

# Background on Higher-Order Fourier Analysis

FOCS '14 Workshop

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# Plan for the day

- **Talk 1:** Mathematical primer (me)
- **Talk 2:** Polynomial pseudorandomness (P. Hatami)
- **Talk 3:** Algorithmic h.o. Fourier analysis (Tulsiani)
- **Talk 4:** Applications to property testing (Yoshida)
- **Talk 5:** Applications to coding theory (Bhowmick)
- **Talk 6:** A different generalization of Fourier analysis and application to communication complexity (Viola)

# Teaser

Given a quartic polynomial  $P: \mathbb{F}_2^n \rightarrow \mathbb{F}_2$ ,  
can we decide in  $\text{poly}(n)$  time whether:

$$P = Q_1 Q_2 + Q_3 Q_4$$

where  $Q_1, Q_2, Q_3, Q_4$  are quadratic polys?

**Yes! [B. '14, B.-Hatami-Tulsiani '15]**

# Some Preliminaries

# Setting

$\mathbb{F}$  = finite field of **fixed**  
**prime** order

- For example,  $\mathbb{F} = \mathbb{F}_2$  or  $\mathbb{F} = \mathbb{F}_{97}$
- Theory simpler for fields of large (but fixed) size

# Functions

Functions are always multivariate,  
on  $n$  variables

$$f: \mathbb{F}^n \rightarrow \mathbb{C} \quad (|f| \leq 1)$$

and

$$P: \mathbb{F}^n \rightarrow \mathbb{F}$$

Current bounds  
aim to be  
efficient wrt  $n$

# Polynomial

**Polynomial of degree  $d$**  is of the form:

$$\sum_{i_1, \dots, i_n} c_{i_1, \dots, i_n} x_1^{i_1} \cdots x_n^{i_n}$$

where each  $c_{i_1, \dots, i_n} \in \mathbb{F}$  and  $i_1 + \cdots + i_n \leq d$

# Phase Polynomial

**Phase polynomial of degree  $d$**  is a function  $f: \mathbb{F}^n \rightarrow \mathbb{C}$  of the form  $f(x) = e(P(x))$  where:

1.  $P: \mathbb{F}^n \rightarrow \mathbb{F}$  is a polynomial of degree  $d$
2.  $e(k) = e^{2\pi i k / |\mathbb{F}|}$

# Inner Product

The **inner product** of two functions  $f, g: \mathbb{F}^n \rightarrow \mathbb{C}$  is:

$$\langle f, g \rangle = \mathbb{E}_{x \in \mathbb{F}^n} [f(x) \cdot \overline{g(x)}]$$

Magnitude captures correlation between  $f$  and  $g$

# Derivatives

## Additive derivative in direction

$h \in \mathbb{F}^n$  of function  $P: \mathbb{F}^n \rightarrow \mathbb{F}$  is:

$$D_h P(x) = P(x + h) - P(x)$$

# Derivatives

**Multiplicative derivative in direction**  $h \in \mathbb{F}^n$  of function  $f: \mathbb{F}^n \rightarrow \mathbb{C}$  is:

$$\Delta_h f(x) = f(x + h) \cdot \overline{f(x)}$$

# Polynomial Factor

**Factor of degree  $d$  and order  $m$**  is a tuple of polynomials

$\mathcal{B} = (P_1, P_2, \dots, P_m)$ , each of degree  $d$ .

As shorthand, write:

$$\mathcal{B}(x) = (P_1(x), \dots, P_m(x))$$

# Fourier Analysis over $\mathbb{F}$

# Fourier Representation

Every function  $f: \mathbb{F}^n \rightarrow \mathbb{C}$  is a linear combination of **linear** phases:

$$f(x) = \sum_{\alpha \in \mathbb{F}^n} \hat{f}(\alpha) e\left(\sum_i \alpha_i x_i\right)$$

# Linear Phases

- The inner product of two linear phases is:

$$\langle e\left(\sum_i \alpha_i x_i\right), e\left(\sum_i \beta_i x_i\right) \rangle = \mathbb{E}_x \left[ e\left(\sum_i (\alpha_i - \beta_i) x_i\right) \right] = 0$$

if  $\alpha \neq \beta$  and is 1 otherwise.

- So:

$$\hat{f}(\alpha) = \langle f, e(\sum_i \alpha_i x_i) \rangle = \text{correlation with linear phase}$$

# Random functions

With high probability, a **random** function  $f: \mathbb{F}^n \rightarrow \mathbb{C}$  with  $|f| = 1$  has each  $\hat{f}(\alpha) \rightarrow 0$ .

