

# Contraction Mapping on Probabilistic Hyperspace and an Application

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## Abstract

We considered a class of precompact subsets of a PM space (here we called probabilistic hyperspace) and applied the fixed point principle in this space to obtain an attractor of the dynamical system defined through the iterated function systems (IFS).

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## 1 Introduction

Let  $\overline{\mathbf{R}} = [-\infty, \infty]$  be the extended real line. A *distance distribution function* (d.d.f)  $F$  is a non-decreasing function from  $\overline{\mathbf{R}}^+ = [0, +\infty]$  into  $[0, 1]$ , which is left continuous on  $(0, \infty)$  and takes on the values  $F(0) = 0$  and  $F(+\infty) = 1$ . The set of all d.d.f. is denoted by  $\Delta^+$ . We shall also consider the subset  $D^+ = \{F \in \Delta^+ | \lim_{x \rightarrow \infty} F(x) = 1\} \subset \Delta^+$ .

For  $a < \infty$ , the element  $\varepsilon_a \in \Delta^+$  is defined as

$$\varepsilon_a(x) = \begin{cases} 0, & t \leq a; \\ 1, & t > a. \end{cases}$$

By setting  $F \leq G$  whenever  $F(x) \leq G(x)$ ,  $\forall x \in \mathbf{R}^+$ , one introduces a natural ordering in  $\Delta^+$ . The maximal element for  $\Delta^+$  in this ordering is the d.d.f  $\varepsilon_0$

and the minimal element is  $\varepsilon_\infty$ . Convergence in  $\Delta^+$  is assumed to be weakly convergence, i.e.  $F_n \rightarrow F$  if and only if  $F_n(x) \rightarrow F(x)$  at each continuity point  $x$  of  $F$ .

A *triangle function* is a mapping  $\tau: \Delta^+ \times \Delta^+ \rightarrow \Delta^+$  that is commutative, associative, nondecreasing in each variable and has  $\varepsilon_0$  as identity. A triangle function  $\tau$  is said to be sup-continuous if for every family  $\{F_\lambda | \lambda \in \Lambda\}$  of d.d.f.'s in  $\Delta^+$  and every  $G \in \Delta^+$ :

$$\sup_{\lambda \in \Lambda} \tau(F_\lambda, G) = \tau \left( \sup_{\lambda \in \Lambda} F_\lambda, G \right).$$

The typical triangle function is the operation  $\tau_T$  which is given by

$$\tau_T(F, G)(x) = \sup_{u+v=x} T(F(u), G(v)),$$

for all  $F, G \in \Delta^+$  and all  $x > 0$  [[4], Secs. 7.2 and 7.3]. Here  $T$  is a *t-norm*, i.e.,  $T$  is binary operation on  $[0, 1]$  that is commutative, associative, nondecreasing in each place and has 1 as identity. The most important t-norms are the function  $W$ ,  $Prod$ , and  $M$  which are defined, respectively, by

$$\begin{aligned} W(a, b) &= \max\{a + b - 1, 0\}, \\ Prod(a, b) &= ab, \\ M(a, b) &= \min(a, b). \end{aligned}$$

Let  $S$  be a nonempty set and  $\tau$  be a triangle function on  $\Delta^+$ .

**Definition 1** A mapping  $F: S \times S \rightarrow \Delta^+$  is called a *probabilistic metric on  $S$*  if  $F$  satisfies the conditions:  $\forall p, q, r \in S$ ,

$$(PM1) F_{pq} = \varepsilon_0 \Leftrightarrow p = q,$$

$$(PM2) F_{pq} = F_{qp},$$

$$(PM3) F_{pr} \geq \tau(F_{pq}, F_{qr}).$$

The triple  $(S, F, \tau)$  is called a *probabilistic metric space (in short PM space)*.

The detailed study of PM spaces can be found in [4].

