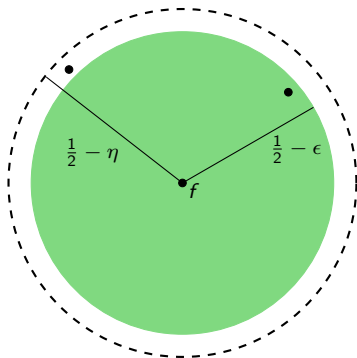


Algorithmic Questions in Higher-Order Fourier Analysis



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Based on joint works with
Arnab Bhattacharyya, Eli
Ben-Sasson, Pooya Hatami,
Noga Ron-Zewi, Luca
Trevisan, Salil Vadhan and
Julia Wolf

Decomposition Theorems

Decomposition Theorems

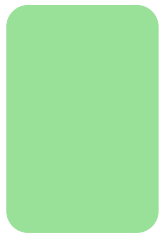


Object of study



Family of
algorithms or
functions

Decomposition Theorems

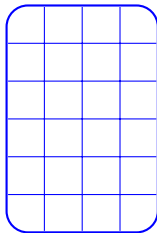


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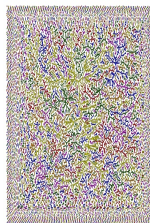
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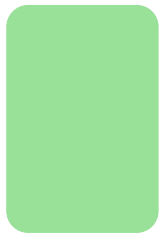
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+



No apparent
structure
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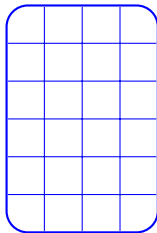


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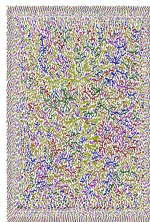
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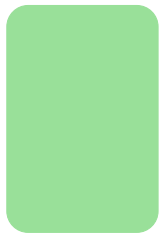
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- Decompose an object in to **structured** and **pseudorandom** parts.

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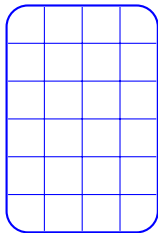


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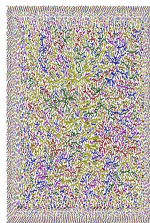
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No apparent
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- Decompose an object in to **structured** and **pseudorandom** parts.
- Can often ignore the pseudorandom part for many applications. Structured part easier to study.

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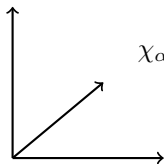
- [Parseval]: $\|g\|^2 = \langle g, g \rangle = \mathbb{E}_x [(g(x))^2] = \sum_\alpha (\hat{g}(\alpha))^2$.

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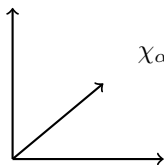


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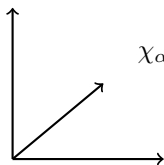


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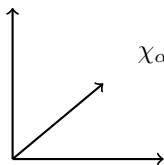
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- $k \leq 1/\epsilon^2$.

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simple structure

- f has small correlation with linear functions.

pseudorandom

$$\forall \alpha, |\langle f, \chi_\alpha \rangle| = |\mathbb{E}_x [f(x) \chi_\alpha(x)]| \leq \epsilon$$

Getting high: Quadratic Fourier Analysis [Gowers 98, Green 07]

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$$\|f\|_{U^2}^4 = \mathbb{E}_{x,y,z} [f(x) \cdot f(x+y) \cdot f(x+z) \cdot f(x+y+z)].$$

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- Can define higher norms similarly

$$\|f\|_{U^3}^8 = \mathbb{E}_{x,y,z,w} \left[\begin{array}{cccc} f(x) & f(x+y) & f(x+z) & f(x+y+z) \\ f(x+w) & f(x+y+w) & f(x+z+w) & f(x+y+z+w) \end{array} \right]$$

Norms, Shnorms... so what?

- $\|f\|_{U^2}$ measures correlation with Fourier characters (linear phase functions).

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 - $\|f\|_{U^3} \geq \epsilon \implies$ for some Q , $|\langle f, (-1)^Q \rangle| \geq \eta(\epsilon)$.