# Decomposition Theorem

$$f(x) = g(x) + h(x)$$

where:

$$g(x) = \sum_{\alpha: \hat{f}(\alpha) \geq \epsilon} \hat{f}(\alpha) \cdot e \left( \sum_i \alpha_i x_i \right)$$

$$h(x) = \sum_{\alpha: \hat{f}(\alpha) < \epsilon} \hat{f}(\alpha) \cdot e \left( \sum_i \alpha_i x_i \right)$$

# Decomposition Theorem

$$g(x) = \sum_{\alpha: \hat{f}(\alpha) \geq \epsilon} \hat{f}(\alpha) \cdot e\left(\sum_i \alpha_i x_i\right)$$
$$h(x) = \sum_{\alpha: \hat{f}(\alpha) < \epsilon} \hat{f}(\alpha) \cdot e\left(\sum_i \alpha_i x_i\right)$$

Every Fourier coefficient of  $h$  is less than  $\epsilon$ , so  $h$  is “pseudorandom”.

# Decomposition Theorem

$$g(x) = \sum_{\alpha: \hat{f}(\alpha) \geq \epsilon} \hat{f}(\alpha) \cdot e\left(\sum_i \alpha_i x_i\right)$$

$$h(x) = \sum_{\alpha: \hat{f}(\alpha) < \epsilon} \hat{f}(\alpha) \cdot e\left(\sum_i \alpha_i x_i\right)$$

$g$  has only  $1/\epsilon^2$  nonzero Fourier coefficients

# Decomposition Theorem

$$g(x) = \sum_{\alpha: \hat{f}(\alpha) \geq \epsilon} \hat{f}(\alpha) \cdot e\left(\sum_i \alpha_i x_i\right)$$
$$h(x) = \sum_{\alpha: \hat{f}(\alpha) < \epsilon} \hat{f}(\alpha) \cdot e\left(\sum_i \alpha_i x_i\right)$$

The nonzero Fourier coefficients of  $g$  can be found in poly time

**[Goldreich-Levin '89]**

# Elements of Higher-Order Fourier Analysis

**Higher-order Fourier analysis** is the interplay between three different notions of pseudorandomness for functions and factors.

1. Bias
2. Gowers norm
3. Rank

# Bias

# Bias

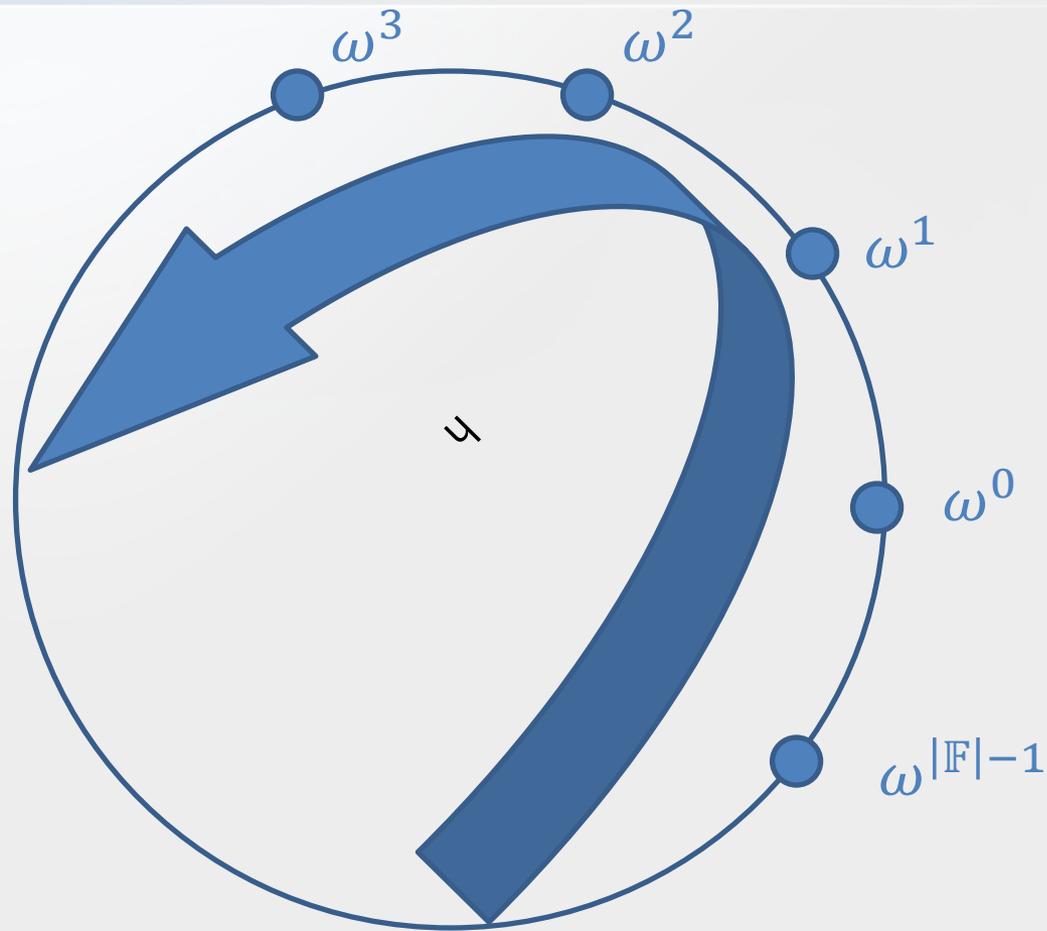
For  $f: \mathbb{F}^n \rightarrow \mathbb{C}$ ,

