

# Lecture B.3

# Low-Degree Testing

*(Locality of the Reed-Muller code)*

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# Recap

## Linearity Testing

A function  $f: \mathbb{F}^n \rightarrow \mathbb{F}$  is linear if  $\exists c \in \mathbb{F}^n$  s.t.  $f(x) = \sum_{i=1}^n c_i x_i$ .  
Equivalently, if  $\forall x, y \in \mathbb{F}^n$   $f(x) + f(y) = f(x+y)$ .



$$\text{ALL} = \{ f: \mathbb{F}^n \rightarrow \mathbb{F} \}$$
$$\text{LIN} = \{ f: \mathbb{F}^n \rightarrow \mathbb{F} \mid f(x) + f(y) = f(x+y) \}$$

We want a  $O(1)$ -query test that, given  $f$ ,  
**But this is impossible:** if  $f$  differs from linear  
but we can't query all points.

So we relax the question: given  $f$ ,  
say YES if  $f \in \text{LIN}$  and NO if  $f \notin \text{LIN}$ .

We count in Hamming distance:

$$\Delta(f, g) := \Pr_{x \in \mathbb{F}^n} [f(x) \neq g(x)] \text{ and}$$

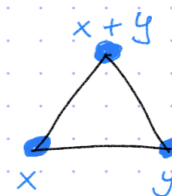
Q1: can we solve the relaxed problem?

## Proof overview

$$V_{\text{BLR}}^f := \Pr_{x, y \in \mathbb{F}^n} [f(x) + f(y) = f(x+y)]$$

Theorem:  $\Pr[V_{\text{BLR}}^f = 0] \geq \min\{1/6, \frac{1}{2} \cdot \Delta(f, \text{LIN})\}$

$g_f: \mathbb{F}^n \rightarrow \mathbb{F}$  is defined as  $g_f(x) := \arg \max_{v \in \mathbb{F}} |\{y \in \mathbb{F}^n \mid v = f(x+y) - f(y)\}|$   
this is the plurality value



Step 1: ~~Bad-triangle~~ is captured by distance from plurality vote

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

Step 2: plurality implies overwhelming majority

$$\Pr_{x, y \in \mathbb{F}^n} [g_f(x) = f(x+y) - f(y)] \geq 1 - 2 \cdot \Pr[V_{\text{BLR}}^f = 0] \geq \frac{2}{3}$$

Step 3: plurality vote ( $g_f$ ) is a linear function

$$\text{if } \Pr[V_{\text{BLR}}^f = 0] < \frac{1}{6} \text{ then } \forall x, y \quad g_f(x) + g_f(y) = g_f(x+y)$$

# Low-Degree Testing

Recall the goal of **linearity testing**:

The goal of **low-degree testing** is:

input:  $\mathbb{F}, n$

oracle:  $f: \mathbb{F}^n \rightarrow \mathbb{F}$

requirement: YES w.p. 1 if  $f \in \text{LIN}(\mathbb{F}, n)$

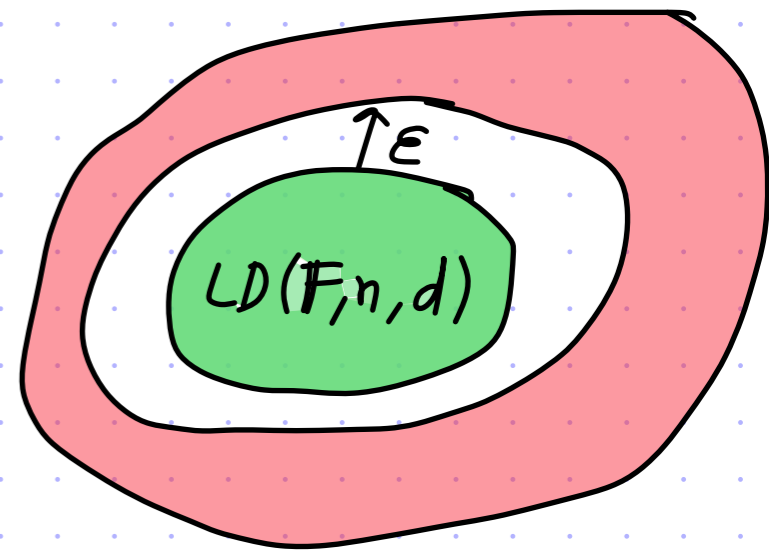
YES w.p.  $\frac{1}{2}$  if  $f$  is  $\frac{1}{10}$ -far from  $\text{LIN}(\mathbb{F}, n)$

