

Ajtai commitment expansion

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What we are trying to do

- ▶ We start off with a relation made hard to 'break' based on the MSIS problem
- ▶ We will show that this relation is equivalent to two other relations, that can be used for folding
- ▶ We then bind these two other relations to the Customisable Constraint System (CCS)
- ▶ This allows for a folding scheme based on ajtai commitments and compatible with CCS

Ajtai Commitments

- ▶ Ajtai commitments allow us to commit to a vector of polynomials
- ▶ We commit to an a vector $\vec{x} \in \mathcal{R}^m$ by multiplying it with a random matrix $\mathbf{A} \in \mathcal{R}_q^{\kappa \times m}$
- ▶ $\|\vec{x}\|_\infty < B$ where B is the norm bound
- ▶ Output of commitment is $cm := \mathbf{A} \cdot \vec{x} \bmod \mathbf{q} \in \mathcal{R}_q^\kappa$
- ▶ This commitment is considered binding because of the assumed hardness of MSIS

Ajtai commitments as a relation

- ▶ We define relation $\mathcal{R}_{MSIS\infty}^B$ between an ajtati commitment and the \vec{x}
- ▶ $\mathcal{R}_{MSIS\infty}^B := (pp, cm \in \mathcal{R}_q^\kappa; \vec{x} \in \mathcal{R}^m : (cm = \mathbf{A} \cdot \vec{x} \bmod \mathbf{q}) \wedge \|\vec{x}\|_\infty < \mathbf{B})$
- ▶ $pp := (\kappa, m, B, \mathbf{A})$ are the public parameters of the relation
- ▶ Public parameters define the 'meta' information of the relation:
 1. The size of the vectors and matrices
 2. The norm limit of \vec{x}
 3. The random matrix \mathbf{A}

$$\vec{x} \in \mathcal{R}_q^m$$

- ▶ We now rewrite $\vec{x} \in \mathcal{R}_q^m$
- ▶
 - ▶ $\|\vec{x}\|_\infty < B$ and $B < \frac{q}{2}$
 - ▶ $\vec{x} \in \mathcal{R}^m$ can be uniquely represented in \mathcal{R}_q^m
 - ▶ We define $\|\vec{x}\|_{\text{infly}} < B$ as the norm after lifting $\vec{x} \in \mathcal{R}_q^m$ to \mathcal{R}
- ▶ We can rewrite our commitment as

$$\mathcal{R}_{MSIS^\infty}^B := (pp, cm \in \mathcal{R}_q^\kappa; \vec{x} \in \mathcal{R}_q^m : (cm = \mathbf{A} \cdot \vec{x}) \wedge \|\vec{x}\|_\infty < \mathbf{B})$$

Coefficient Embeddings and Rotational Matrices

- ▶ For $a \in \mathcal{R}_q$, $\text{vec}(a) \in \mathbb{Z}_q^d$ represents the vectors of coefficients
- ▶ For a vector $\vec{a} \in \mathcal{R}_q^m$, $\text{vec}(\vec{a}) \in \mathbb{Z}^{m \times d}$ represents the coefficient vectors in \vec{a}
- ▶ $\text{fvec}(\vec{a}) \in \mathbb{Z}^{md}$ is the vector that concatenates the rows of \vec{a}
- ▶ $\text{Rot}(\mathbf{a}) := (\text{vec}(\mathbf{a}), \text{vec}(\mathbf{X} \cdot \mathbf{a}), \dots, \text{vec}(\mathbf{X}^{d-1} \cdot \mathbf{a})) \in \mathbb{Z}_q^{d \times d}$.
- ▶ For a matrix $\mathbf{A} \in \mathbb{R}_q^{\kappa \times m}$, we define the rotation matrix $\text{Rot}(\mathbf{A}) \in \mathbb{Z}_q^{\kappa d \times md}$ as

$$\text{Rot}(\mathbf{A}) := \begin{bmatrix} \text{Rot}(\mathbf{A}_{1,1}) & \cdots & \text{Rot}(\mathbf{A}_{1,m}) \\ \vdots & \ddots & \vdots \\ \text{Rot}(\mathbf{A}_{\kappa,1}) & \cdots & \text{Rot}(\mathbf{A}_{\kappa,m}) \end{bmatrix}$$

- ▶ $\text{fvec}(\mathbf{A}\mathbf{f}) = \text{Rot}(\mathbf{A})\text{fvec}(\mathbf{f})$ for any $\mathbf{A} \in \mathbb{R}_q^{\kappa \times m}$ and $\mathbf{f} \in \mathbb{R}_q^m$.

