Large Salem Sets Avoiding Nonlinear Configurations 2

Jacob Denson*

October 10, 2022

Our goal is to improve the result of the last paper. Let us reintroduce notation:

- We consider a family of axis-aligned cubes $Q_1, \ldots, Q_n \subset [0,1]^d$, each with common sidelength s > 0, such that $d(Q_i, Q_j) \ge 10s$ for $i \ne j$.
- We consider a family of density functions $\psi_1, \ldots, \psi_n \in C^{\infty}(\mathbb{T}^d)$ supported on $2Q_i$. We fix a large integer M > 0, and consider a family of independent random variables

$$\{X_i(k): 1 \leqslant i \leqslant n, 1 \leqslant k \leqslant M\}$$

where $X_i(k)$ is chosen with respect to the probability density function ψ_i .

• Let $r = M^{-1/\lambda}$, and consider the random set I of all indices $k_n \in \{1, ..., M\}$ such that there exists indices $k_1, ..., k_{n-1} \in \{1, ..., M\}$ such that

$$|f(X_1(k_1),\ldots,X_n(k_n))| \leqslant r.$$

There are two situations of interest here:

- (Scenario A): $f:(\mathbb{R}^d)^n \to \mathbb{R}^k$ is smooth and has full rank everywhere, so that the zero set is a dn-k dimensional manifold.
- (Scenario B): $f:(\mathbb{R}^d)^n \to \mathbb{R}^d$ is of the form $f(x_1,\ldots,x_n)=x_n-g(x_1,\ldots,x_{n-1})$.
- Finally, for $\xi \in \mathbb{Z}^d$

$$H(\xi) = \frac{1}{M} \sum_{k \in I} e^{2\pi i \xi \cdot X_n(k)}.$$

Our goal is to show that with high probability (so that we can take a union bound over all $|\xi| \leq M^{1/\lambda}$) we have $|H(\xi)| \leq M^{-1/2} + |\xi|^{-\lambda/2}$. In Scenario B, we already know that for the optimal range of λ , we have

$$|H(\xi) - \mathbf{E}[H(\xi)]| \lesssim M^{-1/2}$$

with high probability, so we just need to bound $\mathbb{E}[H(\xi)]$, which is a non-probabilistic quantity. In Scenario A, the novel situation occurs when $\lambda > k/(n-1/2)$, and in Scenario B, the novel situation occurs when $\lambda > d/(n-3/4)$ and n > 2.

In these notes I'll develop loose ideas I have to prove the results above.

^{*}University of Madison Wisconsin, Madison, WI, jcdenson@wisc.edu

1 Review of Concentration of Measure

 \mathbf{S}

2 Review of Expectation Bounds

We start by writing

$$\mathbb{E}[H(\xi)] = \int \psi_n(x_n) p_M(x_n) e^{2\pi i \xi \cdot x_n} dx_n,$$

where $p_M(x_n)$ denotes the probability that there exists $k_1, \ldots, k_{n-1} \in \{1, \ldots, M\}$ such that

$$|x_n - f(X_1(k_1), \dots, X_{n-1}(k_{n-1}))| \le r.$$

In that paper, it was shown using an inclusion exclusion argument that if $E_{x_n} = f^{-1}(x_n)$, then

$$P_M(x_n) = M^{n-1} \int_{E_{x_n}} \psi_1(x_1) \cdots \psi_{n-1}(x_{n-1}) \ dx_1 \dots \ dx_{n-1} + O(M^{2(n-1)-2d/\lambda}).$$

The error here is $O(M^{-1/2})$ provided that $\lambda \leq d/(n-3/4)$. Thus in this situation if we write $\psi = \psi_1 \otimes \cdots \otimes \psi_n$ then

$$\mathbb{E}[H(\xi)] = \left(\int \int_{E_{x_n}} \psi(x) e^{2\pi i \xi \cdot x_n} \right) + O(M^{-1/2}).$$

The integral here can be converted using the coarea formula into an oscillatory integral, which yields the $\delta |\xi|^{-\lambda/2}$ term required.

3 Does Smoothness Help?

The function ψ_n is smooth. Thus if we could show that

$$||p_M||_{W^{s/2,1}(\mathbb{T}^d)} \lesssim M^{-1/2}$$

then the fact that $\mathbb{E}[H(\xi)]$ is M times the Fourier transform of $\psi_n \cdot P_M$ would give the required result. But heuristically, it doesn't seem like the function should be that smooth, though I should redo the calculation just to be sure.

