

## A MODIFIED INERTIAL SUBGRADIENT EXTRAGRADIENT METHOD FOR SOLVING PSEUDOMONOTONE VARIATIONAL INEQUALITIES AND COMMON FIXED POINT PROBLEMS

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**Abstract.** In this paper, we introduce a modified inertial subgradient extragradient method for solving a variational inequality problem with Lipschitz pseudomonotone mapping and a common fixed-point problem of a family of nonexpansive mappings. Under mild conditions, we obtain strong convergence theorems in a real Hilbert space. An application is also provided.

**Key Words and Phrases:** Inertial subgradient extragradient method, variational inequality, pseudomonotone mapping, nonexpansive mapping, fixed point.

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### 1. INTRODUCTION

Let  $H$  be a real Hilbert space with inner product  $\langle \cdot, \cdot \rangle$  and induced norm  $\| \cdot \|$ . Let  $C$  be a nonempty closed convex subset of  $H$  and let  $P_C$  be the metric projection of  $H$  onto  $C$ . Let  $S : C \rightarrow H$  be a nonlinear mapping on  $C$ . We denote by  $\text{Fix}(S)$  the set of fixed points of  $S$ . Let  $A : H \rightarrow H$  be a mapping. Consider the classical variational inequality problem (VIP) of finding  $x^* \in C$  such that  $\langle Ax^*, x - x^* \rangle \geq 0, \forall x \in C$ . The solution set of the VIP is denoted by  $\text{VI}(C, A)$ . Recently, much attention has been focused on solution methods for the VIP; see, e.g., [24, 17, 18, 8, 6, 23, 5, 1, 7, 3, 2] and references therein. One of effective methods to solve the VIP is the extragradient

method, which was introduced by Korpelevich [13] in 1976. It generates a sequence  $\{x_n\}$  in the following manner:  $x_0 \in C$ ,

$$\begin{cases} y_n = P_C(x_n - \tau Ax_n), \\ x_{n+1} = P_C(x_n - \tau Ay_n) \quad \forall n \geq 0, \end{cases} \quad (1.1)$$

where  $A$  is a  $L$ -Lipschitz continuous monotone mapping and  $\tau \in (0, \frac{1}{L})$ .

It deserves mentioning that there are two projections onto  $C$  for each iteration. In most cases, metric projections are not easy to calculate. In 2011, Censor, Gibali and Reich [4] first introduced the subgradient extragradient method, in which the second projection onto  $C$  was replaced by a projection onto a half-space:

$$\begin{cases} y_n = P_C(x_n - \tau Ax_n), \\ C_n = \{w \in H : \langle x_n - \tau Ax_n - y_n, w - y_n \rangle \leq 0\}, \\ x_{n+1} = P_{C_n}(x_n - \tau Ay_n) \quad \forall n \geq 0, \end{cases} \quad (1.2)$$

where  $A$  is a  $L$ -Lipschitz continuous monotone mapping and  $\tau \in (0, \frac{1}{L})$ .

Combining the subgradient extragradient method and the Halpern's iteration method, Kraikaew and Saejung [14] proposed the Halpern subgradient extragradient method for solving the VIP in 2014. For any initial  $x_0 \in H$ , their iterative sequence  $\{x_n\}$  was generated by

$$\begin{cases} y_n = P_C(x_n - \tau Ax_n), \\ C_n = \{x \in H : \langle x_n - \tau Ax_n - y_n, x - y_n \rangle \leq 0\}, \\ z_n = P_{C_n}(x_n - \tau Ay_n), \\ x_{n+1} = \alpha_n x_0 + (1 - \alpha_n) z_n \quad \forall n \geq 0, \end{cases} \quad (1.3)$$

where  $\tau \in (0, \frac{1}{L})$ ,  $\{\alpha_n\} \subset (0, 1)$ ,  $\lim_{n \rightarrow \infty} \alpha_n = 0$  and  $\sum_{n=1}^{\infty} \alpha_n = +\infty$ . They proved the strong convergence of  $\{x_n\}$  to  $P_{VI(C,A)}x_0$ .

