborative proof from its banks to prove that the difference between its debits (assets) and its credits (debts) is larger than some threshold $T$. In Section 7 we evaluate the time required to build this proof.

(3) Private audits of multiple parties The global financial system can be modeled as a transaction graph: every account is a node, and there is an edge from $u$ to $v$ if account $u$ issued a payment to account $v$. Financial regulators often need to answer the following question: did account $u$ at Bank $A$ pay account $v$ at Bank $B$, either directly or via intermediate accounts? If the entire graph were stored at a single location this would be easy to answer: the entity that holds the graph could commit to the graph and provide a succinct proof that there is no chain of transactions $u \rightarrow a_1 \rightarrow \cdots \rightarrow a_t \rightarrow v$.

In reality, the transaction graph is distributed and siloed across a number of institutions. Each financial institution only sees the portion of the graph that touches accounts under its control. Yet, the regulator needs an answer to questions about the entire graph. Previous work applies MPC techniques to this problem [34, 86, 96], but the results are not verifiable by a third party (i.e., the regulator or the public).

Collaborative proofs provide a clean solution. First, every bank commits to its local view of the graph. They then collaboratively generate a succinct proof that there is no path from $u$ to $v$ in the union of their graphs. More generally, this applies to any query about the global graph that can be computed using an arithmetic circuit of reasonable size. The banks can collaboratively compute an answer to the query, along with a proof that the answer is consistent with each bank’s commitment to its local view of the graph. In case of a later dispute, the commitments can be opened to identify who is at fault.

Publicly auditable MPC The three applications above are special cases of publicly auditable MPC (PA-MPC) [7]: a general primitive that extends secure MPC to produce a proof that can be independently checked to verify that the computation was performed correctly. The auditor need not participate in this computation, yet it can still check that the claimed output is consistent with commitments to the inputs.

We will see in Section 4.1 that collaborative proofs yield a PA-MPC with proof size and verification time sublinear in the circuit size.

Reducing the number of parties Some of the applications in this section could involve many online parties, and the resulting MPC can be expensive in communication (§7). However, by having each data provider secret share their data among a small set of non-colluding servers, we can make the cost independent of the number of data providers. Furthermore, only the small set of non-colluding servers need to be simultaneously online. A similar approach has been successfully used in real-world MPC-based auctions [26].

3 Background

Let $\mathbb{F}_p$ denote the finite field of integers mod $p$, a prime. We omit $p$ when unambiguous. Let $G$ be an additive cyclic group of prime order $q$, group operation $+$, inverse operation $-$, identity $0$, and generator $g$.

Let $G_1$, $G_2$, and $G_T$ be cyclic groups of prime order $q$ with generators $g_1 \in G_1$ and $g_2 \in G_2$. Let $e : G_1 \times G_2 \rightarrow G_T$ be an efficiently computable and non-degenerate pairing, so that

$$e(\alpha h_1, \beta h_2) = \alpha \beta e(h_1, h_2)$$
for all \( \alpha, \beta \in \mathbb{F}_q \) and all \( h_1 \in \mathbb{G}_1 \) and \( h_2 \in \mathbb{G}_2 \).

### 3.1 Multi-party computation

Let \( f : X^k \to Y \) be a function and let \( P_1, \ldots, P_n \) be \( N \) parties such that \( P_i \) has input \( x_i \in X \). A multi-party computation (MPC) for \( f \) is a protocol \( \Pi \) among the parties which computes and reveals \( y := f(x_1, \ldots, x_N) \). Informally, \( \Pi \) is secure if it reveals nothing except \( y \), and it requires malicious parties to chose their inputs independently from the honest parties. In more detail, \( \Pi \) is secure against \( t \) dishonest parties if for every efficient adversary \( A \) corrupting \( t \) parties, there is an efficient simulator \( \mathcal{S} \) which can produce a protocol transcript from the output and the inputs of those \( t \) parties, that is indistinguishable from the view of those parties in a real protocol execution.\(^1\) We refer the reader to [81] for the complete definition.

Many protocols are secure with one caveat: aborts. An adversary can cause the protocol to abort and deny the output to all honest parties. This security model is called security-with-aborts and it is sufficient in our setting. Many protocols achieve security-with-abort (or better) assuming an honest majority of parties [9, 10, 20, 22, 27, 30, 31, 39, 44, 45, 54, 55, 57, 62, 64, 82, 89, 94] or even in the presence of a single honest party [46, 58, 73, 80, 83, 84, 85, 88]. The latter is called the dishonest majority setting.

**Arithmetic circuit MPC** We build on MPC protocols that can securely compute any function defined as a bounded size arithmetic circuit. An arithmetic circuit is a directed acyclic graph of gates and wires. Wires carry values in a finite field \( \mathbb{F} \) (e.g., \( \mathbb{F}_p \)). Input wires can be public, or private to a single party. Gates take two inputs and perform multiplication or addition in \( \mathbb{F} \). Certain wires are designated as outputs.

The two protocols we build on—SPDZ [46] and GSZ [64]—are based on secret sharing. They evaluate the circuit by representing each wire’s value using a \( k \)-out-of-\( N \) secret-sharing: \( N \) tokens distributed among the parties such that no \( k - 1 \) tokens reveal the true value, but any \( k \) tokens do. We use \( [y] \) to denote that \( y \) is shared using a secret sharing scheme.

- SPDZ [46] uses additive secret sharing where \( y \in \mathbb{F} \) is represented by random shares \( y_1, \ldots, y_N \) that sum to \( y \). All \( N \) shares are required to recover \( y \).
- GSZ [45, 64] uses Shamir secret sharing [99] where \( y \in \mathbb{F} \) is shared among \( N \) parties as \( y_1, \ldots, y_N \) such that the points \( (1, y_1), \ldots, (N, y_N) \) lie on a polynomial \( p(X) \) of degree at most \( d \), with \( p(0) = y \). Any \( d + 1 \) points suffice to reconstruct the polynomial and \( y \).

SPDZ [46] is a dishonest majority protocol. GSZ [64] is representative of a new generation of state-of-the-art honest majority protocols [22, 27, 30, 31, 62, 64].

### 3.2 zk-SNARKs

Informally, a proof for a relation \( \mathcal{R} \) is a protocol between a prover \( \mathcal{P} \) and an efficient verifier \( \mathcal{V} \) by which \( \mathcal{P} \) convinces \( \mathcal{V} \) that \( \exists w : \mathcal{R}(x, w) = 1 \), where \( x \) is a called an instance, and \( w \) a witness for \( x \). If the proof comprises a single message from \( \mathcal{P} \) to \( \mathcal{V} \), it is said to be non-interactive and has syntax:

- Setup\((1^k, \mathcal{R}) \to \mathbb{P}p) \): setup public parameters for \( \mathcal{R} \).
- Prove\((\mathbb{P}p, x, w) \to \pi) \): if \( (x, w) \in \mathcal{R} \), output a proof \( \pi \), otherwise \( \bot \).
- Verify\((\mathbb{P}p, x, \pi) \to \{0, 1\}) \): check a proof.

Generally, proofs support a class of relations (e.g., bounded size arithmetic circuits). The supported class must also define the size of a relation—denoted \( |\mathcal{R}| \)—the time needed to evaluate it (e.g., the number of gates in the circuit). In the next subsection we discuss common classes of relations.