Decompositions in Quadratic Fourier Analysis

Theorem (Gowers-Wolf 09)

Given $\epsilon > 0$, any $g : \mathbb{F}_2^n \rightarrow [-1, 1]$ can be decomposed as

$$g = \sum_{i=1}^k c_i (-1)^{Q_i} + f + e$$

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Similar to basic Fourier decomposition, where we get

$$g = \sum_{i=1}^k c_i \chi_{\alpha_i}(x) + f,$$

with $|\langle f, \chi_\alpha \rangle| \leq \epsilon$ for all α and $k \leq 1/\epsilon^2$ (also implies $\sum_i |c_i| \leq 1/\epsilon$).

Decompositions in Higher-Order Fourier Analysis

Theorem (Gowers-Wolf 10)

Given $\epsilon > 0$ and $p > d$, there exists $M(\epsilon, p)$ such that any $g : \mathbb{F}_p^n \rightarrow [-1, 1]$ can be decomposed as

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- Stronger decomposition theorems proved by [HL 11] and [BFL 12].
- Decomposition theorems for the case when $p \leq d$ require non-classical polynomials.

Q1: Can we compute these decompositions efficiently?

Algorithmic version of the basic Fourier decomposition

Theorem (Goldreich-Levin 89)

There is a randomized algorithm, which given $\epsilon, \delta > 0$ and oracle access to $g : \mathbb{F}_2^n \rightarrow [-1, 1]$, runs in time $O(n^2 \log n \cdot (1/\epsilon^2) \cdot \log(1/\delta))$ and outputs a decomposition

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- Finding large Fourier coefficients has many applications.

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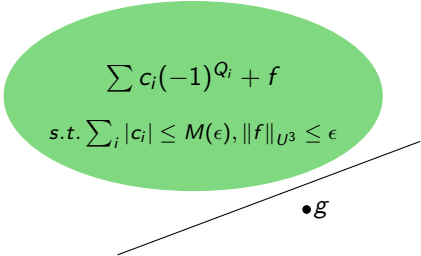
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- Use inverse theorem for Gowers norm to get a contradiction.

A quadratic Goldreich-Levin Theorem

Theorem (T, Wolf 11)

For $M(\epsilon) = \exp(1/\epsilon^C)$, can compute in time $\text{poly}(n, M(\epsilon), \log(1/\delta))$, a decomposition

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such that

- with probability $1 - \delta$, $\|f\|_{U^3} \leq \epsilon$ and $\|e\|_1 \leq \epsilon$.
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Theorem (BRTW 12)

For $M(\epsilon) = O(\exp(\log^4(1/\epsilon)))$, can compute in time $\text{poly}(n, M(\epsilon), \log(1/\delta))$, a decomposition

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The algorithmic problem

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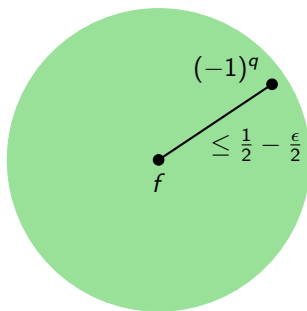
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Truth-tables of **functions $(-1)^Q$ form the Reed-Muller code of order 2.**
Want a codeword inside a ball of distance $1/2 - \epsilon/2$ around f (if one exists).



Q2: Decoding beyond the list-decoding radius

Finding codewords at large distances

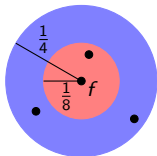
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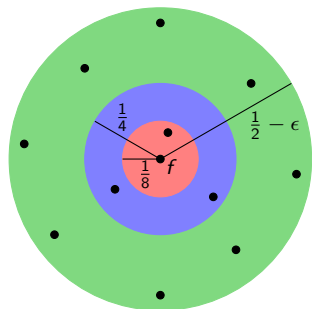


Finding codewords at large distances

- List decoding radius is $\frac{1}{4}$.
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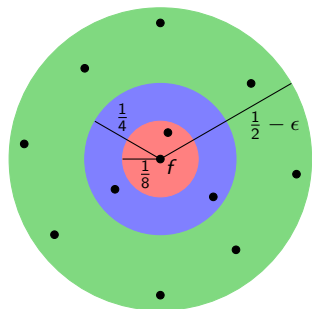


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- But we only need to find **one codeword!** In time $poly(n)$ (polylogarithmic in code length).

Finding codewords at large distances

- Given (the coefficients of) a degree- d polynomial $P : \mathbb{F}_p^n \rightarrow \mathbb{F}_p$, the Reed-Muller encoding of P is of length p^n and is given by the table of values $\{P(x)\}_{x \in \mathbb{F}_p^n}$.

