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# A New Method for a Class of Nonlinear Variational Inequalities with Fuzzy Mappings

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**Abstract**—In this paper, we construct a new iterative algorithm of solution for a new class of nonlinear variational inequalities with fuzzy mappings and give some convergence analysis of iterative sequences generated by algorithm.

**Keywords**—Variational inequality, Fuzzy mapping, Algorithm, Convergence.

## 1. INTRODUCTION

Variational inequalities not only have stimulated new results dealing with nonlinear partial differential equations, but also have been used in a large variety of problems arising in mechanics, physics, optimization and control, nonlinear programming, economics and transportation equilibrium, and engineering sciences, etc. (See [1–6] and the references cited therein.)

Recently, the variational inequalities for fuzzy mappings was introduced and studied by Chang and Zhu [7] and later developed by Noor [8], Chang [9], Chang and Huang [10,11] and Lee *et al.* [12]. On the other hand, in the recent papers [13–16], He proposed some new projection and contraction methods for a class of linear or nonlinear variational inequalities with monotone mappings.

The purpose of this paper is to construct a new iterative algorithm of solution for a new class of nonlinear variational inequalities with fuzzy mappings. We also give some convergence analysis of iterative sequences generated by algorithm. Our results develop some projection and contraction methods for solving monotone variational inequalities.

## 2. PRELIMINARIES

Let  $\mathcal{F}(R^n)$  be the family of all fuzzy sets over  $R^n$ . A mapping  $F$  from  $R^n$  into  $\mathcal{F}(R^n)$  is called fuzzy mapping over  $R^n$ . If  $F$  is a fuzzy mapping over  $R^n$ , the  $F(x)$  (denoted by  $F_x$  in the sequel) is a fuzzy set over  $R^n$ , and  $F_x(y)$  is the membership function of the point  $y$  in  $F_x$ . Let  $B \in \mathcal{F}(R^n)$ ,  $p \in [0, 1]$ . Then  $(B)_p = \{x \in R^n : B(x) \geq p\}$  is called a  $p$ -cut set of  $B$ .

A set-valued mapping  $A : R^n \rightarrow 2^{R^n}$  (where  $2^{R^n}$  denotes the family of all nonempty subsets of  $R^n$ ) is called to be upper semicontinuous [17], if for any  $\{u_k\} \subset R^n$ ,  $y_k \in A(u_k)$ ,

$$u_k \rightarrow u_*, \quad y_k \rightarrow y_* \implies y_* \in A(u_*).$$

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Let  $F : R^n \rightarrow \mathcal{F}(R^n)$  and  $p : R^n \rightarrow [0, 1]$ . We can define a set-valued mapping  $\tilde{F}$  as follows:

$$\tilde{F} : R^n \rightarrow 2^{R^n}, \quad x \mapsto (F_x)_{p(x)}.$$

In the sequel, we shall call  $\tilde{F}$  the set-valued mapping induced by fuzzy mapping  $F$ .

A fuzzy mapping  $F : R^n \rightarrow \mathcal{F}(R^n)$  is called to be upper semicontinuous monotone, if  $\tilde{F}$  is upper semicontinuous monotone.

Let  $K \subset R^n$  be a closed convex set,  $F : R^n \rightarrow \mathcal{F}(R^n)$  and  $p : R^n \rightarrow [0, 1]$ . We consider the following problem: find  $u \in K$ ,  $y \in R^n$  such that  $F_u(y) \geq p(u)$  and

$$(v - u)^\top y \geq 0, \quad \forall v \in K, \quad (2.1)$$

which is called a nonlinear variational inequality for fuzzy mapping.

If  $F : R^n \rightarrow 2^{R^n}$  and  $p(x) \equiv 1$ ,  $\forall x \in R^n$ , then the problem (2.1) is equivalent to finding  $u \in K$ ,  $y \in F(u)$  such that

$$(v - u)^\top y \geq 0, \quad \forall v \in K, \quad (2.2)$$

which is called a set-valued nonlinear variational inequality. Obviously, the problems (2.1) and (2.2) include many variational inequalities as special cases.

Let  $S$  denote the solution set of (2.1) and  $P_K(\cdot)$  denote the projection to  $K$ . Throughout this paper, we assume that  $S \neq \emptyset$ . In following the  $\|\cdot\|$ ,  $G$  and  $\|u\|_G$  will be denoted the Euclidean norm, positive definite matrix and  $(u^\top G u)^{1/2}$  respectively.

