

# Theory and Practice of Succinct Zero Knowledge Proofs

## Lecture 05: Holographic PIOP for R1CS

Pratyush Mishra  
UPenn  
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**A PIOP for R1CS**

# R1CS

An rank-1 constraint system (R1CS) is a generalization of arithmetic circuits

$$(F := (\mathbb{F}, n \in \mathbb{N}, A, B, C), x, w)$$

$$z := \begin{bmatrix} x \\ w \end{bmatrix} \quad \overset{n}{\underbrace{\left[ \begin{array}{c} \phantom{A} \\ \phantom{A} \end{array} \right]}_n} \begin{bmatrix} A \\ z \end{bmatrix} \circ \begin{bmatrix} B \\ z \end{bmatrix} = \begin{bmatrix} C \\ z \end{bmatrix}$$

# What checks do we need?

## **Step 1: Correct Hadamard product**

check that for each  $i$ ,  $z_A[i] \cdot z_B[i] = z_C[i]$

## **Step 2: Correct matrix-vector multiplication**

check that  $Mz = z_M \quad \forall M \in \{A, B, C\}$

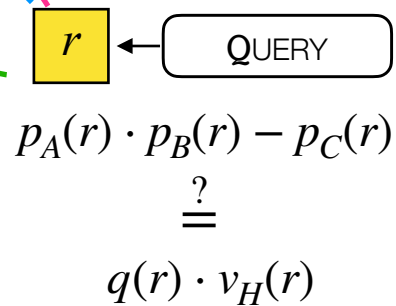
# PIOP for Hadamard Product

**Prover**( $F, x, w$ )

1. Let  $H \subseteq \mathbb{F}$  be a set of size  $n$ .
2. Interpolate  $z_A, z_B, z_C$  over  $H$  to get  $p_A, p_B, p_C$ .
3. Compute quotient  $q = \frac{p_A \cdot p_B - p_C}{v_H}$ .



**Verifier**( $F, x$ )



# PIOP for Matrix-vector products

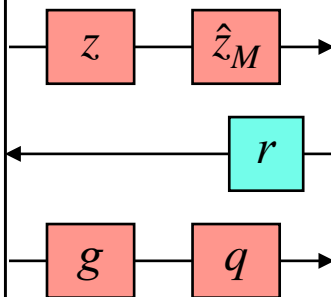
## Prover( $M, z$ )

1. Compute  $z_M := Mz$
2. Interpolate  $z_M$  over  $H$  to get  $\hat{z}_M$
3. Interpolate  $(\vec{r}, \vec{r}^\top M)$  to get  $(\hat{r}, \hat{r}_M)$
4. Use sumcheck lemma to compute  $g, q$  such that

$$\begin{aligned} \hat{r}(X) \cdot \hat{z}_M(X) - \hat{r}_M(X) \cdot \hat{z}(X) \\ = \\ X \cdot g(X) + q(X)v_H(X) \end{aligned}$$

## Verifier( $M$ )

1.  $r \xleftarrow{\$} \mathbb{F}$
2.  $\vec{r} := (1, r, \dots, r^{n-1})$
3. Interpolate  $(\vec{r}, \vec{r}^\top M)$  to get  $(\hat{r}, \hat{r}_M)$
4. Invoke PIOP for ZC!



# New tool: univariate sum check

Lemma:

$$\sum_{h \in H} p(h) = \sigma$$



$\exists q, g$  s.t.

$$p(x) = xg(x) + \frac{\sigma}{|H|} + q(x) \cdot v_H(x)$$

Reduce sum check  
to zero check!

# Why is the verifier slow?

## Prover( $M, z$ )

1. Compute  $z_M := Mz$
2. Interpolate  $z_M$  over  $H$  to get  $\hat{z}_M$
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## Verifier( $M$ )

1.  $r \xleftarrow{\$} \mathbb{F}$
2.  $\vec{r} := (1, r, \dots, r^{n-1})$
3. Interpolate  $(\vec{r}, \vec{r}^\top M)$  to get  $(\hat{r}, \hat{r}_M)$

To check this, it must evaluate  $\hat{r}(X)$  and  $\hat{r}_M(X)$

Must compute  $\vec{r}^\top \cdot M!$

Make PIOP for ZC!