input:  $\mathbb{F}, n, d$

oracle:  $f: \mathbb{F}^n \rightarrow \mathbb{F}$

requirement: YES w.p. 1 if  $f \in \text{LD}(\mathbb{F}, n, d)$

YES w.p.  $\frac{1}{2}$  if  $f$  is  $\frac{1}{10}$ -far from  $\text{LD}(\mathbb{F}, n, d)$



What does degree  $d$  mean?

- **total degree** (e.g. in this case  $\text{LD}(\mathbb{F}, n, \text{tot} \leq 1) = \text{LIN}(\mathbb{F}, n)$ )
- **individual degree** (e.g. in this case  $\text{LD}(\mathbb{F}, n, \text{ind} \leq 1)$  is multilinear polys)

A test for individual degree can be derived from a test for total degree.

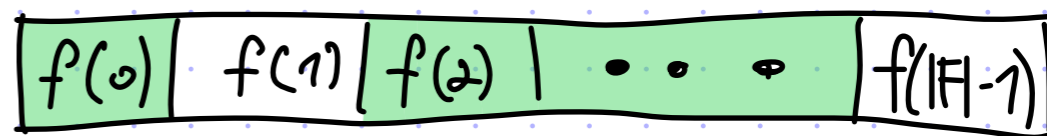
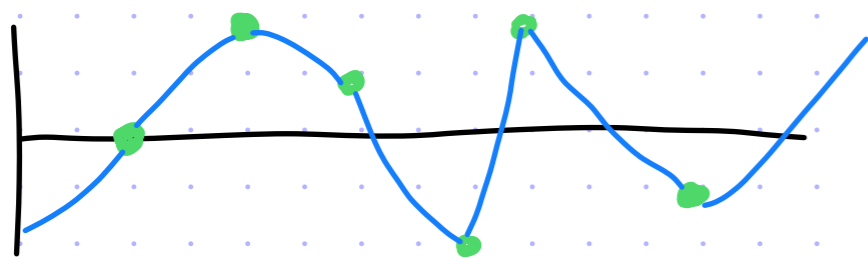
Either way in most applications to PCPs the difference does not matter.

Today we study total degree:

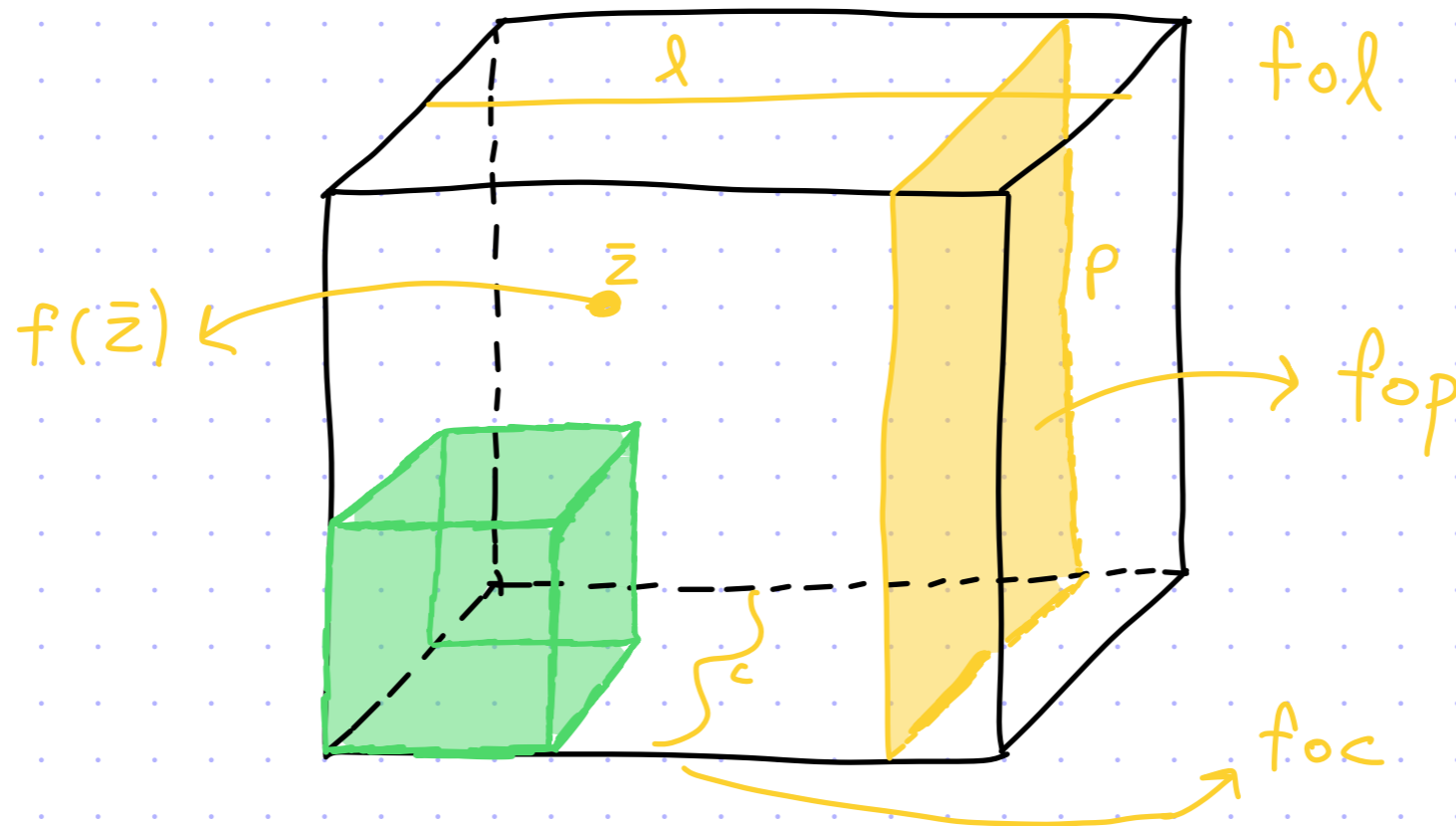
**Step 1:** understand  $n=1$  (univariate polys)    **Step 2:** extend to  $n > 1$  (multivariate polys)

# Interlude: the magic of polynomials

Univariate polynomials of degree  $d$  extend the values of  $d+1$  points to a field  $\mathbb{F}$ .



Consider a polynomial  $f: \mathbb{F}^n \rightarrow \mathbb{F}$  s.t.  $\deg(f) = d$ .

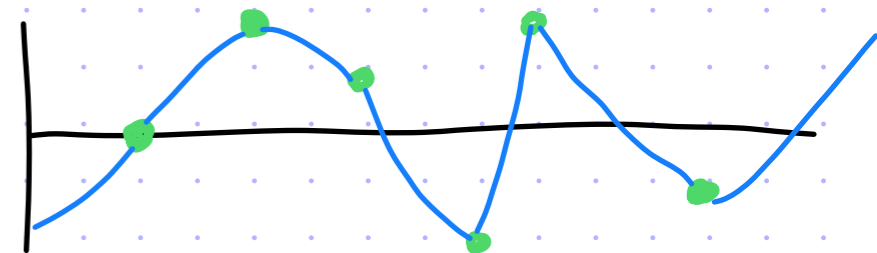


all affine subspaces are codes!