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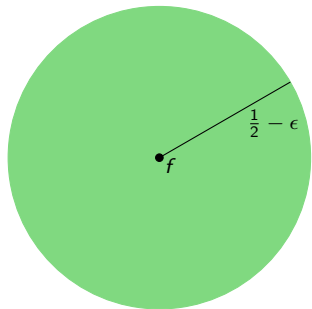
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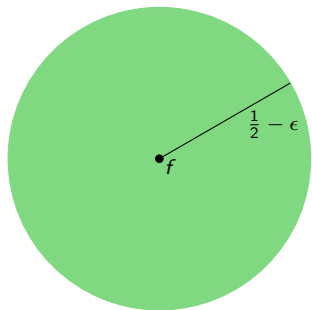
- If there exists a Reed-Muller codeword within a ball of radius $1 - \frac{1}{p} - \epsilon$, find one within a ball of radius $1 - \frac{1}{p} - \eta$.

Finding a single codeword: the quadratic case

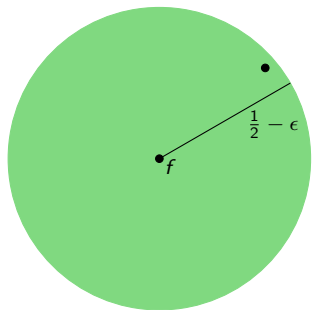


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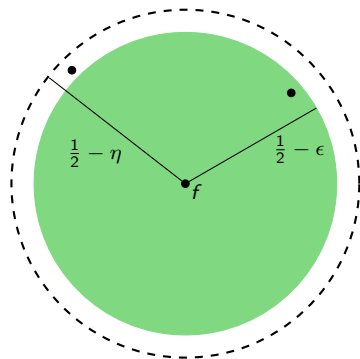
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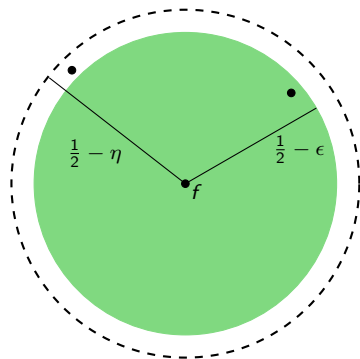
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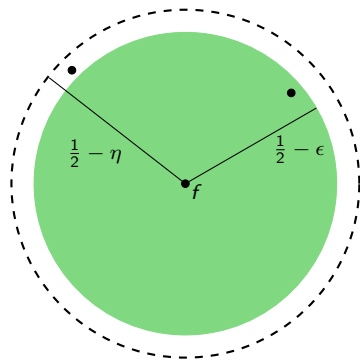
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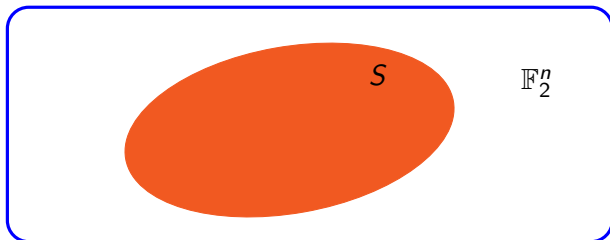
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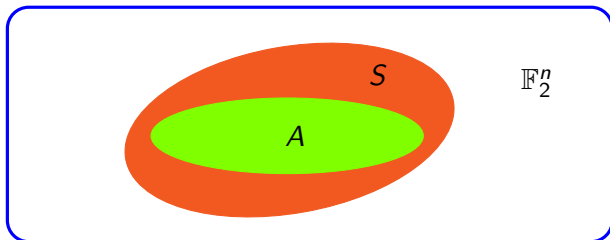
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Algorithmic versions of combinatorial theorems



