

# Ligerito: A Fast and Concretely Small Multilinear Polynomial Commitment Scheme

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**Abstract**

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## 1 Introduction

In this short note, we

## 2 Notes

High level idea: we would like to reduce checking that some vector  $x$  is close to a tensor encoding, in the sense that there exists some  $\tilde{x}$  such that  $(G_\ell \otimes \cdots \otimes G_1)\tilde{x}$  and  $x$  are close. In particular, we will view  $x$  as a tensor which is (hopefully) encoded via the tensor code  $G_\ell \otimes \cdots \otimes G_1$ . We will show a protocol that both verifies that there is a unique correctly-encoded tensor ‘closest’ to  $x$  and that the reductions of each tensor to a smaller correspond to (partial) evaluations of a multilinear polynomial whose coefficients are  $\tilde{x}$ .

**Basic check.** From the tensor distance check of [?], we know that, given uniformly sampled  $r \in \mathbf{F}^k$  and  $S \subseteq \{1, \dots, m\}$ , if a matrix  $X \in \mathbf{F}^{m \times 2^k}$  satisfies

$$X_S \bar{g}_r = \bar{X}_S G'^T \quad \text{and} \quad \bar{X}_S \bar{g}_r = G_S y_r,$$

for some  $y_r$  depending on  $r$  but not on  $S$  and some  $\bar{X}_S$ , then we know that, with high probability, there exists a unique matrix  $\tilde{X}$  such that  $\|X - G\tilde{X}G'^T\| < d/3$  and

$$X_S = G_S \tilde{X} G'^T \quad \text{and} \quad y_r = \tilde{X} \bar{g}_r. \tag{1}$$

The probability of error here is bounded from above by, setting  $\delta = d/m$ ,

$$\left(1 - \frac{\delta}{3}\right)^{|S|} + \frac{k(d+9)}{|\mathbf{F}|}. \tag{2}$$

(Note that the distance of  $G'$  does not matter in this particular test; intuitively this is because we are verifying that the each of the  $S$  rows are completely correctly encoded.)

**Basic operations.** Given a vector  $x \in \mathbf{F}^{mn}$ , we will write the ‘natural’ operation

$$\mathbf{Mat} : \mathbf{F}^{mn} \rightarrow \mathbf{F}^{m \times n}$$

which interprets the  $mn$ -sized vector  $x$  as an  $m \times n$  matrix, row-wise. More specifically, the first row of  $\mathbf{Mat}(x)$  corresponds exactly to the first  $n$  entries of  $x$ , and so on, with the dimensions clear from context. Unfortunately, this ‘row-major’ construction, which clashes with the standard ‘column-major’ construction, is unavoidable here without introducing additional complexity down the line.

Although not used later, it is also possible to define the inverse operation,  $\mathbf{vec} : \mathbf{F}^{m \times n} \rightarrow \mathbf{F}^{mn}$ , which simply stacks the rows of the input matrix into an  $mn$ -sized vector, such that  $\mathbf{vec}(\mathbf{Mat}(x)) = x$ .

**Kronecker product.** The *Kronecker product* of two matrices  $G \in \mathbf{F}^{m \times n}$  and  $G' \in \mathbf{F}^{m' \times n'}$  is written  $G \otimes G' \in \mathbf{F}^{mm' \times nn'}$  and is defined as

$$G' \otimes G = \begin{bmatrix} G'_{11}G & G'_{12}G & \cdots & G'_{1n'}G \\ \vdots & \vdots & \ddots & \vdots \\ G'_{m'1}G & G'_{m'2}G & \cdots & G'_{m'n'}G \end{bmatrix}.$$

First, this operation is easily shown to be associative; *i.e.*, for another matrix  $G'' \in \mathbf{F}^{m'' \times n''}$  we have

$$(G'' \otimes G') \otimes G = G'' \otimes (G' \otimes G),$$

so we may write  $G'' \otimes G' \otimes G$  without ambiguity. We will constantly use the fact that this operation is associative in what follows. Note also that, for any  $x \in \mathbf{F}^{nn'}$ , using the above definition, we know

$$\mathbf{Mat}((G' \otimes G)x) = G' \mathbf{Mat}(x) G^T,$$

which also serves as another definition of the Kronecker product that we use throughout the text. (It is very important to note here that, due to the row-major ordering of the  $\mathbf{Mat}$  operation, the product is reversed from many standard linear algebra texts.) Another useful consequence of this definition is that, for some provided  $x \in \mathbf{F}^{2^k}$  and  $r_1, \dots, r_k \in \mathbf{F}$ , we may interpret

$$x^T((1 - r_1, r_1) \otimes \cdots \otimes (1 - r_k, r_k)), \quad (3)$$

as the evaluation of a multilinear polynomial of  $k$  variables, with coefficients in  $x$ , at the point  $(r_1, \dots, r_k)$ . (Care has to be taken in interpreting the individual entries of  $x$  as coefficients over the different possible monomials, but this is a good exercise for the reader not deeply familiar with linear algebra.)

