

Markov-self-similar sets

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1. Introduction

A theory of non-random self-similar sets has been developed by Moran [11] and Hutchinson [9]. Lately Mauldin-Williams [10], Falconer [5] and Graf [7] investigated random self-similar sets. In this paper we introduce a new concept of *Markov-self-similarity* and investigate deterministic and random Markov-self-similar sets. Takahashi [12] introduced a concept of multi-similarity which is essentially the same concept as Markov-self-similarity. Markov-self-similarity is a natural extension of self-similarity and Markov-self-similar sets appear as the limit sets of cellular automata [12, 15]. Cellular automata are used to model problems in crystal growth and diffusion and other problems of self-organization. Therefore the patterns appeared in these fields are expected to be Markov-self-similar. On the other hand some Markov-self-similar sets can be constructed as recurrent sets defined by Dekking [3]. (See also Bedford [1, 2].)

A Markov-self-similar set is constructed as follows. First we prepare an N -tuple (S_{01}, \dots, S_{0N}) of contraction similarities of \mathbf{R}^d which are initial contractions and used only in the first step. Let F be a non-empty compact subset of \mathbf{R}^d , and set

$$A_1 = \bigcup_{k=1}^N S_{0k}(F).$$

Next we fix a family of N N -tuples $\{(S_{k1}, \dots, S_{kN})\}_{k=1}^N$ of contraction similarities of \mathbf{R}^d which are fundamental contractions and used in the following process repeatedly. We assume that above N N -tuples satisfy the irreducibility condition and the open set condition. (See Section 2.) Set

$$A_2 = \bigcup_{k=1}^N S_{0k}(\bigcup_{i=1}^N S_{ki}(F)).$$

Note that the contractions S_{ki} are selected depending on the index k of S_{0k} . Set

$$A_3 = \bigcup_{k=1}^N S_{0k}(\bigcup_{i=1}^N S_{ki}(\bigcup_{j=1}^N S_{ij}(F))).$$

We continue this process. Let $K = \lim_{n \rightarrow \infty} A_n$ where the limit is taken with respect to the Hausdorff metric. The set K has a Markovian shape structure which is not possessed by a self-similar set constructed in Hutchinson [9].

A random Markov-self-similar set is a probabilistic counterpart of a non-random Markov-self-similar set. The plan of this paper is as follows.

In Section 2 we investigate a Markov-self-similar N -tuple of compact sets which is an extension of a Hutchinson's self-similar set. The fundamental result is as follows: Let $\mathbf{S} = (\underline{S}_1, \dots, \underline{S}_N)$ be an N -tuple of $\underline{S}_k = (S_{k1}, \dots, S_{kN})$, $k = 1, \dots, N$ where S_{ki} , $i = 1, \dots, N$ are contraction similarities of \mathbf{R}^d which satisfy the open set condition. For a non-negative number β , we define an $N \times N$ non-negative matrix $R(\beta) = [R(\beta)_{kj}]$ by

$$R(\beta)_{kj} = r(S_{kj})^\beta \quad k, j = 1, \dots, N$$

where $r(S_{kj})$ is the contraction ratio of S_{kj} . Let $\lambda(\beta)$ be the maximal eigen value of $R(\beta)$. Let F be a non-empty compact set. Set

$$K_k = \lim_{m \rightarrow \infty} \bigcup_{i_1, \dots, i_m = 1}^N S_{ki_1} \circ S_{i_1 i_2} \circ \dots \circ S_{i_{m-1} i_m}(F)$$

for $k = 1, \dots, N$ where the limit is taken with respect to the Hausdorff metric. Then

$$\dim_H(K_k) = \alpha$$

and

$$0 < \mathcal{H}^\alpha(K_k) < \infty$$

for all $k = 1, \dots, N$ where α is such that $\lambda(\alpha) = 1$. Furthermore there exists $c > 0$ such that

$$\mathcal{H}^\alpha(K_k) = cx_k \quad \text{for } k = 1, \dots, N$$

where (x_1, \dots, x_N) is a positive eigenvector of $R(\alpha)$ corresponding to the maximal eigen value $\lambda(\alpha) = 1$. The N -tuple (K_1, K_2, \dots, K_N) of compact sets defined above satisfies the conditions:

$$K_k = \bigcup_{i=1}^N S_{ki}(K_i) \quad \text{for } k = 1, \dots, N,$$

K_k is an α -set and $\mathcal{H}^\alpha(S_{ki}(K_i) \cap S_{kj}(K_j)) = 0$ for all $k = 1, \dots, N$ and $i \neq j$. Such an N -tuple of compact sets is called Markov-self-similar.

In Section 3 we introduce a concept of random Markov-self-similarity and show that the results that correspond to those for the concept of statistical self-similarity obtained in Graf [7] hold. Let (μ_1, \dots, μ_N) be an N -tuple of Borel probability measures on $\text{Con}(X)^N$ where $\text{Con}(X)$ denotes the set of all contractions of a compact set X . Then there exists a unique N -tuple of probability measures (P_1, \dots, P_N) on $\mathcal{K}(X)$, the set of all non-empty compact sets in X , such that for every Borel set $B \subset \mathcal{K}(X)$,

$$(i) \quad P_k(B) = [\mu_k \times \prod_{i=1}^N P_i] (\{((S_1, \dots, S_N), (K_1, \dots, K_N)) \in \text{Con}(X)^N \times \mathcal{K}(X)^N \mid \bigcup_{i=1}^N S_i(K_i) \in B\})$$

for all $k = 1, \dots, N$. An N -tuple (P_1, \dots, P_N) of probability measures on $\mathcal{X}(X)$ which satisfies (i) is called (μ_1, \dots, μ_N) -Markov-self-similar. Furthermore the following holds: Let $R(\beta) = [R(\beta)_{ij}]$ be an $N \times N$ matrix defined by

$$R(\beta)_{ij} = \int r(S_j)^\beta d\mu_i(S_1, \dots, S_N)$$

where $\beta \geq 0$, and let $\lambda(\beta)$ be the maximal eigen value of non-negative matrix $R(\beta)$. Under some conditions, $\dim_H(K) = \alpha$ for P_k -a.e. $K \in \mathcal{X}(X)$ for all $k = 1, \dots, N$ where α is a positive number such that $\lambda(\alpha) = 1$.

In Section 4 we investigate the Hausdorff-measures of random Markov-self-similar sets. The results are as follows: Suppose that there exists a $\delta > 0$ such that if $R(0)_{ki} > 0$, then $r(S_i) \geq \delta$ for μ_k -a.e. (S_1, \dots, S_N) where $k, i = 1, \dots, N$. Let (x_1, \dots, x_N) be a positive eigenvector of $R(\alpha)$ corresponding to the maximal eigen value 1. Then the following statements are equivalent:

- a) $\sum_{i=1}^N r(S_i)^\alpha x_i = x_k$ for μ_k -a.e. (S_1, \dots, S_N) and all $k = 1, \dots, N$.
- b) $\mathcal{H}^\alpha(K) > 0$ for P_k -a.e. $K \in \mathcal{X}(X)$ and all $k = 1, \dots, N$.
- c) $P_j(\{K \in \mathcal{X}(X) | \mathcal{H}^\alpha(K) > 0\}) > 0$ for some $j \in \{1, \dots, N\}$.

This is an extension of the result given by Graf [7]. Furthermore if $P_j(\{K \in \mathcal{X}(X) | \mathcal{H}^\alpha(K) > 0\}) > 0$ for some $j \in \{1, \dots, N\}$, then there exists $c > 0$ such that

$$\mathcal{H}^\alpha(K) = cx_k \quad \text{for } P_k\text{-a.e. } K \in \mathcal{X}(X) \text{ and all } k = 1, \dots, N.$$

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2. Markov-self-similar sets

Let $Y = (Y, d)$ be a complete metric space. A mapping $S: Y \rightarrow Y$ is called a contraction if $d(S(x), S(y)) \leq rd(x, y)$ for all $x, y \in Y$ where $0 < r < 1$, and $r(S) = \inf\{r \geq 0 | d(S(x), S(y)) \leq rd(x, y) \text{ for all } x, y \in Y\}$ is called the contraction ratio of S . By $\text{Con}(Y)$ we denote the set of all contractions of Y . We assume the null contraction ϕ is an element of $\text{Con}(Y)$ where ϕ is such that $\phi(Y) = \emptyset$. Fix a positive integer $N \geq 2$. Let $\text{Con}(Y)^N = \{(S_1, S_2, \dots, S_N) | S_i \in \text{Con}(Y) \text{ for } i = 1, \dots, N, (S_1, S_2, \dots, S_N) \neq (\phi, \phi, \dots, \phi)\}$. Let $\mathcal{X}(Y)$ be the space of all non-empty compact subsets of Y . The topology of $\mathcal{X}(Y)$ is defined by the Hausdorff metric $\rho(A, B) = \sup\{d(a, B), d(A, b) | a \in A, b \in B\}$, $A, B \in \mathcal{X}(Y)$.

Hutchinson [9] proved that for every finite set of contractions S_1, S_2, \dots, S_N of a complete metric space there exists a unique invariant non-empty compact set K , i.e., $K = \bigcup_{i=1}^N S_i(K)$. Furthermore he showed that if S_i are similarities with contraction ratio r_i of \mathbf{R}^d which satisfy the open set

condition, the Hausdorff dimension of K equals to α where α is a number such that $\sum_{i=1}^N r_i^\alpha = 1$. We extend the result as follows.

THEOREM 2.1. *Let $\mathbf{S} = (\underline{S}_1, \dots, \underline{S}_N)$ be an N -tuple of $\underline{S}_k = (S_{k1}, \dots, S_{kN}) \in \text{Con}(Y)^N$ $k = 1, \dots, N$. Then there exists a unique N -tuple (K_1, \dots, K_N) of non-empty compact sets such that*

$$(1) \quad K_k = \bigcup_{i=1}^N S_{ki}(K_i) \quad \text{for } k = 1, \dots, N.$$

Furthermore for any non-empty compact set F

$$(2) \quad \lim_{m \rightarrow \infty} \bigcup_{i_1, \dots, i_m}^N S_{ki_1} \circ S_{i_1 i_2} \circ \dots \circ S_{i_{m-1} i_m}(F) = K_k \quad \text{for } k = 1, \dots, N$$

where the limit is taken with respect to the Hausdorff metric.

The statement (1) of Theorem 2.1 is a special case of Proposition 3.6 in Section 3, and the statement (2) is proved in the same manner as in Hutchinson [9].

REMARKS (i) Associated with $\mathbf{S} = \{\underline{S}_1, \dots, \underline{S}_N\}$, an operator $T_{\mathbf{S}}: \mathcal{K}(Y)^N \rightarrow \mathcal{K}(Y)^N$ is defined by

$$T_{\mathbf{S}}(F_1, \dots, F_N) = (\bigcup_{i=1}^N S_{1i}(F_i), \dots, \bigcup_{i=1}^N S_{Ni}(F_i))$$

for $(F_1, \dots, F_N) \in \mathcal{K}(Y)^N$. Then the equalities (1) imply $T_{\mathbf{S}}(K_1, \dots, K_N) = (K_1, \dots, K_N)$, i.e. (K_1, \dots, K_N) is $T_{\mathbf{S}}$ -invariant.

(ii) Let F be a non-empty compact set in Y and $(\underline{S}_1, \dots, \underline{S}_N)$ and \underline{S}_0 be such that $\underline{S}_k = (S_{k1}, \dots, S_{kN}) \in \text{Con}(Y)^N$, $k = 0, 1, \dots, N$. Let

$$(2') \quad K = \lim_{m \rightarrow \infty} \bigcup_{i_1, \dots, i_m=1}^N S_{0i_1} \circ S_{i_1 i_2} \circ \dots \circ S_{i_{m-1} i_m}(F).$$

Then the set K can be expressed by

$$K = \bigcup_{k=1}^N S_{0k}(K_k)$$

where (K_1, \dots, K_N) is the N -tuple of compact sets that satisfy the equalities (1) with respect to $(\underline{S}_1, \dots, \underline{S}_N)$.

Next we give the lower and upper estimates of the Hausdorff measures of compact sets K_k . We introduce some notation.

Let $E \subset Y$, $\delta > 0$ and $\alpha \geq 0$ be arbitrary. Define

$$\mathcal{H}_\delta^\alpha(E) = \inf \{ \sum_{i=1}^\infty |E_i|^\alpha \mid E \subset \bigcup_{i=1}^\infty E_i, |E_i| \leq \delta \},$$

and

$$\mathcal{H}^\alpha(E) = \sup_{\delta > 0} \mathcal{H}_\delta^\alpha(E)$$

where $|E|$ is the diameter of E . Then \mathcal{H}^α is an outer measure on Y such that

all Borel sets are \mathcal{H}^α -measurable. \mathcal{H}^α is called the α -dimensional measure. The Hausdorff dimension of E is defined by

$$\begin{aligned} \dim_H(E) &= \sup \{ \alpha \geq 0 \mid \mathcal{H}^\alpha(E) > 0 \} \\ &= \inf \{ \alpha \geq 0 \mid \mathcal{H}^\alpha(E) < \infty \}. \end{aligned}$$

An \mathcal{H}^α -measurable set E is called α -set if $0 < \mathcal{H}^\alpha(E) < \infty$.

