

# The Time Average Reward for Some Dynamic Fuzzy Systems

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## Abstract

In this paper, by using a fuzzy relation, we define a dynamic fuzzy system with a bounded convex fuzzy reward on the positive orthant  $\mathbf{R}_+^n$  of an  $n$ -dimensional Euclidean space. As a measure of the system's performance we introduce the time average fuzzy reward, which is characterized by the limiting fuzzy state under the contractive properties of the fuzzy relation. In one-dimensional case, the average fuzzy reward is expressed explicitly by the functional equations concerning the extreme points of its  $\alpha$ -cuts. Also, a numerical example is given to illustrate the theoretical results.

**Keywords:** dynamic fuzzy system, time average fuzzy reward, contractive properties, fuzzy relational equation.

## 1. Introduction and notations

In the previous papers, Kurano etc. [4], [11], [13], [14], we have defined a dynamic fuzzy system using a fuzzy relation and proved a limit theorem for transition of fuzzy states under the contractive properties of the fuzzy relation. Here, the dynamic fuzzy system will be extended to the one with a bounded

fuzzy reward on the positive orthant  $\mathbf{R}_+^n$  of an  $n$ -dimensional Euclidean space and the time average fuzzy reward is introduced as a measure of the system's performance and characterized by the limiting fuzzy state or by various fuzzy relational equations. For sequential decision analyses in a fuzzy environment, refer to Bellman and Zadeh[1], Esogbue and Bellman[2], Kurano etc.[5].

Let  $X$  and  $Y$  be convex subsets of some Banach space. We denote by  $\mathcal{C}(X)$  the collection of all compact convex subsets of  $X$  and  $\rho$  be the Hausdorff metric on  $\mathcal{C}(X)$ . Throughout this paper we denote a fuzzy set on  $X$  by its membership function  $\tilde{s} : X \mapsto [0, 1]$  and its  $\alpha$ -cut by  $\tilde{s}_\alpha$ . For the details, refer to Zadeh[15], Novák[9] and the previous our papers.

A fuzzy set  $\tilde{s}$  on  $X$  is called convex if

$$\tilde{s}(\lambda x + (1 - \lambda)y) \geq \tilde{s}(x) \wedge \tilde{s}(y) \quad \text{for any } x, y \in X \text{ and } \lambda \in [0, 1],$$

where  $a \wedge b = \min\{a, b\}$  for real numbers  $a$  and  $b$ . Also, a fuzzy relation  $\tilde{p}$  defined on  $X \times Y$  is called convex if

$$\tilde{p}(\lambda x_1 + (1 - \lambda)x_2, \lambda y_1 + (1 - \lambda)y_2) \geq \tilde{p}(x_1, y_1) \wedge \tilde{p}(x_2, y_2)$$

for any  $x_1, x_2 \in X, y_1, y_2 \in Y$ , and  $\lambda \in [0, 1]$ .

Let  $\mathcal{F}(X)$  be the set of all convex fuzzy sets  $\tilde{s}$  on  $X$ , which are upper semi-continuous and have a compact support. Clearly  $\tilde{s} \in \mathcal{F}(X)$  implies  $\tilde{s}_\alpha \in \mathcal{C}(X)$  for all  $\alpha \in [0, 1]$ . The addition and the multiplicative operation of fuzzy sets are defined as follows (see Madan etc.[7]) : For any  $\tilde{s}, \tilde{v} \in \mathcal{F}(\mathbf{R}_+^n)$  and  $\lambda \in \mathbf{R}_+ := [0, \infty)$ ,

$$(\tilde{s} + \tilde{v})(x) := \sup_{y, z \in \mathbf{R}_+^n : y+z=x} \{\tilde{s}(y) \wedge \tilde{v}(z)\} \quad x \in \mathbf{R}_+^n \quad (1.1)$$

and

$$(\lambda \tilde{s})(x) := \tilde{s}(x/\lambda) \quad \text{if } \lambda > 0, \quad x \in \mathbf{R}_+^n, \quad (1.2)$$

and  $(\lambda \tilde{s})(x) := I_{\{0\}}(x)$  if  $\lambda = 0$ , where  $I_A(\cdot)$  is the indicator function of a subset  $A$  of  $\mathbf{R}_+^n$ .

It is easily seen that, for  $\alpha \in [0, 1]$ ,

$$(\tilde{s} + \tilde{v})_\alpha = \tilde{s}_\alpha + \tilde{v}_\alpha \quad \text{and} \quad (\lambda \tilde{s})_\alpha = \lambda \tilde{s}_\alpha,$$

where  $A + B := \{x + y \mid x \in A, y \in B\}$  and  $\lambda A := \{\lambda x \mid x \in A\}$  for any subsets  $A, B$  of  $\mathbf{R}_+^n$ .

