

# Lipschitz-Killing curvatures for different classes of fractals

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# 1. Minkowski and S-contents, extension to curvatures

For a compact set  $K \subset \mathbb{R}^d$  and  $r > 0$  the  $r$ -parallel set is given by

$$K(r) := \{x \in \mathbb{R}^d : \min_{y \in K} |x - y| \leq r\},$$

and for  $0 \leq D \leq d$ , the  $D$ -dimensional Minkowski content of  $K$  by

$$\mathcal{M}^D(K) := \lim_{\varepsilon \rightarrow 0} \frac{\mathcal{L}^d(K(\varepsilon))}{\varepsilon^{d-D}},$$

for  $0 \leq D < d$  the  $D$ -dimensional S-content (surface area based) by

$$\mathcal{S}^D(K) := \lim_{\varepsilon \rightarrow 0} \frac{\mathcal{H}^{d-1}(\partial K(\varepsilon))}{(d-D)\varepsilon^{d-1-D}},$$

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**Average versions:**

$$\widetilde{\mathcal{M}}^D(K) := \lim_{\delta \rightarrow 0} \frac{1}{|\ln \delta|} \int_{\delta}^1 \frac{\mathcal{L}^d(K(\varepsilon))}{\varepsilon^{d-D}} \frac{1}{\varepsilon} d\varepsilon$$

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$\mathcal{L}^d(K(\varepsilon))$  and  $\mathcal{H}^{d-1}(\partial K(\varepsilon))$  are completed by the **Lipschitz-Killing curvatures**  $C_k(K(\varepsilon))$ ,  $k = 0, \dots, d-2$ , under some additional conditions of regularity of the boundary  $\partial K(\varepsilon)$ .

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For all  $k = 0, \dots, d$  and compact sets  $\mathcal{K}$  such that  $\mathcal{K}$  or  $\overline{\mathcal{K}^c}$  has positive reach, there exist **measure variants**  $C_k(\mathcal{K}, \cdot)$  with  $C_k(\mathcal{K}, \mathbb{R}^d) = C_k(\mathcal{K})$ . (By a result of Fu, 1985, in space dimensions less than 4 the condition is fulfilled for  $\mathcal{K} = K(\varepsilon)$  for any compact set  $K$  and Lebesgue almost all  $\varepsilon$ .)

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*Essential properties for our purposes:* They are

- ▶ invariant under Euclidean motions,
- ▶ scaling:  $C_k(\lambda\mathcal{K}, \lambda(\cdot)) = \lambda^k C_k(\mathcal{K}, \cdot)$ ,  $\lambda > 0$ ,
- ▶ locally determined:  $C_k(\mathcal{K}', (\cdot) \cap G) = C_k(\mathcal{K}, (\cdot) \cap G)$ , if  $\mathcal{K}' \cap G = \mathcal{K} \cap G$  for an open set  $G$ .



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(For certain classes of sets the total values  $C_k(\mathcal{K})$ ,  $k = 0, 1, \dots, d$ , form a complete system of related invariants under Euclidean motions.)

*Question:* For which classes of fractal sets  $K$  (or domains with fractal boundaries) we can extend the notions of (average) Minkowski- and  $s$ -contents to curvatures in the following sense?

$$C_k^{frac}(K) := \lim_{\varepsilon \rightarrow 0} \frac{C_k(K(\varepsilon))}{\varepsilon^{k-D}}, \text{ or}$$

$$\tilde{C}_k^{frac}(K) := \lim_{\delta \rightarrow 0} \frac{1}{|\ln \delta|} \int_{\delta}^1 \frac{C_k(K(\varepsilon))}{\varepsilon^{k-D}} \frac{1}{\varepsilon} d\varepsilon$$

up to some normalizing constants.

- ▶ **Relationships to spectral analysis, certain Zeta functions and fractal drums**

For the **Minkowski content** see the books of Lapidus and van Frankenhuijsen 2006, Lapidus, Radunovic and Zubrinic 2017, and the references therein.

- ▶ **Self-similar sets**

Falconer 1995 under (SSC), Gatzouras 2000 under (OSC): **existence and integral representation for  $\mathcal{M}^D$ , resp.  $\widetilde{\mathcal{M}}^D$ .**

Winter 2008:  $C_k^{frac}$  for the case of polyconvex parallel sets.

