



FUZZY NONLINEAR MIXED RANDOM VARIATIONAL-LIKE INCLUSIONS

BYUNG-SOO LEE*, M. FIRDOSH KHAN AND SALAHUDDIN

Abstract: In this paper, we consider a class of randomly $h-\eta$ -maximal monotone mappings and a class of generalized nonlinear mixed random variational-like inclusions for random fuzzy mappings and define an iterative algorithm for finding approximate solutions for the class of variational inclusions. By using the random resolvent operator of randomly $h-\eta$ -maximal monotone mappings, we establish the approximate solutions obtained by our algorithm converge to the exact solutions of the generalized nonlinear mixed random variational-like inclusions for random fuzzy mappings.

Key words: nonlinear mixed random variational-like inclusions, random fuzzy mappings, randomly (h_t, η) maximal monotone mappings, randomly strongly monotone mappings, randomly relaxed Lipschitz continuous mappings, randomly relaxed monotone mappings, Hausdorff metric

Mathematics Subject Classification: 49J40, 47H19

1 Introduction

A variational inclusion is one of the useful and important generations of variational inequalities. It was introduced and considered by Hassouni and Moudafi [20] in 1994, and a perturbed algorithm for finding approximate solutions of the variational inclusions was developed by them. May authors [1, 2, 6, 15, 16, 23, 28, 31, 33, 41] have obtained some important results on variational inclusions with their algorithms to obtain approximate solutions to them in various different assumptions.

A fuzzy set introduced by Zadeh [45] is an extension of a crisp set by enlarging the truth valued set $\{0, 1\}$ to the real unit interval [0, 1]. A fuzzy set is characterized by, and identified with a mapping called a membership-grade function from the whole set into [0, 1]. Heilpern [21] introduced the concept of fuzzy mappings and showed a fixed point theorem for fuzzy contraction mappings which is a fuzzy analogue of Nadler's fixed point theorem for multi-valued mappings. In 1989, Chang and Zhu [11] introduced the concept of variational inequalities with fuzzy mappings and extended some of results of Lassonde [30], Shih and Tan [37], Takahashi [38], Yen [43] in the fuzzy setting. Later, they were developed by Agarwal *et al.* [3], Ahmad *et al.* [4], Ding [12, 13], etc..

On the other hand, random variational inequality problems and random quasi-variational inequality problems have been considered by Chang [7], Chang and Huang [9, 10], Huang [24, 25], Husain *et al.* [27], Tan *et al.* [39], Yuan [44], Khan and Salahuddin [29], Salahuddin [36] and Tan [40], etc..

Copyright © 2010 Yokohama Publishers htt

http://www.ybook.co.jp

^{*}Corresponding author.

In 2003, Fang and Huang [17] introduced a class of H-monotone operators and the resolvent operator associated with the operators, with its Lipschitz continuity. They also considered a class of variational inclusions involving H-monotone operators and constructed an algorithm for solving the class of variational inclusions by using their resolvent operator technique.

At the same time, Huang and Fang [26] introduced a class of maximal η -monotone opertors and defined an associated resolvent operator. Using their resolvent operator methods, they developed some iterative algorithms to approximate the solution of a class of variational inclusions involving maximal η -monotone operators. Huang and Fang's method extended the resolvent operator method associated with an η -subdifferential operator.

In 2005, Fang et al. [18] introduced a new class of (H, η) -monotone operators which unify a framework for a class of maximal monotone operators, a class of maximal η -monotone operators and a class of H-monotone operators, and studied a system of variational inclusions by using the resolvent operators associated with (H, η) -monotone operators in Hilbert spaces.

Very recently, Peng and Zhu [34] introduced and studied one new system of generalized mixed quasi-variational inclusions with (H, η) -monotone operators. By using the resolvent technique for the (H, η) -monotone operators, they proved the existence of solutions for the system of generalized mixed quasi-variational inclusions and the convergence of a new iterative algorithm approximating the solution for the system.

In [35], they also, very recently, introduced and studied another new system of set-valued variational inclusions with (H, η) -monotone operators. By using the resolvent technique for the (H, η) -monotone operators, they showed the existence of solutions for the system, and proved the convergence of a new three-step iterative algorithm approximating the solution for the system.

Basing on the notion of (H, η) -monotonicity for solving a generalized inclusion problem, Verma [42] also developed a generalized framework for the Eckstein-Bertsekas proximal point algorithm.

Our aim of this paper is to introduce and study generalized nonlinear mixed random variational-like inclusions for random fuzzy mappings. By using random resolvent operator technique of randomly (h_t, η) -maximal monotone mappings, we prove the approximate solutions obtained by the iterative algorithm converge to the exact solution of the generalized nonlinear mixed random variational-like inclusions for random fuzzy mappings.

2 Preliminaries

Throughout this paper, (Ω, Σ) is a measurable space with a set Ω and a σ -algebra Σ of subsets of Ω . *H* is a real separable Hilbert space endowed with a norm $\|\cdot\|$ and an inner product $\langle \cdot, \cdot \rangle$. Notations $\mathcal{B}(H)$, 2^H and CB(H) denote the class of Borel σ -fields in *H*, the family of all nonempty subsets of *H*, the family of all nonempty closed bounded subsets of *H*, respectively.

Let $D(\cdot, \cdot)$ represent the Hausdorff metric on CB(H) defined by

$$D(A,B) = \max\left\{\sup_{a \in A} d(a,B), \sup_{b \in B} d(A,b)\right\} \text{ for all } A, B \in CB(H),$$

where

$$d(a,B) = d(B,a) = \inf_{b \in B} ||a - b||$$
 for $a \in A$

Definition 2.1. A mapping $u : \Omega \to H$ is said to be measurable if for any $B \in \mathcal{B}(H)$, $u^{-1}(B) = \{t \in \Omega : u(t) \in B\} \in \Sigma$.

Definition 2.2. A mapping $f: \Omega \times H \to H$ is called a random mapping if for each fixed $x \in H$, a mapping $f(\cdot, x) : \Omega \to H$ is measurable. A random mapping f is said to be continuous if for each fixed $t \in \Omega$, a mapping $f(t, \cdot) : H \to H$ is continuous.

Definition 2.3. A multi-valued mapping $T : \Omega \to 2^H$ is said to be measurable if for any $B \in \mathcal{B}(H), T^{-1}(B) = \{t \in \Omega : T(t) \cap B \neq \emptyset\} \in \Sigma.$

Definition 2.4. A mapping $u : \Omega \to H$ is called a measurable selection of a measurable multi-valued mapping $T : \Omega \to 2^H$, if u is measurable and for any $t \in \Omega$, $u(t) \in T(t)$.

Definition 2.5. A mapping $T: \Omega \times H \to 2^H$ is called a random multi-valued mapping if for each fixed $x \in H$, $T(\cdot, x) : \Omega \to 2^H$ is a measurable multi-valued mapping. A random multi-valued mapping $T: \Omega \times H \to CB(H)$ is said to be *D*-continuous if for each fixed $t \in \Omega$, $T(t, \cdot) : H \to 2^H$ is continuous with respect to the Hausdorff metric *D*.

Let $\mathcal{F}(H)$ be a collection of fuzzy sets over H. A mapping F from Ω into $\mathcal{F}(H)$ is called a fuzzy mapping on H. If F is a fuzzy mapping on H, then for any $t \in \Omega$, F(t) (denoted by F_t) is a fuzzy set on H and $F_t(x)$ is the membership-grade of x in F_t .

Let $A \in \mathcal{F}(H)$, $\alpha \in [0, 1]$, then the set

$$(A)_{\alpha} = \{ x \in H : A(x) \ge \alpha \}$$

is called an α -cut of A.

Definition 2.6. A fuzzy mapping $F : \Omega \to \mathcal{F}(H)$ is said to be measurable, if for any $\alpha \in (0, 1]$, a multi-valued mapping $(F(\cdot))_{\alpha} : \Omega \to 2^{H}$ is measurable.

Definition 2.7. A fuzzy mapping $F : \Omega \times H \to \mathcal{F}(H)$ is called a random fuzzy mapping, if for each fixed $x \in H$, $F(\cdot, x) : \Omega \to \mathcal{F}(H)$ is a measurable fuzzy mapping.

Let $A, T, G : \Omega \times H \to \mathcal{F}(H)$ be random fuzzy mappings satisfying the following condition (*);

(*) there exist functions $\alpha, \beta, \gamma : H \to (0, 1]$ such that $(A_{t,x})_{\alpha(x)}, (T_{t,x})_{\beta(x)}$ and $(G_{t,x})_{\gamma(x)} \in CB(H)$ for all $(t, x) \in \Omega \times H$, where $A_{t,x}$ denotes the value of A at (t, x).

