

Large Salem Sets Avoiding Configurations

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1 Introduction

Geometric measure theory explores the relationship between the geometry of subsets of Euclidean spaces, and regularity properties of the family of Borel measures supported on those subsets. From the perspective of harmonic analysis, it is interesting to explore what structural information can be gathered from the Fourier analytic properties of measures supported on a particular subset of Euclidean space. In this paper, we focus on the relationship between the Fourier analytic properties of a set and the existence of patterns on the set. In particular, given a ‘rough pattern’, in the sense of [3], we construct a family of sets which generically avoids this pattern, and which supports measures with fast Fourier decay.

A useful statistic associated with any Borel set $E \subset \mathbf{R}^d$ is its *Fourier dimension*; given a finite Borel measure μ on \mathbf{R}^d , its Fourier dimension, $\dim_{\mathbf{F}}(\mu)$, is the supremum of all $s \in [0, d]$ such that

$$\sup \{ |\widehat{\mu}(\xi)| |\xi|^{s/2} : \xi \in \mathbf{R}^d \} < \infty. \quad (1.1)$$

The Fourier dimension of a Borel set $E \subset \mathbf{R}^d$, denoted $\dim_{\mathbf{F}}(E)$, is then the supremum of $\dim_{\mathbf{F}}(\mu)$, over all Borel probability measures μ supported on E . A particularly tractable family of sets in this scheme are *Salem sets*, those sets whose Fourier dimension agrees with their Hausdorff dimension. Most classical fractal sets are not Salem, often having Fourier dimension zero. Nonetheless, the sets we construct in this paper are Salem.

Theorem 1.1. *Let $0 \leq \alpha < dn$, and let $Z \subset \mathbf{R}^{dn}$ be a countable union of compact sets, each with lower Minkowski dimension at most α . Then there exists a compact Salem set $E \subset [0, 1]^d$ with dimension*

$$\beta = \min \left(\frac{nd - \alpha}{n - 1/2}, d \right)$$

such that for any distinct points $x_1, \dots, x_n \in E$, $(x_1, \dots, x_n) \notin Z$.

Remark 1.2. *Theorem 1.1 is an attempt to strengthen the main result of [3] to give a Fourier dimension bound, albeit under a weaker dimension bound. Unlike in [3], the case of Theorem 1.1 when $0 \leq \alpha < d$ is still interesting, since the trivial construction $[0, \pi]^d - \pi(Z)$ is not necessarily a Salem set, where $\pi(x_1, \dots, x_n) = x_1$ is projection onto the first coordinate. For instance, in Example 8 of [6] it is shown that there exists a compact set $E \subset [0, 1]$ such that $\dim_{\mathbf{M}}(E) < 1$ and $\dim_{\mathbf{F}}([0, 1] - E) < 1$. Setting $Z = E \times \{0\} \cup \{0\} \times E$ shows that neither*

subtracting projections onto the first nor the second coordinate gives the required Fourier dimension bounds.

Because we are working with *compact* sets avoiding patterns, working in the domain \mathbf{R}^d is not significantly different from working in a periodic domain $\mathbf{T}^d = \mathbf{R}^d / \mathbf{Z}^d$, and working in this space has several advantages over the Euclidean case. For a finite measure μ on \mathbf{T}^d , we can define its Fourier dimension $\dim_{\mathbf{F}}(\mu)$ as the supremum of all $0 \leq s \leq d$ such that

$$\sup_{\xi \in \mathbf{Z}^d} |\widehat{\mu}(\xi)| |\xi|^{s/2} < \infty. \quad (1.2)$$

We can then define the Fourier dimension $\dim_{\mathbf{F}}(E)$ of any Borel set $E \subset \mathbf{T}^d$ as the supremum of $\dim_{\mathbf{F}}(\mu)$, over all Borel measures μ supported on E . Since \mathbf{T}^d has a natural metric space structure, we can define the Hausdorff dimension of sets on \mathbf{T}^d . It is a simple consequence of the Poisson summation formula that if μ is a compactly supported measure on \mathbf{R}^d , then (1.1) is equivalent to the more discrete condition

$$\sup_{\xi \in \mathbf{Z}^d} |\widehat{\mu}(\xi)| |\xi|^{s/2} < \infty. \quad (1.3)$$

A proof is given in Lemma 39 of [4]. In particular, if μ^* is the *periodization* of μ , i.e. the measure on \mathbf{T}^d such that for any $f \in C(\mathbf{T}^d)$,

$$\int_{\mathbf{T}^d} f(x) d\mu^*(x) = \int_{\mathbf{R}^d} f(x) d\mu(x), \quad (1.4)$$

then (1.3), together with the Poisson summation formula, implies $\dim_{\mathbf{F}}(\mu^*) = \dim_{\mathbf{F}}(\mu)$. Since μ is compactly supported, it is also simple to see that $\dim_{\mathbf{H}}(\mu^*) = \dim_{\mathbf{H}}(\mu)$. Thus Theorem 1.1 is clearly equivalent to Theorem 1.3, which is its periodic variant.

Theorem 1.3. *Let $0 \leq \alpha < dn$, and let $Z \subset \mathbf{T}^{dn}$ be a countable union of compact sets, each with lower Minkowski dimension at most α . Then there exists a compact Salem set $X \subset \mathbf{T}^d$ with dimension*

$$\beta = \min \left(\frac{dn - \alpha}{n - 1/2}, d \right)$$

such that for any distinct points $x_1, \dots, x_n \in X$, $(x_1, \dots, x_n) \notin Z$.

We construct the set X in Theorem 1.3 by relying on a Baire-category type approach. Thus we consider a complete metric space \mathcal{X}_β , whose elements consist of pairs (E, μ) , where E is a subset of \mathbf{T}^d , and μ is a probability measure supported on E . We then show that for *quasi-all* elements $(E, \mu) \in \mathcal{X}_\beta$, E is a Salem set of dimension β , and for distinct $x_1, \dots, x_n \in E$, $(x_1, \dots, x_n) \notin Z$, in the sense that the set of pairs (E, μ) which do not satisfy these properties is a set of first category in \mathcal{X}_β . It thus follows that the consequences of Theorem 1.3 holds in a ‘generic’ sense for elements of \mathcal{X}_β .

Once we have setup the appropriate metric space \mathcal{X}_β , our approach is quite similar to the construction in [3], relying on a random selection procedure, which is now exploited to give high probability bounds on the Fourier transform of the measures we study. The use of the Baire category approach in this paper, rather than an algorithmic, ‘nested set’ approach as

used in [3], is mostly of an aesthetic nature, avoiding the complex queuing method and dyadic decomposition strategy required in the nested set approach; our approach can, with some care, be converted into a queuing procedure like in [3]. But the Baire category argument makes our proof much simpler to read, and has the advantage that it indicates that Salem sets of a specified dimension ‘generically’ avoid a given rough pattern.

2 Notation

- Given a metric space X , $x \in X$, and $\varepsilon > 0$, we shall let $B_\varepsilon(x)$ denote the open ball of radius ε around x . For a given set $E \subset \Omega$ and $\varepsilon > 0$, we let

$$E_\varepsilon = \bigcup_{x \in E} B_\varepsilon(x),$$

denote the ε -thickening of the set E . A subset of X is of *first category* in X if it is the countable union of closed sets with empty interior. We say a property holds *quasi-always*, or a property is *generic* in X if the set of points in X failing to satisfy that property form a set of first category.

- We let $\mathbf{T}^d = \mathbf{R}^d / \mathbf{Z}^d$. Given $x \in \mathbf{T}$, we let

$$|x| = \min\{|x + n| : n \in \mathbf{Z}\},$$

and for $x \in \mathbf{T}^d$, we let

$$|x| = \sqrt{|x_1|^2 + \cdots + |x_d|^2}.$$

The canonical metric on \mathbf{T}^d is then given by $d(x, y) = |x - y|$, for $x, y \in \mathbf{T}^d$. For an axis-oriented cube Q in \mathbf{T}^d , we let $2Q$ be the cube in \mathbf{T}^d with the same centre and twice the sidelength.

- Suppose $\mathbf{E} = \mathbf{T}^d$ or $\mathbf{E} = \mathbf{R}^d$. For $\alpha \in [0, d]$ and $\delta > 0$, we define the Hausdorff content of a Borel set $E \subset \mathbf{E}$ as

$$H_\delta^\alpha(E) = \inf \left\{ \sum_{i=1}^{\infty} \varepsilon_i^\alpha : E \subset \bigcup_{i=1}^{\infty} B_{\varepsilon_i}(x_i) \text{ and } 0 < \varepsilon_i \leq \delta \text{ for all } i \geq 1 \right\}.$$

The α dimensional Hausdorff measure of E is equal to

$$H^\alpha(E) = \lim_{\delta \rightarrow 0} H_\delta^\alpha(E).$$

The Hausdorff dimension $\dim_{\mathbf{H}}(E)$ of a Borel set E is then the infimum over all $s \in [0, d]$ such that $H^s(E) = \infty$, or alternatively, the supremum over all $s \in [0, d]$ such that $H^s(E) = 0$.

