

# Succinct Arguments

## Lecture 05: Holographic PIOP for R1CS

**A PIOP for R1CS**

# R1CS

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

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

$$z := \begin{bmatrix} x \\ w \end{bmatrix} \quad n \left\{ \begin{matrix} n \\ \left[ A \right] \left[ z \right] \end{matrix} \right\} \circ \left[ B \right] \left[ z \right] = \left[ C \right] \left[ z \right]$$

# 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  $z_A, z_B, z_C$  to get  $p_A, p_B, p_C$ .
3. Run PIOP for zerocheck for polynomial  $p_A \cdot p_B - p_C$ .

$p_A$   $p_B$   $p_C$

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

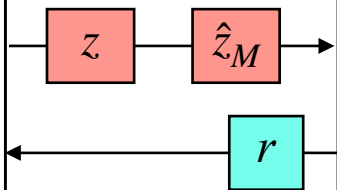
Run PIOP verifier for  
zerocheck for  
polynomial  
 $p_A \cdot p_B - p_C$ .

# Next point: *PIOP* for MV checks

## 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. Invoke sumcheck PIOP prover on

$$\hat{r}(X) \cdot \hat{z}_M(X) - \hat{r}_M(X) \cdot \hat{z}(X)$$



## 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 sumcheck PIOP verifier on

$$\hat{r}(X) \cdot \hat{z}_M(X) - \hat{r}_M(X) \cdot \hat{z}(X)$$

# Why is the verifier slow?

## Prover( $M, z$ )

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$$\hat{r}(X) \cdot \hat{z}_M(X) - \hat{r}_M(X) \cdot \hat{z}(X)$$

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

## 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  $(\hat{r}, \hat{r}_M)$

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

Invoke sumcheck PIOP prover on

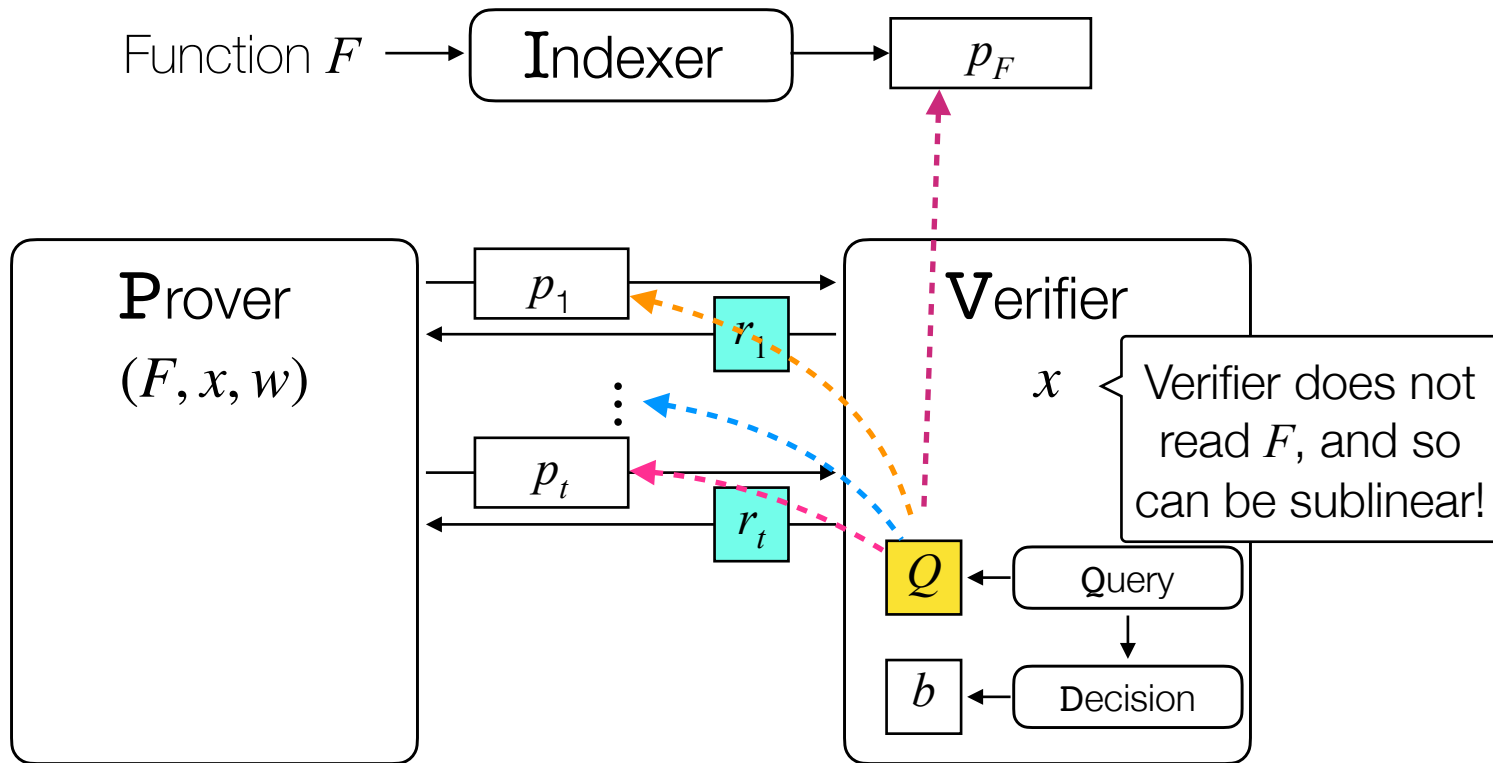
$$\hat{r}(X) \cdot \hat{z}_M(X) - \hat{r}_M(X) \cdot z(X)$$