Induce random multi-valued mappings \tilde{A} , \tilde{T} and \tilde{G} from A, T and G, respectively as follows:

$$\tilde{A}: \Omega \times H \to CB(H), \ (t,x) \mapsto (A_{t,x})_{\alpha(x)},$$

$$\tilde{T}: \Omega \times H \to CB(H), \ (t,x) \mapsto (T_{t,x})_{\beta(x)},$$

and

$$\tilde{G}: \Omega \times H \to CB(H), \ (t,x) \mapsto (G_{t,x})_{\gamma(x)} \text{ for all } (t,x) \in \Omega \times H.$$

Let $N : \Omega \times H \times H \times H \to H$ and $\eta : \Omega \times H \times H \to H$ be random mappings. Let $g : \Omega \times H \to H$ be a random mapping with $g(t, x(t)) \cap \text{Dom}\partial\varphi(\cdot, y(t)) \neq \emptyset$ for $t \in H$, $x(t) \in H$ and fixed $y(t) \in H$, where $\partial\varphi$ denotes the subdifferential of a proper, convex and lower semi-continuous functional $\varphi : H \times H \to \mathbb{R} \cup \{+\infty\}$.

Now we consider the following problem;

Find measurable mappings $x, u, v, w : \Omega \to H$ such that for all $t \in \Omega$ and each fixed $y(t) \in H$, $A_{t,x(t)}(u(t)) \ge \alpha(x(t)), T_{t,x(t)}(v(t)) \ge \beta(x(t)), G_{t,x(t)}(w(t)) \ge \gamma(x(t)),$

$$g(t, x(t)) \cap \text{Dom}\,\partial\varphi(\cdot, y(t)) \neq \emptyset$$

and

$$\langle N(t, u(t), v(t), w(t)), \eta(t, y(t), g(t, x(t))) \rangle \geq \varphi(g(t, x(t)), x(t)) - \varphi(x(t), y(t)),$$

called a fuzzy nonlinear mixed random variational-like inclusion (**FNMRVLI**). The set of measurable mappings (x, u, v, w) is called a random solution of **FNMRVLI**.

If $\alpha(x) = \beta(x) = \gamma(x) = 1$, for all $x \in H$, then **FNMRVLI** is reduced to finding measurable mappings $x, u, v, w : \Omega \to H$ such that for $t \in \Omega$ and fixed $y(t) \in H$, $u(t) \in \tilde{A}(t, x(t)), v(t) \in \tilde{T}(t, x(t)), w(t) \in \tilde{G}(t, x(t)), g(t, x(t)) \cap \text{Dom}\partial\varphi(\cdot, y(t)) \neq \emptyset$ and

 $\langle N(t, u(t), v(t), w(t)), \eta(t, y(t), g(t, x(t))) \rangle \geq \varphi(g(t, x(t)), x(t)) - \varphi(x(t), y(t)),$

called a nonlinear mixed random variational-like inclusion. In fact, **FNMRVLI** includes many kind of variational inequalities, quasi-variational inequalities and variational inclusions as well as quasi-variational inclusions in [5, 17, 19, 24, 36] as special cases.

3 Conceptual Background

We recall some useful concepts and results. Throughout this section, $x, y, u, v, w : \Omega \to H$ denote measurable mappings.

Lemma 3.1 ([8]). Let $G : \Omega \times H \to CB(H)$ be a D-continuous random multi-valued mapping. Then for a measurable mapping $u : \Omega \to H$, a multi-valued mapping $G(\cdot, u(\cdot)) : \Omega \to CB(H)$ is measurable.

Lemma 3.2 ([8]). Let $A, T : \Omega \to CB(H)$ be measurable multi-valued mappings and $u : \Omega \to H$ be a measurable selection of A. Then there exists a measurable selection $v : \Omega \to H$ of T such that for all $t \in \Omega$ and $\epsilon > 0$,

$$||u(t) - v(t)|| \le (1 + \epsilon) D(A(t), T(t)).$$

Definition 3.3. Let $x, y : \Omega \to H$ be random mappings and $t \in \Omega$. A random mapping $\eta : \Omega \times H \times H \to H$ is said to be

(i) randomly monotone if

$$\langle x(t) - y(t), \eta(t, x(t), y(t)) \rangle \geq 0$$
, for all $x(t), y(t) \in H$;

- (ii) randomly strictly monotone if η is randomly monotone and the equality holds if and only if x(t) = y(t) for all $t \in \Omega$;
- (iii) randomly α_n -strongly monotone if there exists a function $\alpha_n: \Omega \to (0, \infty)$ such that

$$\langle x(t) - y(t), \eta(t, x(t), y(t)) \rangle \geq \alpha_{\eta}(t) ||x(t) - y(t)||^2$$
, for all $x(t), y(t) \in H$;

(iv) randomly L_{η} -Lipschitz continuous if there exists a function $L_{\eta}: \Omega \to (0, \infty)$ such that

$$\|\eta(t, x(t), y(t))\| \le L_{\eta}(t) \|x(t) - y(t)\|, \text{ for all } x(t), y(t) \in H.$$

Remark 3.4. If $\alpha_{\eta}(t) = 1$ and $L_{\eta}(t) = 1$ for all $t \in \Omega$, then η is called randomly strongly monotone and randomly continuous, respectively.

Definition 3.5. Let $x, y: \Omega \to H$, $h: \Omega \times H \to H$ and $\eta: \Omega \times H \times H \to H$ be random mappings. If we design $h(t, x(t)) = h_t(x(t))$ for all $t \in \Omega$, h_t is said to be

(i) randomly η -monotone if

$$\langle h_t(x(t)) - h_t(y(t)), \eta(t, x(t), y(t)) \rangle \ge 0$$
, for all $x(t), y(t) \in H$;

(ii) randomly η -strictly monotone if h_t is randomly η -monotone and

$$\langle h_t(x(t)) - h_t(y(t)), \eta(t, x(t), y(t)) \rangle = 0$$
 iff $x(t) = y(t)$, for all $t \in \Omega$;

(iii) randomly α_{h_t} - η -strongly monotone if there exists a function $\alpha_{h_t} : \Omega \to (0, \infty)$ such that

$$\langle h_t(x(t)) - h_t(y(t)), \eta(t, x(t), y(t)) \rangle \ge \alpha_{h_t}(t) \|x(t) - y(t)\|^2$$
, for all $x(t), y(t) \in H$;

(iv) randomly L_{h_t} -Lipschitz continuous if there exists a function $L_{h_t} : \Omega \to (0, \infty)$ such that

$$||h_t(x(t)) - h_t(y(t))|| \le L_{h_t}(t)||x(t) - y(t)||, \text{ for all } x(t), y(t) \in H.$$

Remark 3.6. If $h_t = I$, the identity mapping on H, then conditions (i), (ii) and (iii) in Definition 3.5 reduce to (i), (ii) and (iii) in Definition 3.3, respectively.

Definition 3.7. Let $h: \Omega \times H \to H$ and $\eta: \Omega \times H \times H \to H$ be random mappings and $M: H \to 2^H$ be a multi-valued mapping, M is said to be

(i) randomly η -monotone if

 $\langle u(t) - v(t), \eta(t, x(t), y(t)) \rangle \ge 0,$ for all $x(t), y(t) \in H, \ u(t) \in M(x(t)), \ v(t) \in M(y(t));$

(ii) randomly η -maximal monotone if M is randomly η -monotone and

$$(I + \rho(t)M)(H) = H,$$

where $\rho: \Omega \to (0, \infty)$ is a function.

(iii) randomly (h_t, η) -maximal monotone if M is randomly η -monotone and

$$(h_t + \rho(t)M)(H) = H,$$

where $\rho: \Omega \to (0, \infty)$ is a function.

Remark 3.8. For $h_t = I$, the identity mapping, the randomly I- η -maximal monotonicity coincides with the randomly η -maximal monotonicity. If $\eta(t, x(t), y(t)) = x(t) - y(t)$ for all $x(t), y(t) \in H$, the concept of a randomly (h_t, η) -maximal monotone mapping reduces to that of a random mapping, which is called a randomly h_t -monotone mapping.

Definition 3.9. Let $\eta : \Omega \times H \times H \to H$ be a random mapping and $\varphi : H \to \mathbb{R} \cup \{+\infty\}$ be a proper functional, φ is said to be η -subdifferentiable at a point $x(t) \in H$ if there exists a point $f^* \in H$ such that

$$\varphi(y(t)) - \varphi(x(t)) \geq \langle f^*, \eta(t, y(t), x(t)) \rangle$$
 for all $y(t) \in H$

where f^* is called a η -subgradient of φ at x(t). The set of all η -subgradients of φ at x(t) is designed by $\partial_{\eta}\varphi(x(t))$. A multi-valued mapping $\partial_{\eta}\varphi: H \to 2^H$ defined by

$$\partial_{\eta}\varphi(x(t)) = \{ f^* \in H : \varphi(y(t)) - \varphi(x(t)) \ge \langle f^*, \eta(t, y(t), x(t)) \rangle \text{ for all } y(t) \in H \}$$

is called a η -subdifferential of φ at x(t).