**Definition 2** Let  $(S, F, \tau)$  be a PM space and  $A$  is a nonempty subset of  $S$ .

1.  $A$  is said to be *closed* if and only if  $F_{pA} \neq \varepsilon_0$  whenever  $p \notin A$ , where  $F_{pA}(x) = \sup_{q \in A} F_{pq}(x), \forall x > 0$  is the *probabilistic distance* between a point  $p$  and a set  $A$ .
2. The *strong closure* of  $A$  in the sense of Kuratowski [4] is defined as

$$\overline{A} = \{p \in S | F_{pA} = \varepsilon_0\}.$$

3.  $A$  is said to be bounded if  $\lim_{x \rightarrow +\infty} D_A(x) = 1$ , where  $D_A(x) = \sup_{t < x} \inf_{p, q \in A} F_{pq}(t)$  is the probabilistic diameter of  $A$ . Since for any  $p, q \in A$ ,  $F_{pq}(x) \geq D_A(x)$ , [[4], Theorem 12.4.2], this implies that  $\lim_{x \rightarrow +\infty} F_{pq}(x) = 1$  for every  $p, q \in A$ .

The notion of precompact was introduced by Istratescu [3] as follows:

**Definition 3 ([3])** Let  $(S, F, \tau)$  be a PM space and  $A$  be a bounded subset of  $S$ . The set  $A$  is said to be precompact if for any  $t > 0$ , there exists a finite cover  $\{A_1, A_2, \dots, A_n\}$  of  $A$  such that

$$D_{A_j}(t) > 1 - t, \quad \forall j = 1, 2, \dots, n.$$

**Theorem 1 ([3])** A PM space  $(S, F, \tau)$  is said to be precompact iff for every  $t > 0$ , there exists a finite set  $A_t$  in  $S$  such that for any  $p \in S$ , there is a  $q \in A_t$  such that

$$F_{pq}(t) > 1 - t.$$

Sehgal and Bharucha [5] introduced the notion of a contraction mapping in PM spaces and provided a prove for a fixed point theorem in complete PM spaces. Further work on contraction mappings and fixed point theorem in PM spaces can be found in [[2],[3]].

**Definition 4 ([5])** Let  $(S, F, \tau)$  be a PM space. A mapping  $f: S \rightarrow S$  is called a probabilistic contraction map on  $(S, F, \tau)$  if there exists a  $k \in (0, 1)$ , called a contractive factor such that for every  $p, q \in S$  and  $x > 0$ ,

$$F_{f(p), f(q)}(x) \geq F_{pq}\left(\frac{x}{k}\right).$$

**Theorem 2 ([5])** Let  $(S, F, \tau)$  be a complete PM space. If  $f: S \rightarrow S$  is a probabilistic contraction mapping on  $(S, F, \tau)$  then  $f$  posses exactly one fixed point  $p \in S$ , i.e.,  $f(p) = p$ .

## 2 Probabilistic Hyperspaces

Let  $(S, F, \tau)$  be a PM space and  $K_F(S) = \{A \subset S | A \text{ is nonempty and precompact set}\}$ . For any  $A, B \in K_F(S)$ , the *probabilistic Hausdorff distance* between  $A$  and  $B$  [2] is given by

$$H_{AB}(x) = \text{Min sup}_{t < x} \left\{ \inf_{p \in A} F_{pB}(t), \inf_{q \in B} F_{Aq}(t) \right\}, \quad (1)$$

where  $F_{pA}(x) = \sup_{q \in A} F_{pq}(x)$ .

From [[4], Theorem 12.9.3, 12.9.4 and 12.9.5],  $H_{AB}$  is a probabilistic metric provided  $\tau$  is sup-continuous in  $(S, F, \tau)$ . Whence we conclude that the triple  $(K_F(S), H, \tau)$  is a PM space with  $\tau$  is sup-continuous in  $(S, F, \tau)$ . We shall call the complete PM space  $(K_F(S), H, \tau)$  as a *probabilistic hyperspace*.

We prove the following fixed point theorem which will be the extension of Theorem 2.

**Theorem 3** *Let  $(S, F, \tau)$  be a bounded complete PM space with  $\text{Range}(F) \subset D^+$ . Let  $T: K_F(S) \rightarrow K_F(S)$  (where the set  $K_F(S)$  as defined above) be a mapping such that*

$$H_{TA, TB}(x) \geq H_{AB} \left( \frac{x}{k} \right), \quad (2)$$

for every  $A, B \in K_F(S)$  and for every  $k \in (0, 1)$ . Then  $T$  posses an unique fixed point  $U \in K_F(S)$ .

