

# Universal Sets for Projections

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- 1 Projections in geometric measure theory
- 2 Effective dimension
- 3 Some proof ideas

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1 Projections in geometric measure theory

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**The basic question:** How large are sets?

- Usually, by size we mean measure or some notion of dimension (e.g. Hausdorff, packing, box-counting, Assouad... ).
- More precisely, what can we say about sets that we know have a certain geometric property, or that are obtained by some natural geometric operation on a set with known size?

# Covers and packings

Let  $E \subset \mathbb{R}^n$  and  $\{B_i\}_{i \in \mathbb{N}}$  be a collection of open balls in  $\mathbb{R}^n$ .

We call  $\{B_i\}_{i \in \mathbb{N}}$  a  $\delta$ -cover for  $E$  if

- $E \subseteq \bigcup_{i=1}^{\infty} B_i$
- $\text{diam}(B_i) \leq \delta$

We call  $\{B_i\}_{i \in \mathbb{N}}$  a  $\delta$ -packing for  $E$  if

- The balls are pairwise disjoint
- The balls have centers in  $E$
- $\text{diam}(B_i) \leq \delta$

## Hausdorff dimension

$$\mathcal{H}_\delta^s(E) = \inf_{\delta\text{-covers}} \left\{ \sum_{i=1}^{\infty} \text{diam}(B_i)^s \right\}$$

$$\mathcal{H}^s(E) = \lim_{\delta \rightarrow 0^+} \mathcal{H}_\delta^s(E)$$

$$\dim_H(E) = \inf \{s : \mathcal{H}^s(E) = 0\}$$

## Packing dimension

$$\bar{\mathcal{P}}_\delta^s(E) = \sup_{\delta\text{-packings}} \left\{ \sum_{i=1}^{\infty} \text{diam}(B_i)^s \right\}$$

$$\bar{\mathcal{P}}^s(E) = \lim_{\delta \rightarrow 0^+} \bar{\mathcal{P}}_\delta^s(E)$$

$$\mathcal{P}^s(E) = \inf \left\{ \sum_{i=1}^{\infty} \bar{\mathcal{P}}^s(E_i) : E \subseteq \bigcup_{i=1}^{\infty} E_i \right\}$$

$$\dim_P(E) = \inf \{s : \mathcal{P}^s(E) = 0\}$$

“Regularity” refers to a set looking the same (or at least the same size) at different scales.

- $E$  is **weakly regular** if  $\dim_P(E) = \dim_H(E)$
- $E$  is  **$\alpha$ -AD regular** if there exists some  $C$  such that

$$C^{-1}r^\alpha \leq \mathcal{H}^\alpha(E \cap B(x, r)) \leq Cr^\alpha, \quad x \in A, 0 < r < \text{diam}(A)$$

AD-regularity can be thought of as a generalization of self-similarity. Weak regularity is in turn a generalization of AD regularity.

# Size of projections

We work in  $\mathbb{R}^2$  and consider projections onto lines. In particular, for  $e \in S^1$ , let  $p_e E = \{x \cdot e : x \in E\}$

**Question:** How does the Hausdorff dimension of a set change under projections?

We have the upper bound

$$\dim_H(p_e E) \leq \min\{1, \dim_H(E)\}$$

**Theorem (Marstrand, 1954)**

*Let  $E \subseteq \mathbb{R}^2$  be Borel. For Lebesgue almost every  $e \in S^1$ ,*

$$\dim_H(p_e E) = \min\{1, \dim_H(E)\}$$



# Universal sets

Call  $e \in S^1$  **maximal** for  $E$  if

$$\dim_H(p_e E) = \min\{1, \dim_H(E)\}.$$

Suppose  $\mathcal{C}$  is some class of subsets of  $\mathbb{R}^2$ , i.e. the Borel sets, the weakly regular sets, the AD regular sets...

Call a set of directions  $D \subseteq S^1$  such that every  $E \in \mathcal{C}$  has a maximal direction in  $D$  a  **$\mathcal{C}$ -universal** set.

Are there small universal sets?

# Small universal sets

All the results apply to sets in  $\mathbb{R}^2$

Theorem (F. and Stull, 2024)

*The class of sets with optimal oracles has a Lebesgue measure zero universal set*

Theorem (F. and Stull, 2024)

*For any  $\varepsilon > 0$ , the class of weakly regular sets has a Hausdorff dimension  $\varepsilon$  universal set.*

Theorem (F. and Stull, 2024)

*The class of AD regular sets has a Hausdorff dimension 0 universal set.*

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# Complexity of points in Euclidean space

Fix a universal prefix-free oracle Turing machine  $U$ . Given  $A \subseteq \mathbb{N}$ , the (prefix-free) Kolmogorov complexity of a string  $\sigma$  relative to  $A$  is

$$K^A(\sigma) = \min\{|\pi| : U^A(\pi) = \sigma\}$$

We can encode rational vectors  $q \in \mathbb{R}^n$  as binary strings, and hence can talk about  $K^A(q)$ . This in turn allows us to define the complexity of *arbitrary* points in  $\mathbb{R}^n$  at any given precision.