**Structured randomness.** For convenience, we will define, for a vector  $r \in \mathbf{F}^k$ , the ‘structured random’ vector

$$\bar{g}_r = (1 - r_1, r_1) \otimes \cdots \otimes (1 - r_k, r_k),$$

which results in a vector with dimensions  $\bar{g}_r \in \mathbf{F}^{2^k}$ . When the values  $r \in \mathbf{F}^k$  are uniformly randomly drawn, this type of structured vector is sometimes called ‘logarithmically random’ [?]. From before (3), we can interpret the product  $\bar{g}_r^T x$  as the evaluation of a multilinear polynomial over  $k$  variables, with coefficients in  $x$ , at the point  $(r_1, \dots, r_k)$ .

## 3 Protocol

### 3.1 General reduction

**High level idea.** The reduction is simple at a high level. We begin with any  $x \in \mathbf{F}^{mm'}$  and a tensor code  $G \otimes G'$ . We would like to reduce checking that  $x$  uniquely decodes to some message  $\tilde{x}$  to checking some smaller claim of the same form. Doing this, we will also see that, as a byproduct, we get a partial evaluation of  $\tilde{x}$  on some uniformly randomly chosen coefficients, when  $\tilde{x}$  is interpreted as the coefficients of a multilinear polynomial, at a random point. To do this, we will use the basic check presented previously. First, we may verify that  $x$  is indeed close to some tensor codeword by using the basic check presented previously on  $\mathbf{Mat}(x)$ :

$$\mathbf{Mat}(x)_S = \bar{X}_S G'^T \quad \text{and} \quad \bar{X}_S \bar{g}_r = G_S y_r. \quad (4)$$

If this is true, the claim is satisfied by the conclusions of (1). Unfortunately, it is difficult to verify the latter claim in (4): in the standard protocol (used in the tensor distance check of [?]), verifying this latter claim involves sending the complete vector  $y_r$  which may be very large, as we will see next. Instead, we will have the prover commit to a new vector, which we will call  $z_r$ , which it claims satisfies  $(z_r)_S = G_S y_r$ , for some (unknown to the verifier)  $y_r$ . The claim (4) will then be

$$\mathbf{Mat}(x)_S = \bar{X}_S G'^T \quad \text{and} \quad \bar{X}_S \bar{g}_r = (z_r)_S \quad \text{and} \quad (z_r)_S = G_S y_r. \quad (5)$$

Now, it suffices only to prove that the committed vector  $z_r$ , when opened at the  $S$  positions, is indeed equal to the correct encoding (via  $G$  only, instead of  $G \otimes G'$ ) of some message.

**Observation.** In standard Ligo and its tensor variation used in ZODA, as mentioned before, the standard way of verifying this is to send  $z_r$  (or its corresponding message  $y_r$ ) and have the verifier check the encoding (or encode the message) and verify equality at the corresponding  $S$  entries. On the other hand, if  $G$  is itself a tensor code, say  $G = G_2 \otimes G_1$ , then the proposition  $(z_r)_S = G_S y_r$  is implied by the conclusion of (1) except over  $G_1$  and  $G_2$ . (See figure XXX.) This means that we can use the same check, again, to reduce verifying that  $z_r$  is close to a good tensor encoding down to a simpler claim, which is, in turn, even smaller.