lines  
planes  
hyperplanes  
curves  
manifolds

# Univariate Polynomials: a Basic Test

Idea: any  $d+1$  locations determine a polynomial



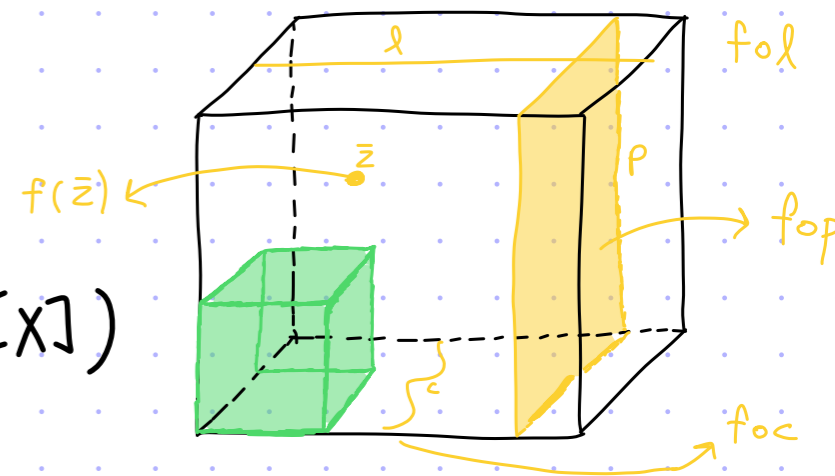
$T_{f: \mathbb{F} \rightarrow \mathbb{F}}(\mathbb{F}, d) :=$

1. sample  $r \in \mathbb{F}$
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:  $\Pr_f[\text{accept}] = \Pr_f[\tilde{p}(r) = f(r)] \leq 1 - \Delta(f, \mathbb{F}^{\leq d}[X])$



The query complexity of  $O(d)$  could be much less than  $|\mathbb{F}|$  (reading all of  $f$ ).  
 Also, one can prove that a query complexity of  $\Omega(d)$  is necessary.

# Univariate Polynomials: a Different Attempt

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

The test is inspired by a different local characterization & 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:  $\forall d < p, \forall f: \mathbb{F}_p \rightarrow \mathbb{F}_p \quad f \in \mathbb{F}^{\leq d}[x]$  iff  $\forall a \in \mathbb{F}_p \quad \sum_{i=0}^{d+1} c_i \cdot f(a+i) = 0$

proof: Induction and 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.,  $\frac{f(a+1)-f(a)}{(a+1)-a} - \frac{f(a+2)-f(a+1)}{(a+2)-(a+1)} = 0$ .

A new proposal:

$T_f: \mathbb{F}_p \rightarrow \mathbb{F}_p$  ( $\mathbb{F}_p, d$ ):

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  $f: \boxed{p_0} \boxed{p_1}$ , which has distance  $1/2$  to  $\mathbb{F}^{\leq d}[x]$ .

But the test rejects only with probability  $\approx d/(|\mathbb{F}|/2)$ .

# Univariate Polynomials: the Rubinfeld-Sudan Test

Check one of the  $|\mathbb{F}_p|^2$  local conditions at random:

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

1. sample  $r, s \in \mathbb{F}_p$
2. query  $f$  at  $r, r+s, \dots, r+(d+1) \cdot s$
3. check that  $\sum_{i=0}^{d+1} \binom{d+1}{i} \cdot f(r+i \cdot s) = 0$

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

Completeness: if  $f \in \mathbb{F}_p^{\leq d}[X]$  then  $\Pr_{r,s} [T^f = 1] = 1$  by corollary

Soundness: if  $f$  is  $\frac{1}{10}$ -far from  $\mathbb{F}_p^{\leq d}[X]$  then  $\Pr [T^f = 1] \leq 1 - O\left(\frac{1}{d^2}\right)$ .

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

Isn't this test worse?

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

But: this test will extend to multivariate polynomials with no changes

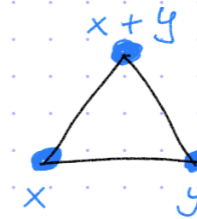
# Recall the linear testing analysis

## Proof overview

$V_{BLR}^f: \mathbb{F}^n \rightarrow \mathbb{F}$   
:= 1. sample  $x, y \in \mathbb{F}^n$   
2. check that  $f(x) + f(y) = f(x+y)$

Theorem:  $\Pr[V_{BLR}^f = 0] \geq \min\left\{\frac{1}{6}, \frac{1}{2} \cdot \Delta(f, \text{LIN})\right\}$

$g_f: \mathbb{F}^n \rightarrow \mathbb{F}$  is defined as  $g_f(x) := \arg \max_{v \in \mathbb{F}} |\{y \in \mathbb{F}^n \mid v = f(x+y) - f(y)\}|$   
this is the plurality value



