

Geometric Measure Theory

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Chapter 1

Fractal Dimensions

The expression of certain geometric properties of subsets of \mathbf{R}^d requires more than can be expressed using the d dimensional Lebesgue measure. For instance, curves and surfaces all have measure zero in two and three dimensions respectively, and so we cannot distinguish them by the Lebesgue measure from any of the other nasty Lebesgue measurable subsets of measure zero. Hausdorff showed that there is a notion of ‘dimension’ of measure zero subsets of \mathbf{R}^d which matches the dimension of corresponding curves and surfaces. Even more interestingly, Hausdorff’s theory of dimension gives certain fractal subsets non-integer dimension. It is very useful when studying non-smooth shapes, like fractals, and in many quantitative questions in harmonic analysis.

1.1 Minkowski Dimension

The easiest fractal dimension to introduce is Minkowski dimension. If E is a bounded set in \mathbf{R}^n , then we can consider the open set

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

which is an ‘ δ thickening’ of E . We define the *lower* and *upper* Minkowski dimension as

$$\underline{\dim}_{\mathbf{M}}(E) = \liminf_{\delta \rightarrow 0} \left(n - \frac{\log |E_\delta|}{\log \delta} \right)$$

and

$$\overline{\dim}_{\mathbf{M}}(E) = \limsup_{\delta \rightarrow 0} \left(n - \frac{\log |E_\delta|}{\log \delta} \right).$$

If $\underline{\dim}_{\mathbf{M}}(E) = \overline{\dim}_{\mathbf{M}}(E)$, then we refer to this common quantity as the Minkowski dimension $\dim_{\mathbf{M}}(E)$. One can interpret that if $\dim_{\mathbf{M}}(E) = \alpha$, then $|E_\delta| = \delta^{n-\alpha+o(1)}$. These notions can also be extended to unbounded sets by considering the supremum over all bounded subsets. For a set E , we define

$$\underline{\dim}_{\mathbf{M}}(E) = \liminf_{r \rightarrow \infty} [\underline{\dim}_{\mathbf{M}}(E \cap B_r(0))]$$

and

$$\overline{\dim}_{\mathbf{M}}(E) = \limsup_{r \rightarrow \infty} [\overline{\dim}_{\mathbf{M}}(E \cap B_r(0))].$$

The Minkowski dimension $\dim_{\mathbf{M}}(E)$ of an unbounded set is defined as the common value of these two functions, if they agree.

Example. If $E = B^k \times \{0\}^{n-k}$, where B^k is the k dimensional unit ball, then

$$B^k \times \delta B^{n-k} \subset E_\delta \subset (1 + \delta) B^k \times \delta B^{n-k}$$

which shows that

$$\delta^{n-k} \lesssim |E_\delta| \lesssim (1 + \delta)^k \delta^{n-k}$$

Thus $\dim_{\mathbf{M}}(E) = k$. In particular, $\dim_{\mathbf{M}}(rE) = k$ for all $r > 0$, and so taking $r \rightarrow \infty$ shows $\dim_{\mathbf{M}}(\mathbf{R}^k \times 0^{n-k}) = k$.

Example. Let

$$C = \left\{ \sum_{i=1}^{\infty} a_i / 4^i : a_i \in \{0, 3\} \right\}$$

be a Cantor set. If $1/4^{N+1} \leq \delta \leq 1/4^N$, then

$$\left\{ \sum_{i=1}^{\infty} a_i / 4^i : a_1, \dots, a_{N+1} \in \{0, 3\} \right\} \subset C_\delta \subset \left\{ \sum_{i=1}^{\infty} a_i / 4^i : a_1, \dots, a_N \in \{0, 3\} \right\}.$$

The latter set has volume $2^N / 4^N = 1/2^N \leq (2\delta)^{1/2}$. The initial set has volume $2^{N+1} / 4^{N+1} = 1/2^{N+1} \geq \delta^{1/2}$. Thus $\log |C_\delta| = \log(\delta)/2 + O(1)$, and so C has Minkowski dimension $1/2$.

Example. We can modify the last example slightly, considering

$$C = \left\{ \sum_{i=1}^{\infty} a_i / 4^i : a_i \in \{0, 3\} \text{ if there is } k \text{ such that } (2k)! \leq i \leq (2k+1)! \right\}.$$

Then C has lower Minkowski dimension $1/2$ and upper Minkowski dimension 1 . If one looks at the iterated construction of C , one sees that we only dissect C at an incredibly sparse range of scales.

Many fractals display self similarity properties. For instance, if C is the classical Cantor set, then $3C$ is the union of two translates of C . The next lemma thus implies the Minkowski dimension of the Cantor set is $\log_3(2)$.

Theorem 1.1. If $E \subset \mathbf{R}^d$ is compact, $r > 1$, and there is r such that rE is the union of k disjoint translates of E , then $\dim_M(E) = \log_r k$.

Proof. For small δ , $(rE)_\delta$ is the union of k disjoint translates of E_δ , so

$$r^d |E_{\delta/r}| = |(rE)_\delta| = k |E_\delta|.$$

In particular, this means that $|E_{1/r^N}|$ is proportional to $(k/r^d)^N$. But this means that for any $1/r^{N+1} \leq \delta \leq 1/r^N$,

$$|E_\delta| \sim (k/r^d)^N \sim (k/r^d)^{-\log_r \delta} = \delta^{d - \log_r k}.$$

Thus $\dim_M(E) = \log_r k$. □

There are several alternate definitions of Minkowski dimension. Given a bounded set E , and $\delta > 0$, we let

- $N_\delta^{\text{Ext}}(E)$ denote the minimum number of δ balls required to cover E .
- $N_\delta^{\text{Int}}(E)$ denotes the minimum number of δ balls with centers in E required to cover E .
- $N_\delta^{\text{Pack}}(E)$ is the largest number of disjoint open balls of radius δ with centers in E .

Given a cover of E by N balls of radius δ , by doubling the radius of the balls, we can cover E by N balls of radius 2δ with centers of E . Thus $N_{2\delta}^{\text{Int}}(E) \leq N_\delta^{\text{Ext}}(E) \leq N_\delta^{\text{Int}}(E)$. Conversely, if we have a maximal packing by

N radius δ balls, then we can cover E by N radius 2δ balls. Thus $N_{2\delta}^{\text{Int}}(E) \leq N_{\delta}^{\text{Pack}}(E)$. On the other hand, $N_{\delta}^{\text{Pack}}(E) \leq |E_{\delta}|/|\delta B^n| \leq N_{\delta}^{\text{Ext}}(E)$, because a packing of balls inside E provides a disjoint subset of balls in E_{δ} , and if we cover E by δ balls, then E_{δ} is covered by the radius 2δ balls with the same centres. In particular, we have shown that as $\delta \rightarrow 0$, all the quantities $\log_{1/\delta} N_{\delta}^*(E)$ are comparable to one another. Since

$$\delta^n |N_{\delta}^{\text{Pack}}(E)| \lesssim |E_{\delta}| \lesssim \delta^n |N_{\delta}^{\text{Ext}}(E)|$$

We find that

$$\underline{\dim}_{\mathbf{M}}(E) = \liminf_{\delta \rightarrow 0} \log_{1/\delta} N_{\delta}^*(E) \quad \text{and} \quad \overline{\dim}_{\mathbf{M}}(E) = \limsup_{\delta \rightarrow 0} \log_{1/\delta} N_{\delta}^*(E).$$

We can use these as definitions of the Minkowski dimension, which has the advantage that they can be defined on subsets of an arbitrary metric space, i.e. we define, for any metric space X ,

$$\underline{\dim}_{\mathbf{M}}(X) = \liminf_{\delta \rightarrow 0} \log_{1/\delta} N_{\delta}^*(X) \quad \text{and} \quad \overline{\dim}_{\mathbf{M}}(X) = \limsup_{\delta \rightarrow 0} \log_{1/\delta} N_{\delta}^*(X).$$

where N_{δ}^* is any of the particular quantities defined above.

If $f : X \rightarrow Y$ is a Lipschitz map between two metric spaces X and Y , then there exists a constant $c > 0$ such that for each $x \in X$ and $r > 0$, $f(B_r(x)) \subset B_{cr}(f(x))$. This can be used to show

$$\underline{\dim}_{\mathbf{M}}(f(X)) \leq \underline{\dim}_{\mathbf{M}}(X) \quad \text{and} \quad \overline{\dim}_{\mathbf{M}}(f(X)) \leq \overline{\dim}_{\mathbf{M}}(X).$$

For example, the graph of a Lipschitz function $f : \mathbf{R} \rightarrow \mathbf{R}$ has Minkowski dimension one. On the other hand, for maps that are not Lipschitz, the Minkowski dimension of a set can be boosted by taking an image under such a map.

Example. Let $S = \{(x, \sin(1/x)) : 0 < x \leq 1\}$. Then S has Minkowski dimension $3/2$. Consider a fixed scale δ . For any $x \in (0, 1)$, Let $f(x) = \sin(1/x)$. Then $|f'(x)| \leq 1/x^2$. If $x_0 \in (\delta, 1]$, then for any $x \in (x_0 - \delta, x_0 + \delta)$, the mean value theorem implies that

$$|f(x) - f(x_0)| \leq \delta(x_0 - \delta)^{-2}.$$

Thus the length of the vertical segment $S_{\delta} \cap \{x = x_0\}$ is $O(\delta(x_0 - \delta)^{-2})$. In particular, for each $\alpha \in (0, 1)$, we may cover $S_{\delta} \cap [0, 1] \times [-1, 1]$ by an initial

cube $[0, \delta^\alpha] \times [-1, 1]$ for $\alpha < 1$, and then an integral over the bound obtained for the lengths of the vertical segments. Thus an application of Fubini's theorem shows that

$$|S_\delta \cap [0, 1] \times [-1, 1]| \lesssim \delta^\alpha + \int_{\delta^\alpha}^1 \frac{2\delta}{(x - \delta)^\alpha} dx \lesssim \delta^\alpha + \delta^{1-\alpha}.$$

Choosing $\alpha = 1/2$ gives $|S_\delta| \lesssim \delta^{1/2}$. But S_δ certainly contains $[0, \delta^{1/2}] \times [-1, 1]$, which gives $|S_\delta| \gtrsim \delta^{1/2}$. In particular, taking limits shows this estimate is enough to conclude S has Minkowski dimension $3/2$.

The Minkowski dimension of a set is easy to calculate, and gives a powerful quantity describing the thickness of a set. However, it is not as stable with respect to many analytic operations, which is why the Hausdorff dimension, introduced in the next section, is often taken as the ‘canonical’ dimension of fractal sets. For instance, it is easy to see that for any sets $E_1, E_2 \subset \mathbf{R}^d$,

$$\overline{\dim}_{\mathbf{M}}(E_1 \cup E_2) = \max(\overline{\dim}_{\mathbf{M}}(E_1), \overline{\dim}_{\mathbf{M}}(E_2)).$$

On the other hand, if $\{E_n\}$ is a countable family of sets, it is *not* true that

$$\overline{\dim}_{\mathbf{M}}\left(\bigcup_{n=1}^{\infty} E_n\right) = \sup_n \overline{\dim}_{\mathbf{M}}(E_n).$$

For instance, any point in \mathbf{R} has Minkowski dimension zero, but \mathbf{Q} , which is a countable union of points, has Minkowski dimension 1. One way to fix this is to consider the packing dimension

$$\dim_{\mathbf{P}}(E) = \left\{ \sup_n \left(\overline{\dim}_{\mathbf{M}}(E_n) \right) : E \subset \bigcup_{n=1}^{\infty} E_n \right\}.$$

Then

$$\dim_{\mathbf{P}}\left(\bigcup_{n=1}^{\infty} E_n\right) = \sup_n (\dim_{\mathbf{P}}(E_n)).$$

Thus the packing dimension is countably stable.

1.2 Hausdorff Dimension

Hausdorff dimension is a more stable version of fractal dimension which is obtained by finding a canonical ‘ s dimensional measure’ H^s on \mathbf{R}^d for each s , and then setting the dimension of E to be the supremum of s such that $H^s(E) < \infty$. Since \mathbf{R}^d assigns mass $V_d r^d$ to each radius r ball, where $V_d = [\pi^{d/2}/\Gamma(1 + s/2)]$, a naive idea to obtain an ‘ s dimensional measure’ would be to assign a mass r^s to each radius r ball in \mathbf{R}^n , and then define

$$H_\infty^s(E) = V_s \cdot \inf \left\{ \sum_{n=1}^{\infty} r_n^s : E \subset \bigcup B_{r_n}(x_n) \right\},$$

where $V_s = \pi^{s/2}/\Gamma(1+s/2)$. This is an outer measure, and so Caratheodory’s extension theorem gives a σ algebra of measurable sets. Unfortunately, intervals aren’t even measurable with respect to this σ algebra, for non-integer values of s .

Example. Let $s = 1/2$, and let $E = (a, b)$. On one hand,

$$H_\infty^s(E) \leq V_{1/2} \left(\frac{b-a}{2} \right)^{1/2}.$$

On the other hand, if (a, b) is covered by a countable collection of balls $\{B_{r_n}(x_n)\}$, then $\sum_{n=1}^{\infty} (2r_n) \geq b - a$, so applying the concavity of $x \mapsto x^{1/2}$, we conclude

$$\sum_{n=1}^{\infty} r_n^{1/2} \geq \left(\sum_{n=1}^{\infty} r_n \right)^{1/2} \geq \left(\frac{b-a}{2} \right)^{1/2}$$

Thus we conclude that

$$H_\infty^s(E) \geq V_s \left(\frac{b-a}{2} \right)^{1/2}.$$

Thus we actually have equality here. But now we see that the additivity property begins to breakdown, since we have $H_\infty^{1/2}[0, 1] = 1$, whereas

$$H_\infty^{1/2}[0, 1/2] = H_\infty^{1/2}[1/2, 1] = 2^{-1/2},$$

and so

$$H_\infty^{1/2}[0, 1] < H_\infty^{1/2}[0, 1/2] + H_\infty^{1/2}[1/2, 1].$$

The reason why intervals fail to be measurable is that $[0, 1]$ is most efficiently coverable by a single large ball, rather than covering the set by the two intervals $[0, 1/2]$ and $[1/2, 1]$. We can fix this by limiting the Hausdorff measure to be the value of the most efficient cover by arbitrarily small balls, which as a consequence, will show all Borel sets are measurable with respect to this new measure.

For a set $E \subset \mathbf{R}^d$, we define

$$H_\delta^s(E) = V_s \cdot \inf \left\{ \sum_{n=1}^{\infty} r_n^s : E \subset \bigcup_{n=1}^{\infty} B_{r_n}(x_n) \text{ and } \sup(r_n) \leq \delta \right\}.$$

We then define $H^s(E) = \lim_{\delta \rightarrow 0} H_\delta^s(E)$. Then H^s is an exterior measure, and $H^s(E \cup F) = H^s(E) + H^s(F)$ if $d(E, F) > 0$. Thus all Borel sets are measurable with respect to H^s , which is certainly more satisfactory than the last definition.

Remark. Though not all sets are measurable with respect to H_∞^s . Nonetheless, since the values $H_\delta^s(E)$ increase to the value $H^s(E)$, if $H^s(E) = 0$, then $H_\delta^s(E) = 0$ for all $\delta > 0$. Thus $H_\infty^s(E) = 0$. Conversely, if $s > 0$ and $H_\infty^s(E) = 0$, then for each $\varepsilon > 0$, we can find a family of balls $\{B_{r_n}(x_n)\}$ covering E with

$$\sum_{n=1}^{\infty} r_n^s \leq \varepsilon V_s^{-1}.$$

In particular if $\varepsilon_0 = \varepsilon^{1/s} V_s^{-1/s}$, then $\sup_n(r_n) \leq \varepsilon_0$, so that $H_{\varepsilon_0}^s(E) \leq \varepsilon$. As $\varepsilon_0 \rightarrow 0$, $\varepsilon \rightarrow 0$, so

$$H^s(E) = \lim_{\delta \rightarrow 0} H_\delta^s(E) = \lim_{\varepsilon_0 \rightarrow 0} H_{\varepsilon_0}^s(E) \leq \lim_{\varepsilon_0 \rightarrow 0} \varepsilon = 0.$$

The result also holds in the case $s = 0$, since if $H_\infty^s(E) = 0$ then $E = \emptyset$, so $H^s(E) = 0$. Thus we conclude that, although the σ algebra of measurable sets with respect to H^s is not the same as the σ algebra of measurable sets with respect to H_∞^s , the null sets of H^s and H_∞^s do agree.

