

# Low degree tests at large distances

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# Notions in CS

Informal - some suspension of disbelief is asked for

- NP is the class of mathematical statements with easily verifiable (short) proofs.
- CL'71: Reduction to verifying that a given 3-CNF boolean formula is satisfiable.
- A...S'92, D'05 PCP: Reduction to distinguishing between a satisfiable 3-CNF boolean formula, and a significantly unsatisfiable formula - an optimal assignment leaves a positive fraction of terms unsatisfied.
- R' 95 Parallel Repetition: Invalid statement is translated into a very unsatisfiable formula - an optimal assignment leaves a  $(1 - \epsilon)$ -fraction of terms unsatisfied.

# Notions in CS - continued

- BGS'95, H'97: A format for proving satisfiability which allows verification by looking at **tiny randomized samples** from the proof.
- A proof is partitioned into 0-1 strings of length  $2^n$  viewed as **functions**  $f_i : \{0, 1\}^n \rightarrow \{0, 1\}$ .
- In a **valid** proof all the functions  $f_i$  are **structured**. In any proof of an **invalid** statement, many of the functions are **not structured**.

# Important building block

$\mathbb{F}$  is a finite field. Given  $f : \mathbb{F}^n \rightarrow \mathbb{F}$ , determine if

- $f$  is a **low-degree  $n$ -variate polynomial**. (**structured**)
- $f$  is  **$\epsilon$ -far** from all low-degree polynomials:

$$\Pr_x\{f(x) \neq g(x)\} \geq \epsilon$$

for any degree- $d$  polynomial  $g$ . (**not structured**)

- Allowed only **local tests** - may query the function only at a few points.
- May use **randomization**.

# Generalization - Property testing

Given a large combinatorial object  $G$ , determine if

- $G$  has a global property  $P$ .
- $f$  is  $\epsilon$ -far from all objects with property  $P$
- Only randomized local queries to  $G$  are allowed.
- Ex. Given a graph  $G$  on  $k$  vertices determine whether  $G$  is bi-partite or requires removal of at least  $\epsilon k^2$  edges to become bi-partite, by querying a small number of edges of  $G$  (AK '02).

# Specification - extremal polynomiality testing

## Main question

Given  $f : \mathbb{F}_2^n \rightarrow \mathbb{F}_2$ , determine if

- $f$  is a **low-degree  $n$ -variate polynomial**.
- $f$  is **very far** from all low-degree polynomials:

$$Pr_x\{f(x) \neq g(x)\} \geq \frac{1}{2} - \epsilon$$

for any degree- $d$  polynomial  $g$ .

- Allowed only **local tests** - may query the function only at a few points.

# Linear polynomials

- Distinguishing between **linear** and **far from linear** functions.
- This case is known. Plays an important role in **PCP** constructions.
- BLR'93, BCHKS'96 - A local test, **querying**  $f$  at **3** points and **returning 1** bit, which behaves
  - **Deterministically** for linear functions
  - **Randomly** for functions far from linear
- The test makes **3** queries and distinguishes linear and far from linear functions w.p.  **$1/2$** .
- H'97: Can be “lifted” to a PCP construction with same parameters.

# Pseudorandomness

- Point of view: Linear functions are **structured**, functions far from linear are **pseudorandom** - allowing to extract one **random** bit.
- In fact, this definition of pseudorandomness for a function  $f$  is equivalent to the usual one:  $f$  has small **Fourier coefficients**.
- Need to distinguish between **pseudorandomness** and **structure**.



# Motivation: stronger linearity tests

- Want to optimize the ratio

$$\rho = \frac{q}{\log_2 1/p}$$

where  $q$  is the number of queries and  $p$  is the probability the test succeeds.

- For the previous test  $\rho = 3/\log(2) = 3$ .
- Want to have a test with  $\rho = 1 + o_q(1)$ .

# Motivation: stronger linearity tests

- Want to have a test with  $\rho = 1 + o_q(1)$ .
- ST'00 - A local test, querying  $f$  at  $q$  points and returning  $q - \sqrt{2q}$  bits, which behaves
  - Deterministically for linear functions
  - Randomly for pseudorandom (far from linear) functions
- The test makes  $q$  queries and distinguishes linear and pseudorandom functions w.p.  $2^{-q+\sqrt{2q}}$ .

$$\rho = \frac{q}{q - \sqrt{2q}} = 1 + o(1)$$

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- The test makes  $q$  queries and distinguishes linear and pseudorandom functions w.p.  $2^{-q+\sqrt{2q}}$ .
- Lifting to a PCP construction with similar parameters.
- Can we squeeze out even **more** randomness? How **powerful** is this notion of pseudorandomness?

# Local tests for pseudorandomness

## Structure vs. pseudorandomness

Let  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  be a boolean function.

- BLR'93, BCHKS'96: Choose  $x, y \in \{0, 1\}^n$  at random.

Compute

$$f(x) + f(y) + f(x + y)$$

Makes 3 queries, returns 1 useful bit.