$$\vec{x} \in \mathbb{Z}^{\kappa d}$$

- ▶ We can uniquely represent $\vec{x} \in \mathcal{R}_q^m$ as $\vec{x} \in \mathbb{Z}^{\kappa d}$ by taking $fvec(\vec{x})$
- ▶ $\bar{\mathbf{A}} = rot(\mathbf{A})$
- ▶ \overline{cm} is the coefficient embedding of cm
- ▶ $\overline{cm} = \bar{\mathbf{A}} \cdot fvec(\vec{x})$

$$\mathcal{R}_{MSIS^\infty}^B := (pp, \overline{cm} \in \mathbb{Z}^{\kappa d}; \vec{x} \in \mathbb{Z}^{md} : (\overline{cm} = \bar{\mathbf{A}} \cdot \vec{x}) \wedge \|\vec{x}\|_\infty < B)$$

Representing $\|\vec{x}\|_\infty < B$ as an hadamard product

$$\mathcal{R}_{\text{MSISProd}}^B := \left\{ (pp, \overline{cm} \in \mathbb{Z}^{\kappa d}; \vec{x} \in \mathbb{Z}^{md} \mid \begin{array}{l} \overline{cm} = \overline{\mathbf{A}} \cdot \vec{x} \\ \wedge \|\vec{x}\| \circ \left[\bigcirc_{i=1}^{B-1} (\vec{x} - \vec{i}) \circ (\vec{x} + \vec{i}) \right] = \vec{0} \end{array} \right\}$$

- To see this see that the biggest coefficient in any of the x matrices is less than B

$$\mathcal{R}_{cm}^B$$

- ▶ We can look at \vec{x} in two ways
- ▶ \vec{x} is a NTT representation of a $\hat{f} \in \mathcal{R}_q^m$
- ▶ \vec{x} is coefficient embedding of a $\vec{f} \in \mathcal{R}_q^m$
- ▶ The Hadamard product of two NTT representation is equivalent to the multiplication of the two elements
- ▶ i.e $\vec{x} \circ \vec{x} \cong \hat{f} \circ \hat{f}$

$$\mathcal{R}_{cm}^B := \left\{ (pp, cm \in \mathcal{R}_q^\kappa; \vec{f} \in \mathcal{R}_q^m \mid \begin{array}{l} \overline{cm} = \overline{\mathbf{A}} \cdot \vec{f} \\ \wedge \|\hat{f}\| \circ \left[\bigcirc_{i=1}^{B-1} (\hat{f} - \hat{i}) \circ (\hat{f} + \hat{i}) \right] = \hat{0} \end{array} \right\}$$

$$\mathcal{R}_{eval}^B$$

- ▶ Essentially the same as before, with an added evaluation statement
- ▶ We supply the relation with variables and an evaluation of the \vec{f} at those variable

$$\mathcal{R}_{eval}^B = \left\{ (pp; (r, v, cm) \in \mathcal{R}_q^{\log m} \times \mathcal{R}_q \times \mathcal{R}_q^\kappa; \vec{f} \in \mathcal{R}_q^m) \mid \begin{array}{l} (pp; cm; \vec{f}) \in \mathcal{R}_{cm}^B \\ \wedge \text{mle}[\hat{f}](\vec{r}) = v \end{array} \right\}$$

Let's take this to CCS

- ▶ We introduce an insane amount of notation
- ▶
 - ▶ Public Paramers (\mathbf{pp}) $:= (n_r, n_c, t, n_s, \deg, l_{in})$
 - ▶ $\overline{\mathcal{R}}$ is an arbitrary ring
 - ▶ \mathfrak{i} consists of
 1. t matrices $M_1..M_t \in \overline{\mathcal{R}}^{n_r \times n_c}$ with $\mathcal{O}(n_r + n_c)$ non-zero entries
 2. n_s multisets $S_1..S_{n_s} \subseteq [t]$ with $|S_i| < \deg$ for all $i \in [n_s]$
 3. n_s scalars $cn_1, \dots, cn_s \in \overline{\mathcal{R}}$
- ▶ We then introduce the relation \mathcal{R}_{CCS}
- ▶
 1. $\mathbf{pp}_{\text{CCS}} := (\mathbf{pp}, \mathfrak{i})$
 2. $(\mathbf{pp}_{\text{CCS}}, \mathbb{X} \in \overline{\mathcal{R}}^{l_{in}}, \mathbb{W} \in \overline{\mathcal{R}}^{n_c - l_{in} - 1})$
 3. $\vec{\mathbb{Z}} := (\mathbb{X}, 1, \mathbb{W}) \in \mathcal{R}^{n_c}$
 4. The condition for the relation is $\sum_{i=1}^{n_s} c_i \cdot \bigcirc_{j \in S_i} (M_j \cdot \vec{\mathbb{Z}}) = 0^{n_r}$

$$\mathcal{R}_{\text{CCS}} := \left\{ (\mathbf{pp}_{\text{CCS}}, \mathbb{X} \in \overline{\mathcal{R}}^{l_{in}}, \mathbb{W} \in \overline{\mathcal{R}}^{n_c - l_{in} - 1}) \text{ such that } \sum_{i=1}^{n_s} c_i \cdot \bigcirc_{j \in S_i} (M_j \cdot \vec{\mathbb{Z}}) = 0^{n_r} \right\}$$

Let's bind \mathcal{R}_{cm} and \mathcal{R}_{ccs} together!

