

Fractals Avoiding Fractal Sets

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Abstract

Hello, here is some text without a meaning. This text should show what a printed text will look like at this place. If you read this text, you will get no information. Really? Is there no information? Is there a difference between this text and some nonsense like “Huardest gefburn”? Kjift – not at all! A blind text like this gives you information about the selected font, how the letters are written and an impression of the look. This text should contain all letters of the alphabet and it should be written in of the original language. There is no need for special content, but the length of words should match the language.

Is it true that sets with large fractal dimension unavoidably contain certain patterns? The Lebesgue density theorem implies this is essentially true for sets with positive Lebesgue measure. But the case of fractal dimension is much more interesting. In [1], a set with full Hausdorff dimension is obtained avoiding translates. TODO: MORE EXAMPLES. In this paper, we construct explicit high dimensional sets avoiding ‘low dimensional’ patterns.

Theorem 1. *Suppose $Z \subset (\mathbf{R}^d)^n$ can be written as the countable union of compact sets with lower Minkowski dimension upper bounded by α . Then there exists a set $X \subset [0, 1]^d$ with*

$$\dim_{\mathbf{H}}(X) = \min \left(\frac{nd - \alpha}{n - 1}, d \right),$$

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

There are already generic pattern avoidance methods in the literature. But they rely on the assumption that Z is a smooth manifold, or an algebraic hypersurface. The novel feature of our method is that we focus on how the fractal dimension of the configuration is exploitable. We are therefore able to find sets avoiding configurations specified by arbitrarily irregular Z . Meanwhile, the Hausdorff dimension of X is competitive in relation to results obtained by more restricted techniques, and even recovers a selection of the avoidance results as special cases. We compare our method to them in more detail in Section 6.

The key idea to avoiding sparse configurations is a random mass equidistribution strategy. This occurs primarily in Section 2, when we solve a discrete

variant of the problem by finding a set minimizing the number of intersections with a discrete variant of Z . Finding an absolute minimizer to this problem is tricky, but a random set has a small number of intersections, which suffices for our purposes. By overlaying the solution to the discretized problems at a sequence of scales, we obtain X as a Cantor-type set, at the end of Section 3. Exploiting the equidistribution of mass at discrete scales, in Section 4, we are able to show the set X has the required Hausdorff dimension regardless of how fast our sequence of scales decay. Though this technique occurs implicitly in at least one other Hausdorff dimension calculation in the literature, as in [2], we do not believe equidistribution has been explicitly identified as a method to balance offsets from a rapid decay of scales in a Cantor set construction.

1 Notation and Terminology

- For a length L , $\mathcal{B}^d(L)$ denotes the partition of \mathbf{R}^d into the family of all half open cubes of sidelength L , with corners on the lattice $(L \cdot \mathbf{Z})^d$, i.e.

$$\mathcal{B}^d(L) = \{[a_1, a_1 + L) \times \cdots \times [a_d, a_d + L) : a_i \in L \cdot \mathbf{Z}\}.$$

- If $E \subset \mathbf{R}^d$, $\mathcal{B}^d(L, E)$ is the family of cubes in $\mathcal{B}^d(L)$ intersecting E , i.e.

$$\mathcal{B}^d(L, E) = \{I \in \mathcal{B}^d(L) : I \cap E \neq \emptyset\}.$$

- The *lower Minkowski dimension* of a compact set $E \subset \mathbf{R}^d$ is

$$\underline{\dim}_{\mathbf{M}}(E) = \liminf_{L \rightarrow 0} \frac{\log(\#\mathcal{B}^d(L, E))}{\log(1/L)}.$$

Thus there is $L_k \rightarrow 0$ with $\#\mathcal{B}^d(L_k, E) = (1/L_k)^{\underline{\dim}_{\mathbf{M}}(E) + o(1)}$.

- Adopting the terminology of [5], we say a collection of sets U_1, U_2, \dots is a *strong cover* of some set E if $E \subset \limsup U_k$, which means every element of E is contained in infinitely many of the sets U_k .
- Given a cube $I \in \mathcal{B}(L, dn)$, there are unique cubes $I_1, \dots, I_n \in \mathcal{B}(L, d)$ such that $I = I_1 \times \cdots \times I_n$. We say I is *non diagonal* if the intervals I_1, \dots, I_n are pairwise distinct.