For a measurable set $E \subset \mathbf{E}$, we let $|E|$ denote its Lebesgue measure. We define the lower Minkowski dimension of a compact Borel set $E \subset \mathbf{E}$ as

$$\underline{\dim}_{\mathbf{M}}(E) = \liminf_{r \rightarrow 0} d - \log_r |E_r|.$$

Thus $\underline{\dim}_{\mathbf{M}}(E)$ is the largest number such that for $\alpha < \underline{\dim}_{\mathbf{M}}(E)$, there exists a decreasing sequence $\{r_i\}$ with $\lim_{i \rightarrow \infty} r_i = 0$ and $|E_{r_i}| \leq r_i^{d-\alpha}$ for each i .

- In this paper we will need to employ probabilistic concentration bounds several times. In particular, we use *McDiarmid's inequality*, trivially modified from the standard theorem to work with complex-valued functions. Let $\mathbf{E} = \mathbf{T}^d$ or $\mathbf{E} = \mathbf{R}^d$. Let $\{X_1, \dots, X_N\}$ be an independant family of random variables, and consider a function $f : \mathbf{E}^N \rightarrow \mathbf{C}$. Suppose that for each $i \in \{1, \dots, N\}$, there exists a constant $A_i > 0$ such that for any $x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_N \in \mathbf{E}$, and for each $x_i, x'_i \in \mathbf{E}$,

$$|f(x_1, \dots, x_i, \dots, x_N) - f(x_1, \dots, x'_i, \dots, x_N)| \leq A_i.$$

Then McDiarmid's inequality guarantees that for all $t \geq 0$,

$$\mathbf{P}(|f(X_1, \dots, X_N) - \mathbf{E}(f(X_1, \dots, X_N))| \geq t) \leq 4 \exp\left(\frac{-2t^2}{A_1^2 + \dots + A_N^2}\right).$$

The complex-valued extension we have just stated is proved easily from the real-valued case by taking a union bound to the inequality for the real and imaginary values of f . Proofs of McDiarmid's inequality are given in many probability texts, for instance, in Theorem 3.11 of [5].

A special case of McDiarmid's inequality is *Hoeffding's Inequality*. The version of Hoeffding's inequality we use states that if $\{X_1, \dots, X_N\}$ is a family of independant random variables, such that for each i , there exists a constant $A_i \geq 0$ such that $|X_i| \leq A_i$ almost surely, then for each $t \geq 0$,

$$\mathbf{P}(|X_1 + \dots + X_N - \mathbf{E}(X_1 + \dots + X_N)| \geq t) \leq 4 \exp\left(\frac{-t^2/2}{A_1^2 + \dots + A_N^2}\right).$$

- Throughout this paper, we will need to consider a standard mollifier. So we fix a smooth, non-negative function $\phi \in C^\infty(\mathbf{T}^d)$ such that $\phi(x) = 0$ for $|x| \geq 2/5$ and

$$\int_{\mathbf{T}^d} \phi(x) dx = 1.$$

For each $r \in (0, 1)$, we can then define $\phi_r \in C^\infty(\mathbf{T}^d)$ by writing

$$\phi_r(x) = \begin{cases} r^{-d} \phi(x/r) & : |x| < r, \\ 0 & : \text{otherwise.} \end{cases}$$

The following standard properties hold:

- (1) For each $r \in (0, 1)$, ϕ_r is a non-negative smooth function with

$$\int_{\mathbf{T}^d} \phi_r(x) dx = 1, \tag{2.1}$$

and $\phi_r(x) = 0$ for $|x| \geq r$.

- (2) For any $r \in (0, 1)$,

$$\|\hat{\phi}_r\|_{L^\infty(\mathbf{Z}^d)} \leq 1. \tag{2.2}$$

(3) For each $\xi \in \mathbf{Z}^d$,

$$\lim_{r \rightarrow 0} \widehat{\phi}_r(\xi) = 1. \quad (2.3)$$

(4) For each $T > 0$, and for all $r > 0$ and any non-zero $\xi \in \mathbf{Z}^d$,

$$|\widehat{\phi}_r(\xi)| \lesssim_T r^{-T} |\xi|^{-T}. \quad (2.4)$$

3 A Metric Space Controlling Fourier Dimension

In order to work with a Baire category type argument, we must construct an appropriate metric space appropriate for our task. In later sections we will fix a specific choice of β , for instance, as in Theorem 1.1. But in this section we let β be an arbitrary element of $(0, d]$. Our approach in this section is heavily influenced by [2]. However, we employ a Frechét space construction instead of the Banach space used in [2], which enables us to use softer estimates in our arguments:

- We let \mathcal{E} denote the family of all compact subsets of \mathbf{T}^d . If, for two compact sets $E, F \in \mathcal{E}$, we consider their Hausdorff distance

$$d_{\mathbf{H}}(E, F) = \inf\{\varepsilon > 0 : E \subset F_\varepsilon \text{ and } F \subset E_\varepsilon\},$$

then $(\mathcal{E}, d_{\mathbf{H}})$ forms a complete metric space.

- We let $M(\beta/2)$ consist of the class of all finite Borel measures μ on \mathbf{T}^d such that for each $\varepsilon \in (0, \beta/2]$, the quantity

$$\|\mu\|_{M(\beta/2-\varepsilon)} = \sup_{\xi \in \mathbf{Z}^d} |\widehat{\mu}(\xi)| |\xi|^{\beta/2-\varepsilon}$$

is finite. Then $\|\cdot\|_{M(\beta/2-\varepsilon)}$ is a seminorm on $M(\beta/2)$, and the collection of all such seminorms for $\varepsilon \in (0, \beta/2]$ gives $M(\beta/2)$ the structure of a Frechét space. Under this topology, a sequence of probability measures $\{\mu_k\}$ converges to a probability measure μ in $M(\beta/2)$ if and only if for any $\varepsilon > 0$, $\lim_{k \rightarrow \infty} \|\mu_k - \mu\|_{M(\beta/2-\varepsilon)} = 0$.

We now let \mathcal{X}_β be the collection of all pairs $(E, \mu) \in \mathcal{E} \times M(\beta/2)$, where μ is a probability measure such that $\text{supp}(\mu) \subset E$. Then \mathcal{X}_β is a closed subset of $\mathcal{E} \times M(\beta/2)$ under the product metric, and thus a complete metrizable space. We remark that for any $\varepsilon > 0$ and $(E, \mu) \in \mathcal{X}_\beta$,

$$\lim_{|\xi| \rightarrow \infty} |\xi|^{\beta/2-\varepsilon} |\widehat{\mu}(\xi)| = 0, \quad (3.1)$$

which follows because $\|\mu\|_{M(\beta/2-\varepsilon/2)}$ is finite. Thus $\dim_{\mathbf{F}}(\mu) \geq \beta$ for each $(E, \mu) \in \mathcal{X}_\beta$.

Lemma 3.1. *The set of all (E, μ) with $\mu \in C^\infty(\mathbf{T}^d)$ is dense in \mathcal{X}_β .*

Proof. Consider $(E, \mu) \in \mathcal{X}_\beta$. For each $r \in (0, 1)$, consider the convolved measure $\mu_r = \mu * \phi_r$. Then $\mu_r \in C^\infty(\mathbf{T}^d)$. We claim that

$$\lim_{r \rightarrow 0} (E_r, \mu_r) = (E, \mu). \quad (3.2)$$

Since $\text{supp}(\mu_r) \subset E_r$, we conclude that

$$d_{\mathbf{H}}(E, E_r) \leq r. \quad (3.3)$$

Thus $E_r \rightarrow E$ as $r \rightarrow 0$ under the Hausdorff distance. Now fix $\varepsilon_1 \in (0, \beta/2]$ and $\varepsilon > 0$. For each $\xi \in \mathbf{Z}^d$, $|\hat{\mu}_r(\xi)| = |\hat{\phi}_r(\xi)| |\hat{\mu}(\xi)|$, so

$$|\xi|^{\beta/2-\varepsilon_1} |\mu_r(\xi) - \mu(\xi)| = |\xi|^{\beta/2-\varepsilon_1} |\hat{\phi}_r(\xi) - 1| |\hat{\mu}(\xi)|. \quad (3.4)$$

Since $(E, \mu) \in \mathcal{X}_\beta$, we can apply (3.1) to conclude that there exists R such that for $|\xi| \geq R$,

$$|\xi|^{\beta/2-\varepsilon_1} |\hat{\mu}(\xi)| \leq \varepsilon. \quad (3.5)$$

Combining (3.4), (3.5), and (2.2), we conclude that for $|\xi| \geq R$,

$$|\xi|^{\beta/2-\varepsilon_1} |\mu_r(\xi) - \mu(\xi)| \leq 2\varepsilon. \quad (3.6)$$