- ▶ We introduce the gadget matrix
- ▶ $l := \frac{m}{n_c}$
- ▶ $\mathbf{G} := \mathbf{I}_{n_c} \otimes [1, B, \dots B^{l-1}] \in \mathbb{Z}_q^{n_c \times m}$
- ▶ We then define $\mathcal{R}_{\text{cmcss}}^B$

We then define

$$\mathcal{R}_{\text{cmcss}}^B := \left\{ (\text{pp}, \mathbb{X} := (\text{cm} \in \mathcal{R}_q^{\kappa}, \mathbb{X}_{\text{ccs}} \in \mathcal{R}_q^{l_{\text{in}}})) ; \mathbb{W} := (\vec{\mathbf{f}} \in \mathcal{R}_q^m, \mathbb{W}_{\text{ccs}} \in \mathcal{R}^{n-l_{\text{in}}-1}) \text{ s.t. } \right. \\ \left. (\text{pp}, \text{cm}, \vec{\mathbf{f}} \in \mathcal{R}_{\text{cm}}^B) \wedge (\text{pp}_{\text{ccs}}, \mathbb{X}_{\text{ccs}}; \mathbb{W}_{\text{ccs}}) \in \mathcal{R}_{\text{ccs}} \wedge (\mathbf{z}_{\text{ccs}} = \mathbf{G} \vec{\mathbf{f}}) \right\}$$

$$\bar{\mathbf{z}}_{\text{ccs}} := (\mathbb{X}_{\text{ccs}}, \mathbf{1}, \mathbb{W}_{\text{ccs}}) \in \mathcal{R}^{n_c}$$

$$\mathcal{R}_{\text{lccs}}^B$$

- ▶ We take the multilinear ccs constraint matrices and provide some values and give expected evaluations
- ▶ We first make sure that the number of constraints n_r as the same as the size of our Ajtai commitment vector m
- ▶ $\vec{\mathbf{r}} \in \mathcal{R}_q^{\log m}$ is the values to be evaluated
- ▶ $[\mathbf{u}_i]_{i=1}^t$ are the expected evaluations
- ▶ $\mathcal{R}_{\text{lccs}} := (\text{pp}_{\text{ccs}}, (\vec{\mathbf{r}}, [\mathbf{u}_i]_{i=1}^t, \mathbb{X}_{\text{ccs}}, \mathbf{h}))$
- ▶ For all $i \in [t]$ it holds that $\mathbf{u}_i = \sum_{\vec{\mathbf{b}} \in \{0,1\}^{\log n_c}} \text{mle}[M_i](\vec{\mathbf{r}}, \vec{\mathbf{b}}) \cdot \text{mle}[\mathbf{z}_{\text{ccs}}](\vec{\mathbf{b}})$

$$\mathcal{R}_{\text{evalccs}}^B$$

- We can now combine $\mathcal{R}_{\text{eval}}^B$ and $\mathcal{R}_{\text{lccs}}^B$ to yield our final relation $\mathcal{R}_{\text{evalccs}}^B$

$$\mathcal{R}_{\text{evalccs}}^B = \left\{ \begin{array}{l} (\text{pp}, \mathbb{X} := (\vec{\mathbf{r}} \in \mathcal{R}_q^{\log m}, \text{cm}, \mathbf{v}, [\mathbf{u}_i]_{i=1}^t, \mathbb{X}_{\text{ccs}}, \mathbf{h})); \\ \mathbb{W} := (\vec{\mathbf{f}} \in \mathcal{R}_q^m, \mathbb{W}_{\text{ccs}} \in \mathcal{R}^{n-l_{\text{in}}-1}) \text{ s.t.} \\ (\mathbf{z}_{\text{ccs}} = \mathbf{G} \vec{\mathbf{f}}) \wedge (\text{pp}_{\text{cm}}, (\text{cm}, \vec{\mathbf{r}}, \mathbf{v}); \vec{\mathbf{f}}) \in \mathcal{R}_{\text{eval}}^B \\ \wedge (\text{pp}_{\text{ccs}}, (\vec{\mathbf{r}}_{\text{ccs}}, [\mathbf{u}_i]_{i=1}^t, \mathbb{X}_{\text{ccs}}, \mathbf{h}); \mathbb{W}_{\text{ccs}}) \in \mathcal{R}_{\text{lccs}} \end{array} \right\}$$