In 2018, Thong and Hieu [19] first proposed the following inertial subgradient extragradient method. For any initial  $x_0, x_1 \in H$ , their iterative sequence  $\{x_n\}$  was generated by

$$\begin{cases} w_n = x_n + \alpha_n(x_n - x_{n-1}), \\ y_n = P_C(w_n - \tau Aw_n), \\ C_n = \{x \in H : \langle w_n - \tau Aw_n - y_n, x - y_n \rangle \leq 0\}, \\ x_{n+1} = P_{C_n}(w_n - \tau Ay_n) \quad \forall n \geq 1, \end{cases} \quad (1.4)$$

with constant  $\tau \in (0, \frac{1}{L})$ . Under suitable conditions, they proved the weak convergence of  $\{x_n\}$  to an element of  $VI(C, A)$ .

Very recently, Thong et al. [20] introduced an inertial subgradient extragradient-type method for solving the VIP with pseudomonotone and Lipschitz continuous mapping in a real Hilbert space. Under appropriate conditions, they proved the strong convergence of  $\{x_n\}$  to an element of  $VI(C, A)$ .

In this paper, we introduce a modified inertial subgradient extragradient method for solving the VIP with a pseudomonotone and Lipschitz continuous mapping and a common fixed point problem (CFFP) of nonexpansive mappings in a real Hilbert space. Our proposed algorithm is based on the inertial subgradient extragradient method, hybrid steepest-descent method, and viscosity approximation method. Under

mild conditions, we prove strong convergence of the proposed algorithm to a common solution of the VIP and CFPP. Our main result can also be applied to common solution problems of a fractional programming and a fixed-point problem.

This paper is organized as follows: In Section 2, we recall some definitions and preliminaries for the sequel use. Section 3 deals with the convergence analysis of the proposed algorithm. Finally, in Section 4, our main result is applied to a common solution problem of the fractional programming and the fixed-point problem.

## 2. PRELIMINARIES

Let  $\{x_n\}$  be a sequence in a Hilbert space  $H$ . We denote by  $x_n \rightarrow x$  (respectively,  $x_n \rightharpoonup x$ ) the strong (respectively, weak) convergence of  $\{x_n\}$  to  $x$ .

A mapping  $T : C \rightarrow H$  is said to be nonexpansive if  $\|Tx - Ty\| \leq \|x - y\|$ ,  $\forall x, y \in C$ . Recall that  $T : C \rightarrow H$  is said to be

(i)  $L$ -Lipschitz continuous (or  $L$ -Lipschitzian) if  $\exists L > 0$  such that

$$\|Tx - Ty\| \leq L\|x - y\|, \forall x, y \in C;$$

(ii) monotone if  $\langle Tx - Ty, x - y \rangle \geq 0$ ,  $\forall x, y \in C$ ;

(iii) pseudomonotone if  $\langle Tx, y - x \rangle \geq 0 \Rightarrow \langle Ty, y - x \rangle \geq 0$ ,  $\forall x, y \in C$ ;

(iv)  $\alpha$ -strongly monotone if  $\exists \alpha > 0$  such that

$$\langle Fx - Fy, x - y \rangle \geq \alpha\|x - y\|^2, \forall x, y \in C;$$

(v) sequentially weakly continuous if  $\forall \{x_n\} \subset C$ , the relation holds:

$$x_n \rightharpoonup x \Rightarrow Tx_n \rightharpoonup Tx.$$

It is easy to see that every monotone operator is pseudomonotone but the converse is not true. For each  $x \in H$ , we know that there exists a unique nearest point in  $C$ , denoted by  $P_C x$ , such that  $\|x - P_C x\| \leq \|x - y\|$ ,  $\forall y \in C$ .  $P_C$  is called a metric projection of  $H$  onto  $C$ .

**Lemma 2.1.** *The following conclusions hold in a Hilbert space  $H$ :*

(i)  $\langle x - P_C x, y - P_C x \rangle \leq 0 \forall x \in H, y \in C$ ;

(ii)  $\|x - y\|^2 \geq \|x - P_C x\|^2 + \|y - P_C x\|^2 \forall x \in H, y \in C$ ;

(iii)  $\|x - y\|^2 = \|x\|^2 - \|y\|^2 - 2\langle x - y, y \rangle \forall x, y \in H$ ;

(iv)  $\|\lambda x + \mu y\|^2 = \lambda\|x\|^2 + \mu\|y\|^2 - \lambda\mu\|x - y\|^2 \forall x, y \in H, \forall \lambda, \mu \in [0, 1]$  with  $\lambda + \mu = 1$ .

