

# Polishchuk-Spielman Bivariate Testing and An Application

Ziyang Jin<sup>1</sup>

<sup>1</sup>Theory Student Seminar  
University of Toronto  
27 Feb 2025

# Table of Contents

Introduction

The Main Theorem

Application of Polishchuk-Spielman

# Definitions

Let  $\mathcal{F}$  be a finite field. We consider bivariate polynomials over a domain  $X \times Y$ , where  $X = \{x_1, \dots, x_n\} \subseteq \mathcal{F}$  and  $Y = \{y_1, \dots, y_n\} \subseteq \mathcal{F}$ .

A polynomial  $p(x, y)$  has degree  $(d, e)$  if it has degree at most  $d$  in  $x$  and degree at most  $e$  in  $y$ . When we say a polynomial of degree  $d$ , we mean a polynomial of degree at most  $d$ . We use them interchangeably.

Suppose we have a function  $f(x, y)$  on  $X \times Y$ . We can represent  $f(x, y)$  in matrix form as follows:

$$M = \begin{pmatrix} f(x_1, y_1) & f(x_1, y_2) & \dots & f(x_1, y_n) \\ f(x_2, y_1) & f(x_2, y_2) & \dots & f(x_2, y_n) \\ \vdots & \vdots & \ddots & \vdots \\ f(x_n, y_1) & f(x_n, y_2) & \dots & f(x_n, y_n) \end{pmatrix}.$$

# Rows and Columns of the Matrix

If we look at each column  $j \in [n]$ , then  $y_j$  is fixed. Each column can be viewed as a univariate function with variable  $x$  evaluated on  $X$ .

$$M = \begin{pmatrix} f(x_1, y_1) & f(x_1, y_2) & \dots & f(x_1, y_n) \\ f(x_2, y_1) & f(x_2, y_2) & \dots & f(x_2, y_n) \\ \vdots & \vdots & \ddots & \vdots \\ f(x_n, y_1) & f(x_n, y_2) & \dots & f(x_n, y_n) \end{pmatrix}.$$

If we look at each row  $i \in [n]$ , then  $x_i$  is fixed. Each row can be viewed as a univariate function with variable  $y$  evaluated on  $Y$ .

# Matrix Representation of a Function

Suppose an adversary gives us a matrix over  $\mathcal{F}$

$$M = \begin{pmatrix} v_{1,1} & v_{1,2} & \dots & v_{1,n} \\ v_{2,1} & v_{2,2} & \dots & v_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ v_{n,1} & v_{n,2} & \dots & v_{n,n} \end{pmatrix}.$$

This can be viewed exactly the same as

$$M = \begin{pmatrix} f(x_1, y_1) & f(x_1, y_2) & \dots & f(x_1, y_n) \\ f(x_2, y_1) & f(x_2, y_2) & \dots & f(x_2, y_n) \\ \vdots & \vdots & \ddots & \vdots \\ f(x_n, y_1) & f(x_n, y_2) & \dots & f(x_n, y_n) \end{pmatrix},$$

because  $M$  uniquely defines  $f(x, y)$ .

# Well-Known Theorem

**Question:** *How do we know if matrix  $M$  represents a bivariate polynomial?*



# Well-Known Theorem

**Question:** *How do we know if matrix  $M$  represents a bivariate polynomial?*

## Theorem (Well-known)

*Let  $f(x, y)$  be a function on  $X \times Y$  such that for  $j \in [n]$ ,  $f(x, y_j)$  agrees with some degree  $d$  polynomial in  $x$  on  $X$ , and for  $i \in [n]$ ,  $f(x_i, y)$  agrees on  $Y$  with some degree  $e$  polynomial in  $y$ . Then, there exists a polynomial  $P(x, y)$  of degree  $(d, e)$  such that  $f(x, y)$  agrees with  $P(x, y)$  everywhere on  $X \times Y$ .*