A zk-SNARK is a proof with the following properties [23]:

- **Completeness:** If \( \mathcal{R}(x, w) = 1 \), then an honest \( \mathcal{P} \) convinces \( \mathcal{V} \) except with negligible probability. If \( \mathcal{R}(x, w) = 0 \), Prove outputs \( \bot \).
- **Zero knowledge:** Informally, the triple \((\mathbb{P}p, x, \pi) \) reveals nothing about the witness \( w \).
- **Knowledge soundness:** Informally, for every \( \mathcal{P}^* \) there exists an efficient algorithm, called an extractor, such that whenever \( \mathcal{P}^* \) convinces \( \mathcal{V} \) to output 1, the extractor can use \( \mathcal{P}^* \) to output \( w \) such that \( \mathcal{R}(x, w) = 1 \) [12].
- ** Succinctness:** proof size and verification time are \( o(|\mathcal{R}|) \). When soundness is computational, the protocol is called an argument, and referred to as a (z)ero-(k)nowledge (S)uccinct (N)on-interactive (AR)gument of (K)nowledge (SNARK).

**R1CS format** Different zk-SNARKs represent the relation \( \mathcal{R} \) in different ways: binary arithmetic circuits [53], quadratic arithmetic programs [91, 97], AIR [14], low-depth circuits [25, 42, 59, 101, 104, 105, 110, 111] and more [18, 19, 56, 91]. We focus on proofs for rank-1 constraint systems (R1CS); these generalize arithmetic circuits, and are the interface to many state-of-the-art proof systems [16, 40, 66, 67].

For R1CS, \( x \) is represented as \( x \in \mathbb{F}^\ell \), \( w \) is represented as \( w \in \mathbb{F}^{m-\ell} \), and the relation itself is defined by three matrices:

\[
A, B, C \in \mathbb{F}^{n \times m}
\]

such that \((Aa) \circ (Ba) = Ca \) where \( a := x \oplus w \) and \( \circ \) is the Hadamard (element-wise) product. This \( a \) is said to be a satisfying assignment for the R1CS relation \((n, \ell, m, A, B, C) \). The matrices are assumed to have only \( \Theta(n) \) non-zero entries, and \( |\mathcal{R}| \) is defined to be \( n \).

The R1CS formalism introduces an additional challenge: each relation \( \mathcal{R} \) must be compiled to an R1CS relation \( R := (n, \ell, m, A, B, C) \). Additionally, the compiler produces a witness extension procedure Extend that maps any satisfying \((x, w) \in \mathcal{R} \) to a satisfying assignment \( a \in \mathbb{F}^m \) for \( R \). Thus, when writing a proof, the prover \( \mathcal{P} \) operates in two steps:

1. \( \mathcal{P} \) extends the pair \((x, w) \) to a satisfying assignment \( a \in \mathbb{F}^m \) for the R1CS relation \( R \),

---

\(^1\)This defines MPC that is computationally—rather than information-theoretically—secure. Computational MPC is more efficient.
The first step is relation-specific, non-cryptographic, and typically inexpensive. The second step is relation-generic, cryptographic.

4 Collaborative zk-SNARKs

A non-interactive collaborative proof for $N$ provers and a relation $R(x, w_1, \ldots, w_N)$ is a tuple $(\text{Setup}, \Pi, \text{Verify})$:
- $\text{Setup}(1^\lambda, R) \rightarrow pp$: setup public parameters.
- $\Pi(pp, x, w_1, \ldots, w_N) \rightarrow \pi$: a protocol for $N$ provers with private inputs $w_1$ through $w_N$ that produces a proof.
- $\text{Verify}(pp, x, \pi) \rightarrow \{0, 1\}$: check the proof.

This primitive is non-interactive in the verification phase: the verifier does not send messages to the provers. Informally, a collaborative proof is secure if it has the following properties:

- **Completeness:** Honest provers with a valid witness produce a valid proof.
- **$t$-zero-knowledge:** Up to $t$ colluding dishonest provers learn nothing about the witnesses of other provers through $\Pi$, other than the validity of the entire witness.
- **Succinctness:** Proof size and verification are $o(|R|)$.
- **Knowledge soundness:** Only proofs that “know” a valid witness can build a valid proof. That is, a valid witness can be efficiently “extracted” given “rewind” access to provers that can build a valid proof.

Definition 1 states security formally, building on prior zk-SNARK security definitions [17]. Since soundness only holds against efficient provers, the protocol is a non-interactive collaborative argument. Regardless, we (informally) refer to it as a “proof” elsewhere in this work.

Importantly, some of the properties in Definition 1 differ from their zk-SNARK analogs. The single-prover definition of zero knowledge requires that the proof $\pi$ for any $(x, w) \in R$
can be simulated without \( w \). The restriction to valid witnesses is acceptable because an honest prover would never write a proof with an invalid witness. However, for collaborative proofs, each prover controls only a part of the witness, so no single prover can guarantee that \((x, w) \in \mathcal{R}\). Moreover, prover \( i \) can choose \( w_i \) arbitrarily, and then learn if the resulting combined witness \( w \) satisfies \((x, w) \in \mathcal{R}\). This is unavoidable, and reveals to prover \( i \) some information about the witnesses held by the other provers. We model this by explicitly providing the validity of the witness, denoted by \( b \leftarrow \mathcal{R}(x, \bar{w}) \in \{0, 1\} \), to the simulator.

In sum, the zero-knowledge property applies to all witnesses, for up to \( t \) malicious provers, and guarantees that only the validity of the witness is revealed.

Knowledge soundness has two subtleties. First, the extractor can program the oracle accessed by all provers participating in \( \Pi \). Second, knowledge soundness establishes only that the provers have distributed knowledge of the \( N \) witnesses. That is, it proves that if they pooled their information, they could determine the \( N \) witnesses.\(^2\) It does not prove that \( \mathcal{P}_i \) knows \( w_i \). This limitation is intrinsic to the non-interactive syntax of the proof, which obscures from \( \mathcal{P}' \) the number of provers participating in the protocol. In some cases, the restriction might be partially circumvented by having \( \mathcal{R} \) itself establish knowledge of secrets believed (for reasons external to the proof) to be in the custody of different entities.

**Theorem 1.** If \((\text{Setup, Prove, Verify}) \) is a zk-SNARK, and \( \Pi \) is an MPC for Prove that is secure-with-abort against up to \( t \) corruptions, then \((\text{Setup, Prove, Verify}) \) is a collaborative zk-SNARK secure against \( t \) malicious provers.

**Proof sketch** Completeness follows from the completeness of the zk-SNARK and the correctness of \( \Pi \) as an MPC. Knowledge soundness follows directly from the knowledge soundness of the zk-SNARK: from the perspective of an extractor, properties that hold for a malicious prover also hold for a malicious collection of provers. Succinctness follows from the succinctness of the zk-SNARK. Security-with-abort is sufficient for \( \Pi \) because the collaborative proof completeness definition does not cover malicious provers.