Let $(\underline{S}_1, \dots, \underline{S}_N)$ be an N -tuple of $\underline{S}_k = (S_{k1}, \dots, S_{kN}) \in \text{Con}(Y)^N$, $k = 1, \dots, N$. For a non-negative number β , we define an $N \times N$ non-negative matrix $R(\beta) = [R(\beta)_{kj}]$ by

$$R(\beta)_{kj} = r(S_{kj})^\beta \quad k, j = 1, \dots, N$$

where $r(S_{kj})$ is the contraction ratio of S_{kj} and $r(\phi) = 0$ where ϕ is the null contraction. Let $\lambda(\beta)$ be the maximal eigenvalue of $R(\beta)$. Assume that $\lambda(0) > 1$. Then there exists a unique $\alpha > 0$ such that $\lambda(\alpha) = 1$.

PROPOSITION 2.2. *Under the assumption of Theorem 2.1, let (K_1, \dots, K_N) be the unique N -tuple of non-empty compact sets which satisfies the equalities (1) of Theorem 2.1, then it holds that*

$$\dim_H(K_k) \leq \alpha \quad \text{for } k = 1, \dots, N$$

where α is such that $\lambda(\alpha) = 1$.

Proposition 2.2 is a special case of Proposition 3.9 in Section 3.

REMARK. If $K = \bigcup_{k=1}^N S_{0k}(K_k)$ for an N -tuple (S_{01}, \dots, S_{0N}) of contractions, then $\dim_H(K) \leq \alpha$.

Now we give the definition of Markov-self-similarity. A mapping $S: Y \rightarrow Y$ is called a similarity if there exists an $r > 0$ such that $d(Sx, Sy) = rd(x, y)$ for all $x, y \in Y$. We define $\text{Sim}(Y)^N$ in the same manner as $\text{Con}(Y)^N$ except that all contractions are contraction similarities.

DEFINITION 2.3. Let $\mathbf{S} = (\underline{S}_1, \dots, \underline{S}_N)$ be an N -tuple of $\underline{S}_k = (S_{k1}, \dots, S_{kN}) \in \text{Sim}(Y)^N$, $k = 1, \dots, N$. An N -tuple (K_1, \dots, K_N) of non-empty compact sets is called *Markov-self-similar* with respect to \mathbf{S} if

$$K_k = \bigcup_{i=1}^N S_{ki}(K_i) \quad \text{for } k = 1, \dots, N$$

and if for some $\alpha \geq 0$, K_k is an α -set and $\mathcal{H}^\alpha(S_{ki}(K_i) \cap S_{kj}(K_j)) = 0$ for all $k = 1, \dots, N$ and $i \neq j$. A non-empty compact set K is called *Markov-self-similar* with respect to \mathbf{S} if there exist a Markov-self-similar N -tuple (K_1, \dots, K_N) with respect to \mathbf{S} and an N -tuple (S_1, \dots, S_N) of contractions such that $K = \bigcup_{k=1}^N S_k(K_k)$ and $\mathcal{H}^\alpha(S_i(K_i) \cap S_j(K_j)) = 0$.

An $N \times N$ matrix R is called irreducible if for any $i, j \in \{1, \dots, N\}$ there exists a positive integer $m = m(i, j)$ such that $(R^m)_{ij} > 0$. For an irreducible non-negative matrix R , the following Frobenius' Theorem holds:

THEOREM 2.4. (Frobenius). *An irreducible non-negative matrix R has a unique maximal positive eigen value λ for which there correspond positive row and column eigenvectors. Furthermore the inequalities*

$$\lambda z \geq Rz \quad \text{for a vector } z \geq 0 \text{ and } z \neq 0$$

or

$$\lambda z \leq Rz \quad \text{for a vector } z \geq 0 \text{ and } z \neq 0$$

imply that $\lambda z = Rz$ and $z > 0$; and the equality

$$Ry = \eta y \quad \text{for a vector } y \geq 0 \text{ and } y \neq 0$$

implies that $\eta = \lambda$. Moreover it holds that

$$\lambda = \max_{z \geq 0} \min_{0 \leq i \leq N} (Az)_i / z_i = \min_{z \geq 0} \max_{0 \leq i \leq N} (Az)_i / z_i$$

where $z = (z_1, \dots, z_N)$.

See Gantmacher [6, Ch. 13, §2].

The following theorem states conditions under which an N -tuple of compact sets satisfying (1) in Theorem 2.1 is Markov-self-similar. See Takahashi [12].

THEOREM 2.5. *Let $\mathbf{S} = (\underline{S}_1, \dots, \underline{S}_N)$ be an N -tuple of $\underline{S}_k = (S_{k1}, \dots, S_{kN}) \in \text{Sim}(\mathbf{R}^d)^N$, $k = 1, \dots, N$ which satisfies the following conditions:*

a) *There exists a non-empty open set V for which*

$$S_{ki}(V) \subset V \text{ and } S_{ki}(V) \cap S_{kj}(V) = \emptyset \text{ if } i \neq j \text{ for all } k = 1, \dots, N.$$

b) *The matrix $R(0)$ is irreducible and the maximal eigen value $\lambda(0) > 1$.*

Let (K_1, \dots, K_N) be the unique N -tuple of compact sets that satisfies the condition (1) of Theorem 2.1. Then (K_1, \dots, K_N) is Markov-self-similar with respect to \mathbf{S} for α such that $\lambda(\alpha) = 1$. Furthermore there exists $c > 0$ such that

$$\mathcal{H}^\alpha(K_k) = cx_k \quad k = 1, \dots, N$$

where (x_1, \dots, x_N) is a positive eigenvector of $R(\alpha)$ corresponding to the maximal eigen value 1.

REMARKS (i) If $\underline{S}_k = \underline{S} = (S_1, \dots, S_N)$ for all $k = 1, \dots, N$, the Hausdorff dimension α is obtained as an α for which $\sum_{i=1}^N r(S_i)^\alpha = 1$, because of Theorem 2.4 (Frobenius).

(ii) For $\mathbf{S} = (\underline{S}_1, \dots, \underline{S}_N)$ such that $\underline{S}_k = (S_{k1}, \dots, S_{kN})$ with $r(S_{ki}) = r_k$ for

$i = 1, \dots, N$ and $k = 1, \dots, N$, the Hausdorff dimension α is obtained as an α for which

$$\sum_{k=1}^N r_k^\alpha = 1,$$

because $(r_1^\alpha, \dots, r_N^\alpha)$ is a positive eigenvector corresponding to the eigen value 1.

(iii) a) Even if $R(0)$ is reducible, there exists at least one $k \in \{1, \dots, N\}$ such that K_k is an α -set.

b) There exists $S = \{\underline{S}_1, \dots, \underline{S}_N\}$ for which $R(0)$ is reducible and $\mathcal{H}^\alpha(K_i) = 0$ and $\mathcal{H}^\alpha(K_j) = \infty$ for some $i, j \in \{1, \dots, N\}$.

For the proof of Theorem 2.5 we need a lemma (cf. Falconer [4]).

LEMMA 2.6. *Under the assumptions of Theorem 2.5 there exists an N -tuple (μ_1, \dots, μ_N) of Borel probability measures such that, for any measurable set F and $k = 1, \dots, N$,*

(ii)
$$\mu_k(F) = \sum_{i=1}^N r(S_{ki})^\alpha \mu_i(S_{ki}^{-1}(F))$$

and

$$\mu_k(\mathbf{R}^d) = x_k$$

where (x_1, \dots, x_N) is a positive eigenvector of $R(\alpha)$ corresponding to the maximal eigen value 1. Furthermore μ_k has the support contained in K_k for $k = 1, \dots, N$.

PROOF. Choose $y \in K_1$ and write

$$y_{i_1 i_2 \dots i_m} = S_{i_1 i_2} \circ S_{i_2 i_3} \circ \dots \circ S_{i_{m-1} i_m}(y)$$

for $i_1, \dots, i_m = 1, \dots, N$. Let us write $r(S_{ij})$ by r_{ij} . For $k = 1, \dots, N$ and $m = 1, 2, \dots$, define positive linear functionals $\varphi_m^{(k)}$ on the space $C(K_k)$ of continuous functions on K_k by

$$\varphi_m^{(k)}(f) = \sum_{i_1 \dots i_m=1}^N (r_{k i_1} r_{i_1 i_2} \dots r_{i_{m-1} i_m})^\alpha x_{i_m} f(y_{k i_1 \dots i_m}).$$

Note that $y_{k i_1 \dots i_m} \in K_k$ or $y_{k i_1 \dots i_m} = \emptyset$ and that $r(\emptyset) = 0$. Usual arguments show that $\lim_{m \rightarrow \infty} \varphi_m^{(k)}$ defines a positive linear functional $\varphi^{(k)}$ on $C(K_k)$. By the Riesz representation theorem, there exists Borel measure μ_k such that

$$\int f d\mu_k = \varphi^{(k)} f = \lim_{m \rightarrow \infty} \varphi_m^{(k)} f$$

for $f \in C(K_k)$. Putting $f \equiv 1$, it follows that $\mu_k(\mathbf{R}^d) = x_k$ because

$$\sum_{j=1}^N r_j^\alpha x_j = x_i.$$

Since $f \in C(K_k)$, μ_k has the support contained in K_k . For $f \in C(K_k)$,

$$\begin{aligned} \varphi_m^{(k)}(f) &= \sum_{i_1=1}^N r_{ki_1}^\alpha \left(\sum_{i_2, \dots, i_m=1}^N (r_{i_1 i_2} \cdots r_{i_{m-1} i_m})^\alpha x_{i_m} f(S_{ki_1}(y_{i_1 \dots i_m})) \right) \\ &= \sum_{i=1}^N r_{ki}^\alpha \varphi_{m-1}^{(i)}(f \circ S_{ki}). \end{aligned}$$

Letting $m \rightarrow \infty$ we get

$$\int f d\mu_k = \sum_{i=1}^N r_{ki}^\alpha \int f \circ S_{ki} d\mu_i,$$

so (ii) follows. This completes the proof.

PROOF OF THEOREM 2.5. The proof is similar to that of Theorem 8.6 of Falconer [4]. The upper bound: Iterating (1) we get

$$K_k = \bigcup_{i_1, \dots, i_m} S_{ki_1} \circ S_{i_1 i_2} \circ \cdots \circ S_{i_{m-1} i_m}(K_{i_m}).$$

Using $\sum_{i=1}^N r_{ki}^\alpha x_i = x_k$, we get

$$\begin{aligned} &\sum_{i_1, \dots, i_m} |S_{ki_1} \circ S_{i_1 i_2} \circ \cdots \circ S_{i_{m-1} i_m}(K_{i_m})|^\alpha \\ &= \sum_{i_1, \dots, i_m} (r_{ki_1} r_{i_1 i_2} \cdots r_{i_{m-1} i_m})^\alpha x_{i_m} |K_{i_m}|^\alpha x_{i_m}^{-1} \\ &\leq \frac{x_k}{\min_i x_i} \max_i |K_i|^\alpha < \infty. \end{aligned}$$

As $|S_{ki_1} \circ S_{i_1 i_2} \circ \cdots \circ S_{i_{m-1} i_m}(K_{i_m})|^\alpha \rightarrow 0$ as $m \rightarrow \infty$, we have $\mathcal{H}^\alpha(K_k) < \infty$.

The lower bound: Using similar arguments as in the proof of Theorem 8.6 of Falconer [4] and Lemma 2.6 instead of Lemma 8.4 of Falconer, we can show that

$$\mathcal{H}^\alpha(K_k) \geq x_k (q \max_i x_i)^{-1} > 0$$

where q is a positive finite constant.

Proof of the facts that $\mathcal{H}^\alpha(K_k) = cx_k$ and that $\mathcal{H}^\alpha(S_{ki}(K_k) \cap S_{kj}(K_k)) = 0$ for $i \neq j$: Using (1) and the fact that S_{ki} are similarities, we get

$$\mathcal{H}^\alpha(K_k) \leq \sum_{i=1}^N \mathcal{H}^\alpha(S_{ki}(K_i)) = \sum_{i=1}^N r(S_{ki})^\alpha \mathcal{H}^\alpha(K_i)$$

for $k = 1, \dots, N$. By Theorem 2.4 (Frobenius) it follows that

$$(a) \quad \mathcal{H}^\alpha(K_k) = \sum_{i=1}^N \mathcal{H}^\alpha(S_{ki}(K_i)) = \sum_{i=1}^N r(S_{ki})^\alpha \mathcal{H}^\alpha(K_i)$$

and that there exists $c > 0$ such that

$$\mathcal{H}^\alpha(K_k) = cx_k \quad \text{for } k = 1, \dots, N$$

where (x_1, \dots, x_N) is a positive eigenvector of $R(\alpha)$ corresponding to the maximal eigen value 1. As $0 < \mathcal{H}^\alpha(K_k) < \infty$, (1) and (a) mean that $\mathcal{H}^\alpha(S_{ki}(K_i) \cap S_{kj}(K_j)) = 0$ for $i \neq j$. This completes the proof.

