

latticefold plus

Folding framework

1. Committed CCS to Linear CCCS
2. Decomposition
3. Fold: LCCCS x LCCCS to LCCCS
 - norm bound
 - row check and linear check

Improvement

1. Multi-foldings
2. Double commitment
3. Range proof

1. Notions

1.1 Cyclotomic rings

- $\mathbb{Z}_q = \mathbb{Z}/q\mathbb{Z}$
- $R = \mathbb{Z}[X]/(X^d + 1)$ is the ring of integers of the $2d$ -cyclotomic field, where d is a power of two.
- $R_q = \mathbb{Z}_q[X]/(X^d + 1)$
- $\|x\|_\infty = \max_{i \in n}(x_i)$ is the ℓ_∞ norm of a vector $\vec{x} \in \mathbb{R}^n$.
- f , an element over R or R_q
- \mathbf{b} , bold, a vector over R or R_q
- $\mathbf{b}[i], \mathbf{b}_i$, the i -th element of a vector \mathbf{b} .
- $\langle \mathbf{a}, \mathbf{b} \rangle$, inner product of two vectors \mathbf{a}, \mathbf{b} .

1.2 Relations

- \mathcal{R} , an NP relation
 - \mathcal{R}_{acc} , accumulation
 - \mathcal{R}_{comp} , linear CCS, evaluation at a random point.
 - $\mathcal{R}_{lin,B}$, a general linear relation with norm bound B .
 - $\mathcal{R}_{open} = \{(\mathbb{x}, \mathbb{w}) = (\text{cm}_{\mathbf{f}}, \mathbf{f})\}$, opening a commitment: \mathbf{f} is a valid opening of the commitment $\text{cm}_{\mathbf{f}}$.
 - \mathcal{R}_{dopen} , opening of a double commitment.
 - $\mathcal{R}_{m,in}$
 - $\mathcal{R}_{m,out}$
 - \mathcal{R}_{rg} , range check, norm check

1.3 Operations

- sgn , the sign of an integer, $\text{sgn}(a) \in \{-1, 0, 1\}$.
- exp , the operation that takes an integer exponent and maps it to a monomial element, $\text{exp}(a) = X^a$.
- EXP , the “border-aware” monomial operator,

$$\text{EXP}(a) := \begin{cases} \{\text{exp}(a)\} & \text{if } a \neq 0 \\ \{0, 1, X^{d/2}\} & \text{if } a = 0 \end{cases}$$

- pow , the vector form of exp : it maps an integer vector to a vector over R_q (or, depending on context, to a matrix).
- com , computes the Ajtai commitment of the input vector (refer to 4.2.1.2)
- dcom , computes the double commitment of the input vector
- flat , flattens a matrix into a long vector by concatenating its column vectors
- split , performs base- d' decomposition on a matrix

2. Multi-Folding

Latticefold+ consists of three reductions of knowledge: commit, fold, and decompose, in the following steps.

Step 1. Commit

Reduce the CCS relation $\mathcal{R}_{\text{comp}}$ to a general linear relation $\mathcal{R}_{\text{lin},B}$.

In this step, the (reduction-of-knowledge) prover outputs the witness \mathbb{w} and the statement \mathbb{x} of the relation $\mathcal{R}_{\text{lin},B}$.

Latticefold+ converts the CCS instances into linear committed CCS, which is called a “general linear relation” $\mathcal{R}_{\text{lin},B}$.

This step is the same as Hypernova and Latticefold, except the commitment is improved to a double commitment.

Step 2. Fold

From $L \geq 2$ instances in $\mathcal{R}_{\text{lin},B} \times \dots \times \mathcal{R}_{\text{lin},B}$ to $\mathcal{R}_{\text{lin},B^2}$.

In this step, it folds the given L instances of $\mathcal{R}_{\text{lin},B}$ algebraically. In latticefold, there are only 2 CCS instances.

Step 3. Decompose

From $\mathcal{R}_{\text{lin},B^2}$ back to 2 $\mathcal{R}_{\text{lin},B}$ instance, using the B-nary decomposition technique (as in latticefold).

$$\left. \begin{matrix} R_{\text{lin}, B}(x, w) \\ R_{\text{lin}, B}(x, w) \\ \vdots \\ R_{\text{lin}, B}(x, w) \end{matrix} \right\} L \xrightarrow{\text{fold}} R_{\text{lin}, B^2}(\sum x, \sum w) \xrightarrow{\text{decompose}} \begin{cases} R_{\text{lin}, B} \\ R_{\text{lin}, B} \end{cases}$$

3. Tools

3.1 MLE over rings

For a function $f(\mathbf{b}) : \{0, 1\}^k \rightarrow R$, or a vector $\mathbf{f} = [f(0), \dots, f(2^k - 1)]$, the MLE function $\tilde{f}(\mathbf{x})$ is

$$\tilde{f}(\mathbf{x}) := \sum_{\mathbf{b} \in \{0,1\}^k} f(\mathbf{b}) \cdot eq(\mathbf{b}, \mathbf{x}),$$

where $eq(\mathbf{b}, \mathbf{x}) := \prod_{i \in [k]} [(1 - \mathbf{b}_i)(1 - \mathbf{x}_i) + \mathbf{b}_i \mathbf{x}_i]$.