2 Avoidance at Discrete Scales

We avoid Z by considering an infinite sequence of scales. At each scale, we solve a discretized version of the problem, and combining these solutions gives a solution to the original problem. This section describes the discretized avoidance technique. The technique is the *core* part of our construction, and the Hausdorff dimension we achieve is a direct result of our success in the discrete setting.

Fix two dyadic sidelengths $L > S$. In the fractal setting, Z is replaced by a union of sidelength S cubes, denoted K . Our goal is to take a set E , which is a union of sidelength L cubes, and carve out a union of sidelength S cubes F such that F^n is disjoint from the non-diagonal cubes of K .

In order to ensure the Hausdorff dimension calculations of X go smoothly, it is crucial in the discrete setting that the mass of F is spread uniformly over E . We can achieve this by trying to include a equal portion of mass in each sidelength R subcube of E , for some intermediary scale $L > R > S$. The next lemma shows that we can select a sidelength S cube from *almost* all of the sidelength R cubes.

Lemma 1. *Fix three dyadic lengths $L > R > S$. Let E be a union of cubes in $\mathcal{B}^d(L)$, and K a union of cubes in $\mathcal{B}^{dn}(S)$. Then there exists a subset F of E , which is a union of cubes in $\mathcal{B}^d(S)$, such that*

- (a) Avoidance: *For any distinct $I_1, \dots, I_n \in \mathcal{B}^d(S, F)$, $I_1 \times \dots \times I_n \notin \mathcal{B}^{dn}(S, K)$.*
- (b) Non Concentration: *$\mathcal{B}^d(S, F)$ contains at most one subcube of each cube in $\mathcal{B}^d(S, E)$.*
- (c) Equidistribution: *$\mathcal{B}^d(R, E)$ contains a subcube from all but at most $|K|R^{-dn}$ of the cubes in $\mathcal{B}^d(R, E)$.*

Proof. Form a random set U by selecting a sidelength S cube from each sidelength R cube uniformly at random. More precisely, set

$$U = \bigcup \{J_I : I \in \mathcal{B}^d(R, E)\},$$

where J_I is an element selected uniformly randomly from $\mathcal{B}^d(S, I)$. U certainly satisfies the equidistribution and non-concentration properties, but not the avoidance property. We will show that with non-zero probability, we can obtain the avoidance property by removing at most $|K|R^{-dn}$ intervals from U .

For any $I \in \mathcal{B}^d(S, E)$, $\mathbf{P}(I \subset U) = (S/R)^d$. Since any two elements of $\mathcal{B}^d(S, U)$ lie in distinct cubes of $\mathcal{B}^d(R)$, the only chance that a *non-diagonal* cube I in $\mathcal{B}^{dn}(S, K)$ is a subset of U^n is if I_1, \dots, I_n all lie in separate cubes of $\mathcal{B}^d(R)$. They each have an independent chance of being added to U , and so

$$\mathbf{P}(I \subset U^n) = \mathbf{P}(I_1 \subset U) \cdots \mathbf{P}(I_n \subset U) = (S/R)^{dn}.$$

If \mathcal{I} denotes the family of all non-diagonal cubes of $\mathcal{B}^{dn}(S, K)$ contained in U^n , then, letting I range over the non-diagonal cubes of $\mathcal{B}^{dn}(S, K)$, we find

$$\mathbf{E}(\#\mathcal{I}) = \sum_I \mathbf{P}(I \subset U^n) \leq |\mathcal{B}^{dn}(S, K)|(S/R)^{dn} = |K|R^{-dn}.$$

In particular, this means that out of all possible outcomes for the set U , there is at least one *particular* U_0 we can choose for which the corresponding set of cubes \mathcal{I}_0 satisfies $\#\mathcal{I}_0 \leq \mathbf{E}(\mathcal{I}) = |\mathcal{B}^{dn}(S, K)|(S/R)^{dn} = |K|R^{-dn}$.