EXAMPLE 1. Let $Y = [0, 1]; N = 2; S_{11}(y) = y/3, S_{12}(y) = (y + 2)/3; S_{21}(y) = y/9, S_{22}(y) = (y + 8)/9$ for $0 \leq y \leq 1$. By Remark (ii) of Theorem 2.5, $\alpha \geq 0$ such that $\lambda(\alpha) = 1$ is obtained as an α for which $(1/3)^\alpha + (1/9)^\alpha = 1$, and it follows that $\alpha = (\log(\sqrt{5} + 1) - \log 2)/(\log 3)$. By Theorem 2.5 we have

$$\mathcal{H}^\alpha(K_1): \mathcal{H}^\alpha(K_2) = (\sqrt{5} - 1): (3 - \sqrt{5}).$$

EXAMPLE 2. Let $Y = [0, 1]; N = 3; S_{11}(y) = S_{21}(y) = y/9, S_{12}(y) = S_{22}(y) = (y + 4)/9, S_{13}(y) = S_{23}(y) = (y + 8)/9, S_{31}(y) = y/4, S_{32}(y) = (y + 3)/4, S_{33} = \phi$ for $0 \leq y \leq 1$. The matrix $R(0) = [r(S_{ki})^0]_{ki}$ is irreducible, $\lambda(1/2) = 1$ and the vector $(1, 1, 1)$ is an eigenvector corresponding to the maximal eigen value 1. Therefore the Hausdorff dimension α equals to $1/2$ and $\mathcal{H}^{1/2}(K_1): \mathcal{H}^{1/2}(K_2): \mathcal{H}^{1/2}(K_3) = 1:1:1$.

3. Random Markov-self-similar sets

Random self-similar sets were investigated by Mauldin-Williams [10], Falconer [5] and Graf [7]. In this section we consider *random* Markov-self-similar sets which are probabilistic counterparts of Markov-self-similar sets defined in Section 2. Our results and techniques were inspired by the work of Graf [7], and all of the results are proved in Appendix.

We introduce the scheme used by Graf [7] with necessary modifications. Let (X, d) be a complete separable metric space whose diameter $|X|$ is *finite*. Fix a positive integer $N \geq 2$. The definition of $\text{Con}(X)^N$ is given in Section 2. Let

$$D = D(N) = \bigcup_{m=0}^\infty C_m$$

where $C_m = C_m(N) = \{1, 2, \dots, N\}^m$ and $C_0 = \{\emptyset\}$. If $\sigma = (\sigma_1, \dots, \sigma_m) \in D$, then $|\sigma| = m$ is the length of σ (in particular $|\emptyset| = 0$), $\sigma|n = (\sigma_1, \dots, \sigma_n)$ where $n \leq m$ and $t(\sigma) = \sigma_m$. Let $\sigma * \tau = (\sigma_1, \dots, \sigma_m, \tau_1, \dots, \tau_r)$ for $\tau = (\tau_1, \dots, \tau_r) \in D$.

Our fundamental space is $\Omega = (\text{Con}(X)^N)^D$ equipped with the product topology. The element of $\Omega = (\text{Con}(X)^N)^D$ will be denoted by

$$\mathcal{S} = (\mathcal{S}_\sigma)_{\sigma \in D}$$

where $\mathcal{S}_\sigma = (S_{\sigma*1}, \dots, S_{\sigma*N}) \in \text{Con}(X)^N$.

Let μ and (μ_1, \dots, μ_N) be a probability measure and an N -tuple of probability measures on $\text{Con}(X)^N$. As a probabilistic counterpart of (2') in Section 2 we define a probability measure $\langle \mu \rangle = \langle \mu: \mu_1, \dots, \mu_N \rangle$ on $\Omega = (\text{Con}(X)^N)^D$ as follows: Let $\{B_\sigma | \sigma \in \bigcup_{k=0}^m C_k\}$ be a collection of Borel sets in $\text{Con}(X)^N$, i.e. $B_\sigma \in \mathcal{B}(\text{Con}(X)^N)$, then

$$\begin{aligned} \langle \mu : \mu_1, \dots, \mu_N \rangle (\{ \mathcal{S} \in \Omega \mid \mathcal{S}_\sigma \in B_\sigma \text{ for } \sigma \in \bigcup_{k=0}^m C_k \}) \\ = \mu(\mathcal{S}_\emptyset \in B_\emptyset) \prod_{\sigma \in \bigcup_{k=1}^m C_k} \mu_{t(\sigma)}(B_\sigma), \end{aligned}$$

and Kolmogorov's extension theory determines $\langle \mu : \mu_1, \dots, \mu_N \rangle$ on Ω . Taking $\mu = \mu_k$, we have $\langle \mu_k \rangle = \langle \mu_k : \mu_1, \dots, \mu_N \rangle$ where $k = 1, \dots, N$.

Consider an $N \times N$ matrix $R(\beta) = [R(\beta)_{ij}]$ corresponding to (μ_1, \dots, μ_N) defined by

$$R(\beta)_{ij} = \int r(S_j)^\beta d\mu_i(S_1, \dots, S_N)$$

where $\beta \geq 0$ and $0^0 = 0$, and let $\lambda(\beta)$ be the maximal eigen value of non-negative matrix $R(\beta)$. Recall that $r(S)$ is the contraction ratio of a contraction S and that $r(\emptyset) = 0$.

In the following we consider an N -tuple of Borel probability measures (μ_1, \dots, μ_N) which satisfies the following conditions (3), (4) and (5):

- (3) $R(0)$ is irreducible.
- (4) If $R(0)_{ij} > 0$, then $r(S_j) > 0$ for μ_i -a.e. (S_1, \dots, S_N) .
- (5) $\lambda(0) > 1$.

Furthermore we assume that μ_0 satisfies the following condition (6):

$$(6) \quad \sum_{i=1}^N r(S_i) > 0 \text{ } \mu_0\text{-a.e. } (S_1, \dots, S_N).$$

REMARK. If $R(0)_{ij} = 0$, then $r(S_i) = 0$ for μ_i -a.e. (S_1, \dots, S_N) , because $R(0)_{ij} = \int r(S_j)^0 d\mu_i(S_1, \dots, S_N)$.

Recall that $\mathcal{X}(X)$ is the space of all non-empty compact sets of X . In order to construct a probability measure $(\mathcal{X}(X), \mathcal{E}, P_{\langle \mu_0 \rangle})$ from $(\Omega = (\text{Con}(X)^N)^D, \mathcal{B}, \langle \mu_0 \rangle)$, we state necessary results. First the following proposition is obvious by the definition of $\langle \mu_0 \rangle = \langle \mu_0 : \mu_1, \dots, \mu_N \rangle$

PROPOSITION 3.1. Define $\varphi : \text{Con}(X)^N \times \Omega^N \rightarrow \Omega$ by

$$\varphi((S_1, \dots, S_N), (\mathcal{S}^{(1)}, \dots, \mathcal{S}^{(N)})) := \mathcal{S}$$

where

$$\mathcal{S}_\emptyset = (S_1, \dots, S_N) \text{ and } \mathcal{S}_{n\sigma} = (\mathcal{S}^{(n)})_\sigma \text{ for } \sigma \in D \text{ and } n = 1, \dots, N.$$

Then φ is Borel measurable and satisfies that for every Borel set $B \subset \Omega$,

$$[\mu_0 \times \prod_{i=1}^N \langle \mu_i \rangle](\varphi^{-1}(B)) = \langle \mu_0 \rangle(B).$$

LEMMA 3.2.

$$\Omega_0 = \{ \mathcal{S} \in \Omega \mid \prod_{n=1}^{\infty} r(S_{\sigma|n}) = 0 \text{ for any } \sigma \in C_{\infty}(N) \}$$

is a Borel set with $\langle \mu_0 \rangle(\Omega_0) = 1$.

By the definition of $\text{Con}(X)^N$, it follows that

$$\bigcap_{m>0} \bigcup_{\sigma \in C_m} \overline{S_{\sigma|1} \circ \dots \circ S_{\sigma|m}}(X) \neq \emptyset.$$

PROPOSITION 3.3. Fix $\tilde{K} \in \mathcal{K}(X)$ and define $\psi: \Omega \rightarrow \mathcal{K}(X)$ by

$$\psi(\mathcal{S}) = \begin{cases} \bigcap_{m>0} \bigcup_{\sigma \in C_m} \overline{S_{\sigma|1} \circ \dots \circ S_{\sigma|m}} & \text{for } \mathcal{S} \in \Omega_0, \\ \tilde{K} & \text{for } \mathcal{S} \notin \Omega_0. \end{cases}$$

Then ψ is a Borel measurable map.

Lemma 3.2 and Proposition 3.3 are proved in Appendix 1.

DEFINITION 3.4. For an N -tuple (μ_1, \dots, μ_N) of Borel probability measures and a Borel probability measure μ_0 on $\text{Con}(X)^N$, let $P_{\langle \mu_0 \rangle}$ be the image measure of $\langle \mu_0 \rangle = \langle \mu_0: \mu_1, \dots, \mu_N \rangle$ with respect to ψ , i.e., for every Borel set $B \subset \mathcal{K}(X)$,

$$P_{\langle \mu_0 \rangle}(B) = \langle \mu_0 \rangle(\psi^{-1}(B)).$$

REMARK. A $P_{\langle \mu_0 \rangle}$ -random set is constructed as follows: Choose an N -tuple (S_1, \dots, S_N) at random with respect to the initial measure μ_0 . Let

$$A_1 = \bigcup_{k=1}^N S_k(X).$$

Then for $k = 1, \dots, N$, choose an N -tuple (S_{k1}, \dots, S_{kN}) with respect to μ_k . Set

$$A_2 = \bigcup_{k=1}^N S_k(\bigcup_{i=1}^N S_{ki}(X)).$$

Continue this process. The limit set $K = \bigcap_{n \in \mathbb{N}} \bar{A}_n$ is a $P_{\langle \mu_0 \rangle}$ -random set. This construction is a stochastic version of that of a Markov-self-similar set in Section 2.

DEFINITION 3.5. Let (μ_1, \dots, μ_N) be an N -tuple of Borel probability measures on $\text{Con}(X)^N$. An N -tuple (P_1, \dots, P_N) of probability measures on $\mathcal{K}(X)$ is called (μ_1, \dots, μ_N) -Markov-self-similar if for every Borel set $B \subset \mathcal{K}(X)$,

$$P_k(B) = [\mu_k \times \prod_{i=1}^N P_i] (\{ ((S_1, \dots, S_N), (K_1, \dots, K_N)) \in \text{Con}(X)^N \\ \times \mathcal{K}(X)^N \mid \bigcup_{i=1}^N S_i(K_i) \in B \})$$

for all $k = 1, \dots, N$.

PROPOSITION 3.6. Let (μ_1, \dots, μ_N) be an N -tuple of Borel probability measures on $\text{Con}(X)^N$. Then the N -tuple $(P_{\langle \mu_1 \rangle}, P_{\langle \mu_2 \rangle}, \dots, P_{\langle \mu_N \rangle})$ is the unique

(μ_1, \dots, μ_N) -Markov-self-similar N -tuple of probability measures on $\mathcal{X}(X)$ where $\langle \mu_k \rangle = \langle \mu_k : \mu_1, \dots, \mu_N \rangle$.

Taking $\mu_k = \delta_{(S_{k1}, \dots, S_{kN})}$ for $k = 1, \dots, N$ in Proposition 3.6, we have the statement (1) of Theorem 2.1. Proposition 3.6 is proved in Appendix 2.

The next theorem assures the existence of α such that $P_{\langle \mu_k \rangle}$ -a.e. compact set has the Hausdorff dimension α for $k = 1, \dots, N$.

THEOREM 3.7. *Let (μ_1, \dots, μ_N) and μ_0 be an N -tuple of probability measures and a probability measure on $\text{Con}(X)^N$ which satisfy the conditions (3), (4), (5) and (6). Suppose that, for $k = 1, \dots, N$, μ_k -a.e. $(S_1, \dots, S_N) \in \text{Con}(X)^N$ and every $i = 1, \dots, N$ such that $R(0)_{ki} > 0$, there exists a $c > 0$ with $d(S_i x, S_i y) \geq cd(x, y)$ for all $x, y \in X$. Then there exists an $\alpha \geq 0$ such that*

$$\dim_H(K) = \alpha$$

for $P_{\langle \mu_0 \rangle}$ -a.e. $K \in \mathcal{X}(X)$. Especially it holds that $\dim_H(K) = \alpha$ for $P_{\langle \mu_k \rangle}$ -a.e. $K \in \mathcal{X}(X)$.