# Key tool: Lagrange polynomials

**Lagrange polynomials for set  $H \subseteq \mathbb{F}$ :**

For each  $i \in H$ ,  $L_H^i(X)$  is 1 at  $i$ , and 0 for all  $j \in H, j \neq i$

**Polynomial Interpolation:**

Given a list  $A = (a_0, \dots, a_d)$ , and a set  $H \subseteq \mathbb{F}$ , the interpolation of  $A$  over  $H$  is

$$\hat{a}(X) := \sum_{i \in H} a_i \cdot L_H^i(X)$$

**Relation to vanishing polynomial:**  $L_H^i(X) := c_i \cdot \frac{v_H(X)}{X - i}$

# Step 1: Efficient $\hat{r}(X)$

Can write  $\hat{r}(X)$  as  $\sum_{i \in H} r^i \cdot L_H^i(X)$

Efficiently evaluating this at a random point  $\beta$  requires efficiently computing each  $r^i$  and  $L_H^i(\beta)$

Let's interpret this as  $\sum_{i \in H} Y^i \cdot L_H^i(X)$

Monomial basis

Lagrange basis

# Step 1: Efficient $\hat{r}(X)$

1. Replace Monomial with Lagrange basis  $\sum_{i \in H} L_H^i(Y) \cdot L_H^i(X)$
2. Can rewrite this as  $\frac{v_H(Y)X - v_H(X)Y}{|H|(X - Y)}$

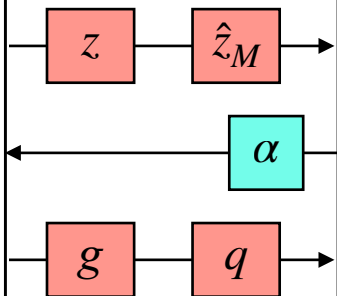
This can be evaluated in time  $O(\log |H|)$ !

# Why is the verifier slow?

## Prover( $M, z$ )

1. Compute  $z_M := Mz$
2. Interpolate  $z_M$  over  $H$  to get  $\hat{z}_M$
3. Use sumcheck lemma to compute  $g, q$  such that

$$\begin{aligned} \hat{r}(\alpha, X) \cdot \hat{z}_M(X) - \hat{r}_M(\alpha, X) \cdot \hat{z}(X) \\ = \\ X \cdot g(X) + q(X)v_H(X) \end{aligned}$$



## Verifier( $M$ )

1.  $\alpha \xleftarrow{\$} \mathbb{F}$
  2. ???
  3. Invoke PIOP for ZC!
- Must compute  $\mathbf{L}_H^i(\alpha)^\top \cdot M!$

# How to use this?

$$\hat{r}(\alpha, \beta) \cdot \hat{z}_M(\beta) - \hat{r}_M(\alpha, \beta) \cdot \hat{z}(\beta)$$

Recall: We have to show

=

$$\beta \cdot g(\beta) + q(\beta)v_H(\beta)$$

Let's expand  $\hat{r}_M(\alpha, \beta) = \sum_{i \in H} \hat{r}(\alpha, i) \cdot \hat{M}(i, \beta)$

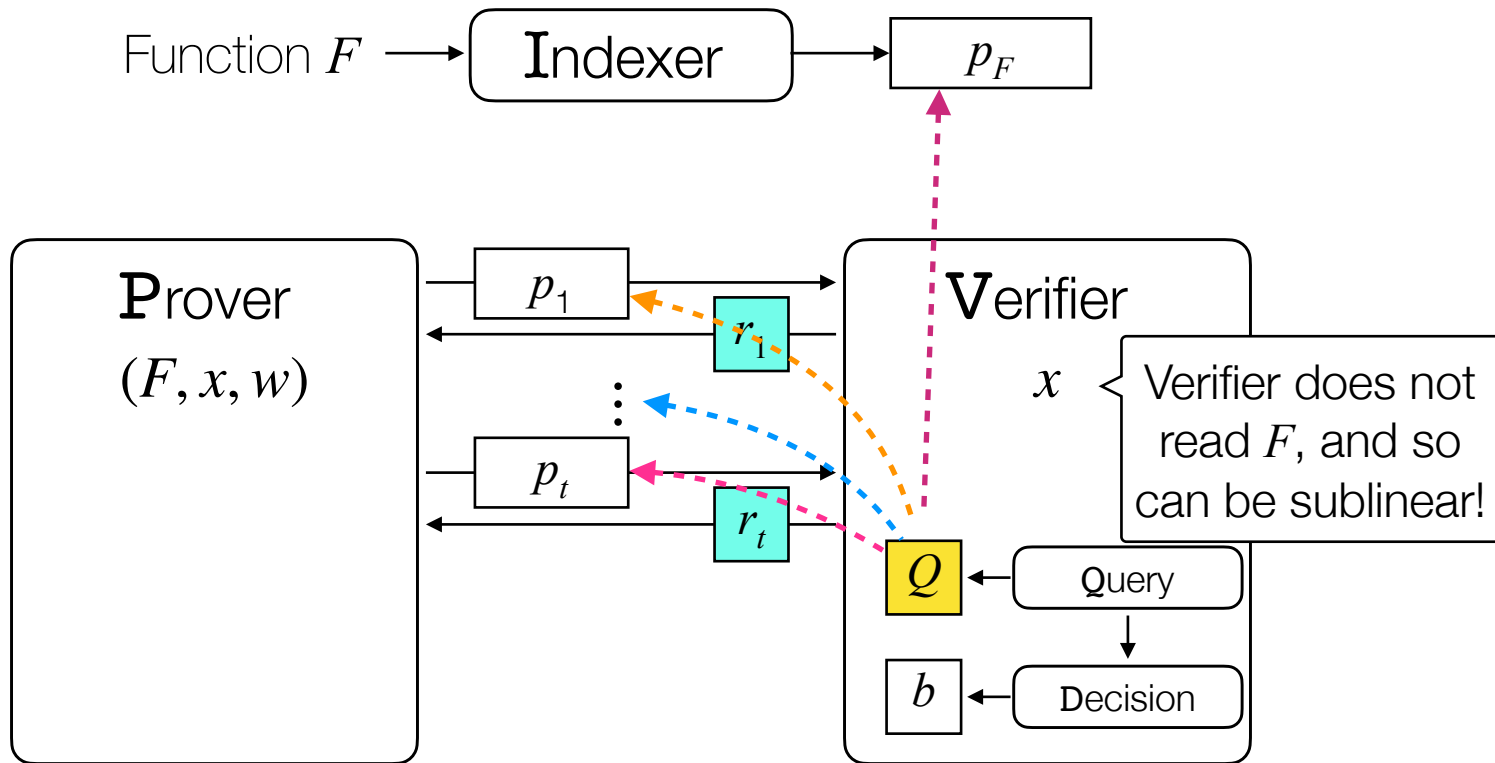
This is yet another sumcheck, so we engage in another sumcheck PIOP, which results in the following check:  $\hat{r}(\alpha, \gamma) \cdot \hat{M}(\gamma, \beta) = \gamma \cdot g'(\gamma) + h'(\gamma)v_H(\gamma)$

