# I.I.D. STATISTICAL CONTRACTION OPERATORS AND STATISTICALLY SELF-SIMILAR SETS\*\*

#### HU DIHE\*

### Abstract

I.i.d. random sequence is the simplest but very basic one in stochastic processes, and statistically self-similar set is the simplest but very basic one in random recursive sets in the theory of random fractal. Is there any relation between i.i.d. random sequence and statistically self-similar set? This paper gives a basic theorem which tells us that the random recursive set generated by a collection of i.i.d. statistical contraction operators is always a statistically self-similar set.

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## §1. Introduction

Hutchinson<sup>[5]</sup> constructed a class of (strictly) self-similar sets by contraction operators and obtained their Hausdorff dimension and Hausdorff exact measure function. Later, Mauldin and Williams<sup>[6]</sup>, Graf<sup>[3]</sup> and Falconer<sup>[2]</sup> independently constructed the statistically self-similar measure and set in different way and investigated their probability properties and fractal properties. But no one has pointed out what is the probability character of statistically self-similar set.

In this paper, we will give a basic theorem which tells us that the random recursive set generated by a collection of i.i.d. statistical contraction operators is always a statistically self-similar set.

Let  $N \geq 2$  be an integer,  $C_0 = \{\emptyset\}$ ,  $C_n = \{0, 1, \dots, N-1\}^n$   $(n \geq 1)$ ,  $D = \bigcup_{n \geq 0} C_n$ ,  $\{f_{\sigma}, \sigma \in D\}$  be a collection of statistical contraction operators. If  $\sup_{\sigma \in D} \operatorname{Lip}(f_{\sigma}) = \alpha < 1$ ,  $\{(f_{(\sigma,0)}, \dots, f_{(\sigma,N-1)}), \sigma \in D\}$  are i.i.d., then

$$K \triangleq \bigcap_{n=1}^{\infty} \bigcup_{(\sigma_1, \dots, \sigma_n) \in C_n} \overline{f_{\sigma_1} \circ f_{(\sigma_1, \sigma_2)} \circ f_{(\sigma_1, \dots, \sigma_n)}(E)}$$

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<sup>\*</sup>College of Mathematics and statistics, Wuhan University, Wuhan 430072, China.

E-mail: dhhu@whu.edu.cn

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is always a statistically self-similar set, where  $\operatorname{Lip}(\cdot)$  is the Lipschitz coefficient,  $(E, \rho)$  is a polish space,  $\overline{A}$  is the closure of A.

### §2. Notations and Preliminaries

Let  $(\Omega, \mathcal{F}, P)$  be a complete probability space,  $(E, \rho)$  be a separable complete metric space.  $\mathcal{K}(E)$  denotes all non-empty compact sets in  $E, \eta$  is the Hausdorff metric on  $\mathcal{K}(E)$ , that is to say,  $I, J \in \mathcal{K}(E)$ ,

$$\eta(I,J) = \sup \{ \rho(x,I), \rho(y,J) : x \in J, y \in I \}, \rho(x,I) \text{ is the distance from } x \text{ to } I.$$

 $(\mathcal{K}(E), \eta)$  is also a separable complete metric space.

For any subset A in a metric space,  $\operatorname{diam}(A)$  denotes the diameter of A.  $\overline{A}$  donotes the closure of A. Let  $\mathbf{R}^d$  be the d-dimensional Euclidean space,  $\mathbf{R} = \mathbf{R}^1$ . We always denote the Borel field of topology space T by  $\mathcal{B}(T)$ , and all Borel probability measures by  $\mathcal{P}(T)$ . Let  $f: E \to E, B \subset E, f(B)$  is the image of f on B.

**Definition 2.1.** Let  $f: E \to E$ .

$$\operatorname{Lip}(f) \triangleq \sup_{x \neq y, x, y \in E} \frac{\rho(f(x), f(y))}{\rho(x, y)}$$

is called the Lipschitz coefficient of f. If  $\operatorname{Lip}(f) < 1$ , then we call f a contraction operator and denote the class of all contraction operators from E to E by con(E).

We always assume that con(E) carries the topology of pointwise convergence.