**Proof:** Recall that a degree  $d$  univariate polynomial is uniquely determined by its values at  $d+1$  points. For  $1 \leq j \leq e+1$ , let  $p_j(x)$  be the degree  $d$  polynomial that agrees with  $f(x, y_j)$ . For  $1 \leq j \leq e+1$ , let  $\delta_j(y)$  be the degree  $e$  polynomial in  $y$  such that

$$\delta_j(y_k) = \begin{cases} 1, & \text{if } j = k, \text{ and} \\ 0, & \text{if } 1 \leq k \leq e+1, \text{ but } j \neq k. \end{cases}$$

We let  $P(x, y) = \sum_{j=1}^{e+1} \delta_j(y)p_j(x)$ . It is clear that  $P$  has degree  $(d, e)$ . Moreover,  $P(x, y_j) = f(x, y_j)$  for all  $x \in X$  and  $1 \leq j \leq d+1$ . To see that in fact  $P(x, y) = f(x, y)$  for all  $(x, y) \in X \times Y$ , observe that  $P$  and  $f$  agree at  $e+1$  points in column  $y$ . Since  $f$  agrees with some degree  $e$  polynomial in column  $y$ , that polynomial must be the restriction of  $P$  to column  $y$ .  $\square$



## Proof Explained

Every column  $p_j$  has degree  $d$ . We pick the first  $e + 1$  columns.

$$M = \begin{pmatrix} p_1(x_1) & p_2(x_1) & \dots & p_{e+1}(x_1) & \dots & f(x_1, y_n) \\ p_1(x_2) & p_2(x_2) & \dots & p_{e+1}(x_2) & \dots & f(x_2, y_n) \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ p_1(x_n) & p_2(x_n) & \dots & p_{e+1}(x_n) & \dots & f(x_n, y_n) \end{pmatrix}.$$

By Lagrange Interpolation, each  $\delta_j$  has degree  $e$ .

$$\delta_1(y) = (1 \quad 0 \quad \dots \quad 0 \quad \dots \quad 0) = \frac{(y - y_2)}{(y_1 - y_2)} \frac{(y - y_3)}{(y_1 - y_3)} \dots \frac{(y - y_{e+1})}{(y_1 - y_{e+1})}$$

$$\delta_j(y_k) = \prod_{j=1, j \neq k}^{e+1} \frac{y - y_j}{y_k - y_j}.$$

The bivariate polynomial is

$$P(x, y) = \sum_{j=1}^{e+1} \delta_j(y) p_j(x).$$

# Applying the Well-Known Theorem

Suppose an adversary gives you a matrix

$$M = \begin{pmatrix} v_{1,1} & v_{1,2} & \dots & v_{1,n} \\ v_{2,1} & v_{2,2} & \dots & v_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ v_{n,1} & v_{n,2} & \dots & v_{n,n} \end{pmatrix},$$

and you want to know if  $M$  represents some bivariate polynomial of degree  $(d, d)$ . *What can you do?*

# Applying the Well-Known Theorem

Suppose an adversary gives you a matrix

$$M = \begin{pmatrix} v_{1,1} & v_{1,2} & \dots & v_{1,n} \\ v_{2,1} & v_{2,2} & \dots & v_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ v_{n,1} & v_{n,2} & \dots & v_{n,n} \end{pmatrix},$$

and you want to know if  $M$  represents some bivariate polynomial of degree  $(d, d)$ . *What can you do?*

- ▶ We can test if a row/column agrees with some polynomial of degree  $d$  by interpolating any  $d + 1$  points and check if all other points lie on the polynomial.
- ▶ If we know that every row agrees with some polynomial of degree  $d$ , and every column agrees with some polynomial of degree  $d$ . We can apply the well-known theorem we just saw.

# An Imperfect World

What if some rows/columns do not fully agree with some polynomial of degree  $d$ ?