**Lemma 2.2.** [9] *For all  $x \in H$  and  $\alpha \geq \beta > 0$  the inequalities hold:*

$$\frac{\|x - P_C(x - \alpha Ax)\|}{\alpha} \leq \frac{\|x - P_C(x - \beta Ax)\|}{\beta}$$

and

$$\|x - P_C(x - \beta Ax)\| \leq \|x - P_C(x - \alpha Ax)\|.$$

**Lemma 2.3.** [4] *Let  $A : C \rightarrow H$  be pseudomonotone and continuous. Then  $x^* \in C$  is a solution to the VIP  $\langle Ax^*, x - x^* \rangle \geq 0 \forall x \in C$ , if and only if*

$$\langle Ax, x - x^* \rangle \geq 0 \forall x \in C.$$

**Lemma 2.4.** [21] *Let  $\{a_n\}$  be a sequence of nonnegative numbers satisfying the conditions:  $a_{n+1} \leq (1 - \lambda_n)a_n + \lambda_n\gamma_n \forall n \geq 1$ , where  $\{\lambda_n\}$  and  $\{\gamma_n\}$  are sequences of real numbers such that*

$$(i) \{\lambda_n\} \subset [0, 1] \text{ and } \sum_{n=1}^{\infty} \lambda_n = \infty, \text{ and}$$

$$(ii) \limsup_{n \rightarrow \infty} \gamma_n \leq 0 \text{ or } \sum_{n=1}^{\infty} |\lambda_n\gamma_n| < \infty.$$

*Then  $\lim_{n \rightarrow \infty} a_n = 0$ .*

**Lemma 2.5.** [10] *Let  $T : C \rightarrow C$  be a nonexpansive mapping with  $\text{Fix}(T) \neq \emptyset$ . Then  $I - T$  is demiclosed at zero, that is, if  $\{x_n\}$  is a sequence in  $C$  such that  $x_n \rightarrow x \in C$  and  $(I - T)x_n \rightarrow 0$ , then  $(I - T)x = 0$ , where  $I$  is the identity mapping of  $H$ .*

**Lemma 2.6.** [22] *Let  $\lambda \in (0, 1]$ ,  $T : C \rightarrow H$  be a nonexpansive mapping, and the mapping  $T^\lambda : C \rightarrow H$  be defined by  $T^\lambda x := Tx - \lambda\mu F(Tx) \forall x \in C$ , where  $F : H \rightarrow H$  is  $\kappa$ -Lipschitzian and  $\eta$ -strongly monotone. Then  $T^\lambda$  is a contraction provided  $0 < \mu < \frac{2\eta}{\kappa^2}$ , i.e.,*

$$\|T^\lambda x - T^\lambda y\| \leq (1 - \lambda\tau)\|x - y\|, \forall x, y \in C,$$

*where  $\tau = 1 - \sqrt{1 - \mu(2\eta - \mu\kappa^2)} \in (0, 1]$ .*

### 3. CONVERGENCE THEOREMS

In this section, let the feasible set  $C$  be a nonempty closed convex subset of a real Hilbert space  $H$ , and always assume that the following hold:

$T_i : H \rightarrow H$  is nonexpansive for  $i = 1, \dots, N$ ;

$A : H \rightarrow H$  is  $L$ -Lipschitz continuous, pseudomonotone monotone on  $H$ , and sequentially weakly continuous on  $C$ , such that  $\Omega = \bigcap_{i=1}^N \text{Fix}(T_i) \cap \text{VI}(C, A) \neq \emptyset$ ;

$f : H \rightarrow H$  is a contraction with constant  $\delta \in [0, 1)$ , and  $F : H \rightarrow H$  is  $\eta$ -strongly monotone and  $\kappa$ -Lipschitzian such that

$$\delta < \tau := 1 - \sqrt{1 - \rho(2\eta - \rho\kappa^2)} \text{ for } \rho \in \left(0, \frac{2\eta}{\kappa^2}\right);$$

$\{\beta_n\}, \{\gamma_n\}, \{\tau_n\}$  are positive sequences such that  $\beta_n + \gamma_n < 1$ ,

$$\sum_{n=1}^{\infty} \beta_n = \infty, \lim_{n \rightarrow \infty} \beta_n = 0, 0 < \liminf_{n \rightarrow \infty} \gamma_n \leq \limsup_{n \rightarrow \infty} \gamma_n < 1$$

and  $\tau_n = o(\beta_n)$ .