**Proposition 2.1.**<sup>[3]</sup> let  $f: E \to E$ .

- (1) If f is continuous, then  $f(\mathcal{K}(E)) \subset \mathcal{K}(E)$ ;
- (2)  $I, J \in \mathcal{K}(E) \Rightarrow \eta(f(I), f(J)) \leq \operatorname{Lip}(f)\eta(I, J)$ ;
- (3)  $\operatorname{Lip}(f) : \operatorname{con}(E) \to [0,1)$  is lower-semicontinuous;
- (4)  $f(J) : con(E) \times \mathcal{K}(E) \to \mathcal{K}(E)$  is continuous;

- (5)  $g(J_1, \dots, J_m) = \bigcup_{i=1}^m J_i : \mathcal{K}(E)^m \to \mathcal{K}(E) \text{ is continuous};$ (6)  $h(f_1, \dots, f_m) = f_1 \circ \dots \circ f_m : \operatorname{con}(E)^m \to \operatorname{con}(E) \text{ is continuous};$ (7)  $I_i, J_i, \bigcup_i I_i, \bigcup_i J_i \in \mathcal{K}(E) \Rightarrow \eta\left(\bigcup_i I_i, \bigcup_i J_i\right) \leq \sup_i, \eta(I_i, J_i).$

**Definition 2.2.** Let  $f^{(\cdot)}: \Omega \to \operatorname{con}(E)$ . We call f a statistical contraction operators, iff f is a random element from  $\Omega$  to con(E), i.e.

$$\{\omega \in \Omega, f^{(\omega)} \in A\} \in \mathcal{F} \ (\forall A \in \mathcal{B}(\operatorname{con}(E))).$$

We denote all statistical contraction operators by  $con(\Omega, E)$ .

We write  $f^{(\omega)} = f(\omega), f^{(\omega)}(x) = f(\omega, x)$  sometimes.

The proofs of the following propositions are straightforward and we leave them to the

**Proposition 2.2.** Let  $(\overline{\Omega}, \overline{\mathcal{F}})$  be a measurable space,  $(\overline{\mathcal{K}}, \overline{\eta})$  be a separable complete metric space.  $\overline{f}(\omega, J) : \overline{\Omega} \times \overline{\mathcal{K}} \to \overline{\mathcal{K}}$  satisfies:  $\forall \omega \in \overline{\Omega}, \overline{f}(\omega, \cdot)$  is uniformly continuous;  $\forall J \in \overline{\mathcal{K}}, \overline{f}(\cdot, J)$ is Borel measurable. Then  $\bar{f}(\cdot,\cdot)$  is Borel measurable.

**Proposition 2.3.** If  $f \in con(\Omega, E)$ , then  $f(\omega, J) : \Omega \times \mathcal{K}(E) \to \mathcal{K}(E)$  is Borel measurable; especially  $f(\cdot, J)$  is Borel measurable for any fixed  $J \in \mathcal{K}(E)$ .

**Proposition 2.4.**  $\{f_1, \dots, f_m\} \subset \operatorname{con}(\Omega, E) \Rightarrow f_1 \circ \dots \circ f_m \in \operatorname{con}(\Omega, E)$ . **Proposition 2.5.**  $\{f_1, \dots, f_m\} \subset \operatorname{con}(\Omega, E) \Rightarrow f_1 \circ \dots \circ f_2 \circ \dots \circ f_m(J)$  is a random element from  $\Omega$  to  $\mathcal{K}(E)$  for any fixed  $J \in \mathcal{K}(E)$ .

We always assume diam $(E) < \infty$  in this paper.

Let us introduce some notations now.  $1_A$  always denotes the indicator function on set  $A, M(\Omega, \mathcal{K}(E))$  denotes the collection of all random elements from  $\Omega$  to  $\mathcal{K}(E)$ .

Let  $N \geq 2$  be an integer,  $C_0 = \{\emptyset\}, C_n = C_n(N) = \{0, 1, \dots, N-1\}^n, (n \geq 1), D = D(N) = \bigcup_{n \geq 0} C_n, C = \{(\sigma_0, \sigma_1, \dots) : 0 \geq \sigma_i < N\}.$ 

 $\forall \sigma = (\sigma_0, \dots, \sigma_{n-1}) \in C_n(N), \tau \in C(N) \bigcup D(N), \tau = (\tau_0, \tau_1, \dots), |\sigma| = n = \text{the length}$ of  $\sigma, \sigma * \tau = (\sigma_0, \dots, \sigma_{n-1}, \tau_0, \tau_1, \dots)$  is the juxtaposition of  $\sigma$  and  $\tau, \tau | k = (\tau_0, \tau_1, \dots, \tau_{k-1})$ (if  $|\tau| \geq k$ ),  $j|\tau = (\tau_j, \tau_{j+1}, \cdots)$  (if  $|\tau| \geq j$ ).  $\{0, 1, \cdots, N-1\}$  carries the discrete topology, and C carries the product topology, so C is compact.

