

# Foundations of Probabilistic Proofs

A course by **Alessandro Chiesa**

## Lecture 09

# Low-Degree Testing



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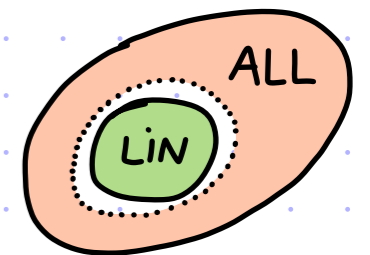
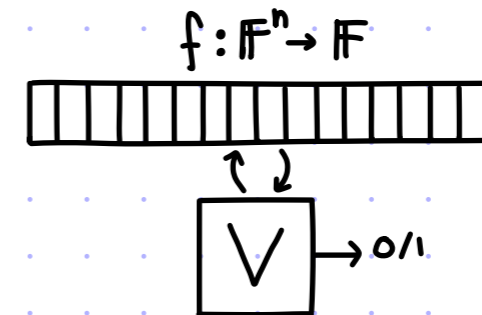
# Low-Degree Testing



Recall the goal of **LINEARITY TESTING**:

A test  $V_{LIN}$  s.t.  $\forall f: \mathbb{F}^n \rightarrow \mathbb{F}$

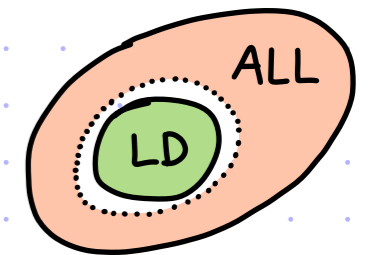
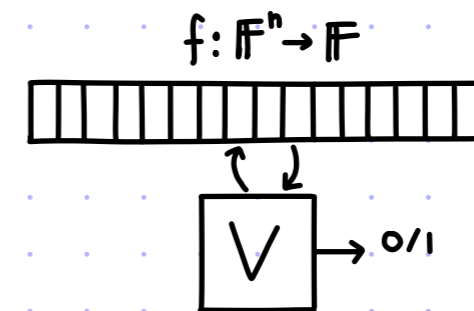
- completeness:  $f \in LIN[\mathbb{F}, n] \rightarrow \Pr[V_{LIN}^f = 1] = 1$
- soundness:  $\Delta(f, LIN[\mathbb{F}, n]) \geq \delta \rightarrow \Pr[V_{LIN}^f = 1] \leq \epsilon(\delta)$



The goal of **LOW-DEGREE TESTING** is:

A test  $V_{LD}$  s.t.  $\forall f: \mathbb{F}^n \rightarrow \mathbb{F}$

- completeness:  $f \in LD[\mathbb{F}, n, d] \rightarrow \Pr[V_{LD}^f = 1] = 1$
- soundness:  $\Delta(f, LD[\mathbb{F}, n, d]) \geq \delta \rightarrow \Pr[V_{LD}^f = 1] \leq \epsilon(\delta)$



What does degree  $d$  mean?

- **total degree**:  $LD[\mathbb{F}, n, \text{tot} \leq d]$  (e.g. in this case  $LD[\mathbb{F}, n, \text{tot} \leq 1] = \text{AFFINE}[\mathbb{F}, n]$ )
- **individual degree**:  $LD[\mathbb{F}, n, \text{ind} \leq d]$  (e.g. in this case  $LD[\mathbb{F}, n, \text{ind} \leq 1] = \text{"multilinear polynomials"}$ )

**EXERCISE**: derive a test for individual degree from a test for total degree.

In most applications the difference total-vs-individual does not matter much.

# The Case of Low Total Degree

Today we study low-degree testing for the case of **TOTAL** degree:

Individual degree is  $< |\mathbb{F}|$   
since  $\forall i \in [n], x_i^{|\mathbb{F}|} \equiv x_i$ .

$$LD[\mathbb{F}, n, \text{tot} \leq d] = \left\{ f: \mathbb{F}^n \rightarrow \mathbb{F} \mid \exists p \in \mathbb{F}[x_1, \dots, x_n] \text{ of TOTAL degree } \leq d \text{ s.t. } p \equiv f \right\}.$$

This set of functions is known as the **Reed-Muller code** (RM code).

The RM code is a **linear (error-correcting) code**:

$$LD[\mathbb{F}, n, \text{tot} \leq d] \text{ is an } \mathbb{F}\text{-linear subspace of } \mathbb{F}^n \quad (\forall f, g \in LD[\mathbb{F}, n, \text{tot} \leq d] \forall \alpha, \beta \in \mathbb{F}, \alpha f + \beta g \in LD[\mathbb{F}, n, \text{tot} \leq d]).$$

The code's parameters depend on the regime. In this course we consider  $d < |\mathbb{F}|$ :

Typically  $\mathbb{F}$  is large enough for soundness reasons.

- message length =  $\binom{n+d}{d}$ . Size of the set  $\{(d_1, \dots, d_n) \in \{0, 1, \dots, |\mathbb{F}|-1\} \mid \sum_{i=1}^n d_i \leq d\}$ .
- block length =  $|\mathbb{F}|^n$ . Size of a function  $f: \mathbb{F}^n \rightarrow \mathbb{F}$ .
- relative distance  $\geq 1 - \frac{d}{|\mathbb{F}|}$ . By the Polynomial Identity Lemma  $(\forall p \neq 0 \Pr_{\alpha \leftarrow \mathbb{F}^n} [p(\alpha) = 0] \leq \frac{\deg_{\text{tot}}(p)}{|\mathbb{F}|})$ .