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Z. 2011:  $C_k^{frac}$  for a more general case.

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Lapidus, Pearse and Winter 2011-2013, Winter 2015: **relationships to fractal tilings.**

► **Extensions to some classes of self-conformal sets**

Kombrink 2011, Bohl 2012, Kesseböhmer and Kombrink 2012, Freiberg and Kombrink 2012.

► **Stochastically self-similar sets (random recursive constructions)**

Gatzouras 2000: existence and integral representation for  $\mathcal{M}^D$ , resp.  $\widetilde{\mathcal{M}}^D$ , almost surely and in the mean sense, equality.

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► **Homogeneous random fractals**

Hambly 1992 (special case), Troscheit 2017:  $\dim_H = \dim_M = D'$  a.s..

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Z. 2020: existence and integral representation for  $\mathbb{E}\mathcal{M}^D$ , resp.  $\mathbb{E}\widetilde{\mathcal{M}}^D$ .

Rataj, Winter and Z. 2023: extension to  $\mathbb{E}C_k^{frac}$ , resp.  $\mathbb{E}\widetilde{C}_k^{frac}$ .

► **V-variable random fractals**

Z. 2023: existence and integral representation for  $\mathbb{E}\mathcal{M}^D$ ,  $\mathbb{E}\mathcal{S}^D$ , resp.  $\mathbb{E}\widetilde{\mathcal{M}}^D$ ,  $\mathbb{E}\widetilde{\mathcal{S}}^D$ , and

$$\mathbb{E}\mathcal{M}^D = \mathbb{E}\mathcal{S}^D, \quad \mathbb{E}\widetilde{\mathcal{M}}^D = \mathbb{E}\widetilde{\mathcal{S}}^D, \text{ if } D < d.$$

► **Domains with piecewise self-similar fractal boundaries**

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Below we will present an extension of the mean value results to a large class of self-similar random code-tree fractals with more dependencies, which contains all former random cases mentioned above (joint work with J. Rataj and S. Winter).



## Random labeled code trees

$(f_1, \dots, f_N)$  random number of random contracting similarities (RIFS) with random contraction ratios  $(r_1, \dots, r_N)$  such that

1.  $0 < r_{min} \leq r_i \leq r_{max} < 1$  w.p.1, for deterministic values  $r_{min}, r_{max}$ .
2.  $1 < \mathbb{E}N < \infty$  (supercritical case)
3.  $\bigcup_{i=1}^N f_i(O) \subset O$  and  $f_i(O) \cap f_j(O) = \emptyset, i \neq j$ , w.p.1, for some fixed open set  $O$  (Uniform Open Set Condition UOSC)

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## Recursive random code tree construction:

$$\Sigma_0 := \{i : 1 \leq i \leq N\}$$

$$\Sigma_n := \{\sigma = \sigma_1 \dots \sigma_n : \sigma_1 \dots \sigma_{n-1} \in \Sigma_{n-1}, 1 \leq \sigma_n \leq N_{\sigma_1 \dots \sigma_{n-1}}\}$$

(codes at levels  $n$ ) for random numbers  $N_{\sigma_1 \dots \sigma_{n-1}}, n \geq 1$

$$\Sigma_* := \bigcup_{n=1}^{\infty} \Sigma_n \quad (\text{random code tree})$$

RIFS's as above are chosen as labels at the nodes  $\sigma \in \Sigma_*$ :

$$\mathcal{F}_\sigma := (f_{\sigma_1}, \dots, f_{\sigma_{N_\sigma}}).$$

Random labeled code tree:

$$\mathcal{T} := \{(\sigma, \mathcal{F}_\sigma) : \sigma \in \Sigma_*\}.$$

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For  $\sigma \in \Sigma_*$  the **subtree** rooted at the node  $\sigma$  is given by

$$\mathcal{T}_\sigma := \{(\tau, \mathcal{F}_{\sigma\tau}) : \sigma\tau \in \Sigma_*\}$$

**Assumption 1 (back path condition):** For each  $i$ , under the condition that  $i \leq N$ , the labeled subtree  $\mathcal{T}_i$  is independent of the preceding mapping  $f_i$  and has the same distribution as  $\mathcal{T}$ .