**Lemma 1.1.**(Chen-wei Xu [3])

(i) For any  $\tilde{s}, \tilde{v} \in \mathcal{F}(\mathbf{R}_+^n)$  and  $\lambda \in [0, \infty)$ ,

$$\tilde{s} + \tilde{v} \in \mathcal{F}(\mathbf{R}_+^n) \quad \text{and} \quad \lambda \tilde{v} \in \mathcal{F}(\mathbf{R}_+^n).$$

(ii) Let  $\tilde{p}$  be any lower semi-continuous convex fuzzy relation on  $X \times Y$ . Then

$$\sup_{x \in X} \tilde{s}(x) \wedge \tilde{p}(x, \cdot) \in \mathcal{F}(Y) \quad \text{for all } \tilde{s} \in \mathcal{F}(X).$$

Here, we give the notion of convergence for a sequence of fuzzy sets, which is used in Section 2.

**Definition 1.1** (Kurano etc.[4], Nanda[8]). Let  $\{\tilde{v}_t\}_{t=0}^\infty$  be a sequence fuzzy sets in  $\mathcal{F}(X)$ . Then we write  $\tilde{v}_t \rightarrow \tilde{v} \in \mathcal{F}(X)$  as  $t \rightarrow \infty$  if

$$\lim_{t \rightarrow \infty} \sup_{\alpha \in [0,1]} \rho(\tilde{v}_{t,\alpha}, \tilde{v}_\alpha) = 0, \quad (1.3)$$

where  $\tilde{v}_{t,\alpha}$  and  $\tilde{v}_\alpha$  are  $\alpha$ -cuts of  $\tilde{v}_t$  and  $\tilde{v}$  respectively.

Note that for a sequence of sets  $\{A_t\}_{t=1}^\infty \subset \mathcal{C}(X)$  and  $A \in \mathcal{C}(X)$ ,

$$\lim_{t \rightarrow \infty} A_t = A$$

means that  $\overline{\lim}_{t \rightarrow \infty} A_t = \underline{\lim}_{t \rightarrow \infty} A_t = A$ , where

$$\overline{\lim}_{t \rightarrow \infty} A_t := \{z \in X \mid \underline{\lim}_{t \rightarrow \infty} d(z, A_t) = 0\},$$

$$\underline{\lim}_{t \rightarrow \infty} A_t := \{z \in X \mid \overline{\lim}_{t \rightarrow \infty} d(z, A_t) = 0\},$$

$d(z, D) := \min_{z' \in D} d(z, z')$   $D \in \mathcal{C}(X)$  and  $d$  is a metric on  $X$ . It is known (Kuratowski[6]) that  $\lim_{t \rightarrow \infty} \rho(A_t, A) = 0$  iff  $\lim_{t \rightarrow \infty} A_t = A$ , so that  $\tilde{v}_t$  converges to  $\tilde{v}$  as  $t \rightarrow \infty$  in the sense of (1.3) means that  $\lim \tilde{v}_{t,\alpha} = \tilde{v}_\alpha$  uniformly for  $\alpha \in [0, 1]$ .

Now, extending a discrete dynamic fuzzy system in Kurano etc. [4], [11], [13], [14], we consider the one with a fuzzy reward, which is characterized with the elements  $(S, \tilde{q}, \tilde{r}, \tilde{s})$  as follows :

- (i) The state space  $S$  is a convex compact subset of some Banach space. In general, the system is fuzzy, so that the state of the system is called a fuzzy state denoted as an element of  $\mathcal{F}(S)$ .
- (ii) The law of the motion and the fuzzy reward for the system are denoted by the time invariant fuzzy relations  $\tilde{q} : S \times S \mapsto [0, 1]$  and  $\tilde{r} : S \times [0, M]^n \mapsto [0, 1]$  respectively, where  $M$  is a fixed positive number and  $n$  is a positive integer. We assume that  $\tilde{q} : S \times S \mapsto [0, 1]$  and  $\tilde{r} : S \times [0, M]^n \mapsto [0, 1]$  are convex and continuous.

If the system is in a fuzzy state  $\tilde{s} \in \mathcal{F}(S)$ , a fuzzy reward  $R(\tilde{s})$  is incurred and we move to a new fuzzy state  $Q(\tilde{s})$ , where  $Q : \mathcal{F}(S) \mapsto \mathcal{F}(S)$  and  $R : \mathcal{F}(S) \mapsto \mathcal{F}([0, M]^n)$  are defined by

$$R(\tilde{s})(z) := \sup_{x \in S} \tilde{s}(x) \wedge \tilde{r}(x, z) \quad z \in [0, M]^n \quad (1.4)$$

and

$$Q(\tilde{s})(y) := \sup_{x \in S} \tilde{s}(x) \wedge \tilde{q}(x, y) \quad y \in S. \quad (1.5)$$

Note that by Lemma 1.1(ii) the maps  $R$  and  $Q$  are well-defined.

(iii) The initial fuzzy state  $\tilde{s} \in \mathcal{F}(S)$  is arbitrary.

For the dynamic fuzzy system  $(S, \tilde{q}, \tilde{r}, \tilde{s})$ , we can define a sequence of fuzzy rewards on  $[0, M]^n$ ,  $\{R(\tilde{s}_t)\}_{t=0}^\infty$ , where

$$\tilde{s}_0 := \tilde{s} \quad \text{and} \quad \tilde{s}_{t+1} := Q(\tilde{s}_t) \quad (t \geq 0). \quad (1.6)$$

In Section 2, we define the time average fuzzy reward, which is characterized by the limiting fuzzy state under the contractive properties of the fuzzy relation  $\tilde{q}$ .