**Question:** *How do we know if matrix  $M$  is “very close” to a bivariate polynomial?*

# An Imperfect World

Maybe we can fix some places such that every row agrees with some polynomial of degree at most  $d$ .

$$M = \begin{pmatrix} v_{1,1} & \color{red}{v_{1,2}} & \cdots & v_{1,n} \\ v_{2,1} & v_{2,2} & \cdots & \color{red}{v_{2,n}} \\ \vdots & \vdots & \ddots & \vdots \\ \color{red}{v_{n,1}} & v_{n,2} & \cdots & \color{red}{v_{n,n}} \end{pmatrix}.$$

Maybe we can fix some places such that every column agrees with some polynomial of degree at most  $d$ .

$$M = \begin{pmatrix} \color{blue}{v_{1,1}} & v_{1,2} & \cdots & v_{1,n} \\ v_{2,1} & \color{blue}{v_{2,2}} & \cdots & \color{blue}{v_{2,n}} \\ \vdots & \vdots & \ddots & \vdots \\ \color{blue}{v_{n,1}} & v_{n,2} & \cdots & v_{n,n} \end{pmatrix}.$$

*Hard to fix both the rows and the columns simultaneously!*

## Rows and Columns

Consider a bivariate polynomial  $R(x, y)$  of degree  $(d, n)$ . Every row  $f_i$  is a univariate polynomial in  $x$  with degree at most  $d$ .

$$R(x, y) = \begin{pmatrix} f_1(x_1) & f_1(x_2) & \dots & f_1(x_n) \\ f_2(x_1) & f_2(x_2) & \dots & f_2(x_n) \\ \vdots & \vdots & \ddots & \vdots \\ f_n(x_1) & f_n(x_2) & \dots & f_n(x_n) \end{pmatrix}.$$

Consider a bivariate polynomial  $C(x, y)$  of degree  $(n, d)$ . Every column  $g_j$  is a univariate polynomial in  $y$  with degree at most  $d$ .

$$C(x, y) = \begin{pmatrix} g_1(y_1) & g_2(y_1) & \dots & g_n(y_1) \\ g_1(y_2) & g_2(y_2) & \dots & g_n(y_2) \\ \vdots & \vdots & \ddots & \vdots \\ g_1(y_n) & g_2(y_n) & \dots & g_n(y_n) \end{pmatrix}.$$



# Polishchuk-Spielman Bivariate Testing Theorem

**Theorem 9 (Bivariate Testing).** *Let  $\mathcal{F}$  be a field, let  $X = \{x_1, \dots, x_n\} \subseteq \mathcal{F}$ , and let  $Y = \{y_1, \dots, y_n\} \subseteq \mathcal{F}$ . Let  $R(x, y)$  be a polynomial over  $\mathcal{F}$  of degree  $(d, n)$  and let  $C(x, y)$  be a polynomial over  $\mathcal{F}$  of degree  $(n, d)$ . If*

$$\text{Prob}_{(x,y) \in X \times Y} [R(x, y) \neq C(x, y)] < \delta^2,$$

*and  $n > 2\delta n + 2d$ , then there exists a polynomial  $Q(x, y)$  of degree  $(d, d)$  such that*

$$\text{Prob}_{(x,y) \in X \times Y} [R(x, y) \neq Q(x, y) \text{ or } C(x, y) \neq Q(x, y)] < 2\delta^2.$$

# What does it mean?

We can fix some places in  $M$  to obtain  $R(x, y)$ , and separately fix some other places in  $M$  to obtain  $C(x, y)$ . If the total number of places we fixed among both  $R(x, y)$  and  $C(x, y)$  is at most  $\delta^2 n^2$ , then  $M$  is actually very close to a bivariate polynomial  $Q(x, y)$  of degree  $(d, d)$ .

$$M = \begin{pmatrix} v_{1,1} & v_{1,2} & \dots & v_{1,n} \\ v_{2,1} & v_{2,2} & \dots & v_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ v_{n,1} & v_{n,2} & \dots & v_{n,n} \end{pmatrix}.$$

(PS: In practice, if  $M$  is really close to  $Q(x, y)$ . Observe that we can view  $M$  as a bivariate Reed-Muller code, then we can recover  $Q(x, y)$ .)