Step 1: ~~\*~~Bad-triangle is captured by distance from plurality vote

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

Step 2: plurality implies overwhelming majority

$$\Pr_{y \in \mathbb{F}^n} [g_f(x) = f(x+y) - f(y)] \geq 1 - 2 \cdot \Pr[V_{BLR}^f = 0] \geq \frac{2}{3}$$

Step 3: plurality vote ( $g_f$ ) is a linear function

$$\text{if } \Pr[V_{BLR}^f = 0] < \frac{1}{6} \text{ then } \forall x, y \quad g_f(x) + g_f(y) = g_f(x+y)$$



# Analysis of the RS Test - Part 1

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

The analysis is analogous to the combinatorial analysis of the BLR test.

We consider the plurality (most popular) values:



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

If  $g_f$  is far from  $f$  then  $T$  must reject with high probability:

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

proof: Letting  $S = \left\{ r \in \mathbb{F}_p \text{ s.t. } \Pr_{s \in \mathbb{F}_p} \left[ f(r) \neq \sum_{i=1}^{d+1} c_i \cdot f(r+is) \right] \geq \frac{1}{2} \right\}$ , we get

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

Also, for every  $r \notin S$  we have  $\Pr_s \left[ f(r) = \sum_{i=1}^{d+1} c_i f(r+is) \right] > \frac{1}{2}$  so  $f(r) = g_f(r)$ .  
This tells us that  $\frac{|S|}{|\mathbb{F}|} \geq \Delta(g_f, f)$ . ■

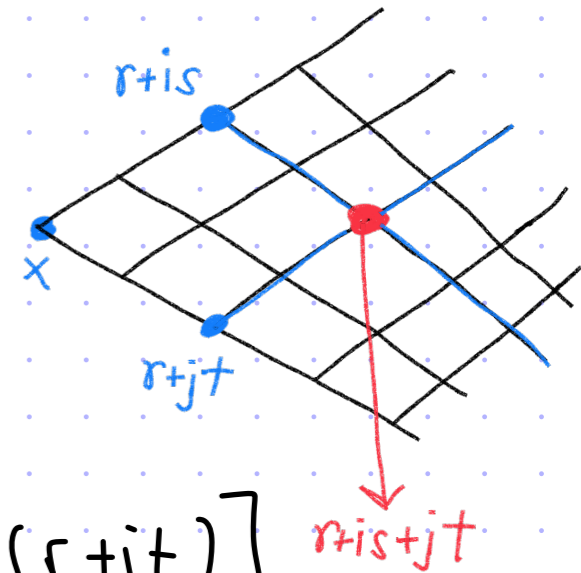
# Analysis of the RS Test - Part 2

claim:  $\forall r \in \mathbb{F}_p, \Pr_s \left[ g_f(r) = \sum_{i=1}^{d+1} c_i f(r+is) \right] \geq 1 - 2 \cdot (d+1) \cdot \Pr[T^f=0]$

proof:

$$\Pr_s \left[ g_f(r) = \sum_{i=1}^{d+1} c_i f(r+is) \right] = \max_{v \in \mathbb{F}_p} \Pr_s \left[ v = \sum_{i=1}^{d+1} c_i \cdot f(r+is) \right]$$

$$\begin{aligned} \sum_i p_i^2 \leq \max_i \{p_i\} \cdot \sum_i p_i &\rightarrow \geq \sum_{v \in \mathbb{F}_p} \Pr_s \left[ v = \sum_{i=1}^{d+1} c_i \cdot f(r+is) \right]^2 \\ &= \Pr_{s,t} \left[ \sum_{i=1}^{d+1} c_i f(r+is) = \sum_{i=1}^{d+1} c_i f(r+it) \right] \\ &\geq 1 - 2(d+1) \Pr[T^f=0] \end{aligned}$$



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

then  $\sum_{i=1}^{d+1} c_i \cdot f(r+is) = \sum_{i=1}^{d+1} c_i \sum_{j=1}^{d+1} c_j f((r+is)+jt) = \sum_{j=1}^{d+1} c_j \sum_{i=1}^{d+1} c_i f((r+js)+is) = \sum_{j=1}^{d+1} c_j f(r+js)$