$$\text{bias}(f) = |\mathbb{E}_x[f(x)]|$$

For  $P: \mathbb{F}^n \rightarrow \mathbb{F}$ ,

$$\text{bias}(P) = |\mathbb{E}_x[\mathbf{e}(P(x))]|$$

[..., Naor-Naor '89, ...]



How well is  $P$  equidistributed?

## Bias of Factor

A factor  $\mathcal{B} = (P_1, \dots, P_k)$  is  **$\alpha$ -unbiased** if every nonzero linear combination of  $P_1, \dots, P_k$  has bias less than  $\alpha$ :

$$\text{bias}\left(\sum_{i=1}^k c_i P_i\right) < \alpha$$
$$\forall (c_1, \dots, c_k) \in \mathbb{F}^k \setminus \{0\}$$

# Bias implies equidistribution

Lemma: If  $\mathcal{B}$  is  $\alpha$ -unbiased and of order  $k$ , then for any  $c \in \mathbb{F}^k$ :

$$\Pr[\mathcal{B}(x) = c] = \frac{1}{|\mathbb{F}|^k} \pm \alpha$$

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$$\Pr[\mathcal{B}(x) = c] = \frac{1}{|\mathbb{F}|^k} \pm \alpha$$

Corollary: If  $\mathcal{B}$  is  $\alpha$ -unbiased and  $\alpha < \frac{1}{|\mathbb{F}|^k}$ , then  $\mathcal{B}$  maps onto  $\mathbb{F}^k$ .

# Gowers Norm

# Gowers Norm

Given  $f: \mathbb{F}^n \rightarrow \mathbb{C}$ , its **Gowers norm of order  $d$**  is:

$$U^d(f) = |\mathbb{E}_{x, h_1, h_2, \dots, h_d} \Delta_{h_1} \Delta_{h_2} \cdots \Delta_{h_d} f(x)|^{1/2^d}$$

# Gowers Norm

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**Observation**: If  $f = e(P)$  is a phase poly, then:

$$U^d(f) = |\mathbb{E}_{x, h_1, h_2, \dots, h_d} e(D_{h_1} D_{h_2} \cdots D_{h_d} P(x))|^{1/2^d}$$

## Gowers norm for phase polys

- If  $f$  is a phase poly of degree  $d$ , then:

$$U^{d+1}(f) = 1$$

- Converse is true when  $d < |\mathbb{F}|$ .

# Other Observations

- $U^1(f) = \sqrt{|\mathbb{E}[f]|^2} = \text{bias}(f)$

- $U^2(f) = \sqrt[4]{\sum_{\alpha} \hat{f}^4(\alpha)}$

- $U^1(f) \leq U^2(f) \leq U^3(f) \leq \dots$  (C.-S.)

# Pseudorandomness

- For random  $f: \mathbb{F}^n \rightarrow \mathbb{C}$  and fixed  $d$ ,  
$$U^d(f) \rightarrow 0$$
- By monotonicity, low Gowers norm implies low bias and low Fourier coefficients.

# Correlation with Polynomials

**Lemma**:  $U^{d+1}(f) \geq \max |\langle f, e(P) \rangle|$   
where max is over all polynomials  $P$  of degree  $d$ .

**Proof**: For any poly  $P$  of degree  $d$ :

$$\begin{aligned} |\mathbb{E}[f(x) \cdot e(-P(x))]| &= U^1(f \cdot e(-P)) \\ &\leq U^{d+1}(f \cdot e(-P)) \\ &= U^{d+1}(f) \end{aligned}$$

# Gowers Inverse Theorem

**Theorem**: If  $d < |\mathbb{F}|$ , for all  $\epsilon > 0$ , there exists  $\delta = \delta(\epsilon, d, \mathbb{F})$  such that if  $U^{d+1}(f) > \epsilon$ , then  $|\langle f, e(P) \rangle| > \delta$  for some poly  $P$  of degree  $d$ .