**Definition 2.3.** Let  $\{f_0, \dots, f_{N-1}\} \subset \operatorname{con}(\Omega, E), Q \in \mathcal{P}(\mathcal{K}(E))$ . We call Q a P –  $(f_0, \dots, f_{N-1})$  statistically self-similar measure, iff  $\forall B \in \mathcal{B}(\mathcal{K}(E))$ , we have

$$Q(B) = P \times Q, N\left(\left\{(\omega; K_0, \cdots, K_{N-1}) \in \Omega \times \mathcal{K}(E)^N : \bigcup_{i=0}^{N-1} f_i(\omega, K_i) \in B\right\}\right).$$

**Definition 2.4.** Let  $K: \Omega \to \mathcal{K}(E)$  be a random set (i.e.  $K^{-1}(\mathcal{B}(\mathcal{K}(E))) \subset \mathcal{F}$ ),  $\{f_0, \dots, f_n\}$  $f_{N-1}) \subset \operatorname{con}(\Omega, E)$ . We call K a  $P - (f_0, \dots, f_{N-1})$  statistically self-similar set, iff  $P \circ K^{-1}$ , the distribution of K, is a  $P - (f_0, \dots, f_{N-1})$  statistically self-similar measure. **Proposition 2.6.** Let  $f_0, \dots, f_{N-1} \subset \operatorname{con}(\Omega, E), K : \Omega \to \mathcal{K}(E)$  be a random set. Then

K is a  $P - (f_0, \dots, f_{N-1})$  statistically self-similar set iff

$$K(\omega) \stackrel{d}{=} \bigcup_{i=0}^{N-1} f_i(\omega, K(\omega_i)).$$

where  $\stackrel{d}{=}$  means equal in distribution.

**Proof.** It is easy to prove by the definition.

#### §3. Main Result

**Theorem 3.1.** Let  $\{f_{\sigma}, \sigma \in D\} \subset \operatorname{con}(\Omega, E), f_{n,\sigma} = f_{\sigma|1} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n} \circ \cdots \circ f_{\sigma|n} (\sigma \in C_n, n \geq 1), \Psi_n = f_{\sigma|n$  $\bigcup_{\sigma \in C_n} f_{n,\sigma}(E), \ K = \bigcap_{n=1}^{\infty} \overline{\Psi}_n.$ (1) If  $\sup_{\sigma \in D} \operatorname{Lip}(f_{\sigma}) = \alpha < 1$  a.s., then, for almost all  $\omega \in \Omega$ ,

$$\lim_{n \to \infty} \eta \Big( K, \bigcup_{\sigma \in C_n} f_{n,\sigma}(J_{\sigma}) \Big) = 0 \ (\forall J_{\sigma} \in M(\Omega, \mathcal{K}(E))); K \in M(\Omega, \mathcal{K}(E)).$$
 (3.1)

(2) If the condition in (1) is satisfied, and  $\{(f_{\sigma*0}, \cdots, f_{\sigma*(N-1)}), \sigma \in D\}$  are i.i.d. random elements from  $\Omega$  to  $\operatorname{con}(E)^N$ , then  $P-(f_0,\cdots,f_{N-1})$  statistically self-similar measure is unique, it is  $P_K \triangleq P \circ K^{-1}$ , so K is a statistically self-similar set.

In order to prove this theorem, we need some lemmas.

We write  $Q^{\#B} = Q^B$  for a measure Q and a set B sometimes. If Q is a set,  $Q^B$  has the same meaning as above.

Let

$$t_{n}(\omega; J_{\sigma}, \sigma \in C_{n}) = \Psi_{n}^{(\omega)}(J_{\sigma}, \sigma \in C_{n})$$

$$= \bigcup_{\sigma \in C_{n}} f_{n,\sigma}^{(\omega)}(J_{\sigma}), (\omega \in \Omega, J_{\sigma} \in \mathcal{K}(E), n \geq 1),$$

$$T_{P}: \mathcal{P}(\mathcal{K}(E)) \to \mathcal{P}(\mathcal{K}(E)),$$
(3.2)

$$T_P: \mathcal{P}(\mathcal{K}(E)) \to \mathcal{P}(\mathcal{K}(E)),$$
  
 $T_P(Q) = (P \times Q^N) \circ t_1^{-1}, \ Q \in \mathcal{P}(\mathcal{K}(E)),$  (3.3)

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 $T_P^{(n)}$  is the *n*-compound mapping of  $T_p$ .

**Lemma 3.1.** Let  $\{f_{\sigma}, \sigma \in D\} \subset \operatorname{con}(\Omega, E)$ ,  $f_{n,\sigma}$ ,  $\Psi_n, t_n, T_P$  and  $T_p^{(n)}$  be defined as before.  $\forall \{Q, Q_k, k \geq 1\} \subset \mathcal{P}(\mathcal{K}(E))$ , if  $Q_k \stackrel{W}{\to} Q$  as  $k \to \infty$  (where  $\stackrel{W}{\to}$  means weak convergence), then  $T_P(Q_k) \stackrel{W}{\to} T_p(Q)$ .