- ST'00: Graph tests. Let  $G = (V, E)$  be a graph on  $k$  vertices. Choose  $x_1 \dots x_k \in \{0, 1\}^n$  at random. For all  $(i, j) \in E$  compute

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- If  $G$  is the **complete graph**: makes  $q$  queries, returns  $q - \sqrt{2q}$  bits.

# Even better tests for pseudorandomness

## Structure vs. pseudorandomness

- Let  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  be a boolean function.
- ST'00: **Hypergraph tests**. Let  $G = (V, E)$  be a **hypergraph** on  $k$  vertices. Choose  $x_1 \dots x_k \in \{0, 1\}^n$  at random. For all  $e = (x_i)_{i \in e} \in E$  compute

$$\sum_{i \in e} f(x_i) + f\left(\sum_{i \in e} x_i\right)$$

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# Even better ?? tests for pseudorandomness

Doesn't work...

- Let  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  be a boolean function.
- ST'00: **Hypergraph tests**. Let  $G = (V, E)$  be a **hypergraph** on  $k$  vertices. Choose  $x_1 \dots x_k \in \{0, 1\}^n$  at random. For all  $e = (x_i)_{i \in e} \in E$  compute

$$\sum_{i \in e} f(x_i) + f\left(\sum_{i \in e} x_i\right)$$

Makes  $|E| + |V|$  queries, returns  $|E|$  **useless** bits.

- If  $G$  is the **complete hypergraph**: makes  $q$  queries, returns  $q - \log q$  **bad** bits.



# An inconvenient example

Let  $n$  be even, and let

$$f(x) = x(1) \cdot x(2) + x(3) \cdot x(4) + \dots + x(n-1) \cdot x(n)$$

- $f$  is **gent** (maximally far from all linear functions).
- ST'00: Any **hypergraph linearity test** with  $q$  queries that accepts linear functions **accepts**  $f$  with probability at least  $2^{-q+\sqrt{2q}}$ .

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- +BHR'03, L'07: Any linearity test with  $q$  queries that accepts linear functions accepts  $f$  with probability at least  $2^{-q+\sqrt{2q}}$ .

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- +BHR'03, L'07: Any linearity test with  $q$  queries that accepts linear functions accepts  $f$  with probability at least  $2^{-q+\sqrt{2q}}$ .
- **What's going on?** The function  $f$  must have a **hidden structure**.

# Property testing

## Low degree polynomials

Let  $\mathbb{F}$  be a finite field. Given  $f : \mathbb{F}^n \rightarrow \mathbb{F}$ , determine if

- $f$  is a polynomial of (low) degree at most  $d$ .
  - $f$  is  $\epsilon$ -far from all degree- $d$  polynomials.
  - Usually the field is large.
- BFL'91: If  $|\mathbb{F}| > d + 1$  - restrict  $f$  to a random line and check it's a degree- $d$  univariate polynomial.
- Always accepts degree- $d$  polynomials.
  - If  $f$  is  $\epsilon$ -far from degree- $d$  polynomials, rejects after  $T(\epsilon, d)$  random restrictions.
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- $f$  is  $\epsilon$ -far from all degree- $d$  polynomials.
- AKKLR'03: What if the field is small,  $\mathbb{F} = \mathbb{F}_2$ ?

Restrict  $f$  to a random  $(d + 1)$ -dimensional affine subspace and check it's a degree- $d$  polynomial.

- Always accepts degree- $d$  polynomials.
- If  $f$  is  $\epsilon$ -far from degree- $d$  polynomials, rejects after  $T(\epsilon, d)$  random restrictions.
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# Low-degree testing over $\mathbb{F}_2$

Following AKKLR'03

- To test if  $f$  is a degree- $d$  polynomial, compute a random  $(d + 1)$ -st directional derivative of  $f$ .
- If  $f$  is degree- $d$  this derivative is **always zero**.
- If it's zero with high probability, then  $f$  is **close to** a degree- $d$  polynomial.



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- The function

$$f(x) = x(1) \cdot x(2) + x(3) \cdot x(4) + \dots + x(n-1) \cdot x(n)$$

is a **quadratic polynomial**

- $f$  is pseudorandom for **linearity tests** but structured for **higher-degree tests**

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is a **quadratic polynomial**

- $f$  is pseudorandom for **linearity tests** but structured for **higher-degree tests**
- **Want stronger notion of pseudorandomness**

# Pseudorandomness I: Balanced derivatives

## A technical notion

- A function is  **$d$ -pseudorandom** if the probability that its random  $(d + 1)$ -st derivative is zero is very close to  $1/2$ .
- An analytic pseudorandomness measure for a boolean function  $f$ :

$d$ -pseudorandomness of  $f$  :

$$(2P(f) - 1)^{1/2^d}$$

where  $P(f)$  is the probability that  $f$  restricted to a random  $(d + 1)$ -dimensional affine subspace of the cube is a degree- $d$  polynomial.

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- The  $1/2^d$ -root is to deal with various notions of derivatives.