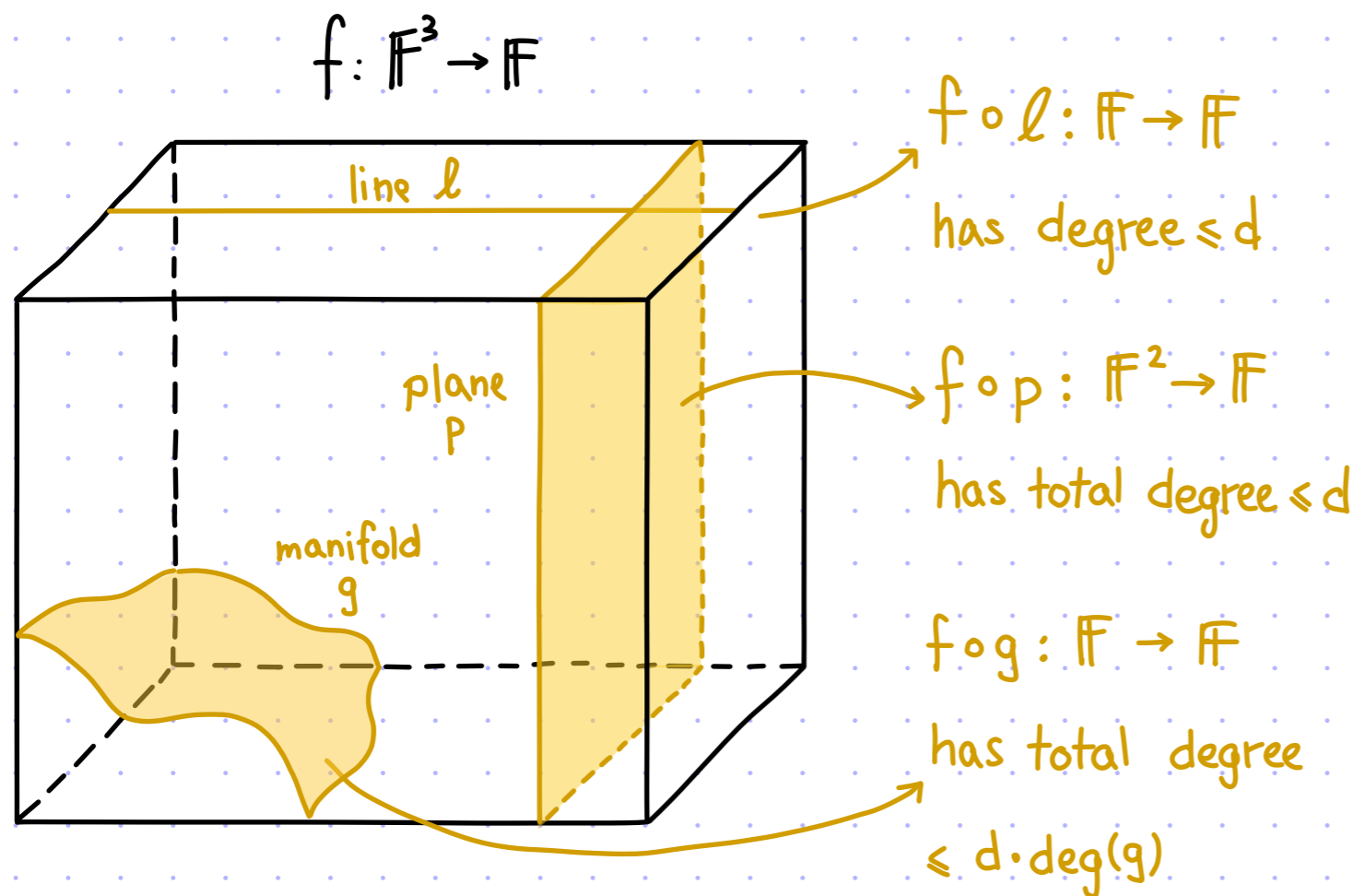
The other regime is  $d \geq |\mathbb{F}|$  (e.g. RM code over  $\mathbb{F}_2$ ).

- In general (any  $d$ ):
- message length  $\leq \min\{|\mathbb{F}|^n, \binom{n+d}{d}\}$ . (Can give better bounds depending on  $|\mathbb{F}|, n, d$ .)
  - relative distance  $\geq \frac{|\mathbb{F}| - d_Q}{|\mathbb{F}|^{d_R+1}}$  where  $d_Q \cdot (|\mathbb{F}|-1) + d_R = d$  for  $d_R < |\mathbb{F}|-1$ .  
implies bound for  $d < |\mathbb{F}|$

# The Rich Structure of the RM Code

The RM code  $LD[\mathbb{F}, n, \text{tot} \leq d]$  has a **rich structure**:

any low-degree (e.g. linear) restriction of the domain  $\mathbb{F}^n$  yields a good subcode ;  
there are many such restrictions, and typically intersect with one another.



A **line** is  $l: \mathbb{F} \rightarrow \mathbb{F}^3$  of total degree 1.  
Hence if  $l(z) = (a_0 + za_1, b_0 + zb_1, c_0 + zc_1)$   
then  $(f \circ l)(z) = f(a_0 + za_1, b_0 + zb_1, c_0 + zc_1)$ .

A **plane** is  $p: \mathbb{F}^2 \rightarrow \mathbb{F}^3$  of total degree 1.  
Hence if  $p(y, z) = (a_0 + a_1y + a_2z, b_0 + b_1y + b_2z, c_0 + c_1y + c_2z)$   
then  $(f \circ p)(y, z) = f(a_0 + a_1y + a_2z, b_0 + b_1y + b_2z, c_0 + c_1y + c_2z)$ .

A  $k$ -variate **manifold** in  $\mathbb{F}^3$  is  $g: \mathbb{F}^k \rightarrow \mathbb{F}^3$ .  
Hence if  $g(z_1, \dots, z_k) = (g_1(z_1, \dots, z_k), g_2(z_1, \dots, z_k), g_3(z_1, \dots, z_k))$   
then  $(f \circ g)(z_1, \dots, z_k) = f(g_1(z_1, \dots, z_k), g_2(z_1, \dots, z_k), g_3(z_1, \dots, z_k))$ .  
[A **curve** is a 1-variate manifold.]

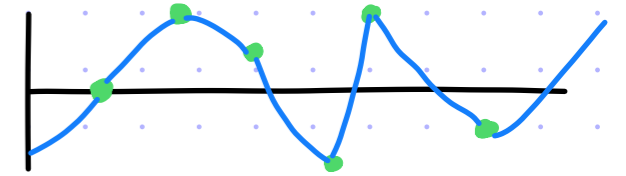
We will see how this rich structure enables proximity testing.

Next: ① low-degree testing for  $LD[\mathbb{F}, n=1, d]$  (univariate polynomials)

② extend to  $LD[\mathbb{F}, n > 1, \text{tot} \leq d]$  (multivariate polynomials)

# Univariate Polynomials: a Trivial Test

Fact: any  $d+1$  locations  $a_0, a_1, \dots, a_d \in \mathbb{F}$  determine a polynomial



A natural idea is to interpolate and test at a random point:

$V^{f: \mathbb{F} \rightarrow \mathbb{F}}(\mathbb{F}, d) :=$

1. Sample  $r \leftarrow \mathbb{F}$
2. Query  $f$  at  $a_0, a_1, \dots, a_d, r$
3. Let  $\tilde{p}(x)$  be the interpolation of  $\{(a_i, f(a_i))\}_{i=0}^d$
4. Check that  $\tilde{p}(r) = f(r)$

query complexity:  
 $d+2 = O(d)$   
[& non-adaptive]

Completeness: if  $f \equiv p$  for a polynomial  $p(x)$  of degree  $\leq d$   
then  $\tilde{p} = p$  and so  $\forall r \in \mathbb{F} \quad \tilde{p}(r) = p(r) = f(r)$

Soundness:  $\Pr_r[V^f = 1] = \Pr_r[\tilde{p}(r) = f(r)] \leq 1 - \Delta(f, LD[\mathbb{F}, n=d])$

The query complexity  $O(d)$  can be much less than  $|\mathbb{F}|$  (reading all of  $f$ ).