By (2.3), we conclude that there exists $r_0 > 0$ such that for $r \leq r_0$ and $|\xi| \leq R$,

$$|\xi|^{\beta/2-\varepsilon_1} |\hat{\phi}_r(\xi) - 1| \leq \varepsilon. \quad (3.7)$$

The (L^1, L^∞) bound for the Fourier transform implies that

$$|\hat{\mu}(\xi)| \leq \mu(\mathbf{T}^d) = 1 \quad (3.8)$$

But from (3.7) and (3.8) applied to (3.4), we find that for $r \leq r_0$ and $|\xi| \leq R$,

$$|\xi|^{\beta/2-\varepsilon_1} |\mu_r(\xi) - \mu(\xi)| \leq \varepsilon. \quad (3.9)$$

Putting together (3.6) and (3.9), we find that for $r \leq r_0$,

$$\|\mu_r - \mu\|_{M(\beta/2-\varepsilon_1)} \leq 2\varepsilon. \quad (3.10)$$

Since ε and ε_1 were arbitrary, we conclude from (3.10) and (3.3) that $(E_r, \mu_r) \rightarrow (E, \mu)$, completing the proof. \square

Remark 3.2. *Let*

$$\tilde{\mathcal{X}}_\beta = \{(E, \mu) \in \mathcal{X}_\beta : \text{supp}(\mu) = E\}.$$

Suppose $(E_0, \mu_0) \in \tilde{\mathcal{X}}_\beta$. Then, in the proof above, one may let E_r be equal to $\text{supp}(\mu_r)$, since it follows from this that $d_{\mathbf{H}}(E_0, E_r) \leq r$. This means that the set of pairs $(E, \mu) \in \tilde{\mathcal{X}}_\beta$ with $\mu \in C^\infty(\mathbf{T}^d)$ are dense in $\tilde{\mathcal{X}}_\beta$.

The reason we don't work with the smaller space $\tilde{\mathcal{X}}_\beta \subset \mathcal{X}_\beta$ in this paper is that $\tilde{\mathcal{X}}_\beta$ is not a closed subset of \mathcal{X}_β , and so is not a complete metrizable space under the product topology. However, as a consolation, quasi-all elements of \mathcal{X}_β belong to $\tilde{\mathcal{X}}_\beta$, so that one can think of \mathcal{X}_β and $\tilde{\mathcal{X}}_\beta$ as being equal 'generically'.

Lemma 3.3. *For quasi-all $(E, \mu) \in \mathcal{X}_\beta$, $\text{supp}(\mu) = E$.*

Proof. For each closed cube $Q \subset \mathbf{T}^d$, let

$$A(Q) = \{(E, \mu) \in \mathbf{T}^d : (E \cap Q) = \emptyset \text{ or } \mu(Q) > 0\}.$$

Then $A(Q)$ is an open set. If $\{Q_k\}$ is a sequence enumerating all cubes with rational corners in \mathbf{T}^d , then

$$\bigcap_{k=1}^{\infty} A(Q_k)$$

is the collection of $(E, \mu) \in \mathcal{X}_\beta$ with $\text{supp}(\mu) = E$. Thus it suffices to show that $A(Q)$ is dense in \mathcal{X}_β for each closed cube Q .

Consider $(E_0, \mu_0) \in \mathcal{X}_\beta - A(Q)$, $\varepsilon_1 \in (0, \beta/2]$, and $\varepsilon > 0$. Our goal is to find $(E, \mu) \in A(Q)$ with $d_{\mathbf{H}}(E, E_0) \leq \varepsilon$ and $\|\mu_0 - \mu\|_{M(\beta/2-\varepsilon_1)} \leq \varepsilon$. Without loss of generality by Lemma 3.1 we may assume that $\mu_0 \in C^\infty(\mathbf{T}^d)$. Because $(E_0, \mu_0) \in \mathcal{X}_\beta - A(Q)$, we know $E \cap Q \neq \emptyset$ and $\mu(Q) = 0$. Find a smooth probability measure ν supported on $E_\varepsilon \cap Q$ and, for $t \in (0, 1)$, define $\mu_t = (1-t)\mu_0 + t\nu$. Then $\text{supp}(\mu_t) \subset E_\varepsilon$, so if we let $E = \text{supp}(\nu) \cup \text{supp}(\mu)$, then $d_{\mathbf{H}}(E, E_0) \leq \varepsilon$. Clearly $(E, \mu_t) \in A(Q)$ for $t > 0$. And

$$\|\mu_t - \mu_0\|_{M(\beta/2-\varepsilon)} \leq t (\|\mu_0\|_{M(\beta/2-\varepsilon)} + \|\nu\|_{M(\beta/2-\varepsilon)}),$$

so if we choose $t \leq \varepsilon(\|\mu\|_{M(\beta/2-\varepsilon)} + \|\nu\|_{M(\beta/2-\varepsilon)})^{-1}$ we find $\|\mu_t - \mu\|_{M(\beta/2-\varepsilon)} \leq \varepsilon$. Since ε was arbitrary, we conclude $A(Q)$ is dense in \mathcal{X}_β . \square

Combining Lemma 3.3 with Remark 3.2 gives the following simple corollary.

Corollary 3.4. *The family of (E, μ) with $\text{supp}(\mu) = E$ and $\mu \in C^\infty(\mathbf{T}^d)$ is dense in \mathcal{X}_β .*

Our main way of constructing approximations to $(E_0, \mu_0) \in \mathcal{X}_\beta$ is to multiply μ_0 by a smooth function f . For instance, we might choose f in such a way as to remove certain points from the support of μ_0 which contribute to the formation of a pattern we are trying to avoid. As long as the Fourier transform of f decays appropriately quickly, we find $f\mu_0 \approx \mu_0$.

Lemma 3.5. *Consider a smooth finite measure μ_0 on \mathbf{T}^d , as well as a smooth probability density function $f \in C^\infty(\mathbf{T}^d)$. If we define $\mu = f\mu_0$, then*

$$\|\mu - \mu_0\|_{M(\beta/2)} \lesssim_{d, \mu_0} \|f\|_{M(\beta/2)}.$$

Proof. Since $\hat{\mu} = \hat{f} * \hat{\mu}_0$, and $\hat{f}(0) = 1$, for each $\xi \in \mathbf{Z}^d$ we have

$$|\xi|^{\beta/2} |\hat{\mu}(\xi) - \hat{\mu}_0(\xi)| = |\xi|^{\beta/2} \left| \sum_{\eta \neq \xi} \hat{f}(\xi - \eta) \hat{\mu}_0(\eta) \right|. \quad (3.11)$$

If $|\eta| \leq |\xi|/2$, then $|\xi|/2 \leq |\xi - \eta| \leq 2|\xi|$, so

$$|\xi|^{\beta/2} |\hat{f}(\xi - \eta)| \leq \|f\|_{M(\beta/2)} |\xi|^{\beta/2} |\xi - \eta|^{-\beta} \leq 2^{\beta/2} \|f\|_{M(\beta/2)} \lesssim_d \|f\|_{M(\beta/2)}. \quad (3.12)$$

Since μ_0 is smooth, for any $T \geq 0$ and $\xi \in \mathbf{Z}^d$,

$$|\xi|^T |\hat{\mu}_0(\xi)| \lesssim_{T, \mu_0} 1. \quad (3.13)$$

Thus we can combine the bounds (3.12) and (3.13), with $T = d + 1$, to conclude that

$$|\xi|^{\beta/2} \left| \sum_{0 \leq |\eta| \leq |\xi|/2} \hat{f}(\eta) \hat{\mu}_0(\xi - \eta) \right| \lesssim_{\mu_0, d} \left(1 + \sum_{0 < |\eta| \leq |\xi|/2} \frac{1}{|\eta|^{d+1}} \right) \|f\|_{M(\beta)} \lesssim_d \|f\|_{M(\beta)}. \quad (3.14)$$

On the other hand, for all $\eta \neq \xi$,

$$|\hat{f}(\xi - \eta)| \leq \|f\|_{M(\beta/2)} |\xi - \eta|^{-\beta} \leq \|f\|_{M(\beta/2)}. \quad (3.15)$$

Thus applying (3.13) and (3.15), with $T = 3d/2$, we conclude that

$$|\xi|^{\beta/2} \left| \sum_{\substack{|\eta| > |\xi|/2 \\ \eta \neq \xi}} \hat{f}(\xi - \eta) \hat{\mu}_0(\eta) \right| \lesssim_{d, \mu_0} |\xi|^{\beta/2} \sum_{|\eta| > |\xi|/2} \frac{\|f\|_{M(\beta/2)}}{|\eta|^{3d/2}} \lesssim_d \|f\|_{M(\beta/2)}. \quad (3.16)$$

Combining (3.11), (3.14) and (3.16) completes the proof. \square

Remark 3.6. In particular, we note that this lemma implies that $1 - \mu(\mathbf{T}^d) \lesssim_{d, \mu_0} \|f\|_{M(0)}$.