**Proof** From (1), the left side of (2) can be expressed as:

$$H_{TA, TB}(x) \geq \text{Min sup}_{t < x} \left\{ \inf_{p \in A} F_{Tp, TB}(t), \inf_{q \in B} F_{TA, Tq}(t) \right\}.$$

Hence, for every  $Tp \in TA$ , there exists  $Tq \in TB$  so that,

$$H_{TA, TB}(x) \geq \sup_{t < x} F_{Tp, Tq}(t). \quad (3)$$

By using (1) again to the right side of (2) and we can conclude that, there exists  $k \in (0, 1)$  such that for every  $p \in A$ , there exist  $q \in B$ , so that

$$F_{AB}(x) \geq \sup_{kt < x} F_{pq}(t). \quad (4)$$

By comparing both (3) and (4) with (2), one can conclude that:

$$\sup_{t < x} F_{Tp, Tq}(t) \geq \sup_{kt < x} F_{pq}(t). \quad (5)$$

Let  $y = kt > 0$ . Then by noting that  $x > t > kt = y$ , we write (5) as:

$$\sup_{y < x} F_{Tp, Tq}(y) \geq \sup_{y < x} F_{pq} \left( \frac{y}{k} \right),$$

which implies that

$$F_{T_p, T_q}(y) \geq F_{pq} \left( \frac{y}{k} \right), \quad (6)$$

for every  $y > 0$  and  $k \in (0, 1)$ .

It is worth noting that, if for each  $p, q \in U \subset S$  with  $\bar{U} \in K_F(S)$ , the mapping  $T: \bar{U} \rightarrow \bar{U}$  is a contraction mapping in  $(S, F, \tau)$  which satisfies the condition (6). Then by Theorem 2,  $T$  has a unique fixed point, say  $u \in \bar{U}$ . Since  $u \in \bar{U}$  is arbitrary in  $\bar{U}$ , we can conclude that the mapping  $T: K_F(S) \rightarrow K_F(S)$  has unique fixed point  $\bar{U} \in K_F(S)$ .

### 3 Application

In this section we consider the iterated function system (IFS) on a precompact PM space which is a generalization of IFS on a complete metric spaces (Barnsley [1]).

**Definition 5** *An iterated function system (IFS) on a precompact PM space  $(S, F, \tau)$  is a finite collection of contraction mappings  $f_i : S \rightarrow S, i = 1, \dots, n$ , with the contractive property: for every  $p, q \in S$  and  $x > 0$ ,*

$$F_{f_i(p), f_i(q)}(x) = F_{p, q} \left( \frac{x}{k_i} \right)$$

where  $k_i, i = 1, \dots, n$ , is the contractivity factors.

Our goal is to describe the *attractor* of an IFS. The basic step in forming this attractor is to start with a precompact set  $A \in K_F(S)$  and apply the contractions  $f_i(A), i = 1, \dots, n$ . We can formulate this process as a mapping of precompact sets:

$$W(A) = \bigcup_{i=1}^n f_i(A) \quad (7)$$

and a dynamical system

$$A \mapsto W(A) \mapsto W^2(A) \mapsto \dots \mapsto W^m(A) \mapsto \dots$$

where the attractor is the limit of this sequence  $\{W^m(A)\}$ .

Let  $(K_F(S), H, \tau)$  be a probabilistic hyperspace as defined in section 2. We say that a sequence  $\{A_n\} \subset K_F(S)$  converges weakly to  $A \in K_F(S)$  if  $F_{A_n A} \rightarrow \varepsilon_0$  as  $n \rightarrow \infty$ .

Returning to the IFS  $\{f_1, \dots, f_n\}$  and the associated map  $W : K_F(S) \rightarrow K_F(S)$  as defined in (7), we claim that the sequence of iterates  $W^m(U)$  will converges weakly to some precompact set  $U$ . To prove this, we need the following theorem.

**Theorem 4** Let  $(K_F(S), H, \tau)$  be a probabilistic hyperspace. The mapping  $W : K_F(S) \rightarrow K_F(S)$  is a probabilistic contraction on  $K_F(S)$  with the contractivity factor  $k \in (0, 1)$ , i.e.,

$$H_{W(A), W(B)}(x) \geq H_{AB} \left( \frac{x}{k} \right)$$

for every  $A, B \in K_F(S)$ .

