

Dimension of diagonal self-affine measures with overlaps

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Self-affine fractals

- *Affine iterated function system (IFS)*: $\Phi = \{\varphi_i\}_{i \in \Lambda}$ with $|\Lambda| < \infty$ and

$$\varphi_i(x) = A_i x + t_i \quad \text{for } x \in \mathbb{R}^d,$$

where $A_i \in \text{GL}_d(\mathbb{R})$, $\|A_i\| < 1$ and $t_i \in \mathbb{R}^d$.

- *Self-affine/Limit set*: the unique nonempty compact set $K \subset \mathbb{R}^d$ such that

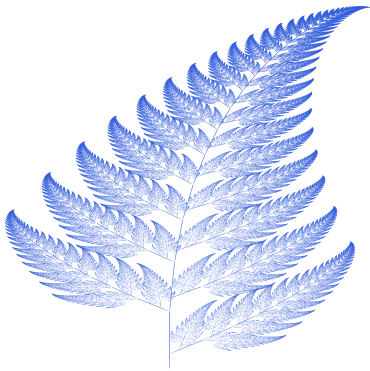
$$K = \bigcup_{i \in \Lambda} \varphi_i(K).$$

- *Self-affine/Stationary measure*: the unique Borel probability measure μ on \mathbb{R}^d such that

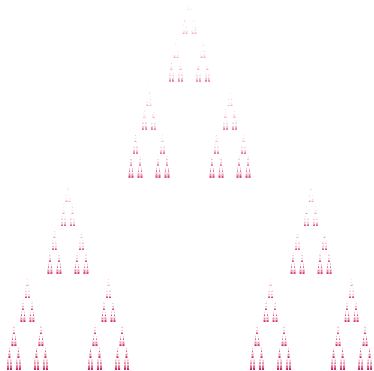
$$\mu = \sum_{i \in \Lambda} p_i \cdot \varphi_i \mu,$$

where $p = (p_i)_{i \in \Lambda}$ is a given probability vector.

- We call K and μ *self-similar* if φ_i are similarities, i.e., $A_i \in \mathbb{R} \cdot \text{O}(d)$.
- We call K and μ *diagonal* if A_i are diagonal matrices.



Barnsley fern



Hironaka curve

Question

How large are K and μ in terms of **dimensions**?

Which dimensions?

The Hausdorff dimension $\dim_{\text{H}} K$ and the (exact) dimension $\dim \mu$:

$$\dim \mu = \lim_{r \rightarrow 0} \frac{\log \mu(B(x, r))}{\log r} \quad \text{for } \mu\text{-a.e. } x.$$

(D.-J. Feng-Hu (2009), Bárány-Käenmäki (2017), D.-J. Feng (2023))

Natural **upper bounds** are defined using *singular value function*:

$$\phi^s(A) = \begin{cases} \sigma_1(A) \cdots \sigma_{\lfloor s \rfloor}(A) \sigma_{\lceil s \rceil}(A)^{s - \lfloor s \rfloor} & \text{if } 0 \leq s \leq d; \\ |\det A|^{s/d} & \text{if } s > d, \end{cases}$$

where $\sigma_1(A) \geq \cdots \geq \sigma_d(A)$ are singular values of $A \in \text{GL}_d(\mathbb{R})$.

Affinity dimension $\dim_A \Phi$ by Falconer ('88): the unique $s \geq 0$ such that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \log \sum_{x_1 \cdots x_n \in \Lambda^n} \phi^s(A_{x_1} \cdots A_{x_n}) = 0.$$

Lyapunov dimension $\dim_L(\Phi, p)$: the unique $t \geq 0$ such that

$$\sum_{i \in \Lambda} -p_i \log p_i + \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{x_1 \cdots x_n \in \Lambda^n} p_{x_1} \cdots p_{x_n} \log \phi^t(A_{x_1} \cdots A_{x_n}) = 0.$$

Relations among dimensions

$$\begin{array}{ccccc}
 \Lambda^{\mathbb{N}} & \dim_L(\Phi, p) & \leq & \dim_A \Phi & \\
 \Pi \downarrow & \vee I & & \vee I & \\
 \mathbb{R}^d & \dim \mu & \leq & \dim_H K & \leq d \\
 P_V \downarrow & \vee I & & \vee I & \\
 V & \text{"dim"} P_V \mu & \leq & \dim_H P_V(K) & \leq m
 \end{array}$$