The bound in Lemma 3.5, if $\|f\|_{M(\beta/2)}$ is taken appropriately small, also implies that the Hausdorff distance between the supports of μ and μ_0 is not too large.

Lemma 3.7. Fix a probability measure $\mu_0 \in C^\infty(\mathbf{T}^d)$. For any $\varepsilon > 0$, there exists $\delta > 0$ depending on μ_0 and ε , such that if $\mu \in C^\infty(\mathbf{T}^d)$, $\text{supp}(\mu) \subset \text{supp}(\mu_0)$, and $\|\mu_0 - \mu\|_{M(\beta/2)} \leq \delta$, then $d_{\mathbf{H}}(\text{supp}(\mu), \text{supp}(\mu_0)) \leq \varepsilon$.

Proof. Consider any cover of $\text{supp}(\mu_0)$ by a family of radius $\varepsilon/3$ balls $\{B_1, \dots, B_N\}$, and for each $i \in \{1, \dots, N\}$, consider a smooth function $f_i \in C_c^\infty(B_i)$ such that there is $s > 0$ with

$$\int f_i(x) d\mu_0(x) \geq s. \quad (3.17)$$

for each $i \in \{1, \dots, N\}$. Fix $A > 0$ with

$$\sum_{\xi \neq 0} |\hat{f}_i(\xi)| \leq A \quad (3.18)$$

for all $i \in \{1, \dots, N\}$ as well. Set $\delta = s/2A$. If $\|\mu_0 - \mu\|_{M(\beta/2)} \leq \delta$, we apply Plancherel's inequality together with (3.17) and (3.18) to conclude that

$$\begin{aligned} \left| \int f_i(x) d\mu(x) dx - \int f_i(x) d\mu_0(x) \right| &= \left| \sum_{\xi \in \mathbf{Z}^d} \hat{f}_i(\xi) (\hat{\mu}(\xi) - \hat{\mu}_0(\xi)) \right| \\ &\leq A \|\mu_0 - \mu\|_{M(\beta/2)} \leq s/2. \end{aligned} \quad (3.19)$$

Thus we conclude from (3.17) and (3.19) that

$$\int f_i(x) d\mu(x) dx \geq \int f_i(x) d\mu_0(x) - s/2 \geq s/2 > 0. \quad (3.20)$$

Since equation (3.20) holds for each $i \in \{1, \dots, N\}$, the support of μ intersects every ball in $\{B_1, \dots, B_N\}$. Combined with the assumption that $\text{supp}(\mu) \subset \text{supp}(\mu_0)$, this implies that $d_{\mathbf{H}}(\mu_0, \mu) \leq \varepsilon$. \square

To obtain the smooth function f in the previous lemmas in our proof, we take a measure η , which is a linear combination of Dirac deltas, and convolve it by the mollifier at the scale ϕ_r . To obtain the appropriate control on \hat{f} , it suffices to have a decay bound for $\hat{\eta}$ for $|\xi| \leq 1/r$, and a weaker bound for $|\xi|$ slightly bigger than $1/r$, which is required before we can take complete advantage of the Fourier decay of the mollifier ϕ_r for $|\xi| \geq 1/r$.

Lemma 3.8. *Fix $C > 0$ and $\varepsilon, \varepsilon_1, \varepsilon_2 > 0$, with $\varepsilon_2 \leq \beta/2$. Then there exists $r_0 > 0$ depending on all these quantities, such that if $0 < r \leq r_0$, then for any Borel probability measure η on \mathbf{T}^d satisfying*

$$|\hat{\eta}(\xi)| \leq \varepsilon \cdot |\xi|^{\varepsilon_2 - \beta/2} \quad \text{for } 0 < |\xi| \leq (1/r) \quad (3.21)$$

and

$$|\hat{\eta}(\xi)| \leq C \cdot r^{\beta/2} \log(1/r)^{1/2} \quad \text{for } (1/r) \leq |\xi| \leq (1/r)^{1+\varepsilon_1}, \quad (3.22)$$

if we define $f(x) = (\eta * \phi_r)(x)$, then $\|f\|_{M(\beta/2 - \varepsilon_2)} \leq 2\varepsilon$.

Proof. For each $\xi \in \mathbf{Z}^d$,

$$\hat{f}(\xi) = \hat{\eta}(\xi) \hat{\phi}_r(\xi). \quad (3.23)$$

For $|\xi| \leq 1/r$ we combine (3.21), (3.23) and (2.2) to conclude that

$$|\hat{f}(\xi)| \leq \varepsilon \cdot |\xi|^{\varepsilon_2 - \beta/2}. \quad (3.24)$$

If $(1/r) \leq |\xi| \leq (1/r)^{1+\varepsilon_1}$, (2.4) implies $|\hat{\phi}_r(\xi)| \lesssim_\beta r^{-\beta/2} |\xi|^{-\beta/2}$, which together with (3.22), (3.23), and (2.2), show that for $r \leq r_1$,

$$\begin{aligned} |\hat{f}(\xi)| &\lesssim_\beta (Cr^{\beta/2} \log(1/r)^{1/2}) (r^{-\beta/2} |\xi|^{-\beta/2}) \\ &\leq C \log(1/r)^{1/2} \cdot |\xi|^{-\beta/2} \\ &\leq Cr^{\varepsilon_2} \log(1/r)^{1/2} \cdot |\xi|^{\varepsilon_2 - \beta/2}. \end{aligned} \quad (3.25)$$

Since $Cr^{\varepsilon_2} \log(1/r)^{1/2} \rightarrow 0$ as $r \rightarrow 0$, so we conclude from (3.25) that there exists $r_1 > 0$ such that for $r \leq r_1$ and $(1/r) \leq |\xi| \leq (1/r)^{1+\varepsilon_1}$

$$|\hat{f}(\xi)| \leq \varepsilon \cdot |\xi|^{\varepsilon_2 - \beta/2}. \quad (3.26)$$

If $|\xi| \geq (1/r)^{1+\varepsilon_1}$, we apply (2.4) for $T \geq \beta/2$ to conclude

$$\begin{aligned} |\hat{f}(\xi)| &\lesssim_T r^{-T} |\xi|^{-T} \\ &= r^{-T} |\xi|^{\beta/2 - T} \cdot |\xi|^{-\beta/2} \\ &\leq r^{-T} (1/r)^{(\beta/2 - T)(1+\varepsilon_1)} \cdot |\xi|^{-\beta/2} \\ &= r^{\varepsilon_1 T - (\beta/2)(1+\varepsilon_1)} \cdot |\xi|^{-\beta/2}. \end{aligned} \quad (3.27)$$

If we choose $T > (\beta/2)(1 + 1/\varepsilon_1)$, then as $r \rightarrow 0$, $r^{\varepsilon_1 T - (\beta/2)(1+\varepsilon_1)} \rightarrow 0$. Thus we conclude from (3.27) that there exists $r_2 > 0$ satisfying such that for $0 < r \leq r_2$ and $|\xi| \geq (1/r)^{1+\varepsilon_1}$,

$$|\hat{f}(\xi)| \leq \varepsilon \cdot |\xi|^{-\beta/2} \leq \varepsilon \cdot |\xi|^{\varepsilon_2 - \beta/2}. \quad (3.28)$$

All that remains is to combine (3.24), (3.26), and (3.28), defining $r_0 = \min(r_1, r_2)$. \square

Corollary 3.9. Fix $C > 0$ and $\varepsilon, \varepsilon_1, \varepsilon_2 > 0$ with $\varepsilon_2 \leq \beta/2$. Then there exists $r_0 > 0$ and $\delta > 0$ depending on these quantities, such that if $0 < r \leq r_0$, then for any Borel probability measure η on \mathbf{T}^d satisfying

$$|\hat{\eta}(\xi)| \leq \delta \cdot |\xi|^{\varepsilon_2 - \beta/2} \quad \text{for } 0 < |\xi| \leq (1/r) \quad (3.29)$$

and

$$|\hat{\eta}(\xi)| \leq C \cdot r^{\beta/2} \log(1/r)^{1/2} \quad \text{for } (1/r) \leq |\xi| \leq (1/r)^{1+\varepsilon_1}, \quad (3.30)$$

if we define $f(x) = (\eta * \phi_r)(x)$, and a probability measure

$$\mu = \frac{f\mu_0}{(f\mu_0)(\mathbf{T}^d)}$$

then $\|\mu - \mu_0\|_{M(\beta/2 - \varepsilon_2)} \leq \varepsilon$.