4 Counting Solutions

The reason the argument in the first section worked was that with high probability, the number of solutions to

$$|x_n - f(X_1(k_1), \dots, X_{n-1}(k_{n-1}))| \le r$$

was equal to M^{n-1} times the surface area of the hyperplane E_{x_n} . Could we do something more robust to get around the inclusion-exclusion argument?

One potential idea, instead of drawing X_1, \ldots, X_{n-1} from a continuous distribution, is to draw the points from a discrete distribution, e.g. uniformly distributed on some rational points with a fixed denominator in the cube Q_i ? If f is an integer valued polynomial, then perhaps the circle method might then be of some use since then we might get much better bounds on the number of solutions to the equation? We could also use probabilistic decoupling to remove the cubes $\{Q_i\}$ from the equation if needed, provided the concentration argument goes through.

5 Can One Connect To The Study of Random Matrices

If we form a $n \times N$ matrix A consisting of the values $\{X_i(k)\}$, is there a way to connect our problem to the study of this random matrix?

We can form the tensor $T = (\sum_k X_1(k)) \otimes \cdots \otimes (\sum_k X_n(k))$. But I'm not sure this is too helpful, since our problem does not really work with linear phenomena that much.

6 Gaussian Chaos

Gaussian Chaos (at least the finite dimensional theory?) studies quantities of the form $\sum a_{ij}X_iX_j$, where $X = (X_1, \ldots, X_n)$ is a random vector with independent entries. Techniques from the study of these quantities might lead to results for the functions I'm studying, since this problem also involves a 'tensor product' of entries.

7 L^2 Trick

We note that the function H is the Fourier transform of the measure

$$\sum_{k\in I} \delta_{X_n(k)},$$

but the goal, of course, is to thicken this measure on an r neighborhood, i.e. considering the probability density

$$\psi(x) = \#(I)^{-1} \sum_{k \in I} \phi\left(\frac{x - X_n(k)}{r}\right).$$

This is a sum of functions with disjoint support, and $\#(I) \approx M$ with high probability for the λ we consider, and so the L^2 norm of ψ is equal to

$$M^{-1} \left(M \cdot O(r^d) \right)^{1/2} \lesssim M^{-1/2} \cdot r^{d/2} = M^{-1/2} M^{-d/2\lambda} = M^{-(1/2)(1+d/\lambda)}.$$

Because ψ is supported on \mathbb{T}^d , we can take Fourier series, and we conclude that the L^2 norm of the Fourier series $\widehat{\psi}$ of ψ is also equal to $M^{-(1/2)(1+d/\lambda)}$,

$$\sum_{\xi \in \mathbb{Z}^d} |\widehat{\psi}(\xi)|^2 \lesssim M^{-(1+d/\lambda)}$$

Thus we get that

$$\sum_{\xi \in \mathbb{Z}^d} \left| H(\xi) \widehat{\phi} \left(rx \right) \right|^2 \lesssim M^{d/\lambda - 1}$$

In particular, we get that

$$\sum_{|\xi| \lesssim M^{1/\lambda}} |H(\xi)|^2 \lesssim M^{1+d/\lambda}$$

Thus $|H(\xi)|^2 \lesssim M^{1/2}$ for $most\ \xi$, as we might expect from e.g. energy integral type relations between Hausdorff dimension and Fourier analysis. In particular, we get that $|H(\xi)| \lesssim (\log M)^{1/2} \cdot M^{1/2}$ for all but $M^{d/\lambda}/\log M$ indices. Maybe this will help with applying a union bound. In particular, we just need to show that for each $\xi \in \mathbb{Z}^d$

$$\mathbb{P}\left(|H(\xi)| \geqslant C \log M \cdot M^{-1/2}\right) \lessapprox M^{-d/\lambda}.$$

But this still seems to require a subexponential / subgaussian bound on $H(\xi)$, i.e. that $||H(\xi)||_{\psi_2} \lesssim M^{-1/2}$.

8 Invariance

Our distributions are a little weird. Can we replace them with something nicer (e.g. a normal distribution), using for instance, Lindeberg's replacement method (1922), or Stein's method (1970), or Tikhomirov's method (1980).

9 Decoupling Methods

Decoupling yields bounds of the form

$$||H(\xi)||_{L^p_{\xi}} = \left\| \sum_{k \in I} e^{2\pi i \xi \cdot x_k} \right\|_{L^p_{\xi}} \lesssim \#(I)^{1/2}.$$

If we could set $p = \infty$, then we would get precisely the bounds we need. The problem is that this bound is not really a bound that standard decoupling techniques obtain when $p = \infty$. Can we still use decoupling type techniques to get something useful here?

References