Example. Let $s = 0$. Then $H_\delta^0(E) = N_\delta^{\text{Ext}}(E)$, which tends to ∞ as $\delta \rightarrow 0$ unless E is finite. For finite sets $E \subset \mathbf{R}^d$,

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

Thus H^0 is just the counting measure. All sets are H^0 measurable.

Example. Let $s = d$. If E has Lebesgue measure zero, then for any $\varepsilon > 0$, there exists a countable collection of balls $\{B_{r_n}(x_n)\}$ covering E with

$$\sum_{n=1}^{\infty} r_n^d < \varepsilon.$$

Then $\sup_n(r_n) < \varepsilon^{1/d}$, so $H_{\varepsilon^{1/n}}^d(E) < \varepsilon$. Letting $\varepsilon \rightarrow 0$, we conclude $H^d(E) = 0$. Thus H^d is absolutely continuous with respect to the Lebesgue measure. The measure H^d is translation invariant, so H^d is actually a constant multiple of the Lebesgue measure. It is simple to verify that $H^d(B)$ is the same as the Lebesgue measure of B , where B is the unit ball, so H^d is actually equal to the Lebesgue measure.

The same proof strategy shows that if V is an m dimensional subspace of \mathbf{R}^n , then H^m , restricted to subsets of V , is equal to the m dimensional Lebesgue measure on V . More generally, in the next theorem we show H^m measures the m dimensional surface area of smooth, m dimensional submanifolds of \mathbf{R}^n .

Theorem 1.2. Let U be an open subset of \mathbf{R}^m , and let $\phi : U \rightarrow \mathbf{R}^d$ be a smooth, injective embedding map. Then for any compact set $K \subset U$,

$$H^m(\phi(K)) = \int_K J_\phi(x) dx$$

where $J_\phi(x)$ is the square root of the sums of squares of the $m \times m$ minors of $D\phi(x)$.

We leave out the proof, since we do not use this theorem for the basic theory, and we will later prove a very useful generalization known as the area formula.

Lemma 1.3. Consider $t < s$, and $E \subset \mathbf{R}^d$.

(i) If $H^t(E) < \infty$, then $H^s(E) = 0$.

(ii) If $H^s(E) \neq 0$, then $H^t(E) = \infty$.

Proof. Suppose that $H^t(E) = A < \infty$. Then for any $\delta > 0$, there is a cover of E by a collection of intervals $\{B_{r_n}(x_n)\}$, such that $\sup_n r_n \leq \delta$, and

$$\sum_{n=1}^{\infty} r_n^t \leq A + 1 < \infty.$$

But then

$$H_\delta^s(E) \leq \sum_{n=1}^{\infty} r_n^s \leq \sum_{n=1}^{\infty} r_n^{s-t} r_n^t \leq \delta^{s-t} (A+1).$$

As $\delta \rightarrow 0$, we conclude $H^s(E) = 0$, proving (i). And (ii) is just the contrapositive of (i), and therefore immediately follows. \square

Corollary 1.4. *If $s > d$ and $E \subset \mathbf{R}^d$, then $H^s(E) = 0$.*

Proof. The measure H^d is just the Lebesgue measure on \mathbf{R}^d , so

$$H^d[-N, N]^d = (2N)^d < \infty.$$

If $s > d$, the last lemma implies $H^s[-N, N]^d = 0$. By countable additivity, taking $N \rightarrow \infty$, we find $H^s(\mathbf{R}^d) = 0$. Since H^s is a positive measure, this implies $H^s(E) = 0$ for all sets E . \square

Given any Borel set E , the previous two results imply there is a unique value $s_0 \in [0, d]$ such that $H^s(E) = 0$ for $s > s_0$, and $H^s(E) = \infty$ for $0 \leq s < s_0$. We refer to s_0 as the *Hausdorff dimension* of E , denoted $\dim_{\mathbf{H}}(E)$. The countable stability of the measures H^s immediately implies that if $\{E_n\}$ is a countable family of sets, and $E = \bigcup_{n=1}^{\infty} E_n$, then

$$\dim_{\mathbf{H}}(E) = \sup_n \dim_{\mathbf{H}}(E_n).$$

This is not true for the Minkowski dimension; for instance, every point in \mathbf{R} has Minkowski dimension zero, but \mathbf{Q} , the countable union of such points, has Minkowski dimension one.

Example. Consider $S = \{(x, \sin(1/x)) : 0 < x \leq 1\}$. Then for each $\delta > 0$, the set $S \cap [\delta, 1] \times \mathbf{R}$ is the graph of a smooth curve, and therefore has Hausdorff dimension 1. Thus for any $\varepsilon > 0$, $H^{1+\varepsilon}(S \cap [\delta, 1] \times \mathbf{R}) = 0$. But then taking limits as $\delta \rightarrow 0$, we conclude $H^{1+\varepsilon}(S) = 0$. Since $H^1(S) > 0$, this shows S has Hausdorff dimension 1. Compare this to the Minkowski dimension $3/2$ result we obtained previously.

An easy way to compare the approaches to fractal dimension given by Minkowski and Hausdorff dimension is that Minkowski dimension measures the efficiency of covers of a set at a fixed scale asymptotically, whereas Hausdorff dimension measures the efficiency of covers of a set simultaneously at various, small scales.

Theorem 1.5. For any bounded set E , $\dim_{\mathbf{H}}(E) \leq \underline{\dim}_{\mathbf{M}}(E)$.

Proof. Given $\delta > 0$, we have a simple bound $H_{\delta}^s(E) \leq N_{\delta}^{\text{Ext}}(E) \cdot l^s$. If $\underline{\dim}_{\mathbf{M}}(E) < s$, then there exists a sequence $\{\delta_n\}$ with $\delta_n \rightarrow 0$, and $N_{\delta_n}^{\text{Ext}}(E) \leq \delta_n^{-s}$. Thus we conclude that

$$H^s(E) = \lim_{n \rightarrow \infty} H_{\delta_n}^s(E) \leq \lim_{n \rightarrow \infty} N_{\delta_n}^*(E) \cdot \delta_n^s \leq 1.$$

Thus $\dim_{\mathbf{H}}(E) \leq s$. Taking infima over $s > \underline{\dim}_{\mathbf{M}}(E)$ shows that $\dim_{\mathbf{H}}(E) \leq \underline{\dim}_{\mathbf{M}}(E)$, completing the proof. \square

1.3 Energy Integrals and Frostman's Lemma

By taking particular covers of a set, it is often easy to upper bound the Hausdorff dimension of a set. On the other hand, finding a lower bound is a little more tricky. One method to finding a lower bound is constructing a measure on our set with a certain 'measure' property. We say a finite Borel measure μ is a *Frostman measure* with dimension s if there exists a constant $C > 0$ such that $\mu(B_r(x)) \leq Cr^s$. For a set E , we let $M(E)$ denote all Borel measures with compact support in E . The proof of Frostman's lemma will utilize a technique often useful, known as the *mass distribution principle*.

Lemma 1.6. Suppose $\{\mu_i\}$ is a Cauchy sequence of non-negative, regular Borel measures on \mathbf{R}^d , in the sense that for any $f \in C_c(\mathbf{R}^d)$, the sequence

$$\left\{ \int f d\mu_i \right\}$$

is Cauchy. Then there is a regular Borel measure μ such that $\mu_i \rightarrow \mu$ vaguely, in the sense that for any $f \in C_c(\mathbf{R}^d)$,

$$\int f d\mu = \lim_{i \rightarrow \infty} \int f d\mu_i.$$

Proof. Fix a compact set K . Then we can find a function $\varphi \in C_c(\mathbf{R}^d)$ such that $\mathbf{I}_K \leq \varphi$. This means that

$$\mu_i(K) = \int \mathbf{I}_K d\mu_i \leq \int \varphi d\mu_i.$$

Since $\{\int \varphi d\mu_i\}$ is Cauchy, the values $\int \varphi d\mu_i$ are uniformly bounded in i . In particular, $\mu_i(K)$ is uniformly bounded in i . Applying the Banach Alaoglu theorem, we conclude that the collection of measures $\{\mu_i|_K\}$ is contained in a compact subset of the space of non-negative Borel measures with respect to the vague topology. But every compact subset of a locally convex space is complete, and so we can therefore find a finite Borel measure μ^K supported on K such that $\mu_i|_K \rightarrow \mu^K$ vaguely. If $f \in C_c(\mathbf{R}^d)$ is supported on $K_1 \cap K_2$, for two compact sets K_1 and K_2 , then

$$\int f d\mu^{K_1} = \lim_{i \rightarrow \infty} \int_{K_1} f d\mu_i = \lim_{i \rightarrow \infty} \int_{K_2} f d\mu_i = \int f d\mu^{K_2}.$$

Thus we can define a measure μ such that for each $f \in C_c(\mathbf{R}^d)$, if f is supported on a compact set K , then

$$\int f d\mu = \int f d\mu^K.$$

For any $f \in C_c(\mathbf{R}^d)$, f is supported on some compact set K , and then

$$\int f d\mu = \int f d\mu^K = \lim_{i \rightarrow \infty} \int_K f d\mu_i = \lim_{i \rightarrow \infty} \int f d\mu_i.$$

Since f was arbitrary, $\mu_i \rightarrow \mu$ in the vague topology. \square

For the next theorem, we work dyadically. Let \mathcal{Q}_n^d be the collection of all sidelength 2^{-n} cubes with corners in the lattice $2^{-n} \cdot \mathbf{Z}^d$, and let $\mathcal{Q}^d = \bigcup_{n=0}^{\infty} \mathcal{Q}_n^d$. For any cube $Q \in \mathcal{Q}_{n+1}^d$, there exists a unique cube $Q^* \in \mathcal{Q}_n^d$ with $Q \subset Q^*$, which we call the *parent* of Q . For a cube Q , we let $l(Q)$ denote its sidelength.

Theorem 1.7 (Mass Distribution Principle). *Let $w : \mathcal{Q}^d \rightarrow [0, \infty)$ be a function such that for any $Q_0 \in \mathcal{Q}^d$,*

$$\sum_{Q^*=Q_0} w(Q) = w(Q_0). \tag{1.1}$$

Then there exists a regular Borel measure μ supported on

$$\bigcap_{k=1}^{\infty} \left[\bigcup \{Q \in \mathcal{Q}_k^d : w(Q) > 0\} \right],$$

such that for each $Q \in \mathcal{Q}^d$,

$$\mu(Q^\circ) \leq w(Q) \leq \mu(Q), \quad (1.2)$$

and for any set E and $n \geq 0$, if $\mathcal{E}_n = \{Q \in \mathcal{Q}_n : Q \cap E_{1/2^n} \neq \emptyset\}$, then

$$\mu(E) \leq \sum_{Q \in \mathcal{E}_k} w(Q). \quad (1.3)$$

Proof. For each i , let μ_i be a regular Borel measure such that for each $j \leq i$, and $Q \in \mathcal{Q}_j^d$, $\mu_i(Q) = \mu_i(Q^\circ) = w(Q)$. One such choice is given, for each $\varphi \in C_c(\mathbf{R}^d)$, by the equation

$$\int \varphi d\mu_i = \sum_{Q \in \mathcal{Q}_i^d} \frac{w(Q)}{|Q|} \int_Q \varphi dx.$$

We claim that $\{\mu_i\}$ is a Cauchy sequence. Fix $\varphi \in C_c(\mathbf{R}^d)$, and choose some N such that φ is supported on $[-N, N]^d$. Set $Q' = [-N, N]^d$, and for each i , set $\mathcal{Q}'_i = \mathcal{Q}_i(Q')$. Since φ is compactly supported, φ is *uniformly continuous*, so for each $\varepsilon > 0$, if i is suitably large, there is a sequence of values $\{a_Q : Q \in \mathcal{Q}'_i(Q')\}$ such that if $x \in Q$, $|\varphi(x) - a_Q| \leq \varepsilon$. But this means

$$\sum_{Q \in \mathcal{Q}'_i} (a_Q - \varepsilon) \mathbf{I}_Q \leq \varphi \leq \sum_{Q \in \mathcal{Q}'_i} (a_Q + \varepsilon) \mathbf{I}_Q.$$

Thus for any $j \geq i$,

$$\int \varphi d\mu_j \leq \sum_{Q \in \mathcal{Q}'_i} (a_Q + \varepsilon) \mu_j(Q) = \sum_{Q \in \mathcal{Q}'_i} (a_Q + \varepsilon) w(Q)$$

and

$$\int \varphi d\mu_j \geq \sum_{Q \in \mathcal{Q}'_i} (a_Q - \varepsilon) \mu_j(Q) = \sum_{Q \in \mathcal{Q}'_i} (a_Q - \varepsilon) w(Q).$$

In particular, if $j, j' \geq i$,

$$\left| \int \varphi d\mu_j - \int \varphi d\mu_{j'} \right| \leq 2\varepsilon \sum_{Q \in \mathcal{Q}'_i} w(Q).$$

Repeated applications of (1.1) show that

$$\sum_{Q \in \mathcal{Q}'_i} w(Q) = \sum_{Q \in \mathcal{Q}'_0} w(Q),$$

which is therefore bounded independently of i . Since ε and φ were arbitrary, this shows $\{\mu_i\}$ is Cauchy.

Applying Lemma 1.6, we find a regular Borel measure μ such that $\mu_i \rightarrow \mu$ vaguely. For any $Q \in \mathcal{Q}^d$, since Q is a closed set,

$$\mu(Q) \geq \limsup_{i \rightarrow \infty} \mu_i(Q) = w(Q),$$

and since Q° is an open set,

$$\mu(Q^\circ) \leq \liminf_{i \rightarrow \infty} \mu_i(Q^\circ) = w(Q).$$

This establishes (1.2). Conversely, if E is arbitrary, then $E \subset E(l_i)^\circ$ for any i , so if $\mathcal{E}_i = \mathcal{Q}_i(E(l_i))$, then

$$\mu(E) \leq \liminf_{i \rightarrow \infty} \mu_i(E(l_i)^\circ) \leq \liminf_{i \rightarrow \infty} \mu_i(E(l_i)) = \sum_{Q \in \mathcal{E}_i} w(Q).$$

This establishes (1.3). □

Remark. The reason why we cannot necessarily find a function μ which *precisely* extends w is that in the weak limit, mass which the weight function w assigns to one cube can ‘leak’ into the mass of adjacent cubes when we take a weak limit. This is why we must use the weaker ‘extension bounds’ (1.2) and (1.3). This doesn’t cause us to ‘gain’ or ‘lose’ any mass in the weak limit, as (1.2) shows, so the result is still a ‘mass distribution’ result. It just means that the mass specified by w may be shared by adjacent cubes in the weak limit, if enough mass is pushed out to the boundary of these cubes. If, for each $Q \in \mathcal{Q}^d$,

$$\lim_{k \rightarrow \infty} \sum_{Q \in \mathcal{Q}'_k} w(Q) = 0,$$

where $\mathcal{Q}'_k = \{Q \in \mathcal{Q}_k : (Q \cap ((Q_0)_{1/2^n} - Q_0) \neq \emptyset)\}$, then (1.2) and (1.3) together imply we actually have $\mu(Q) = w(Q)$ for all $Q \in \mathcal{Q}^d$. Thus μ is a measure extending w . One such condition that guarantees this is that $w(Q) \lesssim l(Q)^s$ for some $s > d - 1$. Another is that for any $k \geq 0$, and distinct $Q_0, Q_1 \in \mathcal{Q}_k^d$ with $Q_0 \cap Q_1 \neq \emptyset$, either $w(Q_0) = 0$ or $w(Q_1) = 0$.

Theorem 1.8 (Frostman's Lemma). *If E is a Borel set, $H^s(E) > 0$ if and only if there exists an s dimensional Frostman measure supported on E .*

Proof. Suppose that μ is s dimensional and supported on E . If $F \subset \mathbf{R}^d$ and $H^s(F) = 0$, then for each $\varepsilon > 0$ there is a sequence of balls $\{B_{r_n}(x_n)\}$ whose union covers F , with

$$\sum_{n=1}^{\infty} r_n^s \leq \varepsilon.$$

But then

$$\mu(F) \leq \sum_{n=1}^{\infty} \mu(B_{r_n}(x_n)) \lesssim \sum_{n=1}^{\infty} r_n^s \leq \varepsilon.$$

Taking $\varepsilon \rightarrow 0$, we conclude $\mu(F) = 0$. Thus μ is absolutely continuous with respect to H^s . Since $\mu(E) > 0$, this means that $H^s(E) > 0$.