Proof. Let

$$\delta = \min \left(\frac{\varepsilon}{2}, \frac{1}{2}, \frac{\varepsilon}{4\|\mu_0\|_{M(\beta/2 - \varepsilon_2)}} \right).$$

Thus if we apply Lemmas 3.5 and 3.8, we conclude there exists r_0 such that for $r \leq r_0$,

$$\|f\mu_0 - \mu_0\|_{M(\beta/2 - \varepsilon_2)} \leq \varepsilon/2, \quad (3.31)$$

and

$$\|f\mu_0 - \mu_0\|_{M(0)} \leq \min \left(\frac{1}{2}, \frac{\varepsilon}{4\|\mu_0\|_{M(\beta/2 - \varepsilon_2)}} \right). \quad (3.32)$$

Equation (3.32) implies that

$$1 - \min \left(\frac{1}{2}, \frac{\varepsilon}{4\|\mu_0\|_{M(\beta/2 - \varepsilon_2)}} \right) \leq (f\mu_0)(\mathbf{T}^d) \leq 1. \quad (3.33)$$

But now (3.31) and (3.33) show that

$$\begin{aligned} \|\mu - \mu_0\|_{M(\beta/2 - \varepsilon_2)} &\leq \|f\mu_0 - \mu_0\|_{M(\beta/2 - \varepsilon)} + \|\mu - f\mu_0\|_{M(\beta/2 - \varepsilon)} \\ &\leq (\varepsilon/2) + \left(1 - \frac{1}{(f\mu_0)(\mathbf{T}^d)} \right) \|\mu_0\|_{M(\beta/2 - \varepsilon)} \\ &\leq (\varepsilon/2) + (\varepsilon/2) \leq \varepsilon. \end{aligned} \quad \square$$

The average function has small Fourier coefficients on average. The next lemma gives a particular instance of this result.

Lemma 3.10. Fix a large integer K . Let X_1, \dots, X_K be independant random variables on \mathbf{T}^d , such that for each nonzero $\xi \in \mathbf{Z}^d$,

$$\sum_{k=1}^K \mathbf{E} \left(e^{2\pi i \xi \cdot X_k} \right) = 0. \quad (3.34)$$

Set

$$\eta(x) = \frac{1}{K} \sum_{k=1}^K \delta(x - x_k)$$

and

$$B = \{\xi \in \mathbf{Z}^d : 0 < |\xi| \leq K^{100/\beta}\}.$$

Then there exists a constant C such that

$$\mathbf{P}(\|\hat{\eta}\|_{L^\infty(B)} \geq CK^{-1/2} \log(K)^{1/2}) \leq 1/10.$$

Remark 3.11. In particular, (3.34) holds if the $\{X_i\}$ are uniformly distributed on \mathbf{T}^d .

Proof of Lemma 3.10. For each $\xi \in \mathbf{Z}^d$ and $k \in \{1, \dots, K\}$, consider the random variable

$$Y(\xi, k) = K^{-1} e^{2\pi i(\xi \cdot X_k)}. \quad (3.35)$$

Then for each $\xi \in \mathbf{Z}^d$,

$$\sum_{k=1}^K Y(\xi, k) = \hat{\eta}(\xi). \quad (3.36)$$

We also note that for each $\xi \in \mathbf{Z}^d$ and $k \in \{1, \dots, K\}$,

$$|Y(\xi, k)| = K^{-1}. \quad (3.37)$$

Moreover,

$$\sum_{k=1}^K \mathbf{E}(Y(\xi, k)) = 0. \quad (3.38)$$

Since the family of random variables $\{Y(\xi, 1), \dots, Y(\xi, K)\}$ is independent for a fixed non-zero ξ , we can apply Hoeffding's inequality together with (3.36) and (3.37) to conclude that for all $t \geq 0$,

$$\mathbf{P}(|\hat{\eta}(\xi)| \geq t) \leq 2e^{-Kt^2/2}. \quad (3.39)$$

A union bound obtained by applying (3.39) over all $|\xi| \leq K^{1/\beta+1}$ shows that if

$$B = \{\xi \in \mathbf{Z}^d : |\xi| \leq K^{100/\beta}\},$$

then there exists a constant $C \geq 1$ such that

$$\mathbf{P}(\|\hat{\eta}\|_{L^\infty(B)} \geq t) \leq \exp\left(C \log(K) - \frac{5Kt^2}{C}\right). \quad (3.40)$$

But then, setting $t = CK^{-1/2} \log(K)^{1/2}$ in (3.40) completes the proof. \square

Let us now consider the consequences of the square-root cancellation bound in Lemma 3.10. Given η as in Lemma 3.10, consider the smooth function $f(x) = (\eta * \phi_r)(x)$. The support of f consists of K radius r balls. Provided $K \approx r^{-\beta}$, the support therefore behaves like an r -thickening of a set with Minkowski dimension β . If

$$\|\hat{\eta}\|_{L^\infty(B)} \leq CK^{-1/2} \log(K)^{1/2}, \quad (3.41)$$

as is guaranteed with high probability in Lemma 3.10, we actually find that the support of f also behaves like an r -thickening of a set with Fourier dimension β . This is the reason why constructions involving some kind of square-root cancellation are often a viable tactic to construct Salem sets, with random examples being an important example of such a construction.

Lemma 3.12. *Fix $\varepsilon, \varepsilon_1 > 0$ and $C > 0$ with $\varepsilon_1 \leq \beta/2$. Then there exists $r_0 > 0$ and $C' > 0$ depending on these quantities such that for $r \leq r_0$, if $K \geq (1/C)r^{-\beta}$ and η is a Borel probability measure with*

$$\|\hat{\eta}\|_{L^\infty(B)} \leq CK^{-1/2} \log(K)^{1/2},$$

then

$$|\hat{\eta}(\xi)| \leq \varepsilon |\xi|^{\varepsilon_1 - \beta/2} \quad \text{for } |\xi| \leq (1/r)$$

and

$$|\hat{\eta}(\xi)| \leq C' r^{\beta/2} \log(1/r)^{1/2} \quad \text{for } (1/r) \leq |\xi| \leq (1/r)^2.$$

Proof. The function $x \mapsto x^{-1/2} \log(x)^{1/2}$ is decreasing for sufficiently large x . Thus if r_0 is chosen appropriately small, then for $|\xi| \leq (1/r)$ we find

$$|\hat{\eta}(\xi)| \leq CK^{-1/2} \log(K)^{1/2} \leq Cr^{\varepsilon_2} \log((1/C)r^{-\beta})^{1/2} \cdot |\xi|^{\varepsilon_2 - \beta/2} \leq \varepsilon \cdot |\xi|^{\varepsilon_2 - \beta/2}. \quad (3.42)$$

If r_0 is appropriately small, then

$$(1/r)^2 \leq C^{2/\beta} K^{2/\beta} \leq K^{100/\beta}$$

Thus for $(1/r) \leq |\xi| \leq (1/r)^2$, $\xi \in B$, and so we find there is $C' > 0$ such that

$$|\hat{\eta}(\xi)| \leq CK^{-1/2} \log(K)^{1/2} \leq Cr^{\beta/2} \log((1/C)r^{-\beta})^{1/2} \leq C' r^{\beta/2} \log(1/r)^{1/2}. \quad (3.43)$$

Together, (3.42) and (3.43) give what was required. \square

It is a general heuristic that quasi-all sets are as ‘thin as possible’ with respect to the Hausdorff metric. In particular, we should expect the Hausdorff dimension and Fourier dimension of a generic element of \mathcal{X}_β to be as low as possible. For each $(E, \mu) \in \mathcal{X}_\beta$, the condition that $\mu \in M(\beta/2)$ implies that $\dim_{\mathbf{F}}(\mu) \geq \beta$, so $\dim_{\mathbf{F}}(E) \geq \beta$. Thus it is natural to expect that for quasi-all $(E, \mu) \in M(\beta/2)$, the set E has both Hausdorff dimension and Fourier dimension equal to β .