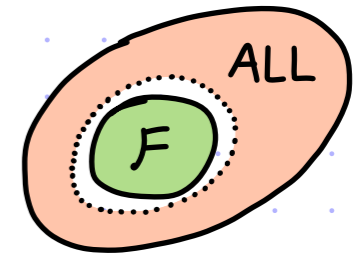
Exercise: prove that a query complexity of  $\Omega(d)$  is necessary.

PROBLEM: the straightforward extension of this idea to  $n > 1$  yields large query complexity.

# Trivial Test for Multivariate Polynomials

The "interpolate-and-test" idea is an example of a **TRIVIAL TEST**.

We describe a trivial test for any property  $F = \{f: D \rightarrow \Sigma\}$ .



A **fixing set**  $S \subseteq D$  for  $F$  is such that,  $\forall a: S \rightarrow \Sigma$ ,  $|\{f \in F \mid f(S) = a\}| \leq 1$ .

- $V_S^{f: D \rightarrow \Sigma}$ :
1. Sample  $r \leftarrow D$ .
  2. Query  $f$  at  $S$  and  $r$ . ← query complexity is  $|S|+1$
  3. Let  $\tilde{p}$  be the unique function in  $F$  s.t.  $\tilde{p}(S) = f(S)$ .
  4. Check that  $\tilde{p}(r) = f(r)$ .

Completeness: if  $f \in F$  then  
 $\Pr[V_S^f = 1] = \Pr[\tilde{p}(r) = f(r)] = 1$ .

Soundness:  
 $\Pr[V_S^f = 1] = \Pr[\tilde{p}(r) = f(r)] \leq 1 - \Delta(f, F)$ .

Examples of trivial tests:

	LECTURE 7			prior slide		
$F$	ALL-ZERO	CONST	LIN( $\mathbb{F}, n$ )	LD[ $\mathbb{F}, n=1, d$ ]	LD[ $\mathbb{F}, n, \text{tot} \leq d$ ]	LD[ $\mathbb{F}, n, \text{ind} \leq d$ ]
$ S $	0	1	$n$	$d+1$	$\binom{n+d}{d}$	$(d+1)^n$

The trivial tests for LD[ $\mathbb{F}, n, \text{tot} \leq d$ ] and LD[ $\mathbb{F}, n, \text{ind} \leq d$ ] are NOT optimal.  
 EXERCISE: where does the proof that the trivial test is optimal for LD[ $\mathbb{F}, n=1, d$ ] break down for  $n > 1$ ?

The BLR test for LIN( $\mathbb{F}, n$ ) has 3 queries, much better than the  $(n+1)$ -query trivial test.

The  $(d+2)$ -query trivial test for LD[ $\mathbb{F}, n=1, d$ ] is optimal.

The trivial tests for LD[ $\mathbb{F}, n, \text{tot} \leq d$ ] and LD[ $\mathbb{F}, n, \text{ind} \leq d$ ] have **large query complexity**.

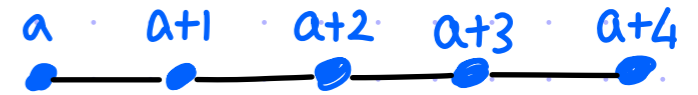
Today we see a total low-degree test that is much better than the trivial test.

# Univariate Polynomials: a Different Attempt

We focus on a special case:  $\mathbb{F} = \mathbb{F}_p$  for prime  $p \geq d+2$ .

The test is inspired by a different local characterization of low-degree polynomials:

def: For  $i=0,1,\dots,d+1$ ,  $C_i := (-1)^{i+1} \binom{d+1}{i} \in \mathbb{F}_p$ .



lemma:  $\forall f: \mathbb{F}_p \rightarrow \mathbb{F}_p \quad \deg(f) \leq d \iff \forall a \in \mathbb{F}_p \quad \sum_{i=0}^{d+1} C_i \cdot f(a+i) = 0$

The proof is by induction, using formal derivatives.

Ex for  $d=0$ :  $(C_0, C_1) = (-1, 1) \rightarrow -f(a) + f(a+1) = 0$ .

Ex for  $d=1$ :  $(C_0, C_1, C_2) = (-1, 2, -1) \rightarrow -f(a) + 2f(a+1) - f(a+2) = 0$ , i.e.,  $\frac{f(a+1)-f(a)}{(a+1)-a} - \frac{f(a+2)-f(a+1)}{(a+2)-(a+1)} = 0$ .

The derivative of  $f(x)$  is

$$f'(x) := \frac{f(x+1)-f(x)}{(x+1)-x} = f(x+1)-f(x).$$

If  $\deg(f) > 0$  then  $\deg(f') = \deg(f) - 1$ .

New proposal:  $\forall f: \mathbb{F}_p \rightarrow \mathbb{F}_p$  ( $\mathbb{F}_p, d$ ):

1. Sample  $r \leftarrow \mathbb{F}_p$
2. Query  $f$  at  $r, r+1, \dots, r+(d+1)$
3. Check that  $\sum_{i=0}^{d+1} C_i \cdot f(r+i) = 0$

**PROBLEM:** it does not work. [Not all local characterizations do!]

Consider  $f = \begin{bmatrix} p_0 & p_1 \end{bmatrix}$ , which has distance  $\geq \frac{1}{2} - \frac{d}{|\mathbb{F}|}$  to  $LD[\mathbb{F}, n=1, d]$ .

This test rejects with probability only  $\Theta(d/|\mathbb{F}|)$ .

# A Refined Local Characterization

def: For  $i=0,1,\dots,d+1$ ,  $c_i := (-1)^{i+1} \binom{d+1}{i} \in \mathbb{F}_p$ .

lemma:  $\forall f: \mathbb{F}_p \rightarrow \mathbb{F}_p \quad \deg(f) \leq d \iff \forall a \in \mathbb{F}_p \quad \sum_{i=0}^{d+1} c_i \cdot f(a+i) = 0$

corollary:  $\forall f: \mathbb{F}_p \rightarrow \mathbb{F}_p \quad \deg(f) \leq d \iff \forall a, b \in \mathbb{F}_p \quad \sum_{i=0}^{d+1} c_i \cdot f(a+i \cdot b) = 0$

proof:

For the direction " $\leftarrow$ " set  $b=1$  and invoke the lemma.

For the direction " $\rightarrow$ ", fix  $a, b \in \mathbb{F}_p$  and consider  $g(x) := f(a+xb)$ .