*Essential consequence:* For all steps  $n \in N$ , under the condition that  $\sigma = \sigma_1 \dots \sigma_n \in \Sigma_n$ , the random labeled subtree  $\mathcal{T}_\sigma$  rooted at  $\sigma$  is independent of the corresponding random mappings along the path  $(f_{\sigma_1}, f_{\sigma_1\sigma_2}, \dots, f_{\sigma_1 \dots \sigma_n})$  and has the same distribution as the primary tree  $\mathcal{T}$ .

*New:* Allows much more dependencies between the different paths than in the former models.

**Iteration of the random function systems along the paths of the random tree:** For  $n \geq 1$  and  $\sigma = \sigma_1 \dots \sigma_n \in \Sigma_n$  denote

$$f^\sigma = f_{\sigma_1} \circ f_{\sigma_1 \sigma_2} \circ \dots \circ f_{\sigma_1 \dots \sigma_n}$$

**Associated self-similar random code-tree fractal:**

$$F := \bigcap_{n=1}^{\infty} \bigcup_{\sigma \in \Sigma_n} f^\sigma(\overline{O})$$

is a.s. determined

**Assumption 2 (USOSC):** Uniform Strong Open Set Condition for some bounded open set  $O$  in  $\mathbb{R}^d$ , i.e., UOSC as above for  $O$  and

$$\mathbb{P}(F \cap O \neq \emptyset) > 0.$$

# Mean Minkowski and S-contents

Let  $(r_1, r_2, \dots, r_N)$  be a random vector with random  $N$  having the same distribution as the contraction ratios of the above RIFSs. Let  $D$  be determined by

$$\mathbb{E} \sum_{i=1}^N (r_i)^D = 1 \quad (\text{mean dimension equation}).$$

Associated probability distribution of the logarithmic contraction ratios:

$$\mu := \mathbb{E} \left( \sum_{i=1}^N \mathbf{1}_{(\cdot)}(|\ln(r_i)|) (r_i)^D \right)$$

Corresponding mean value:

$$\eta := \mathbb{E} \left( \sum_{i=1}^N |\ln(r_i)| (r_i)^D \right)$$

Denote  $\varphi_d(\varepsilon) := \mathbb{E} \mathcal{L}^d(F(\varepsilon))$ . Under the above conditions we obtain the following:

### Theorem 1 [(average) mean Minkowski content]

(i) If the measure  $\mu$  is non-lattice, then

$$\lim_{\varepsilon \rightarrow 0} \frac{\mathbb{E} \mathcal{L}^d(F(\varepsilon))}{\varepsilon^{d-D}} = \frac{1}{\eta} \int_0^1 \varepsilon^{D-d} R_d(\varepsilon) \frac{1}{\varepsilon} d\varepsilon =: M_F^D,$$

where the function  $R_d(\varepsilon)$  is given by

$$R_d(\varepsilon) = \mathbf{1}_{(0,1]}(\varepsilon) \varphi_d(\varepsilon) - \mathbb{E} \sum_{i=1}^N \mathbf{1}_{(0,r_i]}(\varepsilon) r_i^d \varphi_d(r_i^{-1} \varepsilon).$$

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(ii) For general  $\mu$  we get for the average limit

$$\lim_{\delta \rightarrow 0} \frac{1}{|\ln \delta|} \int_{\delta}^1 \frac{\mathbb{E} \mathcal{L}^d(F(\varepsilon))}{(d-D)\varepsilon^{d-D}} \frac{1}{\varepsilon} d\varepsilon = M_F^D.$$



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(iii) The limit value  $M_F^D$  does not vanish.

## Theorem 2 [(average) mean $S$ -content]

Items (i) and (ii) are analogous for the surface area based version  $S_F^D$ .  
Moreover,

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**Methods of proof:** Renewal theorem from classical probability theory, using Markov stops, conditional expectations and Assumption 1 concerning the back path property.