### 3. ITERATIVE ALGORITHM

Let  $\beta > 0$  be a constant,  $F : R^n \rightarrow \mathcal{F}(R^n)$  and  $p : R^n \rightarrow [0, 1]$ . By [11], we know that  $u \in K$ ,  $y \in R^n$  are the solutions of problem (2.1) if and only if  $u \in K$ ,  $y \in R^n$  such that  $F_u(y) \geq p(u)$  and

$$e(u, y, \beta) \stackrel{\text{def}}{=} u - P_K[u - \beta y] = 0. \quad (3.1)$$

Let  $\gamma \in (0, 2)$  be a constant and  $F$  be a upper semicontinuous monotone fuzzy mapping. Since  $F$  is monotone, we know that  $\tilde{F}$  is also monotone. By [18],  $(I + \beta\tilde{F})^{-1}$  is a single-valued nonexpansive mapping.

Now, we give the algorithm of solution for the problem (2.1) as follows.

**ALGORITHM 3.1.** Given  $u^0 \in R^n$ ,  $y^0 \in \tilde{F}(u^0)$ , for  $k = 0, 1, 2, \dots$ , if  $(u^k, y^k) \notin S$  and  $u^k + \beta y^k - \gamma \rho_k G^{-1} e(u^k, y^k, \beta)$  is in the image of  $I + \beta\tilde{F}$ , then

$$u^{k+1} = (I + \beta\tilde{F})^{-1} (u^k + \beta y^k - \gamma \rho_k G^{-1} e(u^k, y^k, \beta)), \quad (3.2)$$

and  $y^{k+1} \in \tilde{F}(u^{k+1})$  such that

$$u^{k+1} + \beta y^{k+1} = u^k + \beta y^k - \gamma \rho_k G^{-1} e(u^k, y^k, \beta), \quad (3.3)$$

where

$$\rho_k = \frac{\|e(u^k, y^k, \beta)\|^2}{e(u^k, y^k, \beta)^\top G^{-1} e(u^k, y^k, \beta)}. \quad (3.4)$$

**REMARK 3.1.**

- (i) If  $F$  is a set-valued mapping, then Algorithm 3.1 reduces to a new iterative algorithm of solution for the set-valued nonlinear variational inequality (2.2).
- (ii) If  $F$  is a single-valued mapping, then Algorithm 3.1 reduces to the PC methods for monotone variational inequalities of [13].
- (iii) If  $F$  is a single-valued mapping,  $G = I$  and  $\gamma = 1$ , then Algorithm 3.1 reduces to the Douglas-Rachford operator splitting methods (see [13,19,20]).
- (iv) In addition, if  $F$  is a single-valued mapping,  $G = I$ ,  $\gamma = 1$ , and  $K = R^n$ , then Algorithm 3.1 reduces the the Levenberg-Marquardt method for unconstrained optimization [13,16].

#### 4. CONVERGENCE

We first give the following lemma.

LEMMA 4.1. *Let  $(u^*, y^*) \in S$ , then*

$$\begin{aligned} (u - u^* + \beta(y - y^*))^\top e(u, y, \beta) &\geq \|e(u, y, \beta)\|^2 + \beta(u - u^*)^\top (y - y^*), \\ \forall u \in R^n, \quad \forall y \in \tilde{F}(u). \end{aligned} \quad (4.1)$$

PROOF. Since  $(u^*, y^*)$  is a solution of the problem (2.1) and  $P_K(\cdot) \in K$ , it follows from (2.1) that

$$\beta y^{*\top} (P_K(u - \beta y) - u^*) \geq 0, \quad \forall u \in R^n, \quad \forall y \in \tilde{F}(u). \quad (4.2)$$

A well know basic property of the projection mapping is

$$(v - P_K v)^\top (P_K v - u) \geq 0, \quad \forall v \in R^n, \quad \forall u \in K. \quad (4.3)$$

Letting  $v := u - \beta y$  and  $u := u^*$  in (4.3), we have

$$(e(u, y, \beta) - \beta y)^\top (P_K(u - \beta y) - u^*) \geq 0. \quad (4.4)$$

Adding (4.2) and (4.4), we obtain

$$(e(u, y, \beta) - \beta(y - y^*))^\top (u - u^* - e(u, y, \beta)) \geq 0. \quad (4.5)$$

From the above inequality follows the assertion of this lemma immediately. This completes the proof.