The degree of  $g$  is at most  $d$ . Hence, by the lemma,

$$\forall e \in \mathbb{F}_p \quad 0 = \sum_{i=0}^{d+1} c_i \cdot g(e+i) = \sum_{i=0}^{d+1} c_i \cdot f(a+(e+i) \cdot b) = \sum_{i=0}^{d+1} c_i \cdot f((a+eb)+i \cdot b).$$

Set  $e:=0$ , and we get the condition for  $a, b$ . ■

The local constraints increased from  $|\mathbb{F}_p|=p$  to  $|\mathbb{F}_p|^2=p^2$ .

The choice of  $b$  randomizes the "step size" and seems to rule out the counterexample.

# Univariate Polynomials: the Rubinfeld–Sudan Test

Check one of the  $|\mathbb{F}_p|^2 = p^2$  local constraints at random.

$V_{RS}^{f: \mathbb{F}_p \rightarrow \mathbb{F}_p}(\mathbb{F}_p, d) :=$

1. Sample  $r, s \leftarrow \mathbb{F}_p$
2. Query  $f$  at  $r, r+s, \dots, r+(d+1) \cdot s$
3. Check that  $\sum_{i=0}^{d+1} C_i \cdot f(r+i \cdot s) = 0$

query complexity:  
 $d+2 = O(d)$   
(& non-adaptive)

Completeness: if  $\deg(f) \leq d$  then  $\Pr[V_{RS}^f = 1] = 1$  by corollary

Soundness: theorem:  $\Pr[V_{RS}^f = 0] \geq \min \left\{ \Omega\left(\frac{1}{d^2}\right), \frac{1}{2} \cdot \Delta(f, LD[\mathbb{F}, n=1, d]) \right\}$

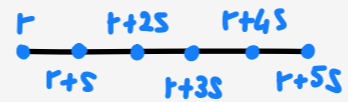
Equivalently:  $\Pr[V_{RS}^f = 1] \leq \max \left\{ 1 - O\left(\frac{1}{d^2}\right), 1 - \frac{1}{2} \cdot \Delta(f, LD[\mathbb{F}, n=1, d]) \right\}$

Isn't this test worse?

- lose a factor of 2 in distance (previously,  $\Pr[V^f = 0] \geq \Delta(f, LD[\mathbb{F}, n=1, d])$ )
- high-agreement regime: even if  $f$  is  $\frac{1}{10}$ -far, we get error only  $\leq 1 - O\left(\frac{1}{d^2}\right)$ , so we need to repeat the test  $O(d^2)$  times for constant error  $\rightarrow O(d^3)$  queries

BUT: RS test extends to multivariate polynomials with no changes

# Proof overview

$$V_{RS}^{f: \mathbb{F}_p \rightarrow \mathbb{F}_p} := \begin{array}{l} 1. \text{ Sample } r, s \in \mathbb{F}_p \\ 2. \text{ Check that } \sum_{i=0}^{d+1} c_i \cdot f(r+i \cdot s) = 0 \end{array}$$


Similar to the case of linearity testing.

theorem:  $\Pr[V_{RS}^f = 0] \geq \min \left\{ \frac{1}{4 \cdot (d+2)^2}, \frac{1}{2} \cdot \Delta(f, \mathbb{F}_p^{\leq d}[X]) \right\}.$

The plurality correction is

$$g_f: \mathbb{F} \rightarrow \mathbb{F} \quad \text{where} \quad g_f(x) := \arg \max_{v \in \mathbb{F}_p} \left| \left\{ s \in \mathbb{F}_p \mid v = \sum_{i=1}^{d+1} c_i \cdot f(x+i \cdot s) \right\} \right|.$$

- Part 1:  $\Pr[V_{RS}^f = 0] \geq \frac{1}{2} \cdot \Delta(f, g_f)$  far from plurality correction  $\rightarrow$  many bad lines
- Part 2:  $\Pr[V_{RS}^f = 0] < \frac{1}{4 \cdot (d+2)^2} \rightarrow \deg(g_f) \leq d$  few bad lines  $\rightarrow$  plurality correction is low-degree

Conclusion:

- If  $\Pr[V_{RS}^f = 0] \geq \frac{1}{4 \cdot (d+2)^2}$  then we are done.
- If  $\Pr[V_{RS}^f = 0] < \frac{1}{4 \cdot (d+2)^2}$  then (by Part 2)  $g_f$  is low-degree and (by Part 1) we get  
$$\Pr[V_{RS}^f = 0] \geq \frac{1}{2} \cdot \Delta(f, g_f) \geq \frac{1}{2} \cdot \Delta(f, \text{LD}[\mathbb{F}, n=1, d]).$$

# Analysis of the RS Test - Part 1

$$\sum_{i=0}^{d+1} c_i \cdot f(r+is) = 0 \Leftrightarrow f(r) = \sum_{i=1}^{d+1} c_i \cdot f(r+is)$$

The plurality correction of  $f$  is  $g_f(x) := \arg \max_{v \in \mathbb{F}_p} \left| \left\{ s \in \mathbb{F}_p \mid v = \sum_{i=1}^{d+1} c_i \cdot f(x+is) \right\} \right|$ .

If  $g_f$  is far from  $f$  then  $V_{RS}^f$  rejects with high probability:

claim:  $\Pr[V_{RS}^f = 0] \geq \frac{1}{2} \cdot \Delta(f, g_f)$

proof: Define  $S := \left\{ r \in \mathbb{F}_p \mid \Pr_s \left[ f(r) \neq \sum_{i=1}^{d+1} c_i \cdot f(r+is) \right] \geq \frac{1}{2} \right\}$ .

For every  $r \notin S$ ,  $\Pr_s \left[ f(r) = \sum_{i=1}^{d+1} c_i \cdot f(r+is) \right] > \frac{1}{2}$  (more than half of  $s$ 's vote for  $f(r)$ )  
so  $f(r) = g_f(r)$ .

Hence  $\Delta(f, g_f) \leq \frac{|S|}{|\mathbb{F}|}$  ( $\forall r$  if  $f(r) \neq g_f(r)$  then  $r \in S$ ).

$$\begin{aligned} \text{So } \Pr[V_{RS}^f = 0] &= \Pr_r[r \in S] \cdot \Pr_{r,s} [V_{RS}^f = 0 \mid r \in S] + \Pr_r[r \notin S] \cdot \Pr_{r,s} [V_{RS}^f = 0 \mid r \notin S] \\ &\geq \frac{|S|}{|\mathbb{F}|} \cdot \min_{r \in S} \left\{ \Pr_s \left[ f(r) \neq \sum_{i=1}^{d+1} c_i \cdot f(r+is) \right] \right\} + 0 \\ &\geq \frac{|S|}{|\mathbb{F}|} \cdot \frac{1}{2} \geq \Delta(f, g_f) \cdot \frac{1}{2}. \quad \blacksquare \end{aligned}$$