Zero-knowledge follows from the zero-knowledge of the zk-SNARK and the security of \( \Pi \). In the case that \( b = 1 \) (the witness is valid), then the SNARK’s zero-knowledge implies that \( \pi \) can be simulated from \( x \). Then, the security of \( \Pi \) implies that the adversary’s view can be simulated from \( \pi \) and the witnesses of the corrupted provers. If \( b = 0 \) (the witness is invalid), then the security of \( \Pi \) implies that adversary’s view can directly be simulated from the witnesses of corrupted provers and \( \perp \).

**Collaborative proofs for R1CS** In this work, we consider non-interactive collaborative arguments for secret-shared R1CS witnesses. That is, we consider an R1CS instance \( x \)

\( \mathcal{R} \leftarrow \text{Compiler} \rightarrow R \)

\( \bar{w} \leftarrow \Pi_{\text{ext}} \rightarrow [w] \)

\( x \leftarrow \Pi_{\text{R1CS}} \rightarrow \pi \)

Figure 2: The role of compilation and extension when using a collaborative zk-SNARK for R1CS. First, the relation \( \mathcal{R} \) is compiled to an R1CS relation \( R \) and a protocol \( \Pi_{\text{ext}} \). Given \((x, \bar{w}) \in \mathcal{R}\), the provers engage in \( \Pi_{\text{ext}} \) to yield \( x \) and \([w]\) such that \( x \parallel w \) satisfies \( R \).

and provers \( \mathcal{P}_1, \ldots, \mathcal{P}_N \) which hold shares \([w]\) of an R1CS witness \( w \), such that \( x \parallel w \) is a satisfying assignment for an R1CS relation \( R \). As in a single prover SNARK for R1CS, this approach requires a compiler to R1CS which also produces an extension protocol that maps the input and witness for \( \mathcal{R} \) to shares of the complete R1CS witness. Figure 2 shows the use of the compiler and extension protocol.

### 4.1 PA-MPC from collaborative proofs

**Publicly-auditable MPC** [7] (PA-MPC) extends MPC with a publicly-verifiable proof that the output of the MPC is correct with respect to commitments to each party’s inputs. In Appendix D we define PA-MPC formally, and show that collaborative proofs can be used to construct a PA-MPC protocol. Specifically, we show how a suitable MPC protocol, a commitment scheme, and a collaborative proof give a PA-MPC protocol. The derived PA-MPC protocol outputs proofs whose size is about the same as those output by the collaborative proof. In particular, if the collaborative proof scheme outputs constant size proofs, then so does the derived PA-MPC protocol. This is a dramatic improvement over the classic construction [7] (where proof size and verification time are linear in the size of the evaluated circuit) and it is competitive with recent [103] and concurrent [75] work.

### 5 Design: efficient MPC for SNARK provers

#### 5.1 Review: two algebraic MPC protocols

We extend two secure MPC protocols for arithmetic circuit evaluation. The first, GSZ (§5.1), is secure if a majority of parties are honest. The second, SPDZ (§5.1), is secure if at least one party is honest. We choose these two protocols because they are simple, efficient, and extend naturally to elliptic curve computations (§5.2). SPDZ [46] is a dishonest majority protocol. GSZ [64] is representative of a new generation of state-of-the-art honest majority protocols [22, 27, 30, 31, 62, 64].

We first describe important features of both protocols. Then in Section 5.2, we discuss how to generalize these protocols
to circuits over elliptic curve points and their scalars.

**Honest majority MPC using GSZ**  The first protocol, GSZ was developed in [64], and based on [27, 45, 55]. The protocol builds on Shamir secret-sharing. It evaluates an arithmetic circuit one gate at a time. Shamir shares can be added, shifted by public values, and scaled by public values using a single field operation and no communication between parties. Shares can be multiplied with an interactive protocol [45]. This protocol is not secure against malicious behavior [55], so all multiplication triples \((x, y, z)\) corresponding to wire values of multiplication gates in the circuit, must be saved and checked for consistency, namely \(xy = z\), before any share is opened. The multiplication checking protocol requires communication sub-linear in the number of triples [64].

**Dishonest majority MPC using SPDZ**  The SPDZ protocol [46] uses additive sharing. Similar to Shamir shares, additive shares can be added, shifted by public values, or scaled by public values with at most one field operation and with no communication between parties. Multiplying shares requires an interactive protocol. SPDZ does not need to check multiplication triples because it maintains a shared message authentication code (MAC) \(\Delta x\) for each shared value \(x\) \((\Delta \in \mathbb{F}, \text{the MAC key, is unknown to all})\). Together, a share and its MAC form an authenticated share. The opening protocol for authenticated shares assures the integrity of the revealed value. However, since each operation on an authenticated share changes the MAC, the operation takes twice as long.

Both GSZ and SPDZ are typically split into a preprocessing phase (which can run before the circuit inputs are known) and an online phase (which runs after the inputs are known). We briefly discuss the relative costs of these phases in Section 7.

### 5.2 MPC using elliptic curve secret sharing

While SPDZ and GSZ natively support arithmetic over a finite field, the zk-SNARKs we consider perform many elliptic curve group operations, such as curve additions and multiplications by scalars.

Since elliptic curve operations decompose into operations over the curve’s base field, there is an obvious way to perform them in an MPC. Unfortunately, this naive approach is very expensive: curve additions require tens of field operations, and scalar products require thousands. Furthermore, curve additions require multiple field multiplications, so even curve additions require rounds of communication.

Fortunately there is a better way: both the SPDZ and GSZ protocols can be adapted to operate directly on elliptic curves, where all secret sharing is done over the elliptic curve.

#### An elliptic curve circuit

(Fig. 3) is a generalization of an arithmetic circuit over \(\mathbb{F}_q\), adding support for computations over an elliptic curve group \(\mathbb{G}\), with order \(q\). In an elliptic curve circuit, wires can hold elliptic curve points from \(\mathbb{G}\) or scalars from \(\mathbb{F}_q\) and the gate set is extended to include curve addition and the multiplication of curve points by scalars. The multiplication of two curve points is prohibited.

![Figure 3: An elliptic curve circuit. The sub-circuit drawn with solid lines is an arithmetic circuit; the dashed extension includes curve operations](image)

Prior work shows that SPDZ generalizes to elliptic curve circuits [100]; we observe that GSZ does too. To start, observe that Shamir secret sharing can be used to share an elliptic curve point: the \(N\) shares are themselves elliptic curve points, and the secret is reconstructed by a weighted linear combination of sufficiently many shares. In Appendix A, we show how to apply GSZ’s sub-protocols—those for computing [45] and checking [64] field products—to scalar-curve products.\(^3\)

### 5.3 Commitments to shared polynomials

SNARK provers have many components. In this section and the next we show efficient MPC protocols for the core building blocks needed to generate a SNARK proof. We begin with an MPC protocol for a polynomial commitment scheme.

**Polynomial commitments** are a primitive—and bottleneck—in many recent SNARKs. SNARKs are built from polynomial commitments using the polynomial interactive oracle proof (P-IOP) paradigm [36, 40]. In a P-IOP, \(P\)’s messages to \(V\) are oracles to polynomials which \(V\) accesses by querying for the outputs at various inputs. The polynomials are encodings of the SNARK witness, and \(V\) checks the SNARK relation using the results of its queries. To compile a P-IOP to a standard protocol, a polynomial commitment scheme is used to implement the oracles. For each oracle, \(P\) sends \(P\) a commitment to the underlying polynomial. For each query, \(V\) sends the input, \(P\) sends the output, and \(P\) and \(V\) engage in an evaluation protocol to validate the claimed evaluation. Informally, the evaluation protocol forces \(P\) to answer queries consistently with the underlying polynomial, so the compiled protocol is equivalent to the P-IOP.