Proposition 3.10. Let $\eta : \Omega \times H \times H \to H$ be a randomly continuous and randomly strongly monotone mapping such that $\eta(t, x(t), y(t)) + \eta(t, y(t), x(t)) = 0$ for all $x(t), y(t) \in H$ and for any given $x(t) \in H$, a function

$$h(y(t), u(t)) = \langle x(t) - u(t), \eta(t, y(t), u(t)) \rangle$$

is 0-diagonally quasi-concave in y(t), where $u: \Omega \to (0, \infty)$ is a function. Let $\varphi: H \to \mathbb{R} \cup \{+\infty\}$ be a lower semi-continuous η -subdifferentiable proper functional. Then $\partial_{\eta}\varphi: H \to 2^{H}$ is randomly η -maximal monotone, hence for any $\rho(t) > 0$, $(I + \rho(t)\partial_{\eta}\varphi)(H) = H$.

Remark 3.11. It may be considered that Theorem 2.8 of Ding and Luo [15] is a deterministic case of Proposition 3.10.

Proposition 3.12. Let a mapping $h : \Omega \times H \to H$ be randomly η -strictly monotone and a multi-valued mapping $M : H \to 2^H$ be randomly (h_t, η) -maximal monotone. Then M is randomly η -maximal monotone.

Proof. Since M is randomly η -monotone. From [14] it is sufficient to prove that

$$\langle u(t) - v(t), \eta(t, x(t), y(t)) \rangle \ge 0$$
 for all $(y(t), v(t)) \in Gr(M)$ implies $u(t) \in M(x(t))$,

where $Gr(M) = \{(x(t), u(t)) \in H \times H : u(t) \in M(x(t))\}$ denotes the graph of M.

Suppose that M is not randomly η -maximal monotone, then there exists $(x_0(t), u_0(t)) \notin$ Gr(M) such that

$$\langle u_0(t) - v(t), \eta(t, x_0(t), y(t)) \rangle \ge 0$$
 for all $(y(t), v(t)) \in Gr(M)$.

By assumption, for any $\rho(t) > 0$, $(h_t + \rho(t)M)(H) = H$, there exists $(x_1(t), u_1(t)) \in Gr(M)$ such that

$$h_t(x_1(t)) + \rho(t)u_1(t) = h_t(x_0(t)) + \rho(t)u_0(t).$$

It follows that

$$\rho(t)\langle u_0(t) - u_1(t), \, \eta(t, \, x_0(t), \, x_1(t)) \rangle = - \langle h_t(x_0(t)) - h_t(x_1(t)), \, \eta(t, \, x_0(t), \, x_1(t)) \rangle \geq 0.$$

Since h_t is randomly η -strictly monotone, we must have $x_0(t) = x_1(t)$ and so $u_0(t) = u_1(t)$. Hence $(x_0(t), u_0(t)) \in \operatorname{Gr}(M)$, which is a contradiction. Therefore M is randomly η -maximal monotone.

Theorem 3.13. Let $\eta : \Omega \times H \times H \to H$ be a random mapping. Let a random mapping $h: \Omega \times H \to H$ be randomly η -strictly monotone and a multi-valued mapping $M: H \to 2^H$ be randomly (h_t, η) -maximal monotone. Then for a function $\rho : \Omega \to (0, \infty)$, the inverse mapping $(h_t + \rho(t)M)^{-1} : H \to H$ is single-valued.

Proof. For any $u(t) \in H$, let $x(t), y(t) \in (h_t + \rho(t)M)^{-1}(u(t))$. Then we have

$$u(t) - h_t(x(t)) \in \rho(t)M(x(t))$$

and

$$u(t) - h_t(y(t)) \in \rho(t)M(y(t)).$$

Since M is random η -monotone, we have

$$0 \leq \langle u(t) - h_t(x(t)) - (u(t) - h_t(y(t))), \eta(t, x(t), y(t)) \rangle$$

= $-\langle h_t(x(t)) - h_t(y(t)), \eta(t, x(t), y(t)) \rangle.$

It follows from the randomly η -strict monotonicity of h_t that x(t) = y(t). Therefore $(h_t + \rho(t)M)^{-1}$ is a single-valued mapping.

Definition 3.14. Let $\eta: \Omega \times H \times H \to H$ be a random mapping. Let a random mapping $h: \Omega \times H \to H$ be randomly η -strictly monotone and a multi-valued mapping $M: H \to 2^H$ be randomly (h_t, η) -maximal monotone. Then for a function $\rho: \Omega \to (0, \infty)$, the resolvent operator $R_{h,\rho}^M: H \to H$ of M is defined by

$$R^M_{h_{t,0}}(x(t)) = (h_t + \rho(t)M)^{-1}(x(t)), \text{ for all } x(t) \in H.$$

Remark 3.15. If $\eta : \Omega \times H \times H \to H$ is a random mapping, $h : \Omega \times H \to H$ is a randomly η -strictly monotone mapping and $\varphi : H \times H \to \mathbb{R} \cup \{+\infty\}$ is a functional such that

$$\operatorname{Range}(h_t(\cdot) + \rho(t)\partial_\eta \varphi(\cdot, \cdot)) = H$$

for any measurable function $\rho:\Omega\to(0,\infty),$ then from Proposition 3.12 and Theorem 3.13, we have

$$R_{h_t,\rho}^{\partial_\eta\varphi(\cdot,\cdot)}(x(t)) \ = \ (h_t(\cdot)+\rho(t)\partial_\eta\varphi(\cdot,\cdot))^{-1}(x(t)), \ \text{ for all } x(t)\in H, \ t\in\Omega.$$

The single-valued mapping $R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)} = (h_t(\cdot) + \rho(t)\partial_\eta \varphi(\cdot,\cdot))^{-1}$ is called a random proximal mapping.

Assumption 3.16. A random mapping $\eta: \Omega \times H \times H \to H$ satisfies the condition

$$\eta(t, x(t), y(t)) + \eta(t, y(t), x(t)) = 0$$
, for all $x(t), y(t) \in H$, $t \in \Omega$.

Theorem 3.17. Let $\varphi : H \times H \to \mathbb{R} \cup \{+\infty\}$ be a functional with $Dom\varphi(\cdot, y(t)) \neq \emptyset$ for fixed $y(t) \in H$. Then $t \in \Omega$, $u(t) \in \tilde{A}(t, x(t)), v(t) \in \tilde{T}(t, x(t)), w(t) \in \tilde{G}(t, x(t))$ is a solution set of **FNMRVLI** if and only if $g(t, x(t)) \in \partial_{\eta}\varphi(\cdot, y(t))$ and

$$N(t, u(t), v(t), w(t)) \in \partial_{\eta} \varphi(\cdot, g(t, x(t))).$$

Proof. This directly follows from the definition of η -subdifferential.

 \square

Theorem 3.18. Let $\eta : \Omega \times H \times H \to H$ be randomly L_{η} -Lipschitz continuous, $h : \Omega \times H \to H$ be randomly α_{h_t} - η -strongly monotone and $\partial_{\eta}\varphi : H \times H \to 2^H$ be randomly (h_t, η) -maximal monotone. Then resolvent operator $R_{h_t,\rho}^{\partial_{\eta}\varphi(\cdot,\cdot)}$ of $\partial_{\eta}\varphi$ is randomly $\frac{L_{\eta}}{\alpha_{h_t}}$ -Lipschitz continuous.

Proof. By the definition of the resolvent operator $R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}$ of $\partial_\eta \varphi(\cdot,\cdot)$, for any $x(t), y(t) \in H$

$$R_{h_t,\rho}^{\partial_\eta\varphi(\cdot,\cdot)}(x(t)) = (h_t + \rho(t)\partial_\eta\varphi(\cdot,\cdot))^{-1}(x(t))$$

and

$$R^{\partial_{\eta}\varphi(\cdot,\cdot)}_{h_{t},\rho}(y(t)) = (h_{t} + \rho(t)\partial_{\eta}\varphi(\cdot,\cdot))^{-1}(y(t)).$$

It follows that

$$\frac{1}{\rho(t)}(x(t) - h_t(R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(x(t)))) \in \partial_\eta \varphi(R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(x(t)))$$

and

$$\frac{1}{\rho(t)}(y(t) - h_t(R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(y(t)))) \in \partial_\eta \varphi(R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(y(t))).$$

Since $\partial_{\eta}\varphi(\cdot, \cdot)$ is randomly η -monotone, we have

$$\begin{aligned} \langle x(t) - h_t(R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(x(t))) - (y(t) - h_t(R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(y(t)))), \\ \\ \eta(t, (R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(x(t))), (R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(y(t)))) \rangle \geq 0. \end{aligned}$$

It follows that

$$\begin{aligned} \langle x(t) - y(t) - (h_t(R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(x(t))) - h_t(R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(y(t)))), \\\\ \eta(t, (R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(x(t))), (R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,\cdot)}(y(t)))) \rangle &\geq 0 \end{aligned}$$

or

$$\begin{aligned} \langle x(t) - y(t), \eta(t, (R_{h_{t},\rho}^{\partial_{\eta}\varphi(\cdot,\cdot)}(x(t))), (R_{h_{t},\rho}^{\partial_{\eta}\varphi(\cdot,\cdot)}(y(t)))) \rangle &\geq \\ \langle h_{t}(R_{h_{t},\rho}^{\partial_{\eta}\varphi(\cdot,\cdot)}(x(t))) - h_{t}(R_{h_{t},\rho}^{\partial_{\eta}\varphi(\cdot,\cdot)}(y(t))), \eta(t, (R_{h_{t},\rho}^{\partial_{\eta}\varphi(\cdot,\cdot)}(x(t))), (R_{h_{t},\rho}^{\partial_{\eta}\varphi(\cdot,\cdot)}(y(t)))) \rangle \end{aligned}$$