# Analysis of the RS Test - Collision Lemma

few bad lines imply many votes for the plurality correction

claim:  $\forall r \in \mathbb{F}_p, \Pr_S \left[ g_f(r) = \sum_{i=1}^{d+1} C_i \cdot f(r+i \cdot s) \right] \geq 1 - 2 \cdot (d+1) \cdot \Pr[V_{RS}^f = 0]$

proof:  $\Pr_S \left[ g_f(r) = \sum_{i=1}^{d+1} C_i \cdot f(r+i \cdot s) \right] = \max_{v \in \mathbb{F}_p} \Pr_S \left[ v = \sum_{i=1}^{d+1} C_i \cdot f(r+i \cdot s) \right]$

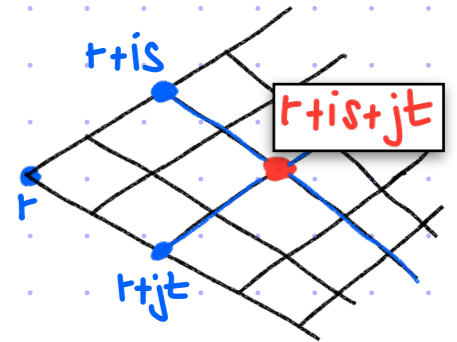
$$\sum_i p_i^2 \leq \max_i \{p_i\} \cdot \sum_i p_i \rightarrow \geq \sum_{v \in \mathbb{F}_p} \Pr_S \left[ v = \sum_{i=1}^{d+1} C_i \cdot f(r+i \cdot s) \right]^2$$

$$= \Pr_{s,t} \left[ \sum_{i=1}^{d+1} C_i \cdot f(r+i \cdot s) = \sum_{i=1}^{d+1} C_i \cdot f(r+i \cdot t) \right]$$

$$\geq 1 - 2 \cdot (d+1) \cdot \Pr[V_{RS}^f = 0].$$

We now analyze the COLLISION PROBABILITY.

For every  $s, t \in \mathbb{F}$  if  $\left\{ \begin{array}{l} \forall i \in \{1, \dots, d+1\} \quad f(r+is) = \sum_{j=1}^{d+1} C_j \cdot f((r+is)+jt) \\ \forall j \in \{1, \dots, d+1\} \quad f(r+jt) = \sum_{i=1}^{d+1} C_i \cdot f((r+jt)+is) \end{array} \right\}$



then  $\sum_{i=1}^{d+1} C_i \cdot f(r+is) = \sum_{i=1}^{d+1} C_i \cdot \sum_{j=1}^{d+1} C_j \cdot f((r+is)+jt) = \sum_{j=1}^{d+1} C_j \cdot \sum_{i=1}^{d+1} C_i \cdot f((r+jt)+is) = \sum_{j=1}^{d+1} C_j \cdot f(r+jt)$ .

Hence

$$\Pr_{s,t} \left[ \sum_{i=1}^{d+1} C_i \cdot f(r+is) \neq \sum_{i=1}^{d+1} C_i \cdot f(r+it) \right] \leq \Pr_{s,t} \left[ \begin{array}{l} \exists i \in \{1, \dots, d+1\} \quad f(r+is) \neq \sum_{j=1}^{d+1} C_j \cdot f((r+is)+jt) \\ \text{or} \\ \exists j \in \{1, \dots, d+1\} \quad f(r+jt) \neq \sum_{i=1}^{d+1} C_i \cdot f((r+jt)+is) \end{array} \right]$$

$$\leq 2(d+1) \cdot \Pr[V_{RS}^f = 0].$$



# Analysis of the RS Test - Part 2

claim: if  $\Pr[V_{RS}^f = 0] < \frac{1}{4 \cdot (d+2)^2}$  then  $\deg(g_f) \leq d$

proof: Fix  $r, s \in \mathbb{F}_p$ . We show that  $\sum_{i=0}^{d+1} c_i \cdot g_f(r+is) = 0$ .

If  $\exists t_1, t_2 \in \mathbb{F}_p$  s.t.  $\begin{cases} \forall i \in \{0, 1, \dots, d+1\} & g_f(r+is) = \sum_{j=1}^{d+1} c_j \cdot f((r+is)+j(t_1+it_2)) \\ \forall j \in \{1, \dots, d+1\} & \sum_{i=0}^{d+1} c_i \cdot f((r+jt_1)+i(s+jt_2)) = 0 \end{cases}$

then

$$\sum_{i=0}^{d+1} c_i \cdot g_f(r+is) = \sum_{i=0}^{d+1} c_i \cdot \left[ \sum_{j=1}^{d+1} c_j \cdot f((r+is)+j(t_1+it_2)) \right] \stackrel{\text{reorder summations}}{=} \sum_{j=1}^{d+1} c_j \cdot \left[ \sum_{i=0}^{d+1} c_i \cdot f((r+jt_1)+j(t_1+it_2)) \right] = \sum_{j=1}^{d+1} c_j \cdot 0 = 0.$$

Hence

$$\Pr_{t_1, t_2} \left[ \sum_{i=0}^{d+1} c_i \cdot g_f(r+is) \neq 0 \right] \stackrel{\text{union bound}}{\leq} \Pr_{t_1, t_2} \left[ \begin{array}{l} \exists i \in \{0, 1, \dots, d+1\} \\ \text{OR} \\ \exists j \in \{1, \dots, d+1\} \end{array} \left[ \begin{array}{l} g_f(r+is) \neq \sum_{j=1}^{d+1} c_j \cdot f((r+is)+j(t_1+it_2)) \\ \sum_{i=0}^{d+1} c_i \cdot f((r+jt_1)+i(s+jt_2)) \neq 0 \end{array} \right] \right]$$

$$\leq (d+2) \cdot \frac{1}{2 \cdot (d+2)} + (d+1) \cdot \frac{1}{4 \cdot (d+2)^2} < 1.$$

$$\Pr_{t_1, t_2} \left[ g_f(r+is) \neq \sum_{j=1}^{d+1} c_j \cdot f((r+is)+j(t_1+it_2)) \right] \stackrel{\text{collision lemma}}{<} 2(d+1) \cdot \Pr[V_{RS}^f = 0] < 2(d+1) \cdot \frac{1}{4 \cdot (d+2)^2} \leq \frac{1}{2 \cdot (d+2)}$$

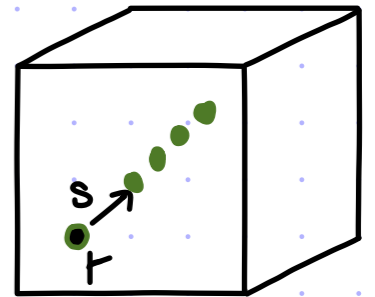
# Extending the RS Test to Multivariate Polynomials

The local characterization holds similarly:

$$\forall d < p-2 \quad \forall f: \mathbb{F}_p^n \rightarrow \mathbb{F}_p, \quad \deg_{\text{tot}}(f) \leq d \iff \forall a, b \in \mathbb{F}_p^n \quad \sum_{i=0}^{d+1} c_i \cdot f(a+i \cdot b) = 0$$