Lemma 3.13. *For quasi-all $(E, \mu) \in \mathcal{X}_\beta$, E is a Salem set of dimension β .*

Proof. We shall assume $\beta < d$ in the proof, since when $\beta = d$, E is a Salem set for any $(E, \mu) \in \mathcal{X}_\beta$, and thus the result is trivial. Since the Hausdorff dimension of a measure is an upper bound for the Fourier dimension, it suffices to show that for quasi-all $(E, \mu) \in \mathcal{X}_\beta$, E has Hausdorff dimension at most β . For each $\alpha > \beta$ and $\delta, s > 0$, and let $A(\alpha, \delta, s) = \{(E, \mu) \in \mathcal{X} : H_\delta^\alpha(E) < s\}$. Then $A(\alpha, \delta, s)$ is an open subset of \mathcal{X}_β , and

$$\bigcap_{n=1}^{\infty} \bigcap_{m=1}^{\infty} \bigcap_{k=1}^{\infty} A(\beta + 1/n, 1/m, 1/k)$$

is precisely the family of $(E, \mu) \in \mathcal{X}_\beta$ such that E has Hausdorff dimension at β . Thus it suffices to show that $A(\alpha, \delta, s)$ is dense in \mathcal{X}_β for $\alpha \in (\beta, d)$ and $\delta, s > 0$. Fix $(E_0, \mu_0) \in \mathcal{X}_\beta$, $\alpha \in (\beta, d)$, $\delta > 0$, $s > 0$, and $\varepsilon_1 > 0$. We aim to show that for each $\varepsilon > 0$, there exists $(E, \mu) \in A(\alpha, \delta, s)$ such that $d_{\mathbf{H}}(E, E_0) \leq \varepsilon$ and $\|\mu - \mu_0\|_{M(\beta/2 - \varepsilon_1)} \leq \varepsilon$. Without loss of generality, in light of Lemma 3.1, we may assume that $\mu_0 \in C^\infty(\mathbf{T}^d)$.

Fix a small value r , and then find an integer K such that $r^{-\beta} \leq K \leq r^{-\beta} + 1$. Lemma 3.10 shows that there exists a constant C depending on β and d , as well as K points $x_1, \dots, x_K \in \mathbf{T}^d$ such that if

$$\eta(x) = \frac{1}{K} \sum_{k=1}^K \delta(x - x_k),$$

then for each $|\xi| \leq (1/r)^{1+1/\beta} \leq K^{1/\beta+1}$,

$$|\hat{\eta}(\xi)| \leq CK^{-1/2} \log(K)^{1/2}. \quad (3.44)$$

Equation (3.44) shows η satisfies the hypotheses of Lemma 3.12, and the result of Lemma 3.12 can then be fed into Corollary 3.9 to conclude that there exists $r_1 > 0$ such that if $r \leq r_1$, if

$$\mu_1(x) = \frac{1}{K} \left(\sum_{k=1}^K \phi_r(x - x_k) \right) \mu_0(x),$$

and if

$$\mu = \mu_1 / \mu_1(\mathbf{T}^d),$$

then

$$\|\mu - \mu_0\|_{M(\beta/2 - \varepsilon_1)} \leq \varepsilon. \quad (3.45)$$

Note that μ is supported on K balls of radius r . Thus for $r \leq \delta$,

$$H_\delta^\alpha(\text{supp}(\mu)) \leq Kr^\alpha \leq (r^{-\beta} + 1)r^\alpha = r^{\alpha-\beta} + r^\alpha. \quad (3.46)$$

Since $\alpha > \beta$, (3.46) implies that there is r_2 depending on α , β , and s such that for $r \leq r_2$,

$$H_\delta^\alpha(\text{supp}(\mu)) \leq s. \quad (3.47)$$

Now let

$$E = \text{supp}(\mu) \cup \{y_1, \dots, y_N\},$$

where $\{y_1, \dots, y_N\} \subset E_0$ is a ε -net of E_0 . Set $r_0 = \min(r_1, r_2, \delta)$ and suppose $r \leq r_0$. Equation (3.47) implies that $H_\delta^\alpha(E) \leq s$, so $(E, \mu) \in A(\alpha, \delta, s)$. And since $\text{supp}(\mu) \subset E$,

$$d_{\mathbf{H}}(E, E_0) \leq \varepsilon. \quad (3.48)$$

Recalling (3.45), we see that we have proved what was required. \square

This concludes the setup to the proof of Theorems 1.3 and TODO. All that remains is to show that quasi-all elements of \mathcal{X}_β avoid the given set Z ; just as with some of the proofs we have given in this section, the advantage of the Baire category approach is that we can reduce our calculations to discussing only a couple scales at once, which allows us to focus solely on the discrete, quantitative question at the heart of the problem.

4 Random Avoiding Sets for Rough Patterns

In the last section, our results held for an arbitrary $\beta \in (0, d]$. But in this section, we assume

$$\beta = \min \left(d, \frac{dn - \alpha}{n - 1/2} \right),$$

which will enable us to generically avoid the pattern Z described in Theorem 1.3. The construction here is very similar to the construction in [3] albeit in the Baire category setting, though we must modify parameters to ensure a Fourier dimension bound rather than just a Hausdorff dimension bound.

Lemma 4.1. *Suppose $Z \subset \mathbf{T}^{dn}$ is a countable union of compact sets, each with lower Minkowski dimension at most α . Then for quasi-all $(E, \mu) \in \mathcal{X}_\beta$, for any distinct points $x_1, \dots, x_n \in E$, $(x_1, \dots, x_n) \notin Z$.*

Proof. The set $Z \subset \mathbf{R}^{dn}$ is the countable union of sets with lower Minkowski dimension at most α . For a closed set $W \subset \mathbf{T}^{dn}$ with lower Minkowski dimension at most α , and $s > 0$, consider the set

$$B(W, s) = \left\{ (E, \mu) \in \mathcal{X}_\beta : \begin{array}{l} \text{for all } x_1, \dots, x_n \in E \text{ such that} \\ |x_i - x_j| \geq s \text{ for } i \neq j, (x_1, \dots, x_n) \notin W \end{array} \right\}.$$

If $(E_0, \mu_0) \in B(W, s)$, then because E_0 is compact, so too is the set

$$F = \{(x_1, \dots, x_n) \in E_0^n : |x_i - x_j| \geq s \text{ for } i \neq j\}$$

Since W is also closed, hence compact, there exists $\varepsilon > 0$ such that if $(x_1, \dots, x_n) \in F$, then $d((x_1, \dots, x_n), W) > \varepsilon$. It follows that if $d_{\mathbf{H}}(E_0, E) \leq \varepsilon$, then for any measure μ supported on E , $(E, \mu) \in B(W, s)$. Thus $B(W, s)$ is an open subset of \mathcal{X}_β . If Z is a countable union of closed sets $\{Z_k\}$ with lower Minkowski at most α , then clearly the set

$$\bigcup_{k=1}^{\infty} \bigcup_{n=1}^{\infty} B(Z_k, 1/n)$$

consists of the family of sets (E, μ) such that for distinct $x_1, \dots, x_n \in E$, $(x_1, \dots, x_n) \notin Z$. Thus it suffices to show that $B(W, s)$ is dense in \mathcal{X}_β for any $s > 0$, and any closed set W with lower Minkowski dimension at most α .

Let us begin by fixing a set $W \subset \mathbf{T}^{dn}$ and a pair $(E_0, \mu_0) \in \mathcal{X}_\beta$. We will show that for any $\varepsilon_1 \in (0, \beta/100]$ and $\varepsilon > 0$, we can find $(E, \mu) \in B(W, s)$ with $d_{\mathbf{H}}(E, E_0) \leq \varepsilon$ and $\|\mu - \mu_0\|_{M(\beta/2 - \varepsilon_1)} \leq \varepsilon$. We may assume by Corollary 3.4 that $\text{supp}(\mu_0) = E$ and $\mu_0 \in C^\infty(\mathbf{T}^d)$.

Since W has lower Minkowski dimension at most α , we can find arbitrarily small $r \in (0, 1)$ such that

$$|W_r| \leq r^{dn-\alpha-\varepsilon_1/4}. \quad (4.1)$$

Assume also that r is small enough that we can find an integer $K \geq 10$ with

$$r^{\varepsilon_1/2-\beta} \leq K \leq r^{\varepsilon_1/2-\beta} + 1. \quad (4.2)$$

Let X_1, \dots, X_K be independent and uniformly distributed on \mathbf{T}^d . For each distinct set of indices $k_1, \dots, k_n \in \{1, \dots, K\}$, the random vector $X_k = (X_{k_1}, \dots, X_{k_n})$ is uniformly distributed on \mathbf{T}^{nd} , and so (4.1) and (4.2) imply that

$$\begin{aligned} \mathbf{P}(d(X_k, W) \leq r) &\leq |W_r| \\ &\leq r^{dn-\alpha-\varepsilon_1/4} \\ &\lesssim_{d,n,\beta} K^{-\frac{(dn-\alpha-\varepsilon_1/4)}{\beta-\varepsilon_1/2}} \\ &= K^{-(n-1/2-\varepsilon_1/4\beta)\left(1+\frac{\varepsilon_1}{2\beta-\varepsilon_1}\right)} \\ &= K^{1/2-n+\varepsilon_1/4\beta-\frac{\varepsilon_1}{2\beta-\varepsilon_1}(n-1/2-6\varepsilon_1/\beta)} \\ &\leq K^{1/2-n+\varepsilon_1/4\beta-\frac{\varepsilon_1}{2\beta-\varepsilon_1}} \leq K^{1/2-n}. \end{aligned} \quad (4.3)$$