Since η is randomly $L_\eta\text{-Lipschitz}$ continuous and h_t is randomly $\alpha_{h_t}\text{-}\eta\text{-strongly}$ monotone, we have

$$L_{\eta}(t) \|x(t) - y(t)\| \|R_{h_{t},\rho}^{\partial_{\eta}\varphi(\cdot,\cdot)}(x(t)) - R_{h_{t},\rho}^{\partial_{\eta}\varphi(\cdot,\cdot)}(y(t))\|$$

$$\geq \alpha_{h_{t}}(t) \|R_{h_{t},\rho}^{\partial_{\eta}\varphi(\cdot,\cdot)}(x(t)) - R_{h_{t},\rho}^{\partial_{\eta}\varphi(\cdot,\cdot)}(y(t))\|^{2}.$$

Hence

$$\|R_{h_t,\rho}^{\partial_\eta\varphi(\cdot,\cdot)}(x(t)) - R_{h_t,\rho}^{\partial_\eta\varphi(\cdot,\cdot)}(y(t))\| \leq \frac{L_\eta(t)}{\alpha_{h_t}(t)} \|x(t) - y(t)\|,$$

for all $x(t), y(t) \in H, t \in \Omega$.

4 Iterative Algorithm

We first give the following lemma.

Lemma 4.1. The set of measurable mappings $x, u, v, w : \Omega \to H$ is a random solution of **FNMRVLI** if and only if for all $t \in \Omega$, $x(t) \in H$, $u(t) \in \tilde{A}(t, x(t))$, $v(t) \in \tilde{T}(t, x(t))$, $w(t) \in \tilde{G}(t, x(t))$ and

$$g(t, x(t)) = R_{h_t, \rho}^{\partial_\eta \varphi(\cdot, x(t))} [g(t, x(t)) - \rho(t) N(t, u(t), v(t), w(t))],$$
(4.1)

where $\rho: \Omega \to (0,\infty)$ is a measurable function and $R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,x(t))} = (h_t(x(t)) + \rho(t)\partial_\eta \varphi(\cdot,x(t)))^{-1}$.

To obtain an approximate solution of **FNMRVLI**, we can apply a successive approximate method to the problem of solving

$$x(t) \in Q(t, x(t)), \text{ for all } t \in \Omega,$$

where

$$Q(t, x(t)) = \left\{ x(t) - g(t, x(t)) + R_{h_t, \rho}^{\partial_\eta \varphi(\cdot, x(t))} [g(t, x(t)) - \rho(t) N(t, u(t), v(t), w(t))] \\ : u(t) \in \tilde{A}(t, x(t)), v(t) \in \tilde{T}(t, x(t)), w(t) \in \tilde{G}(t, x(t)) \right\}.$$

$$(4.2)$$

Based on (4.1) and (4.2), we propose the following random iterative algorithm to compute the approximate solution of **FNMRVLI**.

Algorithmorem 4.2. Suppose that $A, T, G : \Omega \times H \to \mathcal{F}(H)$ be random fuzzy mappings satisfying the condition (*). Let $\tilde{A}, \tilde{T}, \tilde{G} : \Omega \times H \to CB(H)$ be *D*-continuous random multivalued mappings induced by A, T and G, respectively and $g : \Omega \times H \to H$ be a continuous random mapping. Let $\eta : \Omega \times H \times H \to H, N : \Omega \times H \times H \times H \to H$ and $h : \Omega \times H \to H$ be random mappings. For any given measurable mapping $x_0 : \Omega \to H$, multi-valued mappings, $\tilde{A}(\cdot, x_0(\cdot)), \tilde{T}(\cdot, x_0(\cdot)), \tilde{G}(\cdot, x_0(\cdot)) : \Omega \to CB(H)$ are measurable by Lemma 3.1. Hence there exist selections $u_0 : \Omega \to H$ of $\tilde{A}(\cdot, x_0(\cdot)), v_0 : \Omega \to H$ of $\tilde{T}(\cdot, x_0(\cdot))$ and $w_0 : \Omega \to H$ of $\tilde{G}(\cdot, x_0(\cdot))$ by Himmelberg [22]. Let

$$x_1(t) = x_0(t) - g(t, x_0(t)) + R_{h_t, \rho}^{\partial_\eta \varphi(\cdot, x_0(t))} [g(t, x_0(t) - \rho(t)N(t, u_0(t), v_0(t), w_0(t)],$$

then is easy to see that $x_1: \Omega \to H$ is measurable. By Lemma 3.2 and Nadler [32] there exist measurable selections $u_1: \Omega \to H$ of $\tilde{A}(\cdot, x_1(\cdot)), v_1: \Omega \to H$ of $\tilde{T}(\cdot, x_1(\cdot))$ and $w_1: \Omega \to H$ of $\tilde{G}(\cdot, x_1(\cdot))$ such that

$$\begin{aligned} \|u_0(t) - u_1(t)\| &\leq (1+1)D(\tilde{A}(t, x_0(t)), \tilde{A}(t, x_1(t))), \\ \|v_0(t) - v_1(t)\| &\leq (1+1)D(\tilde{T}(t, x_0(t)), \tilde{T}(t, x_1(t))), \\ \|w_0(t) - w_1(t)\| &\leq (1+1)D(\tilde{G}(t, x_0(t)), \tilde{G}(t, x_1(t))). \end{aligned}$$

Let

$$x_2(t) = x_1(t) - g(t, x_1(t)) + R_{h_t, \rho}^{\partial_\eta \varphi(\cdot, x_1(t))}[g(t, x_1(t)) - \rho(t)N(t, u_1(t), v_1(t), w_1(t))].$$

Then $x_2 : \Omega \to H$ is measurable. Continuing the above process inductively we can obtain the following random iterative sequences $\{x_n(t)\}$ of measurable mappings and three

sequences $\{u_n(t)\}, \{v_n(t)\}\$ and $\{w_n(t)\}\$ of measurable selections for solving **FNMRVLI** as follows:

$$\begin{aligned} x_{n+1}(t) &= x_n(t) - g(t, x_n(t)) + R_{h_t, \rho}^{\partial_\eta \varphi(\cdot, x_n(t))}[g(t, x_n(t)) - \rho(t)N(t, u_n(t), v_n(t), w_n(t))]. \quad (4.3) \\ u_n(t) &\in \tilde{A}(t, x_n(t)), \ \|u_n(t) - u_{n+1}(t)\| \leq (1 + (1 + n)^{-1})D(\tilde{A}(t, x_n(t)), \tilde{A}(t, x_{n+1}(t))), \\ v_n(t) &\in \tilde{T}(t, x_n(t)), \ \|v_n(t) - v_{n+1}(t)\| \leq (1 + (1 + n)^{-1})D(\tilde{T}(t, x_n(t)), \tilde{T}(t, x_{n+1}(t))), \\ w_n(t) &\in \tilde{G}(t, x_n(t)), \ \|w_n(t) - w_{n+1}(t)\| \leq (1 + (1 + n)^{-1})D(\tilde{G}(t, x_n(t)), \tilde{G}(t, x_{n+1}(t))) \\ \text{for any } t \in \Omega \text{ and } n = 0, 1, 2, \dots. \end{aligned}$$

Definition 4.3. A random mapping $g: \Omega \times H \to H$ is said to be

(i) randomly r_g -strongly monotone, if there exists a measurable function $r_g: \Omega \to (0, \infty)$ such that

$$\langle g(t, x(t)) - g(t, y(t)), x(t) - y(t) \rangle \ge r_g(t) \|x(t) - y(t)\|^2, \text{ for } x(t), y(t) \in H;$$

(ii) randomly s_g -Lipschitz continuous, if there exists a measurable function $s_g : \Omega \to (0, \infty)$ such that

$$||g(t, x(t)) - g(t, y(t))|| \le s_q(t) ||x(t) - y(t)||^2$$
, for $x(t), y(t) \in H$.

Definition 4.4. Let $N : \Omega \times H \times H \times H \to H$ be a random mapping and $\tilde{A}, \tilde{T}, \tilde{G} : \Omega \times H \to CB(H)$ random multi-valued mappings:

(i) N is said to be randomly $\lambda_{\tilde{T}}$ -relaxed monotone with respect to the second argument for the mapping \tilde{T} , if there exists a measurable function $\lambda_{\tilde{T}} : \Omega \to (0, \infty)$ such that

$$\langle N(t,\cdot,v_1(t),\cdot) - N(t,\cdot,v_2(t),\cdot),x_1(t) - x_2(t) \rangle \geq -\lambda_{\tilde{T}}(t) \|x_1(t) - x_2(t)\|^2$$

for all $x_i(t) \in H$, $v_i(t) \in \tilde{T}(t, x_i(t))$, $t \in \Omega$, i = 1, 2.