Theorem 3.7 is proved in Appendix 3 and the following 0–1 law is used in the proof.

PROPOSITION 3.8. *Assume that an N -tuple (μ_1, \dots, μ_N) of Borel probability measures on $\text{Con}(X)^N$ satisfies the conditions (3) and (5). Let B be a Borel set in $\Omega = (\text{Con}(X)^N)^D$. If*

$$\langle \mu_k \rangle(B) = \prod_{i: R(0)_{ki} > 0} \langle \mu_i \rangle(B)$$

for all $k = 1, \dots, N$, then

$$\langle \mu_k \rangle(B) = 0 \quad \text{for all } k = 1, \dots, N,$$

or

$$\langle \mu_k \rangle(B) = 1 \quad \text{for all } k = 1, \dots, N.$$

PROOF. Assume that $\langle \mu_j \rangle(B) = 0$ for some $j \in \{1, \dots, N\}$. Using the irreducibility of $R(0)$ we deduce that $\langle \mu_k \rangle(B) = 0$ for all $k = 1, \dots, N$. Now assume that $\langle \mu_k \rangle(B) \neq 0$. Note that

$$\prod_{k=1}^N \langle \mu_k \rangle(B) = \prod_{k=1}^N \prod_{i: R(0)_{ki} \neq 0} \langle \mu_i \rangle(B)$$

and that

$$\sum_{k=1}^N \# \{i \mid R(0)_{ki} \neq 0\} > N$$

because $\lambda(0) > 1$. Therefore there exists a $j \in \{1, \dots, N\}$ such that

$$\langle \mu_j \rangle(B) = 1.$$

Using the irreducibility of $R(0)$ we deduce that

$$\langle \mu_k \rangle(B) = 1 \quad \text{for all } k = 1, \dots, N.$$

REMARK. Under the assumptions of Proposition 3.8, the statement in Proposition 3.8 is true for $(P_{\langle \mu_1 \rangle}, \dots, P_{\langle \mu_N \rangle})$: Let B be a Borel set in $\mathcal{K}(X)$. If

$$P_{\langle \mu_k \rangle}(B) = \prod_{i: R(0)_{ki} \neq 0} P_{\langle \mu_i \rangle}(B)$$

for all $k = 1, \dots, N$, then

$$P_{\langle \mu_k \rangle}(B) = 0 \quad \text{for all } k = 1, \dots, N,$$

or

$$P_{\langle \mu_k \rangle}(B) = 1 \quad \text{for all } k = 1, \dots, N.$$

An upper bound for the Hausdorff dimension of $P_{\langle \mu_0 \rangle}$ -random sets is given by the following proposition which is an extension of the result obtained by Mauldin-Williams [10], Falconer [5] and Graf [7].

PROPOSITION 3.9. *Let (μ_1, \dots, μ_N) and μ_0 be an N -tuple of probability measures and a probability measure on $\text{Con}(X)^N$ which satisfy the condition (5). Let α be such that $\lambda(\alpha) = 1$. Then*

$$E_{P_{\langle \mu_0 \rangle}}(\mathcal{H}^\alpha(K)) < \infty.$$

In particular

$$\mathcal{H}^\alpha(K) < \infty \quad \text{for } P_{\langle \mu_0 \rangle}\text{-a.e. } K \in \mathcal{K}(X)$$

and

$$\dim_H(K) \leq \alpha \quad \text{for } P_{\langle \mu_0 \rangle}\text{-a.e. } K \in \mathcal{K}(X).$$

Especially we have the corresponding statements for $P_{\langle \mu_k \rangle}$ -a.e. K .

REMARK. The uniqueness of α for which $\lambda(\alpha) = 1$ follows from the fact that $\lambda(\beta)$ is continuous and strictly decreasing with respect to β .

The proof of Proposition 3.9. is given in Appendix 4. In the proof we use the following martingale convergence theorem (Theorem 3.10). Let Γ be a subset in D , and define $f_{\Gamma, \beta}^{(k)}: (\Omega, \mathcal{B}, \langle \mu_k \rangle) \rightarrow \mathbf{R}_+$ by

$$f_{\Gamma, \beta}^{(k)}(\mathcal{S}) = \sum_{\sigma \in \Gamma} \left[\prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\beta \right] x_{t(\sigma)}$$

and

$$f_{\{\emptyset\}, \beta}^{(k)}(\mathcal{S}) = x_k,$$

for $k = 1, \dots, N$ where (x_1, \dots, x_N) is an positive eigenvector of $R(\alpha)$

corresponding to the maximal eigen value 1. We abbreviate $f_{C_{m,\beta}}^{(k)}$ by $f_{m,\beta}^{(k)}$.

THEOREM 3.10. *Let (μ_1, \dots, μ_N) be an N -tuple of probability measures on $\text{Con}(X)^N$ which satisfies the conditions (3) and (5). Let α be the unique value such that $\lambda(\alpha) = 1$. For $m \in \mathbb{N}$ let \mathcal{B}_m be the σ -field of all Borel subsets in $\Omega = (\text{Con}(X)^N)^{\mathcal{D}}$ depending only on coordinates from $D_m = \bigcup_{k \leq m} C_m$. Then for every $p \in \mathbb{N}$ and $k = 1, \dots, N$, $(f_{m,\alpha}^{(k)})_{m \in \mathbb{N}}$ is an L^p -bounded martingale with respect to $(\mathcal{B}_m)_{m \in \mathbb{N}}$ which converges $\langle \mu_k \rangle$ -a.e. and in $L^p(\Omega, \langle \mu_k \rangle)$ to a function $f^{(k)}$. Furthermore if the condition (4) holds, then $f^{(k)} > 0$ for $\langle \mu_k \rangle$ -a.e. and $k = 1, \dots, N$.*

Theorem 3.10 is proved in Appendix 4.

The following theorem gives conditions which assure that, for $P_{\langle \mu_k \rangle}$ -a.e. compact sets, the Hausdorff dimension is equal to α .

THEOREM 3.11. *Let $X \subset \mathbb{R}^d$ be a compact set with the non-empty interior $\overset{\circ}{X}$. Let (μ_1, \dots, μ_N) and μ_0 be an N -tuple of probability measures and a probability measure on $\text{Con}(X)^N$ which satisfy the conditions (3), (4), (5) and (6). Suppose that, for μ_k -a.e. $(S_1, \dots, S_N) \in \text{Con}(X)^N$ and $k = 1, \dots, N$, the following conditions are satisfied.*

- a) For all $i = 1, \dots, N$, S_i is a contraction similarity or the null contraction ϕ .
- b) (S_1, \dots, S_N) satisfies the following open set condition: $S_i(\overset{\circ}{X}) \cap S_j(\overset{\circ}{X}) = \emptyset$ if $i \neq j$.

Let $\alpha \geq 0$ be such that $\lambda(\alpha) = 1$. Then $\dim_{\text{H}}(K) = \alpha$ for $P_{\langle \mu_0 \rangle}$ -a.e. $K \in \mathcal{X}(X)$. Especially $\dim_{\text{H}}(K) = \alpha$ for $P_{\langle \mu_k \rangle}$ -a.e. $K \in \mathcal{X}(X)$ and $k = 1, \dots, N$.

Theorem 3.11 is proved in Appendix 5.

EXAMPLE. Let $X = [0, 1]$ and $N = 2$. Let T_1, T_2 and T_3 be similarities which map $[0, 1]$ to $[0, 1/3]$, $[1/3, 2/3]$ and $[2/3, 1]$ respectively, and $\tilde{T}_1, \tilde{T}_2, \tilde{T}_3$ and \tilde{T}_4 be similarities which map $[0, 1]$ to $[0, 1/4]$, $[1/4, 1/2]$, $[1/2, 3/4]$ and $[3/4, 1]$ respectively. Let

$$\mu_1 = 3^{-1} \{ \varepsilon_{(T_1, T_2)} + \varepsilon_{(T_2, T_3)} + \varepsilon_{(T_1, T_3)} \}$$

and

$$\mu_2 = 6^{-1} \sum_{1 \leq i < j \leq 4} \varepsilon_{(\tilde{T}_i, \tilde{T}_j)}.$$

Then (μ_1, μ_2) is a pair of probability measures on $\text{Con}(X)^2$, and it satisfies the conditions (3), (4) and (5). By Theorem 3.11,

$$\dim_{\text{H}}(K) = \alpha \text{ for } P_{\langle \mu_k \rangle}\text{-a.e. } K \in \mathcal{X}([0, 1]) \text{ and } k = 1, 2$$

where α is such that $(1/3)^\alpha + (1/4)^\alpha = 1$.

4. Hausdorff measures of random Markov-self-similar sets

First we state a theorem which corresponds to Theorem 7.8 of Graf [7].

THEOREM 4.1. *Let the assumptions of Theorem 3.11 be satisfied. Suppose that there exists a $\delta > 0$ such that if $R(0)_{ki} > 0$, then $r(S_i) \geq \delta$ for μ_k -a.e. (S_1, \dots, S_N) , $k = 1, \dots, N$. Let (x_1, \dots, x_N) be a positive eigenvector of $R(\alpha)$ corresponding to the maximal eigen value 1. Then the following statements are equivalent:*

- a) $\sum_{i=1}^N r(S_i)^\alpha x_i = x_k$ for μ_k -a.e. (S_1, \dots, S_N) and all $k \in \{1, \dots, N\}$.
- b) $\mathcal{H}^\alpha(K) > 0$ for $P_{\langle \mu_k \rangle}$ -a.e. $K \in \mathcal{K}(X)$ and all $k \in \{1, \dots, N\}$.
- c) $P_{\langle \mu_j \rangle}(\{K \in \mathcal{K}(X) | \mathcal{H}^\alpha(K) > 0\}) > 0$ for some $j \in \{1, \dots, N\}$.

Theorem 4.1 is proved in Appendix 6.

The following theorem gives an information about the α -dimensional Haudorff measure $\mathcal{H}^\alpha(K)$ for $P_{\langle \mu_k \rangle}$ -a.e. $K \in \mathcal{K}(X)$ for Markov-self-similar $(P_{\langle \mu_N \rangle}, \dots, P_{\langle \mu_1 \rangle})$. See [13] and [14].

THEOREM 4.2. *Let the assumptions and the condition c) of Theorem 4.1 be satisfied. Then there exists a $c > 0$ such that*

$$\mathcal{H}^\alpha(K) = cx_k$$

for $P_{\langle \mu_k \rangle}$ -a.e. $K \in \mathcal{K}(X)$ and all $k \in \{1, \dots, N\}$.

For the proof of Theorem 4.2 we show the following lemma:

LEMMA 4.3. *Assume that $0 < E_{\langle \mu_k \rangle}(\mathcal{H}^\alpha(K(\mathcal{S}))) < \infty$ for $k = 1, \dots, N$ and that*

$$\sum_{i=1}^N r(S_i)^\alpha x_i = x_k$$

for μ_k -a.e. (S_1, \dots, S_N) and $k = 1, \dots, N$. Then it holds that

$$\mathcal{H}^\alpha(K(\mathcal{S})) = \sum_{i=1}^N r(S_i(\mathcal{S}))^\alpha \mathcal{H}^\alpha(K(\mathcal{S}^{(i)}))$$

for $\langle \mu_k \rangle$ -a.e. \mathcal{S} and $k = 1, \dots, N$. Furthermore there exists a $c > 0$ such that

$$E_{\langle \mu_k \rangle}(\mathcal{H}^\alpha(K(\mathcal{S}))) = cx_k \quad \text{for } k = 1, \dots, N.$$

PROOF. Since

$$K(\mathcal{S}) = \bigcup_{i=1}^N S_i(K(\mathcal{S}^{(i)}))$$

and S_i are similarities, it follows that

$$\mathcal{H}^\alpha(K(\mathcal{S})) \leq \sum_{i=1}^N r(S_i(\mathcal{S}))^\alpha \mathcal{H}^\alpha(K(\mathcal{S}^{(i)})).$$

Integrating the both sides with respect to $\langle \mu_k \rangle$ and using Proposition 3.1,

$$E_{\langle \mu_k \rangle}[\mathcal{H}^\alpha(K(\mathcal{S}))] \leq \sum_{i=1}^N R(\alpha)_{ki} E_{\langle \mu_i \rangle}[\mathcal{H}^\alpha(K(\mathcal{S}))]$$

for $k = 1, \dots, N$. Since $0 < E_{\langle \mu_k \rangle}(\mathcal{H}^\alpha(K(\mathcal{S}))) < \infty$, we deduce, by Theorem 2.4 (Frobenius), that there exists a $c > 0$ such that

$$\begin{aligned} E_{\langle \mu_k \rangle}[\mathcal{H}^\alpha(K(\mathcal{S}))] &= cx_k \quad \text{for } k = 1, \dots, N \\ E_{\langle \mu_k \rangle}[\mathcal{H}^\alpha(K(\mathcal{S}))] &= \sum_{i=1}^N R(\alpha)_{ki} E_{\langle \mu_i \rangle}[\mathcal{H}^\alpha(K(\mathcal{S}))] \end{aligned}$$

for $k = 1, \dots, N$. Therefore

$$\mathcal{H}^\alpha(K(\mathcal{S})) = \sum_{i=1}^N r(S_i(\mathcal{S}))^\alpha \mathcal{H}^\alpha(K(\mathcal{S}^{(i)}))$$

for $\langle \mu_k \rangle$ -a.e. \mathcal{S} and $k = 1, \dots, N$. This completes the proof.