$$K_r^A(x) = \min\{K^A(q) : q \in B_{2^{-r}}(x)\}$$

## Definition

The effective Hausdorff dimension of a point  $x \in \mathbb{R}^n$  relative to an oracle  $A \subseteq \mathbb{N}$  is given by

$$\dim^A(x) = \liminf_{r \rightarrow \infty} \frac{K_r^A(x)}{r}$$

## Definition

The effective packing dimension of a point  $x \in \mathbb{R}^n$  relative to an oracle  $A \subseteq \mathbb{N}$  is given by

$$\text{Dim}^A(x) = \limsup_{r \rightarrow \infty} \frac{K_r^A(x)}{r}$$

# The point-to-set principle

Effective dimension is directly related to classical dimension through the following “point-to-set” principle(s):

Theorem (J. Lutz and N. Lutz, 2015)

For all  $E \subset \mathbb{R}^n$ ,

$$\dim_H(E) = \min_{A \subseteq \mathbb{N}} \sup_{x \in E} \dim^A(x)$$

and

$$\dim_P(E) = \min_{A \subseteq \mathbb{N}} \sup_{x \in E} \text{Dim}^A(x)$$

# Points in regular sets

Let  $E$  be weakly regular. If  $A$  is the join of a Hausdorff and a packing oracle for  $E$ , for any  $\varepsilon > 0$ , there is some  $x \in E$  such that

$$\text{Dim}^A(x) - \dim^A(x) < \varepsilon$$

If  $E$  is AD-regular, we have something stronger: call a *point*  $x \in \mathbb{R}^n$   $\alpha$ -AD regular with respect to an oracle  $A \subseteq \mathbb{N}$  if there exists some  $C$  such that

$$\alpha r - C \log r \leq K_r^A(x) \leq \alpha r + C \log r$$

## Proposition

If  $E$  is compact and  $\alpha$ -AD regular, then there exists an oracle  $A$  relative to which  $\mathcal{H}^\alpha$ -almost every point in  $E$  is  $\alpha$ -AD regular.

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# The reduction

For any  $E$  in our class, it suffices to find some  $e$  in  $D$  with the following property: for every  $\varepsilon > 0$ , there exists some  $x \in E$  such that

$$\dim^B(p_e x) \geq \min\{1, \dim_H(E)\} - \varepsilon$$

where  $B$  is a Hausdorff oracle for  $p_e E$ .

To show this bound holds, we need a few assumptions.

- $e$  has high complexity at certain precisions, and low complexity at other precisions. In particular,  $e$  is the result of appropriately adding 0s to a ML random.
- $A$ , the Hausdorff oracle for the set  $E$ , does not help in the computation of  $e$
- Oracle access to  $e$  does not help in the computation of  $x$

# Partitioning

Let a sequence of precisions  $1 = r_0, r_1, r_2, \dots, r_m = r$  be given. Then

$$\begin{aligned}K_r^A(x) &= \sum_{i=1}^m \left( K_{r_i}^A(x) - K_{r_{i-1}}^A(x) \right) + K_1^A(x) \\ &\approx \sum_{i=1}^m K_{r_i, r_{i-1}}^A(x)\end{aligned}$$

Let  $x \in \mathbb{R}^2$  and  $a \leq b$ . We say that  $[a, b]$  is  $(\sigma, c)$ -teal if

$$K_{b,s}^A(x | x) \leq \sigma(b - s) + c \log b,$$

for all  $a \leq s \leq b$ . We say that  $[a, b]$  is  $(\sigma, c)$ -yellow if

$$K_{s,a}^A(x | x) \geq \sigma(s - a) - c \log b,$$

for all  $a \leq s \leq b$ .

## Lemma

Let  $x \in \mathbb{R}^2$ ,  $e \in \mathcal{S}^1$ ,  $c \in \mathbb{N}$ ,  $\sigma \in \mathbb{Q} \cap (0, 1]$ ,  $A \subseteq \mathbb{N}$  and  $a < b \in \mathbb{R}_+$ . Suppose that  $b$  is sufficiently large (depending on  $e$ ,  $x$ , and  $\sigma$ ) and  $K_{s,b}^A(e \mid x) \geq s - c \log b$ , for all  $s \leq b - a$ . Then the following hold.

- 1 If  $[a, b]$  is  $(\sigma, c)$ -yellow,

$$K_{b,b,b,a}^A(x \mid p_e x, e, x) \leq K_{b,a}^A(x \mid x) - \sigma(b - a) + O_c(\log b)^2.$$

- 2 If  $[a, b]$  is  $(\sigma, c)$ -teal,

$$K_{b,b,b,a}^A(x \mid p_e x, e, x) \leq O_c(\log b)^2.$$

# Symmetry of information

Recall, we are interested in lower bounding the quantity  $K_r^{A,B,e}(p_e x)$ . Using symmetry of information and the assumptions on our points, we have

$$\begin{aligned} K_r^A(x | e, p_e x) &\geq K_r^{A,e}(x | p_e x) - O(\log r) \\ &\geq K_r^{A,B,e}(x | p_e x) - O(\log r) \\ &= K_r^{A,B,e}(x, p_e x) - K_r^{A,B,e}(p_e x) - O(\log r) \\ &\geq K_r^{A,B,e}(x) - K_r^{A,B,e}(p_e x) - O(\log r) \\ &\geq K_r^A(x) - K_r^{A,B,e}(p_e x) - \text{small error} \end{aligned}$$

Then, we can use the upper bound on  $K_r^A(x | e, p_e x)$  that comes from partitioning.

Thank you!