In Section 3, the one-dimensional case is treated and by introducing relative value functions the average fuzzy reward is expressed by the functional equations concerning the extreme points of its  $\alpha$ -cuts.

Also, a numerical example is given to illustrate the theoretical results in this paper.

## 2. The average fuzzy reward

In this paper we specify the time average reward as a measure of the system's performance and discuss its characterization under the contractive assumption given in Kurano etc.[4].

We define the total  $T$ -time fuzzy reward  $\tilde{R}_T(\tilde{s})$  by

$$\tilde{R}_T(\tilde{s}) := \sum_{t=0}^{T-1} R(\tilde{s}_t) \quad T \geq 1, \quad (2.1)$$

where  $\{\tilde{s}_t\}_{t=0}^\infty$  is given in (1.6).

Associated with the fuzzy relation  $\tilde{q}$  and fuzzy reward  $\tilde{r}$ , are the corresponding maps  $Q_\alpha : \mathcal{C}(S) \mapsto \mathcal{C}(S)$  ( $\alpha \in [0, 1]$ ) and  $R_\alpha : \mathcal{C}(S) \mapsto \mathcal{C}([0, M]^n)$  ( $\alpha \in [0, 1]$ ) defined as follows : For  $D \in \mathcal{C}(S)$ ,

$$Q_\alpha(D) := \begin{cases} \{y \in S \mid \tilde{q}(x, y) \geq \alpha \text{ for some } x \in D\} & \alpha > 0 \\ cl\{y \in S \mid \tilde{q}(x, y) > 0 \text{ for some } x \in D\} & \alpha = 0, \end{cases} \quad (2.2)$$

and

$$R_\alpha(D) := \begin{cases} \{z \in [0, M]^n \mid \tilde{r}(x, z) \geq \alpha \text{ for some } x \in D\} & \alpha > 0 \\ cl\{z \in [0, M]^n \mid \tilde{r}(x, z) > 0 \text{ for some } x \in D\} & \alpha = 0. \end{cases} \quad (2.3)$$

The iterates  $Q_\alpha^t$  ( $t \geq 0$ ) are defined by setting  $Q_\alpha^0 := I$  (identity) and iteratively,

$$Q_\alpha^{t+1} := Q_\alpha Q_\alpha^t \quad t \geq 0.$$

We have the following lemma, which is easily verified by the ideas in the proof of Kurano etc.[4, Lemma 1].

**Lemma 2.1.**

- (i)  $\tilde{R}_T(\tilde{s}) \in \mathcal{F}([0, TM]^n)$  for  $T \geq 1$ .
- (ii)  $\tilde{s}_{t,\alpha} = Q_\alpha^t(\tilde{s}_\alpha)$  for  $t \geq 0$ , where  $\tilde{s}_{t,\alpha} = (\tilde{s}_t)_\alpha$ .
- (iii)  $(\tilde{R}_T(\tilde{s}))_\alpha = \sum_{t=0}^{T-1} R_\alpha(\tilde{s}_{t,\alpha})$  for  $T \geq 1$ .

From Lemma 2.1(ii),(iii), the  $\alpha$ -cut of rewards,  $\tilde{R}_T(\tilde{s})$ , can be calculated only through  $\tilde{s}_\alpha$ . So we denote it as

$$\tilde{R}_{T,\alpha}(\tilde{s}_\alpha) := (\tilde{R}_T(\tilde{s}))_\alpha \quad \text{for } T \geq 0 \text{ and } \alpha \in [0, 1].$$

From this  $\alpha$ -cut set of  $\tilde{R}_{T,\alpha}(\tilde{s}_\alpha)$  we try to estimate the increasing amount of fuzzy reward per unit time.

For  $K > 0$  and  $\alpha \in [0, 1]$ , we define

$$G_{K,\alpha} := \left\{ r \in \mathbf{R}_+^n \mid \begin{array}{l} \text{there exists } \{z_T\}_{T=1}^\infty \text{ such that} \\ z_T \in \tilde{R}_{T,\alpha}(\tilde{s}_\alpha) \text{ and } \|z_T - rT\| \leq K \text{ for all } T \geq 1 \end{array} \right\}. \quad (2.4)$$

The properties of  $G_{K,\alpha}$  are formulated in the following lemma. The proof is omitted.

**Lemma 2.2.** *Let  $K > 0$ . Then :*

- (i)  $\{G_{K,\alpha} \mid \alpha \in [0, 1]\} \subset \mathcal{C}(\mathbf{R}_+^n)$ .
- (ii)  $G_{K,\alpha} \subset G_{K,\alpha'}$  for  $0 \leq \alpha' \leq \alpha \leq 1$ .
- (iii)  $\lim_{\alpha' \uparrow \alpha} G_{K,\alpha'} = G_{K,\alpha}$  for  $\alpha \in (0, 1]$ , i.e.,  $\lim_{\alpha' \uparrow \alpha} \delta(G_{K,\alpha'}, G_{K,\alpha}) = 0$ .