A polynomial commit scheme is a tuple \((\text{PC.Setup, PC.Commit, PC.Eval})\) where

\(^3\)Generalizing SPDZ and GSZ to elliptic curves suffices for implementing SNARK provers, but the protocols can do even more: they apply to any bilinear operation, e.g., pairings and polynomial products.
PC.Setup(d) \to pp: prepares public parameters supporting polynomials of degree up to d.
PC.Commit(pp, f, r) \to c: computes a commitment c to an input polynomial f, using randomness r.
PC.Eval(pp, x, y, c, f, r) \to \{0, 1\}: a protocol where P convinces V that c commits to an f such that \( f(x) = y \). Only \( P \) knows f and r.

Prior work \cite{36, 40} gives the security properties that a polynomial commitment must satisfy for the compiled P-IOP to be secure.

For a collaborative proof, the polynomial f is shared since it encodes the witness. However, the commitment to f, and various polynomial evaluations that are part of the proof are public. Hence, our goal is to generalize PC.Commit and PC.Eval to efficiently operate on a shared polynomial.

**KZG commitments** The KZG polynomial commitment \cite{76} is among the most efficient known schemes and in particular has the smallest evaluation proofs. It is defined by four algorithms:

- **PC.Setup(d):** sample \( \alpha \in F \); output 
  \[ pp \leftarrow (\alpha \cdot g_1, \cdots, \alpha^d \cdot g_1, \alpha \cdot g_2) \]

- **PC.Commit(pp, f):** output \( c = f(\alpha) \cdot g_1 \), computed as \( c \leftarrow \sum_{i=0}^d f_i \cdot (\alpha^i \cdot g_1) \)

- **PC.Prove(pp, f, x):** compute the remainder and quotient
  \[ q(X), r(X) \leftarrow (f(X) - f(x))/(X - x); \]
  output \( \pi = q(\alpha) \cdot g_1 \), computed as \( \pi \leftarrow \sum_{i=0}^d q_i \cdot (\alpha^i \cdot g_1) \); the remainder \( r(X) \) must be zero.

- **PC.Check(pp, c, x, y, \pi):** accept if the following pairing equation holds:
  \[ e(\pi, \alpha \cdot g_2 - x \cdot g_2) = e(c - y \cdot g_1, g_2) \]

Here, PC.Setup samples a trapdoor \( \alpha \), which must be discarded for security; for this reason the procedure is called a trusted setup. Together, PC.Prove and PC.Check form the evaluation protocol for the scheme. At a high level, the scheme’s security follows from the fact that \( X - x \) divides \( f(X) - y \) if and only if \( f(x) = y \). See \cite{40, 52, 76} for security proofs and extensions of the scheme.

**Generalizing KZG to multiple provers** We now adapt KZG to polynomials shared among multiple provers.

How can PC.Commit and PC.Prove be implemented for a secret shared f? First, let \([f] \) be represented as a list of shared coefficients: \([f_0 \], [f_1\], \ldots, [f_d \)]. Let \( f_0^{(i)}, f_1^{(i)}, \ldots, f_d^{(i)} \) denote \( P_i \)'s shares of those coefficients, defining a polynomial \( f^{(i)} \), known only to \( P_i \). When using a linear sharing scheme \cite{11} (including the additive and Shamir schemes we use) the provers can correctly create shares of commitments to (respectively, proofs for) \( f \) by locally committing to (respectively, proving evaluations for) their individual \( f^{(i)} \) and interpreting the results as shares of the desired commitment (respectively, proof). More explicitly, Figure 4 holds our multi-party commitment and proving protocols: PC.Commit’ and PC.Prove’.

Let us show that PC.Prove’ is correct; Appendix C shows security and that PC.Commit’ is correct and secure. We argue generically over a linear secret sharing scheme \cite{11}. A linear secret sharing scheme has linear reconstruction: for any sufficient subset of parties there is a linear combination of their shares that reconstructs the value. Let \( \bar{f} \in \mathbb{F}^N \) encode this linear combination, that is, for any value \( v \) with shares \( v^{(i)} \), \( v = \sum_{i=1}^N \gamma_i v^{(i)} \).

We begin with a lemma about the division in PC.Prove’ from Figure 4b.

**Lemma 2.** For any shared polynomial \([f(X)]\) and public polynomial \(d(X)\), if each party \( P_i \) computes \( q^{(i)}, r^{(i)} \leftarrow \frac{f^{(i)}}{d} \), then \( q^{(i)} \) and \( r^{(i)} \) are shares of \( q \) and \( r \) such that \( (q, r) = \frac{f}{d} \).

*Proof.* Per each party’s euclidean division, we have that \( q^{(i)} \cdot d + r^{(i)} = f^{(i)} \). Summing the equations—scaled by the \( \gamma_i \)—yields:

\[
\left( \sum_{i=1}^N \gamma_i q^{(i)} \right) \cdot d + \left( \sum_{i=1}^N \gamma_i r^{(i)} \right) = \sum_{i=1}^N \gamma_i f^{(i)} = f
\]

Since each \( r^{(i)} \) has degree less than \( d \), their scaled sum does too, so \( \sum_{i=1}^N \gamma_i q^{(i)} \) and \( \sum_{i=1}^N \gamma_i r^{(i)} \) are a quotient-remainder pair for \( f \) divided by \( d \). Since polynomial division is unique, these sums must be equal to \( q \) and \( r \) respectively.

**Claim 3.** PC.Prove’ is correct, that is PC.Prove’(pp, f, x) = PC.Prove’(pp, [f,], x).

*Proof.* It suffices to show that \( \sum_{i=1}^N \gamma_i \pi^{(i)} \) is equal to PC.Prove’(pp, f, x). By Lemma 2, we have that for the \( q^{(i)} \) that each \( P_i \) computes, \( \sum_{i=1}^N \gamma_i q^{(i)} = q \). Then we have that \( \sum_{i=1}^N \gamma_i \pi^{(i)} \) is just \( \sum_{i=1}^N \sum_{j=1}^d \gamma_i q^{(i)} \cdot \alpha^j \cdot g_1 \). Swapping the order of the sum and simplifying yields \( \sum_{j=1}^d \sum_{i=1}^N \gamma_i q^{(i)} \cdot \alpha^j \cdot g_1 = \sum_{j=1}^d \gamma_i q^{(i)} (\alpha) \cdot g_1 = \gamma(q(\alpha) \cdot g_1 = PC.Prove'(pp, f, x).

These simple and efficient protocols for commitments to shared polynomials are encouraging: the bottleneck in many SNARKs incurs little overhead in the multi-prover setting. In the next section we’ll see that many other key SNARK operations have similarly efficient protocols.