To prove the converse, we will suppose for simplicity that E is compact. We work dyadically with the family of dyadic cubes \mathcal{Q}^d . By translating, we may assume that $H^s(E \cap [0, 1]^d) > 0$, and so without loss of generality we may assume $E \subset [0, 1]^d$. For each $Q \in \mathcal{Q}^d$, define $w^+(Q) = H_\infty^s(E \cap Q)$. Then

$$w^+(Q) \leq l(Q)^s, \quad (1.4)$$

Moreover, w^+ is subadditive. We now recursively define a function w such that for any $Q \in \mathcal{Q}^d$, (1.1) holds, and

$$w(Q) \leq w^+(Q), \quad (1.5)$$

is satisfied at each stage of the definition of w . We initially define w by setting $w([0, 1]^d) = w^+([0, 1]^d)$. Given $Q \in \mathcal{Q}_k^d$, we enumerate its children as $Q_1, \dots, Q_M \in \mathcal{Q}_{k+1}^d$. We then consider any values $A_1, \dots, A_M \geq 0$ such that

$$A_1 + \dots + A_M = w(Q), \quad (1.6)$$

and for each $n \in \{1, \dots, M\}$,

$$A_n \leq w(Q_n). \quad (1.7)$$

This is feasible to do because $w^+(Q_1) + \dots + w^+(Q_M) \geq w^+(Q)$, and (1.5) holds for parent cubes by induction. We then define $w(Q_n) = A_n$ for each n . Equation (1.6) implies (1.1) holds for this new set of definitions, and

(1.7) implies (1.5) holds. Thus w is a well defined function. Furthermore, (1.6) implies (1.1) of Lemma 1.7, and so the mass distribution principle gives the existence of a measure μ supported on E , satisfying (1.3) and (1.2). In particular, (1.2) implies μ is non-zero. For each ball B of radius $1/2^k$, there are at most $O_d(1)$ cubes in \mathcal{Q}_k intersecting B , and so $\mu(B) \lesssim_d 2^{-sk}$. But now for any $r \in (0, 1)$, and any ball B of radius r , if $1/2^{k+1} \leq r \leq 1/2^k$, then

$$\mu(B) \lesssim_d 2^{-sk} = 2^s 2^{-s(k+1)} \leq 2^s r^{-s}.$$

Thus μ is a Frostman measure of dimension s . \square

Let us consider a simple corollary.

Corollary 1.9. *Let $E \subset \mathbf{R}^n$ and $F \subset \mathbf{R}^m$ be Borel sets. Then $\dim_{\mathbf{H}}(E \times F) \geq \dim_{\mathbf{H}}(E) + \dim_{\mathbf{H}}(F)$.*

Proof. Fix $s_1 < \dim_{\mathbf{H}}(E)$ and $s_2 < \dim_{\mathbf{H}}(F)$. Then we can find a measure μ compactly supported on E and ν compactly supported on F such that for all $r > 0$, $x_1 \in \mathbf{R}^n$ and $x_2 \in \mathbf{R}^m$, $\mu(B_r(x_1)) \lesssim r^{s_1}$ and $\nu(B_r(x_2)) \lesssim r^{s_2}$. Then $B_r((x_1, x_2)) \subset B_r(x_1) \times B_r(x_2)$, so

$$(\mu \times \nu)(B_r((x_1, x_2))) \leq \mu(B_r(x_1))\nu(B_r(x_2)) \lesssim r^{s_1+s_2}.$$

Thus $\dim_{\mathbf{H}}(E \times F) \geq s_1 + s_2$. \square

A fundamental concept in understanding the structure of Frostman measures is to consider the *energy of a measure*. Given a Borel measure μ on \mathbf{R}^d and $s \in [0, d]$, we set

$$I_s(\mu) = \int_{\mathbf{R}^d} \int_{\mathbf{R}^d} |x - y|^{-s} d\mu(x) d\mu(y) = \int_{\mathbf{R}^d} (k_s * \mu) d\mu,$$

where $k_s(x) = |x|^{-s}$ for $x \in \mathbf{R}^d$ is the *Riesz kernel*. If μ is a Frostman measure of dimension s_0 , and $0 < s < s_0$. Then a dyadic decomposition shows that for each $x \in \mathbf{R}^d$,

$$|(k_s * \mu)(x)| \leq \sum_{n=-\infty}^{\infty} \int_{2^{n-1} \leq |y| \leq 2^n} k_s(y) d\mu(x - y) dy.$$

If we define $B_n(x) = \{y \in \mathbf{R}^d : 2^{n-1} \leq |x - y| \leq 2^n\}$, then for $n \geq 0$, we have

$$\int_{2^{n-1} \leq |x| \leq 2^n} k_s(y) d\mu(x - y) dy \lesssim_s 2^{-ns} \mu(B_n(x)) \lesssim_{\mu} 2^{-ns}.$$

and

$$\int_{1/2^{n+1} \leq |x| \leq 1/2^n} k_s(y) d\mu(x-y) dy \lesssim_s 2^{ns} \mu(B_n(x)) \lesssim_\mu 2^{(s-s_0)}.$$

Thus we conclude that

$$\|k_s * \mu\|_{L^\infty(\mathbf{R}^d)} \lesssim_{\mu,s} 1 < \infty.$$

But now we can apply Hölder's inequality, so we conclude that

$$I_s(\mu) \leq \|k_s * \mu\|_{L^\infty(\mathbf{R}^d)} \|\mu\| < \infty.$$

Conversely, if μ is a compactly supported Borel measure with $I_s(\mu) < \infty$, then for μ almost every $x \in \mathbf{R}^d$, $(k_s * \mu)(x) < \infty$. Thus we may find a large value R such that if E is the set of $x \in \mathbf{R}^d$ such that $(k_s * \mu)(x) \leq R$, then $\mu(E) > 0$. Since

$$(k_s * \mu)(x) \geq r^s \mu(B_r(x)),$$

it follows that for $x \in E$, and any $r \geq 0$, $\mu(B_r(x)) \leq Mr^{-s}$. Thus if we define $\nu = \mu|_E$, then ν is a Frostman measure of dimension s .

Theorem 1.10. *For any Borel set E ,*

$$\dim_{\mathbf{H}}(E) = \sup \left\{ s \in [0, d] : \begin{array}{l} \text{there is a Borel measure } \mu \text{ compactly} \\ \text{supported on } E \text{ such that } I_s(\mu) < \infty \end{array} \right\}$$

1.4 Fourier Transforms Salem Sets

Powerful techniques are obtained in geometric measure theory by applying the Fourier transform. The prime source of the connection between geometric measure theory and the Fourier transform is that the energy integral $I_s(\mu)$, for a compactly supported Borel measure μ on \mathbf{R}^d , is defined in terms of a convolution of μ with the Riesz kernel k_s . Naively using the unitary nature of the Fourier transform, and the fact that μ is a positive

measure, we might expect that

$$\begin{aligned}
I_s(\mu) &= \int_{\mathbf{R}^d} (k_s * \mu) \cdot \mu \\
&= \int_{\mathbf{R}^d} \widehat{k_s} \cdot \widehat{\mu} \cdot \overline{\widehat{\mu}} \\
&= \int_{\mathbf{R}^d} \widehat{k_s}(\xi) \widehat{\mu}(\xi) \overline{\widehat{\mu}(-\xi)} d\xi \\
&= \int_{\mathbf{R}^d} \widehat{k_s}(\xi) |\widehat{\mu}(\xi)|^2 d\xi.
\end{aligned}$$

We will show that $\widehat{k_s} = \gamma(d, s) \cdot k_{d-s}$, when we interpret the Fourier transform of k_s in the distributional sense.

Lemma 1.11. *In a distributional sense, for $s \in (0, d)$, there exists $\gamma(d, s)$ such that $\widehat{k_s} = \gamma(d, s) k_{d-s}$.*

Proof. It is easy to see the k_s is a tempered distribution, since k_s is integrable about the origin, i.e. $k_s \mathbf{I}_{|x| \leq 1} \in L^1(\mathbf{R}^d)$, and $k_s(1 - \mathbf{I}_{|x| \leq 1}) \in L^{2d/s}(\mathbf{R}^d)$. Thus $k_s \in L^{1,\infty}(\mathbf{R}^d) + L^{2d/s}(\mathbf{R}^d)$, and is therefore tempered. Now since k_s is a radial distribution, so too is $\widehat{k_s}$. But k_s is also a distribution which is homogenous of degree $d - s$, so it must be a multiple of k_{d-s} . \square

Theorem 1.12. *For any s , there exists $\gamma(d, s)$ such that for any positive, finite, compactly supported Borel measure μ ,*

$$I_s(\mu) = \gamma(d, s) \int |\widehat{\mu}(\xi)|^2 |\xi|^{s-d} d\xi.$$

Proof. We have already argued this theorem is true when μ is absolutely continuous with smooth distribution function. But now if we replace μ with $\mu_\varepsilon = \phi_\varepsilon * \mu$, where $\{\phi_\varepsilon\}$ is a smooth mollifier, then

$$I_s(\mu_\varepsilon) = \gamma(d, s) \int_{\mathbf{R}^d} |\widehat{\phi_\varepsilon}(\xi)|^2 \cdot k_{d-s}(\xi) |\widehat{\mu}(\xi)|^2 d\xi.$$

But

$$\begin{aligned}
I_s(\mu_\varepsilon) &= \int_{\mathbf{R}^d} \int_{\mathbf{R}^d} \left(\int_{\mathbf{R}^d} \int_{\mathbf{R}^d} \phi_\varepsilon(x - z_1) \phi_\varepsilon(y - z_2) |x - y|^{-s} dx dy \right) d\mu(z_1) d\mu(z_2) \\
&= \int_{\mathbf{R}^d} \int_{\mathbf{R}^d} \left(\int_{\mathbf{R}^d} \int_{\mathbf{R}^d} \phi(u) \phi(v) |\varepsilon(u - v) + z_1 - z_2|^{-s} du dv \right) d\mu(z_1) d\mu(z_2).
\end{aligned}$$

As $\varepsilon \rightarrow 0$, for $z_1 \neq z_2$,

$$\int_{\mathbf{R}^d} \int_{\mathbf{R}^d} \phi(u) \phi(v) |\varepsilon(u-v) + z_1 - z_2|^{-s} du dv \rightarrow |z_1 - z_2|^{-s}.$$

On the other hand, if $z_1 - z_2 = w$ and we make a change of variables $s = u - v$, then

$$\begin{aligned} \int_{\mathbf{R}^d} \int_{\mathbf{R}^d} \phi(u) \phi(v) |\varepsilon(u-v) + z_1 - z_2|^{-s} du dv \\ \leq \|\phi\|_{L^\infty(\mathbf{R}^d)}^2 \int_{|u| \leq 1} \int_{|v| \leq 1} |\varepsilon(u-v) + z_1 - z_2|^{-s} du dv \\ \lesssim_\phi \int_{|s| \leq 2} |\varepsilon s + w|^{-s}. \end{aligned}$$

If $\varepsilon \leq |w|/4$, then

$$\int_{|s| \leq 2} |\varepsilon s + w|^{-s} \lesssim |w|^{-s}.$$

On the other hand, a simple dyadic decomposition verifies that for $\varepsilon \geq |w|/4$, then

$$\int_{|s| \leq 2} |\varepsilon s + w|^{-s} \lesssim_d |w|^{-s}.$$

Thus, returning to our original situation, we conclude that

$$\int_{\mathbf{R}^d} \int_{\mathbf{R}^d} \phi(u) \phi(v) |\varepsilon(u-v) + z_1 - z_2|^{-s} du dv \lesssim_d |z_1 - z_2|^{-s}.$$

If $I_s(\mu) < \infty$, then $I_s(\mu_\varepsilon)$ is uniformly integrable, so the dominated convergence theorem implies that

$$I_s(\mu) = \lim_{\varepsilon \rightarrow 0} I_s(\mu_\varepsilon) = \lim_{\varepsilon \rightarrow 0} \gamma(d, s) \int_{\mathbf{R}^d} |\widehat{\phi}_\varepsilon(\xi)|^2 \cdot k_{d-s}(\xi) |\widehat{\mu}(\xi)|^2 d\xi.$$

The monotone convergence theorem implies that

$$\lim_{\varepsilon \rightarrow 0} \gamma(d, s) \int_{\mathbf{R}^d} |\widehat{\phi}_\varepsilon(\xi)|^2 \cdot k_{d-s}(\xi) |\widehat{\mu}(\xi)|^2 d\xi = \gamma(d, s) \int_{\mathbf{R}^d} k_{d-s}(\xi) |\widehat{\mu}(\xi)|^2 d\xi$$

completing the proof in this case. If $I_s(\mu) = \infty$, then Fatou's lemma and monotone convergence implies that

$$\gamma(d, s) \int_{\mathbf{R}^d} k_{d-s}(\xi) |\widehat{\mu}(\xi)|^2 d\xi = \lim_{\varepsilon \rightarrow 0} I_s(\mu_\varepsilon) \geq I_s(\mu) = \infty. \quad \square$$

We therefore conclude that if μ is a Frostman measure of dimension s , then $|\hat{\mu}(\xi)| \lesssim |\xi|^{-s/2}$ for *most* frequencies ξ . However, the fuzzyness of this estimate prevents us from precisely getting a uniform bound

$$|\hat{\mu}(\xi)| \lesssim |\xi|^{-s/2}.$$

If μ is a compactly supported positive Borel measure, then we say μ has *Fourier dimension* s if there exists a constant $C > 0$ such that for all $\xi \in \mathbf{R}^d$,

$$|\hat{\mu}(\xi)| \leq C|\xi|^{-s/2}.$$

Clearly any such measure is a Frostman measure of dimension t for each $t < s$. Thus if we define the Fourier dimension of a Borel set E to be

$$\dim_{\mathbf{F}}(E) = \sup \left\{ s \in [0, d] : \begin{array}{l} \text{There is } \mu \text{ compactly supported on } E \\ \text{such that } \sup_{\xi \in \mathbf{R}^d} |\hat{\mu}(\xi)| |\xi|^{s/2} < \infty \end{array} \right\},$$

then $\dim_{\mathbf{F}}(E) \leq \dim_{\mathbf{H}}(E)$. If $\dim_{\mathbf{F}}(E) = \dim_{\mathbf{H}}(E)$, we say E is a *Salem set*. Often random families of sets are ‘generically’ Salem, but many particular examples of fractals are not Salem.

Example. Let C denote the standard middle thirds Cantor set. Then $\dim_{\mathbf{H}}(C) = \log_3(2)$. However, $\dim_{\mathbf{F}}(C) = 0$, so C is certainly not a Salem set. We rely on a useful fact about C ; for each $x \in C$, and each integer $n \geq 0$, $\{3^n x\} \notin (1/3, 2/3)$, where $\{\cdot\} : \mathbf{R} \rightarrow [0, 1)$ is a map giving the fractional part of any real number. This can be seen from the fact that $C = (C/3) \cup (C/3 + 2/3)$, and that $(1/3, 2/3) \cap C = \emptyset$. Now let ϕ be a smooth, nonnegative bump function compactly supported on $(1/3, 2/3)$ with $\int \phi(x) dx = 1$ and such that for each n , let ϕ_n be the function compactly supported on $[0, 1]$ such that for each $x \in [0, 1]$, $\phi_n(x) = \phi(\{3^n x\})$. By the Fourier inversion formula for Fourier series,

$$\phi_n(x) = \sum_{\xi \in \mathbf{Z}} \hat{\phi}(\xi) e^{2\pi i 3^n \xi x}.$$

For each n ,

$$\int_0^1 \phi_n(x) d\mu(x) = 0,$$

so Parseval’s formula implies that

$$0 = \int_0^1 \phi_n(x) d\mu(x) = \sum_{\xi \in \mathbf{Z}} \hat{\phi}(\xi) \overline{\hat{\mu}(3^n \xi)}.$$

Now suppose $\widehat{\mu}(\xi) \rightarrow 0$ as $|\xi| \rightarrow \infty$. Then for each $\varepsilon > 0$, there exists n such that for each $\xi \neq 0$, $|\widehat{\mu}(3^n \xi)| \leq \varepsilon$. But this means that

$$\begin{aligned} 0 &= \sum_{\xi \in \mathbf{Z}} \widehat{\phi}(\xi) \overline{\widehat{\mu}(3^n \xi)} = |\widehat{\phi}(0)| |\widehat{\mu}(0)| - \sum_{\xi \neq 0} |\widehat{\phi}(\xi)| |\widehat{\mu}(3^n \xi)| \\ &\geq 1 - \varepsilon \sum_{\xi \neq 0} |\widehat{\phi}(\xi)|. \end{aligned}$$

Taking $\varepsilon \rightarrow 0$ gives a contradiction, which shows that for any probability measure μ supported on C ,

$$\limsup_{\xi \in \mathbf{Z}} |\widehat{\mu}(\xi)| > 0.$$

Thus $\dim_{\mathbf{F}}(C) = 0$.