We now define $F = U_0 - \{I_1 : I \in \mathcal{I}_0\}$. As a subset of U_0 , F inherits the non-concentration property. We have removed at most $|K|R^{-dn}$ intervals from

U_0 , and since U_0 contains an interval from *every* cube in $\mathcal{B}^d(R, E)$, F satisfies the equidistribution property. Finally, since we have removed a single side from every non-diagonal cube in U_0^n intersecting K , F satisfies the avoidance property. So our construction is complete. \square

Remark. *Our construction here uses randomness to count the number of intersections with K . Nonetheless, the set F we find is explicitly non-random – it is a set that exists, and can be found by iterating through the finitely many choices of F satisfying the avoidance and non-concentration properties. Thus the set X we obtain is explicit. It is not a random object.*

If Z has dimension α , we will be able to obtain that $|K| \leq 2^{dn} S^{dn-\gamma}$ for values γ converging to α in the limit. We also set R to be the closest dyadic number to S^λ for some $\lambda \in (0, 1)$. The size of λ is directly related to the Hausdorff dimension of our theorem (the larger the better!). The next corollary calculates how large λ can be if F must be equidistributed over a constant fraction of cubes in $\mathcal{B}^d(R, E)$. The error term $5A \log_S |E|$ in the corollary will be made insignificant by the rapid decay of S in the construction.

Corollary 1. *Consider the last lemma's setup, in addition to three additional parameters $\lambda \in (0, 1]$, $\gamma \in [d, dn)$, and A is a positive integer. Suppose $R = 2^{-\lfloor \lambda \log_2(1/S) \rfloor}$, $|E| \leq 1/2$, and $|K| \leq 2^{dn} S^{dn-\gamma}$. If*

$$0 < \lambda \leq \frac{dn - \gamma}{d(n-1)} - 5A \log_S |E|,$$

then F is equidistributed over all but a fraction $1/2^A$ of the cubes in $\mathcal{B}^d(R, E)$.

Proof. The inequality for λ implies

$$dn - \gamma - \lambda d(n-1) \geq 5d(n-1)A \log_S |E|.$$

Since R is within a factor of two from S^λ , we compute

$$\begin{aligned} \frac{|\{I \in \mathcal{B}^d(R, E) : \mathcal{B}^d(S, I) \cap \mathcal{B}^d(S, F) = \emptyset\}|}{|\mathcal{B}^d(R, E)|} &\leq \frac{|K|R^{-dn}}{|E|R^{-d}} \\ &\leq 2^{dn} |E|^{-1} S^{dn-\gamma} R^{-d(n-1)} \leq 2^{dn} |E|^{-1} S^{dn-\gamma} (S/2)^{-\lambda d(n-1)} \\ &\leq 2^{dn+\lambda d(n-1)} |E|^{-1} S^{5d(n-1)A \log_S |E|} \\ &= 2^{dn+\lambda d(n-1)} |E|^{5d(n-1)A-1} \leq 2^{dn+d(n-1)+1-5d(n-1)A} \\ &\leq 2^{2dn+[1-5d(n-1)]} / 2^A \leq 2^{5d-3dn+1} / 2^A \leq 2^{1-d} / 2^A \leq 1/2^A. \end{aligned}$$

The last inequality was obtained because $n \geq 2$. \square

Remark. *As mentioned, this discrete lemma is the core of our avoidance technique. The remaining argument is fairly modular, and can be applied with any other discrete avoidance technique to yield a solution to the fractal avoidance problem. Indeed, the remainder of our paper was inspired by the construction*

method of [2]. If for a special case of Z , one can improve the lemma so fewer intervals are discarded, then one can apply the remaining parts of our paper near verbatim to yield a set X with a larger Hausdorff dimension. For instance, a variation on the argument in [3] shows that if Z is a degree m algebraic hypersurface, and $K = \mathcal{B}(L, Z)$, then one can set $\lambda \approx 1/m$. Following through the remainder of the proof replicates Mathe's result.

3 Fractal Discretization

Now we apply the discrete technique we just described to obtain an actual fractal avoidance set. The fact that Z is the countable union of compact sets with Minkowski dimension α implies that we can find an efficient *strong cover* of Z by cubes restricted to lie at a sequence of dyadic scales L_k .