**Proof** Let  $W(A) = \bigcup_{i=1}^n f_i(A)$  and  $W(B) = \bigcup_{i=1}^n f_i(B)$ . For any  $p \in W(A)$  and  $x > 0$ ,

$$\begin{aligned} F_{p, W(B)}(x) &= \sup\{F_{p, q}(x) | q \in W(B)\} \\ &= \sup\{F_{p, f_i(B)}(x) | i = 1, \dots, n\} \\ &\geq F_{p, f_j(B)}(x) \end{aligned}$$

for some  $j = 1, \dots, n$ . Since  $p \in W(A)$ , we have  $p = f_j(\hat{p})$  for some  $j = 1, \dots, n$  and for some  $\hat{p} \in A$ . Hence, we have

$$F_{p, W(B)}(x) \geq F_{p, f_j(B)}(x) = F_{f_j(\hat{p}), f_j(B)}(x), \quad x > 0.$$

Since  $q = f_j(\hat{q})$ , for some  $\hat{q} \in B$ , by the principle of contraction, we have

$$F_{p, W(B)}(x) \geq F_{f_j(\hat{p}), f_j(\hat{q})}(x) \geq F_{\hat{p}, \hat{q}} \left( \frac{x}{k_j} \right), \quad \forall j = 1, 2, \dots \text{ and } x > 0.$$

By letting  $k = \max\{k_j | j = 1, 2, \dots, n\} < 1$ ,

$$F_{p, W(B)}(x) \geq F_{\hat{p}, \hat{q}} \left( \frac{x}{k} \right), \quad x > 0. \quad (8)$$

Note that, (8) can be written as

$$F_{p, W(B)}(x) \geq \sup_{\hat{q} \in B} F_{\hat{p}, \hat{q}} \left( \frac{x}{k} \right) = F_{\hat{p}, B} \left( \frac{x}{k} \right), \quad \text{for every } x > 0 \text{ and } k \in (0, 1).$$

Since  $p \in W(A)$  and  $\hat{p} \in A$ , one can conclude that:

$$\inf_{p \in W(A)} F_{p, W(B)}(x) \geq \inf_{\hat{p} \in A} F_{\hat{p}, B} \left( \frac{x}{k} \right), \quad \text{for every } x > 0 \text{ and } k \in (0, 1). \quad (9)$$

By the symmetric property of the probabilistic metric, and with  $\hat{q} \in B, q = f_j(\hat{q})$  and  $k = \max k_i$ , one can show that:

$$\inf_{q \in W(B)} F_{W(A), q}(x) \geq \inf_{\hat{q} \in B} F_{A, \hat{q}} \left( \frac{x}{k} \right), \quad \text{for every } x > 0 \text{ and } k \in (0, 1).$$

Thus, by the property of (1), (8) and (9), we have:

$$\begin{aligned} H_{W(A),W(B)}(x) &= \text{Min sup}_{t < x} \left\{ \inf_{p \in W(A)} F_{p,W(B)}(t), \inf_{q \in W(B)} F_{W(A),q}(t) \right\} \\ &\geq \text{Min sup}_{t < x} \left\{ \inf_{\hat{p} \in A} F_{\hat{p},B} \left( \frac{t}{k} \right), \inf_{\hat{q} \in B} F_{A,\hat{q}} \left( \frac{t}{k} \right) \right\} \\ &= H_{AB} \left( \frac{x}{k} \right) \end{aligned}$$

Thus,  $H_{W(A),W(B)}(x) \geq H_{AB} \left( \frac{x}{k} \right)$  for every  $x > 0$  and  $k \in (0, 1)$ . And this completes the proof.

Now, Theorem 3 determines a unique fixed point  $U$  of  $W$  and asserts that all sequences of iterates of  $W^m(U)$  converges upon  $U$ . Hence we have the attractor of the IFS, which is the fixed point  $U \in K_F(S)$ .

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## References

- [1] M. Barnsley, *Fractals Everywhere*, Academic Press, New York, 1993.
- [2] G. Constantin and I. Istratescu, *Elements of Probabilistic Analysis with Application*, Kluwer Academic, Dordrecht, 1989.
- [3] V. I. Istratescu, *Fixed Point Theory*, D. Reidel Publishing Company, Dordrecht, 1981.
- [4] B. Schweizer and A. Sklar, *Probabilistic Metric Spaces*, Elsevier North-Holand, New York, 1983.
- [5] V. M. Sehgal and A. T. Bharucha-Reid, Fixed points of contraction mappings on PM spaces, *Math. System Theory*, 6(1972), 97-102.

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