Unlike pointwise estimates, we can obtain decay estimates for *spherical averages* of the Fourier transform on spheres about the origin. Given a measure μ supported on \mathbf{R}^d , define

$$S_{\mu}(r) = \int_{S^{d-1}} |\widehat{\mu}(\xi s)|^2 d\sigma(s),$$

where σ is the surface measure on the unit sphere.

Theorem 1.13. *Let μ be a compactly supported Borel probability measure, and let $0 \leq s \leq (d-1)/2$. Then*

$$S_{\mu}(r) \lesssim_{n,s} I_s(\mu) r^{-s}.$$

Proof. A change of variables shows that

$$S_{\mu}(r) = \frac{1}{r^{d-1}} \int_{rS^{d-1}} |\widehat{\mu}(\xi)|^2 d\sigma_r(\xi),$$

where σ_r is the surface measure on rS^{d-1} . If f is a smooth, compactly supported function, then the product formula for the Fourier transform shows that

$$S_f(r) = \frac{1}{r^{d-1}} \int (f * \widehat{\sigma}_r)(x) f(x) dx.$$

We note that $\widehat{\sigma}_r(x) = r^{d-1} \widehat{\sigma}(rx)$, so

$$S_f(r) = \int \int \widehat{\sigma}(r(x-y)) f(x) f(y) dx dy.$$

Now the standard decay bound for the spherical measure (established by Bessel function asymptotics or oscillatory integrals) shows that for $r|x - y| \geq 1$,

$$|\widehat{\sigma}(r(x - y))| \lesssim_d (r|x - y|)^{-(d-1)/2} \leq (r|x - y|)^{-s},$$

and trivially, for $r|x - y| \leq 1$,

$$|\widehat{\sigma}(r(x - y))| \leq 1 \leq (r|x - y|)^{-s}.$$

Thus

$$S_f(r) \lesssim_d r^{-s} \iint |x - y|^{-s} f(x) f(y) dx dy \lesssim_{d,s} r^{-s} I_s(f).$$

Thus the theorem is proved in this case. Given any compactly supported measure μ and any a mollifier $\{\phi_\varepsilon\}$, we have

$$S_{\mu * \phi_\varepsilon}(r) \lesssim_{d,s} r^{-s} I_s(\mu * \phi_\varepsilon).$$

As we have seen, if $I_s(\mu) < \infty$, then $I_s(\mu * \phi_\varepsilon) \rightarrow I_s(\mu)$ as $\varepsilon \rightarrow 0$. And since $\widehat{\phi_\varepsilon}(x) = \widehat{\phi}(\varepsilon x)$, $\widehat{\phi}(0) = 1$, and $\widehat{\phi}$ is continuous at the origin,

$$\lim_{\varepsilon \rightarrow 0} \|\widehat{\mu} - \widehat{\mu * \phi_\varepsilon}\|_{L^\infty(S_r)} \leq \lim_{\varepsilon \rightarrow 0} \|1 - \widehat{\phi_\varepsilon}\|_{L^\infty(S_r)} = 0.$$

Thus $S_{\mu * \phi_\varepsilon}(r) \rightarrow S_\mu(r)$, completing the proof in general. \square

Remark. This result clearly cannot hold for all signed measures, because the identity

$$\int_{S^{d-1}} |\widehat{f}(\xi)|^2 d\sigma(\xi) \lesssim_{d,s} \int_{\mathbf{R}^d} |\widehat{f}(x)|^2 |x|^{s-n} dx$$

It is much easier to control averages over balls. For instance, it is immediate that

$$\int_{B_R(0)} |\widehat{\mu}(\xi)|^2 d\xi \leq R^{d-s} \int_{B_R(0)} |\xi|^{s-d} |\widehat{\mu}(\xi)|^2 d\xi \lesssim_{d,s} R^{d-s} I_s(\mu).$$

We can even obtain this directly from a Frostman estimate $\mu(B_r(x)) \lesssim r^s$, which does not necessarily imply that $I_s(\mu) < \infty$.

Theorem 1.14. *Suppose that there is a constant C and a compact Borel measure μ such that for all $x \in \mathbf{R}^d$ and $r > 0$, $\mu(B_r(x)) \leq Cr^s$. Then for all $R > 0$,*

$$\int_{B_R(0)} |\widehat{\mu}(\xi)|^2 d\xi \lesssim_{d,s,C} R^{d-s}.$$

Proof. Choose a non-negative Schwartz function ϕ with $\widehat{\phi(x)} = 1$ for $|x| \leq 1$. Then for each $R > 0$, we can apply the product formula if μ has a smooth density f , to conclude that

$$\begin{aligned} \int_{B(0,R)} |\widehat{f}(\xi)|^2 d\xi &\leq \int \widehat{\phi_R}(\xi) |\widehat{f}(\xi)|^2 d\xi \\ &= \int (\phi_R * f)(x) f(x) dx. \end{aligned}$$

Now

$$\begin{aligned} (\phi_R * f)(x) &= R^n \int \phi(R(x-y)) f(y) dy \\ &\lesssim R^d \int_{B_{1/R}(x)} f(y) dy + R^d \sum_{n=1}^{\infty} 2^{-n(s+1)} \int_{B_{2^n/R}(x)} f(y) dy \\ &\lesssim R^{d-s} + R^d \sum_{n=1}^{\infty} 2^{-n} R^{-s} \lesssim R^{d-s}. \end{aligned}$$

Thus we conclude

$$\int_{B(0,R)} |\widehat{f}(\xi)|^2 d\xi \lesssim R^{d-s}.$$

□

Remark. A very similar proof shows that if there exists $c > 0$ such that $\mu(B_r(x)) \geq cr^s$ for all x in the support of μ and $0 < r < 1$, then for $R > 1$,

$$\int_{B(0,R)} |\widehat{\mu}(\xi)|^2 d\xi \gtrsim R^{d-s}.$$

In particular, if μ is *Ahlfors-David regular* of dimension s , in the sense that $\mu(B_r(x)) \sim r^s$ for all $0 < r < 1$ and x in the support of μ , then

$$\int_{B(0,R)} |\widehat{\mu}(\xi)|^2 d\xi \sim R^{d-s}.$$

Chapter 2

Projection Theorems

In this chapter, we study the dimensional properties of orthogonal projections of sets in \mathbf{R}^d ; we begin with the simplest scenario, projection onto a line. In coordinates, the projection can be modelled by selecting $e \in S^{d-1}$, and studying the map $P_e : \mathbf{R}^d \rightarrow \mathbf{R}$ by setting $P_e(x) = e \cdot x$. For a given set E , we want to study the relation between the dimensions of E and the dimensions of its projections.

Since P_e is Lipschitz, it is easy to see that for any Borel set E , $P_e(E)$ has Hausdorff / Minkowski dimension at most the Hausdorff / Minkowski dimension of E . The Fourier dimension behaves in quite the opposite way. If μ is a probability measure compactly supported on E , then we can define a probability measure μ_e compactly supported on $P(E)$, such that

$$\int f \, d\mu_e = \int (f \circ P_e) \, d\mu.$$

For each $\xi \in \mathbf{R}$, we find that

$$\widehat{\mu_e}(\xi) = \int_{\mathbf{R}} e^{-2\pi i \xi t} d\mu_e(t) = \int_{\mathbf{R}} e^{-2\pi i \xi (e \cdot x)} d\mu(x) = \widehat{\mu}(\xi e).$$

This means that for each $s > 0$,

$$\sup_{\xi \in \mathbf{R}} |\xi|^s |\widehat{\mu_e}(\xi)| = \sup_{\xi \in \mathbf{R}} |\xi e|^s |\widehat{\mu}(\xi e)| \leq \sup_{\eta \in \mathbf{R}^d} |\eta|^s |\widehat{\mu}(\eta)|.$$

These calculations imply that $\dim_{\mathbf{F}}(P_e(E)) \geq \dim_{\mathbf{F}}(E)$. It is certainly possible for the projection of a set to increase the Fourier dimension, or decrease the Hausdorff and Minkowski dimension. However, for Salem sets, the upper and lower bounds play off each other quite nicely.

Theorem 2.1. *If E is a Salem set, then $P_e(E)$ is a Salem set, and $\dim(P_e(E)) = \min(1, \dim(E))$. Furthermore,*

- *If $\dim(E) > 1$, then $P_e(E)$ has positive Lebesgue measure.*
- *If $\dim(E) > 2$, then $P_e(E)$ has nonempty interior.*

Proof. Our calculations show that

$$\min(1, \dim(E)) \leq \dim_{\mathbf{F}}(P_e(E)) \leq \dim_{\mathbf{H}}(P_e(E)) \leq \min(1, \dim(E)).$$

Thus $\dim_{\mathbf{F}}(P_e(E)) = \dim_{\mathbf{H}}(P_e(E)) = \min(1, \dim(E))$.

If $\dim_{\mathbf{F}}(E) > 1$, then our calculations actually show that we can find a probability measure μ compactly supported on $P_e(E)$ such that $|\hat{\mu}(\xi)| \lesssim |\xi|^{-s/2}$ for some $s > 1$. This implies that $\hat{\mu} \in L^2(\mathbf{R}^d)$, so the inversion formula implies that $\mu \in L^2(\mathbf{R}^d)$. Thus the support of μ must have positive Lebesgue measure.

If $\dim_{\mathbf{F}}(E) > 2$, our calculations actually show that we can find a probability measure μ compactly supported on $P_e(E)$ such that $|\hat{\mu}(\xi)| \lesssim |\xi|^{-s/2}$, for $s > 2$. But this implies $\hat{\mu}$ is an integrable function, hence μ has a continuous probability density, implying $P_e(E)$ has non-empty interior. \square

Remark. This result can be used to give a simple proof that certain sets are not Salem sets. If C is the Cantor set, then $C \times C \subset \mathbf{R}^2$ is a set with Hausdorff dimension $2\log_3(2) > 1$, known as the *Cantor dust*. But if $P : \mathbf{R}^2 \rightarrow \mathbf{R}$ is given by $P(x, y) = x$, then $P(C \times C) = C$. Since $\dim(C) \neq 1$, $C \times C$ cannot be a Salem set.

The above theorem certainly doesn't apply for non Salem sets. However, recall that sets behave like Salem sets 'generically'. This doesn't really help us looking at the projections of any *single* set E . But a sneaky way of introducing some 'generic' properties is to consider a 'generic' projection. Thus we might expect that for a *particular* set, a generic projection behaves like a projection of a Salem set.

Theorem 2.2. *Let $E \subset \mathbf{R}^d$ be a Borel set, and let $s = \dim_{\mathbf{H}}(E)$. If $s \leq 1$, then $\dim_{\mathbf{H}}(P_e(E)) = s$ for almost every $e \in S^{d-1}$. If $s > 1$, then $P_e(E)$ has positive Lebesgue measure for almost every $e \in S^{d-1}$. And if $s > 2$, then $P_e(E)$ has nonempty interior for almost every $e \in S^{d-1}$.*

Proof. Fix $t < s$. Then we can find a measure μ compactly supported on E with $I_t(\mu) < \infty$. Now working in spherical coordinates,

$$\begin{aligned} \int_{S^{d-1}} I_t(\mu_e) d\sigma(e) &= \int_{S^{d-1}} \int_{\mathbf{R}} |\widehat{\mu}_e(\xi)|^2 |\xi|^{t-1} d\xi d\sigma(e) \\ &= 2 \int_{S^{d-1}} \int_0^\infty |\widehat{\mu}(\xi e)|^2 |\xi|^{t-1} d\xi d\sigma(e) \\ &\sim_d \int_{\mathbf{R}^d} |\widehat{\mu}(v)|^2 |v|^{t-d} dv = I_t(\mu) < \infty. \end{aligned}$$

Since this integral is finite, the integrand must be finite almost everywhere. Thus $I_t(\mu_e) < \infty$ for almost every e , and, since t was arbitrary, this shows that $\dim_{\mathbf{H}}(P_e(E)) \geq s$ for almost every e . Since P^e is always Lipschitz, we know $\dim_{\mathbf{H}}(P_e(E)) \leq s$ for all e , so we obtain equality.

If $s > 1$, pick $1 < t < s$. Then we can find a measure μ compactly supported on E such that for almost every $e \in S^{d-1}$, $I_1(\mu_e) < \infty$. But since

$$I_1(\mu_e) = \|\widehat{\mu}_e\|_{L^2(\mathbf{R}^d)},$$

the Fourier inversion formula implies that $\mu_e \in L^2(\mathbf{R}^d)$, so its support has positive Lebesgue measure.

If $s > 2$, we pick $t > 0$ such that $2 < 2 + t < s$, and conclude that for almost every $e \in S^{d-1}$,

$$\int |\widehat{\mu}_e(\xi)|^2 |\xi|^{1+t} d\xi < \infty.$$

Since $\|\widehat{\mu}_e\|_{L^\infty(\mathbf{R}^d)} \leq 1$, Hölder's inequality implies that $\widehat{\mu}_e \in L^1(\mathbf{R}^d)$, so that the Fourier inversion theorem implies that μ_e is continuous. Thus the support of μ_e must have nonempty interior. \square

Another consequence is a quantitative estimate for the average Lebesgue measure of the projected sets.

Theorem 2.3. *If $E \in \mathbf{R}^d$ is Lebesgue measurable, μ is a probability measure compactly supported on E with $I_1(\mu) < \infty$, then*

$$\int |P_e(A)| d\sigma(e) \gtrsim_d \frac{1}{I_1(\mu)}.$$

Proof. We see that $\hat{\mu}_e \in L^2(\mathbf{R}^d)$ for almost every $e \in S^{d-1}$, so $\mu_e \in L^2(\mathbf{R}^d)$, and Parseval's inequality implies

$$\|\mu_e\|_{L^2(\mathbf{R}^d)} = \|\hat{\mu}_e\|_{L^2(\mathbf{R}^d)} \sim_d I_1(\mu_e).$$

Cauchy Schwartz implies that

$$1 = \|\mu_e\|_{L^1(\mathbf{R}^d)}^2 \leq |P_e(E)| \|\mu_e\|_{L^2(\mathbf{R}^d)}^2.$$

Thus we conclude that

$$\int_{S^{d-1}} \frac{1}{|P_e(E)|} d\sigma(e) \leq \int_{S^{d-1}} \|\mu_e\|_{L^2(\mathbf{R}^d)}^2 d\sigma(e) \lesssim_d I_1(\mu).$$

Another application of Cauchy Schwartz implies that

$$\begin{aligned} 1 &\lesssim_d \int_{S^{d-1}} d\sigma(e) \\ &= \int_{S^{d-1}} \frac{|P_e(E)|^{1/2}}{|P_e(E)|^{1/2}} d\sigma(e) \\ &\leq \left(\int_{S^{d-1}} |P_e(E)| d\sigma(e) \right)^{1/2} \left(\int_{S^{d-1}} \frac{1}{|P_e(E)|} d\sigma(e) \right)^{1/2} \\ &\leq \left(\int_{S^{d-1}} |P_e(E)| d\sigma(e) \right)^{1/2} \cdot I_1(\mu)^{1/2}. \end{aligned}$$

Rearranging this identity completes the proof. \square

2.1 Applications to Falconer's Conjecture

It seems plausible that if $d \geq 2$, and $E \subset \mathbf{R}^d$ is Borel, with $\dim_{\mathbf{H}}(E) > d/2$, then the *distance set*

$$D(E) = \{|x - y| : x, y \in E\}$$

has positive Lebesgue measure, or maybe even has non-empty interior (we know this to be true if E has positive Lebesgue measure). This is Falconer's conjecture. Much remains open about this problem; we do not even know if $\dim_{\mathbf{H}}(D(E)) = 1$ if $\dim_{\mathbf{H}}(E) > d/2$. Here we prove less than state of the art results related to the Falconer conjecture.

Our strategy involves projecting measures on E onto $D(E)$. Given a probability measure μ compactly supported on E , define a measure $\delta(\mu)$ supported on $[0, \infty)$ such that for each $f \in C[0, \infty)$,

$$\int_0^\infty f d\delta(\mu) = \int_{\mathbf{R}^d} \int_{\mathbf{R}^d} f(|x - y|) d\mu(x) d\mu(y).$$

Obviously, $\delta(\mu)$ is compactly supported on $D(E)$. This projection smooths out μ , so that $\delta(\mu)$ has a density with respect to the Lebesgue measure.