**Construction.** Writing this out, let  $G_1 \otimes \dots \otimes G_\ell$  be a tensor code. Let  $d_1, \dots, d_\ell > 0$  be the distances and  $\delta_1, \dots, \delta_\ell$  be the relative distances of each matrix and define  $d = d_1 \dots d_{\ell-1}$ . We

will show that there exists a unique  $\tilde{x}$  such that  $\|\mathbf{Mat}(x) - (G_1 \otimes \dots \otimes G_{\ell-1})\mathbf{Mat}(\tilde{x})G_\ell^T\| < d/3$  and that  $\tilde{x}^T(\bar{g}_{r_1} \otimes \dots \otimes \bar{g}_{r_k}) = t$  if the following tests pass

$$\mathbf{Mat}(z^{(i)})_{S_i} = \bar{X}_{S_i}^{(i)} G'^T \quad \text{and} \quad \bar{X}_{S_i}^{(i)} \bar{g}_{r_i} = z_{S_i}^{(i+1)},$$

for  $i = 1, \dots, \ell - 1$ , where  $z^{(1)} = x$ , and  $S_{i-1}$  is defined as the subset of rows which contain the indices of  $S_i$  after these indices are passed through the **Mat** operation. (See figure ??.) The error probability of this test is

$$\sum_{i=1}^{\ell-1} \left( \left(1 - \frac{\delta_1 \dots \delta_i}{3}\right)^{|S_i|} + \frac{k_i(d_1 \dots d_i + 9)}{3|\mathbf{F}|} \right). \quad (6)$$

Note that we may choose  $|S_i| \geq -\lambda/\log(1 - \delta_1 \dots \delta_i)$ , where  $\lambda$  is the natural log of the desired error probability; or, if the matrices  $G_i$  are all the same code, this is simply  $|S_i| \geq -\lambda/\log(1 - \delta^i)$ . While this exponential decay in relative distance may seem problematic, we will show that, by appropriately choosing the rank  $\ell$  of the tensor code, we receive a total proof size of roughly  $\lambda C \exp(C' \sqrt{\log(N)})$  that is smaller than  $N^\varepsilon$  for any  $\varepsilon > 0$ , where  $N$  is the number of elements of  $\tilde{x}$ , but asymptotically larger than  $\log^c(N)$  for any  $c > 0$ .

### 3.2 Protocol

**Prover algorithm.** The prover algorithm is relatively simple. It begins with some vector  $\tilde{x} \in \mathbf{F}^{2^{k_1+\dots+k_\ell}}$  and code matrices  $G_i \in \mathbf{F}^{m_i \times 2^{k_i}}$  for  $i = 1, \dots, \ell$ , then proceeds as follows.

1. Set  $x = z^{(1)} = (G_1 \otimes \dots \otimes G_\ell)\tilde{x}$
2. For  $i = 1, \dots, \ell - 1$ , receive uniform randomness  $r_i \in \mathbf{F}^{k_i}$ , compute and commit to

$$z^{(i+1)} = (G_i \otimes \dots \otimes G_\ell)\mathbf{Mat}(\tilde{x})(\bar{g}_{r_i} \otimes \dots \otimes g_{r_1});$$

note that the dimensions for the **Mat** operation are clear from context XXX: Check indices

3. Send the entirety of  $z^{(\ell)}$

Note that, if  $G_i$  are systematic codes, it is not necessary to recompute the complete Kronecker product in step 2 at every iteration: it suffices only to take the systematic part over the  $i$ th axis and take the random linear combination of this result. Also the commitment to  $z^{(\ell)}$  is unnecessary since the entire vector is being sent. XXX: Clarify

**Verifier algorithm.** The verifier algorithm is also relatively simple. We assume that the verifier only has access to a small number of entries of the vectors  $z^{(i)}$ .

1. Uniformly randomly sample some set  $S_1 \subseteq \{1, \dots, (m_1 \dots m_\ell)\}$
2. For  $i = 1, \dots, \ell - 1$  perform the following:

- (a) Receive  $S_i$  rows of  $\mathbf{Mat}(z^{(i)}) \in \mathbf{F}^{(m_{i+1} \dots m_\ell) \times m_i}$
  - (b) Verify that the  $S_i$  rows of  $\mathbf{Mat}(z^{(i)})$  are encodings via  $G_i$  of some message  $\bar{X}_{S_i}^{(i)}$
  - (c) Check that  $\bar{X}_{S_i}^{(i)} \bar{g}_{r_i} = z_{S_i}^{(i+1)}$
  - (d) Set  $S_{i+1}$  to be the indices of rows that contain at least one index of  $S_i$  XXX:  
clarify
3. Verify that  $z^{(\ell)}$  is a codeword of  $G_\ell$

**Proof.** The proof of the error bound is essentially by induction. The base case is exactly the basic check: when the code is  $G_1 \otimes G_2$ , the conclusions of the check (1) is satisfied directly.

## References