**Proof.**  $\forall g : \mathcal{K}(E) \to R$ , if g is continuous and bounded, then

$$\begin{split} \int_{\mathcal{K}(E)} g(J) T_p(Q_k)(dJ) &= \int_{\mathcal{K}(E)} g(J) [(P \times Q_k^N) \circ t_1^{-1}](dJ) \\ &= \int_{\Omega \times \mathcal{K}(E)^N} g(t_1(\omega; J_i, i \in C_1)) (P \times Q_k^N)(d\omega; dJ_i, i \in C_1) \\ &= \int_{\Omega} P(d\omega) \Big( \int_{\mathcal{K}(E)^N} g(t_1(\omega; J_i, i \in C_1)) Q_k^N(dJ_i, i \in C_1) \Big). \end{split}$$

Since  $Q_k \stackrel{W}{\to} Q$ , we have  $Q_k^N \stackrel{W}{\to} Q^N$  by Theorem 3.1 in Chapter 1 in [1]. But for any fixed  $\omega \in \Omega$ ,  $g(t_1(\omega; \cdot, \dots, \cdot)) : \mathcal{K}(E)^N \to R$  is continuous and bounded by Proposition 2.1, hence

$$\lim_{k \to \infty} \int_{\mathcal{K}(E)^N} g(t_1(\omega; J_i, i \in C_1)) Q_k^N(dJ_i, i \in C_1)$$
$$= \int_{\mathcal{K}(E)^N} g(t_1(\omega; J_i, i \in C_1)) Q^N(dJ_i, i \in C_1).$$

It follows from the bounded convergence theorem that

$$\lim_{k \to \infty} \int_{\mathcal{K}(E)^N} g(J) T_p(Q_k)(dJ) = \int_{\Omega} P(d\omega) \int_{\mathcal{K}(E)^N} g(t_1(\omega; J_i, i \in C_1)) Q^N(dJ_i, i \in C_1)$$
$$= \int_{\mathcal{K}(E)} g(J) T_p(Q)(dJ).$$

Lemma 3.1 is proved.

**Lemma 3.2.** Let  $\{f_{\sigma}, \sigma \in D\}$ ,  $f_{n,\sigma}, \Psi_n, t_n, T_P$  and  $T_p^{(n)}$  be difined as in Lemma 3.1. If

$$P^{2}\left(\left\{(\omega, \bar{\omega}): \bigcup_{i \in C_{1}} f_{i}^{(\omega)}(t_{n}(\bar{\omega}; J_{i*\sigma}, \sigma \in C_{n})) \in A\right\}\right)$$

$$= P(\left\{\omega: t_{n+1}(\omega; J_{\tau}, \tau \in C_{n+1}) \in A\right\})$$

$$(\forall A \in \mathcal{B}(\mathcal{K}(E)), J_{\tau} \in \mathcal{K}(E), \tau \in C_{n+1}, \ n \geq 1), \tag{3.4}$$

then

$$T_p^{(n)}(Q) = (P \times Q^{C_n}) \circ t_n^{-1} \ (n \ge 1, \ Q \in \mathcal{P}(\mathcal{K}(E))).$$
 (3.5)

**Proof.** We prove (3.5) by induction. When n = 1, (3.5) is true by the definiton of  $T_P$ . If (3.5) is true for n,  $\forall A \in \mathcal{B}(\mathcal{K}(E))$ , let

$$F(A) = \left\{ (\omega; J_i, i \in C_1) : \bigcup_{i \in C_1} f_i^{(\omega)}(J_i) \in A \right\} = t_1^{-1}(A),$$

$$M(A) = \left\{ (\omega, \bar{\omega}; J_\tau, \tau \in C_{n+1}) : \bigcup_{i \in C_1} f_i^{(\omega)}(t_n(\bar{\omega}; J_{i*\sigma}, \sigma \in C_n)) \in A \right\},$$

then

$$M(A) = \{(\omega, \bar{\omega}; J_{\tau}, \tau \in C_{n+1}) : (\omega; t_n(\bar{\omega}; J_{i*\sigma}, \sigma \in C_n), i \in C_1) \in F(A)\},\$$

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hence

$$1_{F(A)}(\omega; t_n(\bar{\omega}; J_{i*\sigma}, \sigma \in C_n), i \in C_1) = 1_{M(A)}(\omega, \bar{\omega}; J_{\tau}, \tau \in C_{n+1}). \tag{3.6}$$