Lemma 2.4. *For any compactly supported Borel measure μ with $I_{(d+1)/2}(\mu) < \infty$, $\delta(\mu)$ is absolutely continuous with*

$$\delta(\mu)(t) = \int_{\mathbf{R}^d} \hat{\sigma}_t(\xi) |\hat{\mu}(\xi)|^2 d\xi.$$

Proof. Let us begin with the simple argument in the case where μ is given by a smooth, compactly supported density $\phi \in C_c^\infty(\mathbf{R}^d)$. Indeed, a polar coordinate argument shows that for any $f \in C_c[0, \infty)$,

$$\begin{aligned} \int_0^\infty f d\delta(\phi) &= \int_{\mathbf{R}^d} \int_{\mathbf{R}^d} f(|x - y|) \phi(x) \phi(y) dx dy \\ &= \int_0^\infty f(t) \left(\int_{\mathbf{R}^d} \phi(x) \int_{S^{d-1}} \phi(x - e) d\sigma_t(e) dx \right). \end{aligned}$$

It follows that

$$\delta(\phi)(t) = \int_{\mathbf{R}^d} \phi(x) (\phi * \sigma_t)(x) dx.$$

Applying Parseval's formula,

$$\delta(\phi)(t) = \int_{\mathbf{R}^d} \hat{\sigma}_t(\xi) |\hat{\phi}(\xi)|^2 d\xi = \int_{\mathbf{R}^d} \hat{\sigma}(t\xi) |\hat{\phi}(\xi)|^2 d\xi.$$

Now it suffices to approximate general Borel measures. If $\{\phi_\varepsilon\}$ is a mollifier, and $\mu_\varepsilon = \mu * \phi_\varepsilon$, then

$$\delta(\mu_\varepsilon)(t) = \int \hat{\sigma}(t\xi) |\hat{\mu}(\xi)|^2 |\hat{\phi}(\varepsilon t)| d\xi$$

Now oscillatory integral techniques verify that $|\hat{\sigma}(t\xi)| \lesssim_t |\xi|^{-(d-1)/2}$, and since

$$\int_{\mathbf{R}^d} |\xi|^{-(d-1)/2} |\hat{\mu}(\xi)|^2 |\hat{\phi}(\varepsilon t)| d\xi \leq \int_{\mathbf{R}^d} |\xi|^{-(d-1)/2} |\hat{\mu}(\xi)|^2 d\xi \lesssim_d I_{(d+1)/2}(\mu) < \infty,$$

the dominated convergence theorem implies that for each $t \geq 0$,

$$\lim_{\varepsilon \rightarrow 0} \delta(\mu_\varepsilon)(t) \rightarrow \int_{\mathbf{R}^d} \widehat{\sigma}(t\xi) |\widehat{\mu}(\xi)|^2 d\xi.$$

A second use of the dominated convergence implies that

$$\lim_{\varepsilon \rightarrow 0} \int_{\mathbf{R}^d} f(t) \left(\int \widehat{\sigma}(t\xi) |\widehat{\mu}(\xi)|^2 |\widehat{\phi}(\varepsilon\xi)| d\xi \right) dt = \int_{\mathbf{R}^d} f(t) \left(\int_{\mathbf{R}^d} \widehat{\sigma}(t\xi) |\widehat{\mu}(\xi)|^2 d\xi \right) dt.$$

But as $\varepsilon \rightarrow 0$, μ_ε converges weakly to μ , so $\delta(\mu_\varepsilon)$ converges weakly to $\delta(\mu)$. But clearly this implies that for any $f \in C_c[0, \infty)$,

$$\int_{\mathbf{R}^d} f(t) d\delta(\mu)(t) = \int_{\mathbf{R}^d} f(t) \left(\int_{\mathbf{R}^d} \widehat{\sigma}(t\xi) |\widehat{\mu}(\xi)|^2 d\xi \right) dt. \quad \square$$

Theorem 2.5. Suppose $d \geq 2$. If $E \subset \mathbf{R}^d$ is a Borel set, then

$$\dim_{\mathbf{H}}(D(E)) \geq \min(1, \dim_{\mathbf{H}}(E) - (d-1)/2).$$

If, in addition, $\dim_{\mathbf{H}}(E) > (d+1)/2$, then $D(E)$ has nonempty interior.

Proof. If $\dim_{\mathbf{H}}(E) \geq (d+1)/2$, then we can find a probability measure compactly supported on E with $I_{(d+1)/2}(\mu) < \infty$. But then the integral formula implies $\delta(\mu)$ is continuous, hence the support of $\delta(\mu)$, contained in $D(E)$, has nonempty interior.

To establish the Hausdorff dimension bound, we require some deeper estimates. Suppose μ is supported on $B_R(0)$, and, and fix $0 < \eta < r < \infty$. We want to bound

$$\delta(\mu)[r, r+\eta] = \int_{\mathbf{R}^d} (\chi_{r\eta} * \mu) d\mu,$$

where $\chi_{r\eta}$ is the characteristic function of the annulus $\{x \in \mathbf{R}^d : r \leq |x| \leq r+\eta\}$. Again, mollifying and applying Parseval's formula shows that

$$\delta(\mu)[r, r+\eta] = \int_{\mathbf{R}^d} \widehat{\chi_{r\eta}}(\xi) |\widehat{\mu}(\xi)|^2 d\xi.$$

To estimate $\widehat{\chi_{r\eta}}$, we use Bessel function estimates since $\chi_{r\eta}$ is radial. Indeed, we find

$$\begin{aligned} \widehat{\chi_{r\eta}}(\xi) &\lesssim_d |\xi|^{1-d/2} \int_r^{r+\eta} J_{d/2-1}(2\pi|\xi|s) s^{d/2} ds \\ &= |\xi|^{-d} \int_{r|\xi|}^{(r+\eta)|\xi|} J_{d/2-1}(2\pi u) u^{d/2} du. \end{aligned}$$

But $J_{d/2-1}(2\pi u) \lesssim |u|^{-1/2}$, so integration gives

$$|\widehat{\chi_{r\eta}}(\xi)| \lesssim_d r^{(d-1)/2} \eta |\xi|^{-(d-1)/2}.$$

Since $(s^m J_m)' = s^m J_{m-1}$, we can use the fundamental theorem of calculus to also obtain a bound

$$\begin{aligned} |\widehat{\chi_{r\eta}}(\xi)| &\lesssim_d |\xi|^{-d} \left| [(r+\eta)|\xi|]^{d/2} J_{d/2}(2\pi(r+\eta)|\xi|) \right. \\ &\quad \left. - (r|\xi|)^{d/2} J_{d/2}(2\pi r|\xi|) \right| \\ &\lesssim r^{(d-1)/2} |\xi|^{-(d+1)/2}. \end{aligned}$$

We use the first bound for small frequencies, and the second bound for high frequencies. For any $t \in (0, 1)$,

$$\begin{aligned} \delta(\mu)[r, r+\eta] &= \int_{\mathbf{R}^d} \widehat{\chi_{r\eta}}(\xi) |\widehat{\mu}(\xi)|^2 d\xi \\ &= \int_{|\xi| \leq 1/\eta} r^{(d-1)/2} \eta |\xi|^{-(d-1)/2} |\widehat{\mu}(\xi)|^2 d\xi \\ &\quad + \int_{|\xi| > 1/\eta} r^{(d-1)/2} |\xi|^{-(d+1)/2} |\widehat{\mu}(\xi)|^2 d\xi \\ &= r^{(d-1)/2} \eta^t \left(\int_{|\xi| \leq 1/\eta} \eta^{1-t} |\xi|^{-(d-1)/2} |\widehat{\mu}(\xi)|^2 d\xi \right. \\ &\quad \left. + \int_{|\xi| \geq 1/\eta} \eta^{-t} |\xi|^{-(d+1)/2} |\widehat{\mu}(\xi)|^2 d\xi \right) \\ &= r^{(d-1)/2} \eta^t \left(\int_{|\xi| \leq 1/\eta} |\xi|^{(t-1)-(d-1)/2} |\widehat{\mu}(\xi)|^2 d\xi \right. \\ &\quad \left. + \int_{|\xi| \geq 1/\eta} |\xi|^{t-(d+1)/2} |\widehat{\mu}(\xi)|^2 d\xi \right) \\ &= r^{(d-1)/2} \eta^t \int_{\mathbf{R}^d} |\xi|^{t-(d+1)/2} |\widehat{\mu}(\xi)|^2 d\xi \\ &\lesssim_{t,d} r^{(d-1)/2} \eta^t \cdot I_{t+(d-1)/2}(\mu). \end{aligned}$$

This will suffice to prove the remainder of the theorem.

To obtain the Hausdorff dimension bounds, it clearly suffices to assume that $(d-1)/2 \leq \dim_{\mathbf{H}}(E) \leq (d+1)/2$, since we have already verified the claim in the case $\dim_{\mathbf{H}}(E) > (d+1)/2$, and the case $\dim_{\mathbf{H}}(E) < (d-1)/2$ is trivial. For any $t \in (0,1)$ with $(d-1)/2 + t < \dim_{\mathbf{H}}(E)$, we can find a probability measure μ on E with $I_{t+(d-1)/2}(\mu) < \infty$. If μ is supported on $B_R(0)$, then $\delta(\mu)$ is supported on $[0, 2R]$. Thus for any $r < 2R$ and $0 < \eta < r$, $\delta(\mu)[r, r+\eta] \lesssim_{R,\mu} \eta^t$. For any $\varepsilon > 0$ small enough that $\delta(\mu)[\varepsilon, \infty) > 0$, the measure

$$\nu(U) = \frac{\delta(\mu)(U \cap [\varepsilon, \infty))}{\delta(\mu)[\varepsilon, \infty)}$$

is a Frostman measure of dimension t supported on $D(E)$, so $\dim_{\mathbf{H}}(D(E)) \geq t$. \square

Remark. s

Recall that the Grassmanian manifold $G(n, m)$ is a space parameterizing the family of m dimensional hyperplanes in \mathbf{R}^n . The orthogonal group $O(n)$ acts on $G(n, m)$, and we let γ_{nm} denote the resultant Borel probability measure on $G(n, m)$ left-invariant under the action of $O(n)$. We then have Marstrand's projection theorem

Theorem 2.6 (Marstrand). s

Chapter 3

Fourier Dimension

The **Fourier dimension** of a Borel set E is

$$\dim_{\mathbf{F}}(E) = \sup\{s : \text{there is } \mu \in M(E) \text{ s.t. } |\hat{\mu}(\xi)| \leq |\xi|^{-s/2}\}$$

This implies the energy integrals of the right dimension to converge, implying $\dim_{\mathbf{F}}(E) \leq \dim_{\mathbf{H}}(E)$. A set is **Salem** if $\dim_{\mathbf{H}}(E) = \dim_{\mathbf{F}}(E)$.

3.1 Dimensions of Brownian Motion

Consider a one dimensional Brownian motion W . Then almost surely, for each $0 < \alpha < 1/2$, W is locally α Hölder continuous. For a fixed Borel set E , The bound

$$\dim_{\mathbf{H}}(W(E)) \leq \frac{1}{\alpha} \dim_{\mathbf{H}}(E)$$

then holds for almost every path of the motion. Taking $\alpha \uparrow 1/2$, we find the $\dim_{\mathbf{H}}(W(E)) \leq 2 \dim_{\mathbf{H}}(E)$. In this lecture we focus on a converse.

Theorem 3.1 (Mckean, 1955). *Let $A \subset [0, \infty)$ be Borel. Then $\dim_{\mathbf{H}}(W(E)) = 2 \dim_{\mathbf{H}}(E) \wedge 1$ almost surely.*

More generally,

Theorem 3.2 (Kaufman's Dimension Doubling Theorem). *Let B be a Brownian motion in \mathbf{R}^d , for $d \geq 2$, then almost surely, for every Borel set E ,*

$$\dim_{\mathbf{H}}(B(E)) = 2 \dim_{\mathbf{H}}(E)$$

Note that the almost surely condition is now independent of E , so we can apply this theorem to random sets. If $Z = \{t \geq 0 : B_t = 0\}$ is the random zero set of a path of Brownian motion, and $d \geq 2$, then almost surely we find $\dim_{\mathbf{H}}(E) = 0$. For $d = 1$, the zero may not even be zero dimensional, so we know that McKean's theorem cannot take out the almost surely over all subsets. We will follow Kahane's 1966 proof of McKean's result. Consider the following lemma.

Lemma 3.3. *If μ is an s dimensional measure supported on $[0, \infty)$, then almost surely, for all $|\xi| > 2$,*

$$|\widehat{\mu_W}(\xi)| \lesssim \frac{(\log |\xi|)^{1/2}}{|\xi|^{-s}}$$

where μ_W is the random pushforward measure of μ by the random path W , and the constant in the inequality is random.

Proof. □

If E was s dimensional, we could find an s dimensional probability measure on E . Then by Kahane's lemma, we compute, almost surely, that

$$I_s(\mu_W) \lesssim O(1) + \int_{|\xi| > 2} |\widehat{\mu_W}(\xi)|^2 |\xi|^{s-1} \lesssim O(1) + \int_{-\infty}^{\infty} \frac{\log |x|}{|x|^{-1-s}} < \infty$$

so μ_W is s dimensional, and so $\dim_{\mathbf{H}}(W(E)) \geq s$. Note that we actually proved something even stronger. The inequality above implies that $\dim_{\mathbf{F}}(W(E)) \geq s$ almost surely, so $W(E)$ is a Salem set almost surely. In particular, Salem sets exist. An alternate proof is to calculate, using Fubini's theorem, we calculate that if $Z \sim N(0, 1)$, then provided $r < 2s < 1$,

$$\mathbf{E}[I_r(\mu_W)] = \int_{-\infty}^{\infty} \mathbf{E}\left(\frac{1}{|W_t - W_s|^r}\right) dr = \left(\int_{-\infty}^{\infty} \frac{dt ds}{|t - s|^{r/2}}\right) \mathbf{E}\left(\frac{1}{|Z|^r}\right) < \infty$$

This doesn't require the K lemma at all.

Chapter 4

Properties of Generic Sets

In modern analysis, a popular technique is to establish certain properties that hold for a ‘generic point’ lying in some space. For instance, if X is a measure space, we can think of a property holding ‘generically’ if the set of points x upon which that property holds forms a set of measure zero. However, it is often the case that one wants to study the generic behaviour of spaces which have no natural measure theoretic structure. These spaces do have a natural metric space structure, however, and so it suffices to establish a suitably notion of a property holding ‘generically’ in a topological space. This is provided by considering a property to hold ‘generically’ if it holds except on a set of points which are of ‘first category’.

If X is a topological space, we say $E \subset X$ is *nowhere dense* if \bar{E} has non-empty interior. We say $E \subset X$ is of *first category* if it is the countable union of nowhere dense sets. A set E is of *second category* if it is not of first category, and a property on X holds *quasi-always* if it is true except on a set of first category. We recall that the *Baire category theorem* says that, in a complete metric space, or a locally compact topological space, the complement of sets of first category are dense, which supports our notion that such sets are generic.

Now let X be a metric space. For $E \subset X$, and $\varepsilon > 0$, we let

$$E_\varepsilon = \{x \in E : \text{there is } y \in E \text{ such that } d(x, y) < \varepsilon\}.$$

Now let \mathcal{E} , or $\mathcal{E}(X)$, be the family of all compact subsets of X . Given two compact sets $E, F \in \mathcal{E}$, we can define the *Hausdorff distance*

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

It is easy to see that since E and F are compact, $d_H(E, F) < \infty$. Moreover, since $\bar{E} = \bigcap_{\varepsilon > 0} E_\varepsilon$, and the fact that E and F are closed sets, $d_H(E, F) = 0$ if and only if $E = F$. If $E \subset F_\varepsilon$, and $F \subset G_\delta$, then $E \subset G_{\varepsilon+\delta}$, which is a manifestation of the triangle inequality for the Hausdorff distance. Thus \mathcal{E} has the natural structure of a metric space.

Theorem 4.1. *If X is complete, \mathcal{E} is complete.*

Proof. Note that if $E = \lim_{i \rightarrow \infty} E_i$, and if $\{x_i\}$ is a sequence of points with $x_i \in E_i$ for each i , and $\lim x_i = x$ for some $x \in X$, then we actually have $x \in E$. This is because we can find points $\{y_i\} \in E$ with $d(x_i, y_i) \leq d(E, E_i)$, hence $d(y_i, x) \leq d(y_i, x_i) + d(x_i, x) \rightarrow 0$, so $y_i \rightarrow x$, and hence $x \in E$ since E is closed. It will be a consequence of this theorem that E consists of *precisely* the limit points of the form above.