The evaluation of \tilde{f} at the point \mathbf{r} is $\tilde{f}(\mathbf{r}) = \sum_{\mathbf{b} \in \{0,1\}^k} f(\mathbf{b}) \cdot eq(\mathbf{b}, \mathbf{r})$.

We can “eliminate” \mathbf{b} and express $eq(\mathbf{b}, \mathbf{r})$ using only \mathbf{r} .

By the definition of $eq(\mathbf{b}, \mathbf{r})$, when $\mathbf{b} \in \{0, 1\}^k$, all values $eq(\mathbf{b}, \mathbf{r})$ form a length- 2^k vector as follows:

$$\begin{aligned} \mathbf{b} &= (0, \dots, 0), \quad eq(\mathbf{b}, \mathbf{r}) = \prod_{i \in [k]} (1 - \mathbf{r}_i) \\ \mathbf{b} &= (0, \dots, 1), \quad eq(\mathbf{b}, \mathbf{r}) = \mathbf{r}_k \cdot \prod_{i \in [k-1]} (1 - \mathbf{r}_i) \\ &\dots \\ \mathbf{b} &= (1, \dots, 1), \quad eq(\mathbf{b}, \mathbf{r}) = \prod_{i \in [k]} \mathbf{r}_i. \end{aligned}$$

By the definition of the tensor product, the vector

$$(\prod_{i \in [k]} (1 - \mathbf{r}_i), \mathbf{r}_k \cdot \prod_{i \in [k-1]} (1 - \mathbf{r}_i), \dots, \prod_{i \in [k]} \mathbf{r}_i) = \otimes_{i \in [k]} (1 - \mathbf{r}_i, \mathbf{r}_i),$$

denoted by $\text{tensor}(\mathbf{r})$.

Therefore, we have $\tilde{f}(\mathbf{r}) = \langle \mathbf{f}, \text{tensor}(\mathbf{r}) \rangle$.

3.2 Double Commitment

3.2.1 Commit to any (large-norm) vectors

3.2.1.1 Gadget Matrix

For an integer $n \geq 1$, define a “gadget matrix” used to compose base- b digits into a value:

$$\mathbf{G}_{b,n} = \mathbf{I}_n \otimes [1 \ b \ b^2 \ \dots \ b^{k-1}] \in \mathcal{R}_q^{n \times nk}.$$

Correspondingly, we use $\mathbf{G}_{b,k}^{-1} : \mathcal{R}_q^n \rightarrow \mathcal{R}_q^{kn}$ to denote base- b decomposition.

Then, for an arbitrary (large) vector $\mathbf{m} \in R^m$, $\mathbf{G}_{b,k}^{-1}(\mathbf{m})$ means: decompose all coefficients of \mathbf{m} into base b , and then pack (pad) them into a vector (over \mathbb{Z}^{nkd}). Hence $\|\mathbf{G}_{b,k}^{-1}(\mathbf{m})\| < b$.

More generally, the \mathbf{G}^{-1} decomposition can also be applied to a matrix. To make the next step convenient, we concatenate the decomposed matrix's column vectors into a single long vector. In Latticefold+, this process is denoted by $\text{split}(\mathbf{M})$.

3.2.1.2 Ajtai commitment

For a message vector $\mathbf{m} \in R^m$ with $\|\mathbf{m}\|_\infty \leq b$, the Ajtai commitment works as follows:

- KeyGen: given the security parameter λ , generate a commitment key $\mathbf{A} \in R_q^{n \times m}$.
- Com: given the commitment key $\mathbf{A} \in R_q^{n \times m}$ and message $\mathbf{m} \in R^m$, compute the commitment $\text{cm}_{\mathbf{m}} = \mathbf{A}\mathbf{m}$.

Similarly, when the message is a matrix \mathbf{M} , we can still commit to it using the operation defined by Com. In Latticefold+, this is denoted by $\text{com}(\mathbf{M}) = \mathbf{A}\mathbf{M}$.

If $\|\mathbf{m}'\|_\infty > b$, we can apply \mathbf{G}^{-1} decomposition and then compute the commitment to \mathbf{m}' as