### 5.4 Optimizing SNARK provers for MPC

In this subsection we discuss techniques for optimizing four SNARKs: Groth16 \cite{66}, Marlin \cite{40}, Plonk \cite{53}, and Fractal \cite{41} to multiple-provers. Marlin and Plonk are defined
procedure PC.Commit’(pp, [f]) → c:
  each party i:
    c(i) ← \sum_{j=0}^{d} f(j) \cdot (\alpha^i \cdot g_i)
  interpret c(i) as a share of [c]
  output [c]

(a) Commitments

procedure PC.Prove’(pp, [f], x) → π:
  each party i:
    q(i), r(i) ← f / (X - x)  (compute quotient and remainder)
    \pi(i) ← \sum_{j=0}^{d} q(j) \cdot (\alpha^i \cdot g_i)
  interpret \pi(i) as a share of [π]
  output [π]

(b) Evaluation proofs

Figure 4: KZG [76] procedures for shared polynomials

<table>
<thead>
<tr>
<th>SNARK</th>
<th>FFT</th>
<th>MSM</th>
<th>Poly. Division</th>
<th>Vector Commit.</th>
<th>Sum-Check</th>
<th>Product-Check</th>
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<tr>
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<td>✓</td>
</tr>
</tbody>
</table>

Figure 5: Key components of SNARKs

generically over a polynomial commitment scheme. We instantiate them with the KZG-derived scheme from [40], adapted per the previous subsection. Here, we focus on other potential bottlenecks, especially those common to multiple provers. Figure 5 shows the operations we consider and the SNARKs they appear in.

**Traditional bottlenecks: FFTs and MSMs** Fast fourier transforms (FFTs) and multi-scalar multiplications (MSMs) are bottlenecks for most SNARK provers. MSMs are computations of the form \( \sum \gamma_j \cdot g_i \) for scalars \( \gamma_j \) and curve points \( g_i \); informally, provers use them to “commit” to the scalars \( \gamma_j \). The fourier transform computes a matrix-vector product \( M \vec{x} \), where \( M \) is derived from a root of unity \( \omega \). Provers use FFTs to convert between the coefficient- and evaluation-forms of polynomials.

Both the FFT and MSM are linear operators in the scalars \( \gamma_j \) and the vector \( \vec{x} \) respectively. Thus, if other inputs (the curve points \( g_i \) and the root of unity \( \omega \)) are public information (as they are in SNARKs) and a linear sharing scheme is used, each operator can be applied directly to each party’s input shares, yielding shares of the output. Thus, FFTs and MSMs in a multi-prover SNARK reduce to performing the operation share-wise, with no communication required. Share-wise computation is cheap in an MPC: identical to the base cost for Shamir shares and twice that for authenticated shares.

We now discuss other operations that are not traditional bottlenecks, but might become bottlenecks in the multi-prover setting.

**Polynomial division** All provers do large polynomial divisions: a complex and non-linear operation that might have high cost in an MPC. However, most SNARKs (including those we consider) only divide by witness-independent—and therefore public—polynomials. The KZG scheme is one example—the divisor is \( X - x \), for public \( x \). Elsewhere, provers divide by far larger polynomials, but these polynomials are still public, admitting—as shown Section 5.3—a simple protocol.

**Sum-and product-checks** Two of the proof systems listed, Plonk and Marlin, combine the aforementioned components with two different but related proof primitives: Marlin relies on SumCheck while Plonk relies on ProdCheck. To explain these primitives, let \( e_f \) be a commitment to a polynomial \( f \in \mathbb{F}_p[X] \), let \( \Omega := \{1, \omega, \omega^2, \ldots, \omega^{n-1}\} \) be a subset of \( \mathbb{F}_p \), and let \( c \in \mathbb{F}_p \). Then SumCheck and ProdCheck are sub-SNARKs for the relations \( \mathcal{R}_+ \) and \( \mathcal{R}_\times \) respectively, where

- \( \mathcal{R}_+ (e_f, c; f) \) holds when \( \sum_{x \in \Omega} f(x) = c \) and \( e_f \) is a commitment to \( f \);
- \( \mathcal{R}_\times (e_f, c; f) \) holds when \( \prod_{x \in \Omega} f(x) = c \) and \( e_f \) is a commitment to \( f \).

The SumCheck SNARK is described in [40] while the ProdCheck SNARK is described in [53]. An MPC implementation of Plonk and Marlin requires proving SumCheck and ProdCheck for shared polynomials.

Adapting the SumCheck prover to a shared \([f]\) is fairly direct. In SumCheck, the prover commits to \( f'(X) \leftarrow (f(X) - c/|\Omega|)/X \), and the verifier checks that \( f = X \cdot f' + c/|\Omega| \). The coefficients of \( f' \) are just coefficients of \( f \), so the only work happens inside the polynomial commitment.

ProdCheck poses a greater challenge. Its prover first computes all partial products

\[
\tau_i := \prod_{j=0}^{i} f(\omega^j) \quad \text{for } i = 0, \ldots, n - 1. \tag{1}
\]

At first glance, one might expect that that an MPC for the ProdCheck prover would be very costly. Indeed, the
6 Implementation

We implement collaborative proofs based on Groth16, Marlin, and Plonk. We omit Fractal because of the tradeoff between proving time and proof size discussed in the previous section. Our starting point is arkworks [48]: a collection of Rust libraries for implementing cryptographic proofs, used in a few recent works [37, 38, 40]. It includes interfaces and implementations for finite fields, pairing-friendly curves, and polynomial commitments. It also includes SNARKs implemented generically over their primitives; e.g., Groth16 and Marlin implementations applicable to any pairing e.

Collaborative SNARKs through MPC lifting Our implementation pursues three goals. First, to support different secret-sharing schemes, proof systems, and MPC protocols. Second, to achieve proving time concretely competitive with state-of-the-art conventional SNARKs. Third, to focus on the multi-party aspects of the implementation and optimization.

Given these considerations, we adopt a design that lifts an existing SNARK prover implementation into a multi-prover protocol by instantiating the primitives that the prover uses with secure multi-party protocols. At a high level, we:

1. define interfaces for field and curve sharing schemes
2. provide implementations based on the SPDZ and GSZ protocols
3. define wrapper types for field and curve elements which can be either public or shared,
4. implement arkworks field and curve interfaces for the wrapper types,
5. instantiate arkworks-based SNARK provers with the wrapper types (yielding multi-prover protocols!), and
6. optimize the protocols.

Building multi-party primitives We first define interfaces for shares of both field elements and groups of prime order (like pairing-friendly elliptic curves). From here on, we discuss only field support; group support is similar. The field share interface includes functions for scaling by public values, shifting by public values, adding shares, multiplying shares, and opening shares. We implement three sharing schemes generically over their underlying field. They use a network library to communicate and access protocol metadata (e.g., their party number). Figure 6 shows the definitions, and how to shift each type of share by a public value. For GSZ shares, all parties shift their value; for additive shares, only one party does; for authenticated shares, the base share is updated by a single party, but the MAC is updated by everyone using their share of the MAC key.

