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Introduction

### **Distributed Lab**

# zkdl-camp.github.io

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Introduction

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## Recap on Poly-IOPs and NILPs

Almost all protocols we have seen so far work over **univariate**  $\mathbb{F}[X]$  **polynomials**. *Typical idea:* we aggregate information into some polynomial: say, p(X) (which is typically a combination of other polynomials), and then check whether p(u) = 0 for every  $u \in \Omega$ :

$$p(u) = 0$$
 for all  $u \in \Omega \iff p(X) = q(X) \prod_{u \in \Omega} (X - u)$ 

We check this inequality at a random point  $r \leftarrow \mathbb{F}$  and achieve soundness of  $1 - \deg p/|\mathbb{F}|$ : see **Schwartz-Zippel Lemma**.

For appropriate domains (typically  $\Omega = \{\omega^j\}_{j \in [2^j]}$ ) we reduce all polynomials computations to  $O(n \log n)$  complexity.

#### **Motivation**

But what if we could work with much smaller degrees?

## **Sum-Check-based Protocols**

- Instead of *n*-variate univariate polynomials  $\mathbb{F}[X]$  we reduce the problem to  $\log n$ -variate multivariate polynomials  $\mathbb{F}[X_1, \dots, X_v]$ .
- The Schwartz-Zippel Lemma still holds for multivariate case:

$$\Pr_{(r_1,\ldots,r_v)\leftarrow \mathbb{S}}[f(r_1,\ldots,r_v)=0] \leq \frac{\deg f}{|\mathbb{S}|}, \quad \mathbb{S} \subseteq \mathbb{F}^v$$

Bad news: divisibility theorems do not hold:

$$f(s_1,\ldots,s_v)=0\iff (X-s_1)\ldots(X-s_v)\mid f(X_1,\ldots,X_v)$$

## Sum-Check meaning

Instead of divisibility checks, we use the Sum-Check:

$$\sum_{b_1 \in \{0,1\}} \sum_{b_2 \in \{0,1\}} \cdots \sum_{b_v \in \{0,1\}} f(b_1, \dots, b_v) = H$$

## **Summary in a Nutshell**



## **Univariate World:**

$$p(X) = q(X) \prod_{i=1}^{n} (X - u)$$

#### **Multivariate World:**

$$\sum_{\mathbf{b}\in\{0,1\}^{\ell}} f(b_1,\ldots,b_{\ell}) = H$$

**Succinct Arguments:** Ligero (2022), Spartan (2020), Libra (2019), Hyrax(2017), GKR (2008), . . .

**Applications:** zkGPT (2025), deep-prove (using *Ceno*) (2024), vSQL (2017), ...

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## Some technicalities

## **Definition (Monomial)**

By **monomial** in  $\nu$  variables we call expression  $\mu(\mathbf{X}) = X_1^{a_1} \dots X_{\nu}^{a_{\nu}}$  where  $a_1, \dots, a_{\nu}$  are non-negative integers. The **degree** of a monomial is naturally defined as  $\deg \mu \triangleq a_1 + \dots + a_{\nu}$ .

## Definition (Multivariate Polynomial)

Function  $f: \mathbb{F}^{\nu} \to \mathbb{F}$  is called an  $\nu$ -variate polynomial, which is denoted by  $f \in \mathbb{F}[X_1, \dots, X_{\nu}]$ , if  $f(\mathbf{X})$  is a finite linear combination of  $\nu$ -variate monomials  $\{\mu_1(\mathbf{X}), \dots, \mu_n(\mathbf{X})\}$ . The **degree** of f is defined as  $\deg f \triangleq \max_{i \in \{1, \dots, n\}} \deg \mu_i$ .

### Example

 $f(X_1,X_2,X_3) = X_1^3 + 3X_1^2X_2^2 + X_3^2 + X_3$  is a linear combination of 3-variate monomials  $\{X_1^3,X_1^2X_2^2,X_3^2,X_3\}$ , thus  $f \in \mathbb{F}[X_1,X_2,X_3]$ . It has degree  $\deg f = 4$ , corresponding to the second monomial  $X_1^2X_2^2$ .