From Kurano etc.[4, Lemma 3], we can define a fuzzy number

$$\tilde{g}(\tilde{s})(r) := \sup_{\alpha \in [0,1]} \{\alpha \wedge I_{G_{K,\alpha}}(r)\} \quad r \in [0, M]^n \quad \text{for } \tilde{s} \in \mathcal{F}(S). \quad (2.5)$$

Then,  $\tilde{g}(\tilde{s}) \in \mathcal{F}([0, M]^n)$  and  $(\tilde{g}(\tilde{s}))_\alpha = G_{K,\alpha}$  for all  $\alpha \in [0, 1]$ .

We call  $\tilde{g}(\tilde{s})$  an *average fuzzy reward* for the dynamic fuzzy systems, which depends on the initial fuzzy state  $\tilde{s} \in \mathcal{F}(S)$  with suppression of  $K$ . In the remainder of this section, we will investigate the average fuzzy reward from the limiting behavior of the fuzzy states. The following lemma is useful in the sequel.

**Lemma 2.3.** *Let  $\{D_t\}_{t=1}^{\infty} \subset \mathcal{C}(S)$  and  $D \in \mathcal{C}(S)$  such that  $\lim_{t \rightarrow \infty} D_t = D$ . Let  $\alpha \in (0, 1]$ . For any  $\epsilon$  ( $\alpha > \epsilon > 0$ ), there exists  $T \geq 1$  such that*

$$R_{\alpha-\epsilon}(D) \supset R_{\alpha}(D_t) \quad \text{for all } t \geq T.$$

**Proof.** Suppose that for some  $\epsilon$  ( $\alpha > \epsilon > 0$ ), there exist sequences  $\{t_k\}_{k=1}^{\infty}$  and  $\{z_k\}_{k=1}^{\infty}$  such that

$$t_k \rightarrow \infty \quad (k \rightarrow \infty), \quad \text{and} \quad z_k \in R_{\alpha}(D_{t_k}) \setminus R_{\alpha-\epsilon}(D) \quad (k = 1, 2, \dots).$$

Then we have

$$\tilde{r}(x, z_k) < \alpha - \epsilon \quad \text{for all } x \in D, k = 1, 2, \dots, \quad (2.10)$$

and there exists a sequence  $\{x_k\}_{k=1}^{\infty}$  such that

$$x_k \in D_{t_k} \quad \text{and} \quad \tilde{r}(x_k, z_k) \geq \alpha \quad \text{for } k = 1, 2, \dots. \quad (2.11)$$

From the compactness, we may assume that the sequences  $\{x_k\}_{k=1}^{\infty}$  and  $\{z_k\}_{k=1}^{\infty}$  are convergent. We put the limits  $x^* = \lim_{k \rightarrow \infty} x_k$  and  $z^* = \lim_{k \rightarrow \infty} z_k$ . Then we have  $x^* \in D$  since  $\lim_{t \rightarrow \infty} D_t = D$ . From (2.10) and (2.11), we obtain

$$\tilde{r}(x^*, z^*) \geq \alpha \quad \text{and} \quad \tilde{r}(x, z^*) \leq \alpha - \epsilon \quad \text{for all } x \in D.$$

It is a contradiction. Thus we get this lemma.  $\square$

In order to characterizing the average fuzzy reward  $\tilde{g}(\tilde{s})$ , we need the following two assumptions, the first one is a contractive property concerning the fuzzy relation  $\tilde{q}$  which guarantee the existence of the limiting fuzzy state and the second is a Lipschitz condition related with the fuzzy reward  $\tilde{r}$ .

**Assumption A.** (Contraction and ergodic property)  
There exists  $t_0 \geq 1$  and  $\beta$  ( $0 < \beta < 1$ ) satisfying that

$$\rho(Q_{\alpha}^{t_0}(D_1), Q_{\alpha}^{t_0}(D_2)) \leq \beta \rho(D_1, D_2) \quad \text{for all } D_1, D_2 \in \mathcal{C}(S), \alpha \in [0, 1].$$

**Assumption B.** (Lipschitz conditions)  
There exists a constant  $C > 0$  such that

$$\delta(R_{\alpha}(D_1), R_{\alpha}(D_2)) \leq C \rho(D_1, D_2) \quad \text{for all } D_1, D_2 \in \mathcal{C}(S), \alpha \in [0, 1], \quad (2.12)$$

where  $\delta$  is the Hausdorff metric on  $\mathcal{C}([0, M]^n)$ .