# Sublinear verification for PIOP-based SNARKs

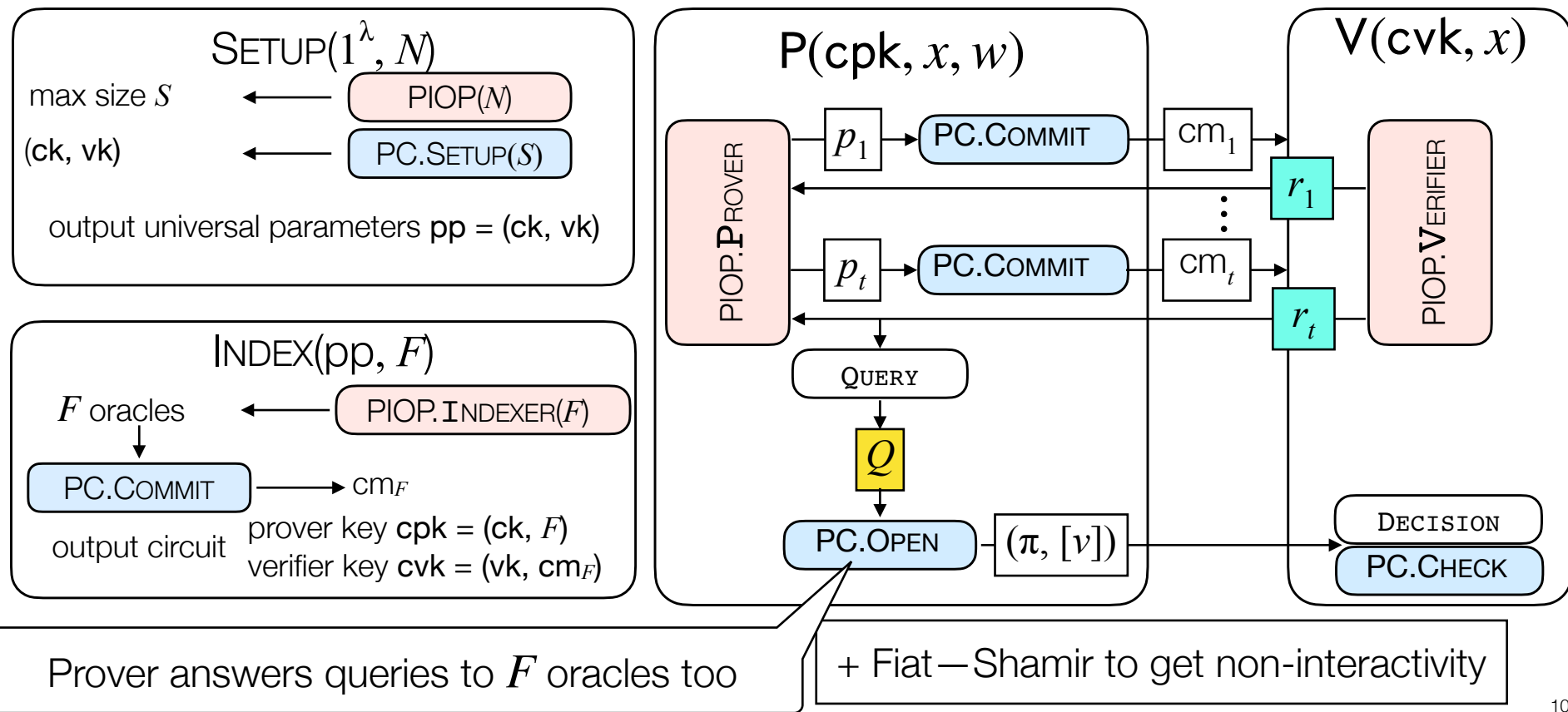


# Holographic PIOPs [CHMMVW20, COS20]

Introduce a new algorithm to preprocess the matrices



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



# 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!

# Holographic PIOP for R1CS

# 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. Interpolate  $(\vec{r}, \vec{r}^\top M)$  to get  $(\hat{r}, \hat{r}_M)$
4. Invoke sumcheck PIOP prover on

$$\hat{r}(X) \cdot \hat{z}_M(X) - \hat{r}_M(X) \cdot \hat{z}(X)$$

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

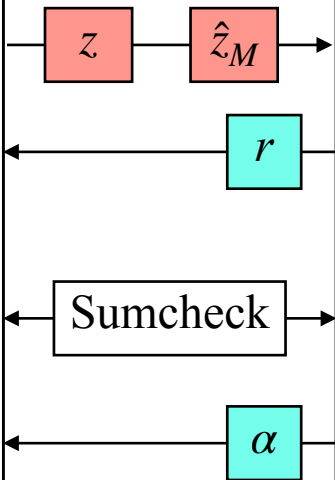