In addition, we write  $T_n := T_{n \bmod N}$  for integer  $n \geq 1$  with the mod function taking values in the set  $\{1, 2, \dots, N\}$ , that is, if  $n = jN + q$  for some integers  $j \geq 0$  and  $0 \leq q < N$ , then  $T_n = T_N$  if  $q = 0$  and  $T_n = T_q$  if  $0 < q < N$ .

**Algorithm 3.1. Initialization.** Let  $\lambda_1 > 0$ ,  $\alpha > 0$ ,  $\mu \in (0, 1)$  and  $x_0, x_1 \in H$  be arbitrary.

**Iterative Steps.** Calculate  $x_{n+1}$  as follows:

**Step 1.** Given the iterates  $x_{n-1}$  and  $x_n$  ( $n \geq 1$ ), choose  $\alpha_n$  such that  $0 \leq \alpha_n \leq \overline{\alpha}_n$ , where

$$\overline{\alpha}_n = \begin{cases} \min\{\alpha, \frac{\tau_n}{\|x_n - x_{n-1}\|}\} & \text{if } x_n \neq x_{n-1}, \\ \alpha & \text{otherwise.} \end{cases} \quad (3.1)$$

**Step 2.** Compute  $w_n = T_n x_n + \alpha_n(T_n x_n - T_n x_{n-1})$  and  $y_n = P_C(w_n - \lambda_n A w_n)$ .

**Step 3.** Construct the half-space  $C_n := \{z \in H : \langle w_n - \lambda_n A w_n - y_n, z - y_n \rangle \leq 0\}$ , and compute  $z_n = P_{C_n}(w_n - \lambda_n A y_n)$ .

**Step 4.** Calculate  $x_{n+1} = \beta_n f(x_n) + \gamma_n x_n + ((1 - \gamma_n)I - \beta_n \rho F)z_n$ , and update

$$\lambda_{n+1} = \begin{cases} \min\{\mu \frac{\|w_n - y_n\|^2 + \|z_n - y_n\|^2}{2\langle A w_n - A y_n, z_n - y_n \rangle}, \lambda_n\} & \text{if } \langle A w_n - A y_n, z_n - y_n \rangle > 0, \\ \lambda_n & \text{otherwise.} \end{cases} \quad (3.2)$$

Let  $n := n + 1$  and return to **Step 1**.

**Remark 3.1.** From (3.1), we get  $\lim_{n \rightarrow \infty} \frac{\alpha_n}{\beta_n} \|x_n - x_{n-1}\| = 0$ . Indeed, we have

$$\alpha_n \|x_n - x_{n-1}\| \leq \tau_n \quad \forall n \geq 1,$$

which together with  $\lim_{n \rightarrow \infty} \frac{\tau_n}{\beta_n} = 0$  implies that

$$\frac{\alpha_n}{\beta_n} \|x_n - x_{n-1}\| \leq \frac{\tau_n}{\beta_n} \rightarrow 0 \text{ as } n \rightarrow \infty.$$

**Lemma 3.1.** *Let  $\{\lambda_n\}$  be generated by (3.2). Then  $\{\lambda_n\}$  is a nonincreasing sequence with  $\lambda_n \geq \lambda := \min\{\lambda_1, \frac{\mu}{L}\} \forall n \geq 1$ , and  $\lim_{n \rightarrow \infty} \lambda_n \geq \lambda := \min\{\lambda_1, \frac{\mu}{L}\}$ .*

*Proof.* First, from (3.2) it is clear that  $\lambda_n \geq \lambda_{n+1} \forall n \geq 1$ . Also, observe that

$$\left. \begin{aligned} \frac{1}{2}(\|w_n - y_n\|^2 + \|z_n - y_n\|^2) &\geq \|w_n - y_n\| \|z_n - y_n\| \\ \langle A w_n - A y_n, z_n - y_n \rangle &\leq L \|w_n - y_n\| \|z_n - y_n\| \end{aligned} \right\} \Rightarrow \lambda_{n+1} \geq \min\left\{\lambda_n, \frac{\mu}{L}\right\}.$$

**Remark 3.2.** In terms of Lemmas 2.2 and 3.1, we claim that if  $w_n = y_n$  or  $A y_n = 0$ , then  $y_n$  is an element of  $\text{VI}(C, A)$ . Indeed, if  $w_n = y_n$  or  $A y_n = 0$ , then

$$0 = \|y_n - P_C(y_n - \lambda_n A y_n)\| \geq \|y_n - P_C(y_n - \lambda A y_n)\|.$$

Thus, the assertion is valid.