It follows from the definitions of  $T_p^{(n)}$  and  $t_n$  and (3.6) that

$$T_{p}^{(n+1)}(Q)(A)$$

$$= T_{p}((P \times Q^{C_{n}}) \circ t_{n}^{-1})(A)$$

$$= (P \times [(P \times Q^{C_{n}}) \circ t_{n}^{-1}]^{N})(t_{1}^{-1}(A))$$

$$= \int_{\Omega} P(d\omega) \prod_{i \in C_{1}} \int_{\mathcal{K}(E)} [(P \times Q^{C_{n}}) \circ t_{n}^{-1}](dJ_{i}) \cdot 1_{F(A)}(\omega; J_{i}, i \in C_{1})$$

$$= \int_{\Omega} P(d\omega) \prod_{i \in C_{1}} \int_{\Omega \times \mathcal{K}(E)^{C_{n}}} (P \times Q^{C_{n}})(d\bar{\omega}, dJ_{i*\sigma}, \sigma \in C_{n})$$

$$\cdot 1_{F(A)}(\omega; t_{n}(\bar{\omega}; J_{i*\sigma}, \sigma \in C_{n}), i \in C_{1})$$

$$\stackrel{(3.6)}{=} \prod_{i \in C_{1}} \int_{\Omega} P(d\omega) \int_{\Omega \times \mathcal{K}(E)^{C_{n}}} (P \times Q^{C_{n}})(d\bar{\omega}, dJ_{i*\sigma}, \sigma \in C_{n}) \cdot 1_{M(A)}(\omega, \bar{\omega}; J_{\tau}, \tau \in C_{n+1})$$

$$= \prod_{\tau \in C_{n+1}} \int_{\mathcal{K}(E)} Q(dJ_{\tau}) \int_{\Omega^{2}} P^{2}(d\omega, d\bar{\omega}) \cdot 1_{M(A)}(\omega, \bar{\omega}; J_{\tau}, \tau \in C_{n+1}). \tag{3.7}$$

But, by the definition of M(A) and (3.4),

$$\int_{\Omega^2} P^2(d\omega, d\bar{\omega}) \cdot 1_{M(A)}(\omega, \bar{\omega}; J_{\tau}, \tau \in C_{n+1})$$

$$= P^2\Big(\Big\{(\omega, \bar{\omega}) : \bigcup_{i \in C_1} f_i^{(\omega)}(t_n(\bar{\omega}; J_{i*\sigma}, \sigma \in C_n)) \in A\Big\}\Big)$$

$$= P(\{\omega : t_{n+1}(\omega, J_{\tau}, \tau \in C_{n+1}) \in A\}).$$

It follows from the above equation and (3.7) that

$$T_P^{(n+1)}(Q)(A) = (P \times Q^{C_{n+1}}) \cdot t_{n+1}^{-1}(A).$$

Lemma 3.2 is proved.

**Lemma 3.3.** If the conditions in Lemma 3.2 are satisfied, and

$$\sup_{\sigma \in D} \operatorname{Lip}(f_{\sigma}) = \alpha < 1 \quad \text{a.s.},$$

- (1)  $T_P^{(n)}(Q) \xrightarrow{W} P_k \triangleq P \circ K^{-1} \text{ as } n \to \infty \ (\forall Q \in \mathcal{P}(\mathcal{K}(E))), K \text{ is defined as in Theorem}$
- (2)  $P (f_0, \dots, f_{N-1})$  statistically self-similar measure is unique, it is  $P_k$ , so K is a statistically self-similar set.

**Proof.** (1) It is enough to prove (1) that for any open set  $G \subset \mathcal{K}(E)$  and closed set  $F \subset \mathcal{K}(E)$  we have

$$\liminf_{n \to \infty} T_P^{(n)}(Q)(G) \ge P_K(G), \tag{3.8}$$

$$\lim_{n \to \infty} T_P^{(n)}(Q)(F) \le P_K(F). \tag{3.9}$$

We only prove (3.9). By Fubini theorem and Lemma 3.2 we have

$$T_P^{(n)}(Q)(F) = P \times Q^D\Big(\Big\{(\omega; J_\sigma, \sigma \in D) \in \Omega \times \mathcal{K}(E)^D : \bigcup_{\sigma \in C_n} f_{n,\sigma}^{(\omega)}(J_\sigma) \in F\Big\}\Big),$$

hence

$$\begin{split} & \limsup_{n \to \infty} T_P^{(n)}(Q)(F) \\ & \leq P \times Q^D \Big( \bigcap_{m=1}^{\infty} \bigcup_{n \geq m} \Big\{ (\omega; J_{\sigma}, \sigma \in D) \in \Omega \times \mathcal{K}(E)^D : \bigcup_{\sigma \in C_n} f_{n,\sigma}^{(\omega)}(J_{\sigma}) \in F \Big\} \Big) \\ & \leq P \times Q^D \Big( \Big\{ (\omega; J_{\sigma}, \sigma \in D) \in \Omega \times \mathcal{K}(E)^D : \lim_{n \to \infty} \bigcup_{\sigma \in C_n} f_{n,\sigma}^{(\omega)}(J_{\sigma}) \in F \Big\} \Big) \\ & = P \times Q^D \big( \{ (\omega; J_{\sigma}, \sigma \in D) \in \Omega \times \mathcal{K}(E)^D : \mathcal{K}(\omega) \in F \} \big) \\ & = P(\{ \omega \in \Omega : \mathcal{K}(\omega) \in F \}) = P_k(F). \end{split}$$

(3.9) is proved.