## **Proof**:

- **[Green-Tao '09]** Combinatorial for phase poly  $f$  (c.f. Madhur's talk later).
- **[Bergelson-Tao-Ziegler '10]** Ergodic theoretic proof for arbitrary  $f$ .

# Small Fields

Consider  $f: \mathbb{F}_2^1 \rightarrow \mathbb{C}$  with:

$$f(0) = 1$$

$$f(1) = i$$

$f$  not a phase poly but  $U^3(f) = 1!$

## Small fields: worse news

Consider  $f = e(P)$  where  $P: \mathbb{F}_2^n \rightarrow \mathbb{F}_2$  is symmetric polynomial of degree 4.

$$U^4(f) = \Omega(1)$$

but:

$$|\langle f, e(C) \rangle| = \exp(-n)$$

for all cubic poly  $C$ .

[Lovett-Meshulam-Samorodnitsky '08, Green-Tao '09]

Nevertheless...

Just **define non-classical phase polynomials of degree  $d$**  to be functions  $f: \mathbb{F}^n \rightarrow \mathbb{C}$  such that  $|f| = 1$  and

$$\Delta_{h_1} \Delta_{h_2} \cdots \Delta_{h_{d+1}} f(x) = 1$$

for all  $x, h_1, \dots, h_{d+1} \in \mathbb{F}^n$

# Inverse Theorem for small fields

**Theorem**: For all  $\epsilon > 0$ , there exists  $\delta = \delta(\epsilon, d, \mathbb{F})$  such that if  $U^{d+1}(f) > \epsilon$ , then  $|\langle f, g \rangle| > \delta$  for some non-classical phase poly  $g$  of degree  $d$ .

## **Proof**:

- [Tao-Ziegler] Combinatorial for phase poly  $f$ .
- [Tao-Ziegler] Nonstandard proof for arbitrary  $f$ .

# Pseudorandomness & Counting

**Theorem**: If  $L_1, \dots, L_m$  are  $m$  linear forms  
( $L_j(X_1, \dots, X_k) = \sum_{i=1}^k \ell_{i,j} X_i$ ), then:

$$\mathbb{E}_{X_1, \dots, X_k \in \mathbb{F}^n} \left[ \prod_{j=1}^m f(L_j(X_1, \dots, X_k)) \right] \leq U^t(f)$$

if  $f: \mathbb{F}^n \rightarrow \mathbb{C}$  and  $t$  is the *complexity* of the linear forms  $L_1, \dots, L_m$ .

# Examples

- If  $f: \mathbb{F}^n \rightarrow \{0,1\}$  indicates a subset and we want to count the number of 3-term AP's:

$$\mathbb{E}_{X,Y}[f(X) \cdot f(X + Y) \cdot f(X + 2Y)] \leq \sum_{\alpha} \hat{f}^3(\alpha)$$

- Similarly, number of 4-term AP's controlled by 3rd order Gowers norm of  $f$ .
- More in Pooya's upcoming talk!

# Rank

# Rank

Given a polynomial  $P: \mathbb{F}^n \rightarrow \mathbb{F}$  of degree  $d$ , its **rank** is the smallest integer  $r$  such that:

$$P(x) = \Gamma(Q_1(x), \dots, Q_r(x)) \quad \forall x \in \mathbb{F}^n$$

where  $Q_1, \dots, Q_r$  are polys of degree  $d - 1$  and  $\Gamma: \mathbb{F}^r \rightarrow \mathbb{F}$  is arbitrary.

# Pseudorandomness

- For random poly  $P$  of fixed degree  $d$ ,  
$$\text{rank}(P) = \omega(1)$$
- High rank is pseudorandom behavior

# Rank & Gowers Norm

If  $P: \mathbb{F}^n \rightarrow \mathbb{F}$  is a poly of degree  $d$ ,  $P$  has high rank **if and only if**  $e(P)$  has low Gowers norm of order  $d$ !

# Low rank implies large Gowers norm

**Lemma:** If  $P(x) = \Gamma(Q_1(x), \dots, Q_k(x))$   
where  $Q_1, \dots, Q_k$  are polys of deg  $d - 1$ , then  
$$U^d(e(P)) \geq \frac{1}{|\mathbb{F}|^{k/2}}.$$

# Low rank implies large Gowers norm

**Lemma:** If  $P(x) = \Gamma(Q_1(x), \dots, Q_k(x))$  where  $Q_1, \dots, Q_k$  are polys of deg  $d - 1$ , then  $U^d(\mathbf{e}(P)) \geq \frac{1}{|\mathbb{F}|^{k/2}}$ .