PROOF OF THEOREM 4.2. Proposition 3.9 and Theorem 4.1 assure the assumptions of Lemma 4.3. Iterating Lemma 4.3, we have

$$\begin{aligned} \mathcal{H}^\alpha(K(\mathcal{S})) &= \sum_{i_1=1}^N r(S_{i_1}(\mathcal{S}))^\alpha \sum_{i_2=1}^N r(S_{i_2}(\mathcal{S}^{(i_1)}))^\alpha \sum \dots \\ &\quad \sum_{i_m=1}^N r(S_{i_m}(\mathcal{S}^{(i_1)\dots(i_{m-1})}))^\alpha \mathcal{H}^\alpha(K(\mathcal{S}^{(i_1)\dots(i_m)})) \end{aligned}$$

for $\langle \mu_k \rangle$ -a.e. \mathcal{S} and $k = 1, \dots, N$ where $\mathcal{S}^{(i_1)(i_2)} = (\mathcal{S}^{(i_1)})^{(i_2)}$ and so on. Consider $E_{\langle \mu_k \rangle}[\mathcal{H}^\alpha(K(\mathcal{S})) | \mathcal{B}_{m-1}]$ where \mathcal{B}_{m-1} are the σ -field of all Borel subsets in $\Omega = (\text{Con}(X)^N)^D$ depending only on coordinates from $\cup_{i \leq m-1} C_i$. Using Proposition 3.1 we have

$$\begin{aligned} E_{\langle \mu_k \rangle}[\mathcal{H}^\alpha(K(\mathcal{S})) | \mathcal{B}_{m-1}] &= \\ &\sum_{i_1=1}^N r(S_{i_1}(\mathcal{S}))^\alpha \sum_{i_2=1}^N r(S_{i_2}(\mathcal{S}^{(i_1)}))^\alpha \dots \\ &\sum_{i_m=1}^N r(S_{i_m}(\mathcal{S}^{(i_1)\dots(i_{m-1})}))^\alpha E_{\langle \mu_{i_m} \rangle}[\mathcal{H}^\alpha(K(\mathcal{S}))]. \end{aligned}$$

Since $\sum_{i=1}^N r(S_i)^\alpha x_i = x_k$ and $E_{\langle \mu_k \rangle}(\mathcal{H}^\alpha(K(\mathcal{S}))) = cx_k$, it follows that

$$E_{\langle \mu_k \rangle}[\mathcal{H}^\alpha(K(\mathcal{S})) | \mathcal{B}_{m-1}] = cx_k.$$

As m is arbitrary, we have

$$\mathcal{H}^\alpha(K) = cx_k \quad \text{for } P_{\langle \mu_k \rangle}\text{-a.e. } K \in \mathcal{K}(X) \text{ and } k = 1, \dots, N.$$

REMARK. In the case of $\mathcal{H}^\alpha(K) = 0$ for a.e. K , the exact Hausdorff dimension of K was investigated by Graf, Mauldin and Williams [8].

EXAMPLE. Consider the example stated at the end of Section 3. Theorem 4.2 implies that

$$\mathcal{H}^\alpha(K) = c(1/3)^\alpha \quad \text{for } P_{\langle \mu_1 \rangle}\text{-a.e. } K \in \mathcal{K}(X)$$

and

$$\mathcal{H}^\alpha(K) = c(1/4)^\alpha \quad \text{for } P_{\langle \mu_2 \rangle}\text{-a.e. } K \in \mathcal{H}(X)$$

for some $c > 0$.

APPENDIX

1. Proof of Lemma 3.2 and Proposition 3.3

PROOF OF LEMMA 3.2 (cf. the proof of Lemma 3.2 of Graf [7]). The result that Ω_0 is a Borel set is proved in Lemma 3.2 of Graf [7]. We show that $\langle \mu_0 \rangle(\Omega_0) = 1$. By Proposition 3.1, it suffices to prove that $\langle \mu_k \rangle(\Omega_0) = 1$ for $k = 1, \dots, N$. For $a > 0$ set

$$B_a = \{ \mathcal{S} \in \Omega \mid \text{there exists } \sigma \in \{1, \dots, N\}^N \text{ such that } \prod_{n=0}^\infty r(S_{\sigma|n}) \geq a \},$$

then the fact that B_a is Borel measurable is also proved in Lemma 3.2 of Graf [7].

Define $p_k: (0, 1) \rightarrow [0, 1]$ by $p_k(a) = \langle \mu_k \rangle(B_a)$ for $k = 1, \dots, N$. It follows that from Proposition 3.1 that, for every $a \in (0, 1)$, we have

$$\begin{aligned} \text{(a1) } p_k(a) &= [\mu_k \times \prod_{i=1}^N \langle \mu_i \rangle](\{((S_1, \dots, S_N), (\mathcal{S}^{(1)}, \dots, \mathcal{S}^{(N)})) \mid \text{there exist} \\ &\quad j \in \{1, \dots, N\} \text{ and } \sigma \in \{1, \dots, N\}^N \text{ such that } r(S_j) \prod_{n=0}^\infty r(\mathcal{S}_{\sigma|n}^{(j)}) \geq a\}) \\ &\leq \sum_{j=1}^N [\mu_k \times \prod_{i=1}^N \langle \mu_i \rangle](\{((S_1, \dots, S_N), (\mathcal{S}^{(1)}, \dots, \mathcal{S}^{(N)})) \mid \text{there exists} \\ &\quad \sigma \in \{1, \dots, N\}^N \text{ such that } r(S_j) \prod_{n=0}^\infty r(S_{\sigma|n}^{(j)}) \geq a\}) \\ &\leq \sum_{j=1}^N \mu_k(\{(S_1, \dots, S_N) \mid r(S_j) \geq a\}) p_j(a). \end{aligned}$$

Since $r(S) < 1$ there exists a $b \in (0, 1)$ such that

$$\mu_j(\{(S_1, \dots, S_N) \mid \max_{1 \leq i \leq N} r(S_i) \geq b\}) < 1/N$$

for all $j \in \{1, \dots, N\}$. If there exists a k such that $p_k(b) > 0$, let k_1 be such that $p_{k_1}(b) = \max_k p_k(b) > 0$. Then it follows from (a1) that $p_k(b) < p_{k_1}(b)$. This contradiction implies that $p_k(b) = 0$ for all $k = 1, \dots, N$.

Let $\eta_k = \inf\{a \in (0, 1) \mid p_k(a) = 0\}$ for $k = 1, \dots, N$, and $\eta = \max_{1 \leq k \leq N} \eta_k < 1$. Assume $\eta > 0$. Then there is an $a > \eta$ with $ab < \eta$. We deduce as before

$$\begin{aligned} p_k(ab) &\leq \sum_{j=1}^N [\mu_k \times \prod_{i=1}^N \langle \mu_i \rangle](\{((S_1, \dots, S_N), (\mathcal{S}^{(1)}, \dots, \mathcal{S}^{(N)})) \mid \text{there exists} \\ &\quad \sigma \in \{1, \dots, N\}^N \text{ such that } r(S_j) \prod_{n=0}^\infty r(S_{\sigma|n}^{(j)}) \geq ab\}). \end{aligned}$$

Since $a > \eta$ we have $p_j(a) = 0$ for $j = 1, \dots, N$, and so

$$\prod_{n=0}^{\infty} r(S_{\sigma(n)}^{(j)}) \leq a \text{ for } \langle \mu_j \rangle\text{-a.e. } \mathcal{S}^{(j)} \text{ and } j = 1, \dots, N.$$

This leads to

$$p_k(ab) \leq \sum_{j=1}^N \mu_k(\{(S_1, \dots, S_N) | r(S_j) \geq b\}) p_j(ab)$$

for $k = 1, \dots, N$. Assume that there exists a k such that $p_k(ab) > 0$. As before this leads to a contradiction, so $p_k(ab) = 0$ for all $k = 1, \dots, N$. This contradicts $ab < \eta$ and the definition of η . Thus $\eta = 0$ and p_k vanishes identically for $k = 1, \dots, N$. This completes the proof.

PROOF OF PROPOSITION 3.3. The proof of Theorem 3.7 of Graf [7] using Lemma 3.2 instead of Lemma 3.2 of Graf [7] implies Proposition 3.3.

2. Proof of Proposition 3.6. (cf. the proof of Theorem 4.5 of Graf [7])

First we give a definition.

DEFINITION. Let (μ_1, \dots, μ_N) be an N -tuple of probability measures on $\text{Con}(X)^N$. For $k = 1, \dots, N$, define $T_k = T_k^{(\mu_1, \dots, \mu_N)}: P(\mathcal{X}(X))^N \rightarrow P(\mathcal{X}(X))$ by

$$[T_k(Q_1, \dots, Q_N)](B) = [\mu_k \times \prod_{i=1}^N Q_i] (\{(S_1, \dots, S_N), (K_1, \dots, K_N) | \bigcup_{1 \leq j \leq N} S_j(K_j) \in B\})$$

where $P(\mathcal{X}(X))$ is the set of all Borel probability measures on $\mathcal{X}(X)$.

REMARK. An N -tuple (P_1, \dots, P_N) of probability measures on $\mathcal{X}(X)$ is (μ_1, \dots, μ_N) -Markov-self-similar if and only if

$$P_k = T_k^{(\mu_1, \dots, \mu_N)}(P_1, \dots, P_N)$$

for all $k = 1, \dots, N$.

PROOF OF PROPOSITION 3.6. The proof of Theorem 4.5 of Graf [7] assures that

$$T_k(P_{\langle \mu_1 \rangle}, \dots, P_{\langle \mu_N \rangle}) = P_{\langle \mu_k \rangle}$$

for $k = 1, \dots, N$.

Define $T: P(\mathcal{X}(X))^N \rightarrow P(\mathcal{X}(X))^N$ by

$$T(Q_1, \dots, Q_N) = (T_1(Q_1, \dots, Q_N), \dots, T_N(Q_1, \dots, Q_N))$$

for $(Q_1, \dots, Q_N) \in P(\mathcal{X}(X))^N$. Let $A \subset \mathcal{X}(X)$ be a closed set. Using induction on n , we have

$$\begin{aligned} & (T^n(Q_1, \dots, Q_N))_k(A) \\ &= [\langle \mu_k \rangle \times (\prod_{i=1}^N Q_i)^D] (\{(\mathcal{S}, (K_{\sigma^*1}, \dots, K_{\sigma^*N})_{\sigma \in D}) \in \Omega \times (\mathcal{X}(X)^N)^D \\ & \quad | \bigcup_{\sigma \in C_{n-1}} \bigcup_{i=1}^N S_{\sigma|1} \circ \dots \circ S_{\sigma|n-1} \circ S_{\sigma^*i}(K_{\sigma^*i}) \in A\}) \end{aligned}$$

Hence we obtain

$$\begin{aligned} & \lim_{n \rightarrow \infty} \sup (T^n(Q_1, \dots, Q_N))_k(A) \\ &= \inf_m \sup_{n \geq m} [\langle \mu_k \rangle \times (\prod_{i=1}^N Q_i)^D] (\{(\mathcal{S}, (K_{\sigma^*1}, \dots, K_{\sigma^*N})_{\sigma \in D}) \in \Omega \times \\ & \quad (\mathcal{X}(X)^N)^D | \bigcup_{\sigma \in C_{n-1}} \bigcup_{i=1}^N S_{\sigma|1} \circ \dots \circ S_{\sigma|n-1} \circ S_{\sigma^*i}(K_{\sigma^*i}) \in A\}) \\ &\leq [\langle \mu_k \rangle \times (\prod_{i=1}^N Q_i)^D] (\bigcap_{n \geq m} \{(\mathcal{S}, (K_{\sigma^*1}, \dots, K_{\sigma^*N})_{\sigma \in D}) \in \Omega \times \\ & \quad (\mathcal{X}(X)^N)^D | \bigcup_{\sigma \in C_{n-1}} \bigcup_{i=1}^N S_{\sigma|1} \circ \dots \circ S_{\sigma|n-1} \circ S_{\sigma^*i}(K_{\sigma^*i}) \in A\}) \\ &\leq [\langle \mu_k \rangle \times (\prod_{i=1}^N Q_i)^D] (\{(\mathcal{S}, (K_{\sigma^*1}, \dots, K_{\sigma^*N})_{\sigma \in D}) \in \Omega \times (\mathcal{X}(X)^N)^D | \\ & \quad \lim_{n \rightarrow \infty} \bigcup_{\sigma \in C_{n-1}} \bigcup_{i=1}^N S_{\sigma|1} \circ \dots \circ S_{\sigma|n-1} \circ S_{\sigma^*i}(K_{\sigma^*i}) \in A\}). \end{aligned}$$

By Theorem 2.2 of Graf [7] and the definition of ψ , the last expression equals to

$$\begin{aligned} & [\langle \mu_k \rangle \times (\prod_{i=1}^N Q_i)^D] (\{(\mathcal{S}, (K_{\sigma^*1}, \dots, K_{\sigma^*N})_{\sigma \in D}) \in \Omega \times (\mathcal{X}(X)^N)^D | \psi(\mathcal{S}) \in A\}) \\ &= \langle \mu_k \rangle (\psi^{-1}(A)). \end{aligned}$$

Therefore it holds that

$$\lim_{n \rightarrow \infty} \sup (T^n(Q_1, \dots, Q_N))_k(A) \leq P_{\langle \mu_k \rangle}(A).$$

Since this is true for an arbitrary closed set A of $\mathcal{X}(X)$, $\{(T^n(Q_1, \dots, Q_N))_k\}_{n \in \mathbb{N}}$ converges to $P_{\langle \mu_k \rangle}$ in the weak topology. This implies the uniqueness of the (μ_1, \dots, μ_N) -Markov-self-similar probability measure.