(2) Now we want to show that  $P - (f_0, \dots, f_{N-1})$  statistically self-similar measure is unique, it is  $P_K$ . It follows from Lemma 3.1 and (1) that

$$P_K = \lim_{n \to \infty} T_P^{(n+1)}(P_k) = \lim_{n \to \infty} T_P(T_P^{(n)}(P_K)) = T_P(\lim_{n \to \infty} T_P^{(n)}(P_k)) = T_P(P_K),$$

this means that  $P_k$  is a statistically self-similar measure.

Suppose Q is any  $P - (f_0, \dots, f_{N-1})$  statistically self-similar measure, then

$$Q = T_P(Q) = \dots = T_P^{(n)}(Q).$$

Let  $n \to \infty$ . By (1) in this lemma, we get

$$Q = \lim_{n \to \infty} T_P^{(n)}(Q) = P_K.$$

Lemma 3.3 is proved.

**Lemma 3.4.** Let  $\{f_{\sigma}, \sigma \in D\} \subset \operatorname{con}(\Omega, E), \{(f_{\sigma*0}, \cdots, f_{\sigma*(N-1)}), \sigma \in D\}$  be a class of i.i.d. random elements from  $\Omega$  to  $\operatorname{con}(E)^N$ . Let

$$\tilde{f}_{\sigma}^{(\omega,\bar{\omega})} = \begin{cases} f_{\sigma}^{(\omega)}, & when \ \sigma \in C_1, (\omega,\bar{\omega}) \in \Omega^2, \\ f_{\sigma}^{\bar{\omega}}, & when \ \sigma \in \bigcup_{k \geq 2} C_k, \ (\omega,\bar{\omega}) \in \Omega^2. \end{cases}$$

Then

(1)  $\{(\tilde{f}_{\sigma*0}, \cdots, \tilde{f}_{\sigma*(N-1)}, \sigma \in D\}$  is a class of i.i.d. random elements from  $\Omega^2$  to  $con(E)^N$ 

$$(\tilde{f}_0, \dots, \tilde{f}_{N-1}) \stackrel{d}{=} (f_0, \dots, f_{N-1}),$$

$$(\tilde{f}_{\sigma}, \sigma \in \bigcup_{k=2}^{n+1} C_k) \stackrel{d}{=} (f_{\sigma}, \sigma \in \bigcup_{k=2}^{n+1} C_k) \ (n \ge 1),$$

where  $\stackrel{d}{=}$  means equal in distribution.

(2) For any fixed  $\{J_i, J_{\sigma}, i \in C_1, \sigma \in \bigcup_{k=2}^{n+1} C_k\} \subset \mathcal{K}(E), \{(f_{\sigma*0}(J_0), \cdots, f_{\sigma*(N-1)}(J_{N-1})), \}$ 

 $\sigma \in D$ } is a class of i.i.d. random elements from  $\Omega$  to  $\mathcal{K}(E)^N$ , and  $\{(\tilde{f}_{\sigma*0}(J_0), \cdots, \tilde{f}_{\sigma*(N-1)}(J_{N-1})), \sigma \in D\}$  is a class of i.i.d. random elements from  $\Omega^2$  to  $\mathcal{K}(E)^N$ , and

$$(\tilde{f}_0(J_0), \dots, \tilde{f}_{N-1}(J_{N-1})) \stackrel{d}{=} (f_0(J_0), \dots, f_{N-1}(J_{N-1})),$$
 (3.10)

$$\left(\tilde{f}_{\sigma}, (J_{\sigma}), \sigma \in \bigcup_{k=2}^{n+1} C_k\right) \stackrel{d}{=} \left(f_{\sigma}(J_{\sigma}), \sigma \in \bigcup_{k=2}^{n+1} C_k\right). \tag{3.11}$$

**Proof.** (1) is obvious, we only need to prove (2). Let  $\mu$  be a probability measure on  $\mathcal{B}(\mathcal{K}(E)^N)$  such that  $\mu(A) = 1$  or 0 according to  $(J_0, \dots, J_{N-1}) \in A$  or not. Let

$$g_{\sigma}: \Omega \times \mathcal{K}(E)^{N} \to \operatorname{con}(E)^{N} \times \mathcal{K}(E)^{N}, \sigma \in D,$$

$$g_{\sigma}(\omega; \tilde{J}_{\sigma}, \cdots, \tilde{J}_{N-1}) = (f_{\sigma*0}^{(\sigma)}, \cdots, f_{\sigma*N-1}^{(\sigma)}; \tilde{J}_{0}, \cdots, \tilde{J}_{N-1}),$$

$$h: \operatorname{con}(E)^{N} \times \mathcal{K}(E)^{N} \to \mathcal{K}(E)^{N},$$

$$h(r_{0}, \cdots r_{N-1}; \tilde{J}_{0}, \cdots, \tilde{J}_{N-1}) = (r_{0}(\tilde{J}_{0}), \cdots, r_{N-1}(\tilde{J}_{N-1})).$$