**Lemma 2.4.** (Kurano etc.[4, Theorem 1]) *Suppose that Assumption A holds.*

- (i) There exists a unique fuzzy state  $\tilde{p} \in \mathcal{F}(S)$ , which is independently of the initial fuzzy state  $\tilde{s}$ , satisfying

$$\tilde{p}(y) = \max_{x \in S} \{\tilde{p}(x) \wedge \tilde{q}(x, y)\} \quad \text{for all } y \in S. \quad (2.13)$$

- (ii) For  $\alpha \in [0, 1]$ , the  $\alpha$ -cut  $\tilde{p}_\alpha$  is a unique set of  $\mathcal{C}(S)$  such that

$$Q_\alpha(\tilde{p}_\alpha) = \tilde{p}_\alpha.$$

- (iii) Let  $\alpha \in [0, 1]$ . It holds that

$$\rho(Q_\alpha^t(D), \tilde{p}_\alpha) \leq \beta^{[t/t_0]} K_\alpha(D, \tilde{p}_\alpha) \quad \text{for all } D \in \mathcal{C}(S), t \geq 1,$$

where  $K_\alpha(D, \tilde{p}_\alpha) := \sum_{l=0}^{t_0-1} \rho(Q_\alpha^l(D), \tilde{p}_\alpha)$  and, for a real number  $c$ ,  $[c]$  is the largest integer equal to or less than  $c$ .

Recently, Yoshida[12] has given the notion of  $\alpha$ -recurrent set for the fuzzy relation and shown that the  $\alpha$ -cut of the limiting fuzzy set  $\tilde{p}$  in Lemma 2.4 is characterized as the maximum  $\alpha$ -recurrent set.

Now, we can state one of main results, which shows that  $\tilde{g}(\tilde{s})$  is represented using the limiting fuzzy state  $\tilde{p}$ .

**Theorem 2.1.** *Suppose that Assumptions A and B hold. For sufficient large all  $K$ , it holds that*

$$\tilde{g}(\tilde{s}) = R(\tilde{p}), \quad (2.14)$$

where  $\tilde{p}$  is the limiting fuzzy state given in Lemma 2.3. Further this is independent of the initial fuzzy state  $\tilde{s}$ .

**Proof.** A rough sketch of the proof is as follows and the details are omitted. First we show that

$$(\tilde{g}(\tilde{s}))_\alpha = G_{K,\alpha} \subset R_\alpha(\tilde{p}_\alpha) = (R(\tilde{p}))_\alpha. \quad (2.15)$$

Suppose that there exists  $r \in G_{K,\alpha} \setminus R_\alpha(\tilde{p}_\alpha)$ . Then  $r \notin R_{\frac{\alpha+\epsilon}{2}}(\tilde{p}_\alpha)$  for some  $\epsilon > 0$ . Since  $R_{\frac{\alpha+\epsilon}{2}}(\tilde{p}_\alpha)$  is closed and convex, there exists a unique  $z_0 \in R_{\frac{\alpha+\epsilon}{2}}(\tilde{p}_\alpha)$  such that

$$0 < \gamma := \|z_0 - r\| \leq \|z - r\| \quad \text{for all } z \in R_{\frac{\alpha+\epsilon}{2}}(\tilde{p}_\alpha). \quad (2.17)$$

By Lemma 2.3, there exists  $T^* > 0$  such that

$$R_{\frac{\alpha+\epsilon}{2}}(\tilde{p}_\alpha) \supset R_\alpha(\tilde{s}_{t,\alpha}) \quad \text{for all } t \geq T^*. \quad (2.18)$$

From  $r \in G_{K,\alpha}$ , there exists  $\{r_T\}_{T=0}^\infty$  such that

$$r_T \in \tilde{R}_{T,\alpha}(\tilde{s}_\alpha) \quad \text{and} \quad \|r_T - rT\| \leq K \quad \text{for all } T \geq 1. \quad (2.19)$$

On the other hand, from Lemma 2.1(iii), there exists a sequence  $\{r_{T,t}\}$  such that

$$r_{T,t} \in R_\alpha(\tilde{s}_{t,\alpha}) \quad (t = 0, 1, 2, \dots, T-1) \text{ and } r_T = \sum_{t=0}^{T-1} r_{T,t} \quad (T \geq 1). \quad (2.20)$$

Noting the supporting hyperplane of  $R_{\frac{\alpha+t}{2}}(\tilde{p}_\alpha)$  at  $z_0$ , we have

$$\langle z_0 - r, r_{T,t} - r \rangle \geq \|z_0 - r\|^2 = \gamma^2 \quad \text{for all } t, T \quad (T > t \geq T^*)$$

and

$$\left\langle z_0 - r, \sum_{t=T^*}^{T-1} (r_{T,t} - r) \right\rangle \geq (T - T^*)\gamma^2 \quad \text{for all } T > T^*.$$

By Cauchy-Schwartz inequality,

$$\left\| \sum_{t=T^*}^{T-1} (r_{T,t} - r) \right\| \geq (T - T^*)\gamma \quad \text{for all } T > T^*. \quad (2.21)$$

After some calculations we see that

$$\|r_T - rT\| = \left\| \sum_{t=0}^{T-1} (r_{T,t} - r) \right\| \rightarrow \infty \quad (T \rightarrow \infty).$$

So this contradicts (2.19) and we obtain (2.15).