This directly motivates the following test: query complexity is  $d+2 = O(d)$

- $V_{RS}^{f: \mathbb{F}_p^n \rightarrow \mathbb{F}_p}(\mathbb{F}_p, n, d) :=$
1. Sample  $r, s \leftarrow \mathbb{F}_p^n$
  2. Query  $f$  at  $r, r+s, \dots, r+(d+1) \cdot s$
  3. Check that  $\sum_{i=0}^{d+1} c_i \cdot f(r+i \cdot s) = 0$



read  $d+2$  locations  
on a random line

The theorem for soundness is also similar:

theorem:  $\Pr[V_{RS}^f = 0] \geq \min \left\{ \frac{1}{4 \cdot (d+2)^2}, \frac{1}{2} \cdot \Delta(f, \text{LD}[\mathbb{F}_p, n, \text{tot} \leq d]) \right\}$

The proof is the same up to syntactic modifications.

By repeating the test  $O(d^2)$  times, we get:

a total low-degree test with query complexity  $O(d^3)$  [independent of  $n$ ]

where "constant relative distance"  $\rightarrow$  "constant soundness error".

# More on Total Low-Degree Testing

[1/2]

A key structure that enables low-degree testing is a **ROBUST LINES CHARACTERIZATION**.

Suppose that  $f: \mathbb{F}^n \rightarrow \mathbb{F}$  has total degree  $\leq d$ .

Then,  $\forall a, b \in \mathbb{F}^n$ , the univariate function  $g_{a,b}(z) := f(a + z \cdot b)$  has degree  $\leq d$ .

Does the converse hold?

**COUNTEREXAMPLE:** Set  $\mathbb{F} := \mathbb{F}_{p^e}$  and fix any  $d$  with  $p^e - p^{e-1} - 1 < d < p^e$ .

Consider the bivariate function  $f(x_1, x_2) := (x_1^{p-1} x_2)^{p^{e-1}}$ .

The total degree of  $f$  is  $p^e > d$ . Yet  $\forall a, b$   $g_{a,b}(z) = f(a_1 + zb_1, a_2 + zb_2) = ((a_1 + zb_1)^{p-1} (a_2 + zb_2))^{p^{e-1}}$

has degree at most  $(p-1)p^{e-1} = p^e - p^{e-1}$ . (Indeed recall that  $z^{p^e} \equiv z$  since  $p^e$  is the field size.)

The converse holds if  $d \leq p^e - p^{e-1} - 1$ . (E.g. if  $\mathbb{F}$  has prime size  $p$  then the condition is  $d \leq p-2$ .)

[Friedl, Sudan 1995]  
for a proof.

In this case, if  $\{g_{a,b}(z) := f(a + zb)\}_{a,b \in \mathbb{F}^n}$  all have degree  $\leq d$  then  $f$  has total degree  $\leq d$ .

Low-degree testing is based on distance variants of such results:

if  $\{g_{a,b}(z) := f(a + zb)\}_{a,b \in \mathbb{F}^n}$  are close to degree  $\leq d$  in expectation

then  $f$  is close to total degree  $\leq d$

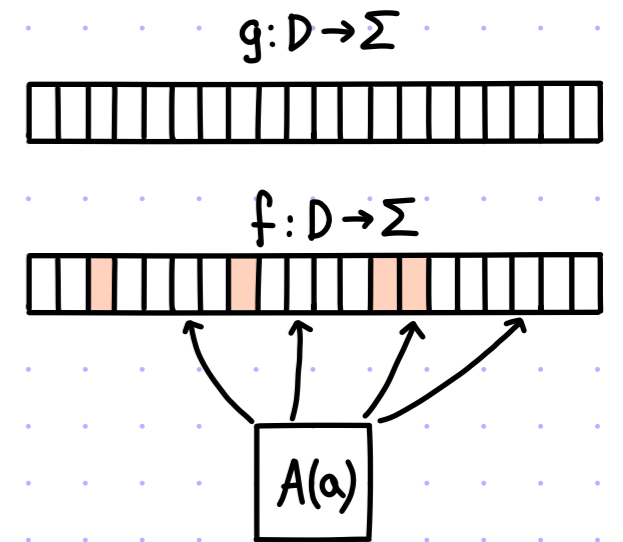
# Local Correction of Low-Degree Functions

[1/4]

Let  $F = \{g: D \rightarrow \Sigma\}$  be a set of functions.

A **local corrector** for  $F$  with **decoding error**  $\epsilon$  is an algorithm  $A$  s.t.

- ①  $\forall g \in F \forall a \in D \Pr[A^g(a) = g(a)] = 1.$
- ②  $\forall g \in F \forall \delta \in [0, 1] \forall f$  with  $\Delta(f, g) \leq \delta \forall a \in D \Pr[A^f(a) \neq g(a)] \leq \epsilon(\delta).$



The local corrector **tolerates** a  $\delta^*$ -fraction of errors if  $\epsilon(\delta) \in [0, \frac{1}{2}) \forall \delta \in [0, \delta^*).$

An error  $< \frac{1}{2}$  enables error reduction via repetition and returning the plurality value.

Let  $\Delta(F) := \min_{f, g \in F} \Delta(f, g)$  be the minimum relative distance of  $F$ .

**claim:**  $\delta^* \leq \frac{1}{2} \Delta(F)$   
 unique decoding radius

**proof:** Fix any  $\delta$  s.t.  $\epsilon(\delta) < \frac{1}{2}$ . Fix arbitrary  $g \in F$ . There cannot be  $g' \in F$  with  $g' \neq g$  and  $\Delta(g', g) \leq \delta$  because, picking  $a \in D$  s.t.  $g'(a) \neq g(a)$ , cannot have  $\Pr[A^{g'}(a) = g(a)] > \frac{1}{2}$  and  $\Pr[A^g(a) = g'(a)] > \frac{1}{2}$ . Hence  $\Delta(F) > 2\delta$ .

We saw a 2-query local corrector for  $\text{LIN}[\mathbb{F}, n]$  with  $\epsilon(\delta) = 2\delta$ , for which  $\delta^* = \frac{1}{4}$ .

We discuss several local correctors for  $\text{LD}[\mathbb{F}, n, \text{tot} \leq d]$ :

- $A_1$  makes  $d+1$  queries and has error  $\epsilon(\delta) = (d+1) \cdot \delta$ , so  $\delta^* = \frac{1}{2(d+1)}$ .
- $A_2$  makes  $q-1$  queries and has error  $\epsilon(\delta) = \frac{2}{1 - \frac{d}{q-1}} \cdot \delta$ , so  $\delta^* = \frac{1}{4} \cdot (1 - \frac{d}{q-1})$ .
- $A_3$  makes  $q-1$  queries and has error  $\epsilon(\delta) = O_{d, \delta}(\frac{1}{q})$  for  $\delta < \frac{1}{2} \cdot (1 - \frac{2d}{q-1})$ , so  $\delta^* = \frac{1}{2} \cdot (1 - \frac{2d}{q-1})$ .