If M_0 denotes the number of indices i such that $d(X_i, W) \leq r$, then by linearity of expectation we conclude from (4.3) that there is a constant C depending only on d, n , and β such that

$$\mathbf{E}(M_0) \leq (C/10)K^{1/2}. \quad (4.4)$$

Applying Markov's inequality to (4.4), we conclude that

$$\mathbf{P}(M_0 \geq CK^{1/2}) \leq 1/10. \quad (4.5)$$

Taking a union bound to (4.5) and the results of Lemma 3.10, we conclude that there exists K points $x_1, \dots, x_K \in \mathbf{T}^d$ and a constant C depending only on d, n , and β such that the following two statements hold:

- (1) Let S be the set of indices $k_1 \in \{1, \dots, K\}$ with the property that we can find distinct indices $k_2, \dots, k_n \in \{1, \dots, K\}$ such that if $X = (X_{k_1}, \dots, X_{k_n})$, then $d(X, W) \leq r$. Then

$$\#(S) \leq CK^{1/2}. \quad (4.6)$$

- (2) If we define

$$\eta_0(x) = \frac{1}{K} \sum_{k=1}^K \delta(x - x_k)$$

then for $0 < |\xi| \leq K^{1/\beta+1}$,

$$|\widehat{\eta}_0(\xi)| \leq CK^{-1/2} \log(K)^{1/2}. \quad (4.7)$$

Thus (4.6) and (4.7) imply that if

$$\eta_1(x) = \sum_{k \notin S} \delta(x - x_k),$$

then for each $|\xi| \leq K^{1/\beta_0+1}$,

$$|\widehat{\eta}_1(\xi)| \leq 2CK^{-1/2} \log(K). \quad (4.8)$$

Since $K \geq r^{\varepsilon_1/2-\beta}$, we can apply Lemma 3.12 to η_1 , and then apply Corollary 3.9 to the result of Lemma 3.12 to conclude that for any $\delta > 0$ to find $r_0 > 0$ such that if $r \leq r_0$ and we define

$$\mu'(x) = \left(\sum_{k \notin S} \phi_{(r/2n^{1/2})}(x - X_k) \right) \mu_0(x),$$

and then set $\mu = \mu'/\mu'(\mathbf{T}^d)$, then

$$\|\mu - \mu_0\|_{M(\beta/2-\varepsilon_1)} \leq \min(\delta, \varepsilon). \quad (4.9)$$

Choosing δ in accordance with Lemma 3.7, (4.9) also implies that

$$d_{\mathbf{H}}(\text{supp}(\mu), \text{supp}(\mu_0)) \leq \varepsilon.$$

Thus all that remains to prove the density statement is to show that if $E = \text{supp}(\mu)$, then $(E, \mu) \in B(W, s)$.

Consider n points $y_1, \dots, y_n \in \text{supp}(\mu)$, with $|y_i - y_j| \geq r$ for any two indices $i \neq j$. We can therefore find distinct indices $k_1, \dots, k_n \in \{1, \dots, K\}$ such that for each $i \in \{1, \dots, n\}$,

$$|x_{k_i} - y_i| \leq (n^{-1/2}/2) \cdot r. \quad (4.10)$$

If we set $x = (x_{k_1}, \dots, x_{k_n})$ and $y = (y_1, \dots, y_n)$, then (4.10) implies

$$|x - y| \leq (r/2). \quad (4.11)$$

Since $i_1 \notin S$, $d(x, W) \geq r$, which combined with (4.11) implies

$$d(y, W) \geq d(x, W) - |x - y| \geq r/2. \quad (4.12)$$

Thus in particular we conclude $y \notin W$, which shows $(E, \mu) \in B(W, s)$. \square

Before we move onto the next proof, let us discuss where the loss in Theorem 1.1 occurs in our proof, as compared to the Hausdorff dimension bound of [3]. In the proof of Lemma 4.1, in order to obtain the bound (4.8), we were forced to choose the parameter r such that $\#(S) \leq K^{1/2}$, so that the trivial bound

$$\left| \sum_{k \in S} e^{2\pi i(\xi \cdot x_k)} \right| \leq \#(S) \quad (4.13)$$

obtained by taking in absolute values is viable to obtain a Fourier dimension bound. On the other hand, if we were able to justify that with high probability, we could obtain a square root cancellation bound

$$\left| \sum_{k \in S} e^{2\pi i(\xi \cdot x_k)} \right| \lesssim \#(S)^{1/2}, \quad (4.14)$$

then we would only need to choose the parameter r such that $\#(S) \lesssim K$, which leads to a set with larger Fourier dimension, matching with the relevant Hausdorff dimension bound obtained in [3]. Under stronger assumptions on the pattern we are trying to avoid, we are roughly able to obtain such a result in the next section (though with an additional term which decays fast as $\xi \rightarrow \infty$), which enables us to improve our Fourier dimension of the sets we obtain.

5 The More Difficult Case

Lemma 5.1. *Let $Z \subset \mathbf{T}^{dn}$ be a countable union of sets $\{W_i\}$, where*

Proof. Fix a manifold $W \subset \mathbf{T}^{dn}$ of dimension $dn-d$ with the appropriate regularity condition, and set

$$\beta = \frac{d}{n-1}.$$

Given any family of disjoint, sidelength s cubes $R_1, \dots, R_n \subset \mathbf{T}^d$ such that $d(R_i, R_j) \geq 10s$, we let

$$H(W; R_1, \dots, R_n) = \{(E, \mu) \in \mathcal{X}_\beta : (R_1 \times \dots \times R_n) \cap E^n \cap W = \emptyset\}. \quad (5.1)$$

Then $H(W; R_1, \dots, R_n)$ is an open subset of \mathcal{X}_β . For the purpose of a Baire category argument, our result will follow by showing $H(W; R_1, \dots, R_n)$ is dense in \mathcal{X}_β for any family of cubes $\{R_1, \dots, R_n\}$. Without loss of generality, possibly permuting coordinates if necessary we may assume that $1/s$ is an integer, and that for the family of cubes $\{R_1, \dots, R_n\}$, if $Q_1 = 2R_1, \dots, Q_n = 2R_n$, then there exists a Lipschitz function $f : Q_2 \times \dots \times Q_n \rightarrow \mathbf{R}^d$ such that

$$W \cap (Q_1 \times \dots \times Q_n) = \{(x_1, \dots, x_d) \in Q_1 \times \dots \times Q_n : x_1 = f(x_2, \dots, x_d)\}. \quad (5.2)$$

Fix a constant $L \geq 0$ such that for any $x, x' \in Q_2 \times \dots \times Q_n$,

$$|f(x) - f(x')| \leq L|x - x'|. \quad (5.3)$$

As with the previous proof, we fix $(E_0, \mu_0) \in \mathcal{X}_\beta$ with $\text{supp}(\mu_0) = E_0$ and $\mu_0 \in C^\infty(\mathbf{T}^d)$, and show that for any $\varepsilon_1 \in (0, \beta/100]$ and $\varepsilon > 0$, there is $(E, \mu) \in H(W; Q_1, \dots, Q_n)$ with $\|\mu - \mu_0\|_{M(\beta/2-\varepsilon_1)} \leq \varepsilon$.

Fix a family of non-negative bump functions $\psi_0, \psi_1, \dots, \psi_n \in C^\infty(\mathbf{T}^d)$, such that for $i \in \{1, \dots, n\}$, $\psi_i(x) = 1$ for $x \in R_i$, $\psi_i(x) = 0$ for $x \notin Q_i$, and $\psi_0(x) + \dots + \psi_n(x) = 1$ for $x \in \mathbf{T}^d$. Let A_0, \dots, A_n denote the total mass of ψ_0, \dots, ψ_n respectively. Now consider a family of independant random variables $\{X_i(k) : 0 \leq i \leq n, 1 \leq k \leq K\}$, where the random variable $X_i(k)$ is continuous with distribution function $A_i^{-1}\psi_i$. Set $r = K^{\varepsilon_1/2-1/\beta}$ and then set

S to be the set of indices $k_1 \in \{1, \dots, K\}$ such that there are indices $k_2, \dots, k_n \in \{1, \dots, K\}$ with the property that

$$|X_1(k_1) - f(X_2(k_2), \dots, X_n(k_n))| \leq r/(L+1). \quad (5.4)$$

A simple argument following from (5.4) shows that if $k_1 \notin S$, then for any $k_2, \dots, k_n \in \{1, \dots, K\}$, if $X = (X_1(k_1), \dots, X_n(k_n))$, then $d(X, W) \geq r$. Thus if we define

$$\mu'(x) = \left(\sum_{i \in \{0, 2, \dots, n\}} A_i \sum_{k=1}^K \phi_{(r/2n^{1/2})}(x - X_i(k)) + A_1 \sum_{k \notin S} \phi_{(r/2n^{1/2})}(x - X_1(k)) \right) \cdot \mu_0(x), \quad (5.5)$$

and then set $\mu = \mu'/\mu'(\mathbf{T}^d)$, $E = \text{supp}(\mu)$, then $(E, \mu) \in H(W; R_1, \dots, R_n)$. The remainder of our argument consists of obtaining control on the exponential sum

$$U(\xi) = \sum_{i \in \{0, 2, \dots, n\}} A_i \sum_{k=1}^K e^{2\pi i \xi \cdot X_i(k)} + A_1 \sum_{k \notin S} e^{2\pi i \xi \cdot X_1(k)}. \quad (5.6)$$

so that we may apply Corollary 3.9.