Proof: By (linear) Fourier analysis:

$$\mathbf{e}(P(x)) = \sum_{\alpha} \hat{\Gamma}(\alpha) \cdot \mathbf{e}\left(\sum_i \alpha_i \cdot Q_i(x)\right)$$

Therefore:

$$|\mathbb{E}_x \sum_{\alpha} \hat{\Gamma}(\alpha) \cdot \mathbf{e}\left(\sum_i \alpha_i \cdot Q_i(x) - P(x)\right)| = 1$$

Then, there's an  $\alpha$  such that

$$\langle \mathbf{e}(P), \mathbf{e}(\sum_i \alpha_i Q_i) \rangle \geq |\mathbb{F}|^{-k/2}.$$

# Inverse theorem for polys

**Theorem**: For all  $\epsilon$  and  $d$ , there exists  $R = R(\epsilon, d, \mathbb{F})$  such that if  $P$  is a poly of degree  $d$  and  $U^d(e(P)) > \epsilon$ , then  $\text{rank}(P) < R$ .

# Bias-rank theorem

**Theorem**: For all  $\epsilon$  and  $d$ , there exists  $R = R(\epsilon, d, \mathbb{F})$  such that if  $P$  is a poly of degree  $d$  and  $\text{bias}(P) > \epsilon$ , then  $\text{rank}(P) < R$ .

[Green-Tao '09, Kaufman-Lovett '08]

# Regularity of Factor

A factor  $\mathcal{B} = (P_1, \dots, P_k)$  is  **$R$ -regular** if every nonzero linear combination of  $P_1, \dots, P_k$  has rank more than  $R$ .

# An Example

**Claim:** If a factor  $\mathcal{B} = (P_1, \dots, P_k)$  of degree  $d$  is sufficiently regular, then for any poly  $Q$  of degree  $d$ , there can be at most one  $P_i$  that is  $\epsilon$ -correlated with  $Q$ .

# An Example

**Claim:** If a factor  $\mathcal{B} = (P_1, \dots, P_k)$  of degree  $d$  is sufficiently regular, then for any poly  $Q$  of degree  $d$ , there can be at most one  $P_i$  that is  $\epsilon$ -correlated with  $Q$ .

**Proof:**

$Q$   $\epsilon$ -correlated with  $P_i$    $\text{bias}(Q - P_i) > \epsilon$

$Q$   $\epsilon$ -correlated with  $P_j$    $\text{bias}(Q - P_j) > \epsilon$

So,  $\text{rank}(Q - P_i), \text{rank}(Q - P_j)$  bounded. But then  $\text{rank}(P_i - P_j)$  bounded, a contradiction.

# Things I didn't talk about

- Decomposition theorem
  - For any  $d, R, \epsilon$ , given function  $f: \mathbb{F}^n \rightarrow \mathbb{C}$ , can find functions  $f_S$  and  $f_R$  such that  $f = f_S + f_R$ ,  $U^{d+1}(f_R) < \epsilon$ , and  $f_S = \Gamma(\mathcal{B})$  for a factor  $\mathcal{B}$  of rank  $R$  and constant order.
- Gowers' proof of Szemerédi's theorem
- Ergodic-theoretic aspects

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# A final example

**Claim**: Let  $\mathcal{B} = (P_1, \dots, P_m)$  be a sufficiently regular factor of degree  $d$ . Define:

$$F(x) = \Gamma(P_1(x), \dots, P_m(x))$$

Then, for any  $Q_1, \dots, Q_m$  with  $\deg(Q_j) \leq \deg(P_j)$ , if

$$G(x) = \Gamma(Q_1(x), \dots, Q_m(x))$$

it holds that:  $\deg(G) \leq \deg(F)$ .

# Sketch of Proof

- Suppose  $D = \deg(F)$ .
- Using (standard) Fourier analysis, write:

$$e(F(x)) = \sum_{\alpha} c_{\alpha} e\left(\sum_i \alpha_i P_i(x)\right)$$

- Now, differentiate above expression  $D + 1$  times to get 1. But all the derivatives of  $\omega^{\sum_i \alpha_i P_i(x)}$  are linearly independent. So, all coefficients of these derivatives cancel formally.
- Can expand out the derivative of  $\omega^{G(x)}$  in the same way to get that it too equals 1.