The following lemmas are quite helpful for the convergence analysis of our algorithm.

**Lemma 3.2.** *Let  $\{w_n\}, \{y_n\}, \{z_n\}$  be the sequences generated by Algorithm 3.1. Then*

$$\begin{aligned} \|z_n - p\|^2 &\leq \|w_n - p\|^2 - \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) \|w_n - y_n\|^2 \\ &\quad - \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) \|z_n - y_n\|^2, \quad \forall p \in \Omega. \end{aligned} \quad (3.3)$$

*Proof.* By the definition of  $\{\lambda_n\}$ , we claim that

$$2\langle Aw_n - Ay_n, z_n - y_n \rangle \leq \frac{\mu}{\lambda_{n+1}} \|w_n - y_n\|^2 + \frac{\mu}{\lambda_{n+1}} \|z_n - y_n\|^2 \quad \forall n \geq 1. \quad (3.4)$$

Indeed, if  $\langle Aw_n - Ay_n, z_n - y_n \rangle \leq 0$ , then inequality (3.4) holds. From (3.2), we get (3.4). Observe that, for each  $p \in \Omega \subset C \subset C_n$ ,

$$\begin{aligned} \|z_n - p\|^2 &= \|P_{C_n}(w_n - \lambda_n Ay_n) - P_{C_n}p\|^2 \leq \langle z_n - p, w_n - \lambda_n Ay_n - p \rangle \\ &= \frac{1}{2} \|z_n - p\|^2 + \frac{1}{2} \|w_n - p\|^2 - \frac{1}{2} \|z_n - w_n\|^2 - \langle z_n - p, \lambda_n Ay_n \rangle, \end{aligned}$$

which hence yields

$$\|z_n - p\|^2 \leq \|w_n - p\|^2 - \|z_n - w_n\|^2 - 2\langle z_n - p, \lambda_n Ay_n \rangle. \quad (3.5)$$

From  $p \in \text{VI}(C, A)$ , we get  $\langle Ap, x - p \rangle \geq 0 \quad \forall x \in C$ . By the pseudomonotonicity of  $A$  on  $C$  we have  $\langle Ax, x - p \rangle \geq 0 \quad \forall x \in C$ . Putting  $x := y_n \in C$  we get  $\langle Ay_n, p - y_n \rangle \leq 0$ . Thus,

$$\langle Ay_n, p - z_n \rangle = \langle Ay_n, p - y_n \rangle + \langle Ay_n, y_n - z_n \rangle \leq \langle Ay_n, y_n - z_n \rangle. \quad (3.6)$$

Substituting (3.6) for (3.5), we obtain

$$\|z_n - p\|^2 \leq \|w_n - p\|^2 - \|z_n - y_n\|^2 - \|y_n - w_n\|^2 + 2\langle w_n - \lambda_n Ay_n - y_n, z_n - y_n \rangle. \quad (3.7)$$

Since  $y_n = P_{C_n}(w_n - \lambda_n Aw_n)$  and  $z_n \in C_n$ , we have

$$2\langle w_n - \lambda_n Ay_n - y_n, z_n - y_n \rangle \leq 2\lambda_n \langle Aw_n - Ay_n, z_n - y_n \rangle,$$

which together with (3.4), implies that

$$2\langle w_n - \lambda_n Ay_n - y_n, z_n - y_n \rangle \leq \mu \frac{\lambda_n}{\lambda_{n+1}} \|w_n - y_n\|^2 + \mu \frac{\lambda_n}{\lambda_{n+1}} \|z_n - y_n\|^2. \quad (3.8)$$

Therefore, substituting (3.8) for (3.7), we infer that inequality (3.3) holds.