(ii) N is said to be randomly $\zeta_{\tilde{G}}$ -relaxed Lipschitz continuous with respect to the third argument for the mapping \tilde{G} , if there exists a measurable function $\zeta_{\tilde{G}} : \Omega \to (0, \infty)$ such that

$$\langle N(t, \cdot, \cdot, w_1(t)) - N(t, \cdot, \cdot, w_2(t)), x_1(t) - x_2(t) \rangle \leq -\zeta_{\tilde{G}}(t) \|x_1(t) - x_2(t)\|^2,$$

for all $x_i(t) \in H$, $w_i(t) \in \tilde{G}(t, x_i(t))$, $t \in \Omega$, i = 1, 2.

(iii) N is said to be randomly Lipschitz continuous with respect to the first, the second and the third arguments, if there exist measurable functions d_N , ϵ_N , $p_N : \Omega \to (0, \infty)$ such that

$$||N(t, x_1(t), x_2(t), x_3(t)) - N(t, y_1(t), y_2(t), y_3(t))||$$

 $\leq d_N(t) \|x_1(t) - y_1(t)\| + \epsilon_N(t) \|x_2(t) - y_2(t)\| + p_N(t) \|x_3(t) - y_3(t)\|$

for $x_i(t)$, $y_i(t) \in H$, i = 1, 2, 3 and $t \in \Omega$.

(iv) N is said to be randomly $\xi_{\tilde{A},g}$ -strongly monotone with respect to the first argument for the mapping \tilde{A} if there exists a measurable function $\xi_{\tilde{A},g}: \Omega \to \infty$ such that

$$\langle N(t, u_1(t), \cdot, \cdot) - N(t, u_2(t), \cdot, \cdot), g(t, x_1(t)) - g(t, x_2(t)) \rangle \geq \xi_{\tilde{A}, q}(t) \|x_1(t) - x_2(t)\|^2$$

for $x_i(t) \in H$, $u_i(t) \in \tilde{A}(t, x_i(t))$, i = 1, 2, where $g : \Omega \times H \to H$ is a random mapping.

(v) Mappings $\tilde{A}, \tilde{T}, \tilde{G}$ are said to be randomly *D*-Lipschitz continuous, if there exist measurable functions $\beta_{\tilde{A}}, \gamma_{\tilde{T}}, \sigma_{\tilde{G}} : \Omega \to (0, \infty)$ such that

$$\begin{aligned} \|u_1(t) - u_2(t)\| &\leq D(A(t, x_1(t)), A(t, x_2(t))) \leq \beta_{\tilde{A}}(t) \|x_1(t) - x_2(t)\|, \\ \|v_1(t) - v_2(t)\| &\leq D(\tilde{T}(t, x_1(t)), \tilde{T}(t, x_2(t))) \leq \gamma_{\tilde{T}}(t) \|x_1(t) - x_2(t)\|, \\ \|w_1(t) - w_2(t)\| &\leq D(\tilde{G}(t, x_1(t)), \tilde{G}(t, x_2(t))) \leq \sigma_{\tilde{G}}(t) \|x_1(t) - x_2(t)\| \end{aligned}$$

for $x_i(t) \in H$, $u_i(t) \in \tilde{A}(t, x_i(t))$, $v_i(t) \in \tilde{T}(t, x_i(t))$, $w_i(t) \in \tilde{G}(t, x_i(t))$, $t \in \Omega$, i = 1, 2.

Theorem 4.5. Let a random mapping $\eta : \Omega \times H \times H \to H$ be randomly L_{η} -Lipschitz continuous and a random mapping $h : \Omega \times H \to H$ be randomly α_{h_t} - η -strongly monotone. Let $N : \Omega \times H \times H \times H \to H$ be a random mapping which is randomly Lipschitz continuous with respect to the first, the second and the third arguments with random coefficients $d_N(t)$, $\epsilon_N(t)$ and $p_N(t)$, respectively. Let $A, T, G : \Omega \times H \to \mathcal{F}(H)$ be random fuzzy mappings satisfying the condition (*). Let $\tilde{A}, \tilde{T}, \tilde{G} : \Omega \times H \to CB(H)$ be random multi-valued mappings induced by A, T and G, respectively, which are randomly D-Lipschitz continuous with random coefficients $\beta_{\tilde{A}}(t), \gamma_{\tilde{T}}(t)$ and $\sigma_{\tilde{G}}(t)$, respectively. Let $g : \Omega \times H \to H$ be randomly r_g -strongly monotone and randomly s_g -Lipschitz continuous. Let N be randomly $\lambda_{\tilde{T}}$ -relaxed monotone with respect to the second argument for the mapping \tilde{T} and randomly $\zeta_{\tilde{G}}$ -relaxed Lipschitz continuous with respect to the third argument for the mapping \tilde{G} . Let N be randomly $\xi_{\tilde{A},g}$ strongly monotone with respect to the first argument for the mapping \tilde{A} . Let $\varphi : H \times H \to$ $\mathbb{R} \cup \{+\infty\}$ be a functional such that for fixed $x(t) \in H$, $\operatorname{Ran}(h_t(x(t)) + \rho(t)\partial_\eta \varphi(\cdot, x(t))) = H$, where $\rho : \Omega \to (0, \infty)$ is a measurable function.

Assume that

$$\|R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,x(t))}(z(t)) - R_{h_t,\rho}^{\partial_\eta \varphi(\cdot,y(t))}(z(t))\| \le \delta(t) \|x(t) - y(t)\| \text{ for } x(t), y(t) \text{ and } z(t) \in H,$$

where $\delta: \Omega \to (0, \infty)$ is a function and the following conditions hold;

$$\left| \rho(t) - \frac{(\xi_{\tilde{A},g}(t)s_{g}^{2}(t)L_{\eta}(t) - \chi(t)\alpha_{h_{t}}(t)(1-\kappa(t)))}{L_{\eta}(t)(d_{N}^{2}(t)\beta_{\tilde{A}}^{2}(t) - \chi^{2}(t))} \right|$$

$$< \frac{\sqrt{(\xi_{\tilde{A},g}(t)s_{g}^{2}(t)L_{\eta}(t) - \chi(t)\alpha_{h_{t}}(t)(1-\kappa(t)))^{2} - (d_{N}^{2}(t)\beta_{\tilde{A}}^{2}(t) - \chi^{2}(t))(L_{\eta}^{2}(t)s_{g}^{2}(t) - \alpha_{h_{t}}^{2}(t)(1-\kappa(t))^{2})}{L_{\eta}^{2}(t)(d_{N}^{2}(t)\beta_{\tilde{A}}^{2}(t) - \chi^{2}(t))}$$

$$\xi_{\tilde{A},g}(t)s_{g}^{2}(t)L_{\eta}(t) > \chi(t)\alpha_{h_{t}}(t)(1-\kappa(t))^{2} + \sqrt{(d_{N}^{2}(t)\beta_{\tilde{A}}^{2}(t) - \chi^{2}(t))(L_{\eta}^{2}(t)s_{g}^{2}(t) - \alpha_{h_{t}}^{2}(t)(1-\kappa(t))^{2})}$$

$$\xi_{\tilde{A},g}(t)s_{g}^{2}(t)L_{\eta}(t) > \chi(t)\alpha_{h_{t}}(t)(1-\kappa(t))^{2}, \quad \kappa(t) < 1, \quad \chi(t) < d_{N}^{2}(t)\beta_{\tilde{A}}^{2}(t),$$

$$(4.4)$$

$$\alpha_{h_t}(t)(1-\kappa(t)) < L_\eta(t)s_g^2(t) \text{ and } \chi(t) = \Delta + \ell,$$

where $\kappa(t) = \sqrt{1 - 2r_g(t) + s_g^2(t)} + \delta(t)$,

$$\Delta = \sqrt{1 - 2\zeta_{\tilde{G}}(t) + \sigma_{\tilde{G}}^2(t)p_N^2(t)} \text{ and } \ell = \sqrt{1 - 2\lambda_{\tilde{T}}(t) + \epsilon_N^2(t)\gamma_{\tilde{T}}^2(t)}.$$

Then there exist measurable mappings $x, u, v, w : \Omega \to H$ such that **FNMRVLI** holds. Moreover $x_n(t) \to x(t), u_n(t) \to u(t), v_n(t) \to v(t)$ and $w_n(t) \to w(t)$ in H, where $\{x_n(t)\}, \{u_n(t)\}, \{v_n(t)\}$ and $\{w_n(t)\}$ are random sequences obtained by Algorithm 4.2.