3. Proof of Theorem 3.7

First we show the following 0–1 law (cf. Theorem 7.2 of Graf [7]):

LEMMA A. For a given $\beta \geq 0$, it holds that

- (a) $P_{\langle \mu_k \rangle}(\{K \in \mathcal{X}(X) | \mathcal{H}^\beta(K) = 0\}) = 0$ for all $k = 1, \dots, N$, or $= 1$ for all $k = 1, \dots, N$,

and that

- (b) $P_{\langle \mu_k \rangle}(\{K \in \mathcal{X}(X) | \mathcal{H}^\beta(K) = \infty\}) = 0$ for all $k = 1, \dots, N$, or $= 1$ for all $k = 1, \dots, N$.

PROOF. By Proposition 3.6 we have

$$\begin{aligned} P_{\langle \mu_k \rangle}(\{K \in \mathcal{K}(X) | \mathcal{H}^\beta(K) = 0\}) &= [\mu_k \times \prod_{i=1}^N P_{\langle \mu_i \rangle}](\{(S_1, \dots, S_N), (K_1, \dots, \\ &\quad K_N) | \mathcal{H}^\beta(\bigcup_{j=1}^N S_j(K_j)) = 0\}) \\ &= [\mu_k \times \prod_{i=1}^N P_{\langle \mu_i \rangle}](\{(S_1, \dots, S_N), (K_1, \dots, K_N) | \mathcal{H}^\beta(S_j(K_j)) = 0 \\ &\quad \text{for } j = 1, \dots, N\}) \\ &= \prod_{i: R(0)_{ki} \neq 0} P_{\langle \mu_i \rangle}(\{K | \mathcal{H}^\beta(K_i) = 0\}). \end{aligned}$$

By the remark of Proposition 3.8 we have (a). The fact (b) follows in the same way because

$$P_{\langle \mu_i \rangle}(\{K | \mathcal{H}^\beta(K_i) = \infty\}) = 1 - P_{\langle \mu_i \rangle}(\{K | \mathcal{H}^\beta(K_i) < \infty\}).$$

PROOF OF THEOREM 3.7. It is easy to prove the theorem using (a) and (b). See the proof of Corollary 7.3 of Graf [7].

4. Proof of Proposition 3.9 and Theorem 3.10

First we prove Theorem 3.10 (cf. the proof of Theorem 6.3 of Graf [7]).

PROOF OF THEOREM 3.10. Since

$$\begin{aligned} E_{\langle \mu_k \rangle} [f_{q+1, \alpha}^{(k)} | \mathcal{B}_q] &= E_{\langle \mu_k \rangle} [\sum_{\tau \in C_{q+1}} \prod_{n=1}^{q+1} r(S_{\tau|n})^\alpha x_{\tau(n)} | \mathcal{B}_q] \\ &= \sum_{\sigma \in C_q} \prod_{n=1}^q r(S_{\sigma|n})^\alpha E_{\langle \mu_{\tau(\sigma)} \rangle} [\sum_{i=1}^N r(S_i)^\alpha x_i] \\ &= \sum_{\sigma \in C_q} \prod_{n=1}^q r(S_{\sigma|n})^\alpha x_{\tau(\sigma)} = f_{q, \alpha}^{(k)} \end{aligned}$$

$\langle \mu_k \rangle$ -a.e. \mathcal{S} for $q \geq 1$ and

$$E_{\langle \mu_k \rangle} [f_{1, \alpha}^{(k)} | \mathcal{B}_0] = \sum_{i=1}^N R(\alpha)_{ki} x_i = x_k = f_{0, \alpha}^{(k)},$$

$(f_{q, \alpha}^{(k)})_{q \in \mathbb{N}}$ is a martingale with respect to $(\mathcal{B}_q)_{q \in \mathbb{N}}$.

By induction on $p \in \mathbb{N}$ we prove $(f_{q, \alpha}^{(k)})_{q \in \mathbb{N}}$ is L^p -bounded. Since $f_{q, \alpha}^{(k)} \geq 0$ and $(f_{q, \alpha}^{(k)})_{q \in \mathbb{N}}$ is a martingale, it is L^1 -bounded. Now assume that $p > 1$ and that for $m < p$, $(f_{q, \alpha}^{(k)})_{q \in \mathbb{N}}$ is L^m -bounded for all $k = 1, \dots, N$. Let

$$M = \sup \{ \|f_{q, \alpha}^{(k)}\|_m | q \in \mathbb{N}, m < p, k = 1, \dots, N \} < \infty,$$

$$L = \max \{ \|f_{0, \alpha}^{(k)}\|_p^p / x_k | k = 1, \dots, N \} < \infty,$$

$$\begin{aligned} C &= \max \left\{ \frac{1}{x_k} \int (\sum_{i=1}^N r(S_i)^\alpha)^p d\mu_k(S_1, \dots, S_N) | k = 1, \dots, N \right\} \\ &\leq N^p / \min_{1 \leq k \leq N} x_k \end{aligned}$$

and

$$\delta = \max \left\{ \int \sum_{i=1}^N r(S_i)^{p\alpha} \frac{x_i}{x_k} d\mu_k(S_1, \dots, S_N) \mid k = 1, \dots, N \right\}.$$

Note that $\delta < 1$ by Theorem 2.4 (Frobenius) because the maximal eigen value of $R(p\alpha)$ is smaller than one. We show by induction on q that

$$(a2) \quad \|f_{q,\alpha}^{(k)}\|_p^p \leq x_k(\delta^q L + M^p \cdot C \cdot \sum_{i=0}^{q-1} \delta^i).$$

For $q = 0$ it is obvious. Assume that (a2) holds for $q = 1, \dots, n$. For $q = n + 1$, we have

$$\begin{aligned} \|f_{n+1,\alpha}^{(k)}\|_p^p &= \int (f_{n+1,\alpha}^{(k)})^p d\langle \mu_k \rangle \\ &= \iint \left\{ \sum_{i=1}^N r(S_i)^\alpha f_{n,\alpha}^{(i)}(\mathcal{S}^{(i)}) \right\}^p \prod_{i=1}^N d\langle \mu_i \rangle(\mathcal{S}^{(i)}) d\mu_k(S_1, \dots, S_N) \\ &= \sum_{v_1 + \dots + v_N = p} \frac{p!}{v_1! \dots v_N!} \int r(S_1)^{v_1\alpha} \dots r(S_N)^{v_N\alpha} \|f_{n,\alpha}^{(1)}\|_{v_1}^{v_1} \dots \|f_{n,\alpha}^{(N)}\|_{v_N}^{v_N} \\ &\quad d\mu_k(S_1, \dots, S_N) \\ &= \int (r(S_1)^{p\alpha} \|f_{n,\alpha}^{(1)}\|_p^p + \dots + r(S_N)^{p\alpha} \|f_{n,\alpha}^{(N)}\|_p^p) d\mu_k(S_1, \dots, S_N) + \sum_{v_1 + \dots + v_N = p} \\ &\quad \frac{P!}{v_1! \dots v_N!} \int r(S_1)^{v_1\alpha} \dots r(S_N)^{v_N\alpha} \|f_{n,\alpha}^{(1)}\|_{v_1}^{v_1} \dots \|f_{n,\alpha}^{(N)}\|_{v_N}^{v_N} d\mu_k(S_1, \dots, S_N) \\ &\leq \int \sum_{i=1}^N r(S_i)^{p\alpha} x_i (\delta^n L + M^p \cdot C \cdot \sum_{i=0}^{n-1} \delta^i) d\mu_k(S_1, \dots, S_N) + \\ &\quad M^p \sum_{v_1 + \dots + v_N = p} \frac{p!}{v_1! \dots v_N!} \int r(S_1)^{v_1\alpha} \dots r(S_N)^{v_N\alpha} d\mu_k(S_1, \dots, S_N) \\ &\leq x_k (\delta^n L + M^p \cdot C \cdot \sum_{i=0}^{n-1} \delta^i) + M^p \int (\sum_{i=1}^N r(S_i)^\alpha)^p d\mu_k(S_1, \dots, S_N) \\ &= x_k (\delta^{n+1} L + M^p C \sum_{i=0}^n \delta^i). \end{aligned}$$

Since $\delta < 1$, we deduce that $(f_{q,\alpha}^{(k)})_{q \in \mathbf{N}}$ is L^p -bounded.

We show that $f^{(k)} > 0$ for $\langle \mu_k \rangle$ -a.e. and $k = 1, \dots, N$ if the condition (4) holds. Using Proposition 3.1 and Lemma 6.4 of Graf [7], we deduce

$$\begin{aligned} \langle \mu_k \rangle(\{\mathcal{S} \mid f^{(k)}(\mathcal{S}) = 0\}) &= [\mu_k \times \prod_{i=1}^N \langle \mu_i \rangle](\{(S_1, \dots, S_N), (\mathcal{S}^{(1)}, \dots, \mathcal{S}^{(N)}) \mid \\ &\quad \sum_{i=1}^N r(S_i)^\alpha f^{(i)}(\mathcal{S}^{(i)}) = 0\}) \end{aligned}$$

$$= \prod_{i:R(0)_{ki} \neq 0} \langle \mu_i \rangle (\{\mathcal{S} | f^{(i)}(\mathcal{S}) = 0\}).$$

By Proposition 3.8 and the fact that $E_{\langle \mu_k \rangle} [f^{(k)}] = x_k > 0$, we deduce that $\langle \mu_i \rangle (\{\mathcal{S} | f^{(i)}(\mathcal{S}) > 0\}) = 1$ for all $i = 1, \dots, N$. This completes the proof.

A subset $\Gamma \subset D$ is called a minimal covering if for each $\eta \in \{1, \dots, N\}^{\mathbb{N}}$ there exists a unique $\sigma \in \Gamma$ such that $\eta|j = \sigma$ for some $j \in \mathbb{N}$. Let $\text{Min} = \{\Gamma \subset D | \Gamma \text{ is a minimal covering}\}$. For $\Gamma_1, \Gamma_2 \subset D$, we write $\Gamma_1 < \Gamma_2$ if for every $\sigma_1 \in \Gamma_1$ there exists $\sigma_2 \in \Gamma_2$ such that $\sigma_2|j = \sigma_1$ for some $j \in \mathbb{N}$.

COROLLARY OF THEOREM 3.10 (cf. Corollary 6.5 of Graf [7]). *Let the assumptions of Theorem 3.10 be satisfied. Then*

$$E_{\langle \mu_k \rangle} [\sup_{\Gamma_0 \in \text{Min}} \inf \{f_{\Gamma, \alpha}^{(k)} | \Gamma \in \text{Min}, \Gamma > \Gamma_0\}] < \infty$$

for $k = 1, \dots, N$. In particular

$$\sup_{\Gamma_0 \in \text{Min}} \inf \{f_{\Gamma, \alpha}^{(k)} | \Gamma \in \text{Min}, \Gamma > \Gamma_0\} < \infty$$

for $\langle \mu_k \rangle$ -a.e. $\mathcal{S} \in \Omega$ and $k = 1, \dots, N$.

PROOF. For $\langle \mu_k \rangle$ -a.e. \mathcal{S} we have

$$\sup_{\Gamma_0 \in \text{Min}} \inf \{f_{\Gamma, \alpha}^{(k)} | \Gamma \in \text{Min}, \Gamma > \Gamma_0\} \leq \sup_{q_0 \in \mathbb{N}} \inf_{q \geq q_0} f_{q, \alpha}^{(k)}(\mathcal{S}) = f^{(k)}(\mathcal{S}).$$

Since $\int f^{(k)} d\langle \mu_k \rangle < \infty$ by Theorem 3.10 the corollary is proved.