**Lemma 3.3.** *Let  $\{w_n\}, \{x_n\}, \{y_n\}$  be the sequences generated by Algorithm 3.1. If  $x_n - x_{n+1} \rightarrow 0$ ,  $w_n - x_n \rightarrow 0$  and  $w_n - y_n \rightarrow 0$  and  $\exists \{w_{n_k}\} \subset \{w_n\}$  such that  $w_{n_k} \rightarrow z \in H$ , then  $z \in \Omega$ .*

*Proof.* From Algorithm 3.1, we get  $w_n - x_n = T_n x_n - x_n + \alpha_n (T_n x_n - T_n x_{n-1}) \quad \forall n \geq 1$ . Hence

$$\|T_n x_n - x_n\| \leq \|w_n - x_n\| + \beta_n \cdot \frac{\alpha_n}{\beta_n} \|x_n - x_{n-1}\|.$$

Utilizing Remark 3.1 and the assumption  $w_n - x_n \rightarrow 0$ , we have

$$\lim_{n \rightarrow \infty} \|x_n - T_n x_n\| = 0. \quad (3.9)$$

Also, from  $y_n = P_C(w_n - \lambda_n Aw_n)$ , we have

$$\langle w_n - \lambda_n Aw_n - y_n, x - y_n \rangle \leq 0 \quad \forall x \in C.$$

Hence

$$\frac{1}{\lambda_n} \langle w_n - y_n, x - y_n \rangle + \langle Aw_n, y_n - w_n \rangle \leq \langle Aw_n, x - w_n \rangle \quad \forall x \in C. \quad (3.10)$$

Note that  $\{w_{n_k}\}$  is bounded. According to the Lipschitz continuity of  $A$ ,  $\{Aw_{n_k}\}$  is bounded. Note that  $\lambda_n \geq \min\{\lambda_1, \frac{\mu}{L}\}$ . So, from (3.10) we get

$$\liminf_{k \rightarrow \infty} \langle Aw_{n_k}, x - w_{n_k} \rangle \geq 0, \quad \forall x \in C.$$

Meantime, observe that

$$\langle Ay_n, x - y_n \rangle = \langle Ay_n - Aw_n, x - w_n \rangle + \langle Aw_n, x - w_n \rangle + \langle Ay_n, w_n - y_n \rangle.$$

Since  $w_n - y_n \rightarrow 0$ , we obtain from  $L$ -Lipschitz continuity of  $A$  that  $Aw_n - Ay_n \rightarrow 0$ , which together with (3.10) yields

$$\liminf_{k \rightarrow \infty} \langle Ay_{n_k}, x - y_{n_k} \rangle \geq 0 \quad \forall x \in C.$$

Next we show that  $\lim_{n \rightarrow \infty} \|x_n - T_l x_n\| = 0$  for  $l = 1, \dots, N$ . Indeed, for  $i = 1, \dots, N$ ,

$$\|x_n - T_{n+i} x_n\| \leq 2\|x_n - x_{n+i}\| + \|x_{n+i} - T_{n+i} x_{n+i}\|.$$

Hence from (3.9) and the assumption  $x_n - x_{n+1} \rightarrow 0$ , we get

$$\lim_{n \rightarrow \infty} \|x_n - T_{n+i} x_n\| = 0$$

for  $i = 1, \dots, N$ . This immediately implies that

$$\lim_{n \rightarrow \infty} \|x_n - T_l x_n\| = 0 \quad \text{for } l = 1, \dots, N. \quad (3.11)$$

We now take a sequence  $\{\varepsilon_k\} \subset (0, 1)$  satisfying  $\varepsilon_k \downarrow 0$  as  $k \rightarrow \infty$ . For all  $k \geq 1$ , we denote by  $m_k$  the smallest positive integer such that

$$\langle Ay_{n_j}, x - y_{n_j} \rangle + \varepsilon_k \geq 0 \quad \forall j \geq m_k. \quad (3.12)$$

Since  $\{\varepsilon_k\}$  is decreasing, it is clear that  $\{m_k\}$  is increasing. Noticing that  $\{y_{m_k}\} \subset C$  guarantees  $Ay_{m_k} \neq 0 \quad \forall k \geq 1$ , we set

$$u_{m_k} = \frac{Ay_{m_k}}{\|Ay_{m_k}\|^2}$$

and get  $\langle Ay_{m_k}, u_{m_k} \rangle = 1 \quad \forall k \geq 1$ . So, from (3.12), we get

$$\langle Ay_{m_k}, x + \varepsilon_k u_{m_k} - y_{m_k} \rangle \geq 0 \quad \forall k \geq 1.$$