Hence:

$$\Pr_{s,t} \left[ \sum_{i=1}^{d+1} c_i f(r+is) \neq \sum_{i=1}^{d+1} c_i f(r+it) \right] \leq \Pr_{s,t} \left[ \begin{array}{l} \exists i f(r+is) \neq \sum_{j=1}^{d+1} c_j f((r+is)+jt) \\ \text{or} \\ \exists j f(r+js) \neq \sum_{i=1}^{d+1} c_i f((r+js)+is) \end{array} \right] \leq 2(d+1) \Pr[T^f=0].$$



# Analysis of the RS Test - Part 3

theorem:  $\Pr[T^f=0] \geq \min\{\Omega(\frac{1}{d^2}), \frac{1}{2} \cdot \Delta(f, \mathbb{F}_p^{\leq d}[X])\}$

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

We proved that  $\Pr[T^f=0] \geq \frac{1}{2} \cdot \Delta(g_f, f)$  &  $\forall r \in \mathbb{F}_p, \Pr_S [g_f(r) = \sum_{i=1}^{d+1} c_i f(r+is)] \geq 1 - 2 \cdot (d+1) \Pr[T^f=0]$ .

If  $\Pr[T^f=0] \geq \frac{1}{4 \cdot (d+2)^2}$  then we are done. So assume that  $\Pr[T^f=0] < \frac{1}{4 \cdot (d+2)^2}$ .

We prove that  $g_f \in \mathbb{F}_p^{\leq d}[X]$ , so we are done as  $\Pr[T^f=0] \geq \frac{1}{2} \Delta(g_f, f) = \frac{1}{2} \cdot \Delta(f, \mathbb{F}_p^{\leq d}[X])$ .

claim: if  $\Pr[T^f=0] < \frac{1}{4 \cdot (d+2)^2}$  then  $\forall r, s \in \mathbb{F}_p \sum_{i=0}^{d+1} c_i g_f(r+is) = 0$

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

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

Hence by union bound:

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

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

# Extending the RS Test to Multivariate Polynomials

The local characterization holds similarly: *refers to total degree*

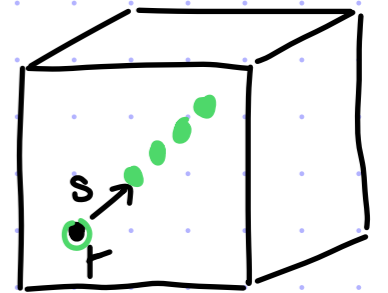
$$\forall d < p, \forall f: \mathbb{F}_p^n \rightarrow \mathbb{F}_p \quad f \in \mathbb{F}^{\leq d}[x_1, \dots, x_n] \text{ iff } \forall a, b \in \mathbb{F}_p \sum_{i=0}^{d+1} c_i f(a+ib) = 0$$

The test is also similar:

*query complexity is  $d+2 = O(d)$*

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

1. sample  $r, s \in \mathbb{F}_p^n$
2. query  $f$  at  $r, r+s, \dots, r+(d+1)s$
3. check that  $\sum_{i=0}^{d+1} c_i \cdot f(r+is) = 0$



random-line test

The theorem for soundness is also similar:

theorem:  $\Pr[T^f = 0] \geq \min\left\{\Omega\left(\frac{1}{d^2}\right), \frac{1}{2} \cdot \Delta(f, \mathbb{F}_p^{\leq d}[x_1, \dots, x_n])\right\}$

And its proof is the same up to syntactic modifications!

In sum, by repeating the test  $O(d^2)$  times, we get:

a low-degree test with query complexity  $O(d^3)$  [independent of  $n$ !] where constant relative distance  $\rightarrow$  constant soundness error.

# 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 these polynomials. We show that these polynomial norms behave “randomly” in some sense, and we prove a form of an inverse theorem for

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

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

July 9, 2003

### Abstract

We describe an efficient randomized algorithm to test if a given binary function  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  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  $O(\frac{1}{\epsilon} + k4^k)$  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.

## 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 can only verify the satisfiability of a boolean formula by reading a 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, each reading logarithmic number of bits, and such that

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).