For the proof of Proposition 3.9 we state a result in Graf [7].

THEOREM 2.4 OF GRAF [7]. *Let $\mathcal{S} \in \Omega_0$ be given. Then, for every $\beta > 0$,*

$$\mathcal{H}^\beta(K(\mathcal{S})) \leq |X|^\beta \sup_{\Gamma_0 \in \text{Min}} \inf \left\{ \sum_{\sigma \in \Gamma} \prod_{n=1}^{\sigma} r(S_{\sigma|n})^\beta | \Gamma \in \text{Min}, \Gamma > \Gamma_0 \right\}.$$

PROOF OF PROPOSITION 3.9 (cf. the proof of Theorem 7.4 of Graf [7]). We show that $E_{P_{\langle \mu_k \rangle}} [\mathcal{H}^\alpha(K)] < \infty$ for $k = 1, \dots, N$. Let $\psi: \Omega \rightarrow \mathcal{H}(X)$ be as defined in Proposition 3.3. Since $P_{\langle \mu_k \rangle} = \langle \mu_k \rangle \circ \psi^{-1}$, it is enough to show that $E_{\langle \mu_k \rangle} [\mathcal{H}^\alpha(\psi(\mathcal{S}))] < \infty$ for $k = 1, \dots, N$. By Lemma 3.2 and Theorem 2.4 of Graf [7] it holds that

$$\begin{aligned} \mathcal{H}^\alpha(\psi(\mathcal{S})) &\leq |X|^\alpha \sup_{\Gamma_0 \in \text{Min}} \inf \left\{ \sum_{\sigma \in \Gamma} \prod_{n=1}^{\sigma} r(S_{\sigma|n})^\alpha | \Gamma \in \text{Min}, \Gamma > \Gamma_0 \right\} \\ &\leq |X|^\alpha \sup_{\Gamma_0 \in \text{Min}} \inf \left\{ f_{\Gamma, \alpha}^{(k)} / \min_{1 \leq i \leq N} x_i | \Gamma \in \text{Min}, \Gamma > \Gamma_0 \right\} \end{aligned}$$

for $\langle \mu_k \rangle$ -a.e. \mathcal{S} . By the last corollary, the expectation of this last expression with respect to $\langle \mu_k \rangle$ is finite. This completes the proof.

5. Proof of Theorem 3.11

For the proof of Theorem 3.11 we need a lemma, Lemma D, which is a

modification of Theorem 6.8 of Graf [7]. To show Lemma *D* we state necessary results. For $\mathcal{S} \in (\text{Con}(X)^N)^D$ and $\sigma \in D$, let $\mathcal{S}^\sigma \in (\text{Con}(X)^N)^D$ defined by $(\mathcal{S}^\sigma)_\tau = \mathcal{S}_{\sigma\tau}$ for $\tau \in D$.

LEMMA *B* (cf. Lemma 6.6. of Graf [7]). *Let (μ_1, \dots, μ_N) satisfy the conditions (3), (4) and (5). Let α be such that $\lambda(\alpha) = 1$. For $\beta < \alpha$, $\langle \mu_k \rangle$ -a.e. $\mathcal{S} \in \Omega$ and $k = 1, \dots, N$, there exists an $m \in \mathbb{N}$ such that, for every $\sigma \in D$ with $|\sigma| \geq m$,*

$$\prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha f^{(t(\sigma))}(\mathcal{S}^\sigma) \leq \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\beta.$$

PROOF. Let $\sigma \in D$ and $p \in \mathbb{N}$ be arbitrary. Using Chebyshev's inequality, we have

$$\begin{aligned} &\langle \mu_k \rangle(\{\mathcal{S} \mid \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^{\alpha-\beta} f^{(t(\sigma))}(\mathcal{S}^\sigma) > 1\}) \\ &\leq \int \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^{p(\alpha-\beta)} d\langle \mu_k \rangle(\mathcal{S}) \int \{f^{(t(\sigma))}(\mathcal{S})\}^p d\langle \mu_{t(\sigma)} \rangle(\mathcal{S}). \end{aligned}$$

Therefore

$$\begin{aligned} &\langle \mu_k \rangle(\{\mathcal{S} \mid \text{there exists a } \sigma \in C_q \text{ such that } \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^{\alpha-\beta} f^{(t(\sigma))}(\mathcal{S}^\sigma) > 1\}) \\ &\leq \int \sum_{\sigma \in C_q} \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^{p(\alpha-\beta)} d\langle \mu_k \rangle(\mathcal{S}) \max_{1 \leq i \leq N} \int \{f(\mathcal{S})^{(i)}\}^p d\langle \mu_i \rangle(\mathcal{S}). \end{aligned}$$

Let $p \in \mathbb{N}$ such that $p(\alpha - \beta) > \alpha$. Then we have $\lambda(p(\alpha - \beta)) < 1$. Let

$$c = \max_{1 \leq i \leq N} \sum_{j=1}^N R(p(\alpha - \beta))_{ij} \frac{x_j}{x_i}$$

where (x_1, \dots, x_N) is a positive eigenvector of $R(\alpha)$ corresponding to the maximal eigen value 1. By Frobenius' theorem we have that $c < 1$. Since $\int (f^{(i)})^p d\langle \mu_i \rangle < \infty$ for $i = 1, \dots, N$ by Theorem 3.10 and

$$\int \sum_{\sigma \in C_q} \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^{p(\alpha-\beta)} d\langle \mu_k \rangle(\mathcal{S}) \leq x_k c^q / (\min_{1 \leq i \leq N} x_i),$$

we deduce

$$\begin{aligned} &\sum_{q=1}^\infty \langle \mu_k \rangle(\{\mathcal{S} \mid \text{there exists a } \sigma \in C_q \text{ such that} \\ &\quad \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^{\alpha-\beta} f^{(t(\sigma))}(\mathcal{S}^\sigma) > 1\}) < \infty. \end{aligned}$$

By the Borel-Cantelli lemma we have

$$\begin{aligned} &\langle \mu_k \rangle(\bigcap_{m \in \mathbb{N}} \bigcup_{q \geq m} \{\mathcal{S} \mid \text{there exists a } \sigma \in C_q \text{ such that} \\ &\quad \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^{\alpha-\beta} f^{(t(\sigma))}(\mathcal{S}^\sigma) > 1\}) = 0. \end{aligned}$$

This completes the proof.

LEMMA C. (cf. Theorem 6.7 of Graf [7]). *Let (μ_1, \dots, μ_N) satisfy the conditions (3), (4) and (5). Let α be such that $\lambda(\alpha) = 1$. For $\beta < \alpha$, $\langle \mu_k \rangle$ -a.e. $\mathcal{S} \in \Omega$ and $k = 1, \dots, N$,*

$$\sup_{\Gamma \in \text{Min}} \inf \{ f_{\Gamma, \beta}^{(k)}(\mathcal{S}) \mid \Gamma \in \text{Min}, \Gamma > \Gamma_0 \} \geq f^{(k)}(\mathcal{S}).$$

PROOF. By Lemma B and Lemma 6.4 of Graf [7] we deduce the result. See the proof of Theorem 6.7 of Graf [7].

LEMMA D. *Let (μ_1, \dots, μ_N) satisfy the conditions (3), (4) and (5). Let $\beta < \alpha$ where $\lambda(\alpha) = 1$. Then, for $\langle \mu_k \rangle$ -a.e. \mathcal{S} and $k = 1, \dots, N$,*

$$\sup_{\Gamma \in \text{Min}} \inf \{ \sum_{\sigma \in \Gamma} r(\mathcal{S}_\sigma)^d \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\beta \mid \Gamma \in \text{Min}, \Gamma > \Gamma_0 \} > 0.$$

PROOF. Since $\lambda(\beta) > 1$, there exists an $\eta > 0$ such that, for $A_k = \{(S_1, \dots, S_N) \mid r(S_i) \geq \eta \text{ for } i = 1, \dots, N \text{ with } R(0)_{ki} > 0\}$ ($k = 1, \dots, N$), the maximal eigen value of a matrix $T = [t_{ki}]$ is greater than 1 where

$$t_{ki} = \int_{A_k} r(S_i)^\beta d\mu_k(S_1, \dots, S_N).$$

Define $r_\eta(S) : \text{Con}(X) \rightarrow [0, 1)$ by

$$r_\eta(S) = \begin{cases} 0, & r(S) < \eta \\ r(S), & r(S) \geq \eta. \end{cases}$$

Let $f_\eta^{(k)}(\mathcal{S}) = \lim_{m \rightarrow \infty} \sum_{\sigma \in C_m} \prod_{n=1}^{|\sigma|} r_\eta(S_{\sigma|n})^\alpha x_{r(\sigma)}$ defined on $(\Omega, \mathcal{B}, \langle \mu_k \rangle)$ for $k = 1, \dots, N$. For $\mathcal{S} \in \Omega$ we have

$$\begin{aligned} & \sup_{\Gamma_0} \inf_{\Gamma > \Gamma_0} \sum_{\sigma \in \Gamma} r(\mathcal{S}_\sigma)^d \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\beta \\ & \geq \sup_{\Gamma_0} \inf_{\Gamma > \Gamma_0} \sum_{\sigma \in \Gamma} r_\eta(\mathcal{S}_\sigma)^d \prod_{n=1}^{|\sigma|} r_\eta(S_{\sigma|n})^\beta \\ & \geq \eta^d \sup_{\Gamma_0} \inf_{\Gamma > \Gamma_0} \sum_{\sigma \in \Gamma} \prod_{n=1}^{|\sigma|} r_\eta(S_{\sigma|n})^\beta \\ & \geq \eta^d f_\eta^{(k)}(\mathcal{S}) / \max_{1 \leq i \leq N} x_i \quad \text{for } \langle \mu_k \rangle\text{-a.e. } \mathcal{S}. \end{aligned}$$

The last inequality follows from Lemma C. Since by Theorem 3.10 $\int f_\eta^{(k)}(\mathcal{S}) d\langle \mu_k \rangle > 0$, we deduce that

$$(a3) \quad \sup_{\Gamma_0} \inf_{\Gamma > \Gamma_0} \sum_{\sigma \in \Gamma} r(\mathcal{S}_\sigma)^d \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\beta > 0$$

with positive probability.

We show that the left-hand side in (a3) is either 0 with probability 1 or > 0 with probability 1. By Proposition 3.1 we have

$$\begin{aligned}
 p_k &:= \langle \mu_k \rangle (\{ \mathcal{S} \mid \sup_{\Gamma_0} \inf_{\Gamma > \Gamma_0} \sum_{\sigma \in \Gamma} r(S_\sigma)^d \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\beta = 0 \}) \\
 &= \mu_k \times \prod_{i=1}^N \langle \mu_i \rangle (\{ (S_1, \dots, S_N), (\mathcal{S}^{(1)}, \dots, \mathcal{S}^{(N)}) \mid \\
 &\quad \sum_{i=1}^N r(S_i)^\beta \sup_{\Gamma_i} \inf_{\Gamma > \Gamma_i} \sum_{\sigma \in \Gamma} r(S_\sigma^{(i)})^d \prod_{n=1}^{|\sigma|} r(S_{\sigma|n}^{(i)})^\beta = 0 \}) \\
 &= \prod_{i: R(0)_{ki} \neq 0} \langle \mu_i \rangle (\{ \mathcal{S} \mid \sup_{\Gamma_i} \inf_{\Gamma > \Gamma_i} \sum_{\sigma \in \Gamma} r(S_\sigma)^d \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\beta = 0 \}),
 \end{aligned}$$

because $r(S_j) > 0$ for $j \in \{1, \dots, N\}$ such that $R(0)_{kj} > 0$. By Proposition 3.8 and (a3) we deduce that

$$p_i = 0 \quad \text{for } i = 1, \dots, N.$$

This completes the proof.

PROOF OF THEOREM 3.11. By Proposition 3.9 we have

$$\dim_H(K) \leq \alpha$$

for $P_{\langle \mu_k \rangle}$ a.e. $K \in \mathcal{K}(X)$ and $k = 1, \dots, N$. The converse inequality is shown in the same way as in the proof of Theorem 7.6 of Graf [7] using Theorem 2.5 of Graf [7] and lemma D.