Proof. From (4.3), for any $t \in \Omega$, we have

$$\begin{aligned} \|x_{n+1}(t) - x_n(t)\| &= \|x_n(t) - x_{n-1}(t) - (g(t, x_n(t)) - g(t, x_{n-1}(t))) \\ &+ R_{h_t,\rho}^{\partial_\eta \varphi(\cdot, x_n(t))}(z_n(t)) - R_{h_t,\rho}^{\partial_\eta \varphi(\cdot, x_{n-1}(t))}(z_{n-1}(t)) \| \\ &\leq \|x_n(t) - x_{n-1}(t) - (g(t, x_n(t)) - g(t, x_{n-1}(t)))\| \\ &+ \|R_{h_t,\rho}^{\partial_\eta \varphi(\cdot, x_n(t))}(z_n(t)) - R_{h_t,\rho}^{\partial_\eta \varphi(\cdot, x_{n-1}(t))}(z_{n-1}(t))\| \\ &\leq \|x_n(t) - x_{n-1}(t) - (g(t, x_n(t)) - g(t, x_{n-1}(t)))\| \\ &+ \|R_{h_t,\rho}^{\partial_\eta \varphi(\cdot, x_n(t))}(z_{n-1}(t)) - R_{h_t,\rho}^{\partial_\eta \varphi(\cdot, x_{n-1}(t))}(z_{n-1}(t))\| \\ &+ \|R_{h_t,\rho}^{\partial_\eta \varphi(\cdot, x_n(t))}(z_{n-1}(t)) - R_{h_t,\rho}^{\partial_\eta \varphi(\cdot, x_{n-1}(t))}(z_{n-1}(t))\| \\ &\leq \|x_n(t) - x_{n-1}(t) - (g(t, x_n(t)) - g(t, x_{n-1}(t)))\| \\ &\leq \|x_n(t) - x_{n-1}(t) - (g(t, x_n(t)) - g(t, x_{n-1}(t)))\| \\ &+ \frac{L_\eta(t)}{\alpha_{h_t}(t)} \|z_n(t) - z_{n-1}(t)\| + \delta(t) \|x_n(t) - x_{n-1}(t)\|, \end{aligned}$$

where

$$z_n(t) = g(t, x_n(t)) - \rho(t)N(t, u_n(t), v_n(t), w_n(t)).$$

Now

$$\begin{aligned} \|z_{n}(t) - z_{n-1}(t)\| \\ = \|g(t, x_{n}(t)) - \rho(t)N(t, u_{n}(t), v_{n}(t), w_{n}(t)) \\ &- g(t, x_{n-1}(t)) + \rho(t)N(t, u_{n-1}(t), v_{n-1}(t), w_{n-1}(t))\| \\ \leq \|g(t, x_{n}(t)) - g(t, x_{n-1}(t)) - \rho(t)(N(t, u_{n}(t), v_{n}(t), w_{n}(t)) - N(t, u_{n-1}(t), v_{n}(t), w_{n}(t)))\| \\ &+ \rho(t)\|x_{n}(t) - x_{n-1}(t) + (N(t, u_{n-1}(t), v_{n}(t), w_{n}(t)) - N(t, u_{n-1}(t), v_{n-1}(t), w_{n}(t)))\| \\ &+ \rho(t)\|x_{n}(t) - x_{n-1}(t) - (N(t, u_{n-1}(t), v_{n-1}(t), w_{n}(t)) - N(t, u_{n-1}(t), v_{n-1}(t), w_{n-1}(t)))\| \\ &+ (4.6) \end{aligned}$$

Adding (4.5) and (4.6), we get

$$\begin{aligned} \|x_{n+1}(t) - x_n(t)\| \\ \leq \|x_n(t) - x_{n-1}(t) - (g(t, x_n(t)) - g(t, x_{n-1}(t)))\| \\ &+ \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} [\|g(t, x_n(t)) - g(t, x_{n-1}(t)) - \rho(t)(N(t, u_n(t), v_n(t), w_n(t))) \\ &- N(t, u_{n-1}(t), v_n(t), w_n(t)))\| + \rho(t)\|x_n(t) - x_{n-1}(t) + (N(t, u_{n-1}(t), v_n(t), w_n(t))) \\ &- N(t, u_{n-1}(t), v_{n-1}(t), w_n(t)))\| + \rho(t)\|x_n(t) - x_{n-1}(t) - (N(t, u_{n-1}(t), v_{n-1}(t), w_n(t))) \\ &- N(t, u_{n-1}(t), v_{n-1}(t), w_{n-1}(t)))\| + \delta(t)\|x_n(t) - x_{n-1}(t)\|. \end{aligned}$$

By the random s_g -Lipschitz continuity and the random r_g -strong monotonicity of g, we have

$$\begin{aligned} \|x_{n}(t) - x_{n-1}(t) - (g(t, x_{n}(t)) - g(t, x_{n-1}(t)))\|^{2} \\ = \|x_{n}(t) - x_{n-1}(t)\|^{2} - 2\langle g(t, x_{n}(t)) - g(t, x_{n-1}(t)), x_{n}(t) - x_{n-1}(t)\rangle \\ + \|g(t, x_{n}(t)) - g(t, x_{n-1}(t))\|^{2} \\ \leq \|x_{n}(t) - x_{n-1}(t)\|^{2} - 2r_{g}(t)\|x_{n}(t) - x_{n-1}(t)\|^{2} + s_{g}^{2}(t)\|x_{n}(t) - x_{n-1}(t)\|^{2} \\ \leq (1 - 2r_{g}(t) + s_{g}^{2}(t))\|x_{n}(t) - x_{n-1}(t)\|^{2}. \end{aligned}$$

$$(4.8)$$

Since N is randomly Lipschitz continuous with measurable mappings $d_N(\cdot)$, $\epsilon_N(\cdot)$, $p_N(\cdot)$: $\Omega \to (0,\infty)$ and $\tilde{A}, \tilde{T}, \tilde{G}$ are randomly D-Lipschitz continuous with mappings $\beta_{\tilde{A}}(\cdot), \gamma_{\tilde{T}}(\cdot), \sigma_{\tilde{G}}(\cdot): \Omega \to (0,\infty)$, respectively, we have

$$\begin{aligned} \|N(t, u_n(t), v_n(t), w_n(t)) - N(t, u_{n-1}(t), v_n(t), w_n(t))\| \\ &\leq d_N(t) \|u_n(t) - u_{n-1}(t)\| \\ &\leq d_N(t) D(\tilde{A}(t, x_n(t)), \tilde{A}(t, x_{n-1}(t))) \\ &\leq d_N(t) \beta_{\tilde{A}}(t) (1 + n^{-1}) \|x_n(t) - x_{n-1}(t)\|, \end{aligned}$$

$$(4.9)$$

$$\|N(t, u_{n-1}(t), v_n(t), w_n(t)) - N(t, u_{n-1}(t), v_{n-1}(t), w_n(t))\|$$

$$\leq \epsilon_N(t) \|v_n(t) - v_{n-1}(t)\|$$

$$\leq \epsilon_N(t) D(\tilde{T}(t, x_n(t)), \tilde{T}(t, x_{n-1}(t)))$$

$$\leq \epsilon_N(t) \gamma_{\tilde{T}}(t) (1 + n^{-1}) \|x_n(t) - x_{n-1}(t)\|,$$
(4.10)

and

$$||N(t, u_{n-1}(t), v_{n-1}(t), w_n(t)) - N(t, u_{n-1}(t), v_{n-1}(t), w_{n-1}(t))||$$

$$\leq p_N(t) ||w_n(t) - w_{n-1}(t)||$$

$$\leq p_N(t) O(\tilde{G}(t, x_n(t)), \tilde{G}(t, x_{n-1}(t)))$$

$$\leq p_N(t) \sigma_{\tilde{G}}(t) (1 + n^{-1}) ||x_n(t) - x_{n-1}(t)||.$$
(4.11)

Since N is randomly $\lambda_{\tilde{T}}$ -relaxed monotone with respect to the second argument, from (4.10), we have

$$\begin{aligned} \|x_{n}(t) - x_{n-1}(t) + (N(t, u_{n-1}(t), v_{n}(t), w_{n}(t)) - N(t, u_{n-1}(t), v_{n-1}(t), w_{n}(t)))\|^{2} \\ &\leq \|x_{n}(t) - x_{n-1}(t)\|^{2} \\ &+ 2\langle N(t, u_{n-1}(t), v_{n}(t), w_{n}(t)) - N(t, u_{n-1}(t), v_{n-1}(t), w_{n}(t)), x_{n}(t) - x_{n-1}(t)\rangle \\ &+ \|N(t, u_{n-1}(t), v_{n}(t), w_{n}(t)) - N(t, u_{n-1}(t), v_{n-1}(t), w_{n}(t))\|^{2} \\ &\leq \|x_{n}(t) - x_{n-1}(t)\|^{2} - 2\lambda_{\tilde{T}}(t)\|x_{n}(t) - x_{n-1}(t)\|^{2} \\ &+ \epsilon_{N}^{2}(t)\gamma_{\tilde{T}}^{2}(t)(1 + n^{-1})^{2}\|x_{n}(t) - x_{n-1}(t)\|^{2} \\ &= (1 - 2\lambda_{\tilde{T}}(t) + \epsilon_{N}^{2}(t)\gamma_{\tilde{T}}^{2}(t)(1 + n^{-1})^{2})\|x_{n}(t) - x_{n-1}(t)\|^{2} . \end{aligned}$$

$$(4.12)$$

Since N is randomly $\zeta_{\tilde{G}}\text{-relaxed Lipschitz continuous with respect to the third argument,$