To represent values in an MPC, we introduce a union type MpcField for either public or shared field elements. MpcField is generic over both the field and the sharing scheme. We implement the Field interface for MpcField, allowing it to be used in any computation that expects to operate on field elements. Figure 7 sketches the code for this. By instantiating the function foo (perhaps not intended as an

4Identical to Shamir shares, save the verification that runs before any share is opened.
with BLS 12-377’s scalar field yields a local computation. Instantiating
the SNARK prover implementations. Our primary changes to the
transform. These new methods and their default implementa-
tion, multi-scalar multiplication, and the Fourier
implementation) for batch multiplication, polynomial division,
relevant algebraic interface, and provide an MPC-optimized
optimization, this entails a sequence of calls to the
Field
optimized protocol here (i.e., the one from Section 5.4), we
situation is available at
arkworks
Plonk implementation (≈2000 lines), an
arkworks Plonk implementation (≈1200 lines), and tests
and benchmarking scripts (≈3000 lines). We also slightly
modify a number of arkworks libraries. Our implementa-
tion is available at https://github.com/alex-ozdemir/
multiprover-snark.

3There are a few exceptions. Some basic operations (polynomial divi-
sion, multi-scalar multiplication) branch on shared data—we provide MPC-
optimized replacements anyway, sidestepping the issue. Some provers also
check intermediate (shared) proof material for consistency—we defer all
checks until the final proof is revealed.

MPC) with an MpcField based on GszShare, we execute it
as an honest-majority secure multi-party computation.

Lifting SNARKs into MPCs and optimizing Similarly, we
can obtain collaborative proofs from prover implementations
that are generic over the arkworks interfaces. For Marlin
and Groth16 we can use implementations from arkworks; Plonk
has no arkworks-based implementation yet, so we build one.

The resulting protocols (with a few small tweaks) are cor-
rect but perform poorly. The problem is that some operations
are not optimized for an MPC. One example is computing the
partial products of a list (discussed in Section 5.4). Without
optimization, this entails a sequence of calls to the Field
interface’s multiplication method; each call requires a round
of communication, which is expensive. To allow for an MPC-
optimized protocol here (i.e., the one from Section 5.4), we
add a partial product method to the Field interface. The de-
fault implementation is based on serial multiplication, but we
also provide an overriding implementation for a sequence of
shared values. We repeat this pattern (find a problematic oper-
ation, identify a better MPC protocol, add the operation to the
relevant algebraic interface, and provide an MPC-optimized
implementation) for batch multiplication, polynomial division,
batch inversion, multi-scalar multiplication, and the Fourier
transform. These new methods and their default implementa-
tions are our primary changes to the arkworks algebraic
interfaces. Calls to these methods are our primary changes to
the SNARK prover implementations.

Limitations Our approach is effective for collaborative
proofs, but cannot be applied to all computations based on
arkworks. Critically, it does not support programs which
branch on shared data or access data at shared locations. The
limitation is fundamental: the MPC protocols we use support
circuit evaluation, not RAM machine execution—which
the Rust programming language and arkworks target. While
a program might perform circuit-incompatible operations,
our approach works because the provers we study do not
do that.3 Techniques for translating RAM programs into cir-
cuits [6, 32, 33, 78, 79, 90, 102] could yield more general
support. Separately, some MPC algorithms [61, 70, 71, 72]
directly support branching programs.

Source code In sum, our implementation comprises a
network library (≈700 lines), interfaces and implementa-
tions of sharing schemes (≈3000 lines), MPC implementa-
tions of arkworks algebraic interfaces (≈2000 lines), an
arkworks Plonk implementation (≈1200 lines), and tests
and benchmarking scripts (≈3000 lines). We also slightly
modify a number of arkworks libraries. Our implementa-
tion is available at https://github.com/alex-ozdemir/
multiprover-snark.
Figure 8: Proving time for varying numbers of parties and protocols, over a 3 gigabit link. $N_{PC}$ denotes a collaboration among $N$ provers. A dishonest majority (SPDZ) protocol is private against $<N$ corrupt provers. An honest majority (GSZ) protocol is private against $<N/2$ corrupt provers. The baseline is a single prover, using the (non-collaborative) zk-SNARK.

7 Evaluation

We evaluate our collaborative SNARKs against conventional SNARKs. Our metric is single-threaded prover runtime; for multiple parties, this is their average runtime. For all configurations, we report the median of three trials. We vary these parameters:

- collaborative proof: based on Groth16, Marlin, or Plonk
- number of rank-1 constraints: from 2 to $2^{20}$. We discuss the number of constraints needed by one application at the end of the section.
- MPC algorithm: GSZ (honest majority) or SPDZ (dishonest majority)
- link capacity: from 1Mb/s to 3Gb/s

As we will see, collaborative proofs are concretely fast: over gigabit links, multi-prover protocols are nearly as fast as a single prover.

Preprocessing We do not evaluate pre-processing. Omitting pre-processing is acceptable for two reasons. First, pre-processing can be performed when spare compute and bandwidth are available. Second, pre-processing is concretely faster than proof generation. The most expensive pre-processing required by either of our base protocols (SPDZ and GSZ) is the multiplication triple generation phase of SPDZ. Yet, with recent protocols [8, 74, 95], this requires only a few microseconds per triple (e.g., [8] generates one million triples in $\approx 5s$). Meanwhile, even the fastest prover (Groth16)—which uses one triple per constraint—requires over 200 $\mu$s per constraint.

High capacity links Our first experiment measures proving time for varying MPC algorithms, numbers of constraints, and proof systems over a high-capacity link. We consider 2 and 3-party proofs, omitting honest majority 2-party proofs, which provide no security. We consider constraints systems of size 2 through $2^{20}$, and collaborative proofs based on Groth16, Marlin, and Plonk. For each proof system, we compare with the single-prover time. Google Cloud Platform (GCP) n2-standard-2 machines are our testbed; these have 1 hardware core (Cascade lake), 8GB of memory, and 3Gb/s links. We run Debian 10 with hyperthreading disabled.

Figure 8 shows the results. Missing data points indicate memory exhaustion. For small constraint systems the multi-prover protocols are slower than a single prover, because network round-trips (which take milliseconds) dominate. However, the number of round-trips is sub-linear in constraints, and for large constraint systems ($\gtrsim 2^{10}$) the protocols perform quite well. The honest majority protocol (3PC GSZ) runs in essentially the same time as a single prover; the dishonest majority protocols (SPDZ) run in essentially twice the time. This is because the computational costs of the protocols dominate the bandwidth costs. SPDZ is twice as slow as a single prover because it applies the computation to the data and its MAC: duplicating the computational work.

Varying party count To show how the different MPC implementations scale with party count, we write Groth16 proofs for $2^{10}$ constraints, with varying numbers of parties. We use the same testbed as the previous experiment, and report slowdown: the time to write the collaborative proof, divided by the time of a single prover. Figure 9 shows the results. As the

6SPDZ is constant-round, while GSZ requires a logarithmic number of rounds to check multiplication triples.
number of parties increases, the performance of SPDZ slowly degrades, while the performance of GSZ degrades faster. The crucial difference is in the number of rounds of inter-prover communication. For SPDZ, there is only a constant number of rounds of communication, since the Groth16 circuit has low multiplicative depth. However, the multiplication checking protocol of GSZ requires a number of rounds logarithmic in the number of constraints. This means that for large numbers of parties, one should use a round-efficient protocol such as SPDZ instead of GSZ, even if one assumes a majority of provers are honest.