## **Multilinear Polynomials**

## **Definition (Multilinear Polynomial)**

Multivariate polynomial  $f \in \mathbb{F}[X_1, \dots, X_{\nu}]$  is called **multilinear** if it is linear in each of the variables. Formally we have:

$$f(X_1,\ldots,X_{\nu})=\mathbf{a}_jX_j+\beta_j,\quad j\in\{1,\ldots,n\},$$

where  $a_j$ ,  $\beta_j$  do not depend on  $X_j$ .

#### Example

For example,  $f(X_1, X_2, X_3) = X_1X_2 + 3X_1X_3 + X_2X_3$  is a multilinear polynomial in 3 variables. For instance, for  $X_1$ :

$$f(X_1, X_2, X_3) = (X_2 + 3X_3)X_1 + X_2X_3.$$

Similarly,  $f(X_1, ..., X_{\nu}) = \prod_{i=1}^{\nu} X_i$  is a multilinear polynomial.

## **Boolean Hypercube**

## Definition (Boolean Hypercube)

By v-dimensional boolean hypercube we simply denote the set  $\{0, 1\}^v$  (which is a set of binary strings of length v).

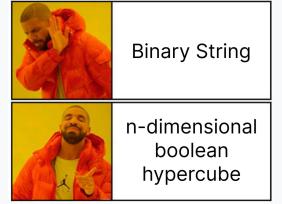
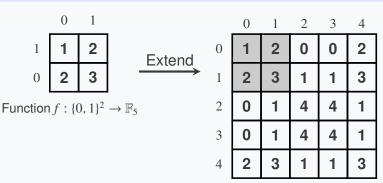


Figure: Cryptographers love overcomplicating things.

## Definition (Boolean Hypercube Extension)

Multivariate Polynomials, Multilinear Extensions,

Suppose we are given the values on the boolean cybercube  $f: \{0,1\}^{\nu} \to \mathbb{F}$ . We call  $\widetilde{f}: \mathbb{F}^{\nu} \to \mathbb{F}$  an **extension** if  $f(\mathbf{b}) = f(\mathbf{b})$  for every **b**  $\in \{0, 1\}^{\nu}$ .



Extension 
$$\widetilde{f}: \mathbb{F}_5^2 \to \mathbb{F}_5$$
  
 $\widetilde{f}(X_1, X_2) = X_1^2 + X_2^2 + 1$ 

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## **Multilinear Extensions**

## **Definition (MLE)**

An extension  $\widetilde{f}: \mathbb{F}^{\nu} \to \mathbb{F}$  of  $f: \{0,1\}^{V} \to \mathbb{F}$  is called **multilinear** if  $\widetilde{f} \in \mathbb{F}[X_1, \dots, X_v]$  is a multilinear polynomial.

## Example

For the previous example f(0,0) = 1, f(1,0) = f(0,1) = 2, f(1,1) = 3(over  $\mathbb{F}_5$ ) the multilinear extension is given by  $f(X_1, X_2) = X_1 + X_2 + 1$ .

The question though is how many extensions  $\tilde{f}$  we can build.

- If  $\widetilde{f}$  is a multivariate polynomial, there might be infinite number of choices: for example above, take  $\widetilde{f}_n(X_1, X_2) = X_1^n + X_2^n + 1$ .
- However, if  $\widetilde{f}$  is multilinear, it is **unique**.
- Additionally, is there an analogy to the Lagrange Interpolation for such case: how to build f practically?

## Lagrange Interpolation of multilinear polynomials

## Theorem (Lagrange Interpolation of Multilinear Polynomials)

Any function over the v-dimensional hypercube  $f: \{0,1\}^v \to \mathbb{F}$  has a unique v-variate multilinear extension  $\widetilde{f} \in \mathbb{F}[X_1,\ldots,X_v]$ . It is defined using the **Lagrange interpolation of multilinear polynomials**:

$$\widetilde{f}(\mathbf{X}) = \sum_{\boldsymbol{b} \in \{0,1\}^v} f(\boldsymbol{b}) \cdot \mathsf{eq}(\mathbf{X}; \boldsymbol{b}),$$