Lemma 2. *Let Z be a countable union of compact sets, each with lower Minkowski dimension at most α , and consider any positive sequence ε_k converging to zero. Then there is a decreasing sequence of lengths L_1, L_2, \dots , and $\mathcal{B}(L_k)$ sets Z_k such that Z is strongly covered by the sets Z_k and $|\mathcal{B}(Z_k, L_k)| \leq 2^{dn}/L_k^{\alpha+\varepsilon_k}$.*

Proof. Let Z be the union of sets Y_i with $\dim_{\mathbf{M}}(Y_i) \leq \alpha$ for each i . Consider any sequence of integers m_1, m_2, \dots which repeats each integer infinitely often. Given k , there are infinitely many lengths L with $\#(\mathcal{B}^{dn}(L, Y_{m_k})) \leq 1/L^{\alpha+\varepsilon_k}$. Replacing L with a dyadic number at most twice the size of L , there are infinitely many *dyadic* lengths L with $\#(\mathcal{B}^{dn}(L, Y_{m_k})) \leq 1/(L/2)^{\alpha+\varepsilon_k} \leq 2^{dn}/L^{\alpha+\varepsilon_k}$. In particular, we may fix a length L_k smaller than L_1, \dots, L_{k-1} . Then the union of the cubes in $\mathcal{B}^{dn}(L_k, Y_{m_k})$ forms the set Z_k . \square

Remark. *In the proof, we are free to make L_k arbitrarily small in relation to the previous parameters L_1, \dots, L_{k-1} we have chosen. For instance, later on when calculating the Hausdorff dimension, we will assume that $L_{k+1} \leq L_k^{k^2}$, and the argument above can be easily modified to incorporate this inequality.*

We can now construct X by avoiding the various discretizations of Z at each scale. The aim is to find a nested decreasing family of discretized sets X_k with $X = \lim X_k$. One condition guaranteeing that X solves the fractal avoidance problem is that X_k^n is disjoint from *non diagonal* cubes in Z_k .

Lemma 3. *If for each k , X_k^n avoids non-diagonal cubes in Z_k , then X solves the fractal avoidance problem for Z .*

Proof. Let $z \in Z$ be given with z_1, \dots, z_n are distinct. Set

$$\Delta = \{w \in (\mathbf{R}^d)^n : \text{there exists } i, j \text{ such that } w_i = w_j\}.$$

Then $d(\Delta, z) > 0$. The point z is covered by cubes in infinitely many of collections Z_{k_m} . For suitably large N , the cube I in $\mathcal{B}(L_{k_N})$ containing z is disjoint from Δ . But this means that I is non diagonal, and so $z \notin X_N^d$. In particular, z is not an element of X^n . \square

It is now simple to see how we must work at the discrete scales. First, we see $X_0 = [0, 1/2]^d$, so that $|X_0| \leq 1/2$. To obtain X_{k+1} from X_k , we apply the discrete argument. We set $E = X_k$ and $W = Z_{k+1}$, with scales $L = L_k$ and $S = L_{k+1}$. We know that we can choose $\gamma = \alpha + \varepsilon_k$, and also pick $R = R_{k+1}$ the closest dyadic number to L_{k+1}^λ , where

$$\lambda = \beta_{k+1} = \frac{dn - \alpha}{d(n-1)} - \frac{\varepsilon_{k+1}}{d(n-1)} - 10(k+1) \log_{L_{k+1}} |X_k|.$$

The discrete lemma then constructs a set F with F^n avoiding non diagonal cubes in Z_{k+1} , and containing a $\mathcal{B}(L_{k+1})$ subcube from all but a fraction $1/2^{2k+2}$ of the $\mathcal{B}(R_{k+1})$ cubes in I . We set $X_{k+1} = F$. Repeatedly doing this builds an infinite sequence of the X_k . Since X_k^n avoids Z_k , X is a solution to the fractal avoidance problem. It now remains to calculate the Hausdorff dimension of X .

4 Dimension Bounds

We now show that the set X has the expected Hausdorff dimension we need. At the discrete scale L_k , X looks like a $d\beta_k$ dimensional set. If the lengths L_k rapidly converge to zero, then we can ensure $\beta_k \rightarrow \beta$, where

$$\beta = \frac{dn - \alpha}{d(n-1)}.$$

Then, in the limit X looks $d\beta$ dimensional on the discrete scales, which is the Hausdorff dimension we want. It then suffices to interpolate this result to get a $d\beta$ dimensional behaviour at all intermediary scales. We won't be penalized here by making the gaps between discrete scales too large, because the uniform way that we have selected cubes in consecutive scales implies that between the scales L_k and L_{k+1}^β , X behaves like a full dimensional set. The remainder of this section fills in the details to this argument.

Lemma 4. $\beta_k \rightarrow \beta$.