# Proof 1

## Lemma (3)

*Let  $S \subset X \times Y$  be a set of size at most  $(\delta n)^2$ , where  $\delta n$  is an integer. Then there exists a non-zero polynomial  $E(x, y)$  of degree  $(\delta n, \delta n)$  such that  $E(x, y) = 0$  for all  $(x, y) \in S$ .*

The proof is obvious:  $E(x, y)$  has  $(\delta n + 1)^2$  unknowns and there are  $(\delta n)^2$  restrictions.

Let  $S$  be the subset of  $X \times Y$  on which  $R$  and  $C$  disagree. Then we have

$$R(x, y)E(x, y) = C(x, y)E(x, y) \text{ for all } (x, y) \in X \times Y.$$

Observe  $C(x, y)E(x, y)$  is a polynomial of degree  $(n + \delta n, d + \delta n)$  and  $R(x, y)E(x, y)$  is a polynomial of degree  $(d + \delta n, n + \delta n)$ .

## Proof 2

By the well-known theorem, there exists a polynomial  $P(x, y)$  of degree  $(d + \delta n, d + \delta n)$  such that

$$R(x, y)E(x, y) = C(x, y)E(x, y) = P(x, y)$$

for all  $(x, y) \in X \times Y$ .

## Proof 2

By the well-known theorem, there exists a polynomial  $P(x, y)$  of degree  $(d + \delta n, d + \delta n)$  such that

$$R(x, y)E(x, y) = C(x, y)E(x, y) = P(x, y)$$

for all  $(x, y) \in X \times Y$ .

It is natural to continue the proof by dividing  $P$  by  $E$ . However, the most we can say is that

$$\frac{P(x, y)}{E(x, y)} = R(x, y) = C(x, y),$$

for all  $(x, y) \in X \times Y$  such that  $E(x, y) \neq 0$ .

## Proof 2

By the well-known theorem, there exists a polynomial  $P(x, y)$  of degree  $(d + \delta n, d + \delta n)$  such that

$$R(x, y)E(x, y) = C(x, y)E(x, y) = P(x, y)$$

for all  $(x, y) \in X \times Y$ .

It is natural to continue the proof by dividing  $P$  by  $E$ . However, the most we can say is that

$$\frac{P(x, y)}{E(x, y)} = R(x, y) = C(x, y),$$

for all  $(x, y) \in X \times Y$  such that  $E(x, y) \neq 0$ . We can show that if  $n$  is sufficiently large, then  $E$  in fact divides  $P$ .

# Proof 3

## Lemma (4)

*Let  $P(x, y), E(x, y), R(x, y), C(x, y)$  be polynomials of degree  $(\delta n + d, \delta n + d), (\delta n, \delta n), (d, n), (n, d)$  respectively such that  $R(x, y)E(x, y) = C(x, y)E(x, y) = P(x, y)$  for all  $(x, y) \in X \times Y$ . If  $|X| > \delta n + d$  and  $|Y| > \delta n + d$ , then for all  $y_0 \in Y$  and for all  $x_0 \in X$ ,  $P(x, y_0) \equiv R(x, y_0)E(x, y_0)$  and  $P(x_0, y) \equiv C(x_0, y)E(x_0, y)$ .*

The proof is obvious: For fixed  $y_0$ ,  $P(x, y_0)$  and  $R(x, y_0)E(x, y_0)$  both have degree  $\delta n + d$ , and they agree on at least  $d + \delta n + 1$  points.