- *Coding map*: $\Pi((x_n)_{n \in \mathbb{N}}) = \lim_{n \rightarrow \infty} \varphi_{x_1} \circ \cdots \circ \varphi_{x_n}(0)$.
- $K = \Pi(\Lambda^{\mathbb{N}})$ and $\mu = \Pi\beta$, where $\beta := p^{\mathbb{N}}$ is the Bernoulli measure.
- P_V denotes orthogonal projection onto m -dim subspace $V \leq \mathbb{R}^d$.

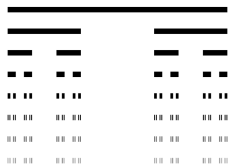
Much research aims to show “=” above across various settings.

Main difficulties: overlaps, nonconformality, saturation...

Overlaps by examples on the line

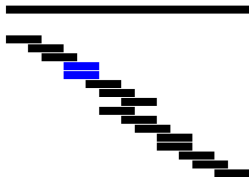
For $0 < \lambda < 1$, consider $\Phi_\lambda = \{\lambda x \pm 1\}$ on \mathbb{R} and $p = (1/2, 1/2)$.

- ($\lambda = 1/3$) Cantor-Lebesgue measure ν :



$$\dim \nu = \frac{\log 2}{\log 3}$$

- ($\lambda > 1/2$) Bernoulli convolutions μ_λ :



$$\dim \mu_\lambda = ?$$

Conjecture (exact overlaps conjecture)

If Ψ on \mathbb{R} generates a *free* semigroup, then $\dim \mu = \min \{1, \dim_L(\Psi, p)\}$.

Research target

- (Falconer '88, Solomyak '94, Jordan-Pollicott-Simon 2007) Assume $\|A_i\| < 1/2$. Then for **Lebesgue almost all translations** $\mathbf{t} = \{t_i\}_{i \in \Lambda}$,
(1) $\dim_{\text{H}} K_{\mathbf{t}} = \min\{d, \dim_A \Phi\}$ and $\dim \mu_{\mathbf{t}} = \min\{d, \dim_L(\Phi, \rho)\}$.
- An amount of progress in understanding **typical** self-affine fractals.

Reveal the **generic** phenomenon and exhibit **new** phenomena.

Want **explicit** examples. The **more**, the better.

Research target

Find **verifiable** and **mild** conditions for **(1)** to hold.

Exact overlaps conjecture is within this theme.

Exponential separation

Given $\psi_1, \psi_2: \mathbb{R} \rightarrow \mathbb{R}$ with $\psi_i(x) = s_i x + b_i$ for $i = 1, 2$, define

$$d(\psi_1, \psi_2) := \begin{cases} \infty & \text{if } s_1 \neq s_2; \\ |b_1 - b_2| & \text{otherwise.} \end{cases}$$

For an IFS $\Psi = \{\psi_i\}_{i \in \Lambda}$ on \mathbb{R} and $n \in \mathbb{N}$, define

$$\Delta_n(\Psi) = \min\{d(\psi_u, \psi_v) : u \neq v \in \Lambda^n\},$$

where $\psi_{u_1 \dots u_n} = \psi_{u_1} \circ \dots \circ \psi_{u_n}$.

Definition (exponential separation)

We call Ψ *exponentially separated* if there exists $c > 0$ such that $\Delta_n(\Psi) > c^n$ for infinitely many $n \in \mathbb{N}$.

- If Ψ is defined by algebraic numbers and has no exact overlaps, then Ψ is exponentially separated.
- Exponential separation allows substantial overlaps and is extremely mild in the sense of small exceptions.

Self-similar:

Hochman (2014): If $d = 1$ and Φ is exponentially separated, then

$$(M) \quad \dim \mu = \min \{d, \dim_L(\Phi, p)\}$$

$$(S) \quad \dim K = \min \{d, \dim_A \Phi\}.$$

Towards exact overlaps conjecture: Varjú (2019), Rapaport (2022), Rapaport-Varjú (2024), D.-J. Feng-F. (2024+) ...

Hochman (2017): Self-similar fractals in \mathbb{R}^d .