Again from the pseudomonotonicity of  $A$ , we have

$$\langle A(x + \varepsilon_k u_{m_k}), x + \varepsilon_k u_{m_k} - y_{m_k} \rangle \geq 0 \quad \forall k \geq 1.$$

This immediately leads to

$$\langle Ax, x - y_{m_k} \rangle \geq \langle Ax - A(x + \varepsilon_k u_{m_k}), x + \varepsilon_k u_{m_k} - y_{m_k} \rangle - \varepsilon_k \langle Ax, u_{m_k} \rangle \quad \forall k \geq 1. \quad (3.13)$$

We claim that

$$\lim_{k \rightarrow \infty} \varepsilon_k u_{m_k} = 0.$$

Indeed, from  $w_{n_k} \rightharpoonup z$  and  $w_n - y_n \rightarrow 0$ , we obtain  $y_{n_k} \rightharpoonup z$ . So,  $\{y_n\} \subset C$  guarantees  $z \in C$ . Again from the sequentially weak continuity of  $A$ , we know that  $Ay_{n_k} \rightharpoonup Az$ . Thus, we have  $Az \neq 0$  (otherwise,  $z$  is a solution). Taking into account the sequentially weak lower semicontinuity of the norm  $\|\cdot\|$ , we get

$$0 < \|Az\| \leq \liminf_{k \rightarrow \infty} \|Ay_{n_k}\|.$$

Note that  $\{y_{m_k}\} \subset \{y_{n_k}\}$  and  $\varepsilon_k \downarrow 0$  as  $k \rightarrow \infty$ . So it follows that

$$0 \leq \limsup_{k \rightarrow \infty} \|\varepsilon_k u_{m_k}\| = \limsup_{k \rightarrow \infty} \frac{\varepsilon_k}{\|Ay_{m_k}\|} \leq \frac{\limsup_{k \rightarrow \infty} \varepsilon_k}{\liminf_{k \rightarrow \infty} \|Ay_{n_k}\|} = 0.$$

Hence we get  $\varepsilon_k u_{m_k} \rightarrow 0$ .

Next we show that  $z \in \Omega$ . Indeed, from  $w_n - x_n \rightarrow 0$  and  $w_{n_k} \rightharpoonup z$ , we get  $x_{n_k} \rightharpoonup z$ . From (3.11) we have  $x_{n_k} - T_l x_{n_k} \rightarrow 0$  for  $l = 1, \dots, N$ . Note that Lemma 2.5 guarantees the demiclosedness of  $I - T_l$  at zero for  $l = 1, \dots, N$ . Thus  $z \in \text{Fix}(T_l)$ . Since  $l$  is an

arbitrary element in the finite set  $\{1, \dots, N\}$ , we get  $z \in \bigcap_{i=1}^N \text{Fix}(T_i)$ .

On the other hand, letting  $k \rightarrow \infty$ , we deduce that the right hand side of (3.13) tends to zero by the uniform continuity of  $A$ , the boundedness of  $\{w_{m_k}\}, \{u_{m_k}\}$  and the limit  $\lim_{k \rightarrow \infty} \varepsilon_k u_{m_k} = 0$ . Thus, we get

$$\langle Ax, x - z \rangle = \liminf_{k \rightarrow \infty} \langle Ax, x - y_{m_k} \rangle \geq 0 \quad \forall x \in C.$$

By Lemma 2.3, we have  $z \in \text{VI}(C, A)$ . Therefore,

$$z \in \bigcap_{i=1}^N \text{Fix}(T_i) \cap \text{VI}(C, A) = \Omega.$$

This completes the proof.