## Proof 4

### Lemma (8)

*Let  $E(x, y)$  be a polynomial of degree  $(b, a)$  and let  $P(x, y)$  be a polynomial of degree  $(b + d, a + d)$ . If there exists distinct  $x_1, \dots, x_n$  such that  $E(x_i, y)$  divides  $P(x_i, y)$  for  $i \in [n]$ , distinct  $y_1, \dots, y_n$  such that  $E(x, y_i)$  divides  $P(x, y_i)$  for  $i \in [n]$  and if*

$$n > \min\{2b + 2d, 2a + 2d\},$$

*then  $E(x, y)$  divides  $P(x, y)$ .*

The proof is not obvious. We will skip it for time sake.

## Recall the main Theorem

**Theorem 9 (Bivariate Testing).** *Let  $\mathcal{F}$  be a field, let  $X = \{x_1, \dots, x_n\} \subseteq \mathcal{F}$ , and let  $Y = \{y_1, \dots, y_n\} \subseteq \mathcal{F}$ . Let  $R(x, y)$  be a polynomial over  $\mathcal{F}$  of degree  $(d, n)$  and let  $C(x, y)$  be a polynomial over  $\mathcal{F}$  of degree  $(n, d)$ . If*

$$\text{Prob}_{(x,y) \in X \times Y} [R(x, y) \neq C(x, y)] < \delta^2,$$

*and  $n > 2\delta n + 2d$ , then there exists a polynomial  $Q(x, y)$  of degree  $(d, d)$  such that*

$$\text{Prob}_{(x,y) \in X \times Y} [R(x, y) \neq Q(x, y) \text{ or } C(x, y) \neq Q(x, y)] < 2\delta^2.$$

## Proof 5

Summary of our proof so far: Let  $S$  be the set of points where  $R(x, y) \neq C(x, y)$ . By Lemma 3, there exists an error correcting polynomial  $E(x, y)$  of degree  $(\delta n, \delta n)$  such that  $E$  vanishes on  $S$ . By Lemma 4 and Lemma 8, there exists a polynomial  $Q(x, y)$  of degree  $(d, d)$  such that

$$R(x, y)E(x, y) = C(x, y)E(x, y) = Q(x, y)E(x, y),$$

for all  $(x, y) \in X \times Y$ .



## Proof 5

Summary of our proof so far: Let  $S$  be the set of points where  $R(x, y) \neq C(x, y)$ . By Lemma 3, there exists an error correcting polynomial  $E(x, y)$  of degree  $(\delta n, \delta n)$  such that  $E$  vanishes on  $S$ . By Lemma 4 and Lemma 8, there exists a polynomial  $Q(x, y)$  of degree  $(d, d)$  such that

$$R(x, y)E(x, y) = C(x, y)E(x, y) = Q(x, y)E(x, y),$$

for all  $(x, y) \in X \times Y$ .

Now we need to show the  $< 2\delta^2$  part. Note that in any row where  $E(x, y) \neq 0$ ,  $Q$  agrees with  $R$  on that entire row. However,  $E$  has degree  $(\delta n, \delta n)$  so it can be (in the worst case) identically zero on at most  $\delta n$  rows. So  $E$  must be non-zero on at least  $(1 - \delta)n$  rows. Thus,  $Q$  must agree with  $R$  on at least  $(1 - \delta)n$  rows. Similarly,  $Q$  must agree with  $C$  on at least  $(1 - \delta)n$  rows.

## Proof 6

Therefore, we have  $R$  and  $C$  agree on the intersection of  $(1 - \delta)n$  columns and rows. This is already a lot of points, but we will show that they agree on many more points.

Recall  $S$  is the set of points where  $R(x, y) \neq C(x, y)$ . Let  $T$  be the set of points where  $R(x, y) = C(x, y)$ , but  $Q(x, y) \neq R(x, y)$  (and also  $Q(x, y) \neq C(x, y)$ ). If we show  $|T| \leq |S|$ , we are done with the  $< 2\delta^2$  part.

We say a row/column is *bad* if  $Q$  disagrees on  $R/C$  on that row/column. Let  $b_r$  be the number of bad rows and let  $b_c$  be the number of bad columns. Call *good* any row/column that is not bad. We say that a row and column disagree if  $R$  and  $C$  take different values at their intersection.