Strongly irreducible and proximal:

For (M): Bárány-Hochman-Rapaport (2019) for $d = 2$ and SSC;
Hochman-Rapaport (2022) for $d = 2$, nontrivial, and exp. sep.;
Rapaport (2024) for $d = 3$ and SSC.

From (M) to (S): Morris-Shmerkin (2019) for $d = 2$,
Morris-Sert (2023+) for $d \geq 3$.

Diagonal:

Various carpet-like cases: Bedford, McMullen, Przytcki-Urbanski, Gatzouras-Lalley, Kenyon-Peres, Barański, Feng-Wang, Fraser...

Some overlapping cases: Fraser-Shmerkin, Bárány-Rams-Simon...

Consider a **diagonal** IFS $\Phi = \{\varphi_i(x) = A_i x + t_i\}_{i \in \Lambda}$ on \mathbb{R}^d :

$$A_i = \text{diag}(r_{i,1}, \dots, r_{i,d}), \quad 0 < |r_{i,j}| < 1 \quad \text{and} \quad t_i = (t_{i,1}, \dots, t_{i,d}).$$

For $1 \leq j \leq d$, define

- the j th Lyapunov exponent: $\chi_j = \sum_{i \in \Lambda} -p_i \log |r_{i,j}|$.
- the induced IFS on j th axis: $\Phi_j = \{x \mapsto r_{i,j}x + t_{i,j}\}_{i \in \Lambda}$.

Theorem (Rapaport 2023+)

If for $1 \leq j_1 < j_2 \leq d$ there is $i \in \Lambda$ with $|r_{i,j_1}| \neq |r_{i,j_2}|$, and Φ_j is exponentially separated for all j , then $\dim_{\text{H}} K_{\Phi} = \min \{d, \dim_{\text{A}} \Phi\}$.

Rapaport deduces it from his analogous result for diagonal self-affine **measures**, under the additional assumption:

$$(\star) \quad \exists \mathbf{c} \in \mathbb{R}^d, \forall i \in \Lambda: (\log |r_{i,1}|, \dots, \log |r_{i,d}|) \in \mathbb{R} \mathbf{c}.$$

In particular, (\star) holds when A_i is the same for $i \in \Lambda$.

The argument **crucially** relies on (\star) . (to be discussed later)

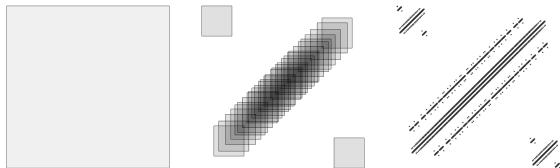
Theorem (F. 2025+)

Suppose

- 1 $\chi_{j_1} \neq \chi_{j_2}$ for $1 \leq j_1 < j_2 \leq d$.
- 2 Φ_j is exponentially separated for $1 \leq j \leq d$.

Then $\dim \mu = \min \{d, \dim_L(\Phi, p)\}$.

- 1 is necessary by **saturation**: e.g.,



- 2 is expected to be relaxed by **no exact overlaps**. When $d = 1$, it is the **exact overlaps conjecture**, which is famous and open.

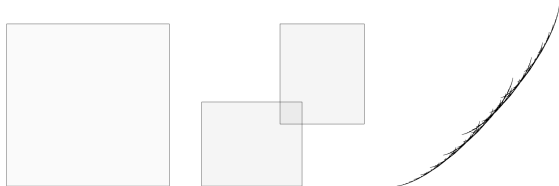
- 1 Determine $\dim \mu$ when Φ is defined by algebraic parameters without exact overlaps.
- 2 Dimension for generic systems with explicit set of exceptions which is small in terms of packing dimension.
- 3 Determine the ergodic measures μ on K such that $\dim \mu = \dim_{\text{H}} K$.
- 4 Dimension of orthogonal projections of certain overlapping self-affine sets.