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**Remark:** Barnsley, Hutchinson and Stenflo 2012: determined the almost sure Hausdorff dimension  $\dim_{\text{H}} F$  via flow matrices for the  $V$ -variable case, Troscheit 2017:  $\dim_{\text{H}} F = \dim_{\text{M}} F = D_{\text{H}}$ , given by

$$\mathbb{E} \ln \left( \sum_{\sigma \in \Sigma_{\text{neck}}} (r^\sigma)^{D_{\text{H}}} \right) = 0.$$

$D_{\text{H}}$  is, in general, less than the above Minkowski dimension  $D$  in the mean sense given by  $\mathbb{E} \sum_{i=1}^N (r_i)^D = 1$ .

# Extensions to fractal Lipschitz-Killing curvatures

Choose an arbitrary constant  $R > \sqrt{2}|O|$ .

**Assumption 3 (regularity of  $F$ ):** w.p.1 for Lebesgue almost all  $r < R$  the sets  $\overline{F(r)^c}$  have positive reach.

(Remark: It is known that, in general for a.a.  $r \geq R$ , and in space dimensions  $d \leq 3$  for a.a.  $r > 0$  this is always fulfilled.)

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Notation (*Markov stop in the code space*):

$$\Sigma(r) := \{\sigma \in \Sigma_* : Rr^\sigma \leq r < Rr^{\sigma||\sigma|-1}\}, 0 < r < R,$$

where  $\sigma||\sigma| - 1$  means  $\sigma_1 \dots \sigma_{n-1}$  if  $\sigma = \sigma_1 \dots \sigma_n$ .

### Theorem 3 [(average) mean Lipschitz-Killing curvatures]

Let  $k \in \{0, 1, \dots, d\}$  and let  $F$  be a random self-similar code tree fractal in  $\mathbb{R}^d$  as above with independence along back paths (Assumption 1) satisfying the Uniform Strong Open Set Condition with basic set  $O \subset \mathbb{R}^d$  (Assumption 2). For  $k \leq d - 2$  we additionally suppose the following:

- (i) if  $d \geq 4$ , then  $F$  is regular ( Assumption 3),
- (ii) there exists a constant  $c_k > 0$  such that w.p.1,

$$C_k^{\text{var}}(F(r), f^\sigma(O)(r) \cap f^\tau(O)(r)) \leq c_k r^k$$

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for Lebesgue a.a.  $r < R$  and all  $\sigma, \tau \in \Sigma(r)$  with  $\sigma \neq \tau$ .

Set for a.a.  $r > 0$ ,

$$R_k(r) := \mathbb{E}C_k(F(r)) - \mathbb{E} \sum_{i=1}^N \mathbf{1}_{(0, Rr_i]}(r) C_k(F_i(r)).$$

Then we get the following:



(I) If the measure  $\mu$  is non-lattice, then

$$\lim_{\varepsilon \rightarrow 0} \varepsilon^{D-k} \mathbb{E} C_k(F(\varepsilon)) = \frac{1}{\eta} \int_0^R r^{D-k-1} R_k(r) dr.$$

(II) If the measure  $\mu$  is lattice with constant  $c$ , then for almost all  $s \in [0, c)$

$$\lim_{n \rightarrow \infty} e^{(k-D)(s+nc)} \mathbb{E} C_k(F(e^{-(s+nc)})) = \frac{1}{\eta} \sum_{m=0}^{\infty} e^{(k-D)(s+mc)} R_k(e^{-(s+mc)}).$$

(III) In general,

$$\lim_{\delta \rightarrow 0} \frac{1}{|\ln \delta|} \int_{\delta}^1 \varepsilon^{D-k} \mathbb{E} C_k(F(\varepsilon)) \varepsilon^{-1} d\varepsilon = \frac{1}{\eta} \int_0^R r^{D-k-1} R_k(r) dr.$$

## Special cases and an example

*Assumption 1* (back path condition) is satisfied for the random recursive case and the  $V$ -variable random fractals.

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*Assumption 2* (USOSC) is always required.

*Assumption 3* (geometric regularity condition) is always fulfilled in  $\mathbb{R}^d$  with  $d \leq 3$ .

Boundedness condition (ii) can be checked for classical examples of random Sierpinski-type fractals (gaskets or carpets)

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THANK YOU !