from (4.11), we obtain

$$\begin{aligned} \|x_{n}(t) - x_{n-1}(t) - (N(t, u_{n-1}(t), v_{n-1}(t), w_{n}(t)) - N(t, u_{n-1}(t), v_{n-1}(t), w_{n-1}(t)))\|^{2} \\ &\leq \|x_{n}(t) - x_{n-1}(t)\|^{2} - 2\langle N(t, u_{n-1}(t), v_{n-1}(t), w_{n}(t)) - N(t, u_{n-1}(t), v_{n-1}(t), w_{n-1}(t)), \\ &x_{n}(t) - x_{n-1}(t)\rangle + \|N(t, u_{n-1}(t), v_{n-1}(t), w_{n}(t)) - N(t, u_{n-1}(t), v_{n-1}(t), w_{n-1}(t))\|^{2} \\ &\leq \|x_{n}(t) - x_{n-1}(t)\|^{2} + 2\zeta_{\tilde{G}}(t)\|x_{n}(t) - x_{n-1}(t)\|^{2} + \sigma_{\tilde{G}}^{2}(t)p_{N}^{2}(t)(1 + n^{-1})^{2}\|x_{n}(t) - x_{n-1}(t)\|^{2} \\ &= (1 + 2\zeta_{\tilde{G}}(t) + \sigma_{\tilde{G}}^{2}(t)p_{N}^{2}(t))(1 + n^{-1})^{2}\|x_{n}(t) - x_{n-1}(t)\|^{2} . \end{aligned}$$

$$(4.13)$$

Again, since N is randomly $\xi_{\tilde{A},g}$ -strongly monotone with respect to the first argument of the mapping \tilde{A} , from (4.9), we obtain

$$\begin{aligned} \|g(t,x_{n}(t)) - g(t,x_{n-1}(t)) - \rho(t)(N(t,u_{n}(t),v_{n}(t),w_{n}(t)) - N(t,u_{n-1}(t),v_{n}(t),w_{n}(t)))\|^{2} \\ &\leq \|g(t,x_{n}(t)) - g(t,x_{n-1}(t))\|^{2} \\ &- 2\rho(t)\langle N(t,u_{n}(t),v_{n}(t),w_{n}(t)) - N(t,u_{n-1}(t),v_{n}(t),w_{n}(t)),g(t,x_{n}(t)) - g(t,x_{n-1}(t))\rangle \\ &+ \rho^{2}(t)\|N(t,u_{n}(t),v_{n}(t),w_{n}(t)) - N(t,u_{n-1}(t),v_{n}(t),w_{n}(t))\|^{2} \\ &\leq s_{g}^{2}(t)\|x_{n}(t) - x_{n-1}(t)\|^{2} - 2\rho(t)\xi_{\tilde{A},g}(t)\|g(t,x_{n}(t)) - g(t,x_{n-1}(t))\|^{2} \\ &+ \rho^{2}(t)d_{N}^{2}(t)\beta_{\tilde{A}}^{2}(t)(1+n^{-1})^{2}\|x_{n}(t) - x_{n-1}(t)\|^{2} \\ &\leq (s_{g}^{2}(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_{g}^{2}(t) + \rho^{2}(t)d_{N}^{2}(t)\beta_{\tilde{A}}^{2}(t)(1+n^{-1})^{2})\|x_{n}(t) - x_{n-1}(t)\|^{2} . \end{aligned}$$
(4.14)

Combining (4.7)-(4.14), we obtain

$$\begin{split} &\|x_{n+1}(t) - x_n(t)\| \\ \leq & \left[\sqrt{1 - 2r_g(t) + s_g^2(t)} + \frac{L_\eta(t)}{\alpha_{h_t}(t)} \left\{ \sqrt{(s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)(1 + n^{-1})^2} \right. \\ & \left. + \rho(t)\sqrt{1 - 2\lambda_{\tilde{T}}(t) + \epsilon_N^2(t)\gamma_{\tilde{T}}^2(t)(1 + n^{-1})^2} \right. \\ & \left. + \rho(t)\sqrt{1 + 2\zeta_{\tilde{G}}(t) + \sigma_{\tilde{G}}^2(t)p_N^2(t)(1 + n^{-1})^2} \right\} \right] \|x_n(t) - x_{n-1}(t)\|, \end{split}$$

$$\begin{split} \|x_{n}(t) - x_{n-1}(t)\| + \delta(t)\|x_{n}(t) - x_{n-1}(t)\| \\ &\leq \left[\sqrt{1 - 2r_{g}(t) + s_{g}^{2}(t)} + \frac{L_{\eta}(t)}{\alpha_{h_{t}}(t)} \left\{\sqrt{(s_{g}^{2}(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_{g}^{2}(t) + \rho^{2}(t)d_{N}^{2}(t)\beta_{\tilde{A}}^{2}(t)(1 + n^{-1})^{2}} \right. \\ &+ \rho(t)\sqrt{1 - 2\lambda_{\tilde{T}}(t) + \epsilon_{N}^{2}(t)\gamma_{\tilde{T}}^{2}(t)(1 + n^{-1})^{2}} \\ &+ \rho(t)\sqrt{1 + 2\zeta_{\tilde{G}}(t) + \sigma_{\tilde{G}}^{2}(t)p_{N}^{2}(t)(1 + n^{-1})^{2}} \right\} + \delta(t) \right] \|x_{n}(t) - x_{n-1}(t)\| \\ &\leq \left[\kappa(t) + \frac{L_{\eta}(t)}{\alpha_{h_{t}}(t)} \left(\sqrt{(s_{g}^{2}(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_{g}^{2}(t) + \rho^{2}(t)d_{N}^{2}(t)\beta_{\tilde{A}}^{2}(t)(1 + n^{-1})^{2}} \right. \\ &+ \rho(t)\chi_{n}(t)\right) \right] \|x_{n}(t) - x_{n-1}(t)\| \\ &\leq \theta_{n}(t)\|x_{n}(t) - x_{n-1}(t)\|, \end{split} \tag{4.15} \\ \text{where } \kappa(t) &= \sqrt{1 - 2r_{g}(t) + s_{g}^{2}(t)} + \delta(t), \ \Delta_{n} = \sqrt{1 + 2\zeta_{\tilde{G}}(t) + \sigma_{\tilde{G}}^{2}(t)p_{N}^{2}(t)(1 + n^{-1})^{2}}, \ \text{and} \\ \ell_{n} &= \sqrt{1 - 2\lambda_{\tilde{T}}(t) + \epsilon_{N}^{2}(t)\gamma_{\tilde{T}}^{2}(t)(1 + n^{-1})^{2}}, \ \text{and} \\ \chi_{n}(t) &= \Delta_{n} + \ell_{n} \end{split}$$

$$\theta_n(t) = \kappa(t) + \frac{L_\eta(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)(1+n^{-1})^2} + \chi_n(t)\rho(t) \right).$$

Letting

$$\theta(t) = \kappa(t) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)s_g^2(t) + \rho^2(t)d_N^2(t)\beta_{\tilde{A}}^2(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} \left(\sqrt{s_g^2(t) - 2\rho(t)\xi_{\tilde{A},g}(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} + \chi(t)\rho(t) + \chi(t)\rho(t) \right) + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} + \frac{L_{\eta}(t)}{\alpha_{h_t}(t)} + \frac{L_{\eta}$$

we know that $\theta_n(t) \to \theta(t)$ for all $t \in \Omega$. It follows from (4.5) that $\theta(t) < 1$ for all $t \in \Omega$. Hence for any $t \in \Omega$, $\theta_n(t) < 1$ for n sufficiently large.

Therefore $\{x_n(t)\}\$ is a Cauchy sequence in H. Since H is complete, there exists a measurable mapping $x: \Omega \to H$ such that $x_n(t) \to x(t) \in H$, for all $t \in \Omega$.

From Algorithm 4.2, we have

$$\begin{aligned} \|u_n(t) - u_{n-1}(t)\| &\leq \beta_{\tilde{A}}(t)(1+n^{-1}) \|x_n(t) - x_{n-1}(t)\|, \\ \|v_n(t) - v_{n-1}(t)\| &\leq \gamma_{\tilde{T}}(t)(1+n^{-1}) \|x_n(t) - x_{n-1}(t)\|, \\ \|w_n(t) - w_{n-1}(t)\| &\leq \sigma_{\tilde{G}}(t)(1+n^{-1}) \|x_n(t) - x_{n-1}(t)\|, \end{aligned}$$

which implies that $\{x_n(t)\}, \{u_n(t)\}, \{v_n(t)\}\)$ and $\{w_n(t)\}\)$ are also Cauchy sequences in H. Let $u_n(t) \to u(t), v_n(t) \to v(t)$ and $w_n(t) \to w(t)$. Since $\{u_n(t)\}, \{v_n(t)\}\)$ and $\{w_n(t)\}\)$ are sequences of measurable mappings. We know that $x, u, v, w : \Omega \to H$ are measurable.