Proof. It suffices to show that the error terms in β_k become negligible over time, i.e. we must show

$$\frac{\varepsilon_{k+1}}{d(n-1)} + 10(k+1) \log_{L_{k+1}} |X_k| = o(1).$$

Since $\varepsilon_{k+1} \rightarrow 0$, the term corresponding to it converges to zero for free. On the other hand, we need the lengths to tend to zero rapidly to make the other error term decay to zero. Since $L_{k+1} \leq L_k^{k^2}$, we find

$$(k+1) \log_{L_{k+1}} |X_k| \leq \frac{(k+1) \log L_k}{\log L_{k+1}} \leq \frac{(k+1) \log L_k}{k^2 \log L_k} = \frac{k+1}{k^2} = O(1/k).$$

Thus both components of the error term converge to zero as $k \rightarrow \infty$. \square

The most convenient way to look at the dimension of X at various scales is to use Frostman's lemma. To understand the behaviour of X , we construct a non-zero measure μ supported on X such that for all $\varepsilon > 0$, for all lengths L , and for all $I \in \mathcal{B}(L)$, $\mu(I) \lesssim_\varepsilon L^{d\beta-\varepsilon}$. We can then understand the behaviour of X at the scale L by looking at μ 's behaviour when restricted to cubes at the particular scale, i.e. cubes in $\mathcal{B}(L)$.

To construct a measure μ naturally reflecting the dimension of X , we rely on a variant of the mass distribution principle. This means we take a sequence of measures μ_k , supported on X_k , and then take a weak limit to form a measure μ . We initialize this construction by setting μ_0 to be the uniform measure on $X_0 = [0, 1/2]^d$. We then define μ_{k+1} , supported on X_{k+1} , by modifying the distribution of μ_k . First, we throw away the mass of the $\mathcal{B}(L_k)$ cubes I for which over half of the $\mathcal{B}(I, R_{k+1})$ cubes fail to contain a part of X_{k+1} . For the cubes I for which more than half of the cubes $\mathcal{B}(I, R_{k+1})$ contain a part of X_{k+1} , we distribute the mass of $\mu_k(I)$ uniformly over the subcubes of I in X_{k+1} . This gives a mass function μ_{k+1} . It is easy to see from the cumulative distribution functions of the μ_k that these measures converge to a function μ such that for any $I \in \mathcal{B}(L_k)$, $\mu(I) \leq \mu_k(I)$, which is useful for passing from bounds on the discrete measures to bounds on the final measure.

Lemma 5. *If $I \in \mathcal{B}(L_k)$, then*

$$\mu(I) \leq \mu_k(I) \leq 2^k \left[\frac{R_k R_{k-1} \dots R_1}{L_{k-1} \dots L_1} \right]^d.$$

Proof. Consider $I \in \mathcal{B}(L_{k+1})$, $J \in \mathcal{B}(L_k)$. If $\mu_k(I) > 0$, this means that J contains a $\mathcal{B}(L_k)$ cube in at least half of the $\mathcal{B}(R_N)$ cubes it contains. Thus the mass of J distributes itself evenly over at least $2^{-1}(L_{k-1}/R_k)^d$ cubes, which gives that $\mu_k(I) \leq 2(R_k/L_k)^d \mu_{k-1}(J)$. But then expanding this recursive inequality, using the fact that μ_0 has total mass one as a base case, we obtain exactly the result we need. \square

Corollary 2. *The measure μ is positive.*

Proof. To prove this result, it suffices to show that the total mass of μ_k is bounded below, independently of k . At each stage k , X_k consists of at most

$$\left[\frac{L_{k-1} \dots L_1}{R_k \dots R_1} \right]^d.$$

$\mathcal{B}(L_k)$ cubes. Since only a fraction $1/2^{2k+2}$ of the $\mathcal{B}(R_k)$ cubes do not contain an interval in X_{k+1} , it is only for at most a fraction $1/2^{2k+1}$ of the $\mathcal{B}(L_k)$ cubes that X_{k+1} fails to contain a $\mathcal{B}(L_{k+1})$ cube from more than half of the $\mathcal{B}(R_{k+1})$ cubes. But this means that we discard a total mass of at most

$$\left(\frac{1}{2^{2k+1}} \left[\frac{L_{k-1} \dots L_1}{R_k \dots R_1} \right]^d \right) \left(2^k \left[\frac{R_k \dots R_1}{L_{k-1} \dots L_1} \right]^d \right) \leq 1/2^{k+1}.$$