where the set  $\{eq(X; b)\}_{b \in \{0,1\}^{\nu}}$  is referred to as **the set of multilinear Lagrange basis polynomials** over  $\{0,1\}^{\nu}$ . Each basis polynomial (among  $2^{\nu}$ ) eq(X; b) is defined as:

$$eq(X; b) \triangleq \prod_{i=1}^{\nu} \{X_i b_i + (1 - X_i)(1 - b_i)\}.$$

## Lagrange Interpolation: Example

Suppose we want to build the MLE for  $f: \{0,1\}^2 \to \mathbb{F}_{11}$  given by:

$$f(0,0) = 3$$
,  $f(0,1) = 4$ ,  $f(1,0) = 1$ ,  $f(1,1) = 2$ 

Step 1. Define multilinear Lagrange basis polynomials:

$$\begin{split} \mathsf{eq}(X_1,X_2;(0,0)) &= (1-X_1)(1-X_2), & \mathsf{eq}(X_1,X_2;(0,1)) &= (1-X_1)X_2, \\ \mathsf{eq}(X_1,X_2;(1,0)) &= X_1(1-X_2), & \mathsf{eq}(X_1,X_2;(1,1)) &= X_1X_2 \end{split}$$

**Step 2.** Find the appropriate linear combination:

$$\begin{split} \widetilde{f}(X_1, X_2) &= \sum_{\boldsymbol{b} \in \{0, 1\}^2} f(\boldsymbol{b}) \cdot \text{eq}(\mathbf{X}; \boldsymbol{b}) \\ &= 3(1 - X_1)(1 - X_2) + 4(1 - X_1)X_2 + X_1(1 - X_2) + 2X_1X_2 \\ &= \boxed{-2X_1 + X_2 + 3} \end{split}$$

**Fact:** Generally,  $\widetilde{f}(r)$  for  $r \leftarrow \$ \mathbb{F}^{v}$  can be computed in  $O(2^{v})$  time.

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## MatMul Sum-Check Protocol

#### **MatMul Protocol**

**Goal:** Verify that C = AB for matrices  $A, B, C \in \mathbb{F}^{n \times n}$ .

**Naïve approach:** Send A, B to the verifier  $\mathcal{V}$ , then  $\mathcal{V}$  computes AB.

Time Complexity:  $O(n^3)$ , Space Complexity:  $O(n^2)$ .

**Freiveld's Protocol:** Send A, B, C to the verifier  $\mathcal{V}$ . Sample  $r \leftarrow \mathbb{F}^n$  and verify that A(Br) = Cr.

Time Complexity:  $O(n^2)$ , Space Complexity:  $O(n^2)$ .

**Sum-Check Protocol.** Apply the sum-check to some particular equation formed by multilinear extensions of matrices.

Time Complexity:  $O(n^2)$ , Space Complexity:  $O(\log n)$ .

#### Question

Where the hell should the Sum-Check be applied here?

## **Matrices Multilinear Extensions**

For simplicity assume  $n = 2^k$  for some k.

**Idea:** Instead of perceiving A, B, C as the collection of  $n^2$  field elements, perceive them as functions

 $f_A, f_B, f_C : \{0, 1\}^{\log n} \times \{0, 1\}^{\log n} \to \mathbb{F}$ , mapping two binarized indices of the matrix to the corresponding value. For example,

$$f_A(i,j) = A_{i,j}$$
, where  $i = (i_1, \dots, i_{\log n}), j = (j_1, \dots, j_{\log n}).$ 

## Example

Suppose 
$$A = \begin{bmatrix} 0 & 1 \\ 2 & 0 \end{bmatrix} \in \mathbb{F}_{13}^{2 \times 2}$$
. Then,  $f_A$  is defined as:

$$f_A(0,0) = f_A(1,1) = 0$$
,  $f_A(0,1) = 1$ ,  $f_A(1,0) = 2$ 

Its MLE is given by  $\widetilde{f}_A(X, Y) = (1 - X)Y + 2X(1 - Y) = 2X + Y - 3XY$ .

## The Trick

Given functions  $f_A, f_B, f_C : \{0, 1\}^{\log n} \times \{0, 1\}^{\log n} \to \mathbb{F}$ , we build the corresponding MLEs  $\widetilde{f}_A, \widetilde{f}_B, \widetilde{f}_C : \mathbb{F}^{\log n} \times \mathbb{F}^{\log n} \to \mathbb{F}$ . Now what?