Now we will prove that $u(t) \in \hat{A}(t, x(t)), v(t) \in \hat{T}(t, x(t))$ and $w(t) \in \hat{G}(t, x(t))$. For any $t \in \Omega$, we have

$$\begin{aligned} d(u(t), A(t, x(t)) &= \inf\{\|u(t) - z\| : z \in A(t, x(t))\} \\ &\leq \|u(t) - u_n(t)\| + d(u_n(t), \tilde{A}(t, x(t))) \\ &\leq \|u(t) - u_n(t)\| + D(\tilde{A}(t, x_n(t)), \tilde{A}(t, x(t))) \\ &\leq \|u(t) - u_n(t)\| + \beta_{\tilde{A}}(t)\|x_n(t) - x(t)\| \to 0 \text{ as } n \to \infty. \end{aligned}$$

Hence, $u(t) \in \tilde{A}(t, x(t))$, for all $t \in \Omega$. Similarly, we can prove that $v(t) \in \tilde{T}(t, x(t))$ and $w(t) \in \tilde{G}(t, x(t))$ for all $t \in \Omega$. This completes the proof.

Acknowledgement

The authors thank the referees for their valuable useful suggestions that improved the content of the paper.

References

- S. Adly, Perturbed algorithms and sensitivity analysis for a general class of variational inclusions, J. Math. Anal. Appl. 201 (1996) 609–630.
- [2] R.P. Agarwal, N.J. Huang and M.Y. Tan, Sensitivity analysis for a new system of generalized nonlinear mixed quasi-variational inclusions, *Appl. Math. Lett.* 17 (2004) 345–352.
- [3] R.P. Agarwal, M.F. Khan, D.O'Regan and Salahuddin, On generalized multivalued nonlinear variational inclusions with fuzzy mappings, *Adv. Nonlinear Var. Inequal.* 8 (2005) 41–55.

- [4] R. Ahmad, S. Hussain and Salahuddin, Generalized nonlinear variational inclusions for fuzzy mappings, *Indian J. Pure Appl. Math.* 32 (2001) 943–947.
- [5] C. Baiocchi and A. Capelo, Variational and Quasivariational Inequalities, Application to Free Boundary Problems, John Wiley & Sons, New York, 1984.
- [6] S.S. Chang, Set-valued variational inclusions in Banach spaces, J. Math. Anal. Appl. 248 (2000) 438–454.
- [7] S.S. Chang, Variational Inequality and Complementarity Problem Theory with Application, Shanghai Scientific and Tech. Literature Publ. House, Shanghai, 1991.
- [8] S.S. Chang, Fixed Point Theory with Applications, Chongging Publishing House, Chongging, 1984.
- [9] S.S. Chang and N.J. Huang, On the problems for a class of random variational inequalities and quasi-variational inequalities, J. Math. Res. Exposition 9 (1989) 385–393.
- [10] S.S. Chang and N.J. Huang, Random generalized set-valued quasi-complementarity problems, Acta Math. Appl. Sinica 16 (1993) 396–405.
- [11] S.S. Chang and Y.G. Zhu, On variational inequalities for fuzzy mappings, Fuzzy Sets and Systems 32 (1989) 359–367.
- [12] X.P. Ding, Generalized implicit quasi-variational inequalities with fuzzy set-valued mappings, Comput. Math. Appl. 38 (1999) 71–79.
- [13] X.P. Ding, Algorithm of solutions for mixed implicit quasi-variational inequalities with fuzzy mappings, *Comput. Math. Appl.* 38 (1999) 231–241.
- [14] X.P. Ding, Chinsan Lee and Su-Jane Yu, Algorithm of solutions for a system of generalized mixed implicit quasi-variational inclusions involving h- η -maximal monotone mappings, (preprint).
- [15] X.P. Ding and C.L. Luo, Perturbed proximal point algorithms for general quasivariational-like inclusions, J. Comput. Appl. Math. 113 (2000) 153–165.
- [16] Y.P. Ding, Generalized quasi-variational-like inclusions with nonconvex functionals, Appl. Math. Comp. 122 (2001) 267–282.
- [17] Y.P. Fang and N.J. Huang, H-monotone operator and resolvent operator technique for variational inclusions, Appl. Math. Comput. 145 (2003) 795–803.
- [18] Y.P. Fang, N.J. Huang and H.B. Thompson, A new system of variational inclusions with (H, η) -monotone operators in Hilbert spaces, *Comput. Math. Appl.* 49 (2005) 365–374.
- [19] F. Giannessi and A. Maugeri, Variational Inequalities and Network Equilibrium Problems, New York, 1995.
- [20] A. Hassouni and A. Moudafi, A perturbed algorithm for variational inclusions, J. Math. Anal. Appl. 185 (1994) 706–712.
- [21] S. Heilpern, Fuzzy mappings and fixed point theorems, J. Math. Anal. Appl. 83 (1981) 566–569.

- [22] G.J. Himmelberg, Measurable relations, Fund. Math. 87 (1975) 53-72.
- [23] N.J. Huang, Generalized nonlinear variational inclusions with noncompact valued mappings, Appl. Math. Lett. 9 (1996) 25–29.
- [24] N.J. Huang, Random generalized set-valued implicit variational inequalities, J. Liaonag Normal Univ. 31 (1994) 420–425.
- [25] N.J. Huang, Random generalized nonlinear variational inclusions for random fuzzy mappings, Fuzzy Sets and Systems 105 (1999) 437–444.
- [26] N.J. Huang and Y.P. Fang, A new class of general inclusions involving maximal ηmonotone mappings, Publ. Math. Debrecen 62 (2003) 83–98.
- [27] T. Hussain, E. Tarafdar and X.Z. Yuan, Some results on random generalized games and random quasi-variational inequalities, *Far East J. Math. Sci.* 21 (1994) 35–55.
- [28] K.R. Kazmi, Mann and Ishikawa type perturbed iterative algorithms for generalized quasi-variational inclusions, J. Math. Anal. Appl. 209 (1997) 572–584.
- [29] M.F. Khan and Salahuddin, Completely generalized nonlinear random variational inclusions, South East Asian Bulletin of Mathematics 30 (2006) 261–276.
- [30] M. Lassonde, On the use of KKM-multi function in fixed point theory and related topics, J. Math. Anal. Appl. 97 (1983) 151–206.
- [31] L.J. Lin, Q.H. Ansari and Y.J. Huang, System of vector quasi-variational inclusions with some applications, *Nonlinear Analysis* 69 (2008) 2812–2824.
- [32] Jr. S.B. Nadler, Multivalued contraction mappings, Pacific J. Math. 30 (1969) 475-485.
- [33] M.A. Noor, Generalized set-valued variational inclusions and resolvent equations, J. Math. Anal. Appl. 228 (1998) 206–220.
- [34] J.W. Peng and D.L. Zhu, A new system of generalized mixed quasi-variational inclusions with (H, η) -monotone operators, J. Math. Anal. Appl. 327 (2007) 175–187.
- [35] J.W. Peng and D.L. Zhu, Three-step iterative algorithm for a system of set-valued variational inclusions with (H, η) -monotone operators, *Nonlinear Analysis* 68 (2008) 139–153.
- [36] Salahuddin, Some Aspects of Variational Inequalities, Ph.D. Thesis, A.M.U., Aligarh, India, 2000.
- [37] H.M. Shih and K.K. Tan, Generalized quasi-variational inequalities in locally convex topological vector spaces, J. Math. Anal. Appl. 108 (1985) 333–343.
- [38] W. Takahashi, Nonlinear variational inequalities and fixed point theorem, J. Math. Soc. Japan 28 (1976) 168–181.
- [39] K.K. Tan, E. Tarafdar and X.Z. Yuan, Random variational inequalities and applications to random minimization and nonlinear boundary problems, *Pan Amer. Math. J.* 4 (1994) 55–71.
- [40] X.Z. Tan, Random quasi-variational inequality, Math. Nachr. 125 (1986) 319–328.

- [41] R.U. Verma, A-monotonicity and its role in nonlinear variational inclusions, J. Optim. Th. Appl. 129 (2006) 457–467.
- [42] R.U. Verma, Generalized Eckstein-Bertsekas proximal point algorithm involving (H, η) -monotonicity framework, *Math. Comput. Model.* 45 (2007) 1214–1230.
- [43] C.L. Yen, A minimax inequality and its applications to variational inequality, Pacific J. Math. 97 (1981) 142–150.
- [44] X.Z. Yuan, Non compact random generalized games and random quasi-variational inequalities, J. Appl. Stochastic Anal. 7 (1994) 467–486.
- [45] L.A. Zadeh, Fuzzy sets, Inform. Control 8 (1965) 338–353.

Manuscript received 27 August 2008 revised 12 January 2009, 9 April 2009 accepted for publication 11 June 2009

BYUNG-SOO LEE Department of Mathematics, Kyungsung University, Busan 608-736, Korea E-mail address: bslee@ks.ac.kr

M. FIRDOSH KHAN Department of Mathematics Aligarh Muslim University, Aligarh-202002, India E-mail address: khan_mfk@yahoo.com

SALAHUDDIN Department of Mathematics Aligarh Muslim University, Aligarh-202002, India E-mail address: salahuddin12@mailcity.com