## Verifier( $M$ )

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

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

4. Invoke sumcheck PIOP verifier on

$$\hat{r}(X) \cdot \hat{z}_M(X) - \hat{r}_M(X) \cdot z(X)$$



# 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( $M, z$ )

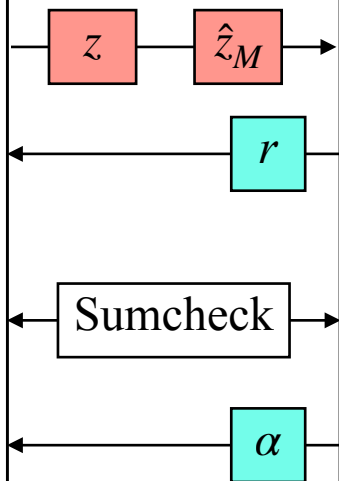
1. Compute  $z_M := Mz$
2. Interpolate  $z_M$  over  $H$  to get  $\hat{z}_M$
3. Interpolate  $(\vec{r}, \vec{r}^\top M)$  to get  $(\hat{r}, \hat{r}_M)$
4. Invoke sumcheck PIOP prover on

$$\hat{r}(X) \cdot \hat{z}_M(X) - \hat{r}_M(X) \cdot \hat{z}(X)$$

V can efficiently evaluate  $\hat{r}(\alpha)$ ; what about  $\hat{r}_M(\alpha)$  ?