(A concrete example is in the next page to illustrate the last two applications:)

A concrete example

Consider the IFS on \mathbb{R}^2 :

$$\Phi = \left\{ \begin{pmatrix} x \\ y \end{pmatrix} \mapsto \begin{pmatrix} \frac{5}{7}x \\ \frac{3}{5}y \end{pmatrix}, \begin{pmatrix} x \\ y \end{pmatrix} \mapsto \begin{pmatrix} \frac{3}{5}x \\ \frac{5}{7}y \end{pmatrix} + \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}.$$



- $\dim_A \Phi = s \approx 1.66$, where $\frac{5}{7} \left(\frac{3}{5}\right)^{s-1} + \frac{3}{5} \left(\frac{5}{7}\right)^{s-1} = 1$.
- There are exactly **two** ergodic measures μ_1, μ_2 which are **self-affine** such that

$$\dim \mu_1 = \dim \mu_2 = \dim_H K = \dim_A \Phi.$$

- (Pyörälä (2025) + above) For **all** 1-dim subspace $V < \mathbb{R}^2$,
 $\dim_H P_V(K) = 1$.

About the proof

- Why assume condition (★)?

To establish the **uniformness** of self-affine measures across scales, which is one of the key ingredients for proving **entropy increase**.

- The **disintegration**: Fix $N \in \mathbb{N}$. Partition $\Lambda^{\mathbb{N}}$ into \mathcal{A} such that

$$\mathcal{A}(x) = \mathcal{A}(y) \iff A_{x_{kN+1}} \cdots A_{x_{(k+1)N}} = A_{y_{kN+1}} \cdots A_{y_{(k+1)N}} \text{ for } k \geq 0.$$

Let $\{\beta^\omega\}_{\omega \in \Omega}$ be the disintegration of β w.r.t. \mathcal{A} . Then $\beta = \int \beta^\omega d\mathbf{P}(\omega)$.

Define $\mu^\omega = \Pi \beta^\omega$. Applying Π gives $\mu = \int \mu^\omega d\mathbf{P}(\omega)$.

Saglietti-Shmerkin-Solomyak (2018), Galicer-Saglietti-Shmerkin-Yavicoli (2016)

- **Exact dimensionality** and **Ledrappier-Young formula** for random measures:

D.-J. Feng (2023) + Falconer-Jin (2014) + handle multiple disintegrations

It also holds for partitions like $\bigvee_{i=0}^{\infty} \sigma^{-iN} \Gamma_N$ for any partition Γ_N of Λ^N .

- **Entropy increase** for random measures:

Strategy: Rapaport (2023+)

Difficulty: Random measures μ^ω are merely *dynamically self-affine* and nonconformal partitions $\mathcal{E}_n^\omega := A^{\omega|n} \mathcal{D}$ depend on ω .

Method: Feel multiscale estimates and use dynamics on (Ω, \mathbf{P}, T) .

Exact dimensionality for random measures

Theorem (Exact dimensionality for disintegrations)

*There exists $\dim \mathcal{A} \geq 0$ such that for \mathbf{P} -a.e. ω , μ^ω is *exact dimensional* with dimension given by $\dim \mathcal{A}$ satisfying a *Ledrappier-Young formula*.*

$$\dim \mathcal{A} = \sum_{j=1}^d \frac{h_{[j-1]}^{C, \mathcal{A}} - h_{[j]}^{C, \mathcal{A}}}{\chi_j}.$$

For example: $0 < \text{z-axis} < \text{xz-plane} < \mathbb{R}^3$.

Entropy increase for random measures

Theorem (Entropy increase for random measures)

Suppose $\dim \mathcal{A} < d$ and $\dim \pi_J \mathcal{A} = |J|$ for each $J \subsetneq [d]$. For $\varepsilon \in (0, 1)$ there exists $\delta = \delta(\varepsilon) > 0$ such that: Let $\eta \in (0, 1)$ be with $\varepsilon^{-1} \ll \eta^{-1}$. There exists $\Omega' \subset \Omega$ with $\mathbf{P}(\Omega') > 1 - \eta$ so that for $\omega \in \Omega'$ and $n \in \mathbb{N}$ with $\eta^{-1} \ll n$: Let $\theta \in \mathcal{M}_c(\mathbb{R}^d)$ with $\text{diam}(\text{supp } \theta) \leq 1/\varepsilon$ and $\frac{1}{n}H(\theta, \mathcal{E}_n^\omega) > \varepsilon$. Then

$$\frac{1}{n}H(\theta * \mu^\omega, \mathcal{E}_n^\omega) \geq \frac{1}{n}H(\mu^\omega, \mathcal{E}_n^\omega) + \delta.$$

Thank you for listening!