Next we prove

$$R_\alpha(\tilde{p}_\alpha) \subset G_{K,\alpha} \quad \text{for sufficient large all } K. \quad (2.22)$$

From Assumption B, we have

$$\delta(R_\alpha(\tilde{s}_{t,\alpha}), R_\alpha(\tilde{p}_\alpha)) \leq C\rho(\tilde{s}_{t,\alpha}, \tilde{p}_\alpha) \quad \text{for } t \geq 0. \quad (2.23)$$

Also, from Lemmas 2.1(ii) and 2.4(iii),

$$\rho(\tilde{s}_{t,\alpha}, \tilde{p}_\alpha) \leq \beta^{[t/t_0]} K_\alpha(\tilde{s}_\alpha, \tilde{p}_\alpha) \quad \text{for } t \geq 0. \quad (2.24)$$

Since  $S$  is compact, there exists a constant  $C^* > 0$  such that

$$\delta(R_\alpha(\tilde{s}_{t,\alpha}), R_\alpha(\tilde{p}_\alpha)) \leq C^* \beta^t \quad \text{for } t \geq 0$$

by using (2.23),(2.24). Therefore, for any  $r \in R_\alpha(\tilde{p}_\alpha)$ , there exists  $\{r_t\}_{t=0}^\infty$  such that

$$r_t \in R_\alpha(\tilde{s}_{t,\alpha}) \text{ and } \|r_t - r\| \leq C^* \beta^t \quad (2.25)$$



for  $t \geq 0$ . Then

$$\left\| \sum_{t=0}^{T-1} r_t - rT \right\| = \left\| \sum_{t=0}^{T-1} (r_t - r) \right\| \leq \sum_{t=0}^{T-1} \|r_t - r\| \leq \sum_{t=0}^{T-1} C^* \beta^t \leq C^*/(1 - \beta)$$

for all  $T \geq 1$ . Thus we get  $r \in G_{K,\alpha}$  for all  $K \geq C^*/(1 - \beta)$ . Therefore (2.22) holds for all  $K \geq C^*/(1 - \beta)$ . Together with (2.15), we get (2.14) for sufficient large all  $K$ . It is trivial that (2.14) is independent of the initial fuzzy state  $\tilde{s}$  from Lemma 2.4(i).  $\square$

From now on we take  $K \geq C^*/(1 - \beta)$ . The following corollary shows that  $\tilde{g}(\tilde{s})$  is given as the limit of  $\{R(\tilde{s}_t)\}_{t=0}^{\infty}$  by the method of Cesaro averaging. The proof is omitted.

**Corollary 2.1.** *Under the same condition as Theorem 2.1, it holds that*

$$\lim_{T \rightarrow \infty} \frac{1}{T} \tilde{R}_{T,\alpha}(\tilde{s}_\alpha) = (\tilde{g}(\tilde{s}))_\alpha \quad \text{for all } \alpha \in [0, 1]. \quad (2.26)$$

### 3. One-Dimensional Case

In this section we consider the case of  $n = 1$ , i.e.  $\tilde{r} \in \mathcal{F}(S \times [0, M])$ , and characterize an average fuzzy reward  $\tilde{g}(\tilde{s})$  by the functional equations concerning with the extremal points of its  $\alpha$ -cuts. Throughout this section it is assumed that Assumptions *A* and *B* hold.

Since  $\mathcal{C}([0, M])$  is the set of all closed intervals, we can write the map  $R_\alpha : \mathcal{C}(S) \mapsto \mathcal{C}([0, M])$  by the following notation :

$$R_\alpha(D) := [\min R_\alpha(D), \max R_\alpha(D)] \quad \text{for all } D \in \mathcal{C}(S). \quad (3.1)$$

Let

$$\tilde{R}_{T,\alpha}(D) := \sum_{t=0}^{T-1} R_\alpha(Q_\alpha^t(D)) \quad \text{for } D \in \mathcal{C}(S).$$

Then, by Lemma 2.1(iii), it holds that

$$\min \tilde{R}_{T,\alpha}(D) = \sum_{t=0}^{T-1} \min R_\alpha(Q_\alpha^t(D)) \quad (3.2)$$

and

$$\max \tilde{R}_{T,\alpha}(D) = \sum_{t=0}^{T-1} \max R_\alpha(Q_\alpha^t(D)), \quad (3.3)$$

where

$$\tilde{R}_{T,\alpha}(D) = [\min \tilde{R}_{T,\alpha}(D), \max \tilde{R}_{T,\alpha}(D)].$$

From Lemma 2.4(iii) and Assumption  $B$  we observe that  $R_\alpha(Q_\alpha^t(D))$  converges to  $R_\alpha(\tilde{p}_\alpha)$  exponentially first as  $t \rightarrow \infty$ . Thus, by (3.2) and (3.2),

$$\underline{h}_\alpha(D) := \lim_{T \rightarrow \infty} (\min \tilde{R}_{T,\alpha}(D) - T \times \min R_\alpha(\tilde{p}_\alpha)) \quad (3.4)$$

and

$$\bar{h}_\alpha(D) := \lim_{T \rightarrow \infty} (\max \tilde{R}_{T,\alpha}(D) - T \times \max R_\alpha(\tilde{p}_\alpha)) \quad (3.5)$$

converge for all  $D \in \mathcal{C}(S)$ . The function  $\underline{h}_\alpha$  ( $\bar{h}_\alpha$  resp.) is called a *lower (upper) relative value function*, whose basic ideas are appearing in the theory of Markov decision processes (c.f. [10]). By Theorem 2.1, we have

$$\tilde{g}(\tilde{p})_\alpha = [\min R_\alpha(\tilde{p}_\alpha), \max R_\alpha(\tilde{p}_\alpha)], \quad (3.6)$$

where the extremal points are characterized in the following theorem.