How to evaluate  $\hat{M}(\gamma, \beta)$ ?

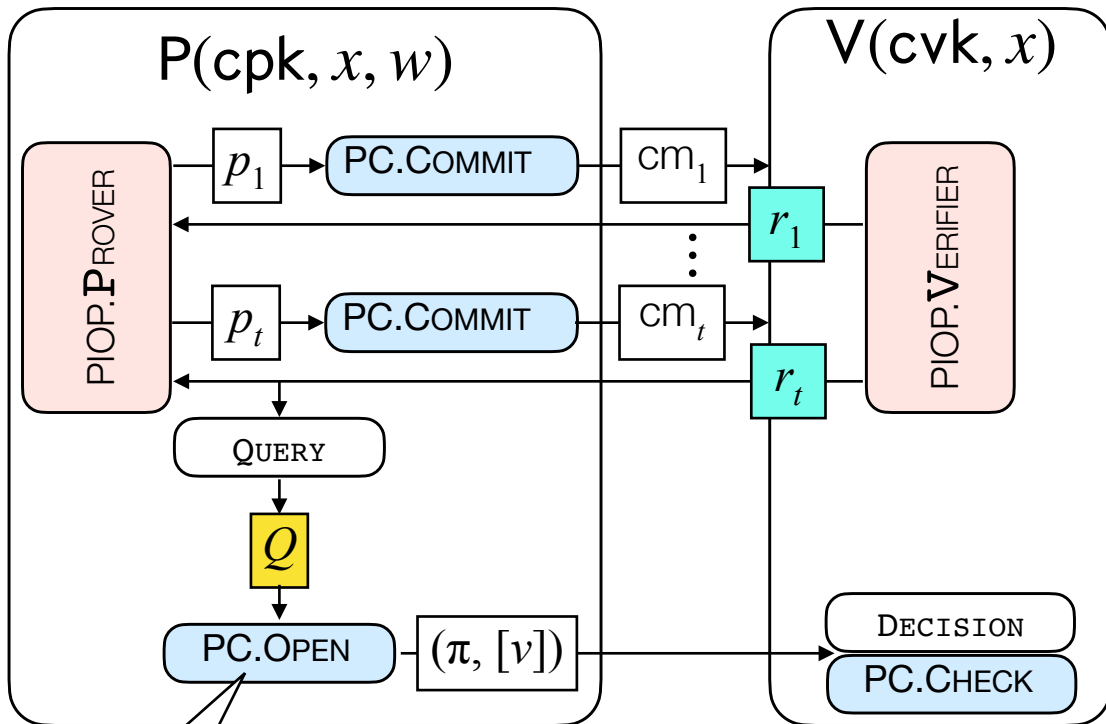
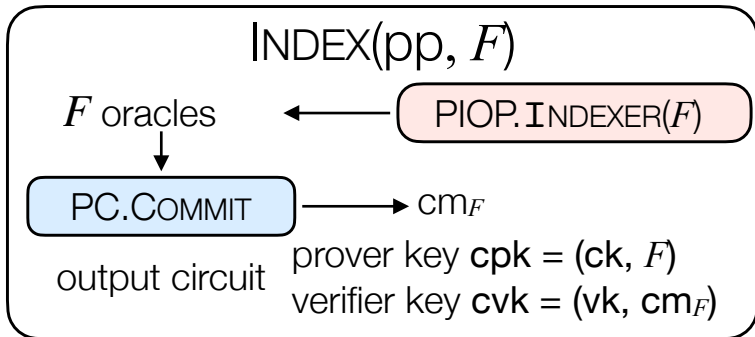
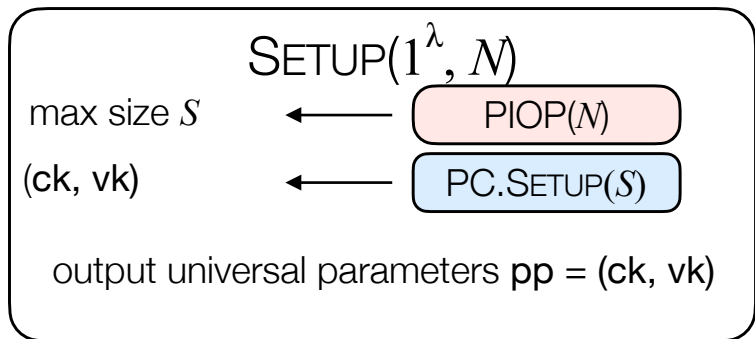
# Sublinear verification for PIOP-based SNARKs