We let  $q := |F|$  here and in next few slides.

# Local Correction of Low-Degree Functions

[2/4]

Suppose that  $f: \mathbb{F}^n \rightarrow \mathbb{F}$  is  $\delta$ -close to  $g \in \text{LD}[\mathbb{F}, n, \text{tot} \leq d]$ .

We wish to correct the value at the point  $a \in \mathbb{F}^n$ .

IDEA: focus on the values of  $f$  on the line  $\ell_{a,b} := \{a + \lambda b\}_{b \in \mathbb{F}}$  for a random  $b \in \mathbb{F}^n$ .

Recall that a polynomial of degree  $d$  is determined by its evaluations at any  $d+1$  points.

Let  $\{\lambda_1, \dots, \lambda_{d+1}\} \subseteq \mathbb{F} \setminus \{0\}$  be a set of size  $d+1$ .

We attempt to recover  $f(a)$  from  $f(a + \lambda_1 b), \dots, f(a + \lambda_{d+1} b)$ .

- $A_1^f(a) :=$
1. Sample  $b \leftarrow \mathbb{F}^n$ .
  2. Query  $f$  at  $\{a + \lambda_i b\}_{i=1}^{d+1}$ .
  3. Let  $p(x)$  be the interpolation of  $\{(a + \lambda_i b, f(a + \lambda_i b))\}_{i=1}^{d+1}$ .
  4. Output  $p(0)$ .

claim:  $\forall a \in \mathbb{F}^n \Pr[A_1^f(a) \neq g(a)] \leq (d+1)\delta$ .

proof: For every  $i \in [d+1]$ ,  $\Pr_b[f(a + \lambda_i b) \neq g(a + \lambda_i b)] \leq \delta$ .

Hence  $\Pr[p(0) \neq g(a)] = \Pr[\exists i \in [d+1] f(a + \lambda_i b) \neq g(a + \lambda_i b)] \leq (d+1) \cdot \delta$ . ■

# Local Correction of Low-Degree Functions

[3/4]

We see a local corrector that makes  $q-1$  queries and tolerates more errors. (The special case  $q=d+2$  is the previous local corrector.)

IDEA: the values along a line have high redundancy and it suffices to find a line with few enough errors.

- $A_2^f(a) :=$
1. Sample  $b \leftarrow \mathbb{F}^n$ .
  2. Query  $f$  at  $\{a + \lambda b\}_{\lambda \in \mathbb{F} \setminus \{0\}}$ .
  3. Let  $p(x)$  be the unique polynomial of degree  $d$  s.t.  $|\{\lambda \in \mathbb{F} \setminus \{0\} : p(\lambda) \neq f(a + \lambda b)\}| < \frac{q-1-d}{2}$ .
  4. Abort if no  $p(x)$  is found; else output  $p(0)$ .

← can efficiently find  $p(x)$  via the Berlekamp-Welch algorithm

claim:  $\forall a \in \mathbb{F}^n \Pr[A_2^f(a) \neq g(a)] \leq \frac{2\delta}{1 - \frac{d}{q-1}}$ .

More generally, can modify  $A_2$  to obtain  $p(x)$  by querying  $f$  at  $\{a + \lambda b\}_{\lambda \in S}$  for any  $S \subseteq \mathbb{F} \setminus \{0\}$  with  $|S| \geq d+1$ . In this case  $A_2$  searches for  $p$  of degree  $d$  s.t.  $|\{\lambda \in S : p(\lambda) \neq f(a + \lambda b)\}| < \frac{|S|-d}{2}$  and errs w.p.  $\leq \frac{2\delta}{1 - \frac{d}{|S|}}$ .

proof: For every  $\lambda \in \mathbb{F} \setminus \{0\}$ , let  $Z_\lambda$  be the indicator for the event  $f(a + \lambda b) \neq g(a + \lambda b)$ , and note that  $\mathbb{E}[Z_\lambda] = \Pr[Z_\lambda = 1] \leq \delta$ . Define  $Z := \sum_{\lambda \in \mathbb{F} \setminus \{0\}} Z_\lambda$  and note that  $\mathbb{E}[Z] \leq (q-1)\delta$ . If  $Z < \frac{q-1-d}{2}$  then  $p(x) \equiv g(a + xb)$  so  $p(0) = g(a)$  (correction succeeds).

Hence  $\Pr[p(0) \neq g(a)] \leq \Pr[Z \geq \frac{q-1-d}{2}] \leq \frac{2}{q-1-d} \cdot \mathbb{E}[Z] \leq \frac{2}{q-1-d} \cdot (q-1) \cdot \delta = \frac{2\delta}{1 - \frac{d}{q-1}}$ . ■

↑ Markov's inequality: for a random variable  $Z \geq 0$  and  $c > 0$   $\Pr[Z \geq c] \leq \frac{1}{c} \mathbb{E}[Z]$ .

# Local Correction of Low-Degree Functions

[4/4]

An additional improvement tolerates even more errors, close to the unique decoding radius  $\frac{1}{2} \cdot (1 - \frac{d}{q})$ .

The idea is to correct on a random quadratic curve, whose locations are pairwise independent.

- $A_3^f(a) :=$
1. Sample  $b, c \leftarrow \mathbb{F}^n$ .
  2. Query  $f$  at  $\{a + \lambda b + \lambda^2 c\}_{\lambda \in \mathbb{F} \setminus \{0\}}$ .
  3. Let  $p(x)$  be the unique polynomial of degree  $2d$  s.t.  $|\{\lambda \in \mathbb{F} \setminus \{0\} : p(\lambda) \neq f(a + \lambda b + \lambda^2 c)\}| < \frac{q-1-2d}{2}$ .
  4. Abort if no  $p(x)$  is found; else output  $p(0)$ .

← can efficiently find  $p(x)$  via the Berlekamp-Welch algorithm

claim:  $\forall a \in \mathbb{F}^n \Pr[A_3^f(a) \neq g(a)] \leq \frac{1}{q-1} \cdot \frac{4 \cdot (\delta - \delta^2)}{(1 - \frac{2d}{q-1} - \delta)^2}$

Can similarly modify  $A_3$  to obtain  $p(x)$  by querying  $f$  at  $\{a + \lambda b + \lambda^2 c\}_{\lambda \in S}$  for any  $S \subseteq \mathbb{F} \setminus \{0\}$  with  $|S| \geq d+1$ .

proof: For every  $\lambda \in \mathbb{F} \setminus \{0\}$ , let  $Z_\lambda$  be the indicator for the event  $f(a + \lambda b + \lambda^2 c) \neq g(a + \lambda b + \lambda^2 c)$

and note that  $\mathbb{E}[Z_\lambda] = \Pr[Z_\lambda = 1] \leq \delta$  and  $\text{Var}(Z_\lambda) = \mathbb{E}[Z_\lambda] - \mathbb{E}[Z_\lambda]^2 \leq \delta - \delta^2$  ( $x - x^2$  is increasing on  $[0, \frac{1}{2}]$ ).