# Pseudorandomness I: Gowers Uniformity

## A technical notion

- Defined in G'01 - for functions on  $\mathbb{Z}_n$ .
- An analytic pseudorandomness measure for a boolean function  $f$ :

Gowers uniformity of degree  $d$  of  $f$ :

$$\left( \mathbb{E}_{x, y_1, \dots, y_d} (-1)^{\sum_{S \subseteq [d]} f(x + \sum_{i \in S} y_i)} \right)^{1/2^d}$$

- A function is **pseudorandom** if its Gowers uniformity is small.

# A stronger linearity test given low Gowers uniformity

- S'05, ST'06 - A local test, querying  $q$  bits and returning  $q - q^{1/d}$  bits, which behaves
  - Deterministically for linear functions
  - Randomly for pseudorandom (low degree- $d$  Gowers uniformity) functions.
- The test makes  $q$  queries and distinguishes linear and pseudorandom functions w.p.  $2^{-q+q^{1/d}}$ .
- ST'06 Conditional lifting to a PCP construction with similar parameters.



# Pseudorandomness II - Polynomial pseudorandomness

## Structure

**Definition:** A function is **degree- $d$  pseudorandom** if it is **far** from degree- $d$  polynomials.

- The "**right**" notion we seem to be looking for.
- Additional dividends: **explicit** degree- $d$  pseudorandom functions for large  $d$  lead to interesting **lower bounds** and **pseudorandom generator constructions**.

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- The **"right"** notion we seem to be looking for.
- Additional dividends: **explicit** degree- $d$  pseudorandom functions for large  $d$  lead to interesting **lower bounds** and **pseudorandom generator constructions**.
- Can we compare the two notions of pseudorandomness?
- G'01, GT'05: Low Gowers Uniformity of degree  $d$  implies polynomial degree- $d$  pseudorandomness.
- **The other direction?**

# Lack of pseudorandomness should imply structure

## Inverse claims

- What if a function  $f$  has a **non-negligible** Gowers Uniformity?

$$\|f\|_{U_d} > \epsilon$$

- $d = 2$ : In this case  $f$  is  $1/2 - \epsilon$  close to a linear function BLR'93, BCHKS'96.
- $\epsilon$  is **BIG**,  $\epsilon = 1 - \delta$ . In this case  $f$  is  $\delta'$ -close to a degree- $(d - 1)$  polynomial AKKLR'03.
- $d = 3$ : In this case  $f$  is  $1/2 - \epsilon'$  close to a degree-2 polynomial GT'05, S'05.
- Any  $d$ :  $f$  has a variable whose **influence** is at least  $\epsilon'/2^d$  ST'06.

# An inverse conjecture for Gowers uniformity

**Conjecture** T'07 (GT'05), S'05: The two notions of pseudorandomness are **equivalent**:  $\|f\|_{U_{d+1}} > \epsilon$  implies  $f$  is  $1/2 - \epsilon'$  close to a degree- $d$  polynomial.

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- Discussion
  - If this conjecture were true, this would give a concise description of Gowers uniformity.
  - It is “equivalent” to **low-degree testing at large distances**.

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- BV'07: A **weaker conjecture**, useful for constructing **pseudorandom generators**: May also assume  $f$  is a polynomial of degree  $d + 1$ .

# The conjecture is false

- GT'07, LMS'07: The conjecture is false, even for  $d = 4$  and for  $f$  a polynomial of degree 4.
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- GT'07: Partial positive results for larger fields.
- Counterexample:  $f = S_4$  is a symmetric polynomial of degree 4.

$$f(x) = \sum_{|S|=4} \prod_{i \in S} x(i)$$

- $\|f\|_{U_4} > 0.9$
- $f$  is  $(\frac{1}{2} - \exp\{-cn\})$ -far from cubic polynomials.



# The conjecture is false

- GT'07, LMS'07: The conjecture is false, even for  $d = 4$  and for  $f$  a polynomial of degree 4.
- GT'07: Partial positive results for larger fields.
- Question. Assume a big family of degree-4 derivatives of  $f$  are non-negligibly imbalanced. Does this imply  $f$  is somewhat close to a cubic polynomial?
- BL'08: There is a version of a degree-4 derivative which is negligible for  $S_4$ .

# Some details

$S_4$  has large 4-uniformity.

- The directional derivative of  $S_4$  in directions  $y_1, y_2, y_3, y_4$  is:

$$\sum_{|S|=4} \text{Det}_S(y_1 \cdots y_4) = \sum_{|S|=4} \text{Det}_S^2(y_1 \cdots y_4) = \text{Det}(\langle y_i, y_j \rangle)$$

- The behavior of a random  $4 \times 4$  matrix  $(\langle y_i, y_j \rangle)$  is not hard to analyze.

# Some details

$S_4$  is far from cubics.

- A correlation between functions is upperbounded by average correlation between their derivatives.

$$\langle f, g \rangle^8 \leq \mathbb{E}_{y,z} \langle f_{y,z}, g_{y,z} \rangle^2$$

- Let  $f = S_4$ ,  $g$  a cubic. Second derivative of  $f$  is quadratic, depending on  $y, z$ . Second derivative of  $g$  is linear.
- By Dixon's theorem know the Fourier spectrum of quadratic polynomials. Need multilinear algebra to wrap this together.