Then, by Proposition 2.1,  $\{g_{\sigma}, \sigma \in D\}$  is a class of i.i.d. random elements from probability space  $(\Omega \times \mathcal{K}(E)^N, \mathcal{F} \times \mathcal{B}(\mathcal{K}(E)^N), P \times \mu)$  to  $con(E)^N \times \mathcal{K}(E)^N$ , h is a continuous operator. Hence

$$\{(f_{\sigma*0}^{(\omega)}(\tilde{J}_0), \cdots, f_{\sigma*(N-1)}^{(\omega)}(\tilde{J}_{N-1})) = h(g_{\sigma}(\omega; \tilde{J}_0, \cdots, \tilde{J}_{N-1})), \ \sigma \in D\}$$

is a class of i.i.d. random elements from  $(\Omega \times \mathcal{K}(E)^N, \mathcal{F} \times B(\mathcal{K}(E)^N), P \times \mu)$  to  $\mathcal{K}(E)^N$ ; especially, for fixed  $(J_0, \dots, J_{N-1})$ ,

$$\{(f_{\sigma*0}^{(\omega)}(J_0)\cdots,f_{\sigma*(N-1)}^{(\omega)}(J_{N-1}))=h(g_{\sigma}(\omega;J_0,\cdots,J_{N-1})),\sigma\in D\}$$

is a class of i.i.d. random elements from  $(\Omega, \mathcal{F}, P)$  to  $\mathcal{K}(E)^N$ .

We can prove  $\{(\tilde{f}_{\sigma*0}(J_0), \cdots, \tilde{f}_{\sigma*(N-1)}(J_{N-1})), \sigma \in D\}$  is a class of i.i.d. random elements from  $(\Omega^2, \mathcal{F}^2, P^2)$  to  $\mathcal{K}(E)^N$  in the same way. (3.10) and (3.11) are obvious. Lemma 3.4 is proved.

Now let us prove Theorem 3.1. (1) is a special case of Theorem 2.1 in [4]. In order to prove the conclusion (2) of Theorem 3.1, using Lemma 3.3, it is enough to prove (3.4) is true under the conditions in the Theorem 3.1. Let  $B(n+1) = \bigcup_{k=1}^{n+1} C_k$ ,

$$q: \operatorname{con}(E)^{B(n+1)} \times \mathcal{K}(E)^{B(n+1)} \to \mathcal{K}(E),$$

$$q(S_{\tau}; J_{\tau}, \tau \in B(n+1)) = \bigcup_{i \in C_1} \bigcup_{\sigma \in C_n} S_i \circ S_{i*(\sigma|1)} \circ \cdots \circ S_{i*(\sigma|n)}(J_{i*\sigma}).$$

Then q is a continuous operator by Proposition 2.1. Hence  $q(\tilde{f}_{\tau}^{(\omega,\tilde{\omega})}; J_{\tau}, \tau \in B(n+1))$  and  $q(f_{i}^{(\omega)}, f_{i*\sigma}^{(\tilde{\omega})}; J_{i}, J_{i*\sigma}, i \in C_{1}, \ \sigma \in B(n))$  are random elements from  $(\Omega^{2}, \mathcal{F}^{2}, P^{2})$  to  $\mathcal{K}(E)$  for fixed  $J_{\tau}, (\tau \in B(n+1))$ .

Since  $\{(f_{\sigma*0}^{(\omega)}, \cdots, f_{\sigma*(N-1)}^{(\omega)}), \sigma \in D\}$  are i.i.d., it follows that

$$\forall \left\{ A_i, B_\tau, i \in C_1, \tau \in \bigcup_{k=2}^{n+1} C_k \right\} \subset \mathcal{B}(\operatorname{con}(E))$$

we have

$$P^{2}(f_{i}^{(\omega)} \in A_{i}, f_{i*(\sigma|j)}^{(\bar{\omega})} \in B_{i*(\sigma|j)}, i \in C_{1}, \sigma \in C_{n}, 1 \leq j \leq n)$$

$$= P^{2}(f_{i}^{(\omega)} \in A_{i}, i \in C_{1}, f_{\sigma*j}^{(\bar{\omega})} \in B_{\sigma*j}, \sigma \in B(n), j \in C_{1})$$

$$= P^{2}(f_{i}^{(\omega)} \in A_{i}, i \in C_{1})P^{2}(f_{\sigma*j}^{(\bar{\omega})} \in B_{\sigma*j}, \sigma \in B(n), j \in C_{1})$$

$$= P^{2}(f_{i}^{(\omega)} \in A_{i}, i \in C_{1})P^{2}(f_{(|\tau|-1)|\tau}^{(\bar{\omega})} \in B_{\tau}, \tau \in \bigcup_{k=2}^{n+1} C_{k}).$$
(3.12)