# Holographic PIOPs [CHMMVW20, COS20]

Introduce a new algorithm to preprocess the matrices



# Holographic PIOPs + PC Schemes $\rightarrow$ Preprocessing SNARKs



Prover answers queries to  $F$  oracles too

+ Fiat–Shamir to get non-interactivity



# Verifier Complexity of Holographic PIOP-based SNARKs

$$T(\text{SNARK.V}) = T(\text{CHECK}) + T(\text{HIOP.V})$$

Now sublinear!

Holography enables sublinear verification for arbitrary circuits computations!

# How to encode matrix?

## ***Polynomial Interpolation of Lists:***

Given a list  $A = (a_0, \dots, a_d)$ , and a set  $H \subseteq \mathbb{F}$ , the interpolation of  $A$  over  $H$  is

$$\hat{a}(X) := \sum_{i \in H} a_i \cdot L_H^i(X)$$

## ***Polynomial Interpolation of Matrices?:***

Given a list  $M \in \mathbb{F}^{n \times n}$ , and a set  $H \subseteq \mathbb{F}$ , the bivariate interpolation of  $A$  over  $H$  is

$$\hat{M}(X, Y) := \sum_{i \in H} \sum_{j \in H} M_{ij} \cdot L_H^i(X) \cdot L_H^j(Y)$$

Problem: computing this requires  $O(|H|^2)$  work

# Insight: The matrices are sparse!

## ***Polynomial Interpolation of Matrices?:***

Given a list  $M \in \mathbb{F}^{n \times n}$ , and a set  $H \subseteq \mathbb{F}$ , the bivariate interpolation of  $A$  over  $H$  is

$$\hat{M}(X, Y) := \sum_{i \in H} \sum_{j \in H} M_{ij} \cdot L_H^i(X) \cdot L_H^j(Y)$$

Most  $M_{ij}$  are zero!

Can rewrite as 
$$\hat{M}(X, Y) := \sum_{i \in H} \sum_{j \in H} M_{ij} \cdot \frac{v_H(X)}{X - i} \cdot \frac{v_H(Y)}{Y - j},$$

Additionally, sum only over non-zero entries!

# Final Matrix Encoding

Let  $m$  be the number of non-zero entries, and  $K \subset \mathbb{F}$  be a subset of size  $m$ . Then, a *sparse* bivariate interpolation of  $A$  over  $K$  is

$$\hat{M}(X, Y) := \sum_{k \in I} v(k) \cdot \frac{v_H(X)}{X - r(k)} \cdot \frac{v_H(Y)}{Y - c(k)}$$

Actually, we need *univariate* interpolations over  $K$ , i.e.  $\hat{r}, \hat{c}, \hat{v}$  zero entry,  $\hat{r}, \hat{c}, \hat{v}$  non-zero entry,  $\hat{r}, \hat{c}, \hat{v}$  non-zero entry

$$\hat{M}(X, Y) := \sum_{k \in K} \hat{v}(k) \cdot \frac{v_H(X)}{X - \hat{r}(k)} \cdot \frac{v_H(Y)}{Y - \hat{c}(k)}$$