If $\{E_i\}$ is a Cauchy sequence in \mathcal{E} , we let E denote the family of limits of Cauchy sequences $\{x_i\}$ with $x_i \in E_i$ for each i . If $\{x_{i_j}\}$ is a Cauchy subsequence with $x_{i_j} \in E_{i_j}$ for each j , then we can extend it to a Cauchy sequence $\{x_i\}$ with $x_i \in E_i$ for each i , so without loss of generality, we may assume by thinning to a subsequence such that $d_H(E_i, E_{i+1}) < 1/2^{i+1}$. In this case, it is easy to see that $E = \bigcap_{i=1}^{\infty} (E_i)_{1/2^i}$. It is certainly true that $\bigcap_{i=1}^{\infty} (E_i)^{1/2^i} \subset E$. Conversely, if $\{x_i\}$ is a Cauchy sequence with $x_i \in E_i$ for each i , and if $x = \lim_{i \rightarrow \infty} x_i$, then for each i , there is $\varepsilon_i > 0$, such that $d_H(E_i, E_{i+j}) + \varepsilon_i < 1/2^i$. If we fix $j > i$ large enough that $d(x_j, x) < \varepsilon_i$, then we can find $y_i \in E_i$ with $d(y_i, x_j) < 1/2^i - \varepsilon_i$, hence $d(y_i, x) < 1/2^i$. Since i was arbitrary, $x \in \bigcap_{i=1}^{\infty} (E_i)_{1/2^i}$. The family of sets $\{(E_i)_{1/2^i}\}$ is nested. Moreover, if $x_i \in E_i$, then a sequence $\{x_j\}$ extending x_i with $x_j \in E_j$ for each j , and $d(x_j, x_{j+1}) < 1/2^{j+1}$, which is therefore Cauchy, and if $x = \lim x_i$ is in E , $d(x, x_i) < 1/2^i$. Thus $E_i \subset E_{1/2^i}$, and since E is obviously a subset of $E_{1/2^i}$, $E_{1/2^i} \subset (E_i)_{1/2^i}$. Thus $d_H(E, E_i) \leq 1/2^i$. Our proof will be finished if we show E is a compact set. Let us now show E is a compact set. Let $A \subset E$ be an infinite subset of E . We define a nested family of infinite sets $\{A_i\}$ inductively. by first defining $A_i = A_0$. Then, given A_i , there are finitely many balls $\{B(x_1, 1/2^{i+1}), \dots, B(x_N, 1/2^{i+1})\}$ which cover E_{i+1} , and then $\{B(x_1, 1/2^i), \dots, B(x_N, 1/2^i)\}$ is a cover of $(E_{i+1})_{1/2^{i+1}}$, and therefore a cover of A_i . Thus there is some j such that $B(x_j, 1/2^i) \cap A_i$ contains infinitely many points, and we set $A_{i+1} = B(x_j, 1/2^i) \cap A_i$. Clearly this family of sets has a limit point, so A has a limit point. This shows E is compact. \square

The proof of this theorem gives the following corollary.

Corollary 4.2. *If $\{E_i\}$ converges to E in the Hausdorff distance, then*

$$E = \{x \in X : \text{there are } x_i \in E_i \text{ s.t. } x = \lim x_i\}.$$

Since we have assigned a complete metric space structure to the family of compact sets in a metric space, it is now natural to discuss whether a property holds for ‘quasi-all’ compact subsets of a space.

Remark. There is an alternate family of sets which can be used with Baire category techniques. If X is a measure space, let \mathcal{E} denote the family of all finite measure subsets of X . We could define a metric, by letting the distance between two sets $E, F \in \mathcal{E}$ by equal to

$$d(E, F) = |E \Delta F| = |E - F| + |F - E|.$$

Then $d(E, F) = 0$ if and only if E and F differ by a set of measure zero. The map $E \mapsto \mathbf{I}_E$ embeds \mathcal{E} in $L^1(X)$. If \mathcal{I}_{E_i} is a sequence of sets and $\|\mathbf{I}_{E_i} - f\|_{L^1(X)} \rightarrow 0$, then a simple approximation argument can show $\{x \in X : f(x) \notin \{0, 1\}\}$ is a set of measure zero, so there exists a finite measure set F such that \mathbf{I}_F equals f almost everywhere. Thus \mathcal{E} is a complete metric space. However, most properties of sets which can be tested generically in \mathcal{E} also tend to extend to generic properties in $L^1(X)$, and Baire-category arguments tend to be extendable to $L^1(X)$ in such a way that we can apply one of the three main principles of functional analysis may be applied.

We begin with a basic, natural example, which shows that if X is a metric space with no isolated points, then quasi-all compact subsets of X have no isolated points.

Lemma 4.3. *Fix $\delta > 0$ and $0 < \varepsilon < \delta/3$. If $E, F \in \mathcal{E}$, $d_H(E, F) < \varepsilon$, and there is $x \in E$ such that $B(x, \delta) \cap E = \{x\}$, then there is $y \in F$ such that $d(x, y) < \varepsilon$, and moreover, for any such y ,*

$$B(y, \delta - 3\varepsilon) \cap F \subset B(x, \varepsilon) \subset B(y, 2\varepsilon).$$

Proof. There must certainly exist $y \in F$ with $d(x, y) < \varepsilon$. If $y' \in F$ satisfies $d(y, y') < \delta - 3\varepsilon$, then because $d_H(E, F) < \varepsilon$, there must be $x' \in E$ such that $d(x', y') < \varepsilon$. But the triangle inequality implies that $d(x, x') < 2\varepsilon < \delta$, so $x = x'$, and thus $d(y, y') < 2\varepsilon$. \square

Theorem 4.4. *Let X be a metric space with no isolated points. Then quasi-all compact subsets of X have no isolated points.*

Proof. For each $\delta > 0$, let $A_\delta = \{E \in \mathcal{E} : \text{there is } x \in E \text{ such that } B(x, \delta) \cap E = \{x\}\}$. Our goal is to show A_δ is closed and nowhere dense. First, we show A_δ is closed. If $\{E_i\}$ is a sequence in A_δ converging to some $E \in \mathcal{E}$. Pick $x_i \in E_i$ for each i such that $B(x_i, \delta) \cap E_i = \{x_i\}$. Without loss of generality, assume that $d(E_i, E) < \delta/3$, and we can choose a decreasing sequence $\varepsilon_i < \delta/3$ such that $d_H(E_i, E) < \varepsilon_i$. Then for each i , we can find $y_i \in E$ such that $d(x_i, y_i) < \varepsilon_i$ and $B(y_i, \delta - 3\varepsilon_i) \cap F \subset B(y_i, 2\varepsilon_i)$. Since F is compact, we can assume $\{y_i\}$ converges to some point y . But if $d(y_i, y) < \varepsilon'$ and $B(y_i, \delta - 3\varepsilon) \cap F \subset B(y_i, 2\varepsilon)$, then $B(y, \delta - 3\varepsilon - \varepsilon') \subset B(y, 2\varepsilon + \varepsilon')$. Taking $i \rightarrow \infty$, and thus $\varepsilon, \varepsilon' \rightarrow 0$, we conclude that $B(y, \delta) = \{y\}$. Thus $E \in A_\delta$. It is easy to show A_δ is nowhere dense. If $E \in A_\delta$, it can only have finitely many elements satisfying the property above by compactness, and we can always add a single point within a ε neighbourhood of each such point to find $F \notin A_\delta$ with $d_H(E, F) < \varepsilon$. Thus $\bigcup_{\delta>0} A_\delta = \bigcup_{n=1}^{\infty} A_{1/n}$ is a set of first category, and its complement is the set of all compact sets with no isolated points. \square

4.1 Kronecker Sets

Let us now also discuss an application to harmonic analysis on $\mathbf{T} = \mathbf{R}/\mathbf{Z}$. Recall that distinct values $x_1, \dots, x_n \in \mathbf{T}$ are *independent* if for any $m_1, \dots, m_n \in \mathbf{Z}$, $m_1 x_1 + \dots + m_n x_n = 0$ only if $m_1, \dots, m_n = 0$. It is a consequence of the classical theorem of Kronecker and Weyl that if $x_1, \dots, x_n \in \mathbf{T}$ are independent, then for any values $a_1, \dots, a_n \in \mathbf{C}$ with $|a_i| = 1$ for all i , and for any $\varepsilon > 0$, there exists an integer N such that $|\exp(2\pi N i x_i) - a_i| < \varepsilon$ for all $i \in \{1, \dots, n\}$. We wish to try and understand this result for infinite families of sets. So we define $E \subset \mathbf{T}$ to be *independent* if all finite subsets of E are independent, and *Kronecker* if for any $f \in C(\mathbf{T})$ with $|f(x)| = 1$ for all $x \in \mathbf{T}$, and for any $\varepsilon > 0$, there exists N such that $|\exp(2\pi N i x) - f(x)| < \varepsilon$ for all $x \in E$.

Lemma 4.5. *If E is a Kronecker set, then E is independent.*

Proof. If $m_1 x_1 + \dots + m_n x_n = 0$, for distinct x_i , then for any N ,

$$\exp(2\pi m_1 N x_1) \dots \exp(2\pi m_n N x_n) = 1.$$

If $|m_1| + \cdots + |m_n| = M$, and if we choose $f \in C(\mathbf{T})$ such that $f(x_k) = \exp(i\pi/m_k n)$ for each k , and if N is chosen such that $|f(x_k) - \exp(2\pi N i x_k)| < 1/M$, then

$$\begin{aligned} 2 &= |1 - (-1)| \\ &= \left| \prod_{k=1}^n \exp(2\pi N i x_k)^{m_k} - \prod_{k=1}^n \exp(i\pi/m_k n)^{m_k} \right| \\ &\leq \sum_{k=1}^n |m_k| |\exp(2\pi N i x_k) - \exp(i\pi/m_k n)| \leq 1, \end{aligned}$$

which gives a contradiction. \square

Remark. The converse, unfortunately, is not true. There are independent sets which are not Kronecker.

It is a general heuristic that a generic compact subset of a metric space is as ‘thin as possible’. Kronecker sets are very thin from the perspective of harmonic analysis, and so we expect quasi-all compact subsets of \mathbf{T} are Kronecker sets. To prove this statement, we note that the family of all functions $f \in C(\mathbf{T})$ with $|f(x)| = 1$ for all $x \in \mathbf{T}$ is separable, and if $\{f_i\}$ is a countable dense set, then E is Kronecker if and only if each of the functions f_i is arbitrarily approximable by a character on \mathbf{T} .

Theorem 4.6. *Quasi-all compact subsets of \mathbf{T} are Kronecker.*

Proof. For each function $f \in C(\mathbf{T})$ with $|f(x)| = 1$ for all $x \in \mathbf{T}$, let $A(f, \varepsilon)$ be the family of all sets E such that there exists an integer N such that $|f(x) - \exp(2\pi i N x)| < \varepsilon$ for each $x \in E$. We claim $A(f, \varepsilon)$ is a dense, open subset of \mathcal{E} . If E is fixed, since E is compact and f is continuous, there is $\delta > 0$ such that $|f(x) - \exp(2\pi i N x)| < \varepsilon$ for each $x \in E_\delta$. But this implies that $F \in A(f, \varepsilon)$ for each F with $d_H(E, F) < \delta$. Thus $A(f, \varepsilon)$ is open. If $\{f_i\}$ is a dense family of continuous functions in the L^∞ norm, then $\bigcap_{i=1}^\infty \bigcap_{n=1}^\infty A(f_i, 1/n)$ is the family of all Kronecker sets. Hence to complete our proof it suffices to show $A(f, \varepsilon)$ is open and dense for any $f \in C(\mathbf{T})$ and $\varepsilon > 0$.

To show density, let E be an arbitrary set. Since f is continuous, it is actually uniformly continuous, so we can find δ such that if $|x - y| < \delta$, then $|f(x) - f(y)| < \sqrt{2}$. This forces the values of f to lie on a particular half of the circle over any δ interval. Thus if N is a large integer such that

$2/N < \delta$, then $\{x \in \mathbf{T} : \exp(2\pi i N x) = f(x)\}$ contains a point in any length $2/N$ interval of \mathbf{T} . In particular, this means that if

$$F = \{x \in E_{2/N} : \exp(2\pi i N x) = f(x)\},$$

then $d_H(F, E) < 2/N$ and $F \in A(f, \varepsilon)$ for all $\varepsilon > 0$. Since N is arbitrary, $A(f, \varepsilon)$ is dense. \square

This theorem also even holds under more algebraically restricted circumstances. We note that

$$\{(E_1, E_2) \in \mathcal{E}^2 : E_1 + E_2 = \mathbf{T}\}$$

is a closed subset of \mathcal{E}^2 , and is therefore complete. We shall show that quasi-all elements of this set are pairs of Kronecker sets.

Theorem 4.7. *Quasi-all $(E_1, E_2) \in \mathcal{E}^2$ with $E_1 + E_2 = \mathbf{T}$ are Kronecker sets.*

Proof. For each f , $\varepsilon > 0$, and $i \in \{0, 1\}$, let $A(f, \varepsilon, i)$ be the family of all pairs of sets (E_1, E_2) with $E_1 + E_2 = \mathbf{T}$ for which there exists an integer N such that for each $x \in E_i$, $|f(x) - \exp(2\pi i N x)| < \varepsilon$. Then

$$\bigcap_{i \in \{0, 1\}} \bigcap_{j=1}^{\infty} \bigcap_{k=1}^{\infty} A(f_j, 1/k, i)$$

is the family of all pairs (E_1, E_2) where $E_1 + E_2 = \mathbf{T}$ and E_1 and E_2 are Kronecker. Thus it suffices to show $A(f, \varepsilon, i)$ is open and dense for each f , i , and $\varepsilon > 0$. It is easy to see that $A(f, \varepsilon, i)$ is open, as in the last proof. Without loss of generality, we can assume $i = 1$. So fix $\delta > 0$, and consider the sets $\tilde{E}_i = (E_i)_{\delta/2}$. Then if $d(E_1, F) < \delta/4$, $F + \tilde{E}_2 = \mathbf{T}$. Thus choosing F as in the last theorem gives a pair (F, \tilde{E}_2) within a δ neighbourhood of (E_1, E_2) . Thus $A(f, \varepsilon, i)$ is open and dense, which proves the theorem. \square

4.2 Besicovitch Sets

We can also use Baire category techniques to study the generic behaviour of Besicovitch sets, subsets of the plane containing a unit line segments in each direction. For each $a, b \in [-2, 2]$, we let l_{ab} be the line segment connecting $(a, 0)$ to $(b, 1)$. We let \mathcal{F} consist of the family of all compact

subsets E of $[-2, 2] \times [0, 1]$ which are a union of line segments of the form l_{ab} , and such that for each $v \in [-1, 1]$, there exists $a \in [-2, 2]$ such that $l_{a(a+v)} \subset \mathcal{F}$. Such a set contains a unit line segment from any angle in a 45 degree arc, and so two copies of such a set form a Besicovitch set.

Lemma 4.8. \mathcal{F} is a closed subset of the family \mathcal{E} of all compact subsets of $[-2, 2] \times [0, 1]$.

Proof. Suppose $E = \lim_{i \rightarrow \infty} E_i$, with $E_i \in \mathcal{F}$ for all i . We begin by showing E is a union of line segments. Suppose that $x \in E$. Then there are points $x_i \in E_i$ for each i such that $x = \lim x_i$. Moreover, we can find pairs a_i and b_i such that $x_i \in l_{a_i b_i}$. Since $[-2, 2]$ is compact, we can assume without loss of generality that a_i and b_i are convergent sequences to values a and b . But this means that $l_{a_i b_i}$ converges in the Hausdorff metric to l_{ab} , and thus $x \in l_{ab}$. Moreover, $l_{ab} \subset E$. Thus E is a union of lines. Similar techniques can be used to show E contains a line segment corresponding to each $v \in [-1, 1]$. \square

Theorem 4.9. \mathcal{F} is a closed subset of the family \mathcal{E} of all compact subsets of $[-2, 2] \times [0, 1]$, and quasi-all elements of \mathcal{F} have Lebesgue measure zero.