Similarly, we also have

$$P^{2}(f_{i}^{(\omega)} \in A_{i}, f_{\sigma|j}^{(\bar{\omega})} \in B_{i*(\sigma|j)}, \sigma \in C_{n}, i \in C_{1}, 1 \leq j \leq N)$$

$$= P^{2}(f_{i}^{(\omega)} \in A_{i}, i \in C_{1})P^{2}\left(f_{(|\tau|-1)|\tau}^{(\bar{\omega})} \in B_{\tau}, \tau \in \bigcup_{k=2}^{n+1} C_{k}\right). \tag{3.13}$$

But, by the definition of  $\{\tilde{f}_{\sigma}, \sigma \in D\}$  and Lemma 3.4, we have

$$(f_0^{(\omega)}, \dots, f_{N-1}^{(\omega)}, f_{i*(\sigma|1)}^{(\bar{\omega})}, \dots, f_{i*\sigma}^{(\bar{\omega})}, i \in C_1, \sigma \in C_n)$$

$$\equiv (\tilde{f}_0^{(\omega,\bar{\omega})}, \dots, \tilde{f}_{N-1}^{(\omega,\bar{\omega})}, \tilde{f}_{i*(\sigma|1)}^{(\omega,\bar{\omega})}, \dots, \tilde{f}_{i*\sigma}^{(\omega,\bar{\omega})}, i \in C_1, \sigma \in C_n)$$

$$\stackrel{d}{=} (f_0^{(\omega)}, \dots, f_{N-1}^{(\omega)}, f_{i*(\sigma|1)}^{(\omega)}, \dots, f_{i*\sigma}^{(\omega)}, i \in C_1, \sigma \in C_n). \tag{3.14}$$

It follows from (3.12)–(3.14) that

$$((f_i^{(\omega)}, f_{\sigma|1}^{(\bar{\omega})}, \cdots, f_{\sigma}^{(\bar{\omega})}), \sigma \in C_n, i \in C_1)$$

$$\stackrel{d}{=} ((f_i^{(\omega)}, f_{i*(\sigma|1)}^{(\omega)}, \cdots, f_{i*\sigma}^{(\omega)}), \sigma \in C_n, i \in C_1). \tag{3.15}$$

Hence

$$q(f_{i}^{(\omega)}, f_{i*(\sigma|j)}^{(\omega)}; J_{i}, J_{i*(\sigma|j)}, i \in C_{1}, \sigma \in C_{n}, 1 \ge j \le n)$$

$$\stackrel{d}{=} q(f_{i}^{(\omega)}, f_{\sigma|j}^{(\bar{\omega})}; J_{i}, J_{i*(\sigma|j)}, i \in C_{1}, \sigma \in C_{n}, 1 \le j \le n).$$
(3.16)

But

$$q(f_{i}^{(\omega)}, f_{\sigma|j}^{(\bar{\omega})}; J_{i}, J_{i*(\sigma|j)}, i \in C_{1}, \sigma \in C_{n}, 1 \leq j \leq n)$$

$$= \bigcup_{i \in C_{1}} f_{i}^{(\omega)} \Big( \bigcup_{\sigma \in C_{n}} f_{\sigma|1}^{(\bar{\omega})} \circ \cdots \circ f_{\sigma|n}^{(\bar{\omega})} (J_{i*\sigma}) \Big)$$

$$= \bigcup_{i \in C_{1}} f_{i}^{(\omega)} (t_{n}(\bar{\omega}; J_{i*\sigma}, \sigma \in C_{n})),$$

$$q(f_{i}^{(\omega)}, f_{i*(\sigma|j)}^{(\omega)}; J_{i}, J_{i*(\sigma|j)}, i \in C_{1}, \sigma \in C_{n}, 1 \leq j \leq n)$$

$$= \bigcup_{i \in C_{1}} \bigcup_{\sigma \in C_{n}} f_{i}^{(\omega)} \circ f_{i*(\sigma|1)}^{(\omega)} \circ \cdots \circ f_{i*\sigma}^{(\omega)} (J_{i*\sigma})$$

$$= t_{n+1}(\omega, J_{\tau}, \tau \in C_{n+1}),$$

hence (3.16) means (3.4) is true. The theorem is proved.

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