Thus

$$\mu_k(\mathbf{R}^d) \geq 1 - \sum_{i=0}^k \frac{1}{2^{i+1}} \geq 1/2.$$

This implies $\mu(\mathbf{R}^d) \geq 1/2$, and in particular, $\mu \neq 0$. \square

Ignoring all parameters in the inequality for I which depend on indices $< k$, we ‘conclude’ that $\mu_k(I) \lesssim R_k^d \lesssim L_k^{\beta_k d}$. The fact that $L_{k+1} \leq L_k^{k^2}$ has such a rapid decay essentially enables us to ignore quantities depending on previous indices, and obtain a true inequality.

Corollary 3. *For all $I \in \mathcal{B}(L_k)$, $\mu(I) \leq \mu_k(I) \lesssim L_k^{d\beta_k - k^{-1/2}}$.*

Proof. Given ε , we find

$$\begin{aligned} \mu_k(I) &\leq 2^k \left[\frac{R_k \dots R_1}{L_{k-1} \dots L_1} \right]^d \leq \left(\frac{2^{d+k}}{L_{k-1}^{d(1-\beta_{k-1})} \dots L_1^{d(1-\beta_1)}} \right) L_k^{d\beta_k} \\ &\leq \left(2^{d+k} L_k^\varepsilon / L_{k-1}^{d(k-1)} \right) L_k^{d\beta_k - \varepsilon} \leq \left(2^{d+k} L_{k-1}^{\varepsilon k^2 - d(k-1)} \right) L_k^{d\beta_k - \varepsilon}. \end{aligned}$$

The open bracket term decays as $k \rightarrow \infty$ so fast that it still tends to zero if ε is not fixed, but is instead equal to $k^{-1/2}$, giving the required inequality. \square

This is close to the cleanest expression of the $d\beta$ dimensional behaviour at discrete scales. To get a general inequality of this form, we use the fact that our construction distributes uniformly across all intervals.

Theorem 2. *If $L \leq L_k$ is dyadic and $I \in \mathcal{B}(L)$, then $\mu(I) \lesssim L^{d\beta_k - k^{-1/2}}$.*

Proof. We break our analysis into three cases, depending on the size of L in proportion to L_k and R_k :

- If $R_{k+1} \leq L \leq L_k$, we can cover I by $(L/R_{k+1})^d$ cubes in $\mathcal{B}(R_{k+1})$. For each of these cubes, because the mass is uniformly distributed over R_{k+1} cubes, we know the mass is bounded by at most $2(R_{k+1}/L_{k+1})^d$ times the mass of a $\mathcal{B}(L_k)$ cube. Thus

$$\begin{aligned} \mu(I) &\lesssim [(L/R_{k+1})^d][2(R_{k+1}/L_k)^d][L_k^{d\beta_k - k^{-1/2}}] \\ &\leq 2L^d / L_k^{d+k^{-1/2} - d\beta_k} \leq 2L^{d\beta_k - k^{-1/2}}. \end{aligned}$$

- If $L_{k+1} \leq L \leq R_{k+1}$, we can cover I by a single cube in $\mathcal{B}(R_{k+1})$. Each cube in $\mathcal{B}(R_{k+1}, d)$ contains at most one cube in $\mathcal{B}(L_{k+1}, d)$ which is also contained in X_{k+1} , so

$$\mu(I) \lesssim L_{k+1}^{d\beta_{k+1} - (k+1)^{-1/2}} \leq L^{d\beta_k - k^{-1/2}}.$$

- If $L \leq L_{k+1}$, there certainly exists M such that $L_{M+1} \leq L \leq L_M$, and one of the previous cases yields that

$$\mu(I) \lesssim 2L^{d\beta_M - M^{-1/2}} \leq 2L^{d\beta_k - k^{-1/2}}.$$

The three bulletpoints address all cases considered in the theorem. \square

To use Frostman's lemma, we need the result $\mu(I) \lesssim L^{d\beta_k - k^{-1/2}}$ for an *arbitrary* interval, not just one with $L \leq L_k$. But this is no trouble; it is only the behavior of the measure on arbitrarily small scales that matters. This is because if $L \geq L_k$, then $\mu(I)/L^{d\beta_k - k^{-1/2}} \leq 1/L_k^{d\beta_k - k^{-1/2}} \lesssim_k 1$, so $\mu(I) \lesssim_k L^{d\beta_k - k^{-1/2}}$ holds automatically for all sufficiently large intervals. Thus we have shown that $\dim_{\mathbf{H}}(X) \geq d\beta_k - k^{-1/2}$, and letting $k \rightarrow \infty$ gives $\dim_{\mathbf{H}}(X) \geq d\beta$. It is also easy to see X has *precisely* this dimension.