#### Lemma

$$\widetilde{f}_C(\mathbf{x}, \mathbf{y}) = \sum_{\boldsymbol{b} \in \{0,1\}^{\log n}} \widetilde{f}_A(\mathbf{x}, \boldsymbol{b}) \cdot \widetilde{f}_B(\boldsymbol{b}, \mathbf{y})$$

**Reasoning.** Both sides are multilinear polynomials in  $\mathbf{x}$  and  $\mathbf{y}$ . Since MLE over  $\{0,1\}^{2\log n}$  is unique, it suffices to check the equality only over  $i,j\in\{0,1\}^{\log n}$ . Then,

$$\widetilde{f}_C(i,j) = \sum_{m{b} \in \{0,1\}^{\log n}} \widetilde{f}_A(i,m{b}) \cdot \widetilde{f}_B(m{b},j)$$

Note that this is exactly the check  $C_{i,j} = \sum_{b=1}^{n} A_{i,b} B_{b,j}!$ 

Now, apply Sum-Check on  $h(\mathbf{z}) = \widetilde{f}_A(r_1, \mathbf{z})\widetilde{f}_B(\mathbf{z}, r_2)$  for  $r_1, r_2 \leftarrow \mathbb{F}^{\log n}$ .

## Idea of Spartan

This protocol might sound too abstract, but this idea of using matrix MLEs is used in **Spartan**! Recall that in QAP we check:

QAP Check: 
$$\sum_i z_i \ell_i(X) \cdot \sum_i z_i r_i(X) - \sum_i z_i o_i(X) = 0, \quad X \in \Omega$$

## Spartan General Idea:

- 1. Commit to the MLE extension  $f_Z(Y)$  of the solution-witness z.
- 2. In universal setup, find MLEs  $\widetilde{f}_L$ ,  $\widetilde{f}_R$ ,  $\widetilde{f}_O$ .
- 3. Reduce R1CS satisfability to zero-check on

$$\zeta(\mathbf{X}) = \sum_{\boldsymbol{b}} \widetilde{f}_L(\mathbf{X}, \boldsymbol{b}) \widetilde{f}_Z(\boldsymbol{b}) \cdot \sum_{\boldsymbol{b}} \widetilde{f}_R(\mathbf{X}, \boldsymbol{b}) \widetilde{f}_Z(\boldsymbol{b}) - \sum_{\boldsymbol{b}} \widetilde{f}_O(\mathbf{X}, \boldsymbol{b}) \widetilde{f}_Z(\boldsymbol{b})$$

4. Apply some dark magic to make above work.

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## **Sum-Check Protocol Goal**

#### Sum-Check Goal

Prover  $\mathcal{P}$  wants to convince the verifier  $\mathcal{V}$  that:

$$\sum_{b_1 \in \{0,1\}} \sum_{b_2 \in \{0,1\}} \cdots \sum_{b_{\nu} \in \{0,1\}} f(b_1, \dots, b_{\nu}) = H$$

**Note:**  $f \in \mathbb{F}[X_1, \dots, X_{\nu}]$  is not necessarily a multilinear polynomial.

**Naïve IP:**  $\mathcal{V}$  takes f and computes the sum. It requires the time and space  $O(2^{\nu})$  — not gud for *succinct* argument systems.

**Round 1.** The prover  $\mathcal{P}$  sends the value  $C_1 \in \mathbb{F}$  which is the claimed value of H. Then, the prover computes:

$$f_1(X_1) := \sum_{(b_2,\dots,b_{\nu})\in\{0,1\}^{\nu-1}} f(X_1,b_2,\dots,b_{\nu})$$

**Question:** If f is multilinear, then what form does  $f_1$  have?

## Sum-Check, Round #1

Assume prover sends  $s_1(X_1)$ . Verify  $\mathcal V$  needs to check:

- 1.  $s_1(X_1)$  is indeed  $f_1(X_1)$ .
- 2.  $s_1(X_1)$  (and thus  $f_1(X_1)$ ) is consistent with the claimed  $C_1$ .