Varying link capacity While gigabit links are common in LANs, within datacenters, and between data centers, lower capacity links are common elsewhere. Our third experiment measures the relationship between proving time and link capacity. We cohost two provers on an Arch Linux (kernel 5.12.15) machine with an AMD Ryzen-2700 CPU and 32GB of RAM. We use the Mahimahi network shell suite [87] to emulate low-capacity links. We report slowdown for SPDZ evaluating varying provers on $2^{10}$ constraints with varying network capacity.

Figure 9 shows the results. At low bandwidths, data transfer dominates proving time, so slowdown grows linearly with the reciprocal of link capacity. At higher bandwidths, computational costs dominate, so slowdown approaches 2 (recall that SPDZ duplicates the computation). Plonk’s slowdown is higher than the other proof systems. This is due to the batch multiplications and divisions in its ProofCheck; while our optimization (Section 5.4) keeps round complexity constant, the communication volume is still substantial.

Price We also estimate the price of writing collaborative proofs. We consider 2-prover Groth16 proofs using the SPDZ protocol. Communication is dominated by 256-bit field elements: $4n$ are transferred. Figure 8 shows that the computation requires $\approx 488 \mu s$ per constraint. Currently, GCP e2-standard-2 VMs cost $0.067/hr$ [93], and cross-continent traffic between VMs costs $0.08/GB$ [92]. At these rates, Figure 11 summarizes net costs for proofs of varying size and type. Collaborative proofs (2PC) have quadruple the compute costs of conventional ones (1PC), because they involve twice the wall-clock time and two machines. Ultimately, net costs are tiny: million-constraint collaborative proofs cost only 3 cents.

Proofs about net assets In Section 2 we discussed a setting where a client of multiple banks wants a publicly verifiable proof that the net of its debits and credits across all its banks exceeds a threshold $T$. Here, we discuss the number of constraints required for such a proof.

To use our collaborative proofs, that claim must be written as a rank-1 constraint system (R1CS). How does the number of constraints depend on the number of transactions (credits and debits)? The constraints must (a) open the bank’s commitments to all the customer’s transactions, (b) check that each transaction is well-formed, (c) sum their values, and (d) compare that sum to $T$.

Let there be $n$ transactions split among $N$ banks. Each transaction is a 64-bit signed integer. Let bank $i$ have $n_i$ transactions in a Merkle tree of $n_i + 1$ leaves—the extra leaf is fully random to make the commitment hiding. Opening bank $i$’s commitment uses $n_i$ hash evaluations: one for each internal

<table>
<thead>
<tr>
<th>Type</th>
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<th>Total Cost</th>
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<tbody>
<tr>
<td>2PC</td>
<td>$2^{10}$</td>
<td>$1.9 \times 10^{-5}$</td>
<td>$1.0 \times 10^{-5}$</td>
<td>$2.9 \times 10^{-5}$</td>
</tr>
<tr>
<td>1PC</td>
<td>$2^{10}$</td>
<td>$4.6 \times 10^{-6}$</td>
<td>0</td>
<td>$4.6 \times 10^{-6}$</td>
</tr>
<tr>
<td>2PC</td>
<td>$2^{20}$</td>
<td>0.019</td>
<td>0.011</td>
<td>0.030</td>
</tr>
<tr>
<td>1PC</td>
<td>$2^{20}$</td>
<td>0.0048</td>
<td>0</td>
<td>0.0048</td>
</tr>
</tbody>
</table>

Figure 11: Price estimates (in US$) for collaborative (2PC) and conventional (1PC) proofs. The “size” column refers to the number of constraints.
node in the binary Merkle tree. Thus, opening all the commitments uses \( n = \sum n_i \) hash evaluations. The Poseidon [65] hash function requires \( \approx 300 \) constraints per evaluation.

Furthermore, checking that each committed transaction has a value representable as a 64-bit signed integer requires \( \approx 64 \) constraints. In R1CS, computing the sum requires essentially no constraints, and comparing with \( T \) requires a small constant number of constraints: \( \approx 64 + \log_2 n \). In sum, a proof about \( n \) transactions requires \( \approx 364n \) constraints. Thus, \( 2^{20} \) constraints suffices for proofs about \( \approx 2900 \) customer transactions. This means that three banks could write a proof about \( \approx 2900 \) transactions in \( \approx 165 \) s. This time could be further reduced with parallelism (§9).

8 Related work

The most closely related work is [75] (building on [103]), which is concurrent and independent. Their focus is on en- actively shared, where \( \Omega \) might try to reduce it. However, we show a partial barrier: \( \Omega(n) \) communication is required if the R1CS witness is addi- tively shared, where \( n \) is the number of rank-1 constraints. We show this, in Appendix B, through a reduction from disjoint: a known communication-hard problem.

Nonetheless, constant-factor reductions in communication are possible. For example, [22, 30, 31, 62] communicate less than GSZ, with varying assumptions about the numbers of honest and dishonest parties. We leave the empirical evaluation of these protocols in our setting to future work.

Parallelization For simplicity, we evaluate single-threaded provers. However, our techniques should yield multi-threaded multi-prover protocols with similarly good performance compared to a multi-threaded single prover. In most provers, parallelism accelerates Fourier transforms and multi-scalar multiplications. Our protocols perform these operations locally, so the same benefits should be achievable. Similarly, distributed computing [109] and hardware acceleration [112] could be used. However, parallelism would be less helpful over low-capacity links, where bandwidth becomes the bottleneck (§7).

Latency Our evaluation considered the effect of link capacity—but not latency—on proving-time. All protocols have round-complexity sub-linear in the constraint count, so latency has an insignificant effect as constraint count grows.

Trusted Setup Our implemented collaborative proofs all require the same trusted setup as the zk-SNARK they build on. For those based on Plonk or Marlin that setup is universal (independent of the relation) and updatable (securely re-randomizable by any party). A Fractal-based collaborative zk-SNARK would not require a trusted setup.

Leaking relation membership We emphasize that a collabora- tive proof for \((x, w)\) reveals whether \((x, w)\) is in \( \mathcal{R} \). This is unavoidable: completeness and soundness require it. For some relations, this enables malicious provers to glean information about another prover’s witness. As a simple example, the relation that checks whether two witnesses agree in the first bit lets one prover learn the first bit of another prover’s witness. When using a collaborative proof, one must consider the worst-case leakage from the relation itself.

Significance We’ve shown that collaborative proofs can be constructed with little to no computational overhead compared to single-prover proofs. Communication costs are asymptotically \( \Theta(\lambda n) \), but concretely small over commodity high-capacity links. In sum, most server-side applications that can tolerate the (considerable) cost of a single-prover proof should also be able to tolerate the cost of a collaborative proof. We hope this makes succinct, public verifiability available to more applications—even when the secrets of multiple parties are involved.

9 Discussion

A partial barrier to reducing communication Since commu- nication is a bottleneck over low-capacity networks, one might try to reduce it. However, we show a partial barrier: \( \Omega(n) \) communication is required if the R1CS witness is addi- tively shared, where \( n \) is the number of rank-1 constraints. We...
Mishra, Dev Ohja, and others for developing and maintaining arkworks.

This work was funded by NSF, DARPA, a grant from ONR, the Simons Foundation, and a Google Cloud Platforms grant. Opinions, findings, and conclusions or recommendations expressed in this material are those of the authors and do not necessarily reflect the views of DARPA.

References

2012.


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rather than field elements.