THEOREM 4.1. *The sequences  $\{u^k\}$  and  $\{y^k\}$  generated by Algorithm 3.1 satisfy the following inequality:*

$$\begin{aligned} \|u^{k+1} - u^* + \beta(y^{k+1} - y^*)\|_G^2 &\leq \|u^k - u^* + \beta(y^k - y^*)\|_G^2 \\ &\quad - \gamma(2 - \gamma)\rho_k \|e(u^k, y^k, \beta)\|^2 \\ &\quad - 2\gamma\rho_k\beta (u^k - u^*)^\top (y^k - y^*), \quad \forall (u^*, y^*) \in S. \end{aligned} \quad (4.6)$$

PROOF. From (3.2) and (3.3), we have

$$u^{k+1} - u^k + \beta(y^{k+1} - y^k) = -\gamma\rho_k G^{-1}e(u^k, y^k, \beta). \quad (4.7)$$

It follows from (3.4), (4.1), and (4.7) that

$$\begin{aligned} \|u^{k+1} - u^* + \beta(y^{k+1} - y^*)\|_G^2 &= \|u^k - u^* + \beta(y^k - y^*) - \gamma\rho_k G^{-1}e(u^k, y^k, \beta)\|_G^2 \\ &= \|u^k - u^* + \beta(y^k - y^*)\|_G^2 \\ &\quad - 2\gamma\rho_k (u^k - u^* + \beta(y^k - y^*))^\top e(u^k, y^k, \beta) \\ &\quad + \gamma^2 \rho_k^2 e(u^k, y^k, \beta)^\top G^{-1}e(u^k, y^k, \beta) \\ &\leq \|u^k - u^* + \beta(y^k - y^*)\|_G^2 \\ &\quad - \gamma(2 - \gamma)\rho_k \|e(u^k, y^k, \beta)\|^2 \\ &\quad - 2\gamma\rho_k\beta (u^k - u^*)^\top (y^k - y^*), \quad \forall (u^*, y^*) \in S. \end{aligned}$$

This completes the proof.

**THEOREM 4.2.** *The sequences  $\{u^k\}$  and  $\{y^k\}$  generated by Algorithm 3.1 converge to  $\bar{u}$  and  $\bar{y}$ , respectively, and  $(\bar{u}, \bar{y})$  is a solution of the problem (2.1).*

**PROOF.** Let  $(\hat{u}, \hat{y})$  be a solution of the problem (2.1). Since  $\tilde{F}$  is monotone, it follows from (4.6) that the sequences  $\{u^k\}$  and  $\{y^k\}$  are bounded. Also from (4.6), we have

$$\sum_{k=0}^{\infty} \gamma(2-\gamma)\rho_k \|e(u^k, y^k, \beta)\|^2 \leq \|u^0 - \hat{u} + \beta(y^0 - \hat{y})\|_G^2.$$

This implies that

$$\lim_{k \rightarrow \infty} e(u^k, y^k, \beta) = 0.$$

Let  $\bar{u}$  be a cluster point of  $\{u^k\}$ , the subsequence  $\{u^{k_j}\}$  converge to  $\bar{u}$ , and  $\bar{y}$  be a cluster point of  $\{y^k\}$ , the subsequence  $\{y^{k_j}\}$  converge to  $\bar{y}$ . Since  $P_K(\cdot)$  is continuous, (3.1) shows that  $e(u, y, \beta)$  is continuous, and therefore,

$$e(\bar{u}, \bar{y}, \beta) = \lim_{j \rightarrow \infty} e(u^{k_j}, y^{k_j}, \beta) = 0. \quad (4.8)$$

Since  $y^k \in \tilde{F}(u^k)$ ,  $u^{k_j} \rightarrow \bar{u}$ ,  $y^{k_j} \rightarrow \bar{y}$ , and  $\tilde{F}$  is upper semicontinuous, we have  $\bar{y} \in \tilde{F}(\bar{u})$  and so (4.8) implies that  $(\bar{u}, \bar{y})$  is a solution of the problem (2.1).

Now, we prove that  $u^k \rightarrow \bar{u}$  and  $y^k \rightarrow \bar{y}$ . In fact, from  $(\bar{u}, \bar{y}) \in S$  and (4.6), we get

$$\|u^{k+1} - \bar{u} + \beta(y^{k+1} - \bar{y})\|_G^2 \leq \|u^k - \bar{u} + \beta(y^k - \bar{y})\|_G^2,$$

and this yields that

$$\lim_{k \rightarrow \infty} u^k = \bar{u}, \quad \lim_{k \rightarrow \infty} y^k = \bar{y}.$$

This completes the proof.

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