**Theorem 3.1.** *Let  $\alpha \in [0, 1]$ . Then the following (i) and (ii) hold.*

- (i) *Let  $\underline{h}_\alpha$  and  $\bar{h}_\alpha$  be defined by (3.4) and (3.5). Then, the following equations hold:*

$$\underline{h}_\alpha(D) + \min R_\alpha(\tilde{p}_\alpha) = \min R_\alpha(D) + \underline{h}_\alpha(Q_\alpha(D)) \quad (3.7)$$

and

$$\bar{h}_\alpha(D) + \max R_\alpha(\tilde{p}_\alpha) = \max R_\alpha(D) + \bar{h}_\alpha(Q_\alpha(D)) \quad (3.8)$$

for all  $D \in \mathcal{C}(S)$ .

- (ii) *Conversely, if there exist bounded functions  $\underline{h}_\alpha$  and  $\bar{h}_\alpha$  on  $\mathcal{C}(S)$  and constants  $\underline{K}_\alpha$  and  $\bar{K}_\alpha$  satisfying that*

$$\underline{h}_\alpha(D) + \underline{K}_\alpha = \min R_\alpha(D) + \underline{h}_\alpha(Q_\alpha(D)) \quad (3.9)$$

and

$$\bar{h}_\alpha(D) + \bar{K}_\alpha = \max R_\alpha(D) + \bar{h}_\alpha(Q_\alpha(D)) \quad (3.10)$$

for all  $D \in \mathcal{C}(S)$ , then  $\tilde{g}(\tilde{s})_\alpha = [\underline{K}_\alpha, \bar{K}_\alpha]$ .

**Proof.** (i) By the definition of (3.4), it implies

$$\begin{aligned} \underline{h}_\alpha(D) &= \lim_{T \rightarrow \infty} \sum_{t=0}^{T-1} (\min R_\alpha(Q_\alpha^t(D)) - \min R_\alpha(\tilde{p}_\alpha)) \\ &= \min R_\alpha(D) - \min R_\alpha(\tilde{p}_\alpha) \\ &\quad + \sum_{t=1}^{\infty} (\min R_\alpha(Q_\alpha^{t-1}(Q_\alpha(D))) - \min R_\alpha(\tilde{p}_\alpha)) \\ &= \min R_\alpha(D) - \min R_\alpha(\tilde{p}_\alpha) + \underline{h}_\alpha(Q_\alpha(D)), \end{aligned}$$

which leads to (3.7). Also, (3.8) can be shown analogously to (3.7).

- (ii) Let  $\underline{h}_\alpha(D)$  and  $\underline{K}_\alpha$  be as in (3.9). Then, it holds that for each  $t$  ( $t \geq 0$ ),

$$\underline{h}_\alpha(Q_\alpha^t(D)) + \underline{K}_\alpha = \min R_\alpha(Q_\alpha^t(D)) + \underline{h}_\alpha(Q_\alpha^{t+1}(D)). \quad (3.11)$$

By summing (3.11) for  $t = 0, 1, \dots, T-1$ , we get

$$\underline{h}_\alpha(D) + T \times \underline{K}_\alpha = \sum_{t=0}^{T-1} \min R_\alpha(Q_\alpha^t(D)) + \underline{h}_\alpha(Q_\alpha^T(D)).$$

So

$$\underline{K}_\alpha = \lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=0}^{T-1} \min R_\alpha(Q_\alpha^t(D)) \quad \text{for } D \in \mathcal{C}(S).$$

Thus, from Theorem 2.1 and Corollary 2.1,

$$\underline{K}_\alpha = \min R_\alpha(\tilde{p}_\alpha).$$

We also obtain  $\overline{K}_\alpha = \max R_\alpha(\tilde{p}_\alpha)$  similarly. Therefore we get  $\tilde{g}(\tilde{s})_\alpha = [\underline{K}_\alpha, \overline{K}_\alpha]$  by (3.6).  $\square$

Here we give a numerical example to illustrate the theoretical results in this section. Let  $S := [0, 1]$ ,  $M := 1$ . Take the fuzzy relation and the fuzzy reward by

$$\tilde{q}(x, y) = (1 - 3|y - x|/2) \vee 0, \quad x, y \in [0, 1], \quad (3.12)$$

and

$$\tilde{r}(x, z) = (1 - 6|x - z|) \vee 0, \quad x, z \in [0, 1]. \quad (3.13)$$

We observe that  $\tilde{q}$  and  $\tilde{r}$  satisfy Assumptions  $A$  for  $t_0 = 1$ , and  $B$  respectively. Let  $\alpha \in [0, 1]$ . From (2.2) and (2.3),

$$Q_\alpha(\{x\}) = [(x/2 - (1 - \alpha)/3) \vee 0, (x/2 + (1 - \alpha)/3)]$$

for  $x \in [0, 1]$ . So, for  $0 \leq a \leq b \leq 1$ ,

$$Q_\alpha([a, b]) = \bigcup_{x \in [a, b]} Q_\alpha(\{x\}) = [T_1(a), T_2(b)] \quad (3.14)$$

where maps  $T_i; i = 1, 2$  on  $[0, 1]$  are given by  $T_1(x) := (x/2 - (1 - \alpha)/3) \vee 0$ ,  $T_2(x) := x/2 + (1 - \alpha)/3$ . Similarly we have

$$R_\alpha([a, b]) = [(a - (1 - \alpha)/6) \vee 0, (b + (1 - \alpha)/6) \wedge 1]. \quad (3.15)$$