Proof. We show the stronger statement that quasi-all elements E of \mathcal{F} are such that, for all $y \in [0, 1]$, $E \cap (\mathbf{R} \times \{y\})$ has length zero (with respect to the Hausdorff measure), from which the general statement follows by Fubini's theorem. For each interval $I \subset [0, 1]$ with length ε , we set $A(I)$ to be the set of all compact sets $E \in \mathcal{F}$ such that, for each $y \in I$, $E \cap [-2, 2] \times \{y\}$ has length less than 100ε . The set

$$\bigcap_{n=1}^{\infty} \bigcap_{k=0}^{n-1} A([k/n, (k+1)/n])$$

is the class of sets whose slices have length zero, so it suffices to show $A(I)$ is open and dense for each interval I . It is obvious that the family of such sets is open, since if $E \in A(I)$, there exists a sufficiently small δ such that $E_\delta \in A(I)$, and so if $d_H(E, F) < \delta$, $F \in A(I)$. Thus we must concentrate on showing $A(I)$ is dense in \mathcal{F} . So consider a set $E \in \mathcal{F}$, and fix a large integer $N > 4$, defining $\delta = 1/N$. If $E = \bigcup_{\alpha} l_{a_{\alpha} b_{\alpha}}$, and we define $x_{\alpha} = a_{\alpha} + 2\delta$ and $y_{\alpha} = b_{\alpha} + 2\delta$ if $a_{\alpha} \leq -1/2$, and $x_{\alpha} = a_{\alpha} - 2\delta$ and $y_{\alpha} = b_{\alpha} - 2\delta$ if $a_{\alpha} > -1/2$. Then $E' = \bigcup_{\alpha} l_{x_{\alpha} y_{\alpha}}$ is an element of \mathcal{F} , a subset of $[-2 + \delta, 2 - \delta] \times \{y\}$, and

$d(E, E') \leq 2\delta$. Thus to prove density, it suffices to show that given $E \in \mathcal{F}$ which is also a subset of $[-2 + 2\delta, 2 - 2\delta] \times \{y\}$, we can find $F \in A(I)$ with $d_H(E, F) \lesssim \delta$. We begin by finding $\{a_1, \dots, a_N\}$ such that $l_{a_k(a_k+k/N)}$ for each $k \in \{-N, \dots, N\}$. Fix a point $u_k \in l_{a_k(a_k+k/N)} \cap ([-2, 2] \times I)$, and consider the family of all lines through u_k which lie in a δ thickening of $l_{a_k(a_k+k/N)}$. This set has length at most δ on each slice of $[-2, 2] \times I$, and contains a line $l_{x(x+v)}$ for each $v \in [(k-1)/N, (k+1)/N]$. The union G of such lines lies in \mathcal{F} , and is a subset of the δ thickening of E . If we consider a finite union of lines H such that $d_H(H, E) < \delta$, then $G \cup H \in A(I)$ and $d(H, E) < \delta$. \square

4.3 Independent Sets With Sharp Fourier Decay

We recall that if $E \subset \mathbf{T}$ has Fourier dimension exceeding $2/n$, and we consider n integers $m_1, \dots, m_n \in \mathbf{Z}$, then $m_1 E + \dots + m_n E$ contains a non-empty interval. It follows that by a slight modification of these integers, one can find $m_1, \dots, m_n \in \mathbf{Z}$ and distinct $x_1, \dots, x_n \in E$ such that

$$m_1 x_1 + \dots + m_n x_n = 0. \quad (4.1)$$

In particular, this implies that if μ is supported on an independent set, then for any $\alpha > 0$,

$$\sup_{n \neq 0} |\hat{\mu}(n)| |n|^\alpha = \infty.$$

In this section, we address the tightness of this result. In particular, we prove the following results:

- We use Baire category methods to construct probability measures μ with Fourier dimension $1/n$ supported on sets containing no solutions to (4.1) for any $m_1, \dots, m_n \in \mathbf{Z}$. A slightly more complicated argument constructs measures with Fourier dimension $1/(n-1)$. This result is far from tight, and improving this bound remains an open question.
- Let $\{A(k) : k \neq 0\}$ be any sequence of positive scalars such that for each $\alpha > 0$,

$$\lim_{|k| \rightarrow \infty} |A(k)| |k|^\alpha = \infty.$$

Then there exists a probability measure μ with independent support, but for each nonzero $k \in \mathbf{Z}$, $|\hat{\mu}(k)| \leq A(k)$.

Since we employ Baire category methods, our theorem involves finding an appropriate complete metric space on which to study generic behaviour. Given a particular sequence $\{A(k) : k \neq 0\}$, we extend the definition by setting $A(0) = 1$. We then let $M(A)$ be the family of all finite measures $\mu \in M(\mathbf{T})$ such that

$$\|\mu\|_A = \sup_{n \neq 0} \frac{|\widehat{\mu}(n)|}{|A(n)|} < \infty.$$

The family of all such measures forms a vector space, and in particular, a Banach space under the norm $\|\cdot\|_A$. To see this, we note that if $\{\mu_i\}$ is a sequence in this space, then $\mu_i(\mathbf{T}) = \widehat{\mu}_i(0)$ is uniformly bounded, so by thinning the sequence, there exists a finite measure $\mu \in M(\mathbf{T})$ such that $\{\mu_i\}$ converges weakly to μ . Thus the Fourier coefficients of the sequence $\{\mu_i\}$ converge pointwise to μ , and since μ_i is Cauchy in the norm on $M(A)$, this implies that $\mu \in M(A)$, and $\|\mu_i - \mu\|_A \rightarrow 0$.

Remark. We have shown that if $\mu_i \rightarrow \mu$ in $M(A)$, then μ_i converges weakly to μ . In particular, this means the set $P(A)$ of finite probability measures in $M(A)$ is a closed, convex subset of $M(A)$, since $P(A)$ can be described as the family of all measures μ such that

$$\int_{\mathbf{T}} d\mu = 1,$$

and for any non-negative $f \in C(\mathbf{T})$, $\int f d\mu \geq 0$.

Remark. The above definition works for any sequence $\{A(k)\}$. Nonetheless, we will often assume in the following discussion that there exists $\alpha > 0$ such that

$$A(k) \gtrsim |k|^{-\alpha},$$

so that $C^\infty(\mathbf{T}) \subset M(A)$, since such functions have rapid Fourier decay.

Convergence of measures in $M(A)$ does not imply enough information about how similar the supports of these measures are. Since we would like to use Baire category arguments to infer information about the supports of measures, the norm on $M(A)$ therefore must be augmented. We might want to define a metric on $M(A)$ by setting

$$d(\mu, \nu) = \|\mu - \nu\|_A + d_H(\text{supp}(\mu), \text{supp}(\nu)),$$

so that we may control both Fourier coefficients and geometric information about the support of a measure. But unfortunately, this metric is not necessarily complete.

Example. Fix an interval $I \subset \mathbf{T}$, and pick non-negative functions $\phi, \eta \in C^\infty(\mathbf{T})$ such that $\text{supp}(\phi) \cap I = \emptyset$, and $I \subset \text{supp}(\eta)$, then the sequence $\psi_n = (1/n)\phi + (1 - 1/n)\eta$ has support $\text{supp}(\phi) \cup \text{supp}(\eta)$ for all n , and if $n, m > 0$,

$$\|\psi_{n+m} - \psi_n\|_A = \frac{m}{n(n+m)} \|\eta - \phi\|_A \lesssim 1/n,$$

so $\{\psi_n\}$ is a Cauchy sequence under the metric d just defined. However, ψ_n converges to η in $M(A)$, which must be the limit under the metric d , if such a limit exists. But $d_H(\text{supp}(\psi_n), \eta) \gtrsim |I| > 0$ for each n , so ψ_n does not converge to η under the metric d , and so this metric is not complete.

Fortunately, quasi-all elements of the completion of $M(A)$ under this metric belong to $M(A)$, as we see below, so we can ‘essentially’ treat $M(A)$ as a complete metric space. Since the space \mathcal{E} of compact subsets of \mathbf{T} is complete under the Hausdorff metric, so too is the space $\mathcal{E} \times M(A)$ under the product metric. The subset

$$\mathcal{G} = \{(E, \mu) \in \mathcal{E} \times P(A) : \text{supp}(\mu) \subset E\}$$

is a closed subset of $\mathcal{E} \times P(\mu)$, and thus complete in its own right. If (E_n, μ_n) are elements of \mathcal{G} converging to (E, μ) in $\mathcal{E} \times P(\mu)$, and $\varepsilon > 0$ is fixed, then for suitably large n , $E_n \subset E_\varepsilon$. We also know $\mu_n \rightarrow \mu$, and since $\text{supp}(\mu_n) \subset E_\varepsilon$ for suitably large n , $\text{supp}(\mu) \subset E_\varepsilon$. And since $\varepsilon > 0$ was arbitrary, $\text{supp}(\mu) \subset E$, so $(E, \mu) \in \mathcal{G}$. The next lemma shows that \mathcal{G} is the metric completion of $M(A)$.

Lemma 4.10. For quasi-all $(E, \mu) \in \mathcal{G}$, $\text{supp}(\mu) = E$.

Proof. For each interval I , let $A(I)$ be the family of all $(E, \mu) \in \mathcal{G}$ such that $E \cap I \neq \emptyset$ and $\mu(I) = 0$. Clearly this set is closed, using Hausdorff convergence of sets, and weak convergence of measures. Moreover, this set is nowhere dense; if $\varepsilon > 0$ is fixed, then we can find $F \in \mathcal{E}$ with $E \subset F$ and $d_H(E, F) < \varepsilon$, where $F \cap I$ contains a length $\varepsilon/2$ interval J . If $\psi \in C^\infty(\mathbf{T})$ is a non-negative probability density function supported on J , define $\mu_\varepsilon = (1 - \varepsilon)\mu + \varepsilon\psi$. Then $\|\mu_\varepsilon - \mu\| \lesssim \varepsilon$, so $d((E, \mu), (F, \mu_\varepsilon)) \lesssim \varepsilon$. Since

(F, μ_ε) is an element of \mathcal{G} , but not an element of $A(I)$, this proves nowhere density. Since $\bigcap_I A(I)$ is the family of pairs (E, μ) with $\text{supp}(\mu) = E$, and the uncountable intersection can be replaced with a countable intersection by a base of intervals, this completes the proof of the claim. \square

The set $\mathcal{G}_0 = \{(E, \mu) \in \mathcal{G} : \widehat{\mu}(n) = o(A_n)\}$ is a closed subset of \mathcal{G} . Moreover, quasi-all pairs $(E, \mu) \in \mathcal{G}_0$ satisfy $\text{supp}(\mu) = E$, by the same construction in the last lemma. If we show that E is an independent set for quasi-all $(E, \mu) \in \mathcal{G}_0$, then our proof will be complete, since if μ_0 is the uniform probability measure on \mathbf{T} , then $(\mathbf{T}, \mu_0) \in \mathcal{G}_0$, so we can find $(E, \mu) \in \mathcal{G}_0$, with E independent, such that $\|\mu - \mu_0\|_A \leq 1$, which implies $|\widehat{\mu}(n)| \leq A(n)$ for each $n \neq 0$. The main reason that we use \mathcal{G}_0 rather than \mathcal{G} is so that we can approximate measures by smooth functions.

We will have need to consider mollifications, i.e. convolutions with smooth approximation to the identity, so we begin by fixed notation. Choose a non-negative bump function $\phi \in C^\infty(\mathbf{T})$ supported on $[-1, 1]$ with

$$\int \phi(x) dx = 1.$$

For each $\varepsilon > 0$, define

$$\phi_\varepsilon(x) = \begin{cases} \varepsilon^{-1} \phi(x/\varepsilon) & : x \in [-\varepsilon, \varepsilon], \\ 0 & : \text{otherwise.} \end{cases}$$

Then ϕ_ε is non-negative and supported on $[-\varepsilon, \varepsilon]$, and $\int \phi_\varepsilon(x) dx = 1$. As $\varepsilon \rightarrow 0$, ϕ_ε converges weakly to the Dirac delta function at the origin, so for each $k \in \mathbf{Z}$,

$$\lim_{\varepsilon \rightarrow 0} \widehat{\phi_\varepsilon}(k) = 1.$$

Moreover, for each $\alpha > 0$, there exists $C_\alpha > 0$ such that for all $\varepsilon > 0$ and non-zero $k \in \mathbf{Z}$,

$$|\widehat{\phi_\varepsilon}(k)| \leq \frac{C_\alpha}{\varepsilon^\alpha |k|^\alpha}.$$

As $\varepsilon \rightarrow 0$, $\{\phi_\varepsilon\}$ is an approximation to the identity.

Lemma 4.11. *The family of all $(E, \mu) \in \mathcal{G}_0$ with $\mu \in C^\infty(\mathbf{T})$ is dense in \mathcal{G}_0 .*

Proof. Fix $(E, \mu) \in \mathcal{G}_0$. For each $\varepsilon > 0$, define $\mu_\varepsilon = \mu * \phi_\varepsilon$. Then $\mu_\varepsilon \in C^\infty(\mathbf{T})$. Since μ is supported on E , μ_ε is supported on $\overline{E_\varepsilon}$. Given δ , since $\mu \in \mathcal{G}_0$, we can pick N suitably large that $|\widehat{\mu}(n)| \leq \delta A(n)$ for $|n| \geq N$. Since $\widehat{\phi_\varepsilon}(k)$ converges to 1 for all k , there is $\varepsilon(\delta, N)$ such that if $\varepsilon < \varepsilon(\delta)$, $|\widehat{\phi_\varepsilon}(k) - 1| < \delta$ for all $|k| \leq N$. This means that for $|k| \leq N$

$$\begin{aligned} |\widehat{\mu_\varepsilon}(k) - \widehat{\mu}(k)| &= |\widehat{\mu}(k)\widehat{\phi_\varepsilon}(k) - \widehat{\mu}(k)| \\ &= |\widehat{\mu}(k)| |\widehat{\phi_\varepsilon}(k) - 1| \\ &< \delta |\widehat{\mu}(k)| < \delta \|\mu\|_A \cdot A(k). \end{aligned}$$

But if $|k| \geq N$,

$$|\widehat{\mu_\varepsilon}(k) - \widehat{\mu}(k)| \leq 2|\widehat{\mu}(k)| \leq 2\delta A(k).$$

Thus we have shown that $\|\mu - \mu_\varepsilon\|_A \lesssim \delta$ for suitably small ε . If $\varepsilon < \delta$, then μ_ε is supported on E_δ , and $d(E, E_\delta) \lesssim \delta$. Thus we conclude that $d((E, \mu), (\overline{E_\delta}, \mu_\varepsilon)) \lesssim \delta$. \square

Remark. The last lemma implies that to show a set $A \subset \mathcal{G}_0$ is dense, it suffices to show that \overline{A} contains all (E, μ) , where $\mu \in C^\infty(\mathbf{T})$.

To prove that E is an independent set for quasi-all elements $(E, \mu) \in \mathcal{G}_0$, we can by taking countable unions restrict ourselves to avoiding a particular equation $m_1 x_1 + \dots + m_n x_n = 0$ for some non-zero $m \in \mathbf{Z}^n$, where x_1, \dots, x_n are distinct. Moreover, we can also restrict ourselves to avoiding solutions (x_1, \dots, x_n) such that $|x_i - x_j| \geq \delta$ for all $i \neq j$, for a fixed $\delta > 0$. So we set $A(m, \delta)$ to be the family of all $(E, \mu) \in \mathcal{G}_0$, where there is $x_1, \dots, x_n \in E$ with $|x_i - x_j| \geq \delta$, and $m_1 x_1 + \dots + m_n x_n = 0$. $A(m, \delta)$ is clearly closed, since if (E_i, μ_i) converges to (E, μ) in \mathcal{G}_0 , and there are $x_i = (x_{i1}, \dots, x_{in}) \in E_i^n$ for each i with $m \cdot x_i = 0$ and $|x_{ij} - x_{ik}| \geq \delta$ for each j, k , then by compactness of \mathbf{T}^n we may thin the sequence, assuming x_i converges to some $x \in E^n$. It is then easy to see that $|x_j - x_k| \geq \delta$ for each j, k , and $m \cdot x = 0$, so $(E, \mu) \in \mathcal{G}_0$. All that remains is to show $A(m, \delta)$ is nowhere dense in \mathcal{G}_0 . We rely on a randomized construction procedure to establish density.