Theorem 3. $\dim_{\mathbf{H}}(X) = (dn - \alpha)/(n - 1)$.

Proof. X_k is covered by at most

$$\left[\frac{L_{k-1} \dots L_1}{R_k \dots R_1} \right]^d$$

sidelength L_k cubes. It follows that if $\gamma > \beta_k$, then

$$H_{L_k}^{d\gamma}(X) \leq \left[\frac{L_{k-1} \dots L_1}{R_k \dots R_1} L_k^\gamma \right]^d \lesssim \left[\frac{L_{k-1} \dots L_1}{R_{k-1} \dots R_1} L_k^{\gamma - \beta_k} \right]^d \leq L_k^{d(\gamma - \beta_k)}.$$

Since $L_k \rightarrow 0$ as $k \rightarrow \infty$, $H^\gamma(X) = 0$. Since γ was arbitrary, taking it to β allows us to conclude that $\dim_{\mathbf{H}}(X) \leq d\beta$. We have already justified that $\dim_{\mathbf{H}}(X) \geq d\beta$, and so $\dim_{\mathbf{H}}(X) = d\beta$. \square

5 Applications

Of course, our result already generalizes methods with interesting applications. But the most novel applications of our method occur when the configurations truly are a fractal set.

Example. Let $Y \subset \mathbf{R}^d$ be the countable union of sets with lower Minkowski dimension upper bounded by α . Then the set $Y_0 = \{(x, y) : x + y \in Y\}$ is a countable union of sets with lower Minkowski dimension upper bounded by $d + \alpha$. Applying our lemma then gives a set X with Hausdorff dimension $d - \alpha$ such that for any distinct $x_1, x_2 \in X$, $x_1 + x_2 \notin Y$. Modifying our construction slightly makes it possible to construct X with $X + X$ avoiding Y completely. Less elegantly, we can also consider

$$Y_1 = \{(x, y) : x + y \in Y\} \cup \{(x, y) : x \in Y/2\}$$

Then Y_1 is also the countable union of sets with lower Minkowski dimension bounded by $1 + \alpha$, and X avoiding Y_1 has $X + X$ disjoint from Y .

We have ideas on fusing our result with inspiration from the result of [3] to obtain the more impressive result which will show, given a set Y with fractal dimension α , how to construct a set X , which is a \mathbf{Q} vector space, disjoint from Y , with Hausdorff dimension $d - \alpha$. Thus given a \mathbf{Q} subspace V of \mathbf{R}^d , we can always find a complementary \mathbf{Q} vector space W with a complementary fractal dimension.

Example. In [2], one shows that we can find a dimension $1/2$ subset of any smooth curve avoiding isosceles triangles. Applying much the same techniques as in [2], but applying our result (though we do not need to be as careful, since we do not care about smoothness), we can extend this result to find a dimension $1/2$ subset of any bi-Lipschitz curve avoiding isosceles triangles.

Example. Suppose we have a fractal set Y , together with an orthogonal projection π such that $\pi(Y) = \mathbf{R}^d$. Then we can form the set

$$\{(x_0, x_1, x_2) : \text{There is } x'_0, x'_1, x'_2 \in Y \text{ s.t. } \pi(x'_i) = x_i, d(x', y') = d(x', z')\}$$

Shall we work on the Koch Snowflake for explicitness?

6 Relation to Literature, and Future Work

TODO: MOVE THIS This result is part of a growing body of work finding general methods to avoid patterns with particular geometric features. In [2] and [3], sets with large Hausdorff dimension are constructed avoiding patterns specified by smooth low-variable functions, and low degree polynomials.

The technical skeleton of our construction are heavily modelled after [2]. Reading this paper in tandem with ours provides an interesting contrast between the techniques of the function oriented configuration avoidance result, and the fractal avoidance result we use. Because of its heavy influence on our result, we begin our discussion of the literature with an in depth comparison of our method to theirs.