## Proof 7

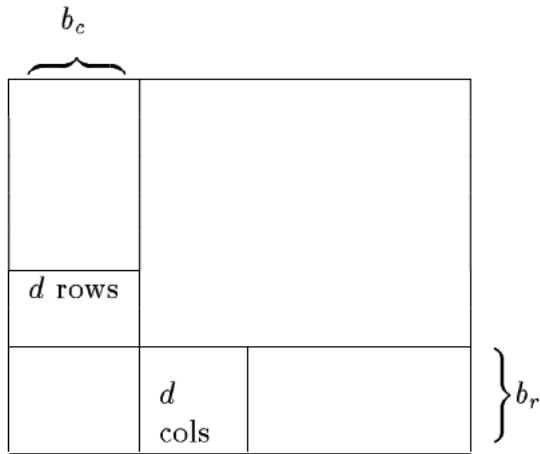
Observe there can be at most  $d + b_r$  points of  $T$  in any bad column: if a column has more than  $d + b_r$  points (e.g.  $d + b_r + 1$  points) of  $T$ , note that  $R(x, y) = C(x, y)$  in  $T$ , then it must have at least  $d + 1$  points in good rows where  $Q$  agrees with  $R$  and therefore  $Q$ , implying that column is in fact good.

Recall  $n > 2\delta n + 2d$ . Thus, every bad column must have at least  $n/2$  points of  $S$  in the intersection of that column with the good rows. Similarly, every bad row must have at least  $n/2$  points of  $S$  in the intersection of that row with the good columns.

Hence, the points of  $T$  in every column is less than the points of  $S$ ; the points of  $T$  in every row is less than the points of  $S$ .  
Therefore,  $|T| \leq |S|$ .

## Proof 8

Here is a picture illustration. The basic idea is that the points of  $T$  must lie in the lower left-hand corner.



# Testing Reed-Solomon Codeword

Let  $n$  be a natural number and let  $\Sigma$  be an alphabet. Let  $x \in \Sigma^n$  be a string, and we use  $x_i$  to denote the  $i$ th symbol of  $x$ . We say  $x$  is  $\delta$ -close to a string  $y \in \Sigma^n$  if  $|\{i \mid x_i \neq y_i\}| \leq \delta n$ . I.e.  $x$  and  $y$  agree on all but at most a  $\delta$ -fraction of the symbols.

Suppose you are given an array of values. How do you test that it is a Reed-Solomon codeword?

# Testing Reed-Solomon Codeword

Let  $n$  be a natural number and let  $\Sigma$  be an alphabet. Let  $x \in \Sigma^n$  be a string, and we use  $x_i$  to denote the  $i$ th symbol of  $x$ . We say  $x$  is  $\delta$ -close to a string  $y \in \Sigma^n$  if  $|\{i \mid x_i \neq y_i\}| \leq \delta n$ . I.e.  $x$  and  $y$  agree on all but at most a  $\delta$ -fraction of the symbols.

Suppose you are given an array of values. How do you test that it is a Reed-Solomon codeword?

What if you are only allowed to query very few values from the array?

# Testing Reed-Solomon Codeword

Let  $n$  be a natural number and let  $\Sigma$  be an alphabet. Let  $x \in \Sigma^n$  be a string, and we use  $x_i$  to denote the  $i$ th symbol of  $x$ . We say  $x$  is  $\delta$ -close to a string  $y \in \Sigma^n$  if  $|\{i \mid x_i \neq y_i\}| \leq \delta n$ . I.e.  $x$  and  $y$  agree on all but at most a  $\delta$ -fraction of the symbols.

Suppose you are given an array of values. How do you test that it is a Reed-Solomon codeword?

What if you are only allowed to query very few values from the array? Not possible!