A unique fixed point  $\tilde{p}_\alpha$  of the map  $Q_\alpha : \mathcal{C}([0, 1]) \mapsto \mathcal{C}([0, 1])$  is given as  $\tilde{p}_\alpha = [\min \tilde{p}_\alpha, \max \tilde{p}_\alpha] = [0, 2(1 - \alpha)/3]$ , by solving  $T_1(\min \tilde{p}_\alpha) = \min \tilde{p}_\alpha$  and  $T_2(\max \tilde{p}_\alpha) = \max \tilde{p}_\alpha$  from (3.14). Therefore, from (3.15) and Theorem 2.1, we get  $\tilde{g}(\tilde{s})_\alpha = R_\alpha([\min \tilde{p}_\alpha, \max \tilde{p}_\alpha]) = [0, 5(1 - \alpha)/6]$ . By (2.9), the average fuzzy reward is

$$\tilde{g}(\tilde{s})(x) = \begin{cases} 1 - 6x/5 & 0 \leq x \leq 5/6 \\ 0 & 5/6 < x \leq 1. \end{cases} \quad (3.18)$$

Finally we calculate the lower and the upper relative value functions  $\underline{h}_\alpha$  and  $\bar{h}_\alpha$ . We put

$$Q_\alpha^t([a, b]) = [T_1^t(a), T_2^t(b)] \quad \text{for } 0 \leq a \leq b \leq 1 \text{ and } t \geq 0, \quad (3.19)$$

where maps  $T_i^t; i = 1, 2, (t \geq 0)$  on  $[0, 1]$  are

$$T_i^0(x) = x, \quad T_i^{t+1}(x) = T_i T_i^t(x).$$

Then we can easily check

$$T_1^t(x) = (2^{-t}x - 2(1-\alpha)(1-2^{-t})/3) \vee 0 \quad \text{for } x \in [0, 1]. \quad (3.20)$$

Similarly

$$T_2^t(x) := (2^{-t}x + 2(1-\alpha)(1-2^{-t})/3) \quad \text{for } x \in [0, 1]. \quad (3.21)$$

Let  $0 \leq a \leq b \leq 1$ . From (3.2) and (3.3), we get

$$\min \tilde{R}_{T,\alpha}([a, b]) = \sum_{t=0}^{T-1} \{ (T_1^t(a) - (1-\alpha)/6) \vee 0 \}$$

and

$$\max \tilde{R}_{T,\alpha}([a, b]) = \sum_{t=0}^{T-1} \{ (T_2^t(b) + (1-\alpha)/6) \wedge 1 \}.$$

From the definition of  $\underline{h}_\alpha$  and  $\bar{h}_\alpha$  and (3.17), the lower relative value function is

$$\begin{aligned} \underline{h}_\alpha(a) &:= \underline{h}_\alpha([a, b]) \\ &= \begin{cases} 2(1-2^{-t^*})(a+2(1-\alpha)/3) - 5t^*(1-\alpha)/6 & \alpha < 1 \\ 2a & \alpha = 1, \end{cases} \end{aligned}$$

where  $t^*$  is the smallest non-negative integer such that

$$2^{-t^*}(a+2(1-\alpha)/3) - 5(1-\alpha)/6 < 0.$$

And the upper relative value function is

$$\begin{aligned} \bar{h}_\alpha(b) &:= \bar{h}_\alpha([a, b]) \\ &= \begin{cases} 2b - 4(1-\alpha)/3 & \text{if } 0 \leq b < (5+\alpha)/6 \\ b + (3\alpha-1)/2 & \text{if } (5+\alpha)/6 \leq b \leq 1. \end{cases} \end{aligned}$$

When  $\alpha = 1/2$ , the lower and the upper relative value functions are

$$\underline{h}_\alpha(x) = \begin{cases} 0 & \text{if } 0 \leq x < 1/12 \\ x - 1/12 & \text{if } 1/12 \leq x < 1/2 \\ 3x/2 - 1/3 & \text{if } 1/2 \leq x \leq 1 \end{cases}$$

and

$$\bar{h}_\alpha(x) = \begin{cases} 2x - 2/3 & \text{if } 0 \leq x < 11/12 \\ x + 1/4 & \text{if } 11/12 \leq x \leq 1 \end{cases}$$

We also find that, when  $\alpha = 1$ ,

$$\underline{h}_\alpha(x) = \bar{h}_\alpha(x) = 2x \quad \text{for } 0 \leq x \leq 1.$$

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