## 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 sumcheck PIOP verifier on
- $$\hat{r}(X) \cdot \hat{z}_M(X) - \hat{r}_M(X) \cdot \hat{z}(X)$$





For all  $j \in H$ ,  $\hat{r}_M(j) = \sum_{i \in H} \hat{r}(i) \cdot \hat{M}(i, j)$ .

So, therefore the interpolation looks like  $\hat{r}_M(X) = \sum_{i \in H} \hat{r}(i) \cdot \hat{M}(i, X)$ , and so

$$\hat{r}_M(\alpha) = \sum_{i \in H} \hat{r}(i) \cdot \hat{M}(i, \alpha)$$

This is yet another sumcheck, so we engage in another sumcheck PIOP, which will eventually result in requiring an evaluation  $\hat{M}(\beta, \alpha)$ , where  $\beta$  is random

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

# 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 *only*  $m$  terms, so we only need  $m$  interpolations over  $K$ , i.e.  $\hat{r}, \hat{c}, \hat{v}$

Will replace  $r, c, v$  with their index of  $k$ -th non-zero entry

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# Final Matrix Encoding

Let  $m$  be the number of non-zero entries in the set of size  $m$ .  
Then, a *sparse* bivariate interpolation

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

Actually, we need polynomials, so we need bivariate  
interpolations over  $K$ , i.e.  $\hat{r}, \hat{c}, \hat{v}$

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

Q: How to do this sumcheck?

This is a rational function!

We only know how to do  
sumcheck for polynomials!

# Final Matrix Encoding

Let  $m$  be the number of non-zero entries in  $H$ , a set of size  $m$ .  
Then, a *sparse* bivariate interpolation

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

Actually, we need polynomials, so we need bivariate  
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Q: How to do this sumcheck?

This is a rational function!

We only know how to do  
sumcheck for polynomials!

# Final Matrix Encoding

Let  $m$  be the number of non-zero entries in  $A$ . Let  $K$  be a set of size  $m$ .  
Then, a *sparse* bivariate interpolation

$$\hat{M}(X, Y) := \sum_{k \in K} A_k \cdot \prod_{i=1}^n (X - \hat{r}_i(k)) \cdot \prod_{j=1}^n (Y - \hat{c}_j(k))$$

Actually, we need polynomials, so we use *polynomial* bivariate interpolations over  $K$ , i.e.  $\hat{r}, \hat{c}, \hat{v}$

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

A: interpolate into a polynomial!

# Rational $\rightarrow$ Polynomial

Key point: just like for other functions, here we just care about behavior of  $\frac{v_H(X)}{X - \hat{r}(k)}$  over  $H$

So we will replace with an interpolation  $p$ , and

1. Perform sumcheck with  $p$ , and
2. Check that  $p - \frac{v_H(X)}{X - \hat{r}(k)} = 0$  over  $H$ .



# What to do for multilinear case?

## ***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

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

Most  $M_{ij}$  are zero!

**Cannot** rewrite as  $\hat{M}(X, Y) := \sum_{i \in H} \sum_{j \in H} M_{ij} \cdot \frac{v_H(X)}{X - i} \cdot \frac{v_H(Y)}{Y - j} !$

# We can still try to exploit sparsity!

## ***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

$$\hat{M}(X, Y) := \sum_{i \in K} M_k \cdot \text{eq}(i, r(k)) \cdot \text{eq}(j, c(k))$$

# 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 *only*  $m$  terms, so we will replace  $r, c, v$  with  $\hat{r}, \hat{c}, \hat{v}$  interpolations over  $K$ , i.e.  $\hat{r}, \hat{c}, \hat{v}$

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