$$\text{cm}_{\mathbf{m}} = \mathbf{A}\mathbf{G}^{-1}(\mathbf{m}').$$

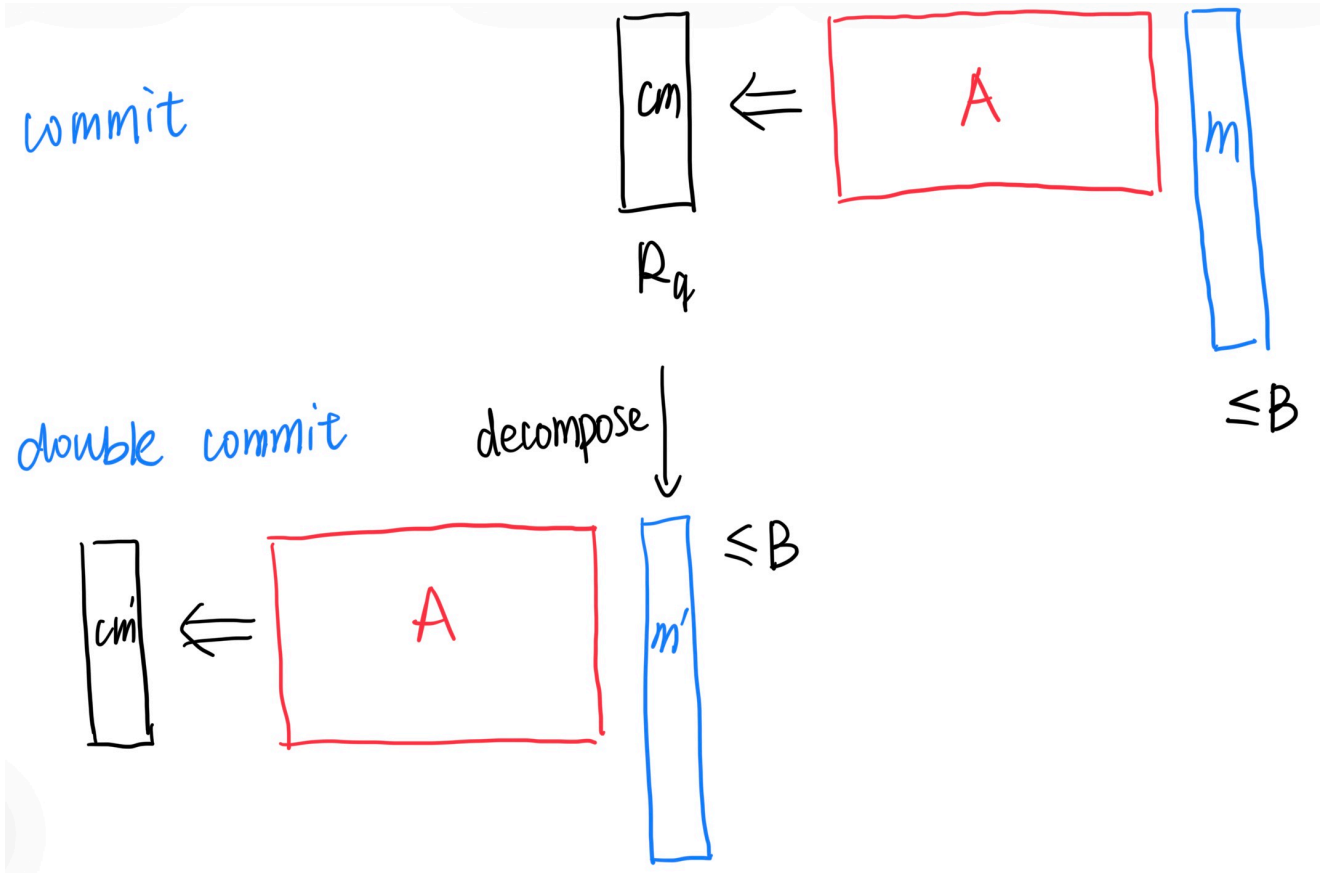
For a matrix \mathbf{M}' with $\|\mathbf{M}'\|_\infty > b$, we can commit via $\text{com}(\text{split}(\mathbf{M}'))$.

3.2.2 Double Commitment

Like Greyhound and Labrador, Latticefold+ applies a double commitment to reduce communication costs.

In a word, a double commitment is a “commitment of a commitment.” We use dcom (double commit) to denote this process. For a message matrix \mathbf{M} ,

$$\text{dcom} := \text{com}(\text{split}(\text{com}(\mathbf{M}))).$$



This double commitment can reduce proof size but it also eliminates the additive homomorphism.

4.2.3 Prove the consistency of a double commitment

- Consistency: dcom and com are computed as in the above figure.

Folding

Formally, $\mathcal{R}_{lin,B}$ is defined below.

$$\mathcal{R}_{lin,B} = \left\{ (\mathfrak{i}, \mathfrak{x}, \mathfrak{w}) \left| \begin{array}{l} \mathfrak{x} = (cm_{\mathbf{f}}, \mathbf{r}, \mathbf{v}_1, \dots, \mathbf{v}_{n_{lin}}), \\ \mathfrak{w} = \mathbf{f}, \\ cm_{\mathbf{f}} = \text{Com}(\mathbf{f}), \\ \|\mathbf{f}\|_{\infty} < B, \\ \forall i \in [n_{lin}], \langle \mathbf{M}^{(i)} \mathbf{f}, \text{tensor}(\mathbf{r}) \rangle = \mathbf{v}_i \end{array} \right. \right\}, \quad (2)$$

where $\langle \mathbf{M}^{(i)} \mathbf{f}, \text{tensor}(\mathbf{r}) \rangle = \mathbf{v}_i$ means that $\mathbf{M}^{(i)} \mathbf{f}$ evaluates to \mathbf{v}_i at the point \mathbf{r} , and \mathfrak{i} is an index that includes the public parameters of the relation.