6. Proof of Theorem 4.1

Our fundamental lemma is as follows:

LEMMA E (cf. Lemma 6.10 of Graf [7]). *Let (μ_1, \dots, μ_N) be an N -tuple of probability measures on $\text{Con}(X)^N$ which satisfies the conditions (3), (4) and (5) in Section 3. Let $\alpha > 0$ be such that $\lambda(\alpha) = 1$. For $n \in \mathbb{N}$ define $h_n: \Omega \rightarrow \mathbf{R}_+$ by*

$$h_n(\mathcal{S}) = \inf \{ f_{\Gamma, \alpha}(\mathcal{S}) \mid \Gamma \in \text{Min}, \Gamma \neq \{\emptyset\}, |\Gamma| \leq n \}$$

where

$$f_{\Gamma, \alpha}(\mathcal{S}) = \sum_{\sigma \in \Gamma} \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha x_{t(\sigma)}$$

and $|\Gamma| = \max \{ |\sigma| : \sigma \in \Gamma \}$. (Note that for all $k = 1, \dots, N$, $f_{\Gamma, \alpha}(\mathcal{S}) = f_{\Gamma, \alpha}^{(k)}(\mathcal{S})$ for $\Gamma \neq \emptyset$.) Then $(h_n)_{n \in \mathbb{N}}$ are non-increasing sequences of Borel measurable functions which satisfy the following properties:

(i) $h_{n+1}(\mathcal{S}) = \sum_{i=1}^N r(S_i)^\alpha \min(x_i, h_n(\mathcal{S}^{(i)}))$ for all $n \in \mathbb{N}$ and $\mathcal{S} \in \Omega$.

(ii) $h := \inf_{n \in \mathbb{N}} h_n = \inf_{\Gamma \in \text{Min} \setminus \{\emptyset\}} f_{\Gamma, \alpha}$.

(iii) If the condition $\langle \mu_j \rangle (\{h > 0\}) > 0$ for some $j \in \{1, \dots, N\}$ holds, then

$$\sum_{i=1}^N r(S_i)^\alpha x_i = x_k \text{ for } \mu_k\text{-a.e. } (S_1, \dots, S_N) \text{ and all } k = 1, \dots, N.$$

PROOF. We only show (iii) since (i) and (ii) is trivial. It follows from (i) and (ii) that

$$(a4) \quad h(\mathcal{S}) = \sum_{i=1}^N r(S_i)^\alpha \min(x_i, h(\mathcal{S}^{(i)}))$$

for all $\mathcal{S} \in \Omega$. Let (y_1, \dots, y_N) be a positive vector such that $(y_1, \dots, y_N) = (y_1, \dots, y_N)R(\alpha)$ and $\sum_{k=1}^N y_k = 1$. Integrating the both sides of (a4) with respect to $\sum_{k=1}^N y_k \langle \mu_k \rangle$, we have by Proposition 3.1 that

$$\begin{aligned} \sum_{k=1}^N y_k \int h(\mathcal{S}) d\langle \mu_k \rangle &= \sum_{k=1}^N y_k \iint \sum_{i=1}^N r(S_i)^\alpha \min(x_i, h(\mathcal{S}^{(i)})) d\langle \mu_i \rangle (\mathcal{S}^{(i)}) \\ &\quad d\mu_k(S_1, \dots, S_N) \\ &= \sum_{i=1}^N \int \min(x_i, h(\mathcal{S})) d\langle \mu_i \rangle \sum_{k=1}^N y_k \int r(S_i)^\alpha \\ &\quad d\mu_k(S_1, \dots, S_N) \\ &= \sum_{i=1}^N \int \min(x_i, h(\mathcal{S})) y_i d\langle \mu_i \rangle. \end{aligned}$$

Since $y_k > 0$, we deduce that

$$h(\mathcal{S}) \leq x_k \quad \text{for } \langle \mu_k \rangle\text{-a.e. } \mathcal{S} \text{ and } k = 1, \dots, N.$$

Therefore (a4) implies that

$$(a5) \quad h(\mathcal{S}) = \sum_{i=1}^N r(S_i)^\alpha h(\mathcal{S}^{(i)})$$

for $\langle \mu_k \rangle$ -a.e. \mathcal{S} and $k = 1, \dots, N$. Let η_k be the essential supremum of $h(\mathcal{S})$ with respect to $\langle \mu_k \rangle$ for $k = 1, \dots, N$. Using (a5) and Proposition 3.1 we obtain that

$$\eta_k \geq \sum_{i=1}^N r(S_i)^\alpha \eta_i$$

for $\langle \mu_k \rangle$ -a.e. \mathcal{S} and $k = 1, \dots, N$. Integrating the both sides with respect to $\langle \mu_k \rangle$, we have

$$\eta_k \geq \sum_{i=1}^N R(\alpha)_{ki} \eta_i \quad \text{for } k = 1, \dots, N$$

where (η_1, \dots, η_N) is non-negative nonzero vector by our assumption (iii). By Theorem 2.4 (Frobenius),

$$\eta_k = \sum_{i=1}^N R(\alpha)_{ki} \eta_i \quad \text{for } k = 1, \dots, N$$

and (η_1, \dots, η_N) is positive eigenvector of $R(\alpha)$ corresponding to the maximal eigen value. This implies that

$$\eta_k = \sum_{i=1}^N r(S_i)^\alpha \eta_i \quad \text{for } \langle \mu_k \rangle\text{-a.e. } \mathcal{S} \text{ and } k = 1, \dots, N.$$

Since $\eta_1 : \dots : \eta_N = x_1 : \dots : x_N$, we have

$$x_k = \sum_{i=1}^N r(S_i)^\alpha x_i \quad \text{for } \mu_k\text{-a.e. } (S_1, \dots, S_N) \text{ and } k = 1, \dots, N.$$

This completes the proof.

Using Lemma E and the similar arguments to the proof of Theorem 6.11 of Graf [7], we have the following proposition.

PROPOSITION F. *Assume the condition of Lemma E are satisfied. Let $\alpha > 0$ be such that $\lambda(\alpha) = 1$ and (x_1, \dots, x_N) be a positive eigenvector of $R(\alpha)$ corresponding to the maximal eigen value 1. Then the following conditions are equivalent:*

- a) For all $k \in \{1, \dots, N\}$, $\sum_{i=1}^N r(S_i)^\alpha x_i = x_k$ for μ_k -a.e. (S_1, \dots, S_N) .
- b) For all $k \in \{1, \dots, N\}$, $\sup_{\Gamma_0 \in \text{Min}} \inf \{f_{\Gamma, \alpha}^{(k)}(\mathcal{S}) \mid \Gamma \in \text{Min}, \Gamma > \Gamma_0\} > 0$ for $\langle \mu_k \rangle$ -a.e. \mathcal{S} .
- c) $\langle \mu_j : \mu_1, \dots, \mu_N \rangle (\{\mathcal{S} \mid \sup_{\Gamma_0 \in \text{Min}} \inf \{f_{\Gamma, \alpha}^{(j)}(\mathcal{S}) \mid \Gamma \in \text{Min}, \Gamma > \Gamma_0\} > 0\}) > 0$ for some $j \in \{1, \dots, N\}$.

PROOF. (a) \rightarrow (b): Under the assumption (a), it holds that $f_{\Gamma, \alpha}^{(k)}(\mathcal{S}) = x_k$ for $\langle \mu_k \rangle$ -a.e. \mathcal{S} . This means

$$\sup_{\Gamma_0} \inf_{\Gamma > \Gamma_0} f_{\Gamma, \alpha}^{(k)}(\mathcal{S}) = x_k > 0 \quad \langle \mu_k \rangle\text{-a.e. } \mathcal{S}.$$

(b) \rightarrow (c) is trivial.

(c) \rightarrow (a): Fix $\Gamma_0 \in \text{Min}$ for $\Gamma \in \text{Min}$ with $\Gamma > \Gamma_0$ and $\sigma \in \Gamma_0$, let $\Gamma_\sigma = \{\tau \in D \mid \sigma * \tau \in \Gamma\}$, then $\Gamma_\sigma \in \text{Min}$. It holds that

$$\begin{aligned} \text{(a6)} \quad \inf_{\Gamma > \Gamma_0} f_{\Gamma, \alpha}^{(j)}(\mathcal{S}) &= \inf_{\Gamma > \Gamma_0} \sum_{\sigma \in \Gamma_0} \left[\prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha \sum_{\tau \in \Gamma_\sigma} \prod_{m=1}^{|\tau|} r(S_{\sigma*(\tau|m)})^\alpha x_{\tau(\sigma*\tau)} \right] \\ &= \sum_{\sigma \in \Gamma_0} \left[\prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha \inf_{\Gamma > \Gamma_\sigma} \sum_{\tau \in \Gamma_\sigma} \prod_{m=1}^{|\tau|} r(S_{\sigma*(\tau|m)})^\alpha x_{\tau(\sigma*\tau)} \right] \\ &= \sum_{\sigma \in \Gamma_0} \left[\prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha \min(x_{\tau(\sigma)}, \inf_{\Gamma \in \text{Min} \setminus \{\emptyset\}} f_{\Gamma, \alpha}(\mathcal{S}^\sigma)) \right] \\ &= \sum_{\sigma \in \Gamma_0} \left[\prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha \min(x_{\tau(\sigma)}, h(\mathcal{S}^\sigma)) \right]. \end{aligned}$$

By (c), there exists a Borel set $B \subset \Omega$ with $\langle \mu_j \rangle(B) > 0$ such that, for any $\mathcal{S} \in B$, there is a Γ_0 with $\inf_{\Gamma > \Gamma_0} f_{\Gamma, \alpha}^{(j)}(\mathcal{S}) > 0$. By (a6), it holds that for any $\mathcal{S} \in B$, there exist $\Gamma_0 \in \text{Min}$ and a $\sigma \in \Gamma_0$ such that $\prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha > 0$ and $h(\mathcal{S}^\sigma) > 0$. For $\sigma \in D$, let $\Omega(\sigma) = \{\mathcal{S} \mid \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha > 0 \text{ and } h(\mathcal{S}^\sigma) > 0\}$. Note that $\langle \mu_j \rangle(\cup_{\sigma \in D} \Omega(\sigma)) > 0$, because $B \subset \cup_{\sigma \in D} \Omega(\sigma)$. Hence there exists a $\sigma \in D$ such that $\langle \mu_j \rangle(\Omega(\sigma)) > 0$. Since $\langle \mu_j \rangle(\Omega(\sigma)) = \langle \mu_j \rangle(\{\prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha > 0\}) \langle \mu_{\tau(\sigma)} \rangle(\{\mathcal{S} \mid h(\mathcal{S}^\sigma) > 0\}) > 0$, it holds that $\langle \mu_{\tau(\sigma)} \rangle(\{\mathcal{S} \mid h(\mathcal{S}^\sigma) > 0\}) > 0$. Therefore Lemma E implies the condition (a).

PROOF OF THEOREM 4.1. (a) \rightarrow (b). By Theorem 2.5 of Graf [7] and Lemma 3.2, there exists a $c > 0$ such that

$$c |X|^\alpha \sup_{r_0} \inf_{r > r_0} \sum_{\sigma \in \Gamma} r(S_\sigma)^d \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha \leq \mathcal{H}^\alpha(\psi(\mathcal{S}))$$

for $\langle \mu_k \rangle$ -a.e. \mathcal{S} . Using the assumptions of Theorem 4.1 we have

$$c \delta^\alpha |X|^\alpha \sup_{r_0} \inf_{r > r_0} \sum_{\sigma \in \Gamma} \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha \leq \mathcal{H}^\alpha(\psi(\mathcal{S}))$$

for $\langle \mu_k \rangle$ -a.e. \mathcal{S} . Proposition F yields $\mathcal{H}^\alpha(\psi(S)) > 0$ for $\langle \mu_k \rangle$ -a.e. \mathcal{S} and by the definition of $P_{\langle \mu_k \rangle}$, we have (b).

(b) \rightarrow (c) is trivial.

(c) \rightarrow (a). By Theorem 2.4 of Graf [7] and Lemma 3.2 in Section 3 it follows that

$$\mathcal{H}^\alpha(\psi(\mathcal{S})) \leq |X|^\alpha \sup_{r_0} \inf_{r > r_0} \sum_{\sigma \in \Gamma} \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha$$

for $\langle \mu_k \rangle$ -a.e. \mathcal{S} and $k = 1, \dots, N$. Assume that $\mu_j((S_1, \dots, S_N) | \sum_{i=1}^N r(S_i)^\alpha x_i \neq x_j) > 0$ for some $j \in \{1, \dots, N\}$. Then Proposition F implies $\sup_{r_0} \inf_{r > r_0} f_{r, \alpha}^{(k)}(\mathcal{S}) := \sup_{r_0} \inf_{r > r_0} \sum_{\sigma \in \Gamma} \prod_{n=1}^{|\sigma|} r(S_{\sigma|n})^\alpha x_{i(\sigma)} = 0$ for $\langle \mu_k \rangle$ -a.e. \mathcal{S} and $k = 1, \dots, N$. It follows that $\mathcal{H}^\alpha(\psi(\mathcal{S})) = 0$ for $\langle \mu_k \rangle$ -a.e. \mathcal{S} and $k = 1, \dots, N$. By the definition of $P_{\langle \mu_k \rangle}$ we have $\mathcal{H}^\alpha(K) = 0$ for $P_{\langle \mu_k \rangle}$ -a.e. $K \in \mathcal{X}(X)$ and all $k = 1, \dots, N$. This completes the proof.

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