# Testing Reed-Solomon Codeword

Let  $n$  be a natural number and let  $\Sigma$  be an alphabet. Let  $x \in \Sigma^n$  be a string, and we use  $x_i$  to denote the  $i$ th symbol of  $x$ . We say  $x$  is  $\delta$ -close to a string  $y \in \Sigma^n$  if  $|\{i \mid x_i \neq y_i\}| \leq \delta n$ . I.e.  $x$  and  $y$  agree on all but at most a  $\delta$ -fraction of the symbols.

Suppose you are given an array of values. How do you test that it is a Reed-Solomon codeword?

What if you are only allowed to query very few values from the array? Not possible! What if you can relax the requirement? You can build a **proof system** that shows the array of values is  $\delta$ -close to some Reed-Solomon codeword.



## Definition (PCP of Proximity)

A *probabilistically checkable proof of proximity* (PCPP) system with soundness error  $s \in (0, 1)$  and proximity parameter  $\delta \in (0, 1)$  is a probabilistic proof system  $(P, V)$  in which the prover  $P$  on input  $(x, w)$  generates a proof  $\pi$ , and the verifier  $V$  can make at most  $q$  queries to the combined oracle  $(x, \pi)$ , and the following holds.

- ▶ **Completeness:** For every  $(x, w) \in \mathcal{R}$  (which means  $x \in L_{\mathcal{R}}$ ),  $V$  accepts with probability 1.
- ▶ **Soundness:** For every  $x$  that is  $\delta$ -far from  $L_{\mathcal{R}}$ ,  $V$  accepts with probability at most  $s$ , regardless of the proof oracle  $\pi$ .

In this case, we write  $L_{\mathcal{R}} \in \text{PCPP}[r, q, \delta, s, \ell]$  where  $r$  is the verifier's randomness complexity,  $q$  is the query complexity, and  $\ell$  is the length of the proof. We say a PCPP is an **exact PCPP** if the proximity parameter  $\delta = 0$ .

# Application of Polishchuk-Spielman

## Theorem (Theorem 3.2 in Ben-Sasson Sudan 05)

*Let  $\mathbb{F}_q$  be a finite field of order  $q = 2^w$ . Let  $S$  be a subset of  $\mathbb{F}_q$  and  $S$  is  $\mathbb{F}_2$ -linear (i.e. for all  $a, b \in S$ , we have  $a + b \in S$ ). Then, for any soundness error  $s \in (0, 1)$  and any proximity parameter  $\delta \in (0, 1)$ , there exists an explicit construction of a PCPP to test if an array of values  $r_1, \dots, r_{|S|} \in \mathbb{F}_q$  is  $\delta$ -close to some univariate polynomial of degree  $d$  evaluated at  $S$ , and the PCPP has randomness complexity  $\log(q \cdot \text{polylog}(q))$ , query complexity  $\text{polylog}(q)$ , and proof length  $q \cdot \text{polylog}(q)$ .*

# Application of Polishchuk-Spielman

## Theorem (Theorem 3.2 in Ben-Sasson Sudan 05)

*Let  $\mathbb{F}_q$  be a finite field of order  $q = 2^w$ . Let  $S$  be a subset of  $\mathbb{F}_q$  and  $S$  is  $\mathbb{F}_2$ -linear (i.e. for all  $a, b \in S$ , we have  $a + b \in S$ ). Then, for any soundness error  $s \in (0, 1)$  and any proximity parameter  $\delta \in (0, 1)$ , there exists an explicit construction of a PCPP to test if an array of values  $r_1, \dots, r_{|S|} \in \mathbb{F}_q$  is  $\delta$ -close to some univariate polynomial of degree  $d$  evaluated at  $S$ , and the PCPP has randomness complexity  $\log(q \cdot \text{polylog}(q))$ , query complexity  $\text{polylog}(q)$ , and proof length  $q \cdot \text{polylog}(q)$ .*

The core idea in the construction is to lay out the array of values as a bivariate polynomial and apply Polishchuk-Spielman! Maybe a good topic for my next TSS.



Thank you

Thank you!