Second is easy: check  $s_1(0) + s_1(1) = H$ . Indeed:

$$s_{1}(0) + s_{1}(1)$$

$$= \sum_{(b_{2},...,b_{\nu})\in\{0,1\}^{\nu-1}} f(0,b_{2},...,b_{\nu}) + \sum_{(b_{2},...,b_{\nu})\in\{0,1\}^{\nu-1}} f(1,b_{2},...,b_{\nu})$$

$$= \sum_{(b_{1},...,b_{\nu})\in\{0,1\}^{\nu}} f(b_{1},...,b_{\nu}) = H$$

For the second, apply the Schwartz-Zippel Lemma: pick  $r_1 \leftarrow \mathbb{F}$  and check whether  $s_1(r_1) = f_1(r_1)$ . Computing  $s_1(r_1)$  is trivial, but how to compute  $f_1(r_1)$  effectively?

## Sum-Check, Subsequent Rounds

Idea: Apply the same procedure again!

**Round 2.** The prover  $\mathcal{P}$  computes  $s_2(X_2)$  claimed to equal:

$$f_2(X_2) = \sum_{\substack{(b_3,\dots,b_{\nu})\in\{0,1\}^{\nu-2}}} f(r_1,X_2,b_3,\dots,b_{\nu}).$$

For the consistency check,  $\mathcal{V}$  verifies that  $s_2(0) + s_2(1) = s_1(r_1)$ . Then, the verification boils down to checking whether  $s_2(r_2) = f_2(r_2)$ .

**Round** *j*. The prover  $\mathcal{P}$  computes  $s_j(X_j)$  claimed to equal:

$$f_j(X_j) = \sum_{(b_{j+1},\ldots,b_{\nu})\in\{0,1\}^{\nu-j}} f(r_1,\ldots,r_{j-1},X_j,b_{j+1},\ldots,b_{\nu}).$$

For the consistency check,  $\mathcal{V}$  verifies that  $s_j(0) + s_j(1) = s_{j-1}(r_{j-1})$ . Then, the verification boils down to checking whether  $s_j(r_j) = f_j(r_j)$ .

## Sum-Check, Wrap-up

**<u>Last Round.</u>** The verifier  $\mathcal{V}$  picks a random  $r_{\nu} \leftarrow \mathbb{F}$  and verifies whether  $s_{\nu}(r_{\nu}) = O^f(r_1, \ldots, r_{\nu})$  where  $O^f(\cdot)$  is an oracle access to the function f (e.g., commitment + opening).

### Lemma (Sum-Check Soundness Lemma)

Let  $f \in \mathbb{F}[X_1, \dots, X_{\nu}]$  be a multivariate polynomial of degree at most d in each variable, defined over the finite field  $\mathbb{F}$ . For any given  $H \in \mathbb{F}$ , let  $\mathcal{L}$  be the language of all polynomials f (given as an oracle) such that

$$H = \sum_{b_1 \in \{0,1\}} \sum_{b_2 \in \{0,1\}} \cdots \sum_{b_{\nu} \in \{0,1\}} f(b_1, \dots, b_{\nu}).$$

The sumcheck protocol is an IOP for  $\mathcal{L}$  with the completness error  $\delta_C = 0$  and the soundness error  $\delta_S \leq vd/|\mathbb{F}|$ .

## **Sum-Check Performance**

### Lemma (Sum-Check Performance)

Assume the average cost of calling  $O^f(\cdot)$  is T,  $d = \deg f$ , and  $n = 2^{\nu}$ . Then, the following is true about the performance of Sum-Check:

- **Proof Size:**  $O(d \log n)$ .
- *Verifier Time:*  $O(d \log n) + T$ .
- Prover Time: O(nT).

However, this is an IP. How to turn it to the *non-interactive* protocol?

Simply apply the **Fiat-Shamir heuristic**! At round j, the transcript is  $\tau = (H, s_1, r_1, \dots, s_{j-1}, r_{j-1}, s_j)$ , thus the randomness can be sampled simply as  $r_j \leftarrow O^R(\tau)$  for a random oracle  $O^R(\cdot)$ .

### Now the coding time!

# Thank you for your attention