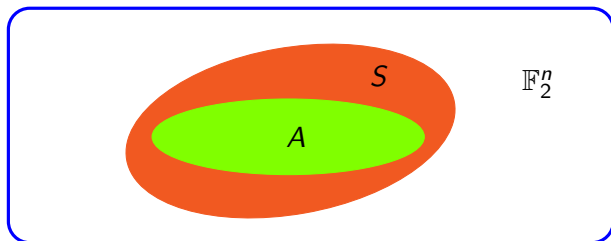
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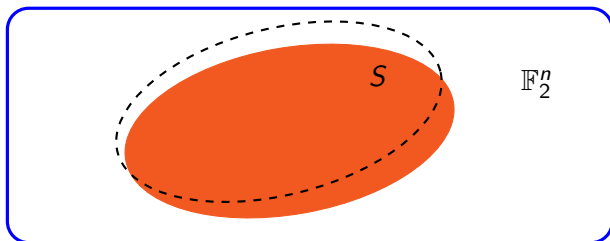
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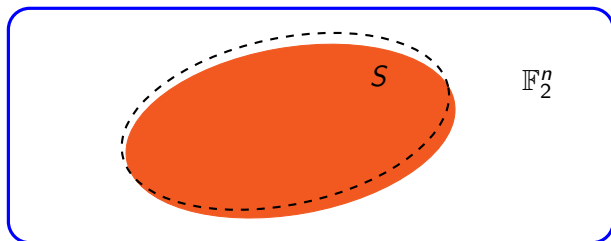
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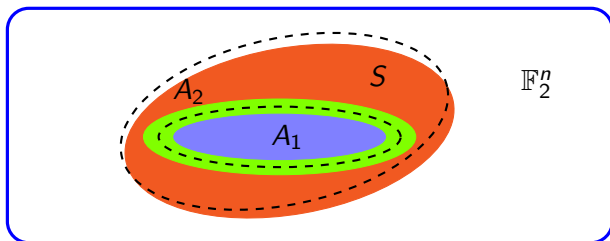
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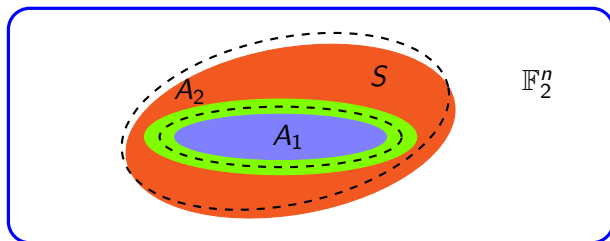
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- Prove “robust” versions of theorems from additive combinatorics.

Finding subspace structure

Most combinatorial results used here find and refine subspace structure in $S \subseteq \mathbb{F}_2^n$.

- [BSG]: If $\mathbb{P}_{x,y \in S} [x + y \in S] \geq \epsilon$ then $\exists A \subseteq S$ s.t.

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- [CS 09]: If $|A + A| \leq K \cdot |A|$, then $\mathbf{1}_A * \mathbf{1}_A$ has a large set of “almost periods” i.e., there is a large set $X \subseteq \mathbb{F}_2^n$ s.t

$$\mathbf{1}_A * \mathbf{1}_A(\cdot) \approx \mathbf{1}_A * \mathbf{1}_A(\cdot + x) \quad \forall x \in X$$

$\mathbf{1}_A * \mathbf{1}_A(\cdot) \approx$ distribution of sum of two random elements from A .

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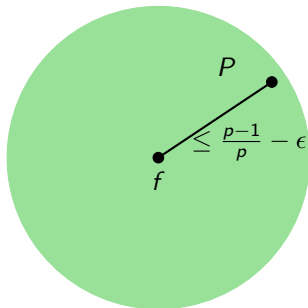
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- **Question:** Can sampling based proofs be used to find better subspace structure?

Decompositions for higher-degrees

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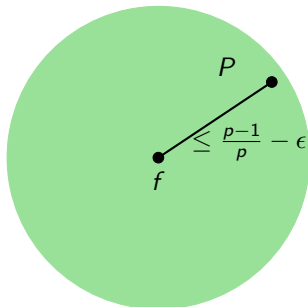
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- Can be solved for the special case when $F \in \mathcal{P}_k$ and $p > k$, inverse theorem by [GT 09].

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- Proof by [GT 09] and many other applications require the factor $\mathcal{B} = \{P_1, \dots, P_m\}$ to satisfy certain “regularity” properties. Obtaining regularity is the main challenge in converting their proof to an algorithm.

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- Like Szemerédi's regularity lemma, proofs find a certificate of non-regularity and make progress by local modification.

Q3: Algorithmic Regularity Lemmas

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- Show these notions provide required equidistribution for various known applications.

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- Regularity lemmas give terrible quantitative bounds. Is there a way to use weaker regularity properties and obtain better bounds?

Thank You

Questions?