Define  $Z := \sum_{\lambda \in \mathbb{F} \setminus \{0\}} Z_\lambda$  and note that  $\mathbb{E}[Z] \leq (q-1)\delta$  and  $\text{Var}(Z) = \sum_{\lambda \in \mathbb{F} \setminus \{0\}} \text{Var}(Z_\lambda) \leq (q-1) \cdot (\delta - \delta^2)$ .  
↑ pairwise independence

If  $Z < \frac{q-1-2d}{2}$  then  $p(x) \equiv g(a + xb + x^2c)$  so  $p(0) = g(a)$  (correction succeeds).

Hence  $\Pr[p(0) \neq g(a)] \leq \Pr[Z \geq \frac{q-1-2d}{2}] \leq \Pr[|Z - \mathbb{E}[Z]| \geq \frac{q-1-2d}{2} - \mathbb{E}[Z]]$

$$\leq \Pr[|Z - \mathbb{E}[Z]| \geq \frac{q-1-2d}{2} - (q-1)\delta] \leq \frac{\text{Var}(Z)}{(\frac{q-1-2d}{2} - (q-1)\delta)^2} \leq \frac{(q-1)(\delta - \delta^2)}{(\frac{q-1-2d}{2} - (q-1)\delta)^2} = \frac{1}{q-1} \cdot \frac{4 \cdot (\delta - \delta^2)}{(1 - \frac{2d}{q-1} - \delta)^2}$$

Chebyshev's inequality: for a random variable  $Z$  and  $c > 0$   $\Pr[|Z - \mathbb{E}[Z]| \geq c] \leq \frac{1}{c^2} \cdot \text{Var}(Z)$ .

# A few words about low-degree testing

## Low-degree testing for quantum states, and a quantum entangled games PCP for QMA

Anand Natarajan\* Thomas Vidick†

### Abstract

We show that given an explicit description of a multiplayer game, with a classical verifier and a constant number of players, it is QMA-hard, under randomized reductions, to distinguish between the cases when the players have a strategy using entanglement that succeeds with probability 1 in the game, or when no such strategy succeeds with probability larger than  $\frac{1}{2}$ . This proves the “games quantum PCP conjecture” of Fitzsimons and the second author (ITCS’15), albeit under randomized reductions.

The core component in our reduction is a construction of a family of two-player games for testing  $n$ -qubit maximally entangled states. For any integer  $n \geq 2$ , we give such a game in which questions from the verifier are  $O(\log n)$  bits long, and answers are  $\text{poly}(\log \log n)$  bits long. We show that for any constant  $\epsilon \geq 0$ , any strategy that succeeds with probability at least  $1 - \epsilon$  in the test must use a state that is within distance  $\delta(\epsilon) = O(\epsilon^c)$  from a state that is locally equivalent to a maximally entangled state on  $n$  qubits, for some universal constant  $c > 0$ . The construction is based on the classical plane-vs-point test for multivariate low-degree polynomials of Raz and Safra (STOC’97). We extend the classical test to the quantum regime by executing independent copies of the test in the generalized Pauli X and Z bases

## Low-degree tests at large distances

Alex Samorodnitsky\*

September 27, 2018

### Abstract

We define tests of boolean functions which distinguish between linear (or quadratic) in some sense, from those polynomially soundness and the

## Testing Low-Degree Polynomials over $GF(2)$

Noga Alon\* Tali Kaufman† Michael Krivelevich‡ Dana Ron§

July 9, 2003

### Abstract

We describe an efficient randomized algorithm to test if a given binary function  $f : \{0, 1\}^k \rightarrow \{0, 1\}$  is a low degree polynomial (that is, a sum of low degree monomials). For a given integer  $k \geq 1$  and a given real  $\epsilon > 0$ , the algorithm queries  $f$  at  $O(\frac{1}{\epsilon} + k^k)$  points. If  $f$  is a polynomial of degree at most  $k$ , the algorithm always accepts, and if the value of  $f$  has to be modified on at least an  $\epsilon$  fraction of all inputs in order to transform it to such a polynomial, then the algorithm rejects with probability at least  $2/3$ . Our result is essentially tight: Any algorithm for testing degree- $k$  polynomials over  $GF(2)$  must perform  $\Omega(\frac{1}{\epsilon} + 2^k)$  queries.

ximity norms behave “randomly” of an inverse theorem for

## Improved low-degree testing and its applications

Sanjeev Arora\* Princeton University

Madhu Sudan† IBM T. J. Watson Research Center

### Abstract

$NP = PCP(\log n, 1)$  and related results crucially depend upon the close connection between the probability with which a function passes a low degree test and the distance of this function to the nearest degree  $d$  polynomial. In this paper we study a test proposed by Rubinfeld and Sudan [29]. The strongest previously known connection for this test states that a function passes the test with probability  $\delta$  for some  $\delta > 7/8$  iff the function has agreement  $\approx \delta$  with a

### 1 Introduction

The use of algebraic techniques has led to probabilistic characterizations of complexity classes. These characterizations involve an untrustworthy prover (or polynomial-time verifier). In  $MIP = NP = PCP(\log n, 1)$  [6, 5] the verifier locally verifies the satisfiability of a boolean formula by checking a very few bits in a “proof string”. In  $IP = PSPACE$  [24, 31] the verifier has

## A Sub-Constant Error-Probability Low-Degree Test, and a Sub-Constant Error-Probability PCP Characterization of NP \*

Ran Raz†

Shmuel Safra‡

### Abstract

We introduce a new low-degree test, one that uses the restriction of low-degree polynomials to planes (i.e., affine sub-spaces of dimension 2), rather than the restriction to lines (i.e., affine sub-spaces of dimension 1). We prove the new test to be of a very small error-probability (in particular, much smaller than constant).

The new test enables us to prove a low-error characterization of NP in terms of PCP. Specifically, our theorem states that, for any given  $\epsilon > 0$ , membership in any NP language can be verified with  $O(1)$  accesses,

of the most fundamental avenues of research in theory of computer-science.

Since the early days, when the classes P and NP were defined, and the question was posed as to whether they are the same or do they differ, many problems were shown to be NP-complete, thereby increasing the weight on finding stricter characterization for the class NP.

NP has since been given a few alternative characterizations. The one most commonly applied being Cook’s [Coo71], which characterizes NP in terms of efficient verification of proofs (or nondeterministic computations).

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