Our result is a direct generalization of the main result of [2], which says that if $Z \subset (\mathbf{R}^d)^n$ is a smooth surface of dimension $nd - d$, then we can find X with dimension $(n - 1)^{-1}$ solving the fractal avoidance problem. Of course, such a Z has Minkowski dimension $nd - d$, and our result achieves the same dimension for X . In response to [2], our result says that the only really necessary feature of a smooth hypersurface to the avoidance problem, aside from other geometric features, is its dimension. Not only is our result more flexible, enabling the surface Z to have non smooth points, but we can also take advantage of the fact that the surface might have dimension different from $nd - d$. Better yet, we can ‘thicken’ or ‘thin’ Z by slightly increasing or decrease the Minkowski dimension, while stably affecting the Hausdorff dimension of the solution X we construct.

The technique leading to this generalization can be compared to a phenomenon that has recently been noticed in the discrete setting, i.e. [4]. There, certain combinatorial problems can be rephrased as abstract problems on hypergraphs, and by doing this one can often generalize the solutions of these

problems into analogues on ‘sparse versions’ of these hypergraphs. One can see our result as a continuous analogue to this phenomenon, where sparsity is represented by the dimension of the set Z we are trying to avoid. One can even view Lemma 1 as a solution to a problem about independent sets in hypergraphs. In particular, we can form a hypergraph by taking the intervals $\mathcal{B}(F, S)$ as vertices, and adding an edge (I_1, \dots, I_n) between n distinct cubes if $I_1 \times \dots \times I_n$ intersects W . Then the union of an independent set of cubes in this graph is precisely a set F with F^n disjoint except on the discretization of the diagonal. And so the goal of Lemma 1 is essentially to find a ‘uniformly chosen’ independent set in this graph. Thus we even applied the discrete phenomenon at many scales to obtain the continuous version of the phenomenon.

A useful technique used in [2], and its predecessor [1], is a Cantor set construction ‘with memory’; a queue in their construction algorithm allows storage of particular configurations, to be retrieved and avoided at a much, much later step of the building process. The fact that our result is more general, yet we can discard the queueing method from our proof, is an interesting anomaly. Adding memory to the queueing set is certainly an important trick to remember when thinking of new constructions for fractal avoiding sets. It enables one to restrict the requirements of an analogy to Lemma 1 from carving out an avoiding set F from a single set E , to carving F_1, \dots, F_n out of disjoint sets E_1, \dots, E_n , such that $F_1 \times \dots \times F_n$ avoids W . Nonetheless, it makes the construction much more complicated to describe, which makes understanding dimension bounds slightly more complicated, because it’s hard to ‘grasp’ precisely what configuration we are avoiding at each step of the construction. The fact that our algorithm is more general than [2], yet we can discard the queueing method, is an interesting anomaly. We have ideas on how to exploit the fact that we do not use queueing to generalize our theorem to much more wide family of ‘dimension α ’ sets Z , which we plan to publish in a later result.

Aside from [2], another paper that takes the perspective of solving a generic fractal avoidance problem is [3], who finds a solution X to an avoidance problem with Z a degree k hypersurface with Hausdorff dimension d/k . If $k \geq n - 1$, then our result does better than Mathe’s result, so where Mathe’s result excels is when Z is a low dimensional hypersurface. Just like how the result of this paper is a sparse analogue of [2], we would like to publish a follow up result giving a sparse analogue to [3]. Just as our result is obtained by assuming Z is covered by a sparse family of cubes, a sparse analogue of [3] would give a result if Z is covered by a sparse family of thickened varieties from a pencil of low degree surfaces. We already have ideas we are refining on how to achieve this.

References

- [1] Tamás Keleti *A 1-Dimensional Subset of the Reals that Intersects Each of its Translates in at Most a Single Point*
- [2] Robert Fraser, Malabika Pramanik *Large Sets Avoiding Patterns*

- [3] A. Mathé *Sets of Large Dimension Not Containing Polynomial Configurations*
- [4] József Balogh, Robert Morris, Wojciech Samotij *Independent Sets in Hypergraphs*
- [5] Nets Hawk Katz, Terence Tao *Some connections between Falconer's distance set conjecture, and sets of Furstenberg type*