For $\xi \in \mathbf{Z}^d$, set

$$V(\xi) = \frac{1}{K} \sum_{i=0}^n A_i \sum_{k=1}^K e^{2\pi i \xi \cdot X_i(k)}. \quad (5.7)$$

Then for non-zero $\xi \in \mathbf{Z}^d$,

$$\mathbf{E}(V(\xi)) = \sum_{i=0}^n \int \psi_i(x) e^{2\pi i \xi \cdot x} dx = K \int_{\mathbf{T}^d} e^{2\pi i \xi \cdot x} dx = 0. \quad (5.8)$$

Applying Lemma 3.10, we conclude that if $B = \{\xi \in \mathbf{Z}^d : 0 < |\xi| < K^{100/\beta}\}$, then there is $C > 0$ such that

$$\mathbf{P}(\|V\|_{L^\infty(B)} \geq CK^{-1/2} \log(K)^{1/2}) \leq 1/10. \quad (5.9)$$

It remains to analyze

$$Y(\xi) = V(\xi) - U(\xi) = \frac{A_1}{K} \sum_{k \in S} e^{2\pi i \xi \cdot X_k}. \quad (5.10)$$

Let us briefly discuss our strategy to obtain a concentration bound for $Y(\xi)$. The random quantity $Y(\xi)$ is a function of the independent random quantities $\{X_i(k)\}$, and so we would like to apply McDiarmid's inequality to understand the concentration properties of $Y(\xi)$ about its mean. However, a naive application of McDiarmid's inequality fails here, because changing a single random variable $X_i(k)$ for $i \geq 2$ can change the quantity $Y(\xi)$ by $\Omega(1)$, which is far too much square root cancellation (TODO: Diagram). On the other hand, it is highly unlikely that the random variables $\{X_1(k)\}$ are arranged in such a way that $X_i(k)$ can change $Y(\xi)$ by $\Omega(1)$, so we should expect that adjusting a single random variable $X_i(k)$ does not influence the value of $Y(\xi)$ very much if the quantity $Y(\xi)$ is *averaged* over the possible choices of $\{X_1(k)\}$.

Consider the random set $\Omega \subset Q_1$, which is the set of values $x \in Q_1$ such that there are $k_2, \dots, k_n \in \{1, \dots, K\}$ with

$$|x - f(X_2(k_2), \dots, X_n(k_n))| \leq r. \quad (5.11)$$

Then

$$Y(\xi) = \frac{1}{K} \sum_{k=1}^K Z_k. \quad (5.12)$$

where

$$Z(k) = \begin{cases} e^{2\pi i \xi \cdot X_1(k)} & : X_1(k) \in \Omega, \\ 0 & : X_1(k) \notin \Omega \end{cases}. \quad (5.13)$$

If Σ is the σ algebra generated by the random variables $\{X_i(k) : i \geq 2, k \in \{1, \dots, K\}\}$, then the random variables $\{Z(k)\}$ are *conditionally independant* (and identically distributed) given Σ . Since we have $|Z(k)| \leq 1$ almost surely, Hoeffding's inequality thus implies that for all $t \geq 0$,

$$\mathbf{P}(|Y(\xi) - \mathbf{E}(Y(\xi)|\Sigma)| \geq t) \leq 4 \exp\left(\frac{-Kt^2}{2}\right). \quad (5.14)$$

It is simple to see that

$$\mathbf{E}(Y(\xi)|\Sigma) = \int_{\Omega} \psi_1(x) e^{2\pi i \xi \cdot x} dx. \quad (5.15)$$

Since

$$\Omega = \bigcup \{B_r(f(X_2(k_2), \dots, X_n(k_n))) : 1 \leq k_2, \dots, k_n \leq K\}. \quad (5.16)$$

we see that varying each random variable $X_i(k)$ adjusts at most K^{n-2} of the balls forming Ω , and thus varying $X_i(k)$ independantly of the other random variables changes $\mathbf{E}(Y(\xi)|\Sigma)$ by at most

$$2 \cdot (2r)^d \cdot K^{n-2} = 2^{d+1} \cdot r^d \cdot K^{n-2} = \frac{2^{d+1} \cdot r^{(n-1)\varepsilon_1/2}}{K} \leq \frac{2^{d+1}}{K}. \quad (5.17)$$

Thus McDiarmid's inequality shows that for any $t \geq 0$,

$$\mathbf{P}(|\mathbf{E}(Y(\xi)|\Sigma) - \mathbf{E}(Y(\xi))| \geq t) \leq 4 \exp\left(\frac{-Kt^2}{2^{2d+1}}\right). \quad (5.18)$$

Combining (5.14) and (5.18), we conclude that for each $\xi \in \mathbf{Z}^d$,

$$\mathbf{P}(|Y(\xi) - \mathbf{E}(Y(\xi))| \geq t) \leq 8 \exp\left(\frac{-Kt^2}{2^{2d+1}}\right). \quad (5.19)$$

Applying a union bound to (5.19) over all $\xi \in B$ shows that there exists a constant $C > 0$ such that

$$\mathbf{P}(\|Y - \mathbf{E}(Y)\|_{L^\infty(B)} \geq CK^{-1/2} \log(K)^{1/2}) \leq 1/10. \quad (5.20)$$

Combining (5.9) and (5.20), we conclude that there exists a constant $C > 0$ such that

$$\mathbf{P}(\|U - \mathbf{E}(Y)\|_{L^\infty(B)} \geq CK^{-1/2} \log(K)^{1/2}) \leq 1/2. \quad (5.21)$$

Our calculation will therefore be complete if we can bound the quantity $\mathbf{E}(Y)$.

For each non-zero $\xi \in \mathbf{Z}^d$, let η_W be the surface measure on W obtained from the pullback of the Lebesgue measure via the projection onto (x_2, \dots, x_d) , let σ_r be the measure of the ball of radius r in $\mathbf{T}^d \times \{0\}^{n-1}$, and let $\psi = \psi_1 \otimes \dots \otimes \psi_n$. We calculate that

$$\begin{aligned}
\mathbf{E}(Y(\xi)) &= \frac{K^{n-1}}{A_1 \dots A_n} \int_W \int_{B_r(0)} \psi(x_1 + y, x_2, \dots, x_n) e^{2\pi i \xi \cdot (x_1 + y)} dy dx \\
&= \frac{K^{n-1}}{A_1 \dots A_n} ((\psi \cdot \eta_W) * \sigma_r)^\wedge(-\xi, 0, \dots, 0) \\
&= \frac{K^{n-1}}{A_1 \dots A_n} \widehat{\psi \cdot \eta_W}(-\xi, 0, \dots, 0) \cdot \widehat{\sigma_r}(-\xi, 0, \dots, 0) \\
&= \frac{r^d K^{n-1}}{A_1 \dots A_n} \widehat{\psi \cdot \eta_W}(-\xi, 0, \dots, 0) \cdot \widehat{\sigma_1}(-r\xi, 0, \dots, 0) \\
&= \frac{r^{(n-1)\varepsilon_1/2}}{A_1 \dots A_n} \widehat{\psi \cdot \eta_W}(-\xi, 0, \dots, 0) \cdot \widehat{\sigma_1}(-r\xi, 0, \dots, 0).
\end{aligned} \tag{5.22}$$

Assuming $\nabla f \neq 0$ on $Q_2 \times \dots \times Q_d$, we conclude that

$$\widehat{\psi \cdot \eta_W}(-\xi, 0, \dots, 0) \lesssim |\xi|^{-\beta/2}. \tag{5.23}$$

Thus for any $\delta > 0$, there exists r_1 such that for any $r \leq r_1$, and for any nonzero $\xi \in \mathbf{Z}^d$,

$$|\mathbf{E}(Y(\xi))| \leq \delta |\xi|^{-\beta/2}. \tag{5.24}$$

From (5.21) and (5.24), we conclude there exists a choice of values $\{X_i(k)\}$ such that for any $\xi \in B$,

$$|U(\xi)| \leq CK^{-1/2} \log(K)^{1/2} + \delta |\xi|^{-\beta/2}. \tag{5.25}$$

Applying Lemma 3.12 and then Corollary 3.9, we conclude that if δ is chosen appropriately, then $\|\mu - \mu_0\|_{M(\beta/2 - \varepsilon_1)} \leq \varepsilon$, which completes the proof. \square

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