The rest of the protocols are adapted to scalar-group products as follows. DE-LINEARIZATION (Protocol 13, [63]) reduces checking a sequence of multiplication tuples (a Hadamard product); it is adapted to G-HADAMARD-Ck, which reduces checking a sequence of scalar-point multiplication triples to checking a scalar-point inner product.

\[
G\text{-HADAMARD-Ck}(\mathbf{a} \in \mathbb{F}^n, [\mathbf{A}], [\mathbf{B}] \in \mathbb{G}^n)
\]

\[
\text{Shows: } a \circ \mathbf{A} = \mathbf{B}
\]

\[
\alpha \leftarrow \text{COIN}()
\]

\[
\alpha \leftarrow (\alpha')^n_{i=1}
\]

\[
G\text{-IP-Ck}(\alpha \circ [\mathbf{a}], [\mathbf{A}], \alpha \cdot [\mathbf{B}])
\]

G-IP-COMPUTE adapts EXTEND-MULT (Protocol 10, [63]) from computing a field inner product to computing a scalar-point inner product. EXTEND-MULT is itself a generalization of the multiplication and degree reduction protocols of [45].

\[
G\text{-IP-COMPUTE}([\mathbf{a}] \in \mathbb{F}^n, [\mathbf{A}] \in \mathbb{G}^n)
\]

\[
\text{Computes } a \circ \mathbf{A} = B
\]

Everyone computes the inner product on their shares

Interpret the results as \([B_1, X_1, Z_1, A_1, R_1] \leftarrow G\text{-DOUBLE-RAND}()
\]

All parties send \([B-X]_1, Z_1, A_1, R_1\) to Party 1

Party 1 opens \(B-X\) and reshares it as \([B-X]_1, Z_1, A_1, R_1\)

return \([B] \leftarrow [B-X]_1 + [X]_1, Z_1, A_1, R_1\)

G-IP-FOLD adapts EXTEND-COMPRESS (Protocol 12, [63]). The original protocol compresses \(N\) different inner product checks into one. The new one compresses 2 point-scalar inner product checks into one.

\[
G\text{-IP-FOLD}([\mathbf{x}] \in \mathbb{F}^n, [\mathbf{X}] \in \mathbb{G}^n, [\mathbf{P}] \in \mathbb{G}, [\mathbf{Y}] \in \mathbb{F}^n, [\mathbf{Y}] \in \mathbb{G}^n, [\mathbf{Q}] \in \mathbb{G})
\]

\[
\text{Reduces } \mathbf{x} \cdot \mathbf{X} = P \land \mathbf{y} \cdot \mathbf{Y} = Q \text{ to } a \circ \mathbf{A} = R
\]

Interpolate shares of lines \([\mathbf{f}]\) such that

\[
f_i(1) = x_i \text{ and } f_i(2) = y_i \text{ for } i \in [n]
\]

Interpolate shares of lines \([\mathbf{f}]\) such that

\[
F_i(1) = X_i \text{ and } F_i(2) = Y_i \text{ for } i \in [n]
\]

\[
[z] \leftarrow [F(3)]; \quad [Z] \leftarrow [F(3)]
\]

\[S \leftarrow G\text{-IP-COMPUTE}([z], [Z])
\]

Interpolate shares of a parabola \([f_p]\) such that

\[
f_p(1) = P, f_p(2) = Q, \text{ and } f_p(3) = S
\]

\[
\alpha \leftarrow \text{COIN}()
\]

\[
[a] \leftarrow [f(\alpha)]; \quad [A] \leftarrow [F(\alpha)]; \quad [R] \leftarrow [f_p(\alpha)]
\]

return \(([a], [A], [R])\)

Finally, G-IP-Ck adapts MULT-VERIFICATION (Protocol 17, [63]), from a recursive protocol for product checking to a recursive protocol for point-scalar inner product checking. G-PROD-Ck is the adapted base case (Protocol 15, [63]).
B Communication Lower Bound

In this section, we show a $\Omega(n)$ lower bound on the communication needed to compute a collaborative proof when the witness is additively shared among two parties.

Our bound applies to proof systems that are honestly sound: a prover following the proving protocol with an invalid witness produces a valid proof with negligible probability.

Definition 2. A proof $(\text{Setup}, \text{Prove}, \text{Verify})$ is honestly sound if for all $(x, w) \notin \mathcal{R}$,

$$\Pr \left[ \text{Verify}(pp, x, \pi) = 1 : \pi \leftarrow \text{Prove}(pp, x, w) \right] = \text{negl}(\lambda)$$

Surprisingly, knowledge soundness does not imply honest soundness. While the former guarantees that a veriﬁer can extract a valid witness from an acceptable proof, it makes no guarantees about the witness that the prover used.

As an example, consider the (trivial) relation

$$\mathcal{R}_x = \{(x \in \mathbb{F}, w \in \mathbb{F}) : x = w\}$$

For this relation, deﬁne the proof system $(\text{Setup}, \text{Prove}, \text{Verify})$ where Setup is a no-op, Prove always returns a null proof $\pi = \bot$, and Verify always return 1. This proof system is knowledge-sound, but not honestly sound.

Regardless, it’s easy to show that the SNARKs we study are honestly sound.

It sufﬁces to show a $\Omega(n)$ communication bound for checking a shared R1CS witness. If there were a low-communication protocol for generating a proof, parties could just check a shared witness by constructing and checking the proof. Per completeness and honest soundness, the proof is valid if and only if the witness was valid, except with negligible probability.

Unfortunately, the communication-hard problem DISJ$^n$ reduces to checking additively shared R1CS witnesses. DISJ$^n$ asks whether the two length-$n$ bit-strings do not share a one at any index; i.e., $\text{DISJ}^n(a, b) = \bigwedge_{i=1}^n (a_i \land b_i)$. DISJ$^n$ has been shown to have $\Omega(n)$ randomized 2-party communication complexity [5, 69]. DISJ$^n(a, b)$ reduces to R1CS checking as follows. $\mathcal{P}_1$, who has $a$, interprets it as a zero-one vector in $\mathbb{F}^n$ and sets $x_0 \leftarrow a; y_0 \leftarrow 0; z_0 \leftarrow 0$. Simultaneously, $\mathcal{P}_2$, interprets $b$ as a zero-one vector in $\mathbb{F}^n$ and sets $x_1 \leftarrow 0; y_1 \leftarrow b; z_1 \leftarrow 0$. With this reduction, $a$ and $b$ are disjoint if and only if $(x_0, x_1)$, $(y_0, y_1)$, and $(z_0, z_1)$ are sharings of $x, y, z$ such that $x \land y = z$, an R1CS relation.

Note that this is only a limited bound. First, it requires $\Omega(n)$—not $\Omega(n \log |\mathbb{F}|)$—communication. We suspect the stronger bound holds as well, but do not know of an existing communication complexity result that immediately follows from. Second, the bound applies only to additively shared witnesses.

C KZG for Shared Polynomials

This appendix is available in the paper’s full version: https://ia.cr/2021/1530.

D PA-MPC from collaborative proofs

This appendix is available in the paper’s full version: https://ia.cr/2021/1530.

E Unrolled Groth16/SPDZ Protocol

This appendix is available in the paper’s full version: https://ia.cr/2021/1530.