Theorem 4.12. Fix $n > 0$ and non-zero $m \in \mathbf{Z}^n$. Then for suitably large integers M , there are $X_1, \dots, X_M \in \mathbf{T}$ such that for any $j_1, \dots, j_n \in \{1, \dots, M\}$,

$$|m_1 X_{j_1} + \dots + m_n X_{j_n}| \geq \frac{1}{10M^n},$$

and if $\mu = M^{-1}(\delta_{X_1} + \cdots + \delta_{X_M})$, then for any $|k| \leq M^{10n}$,

$$|\hat{\mu}(k)| \leq 10n^{1/2} \log(M)^{1/2} M^{-1/2}.$$

Proof. Consider M random variables X_1, \dots, X_M , independant and uniformly distributed on \mathbf{T} . Then for each $j_1, \dots, j_n \in \{1, \dots, M\}$, $m_1 X_{j_1} + \cdots + m_n X_{j_n}$ is uniformly distributed on \mathbf{T} , so

$$\mathbf{P}(|m_1 X_{j_1} + \cdots + m_n X_{j_n}| \leq \varepsilon) = 2\varepsilon.$$

Applying a union bound, we conclude that

$$\mathbf{P}(\text{for all } j_1, \dots, j_n \in \{1, \dots, M\} : |m_1 X_{j_1} + \cdots + m_n X_{j_n}| \geq \varepsilon) \geq 1 - 2M^n \varepsilon.$$

Consider the random measure $\mu = M^{-1}(\delta_{X_1} + \cdots + \delta_{X_M})$. Then

$$\hat{\mu}(k) = M^{-1}(e^{-2\pi i k X_1} + \cdots + e^{-2\pi i k X_M}).$$

Thus $\hat{\mu}(k)$ is an average of M bounded random variables, so we can apply Hoeffding's inequality to conclude that for any $\delta > 0$,

$$\mathbf{P}(|\hat{\mu}(k)| \geq \delta) \leq 2 \exp(-2M\delta^2).$$

Thus another union bound shows that

$$\mathbf{P}(\text{for all } k \in [-K, K] : |\hat{\mu}(k)| \leq \delta) \geq 1 - 5K \exp(-2M\delta^2)$$

Thus, assuming

$$2M^n \varepsilon + 5K \exp(-2M\delta^2) < 1,$$

then there exists $X_1, \dots, X_M \in \mathbf{T}$ such that $m_1 X_{j_1} + \cdots + m_n X_{j_n} \geq \varepsilon$ for all $j_1, \dots, j_n \in \{1, \dots, M\}$, and if $\mu = M^{-1}(\delta_{X_1} + \cdots + \delta_{X_M})$, then $|\hat{\mu}(k)| \leq \delta$ for all $k \in [-K, K]$. In particular, the result holds if we choose $\varepsilon = 1/10M^n$, $K = M^{10n}$, and $\delta = 10n^{1/2} \log(M)^{1/2} M^{-1/2}$. \square

Lemma 4.13. Suppose $\varepsilon > 0$, $m \in \mathbf{Z}^n$, and fix an arbitrary sequence $\{B(k) : k \neq 0\}$ such that $\lim_{|k| \rightarrow \infty} B(k) = \infty$. Then there exists a measure $\mu \in M(\mathbf{T})$ such that for each x_1, \dots, x_n in the support of μ , $m_1 x_1 + \cdots + m_n x_n \neq 0$, and for all nonzero $k \in \mathbf{Z}$,

$$|\hat{\mu}(k)| \leq \varepsilon \cdot B(k) \log(1 + |k|)^{1/2} \cdot |k|^{-1/2n}.$$

Proof. Let μ be as in the proof above. If

$$\delta = \frac{1}{20(|m_1| + \dots + |m_n|)M^n},$$

then for any x_1, \dots, x_n in the support of $\mu_\delta = \mu * \phi_\delta$, $m_1 x_1 + \dots + m_n x_n \neq 0$. Since $\{B(k)\}$ converges to infinity, it is lower bounded, and so for each $\varepsilon > 0$, there exists a small constant $c(\varepsilon)$ independant of M such that if $|k| \leq c(\varepsilon)M^n$, then

$$|\widehat{\mu_\delta}(k)| \leq |\widehat{\mu}(k)| \leq 10 \log(M^n)^{1/2} (M^n)^{-1/2n} \leq \varepsilon B(k) \log(1 + |k|)^{1/2} |k|^{-1/2n}.$$

On the other hand, if $c(\varepsilon)M^n \leq |k| \leq M^{10n}$, then

$$\begin{aligned} |\widehat{\mu_\delta}(k)| &\leq |\widehat{\mu}(k)| |\widehat{\phi_\delta}(k)| \\ &\lesssim \frac{\log(M)^{1/2} M^{-1/2}}{\delta |k|} \\ &\lesssim_m \frac{\log(M)^{1/2} M^{n-1/2}}{\log(1 + |k|)^{1/2} \cdot |k|^{1-1/2n}} \log(1 + |k|)^{1/2} |k|^{-1/2n} \\ &\lesssim_\varepsilon \log(1 + |k|)^{1/2} |k|^{-1/2n}. \end{aligned}$$

If M is suitably large that $B(k) \geq 1/\varepsilon$ for $|k| \geq M^n$, then for all k with $M^n \leq |k| \leq M^{10n}$,

$$|\widehat{\mu_\delta}(k)| \leq \varepsilon B(k) \log(1 + |k|)^{1/2} |k|^{-1/2n}.$$

If $|k| \geq M^{10n}$,

$$|\widehat{\mu_\delta}(k)| \leq |\widehat{\phi_\delta}(k)| \lesssim_m M^n / |k| \leq (M^n / |k|^{1-1/2n}) |k|^{-1/2n} \leq |k|^{-1/2},$$

so if M is suitably large, and $|k| \geq M^{10n}$,

$$|\widehat{\mu_\delta}(k)| \leq \varepsilon B(k) \log(1 + |k|)^{1/2} |k|^{-1/2}.$$

This addresses all possible values of k , completing the proof. \square

If $\{B(k)\}$ is a sequence chosen carefully enough that

$$\lim_{|k| \rightarrow \infty} B(k) \log(1 + |k|)^{1/2} |k|^{-1/2} = 0,$$

and if μ is chosen such that for $k \neq 0$,

$$\widehat{\mu}(k) \leq \varepsilon B(k) \log(1 + |k|)^{1/2} |k|^{-1/2},$$

then $|\widehat{\mu}(k)| \lesssim \varepsilon$ for all $k \neq 0$, in which case the following lemma comes into play, showing a quantitative density statement about the support of μ .

Lemma 4.14. *For any $\varepsilon > 0$, there exists $\delta > 0$ such that if μ is a probability measure on \mathbf{T} such that $|\widehat{\mu}(k)| \leq \delta$ for all $k \neq 0$, then the support of μ contains a point in every length ε interval.*

Proof. By translation, it suffices to show that the support of μ intersects $(-\varepsilon/2, \varepsilon/2)$. But if μ did not intersect this interval, then

$$\int \phi_{\varepsilon/2} d\mu = 0.$$

Applying Parseval's equality, this means that

$$\sum_{k \in \mathbf{Z}} \widehat{\phi_{\varepsilon/2}}(k) \widehat{\mu}(k) = 0.$$

But

$$\left| \sum_{k \in \mathbf{Z}} \widehat{\phi_{\varepsilon/2}}(k) \widehat{\mu}(k) \right| \geq 1 - \delta \sum_{k \neq 0} |\widehat{\phi_{\varepsilon/2}}(k)| = 1 - O_\varepsilon(\delta),$$

where the last equality holds because $\phi_{\varepsilon/2}$ is smooth, and so its Fourier coefficients are rapidly decreasing. But this is impossible if δ is sufficiently small, depending on ε . \square

Lemma 4.15. *For any nonzero $m \in \mathbf{Z}^n$, and $\delta > 0$, $A(m, \delta)$ is dense in \mathcal{G}_0 .*

Proof. Fix $\varepsilon_0 > 0$, and $(E, \mu_0) \in \mathcal{G}_0$. Without loss of generality, we may assume that $\mu_0 \in C^\infty(\mathbf{T})$, and that ε_0 is suitably small that the support of μ_0 is a union of length ε_0 intervals. For any $\varepsilon > 0$, and for any sequence $\{B(k)\}$ converging to ∞ as $|k| \rightarrow \infty$, we can find a measure $\eta \in M(\mathbf{T})$ such that for each x_1, \dots, x_n in the support of η , $m_1 x_1 + \dots + m_n x_n \neq 0$, and for any nonzero $k \in \mathbf{Z}$,

$$|\widehat{\eta}(k)| \leq \varepsilon B(k) \log(1 + |k|)^{1/2} |k|^{-1/2n}.$$

Assume, without loss of generality, that $B(k) \lesssim_\varepsilon |k|^\varepsilon$ for each $\varepsilon > 0$, and that there exists a universal constant $C > 0$ such that for each $|k|$, if $|k|/2 \leq |k'| \leq 2|k|$, then

$$(1/C)B(k) \leq B(k') \leq CB(k).$$

Such a choice is provided, for instance, by setting $B(k) = \log(1 + |k|)$, with $C = 2$, but can be chosen as slowly growing as necessary. Define $\mu = \mu_0 \eta$. If ε is suitably small, depending on ε_0 , this implies that the support of η contains a point in each length ε_0 interval. Thus

$$d_H(\text{supp}(\mu), \text{supp}(\mu_0)) \leq \varepsilon_0.$$

Since $\mu_0 \in C^\infty(\mathbf{T})$, we calculate that

$$\begin{aligned} |\widehat{\mu}(k) - \widehat{\mu}_0(k)| &= |(\widehat{\mu}_0 * \widehat{\eta})(k) - \widehat{\mu}_0(k)| \\ &= \left| \sum_{k' \neq 0} \widehat{\mu}_0(k') \widehat{\eta}(k - k') \right| \\ &\lesssim_{\mu_0} \sum_{k' \neq 0} \frac{|\widehat{\eta}(k - k')|}{|k'|^2} \\ &\leq \sum_{|k'| \leq |k|/2} \frac{|\widehat{\eta}(k - k')|}{|k'|^2} + \sum_{|k'| > |k|/2} \frac{|\widehat{\eta}(k - k')|}{|k'|^2} \\ &\lesssim \varepsilon B(k) \log(1 + |k|)^{1/2} |k|^{-1/2n} \sum_{|k'| \leq |k|/2} \frac{1}{|k'|^2} + \sum_{|k'| > |k|/2} \frac{\varepsilon}{|k'|^2} \\ &\lesssim \varepsilon B(k) \log(1 + |k|)^{1/2} |k|^{-1/2n} + \varepsilon |k|^{-1} \\ &\lesssim \varepsilon B(k) \log(1 + |k|)^{1/2} |k|^{-1/2n}. \end{aligned}$$

We recall that for the sequence $\{A(k)\}$ that we chose, for any $\alpha > 0$,

$$\lim A(k) |k|^\alpha = \infty.$$

If $\alpha < 1/2n$, we therefore find that

$$B(k) \log(1 + |k|)^{1/2} |k|^{-1/2n} \lesssim |k|^{-\alpha} \lesssim A(k).$$

Thus $d_A(\mu, \mu_0) \lesssim \varepsilon \lesssim \varepsilon_0$. Thus we conclude that

$$d((E, \mu), (\overline{E_{\varepsilon_0}}, \mu_0)) \lesssim \varepsilon_0,$$

which completes the proof of density since ε_0 was arbitrary. \square

This completes our proof that quasi-all (E, μ) have $\text{supp}(\mu) = E$, and such that E is an independent set. We note that the decay of the measure μ constructed using the probabilistic construction has a constant decay depending only on the number n of coefficients in the linear equation

$$m_1 x_1 + \cdots + m_N x_N.$$

Working with the sequence $A(k) = |k|^{-1/2n} \log(1 + |k|)^{1/2+\varepsilon}$, for any $\varepsilon > 0$, one can follow through the proof above to show that for quasi all elements $(E, \mu) \in \mathcal{G}_0$, $\text{supp}(\mu) = E$, and for any $m \in \mathbf{Z}^n$, $m_1 x_1 + \cdots + m_n x_n \neq 0$. In particular, this implies that E is a set with Fourier dimension $\geq 1/n$. This is far from the result which guarantees the existence of solutions of such an equation in sets with Fourier dimension greater than $2/n$, but it is a useful start. A slightly more complicated probabilistic construction improves this construction to obtain a set with Fourier dimension $1/(n-1)$ avoiding solutions to the equation $m_1 x_1 + \cdots + m_n x_n = 0$ for all non-zero $m \in \mathbf{Z}^n$.

Theorem 4.16. *Fix $n > 0$ and non-zero $m \in \mathbf{Z}^n$. Then for suitably large integers M , there are $X_1, \dots, X_M \in \mathbf{T}$ such that for any $j_1, \dots, j_n \in \{1, \dots, M\}$,*

$$|m_1 X_{j_1} + \cdots + m_n X_{j_n}| \geq \frac{1}{10M^{n-1/2}},$$

and if $\mu = M^{-1}(\delta_{X_1} + \cdots + \delta_{X_M})$, then for any $|k| \leq M^{10n}$,

$$|\hat{\mu}(k)| \leq 10n^{1/2} \log(M)^{1/2} M^{-1/2}.$$

Proof. Consider independent random variables X_1, \dots, X_M uniformly distributed on \mathbf{T} , and consider the measure

$$\mu = \delta_{X_1} + \cdots + \delta_{X_M}.$$

Let Z denotes the number of tuples $i \in \{1, \dots, M\}^n$ such that

$$|m_1 X_{i_1} + \cdots + m_n X_{i_n}| \leq \varepsilon.$$

Then $\mathbf{E}(Z) = 2M^n \varepsilon$, Markov's inequality implies that

$$\mathbf{P}(Z \geq 10M^n \varepsilon) \leq 1/5.$$

Setting $\varepsilon = 1/10M^{n-1/2}$ and $t = M^{1/2}$, we conclude

$$\mathbf{P}(Z \geq M^{1/2}) \leq 1/4.$$

For each k ,

$$\hat{\mu}(k) = \exp(2\pi i X_1) + \cdots + \exp(2\pi i X_M),$$

is a sum of M independant random variables, so we can apply Hoeffding's inequality to conclude

$$\mathbf{P}(|\hat{\mu}(k)| \geq \delta) \leq 2 \exp(-2\delta^2/M).$$

A union bound thus shows

$$\mathbf{P}(\text{for all } k \in [-K, K] : |\hat{\mu}(k)| \leq \delta) \geq 1 - 5K \exp(-2\delta^2/M).$$

In particular, if $K = M^{10n}$, and $\delta = 5n^{1/2}M^{1/2}\log(M)^{1/2}$, we conclude that

$$\mathbf{P}(\text{for all } |k| \leq M^{10n} : |\hat{\mu}(k)| \leq 5n^{1/2}M^{1/2}\log(M)^{1/2}) \leq 1/4.$$

Applying a union bound, we therefore conclude that there exists X_1, \dots, X_M such that there are at most $3M^{1/2}$ tuples i such that $|m_1 X_{i_1} + \cdots + m_n X_{i_n}| \leq 1/M^{n-1/2}$, and for all $|k| \leq M^{10n}$,

$$|\hat{\mu}(k)| \leq 5n^{1/2}M^{1/2}\log(M)^{1/2}.$$

If we let

$$I = \{1, \dots, M\} - \{i_1 : |m_1 X_{i_1} + \cdots + m_n X_{i_n}| \leq 1/M^{n-1/2}\},$$

This means that

$$\left\| \mu - \sum_{i \in I} X_i \right\|_{M(\mathbf{T})} \leq |I|^c \leq 3M^{1/2},$$

so for each $|k| \leq M^{10n}$, if $\nu = 1/|I| \sum_{i \in I} X_i$, and if M is sufficiently large, then

$$\begin{aligned} |\hat{\nu}(k)| &\leq \frac{1}{|I|} \left(3M^{1/2} + |\hat{\mu}(k)| \right) \\ &\leq \frac{1}{M - 3M^{1/2}} \left(3M^{1/2} + 5n^{1/2}M^{1/2}\log(M)^{1/2} \right) \\ &\leq 10n^{1/2}\log(M)^{1/2}M^{-1/2}. \end{aligned}$$

This completes the proof. □

In the last result, we were able to choose ε proportionally larger than in our previous construction. Thus we are free to convolve our measure with a less convoluted smooth function while still avoiding solutions to the linear equations. Going through the remainder of the construction as we have just done shows that for quasi-all pairs (E, μ) , $\text{supp}(\mu) = E$, and E has Fourier dimension greater than

$$\frac{1}{n - 1/2}.$$

Perform a decomposition on E_ε , writing it as the countable union of small boxes, which tensorize nicely? But the probability of a union of boxes is hard to calculate.