**Theorem 3.1.** *Let  $\{x_n\}$  be the sequence generated by Algorithm 3.1.*

*Then  $x_n \rightarrow x^* \in \Omega \Leftrightarrow x_n - x_{n+1} \rightarrow 0$ , where  $x^* \in \Omega$  is a unique solution to the VIP:*

$$\langle (\rho F - f)x^*, p - x^* \rangle \geq 0 \quad \forall p \in \Omega.$$

*Proof.* We show that  $P_\Omega(f + I - \rho F)$  is a contraction. Indeed, for any  $x, y \in H$ , by Lemma 2.6, we have

$$\|P_\Omega(f + I - \rho F)x - P_\Omega(f + I - \rho F)y\| \leq [1 - (\tau - \delta)]\|x - y\|,$$

which implies that  $P_\Omega(f + I - \rho F)$  is a contraction. Banach's Contraction Mapping Principle guarantees that  $P_\Omega(f + I - \rho F)$  has a unique fixed point. Say  $x^* \in H$ , that is,  $x^* = P_\Omega(f + I - \rho F)x^*$ . Thus, there exists a unique solution

$$x^* \in \Omega = \bigcap_{i=0}^N \text{Fix}(T_i) \cap \text{VI}(C, A)$$



to the VIP

$$\langle (\rho F - f)x^*, p - x^* \rangle \geq 0 \quad \forall p \in \Omega. \quad (3.14)$$

It is clear that the necessity of the theorem is valid. Next we show the sufficiency of the theorem. To the aim, we assume  $x_n - x_{n+1} \rightarrow 0$  and divide the proof of the sufficiency into several steps.

**Step 1.** We show that  $\{x_n\}$  is bounded. Indeed, since

$$0 < \liminf_{n \rightarrow \infty} \gamma_n \leq \limsup_{n \rightarrow \infty} \gamma_n < 1,$$

we may assume, without loss of generality, that  $\{\gamma_n\} \subset [a, b] \subset (0, 1)$ . Take an arbitrary

$$p \in \Omega = \bigcap_{i=1}^N \text{Fix}(T_i) \cap \text{VI}(C, A).$$

Then  $T_n p = p \forall n \geq 1$ , and inequality (3.3) holds, i.e.,

$$\begin{aligned} \|z_n - p\|^2 &\leq \|w_n - p\|^2 - \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) \|w_n - y_n\|^2 \\ &\quad - \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) \|z_n - y_n\|^2 \quad \forall p \in \Omega. \end{aligned} \quad (3.15)$$

Since

$$\lim_{n \rightarrow \infty} \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) = 1 - \mu > 0,$$

we may assume, without loss of generality, that

$$1 - \mu \frac{\lambda_n}{\lambda_{n+1}} > 0 \quad \forall n \geq 1.$$

Therefore, we have

$$\|z_n - p\| \leq \|w_n - p\| \quad \forall n \geq 1. \quad (3.16)$$

It follows that

$$\begin{aligned} \|w_n - p\| &\leq \|T_n x_n - p\| + \alpha_n \|T_n x_n - T_n x_{n-1}\| \\ &\leq \|x_n - p\| + \beta_n \cdot \frac{\alpha_n}{\beta_n} \|x_n - x_{n-1}\|. \end{aligned} \quad (3.17)$$

According to Remark 3.1, we have

$$\frac{\alpha_n}{\beta_n} \|x_n - x_{n-1}\| \rightarrow 0$$

as  $n \rightarrow \infty$ , it follows that there exists a constant  $M_1 > 0$  such that

$$\frac{\alpha_n}{\beta_n} \|x_n - x_{n-1}\| \leq M_1 \quad \forall n \geq 1. \quad (3.18)$$

Combining (3.16), (3.17) and (3.18), we obtain

$$\|z_n - p\| \leq \|w_n - p\| \leq \|x_n - p\| + \beta_n M_1 \quad \forall n \geq 1. \quad (3.19)$$

Since  $\beta_n + \gamma_n < 1 \forall n \geq 1$ , we get

$$\frac{\beta_n}{1 - \gamma_n} < 1 \forall n \geq 1.$$

So, from Lemma 2.6 and (3.19) it follows that

$$\begin{aligned} \|x_{n+1} - p\| &\leq \beta_n \|f(x_n) - p\| + \gamma_n \|x_n - p\| \\ &\quad + (1 - \beta_n - \gamma_n) \left\| \left( \frac{1 - \gamma_n}{1 - \beta_n - \gamma_n} I - \frac{\beta_n}{1 - \beta_n - \gamma_n} \rho F \right) z_n - p \right\| \\ &\leq \beta_n (\|f(x_n) - f(p)\| + \|f(p) - p\|) + \gamma_n \|x_n - p\| \\ &\quad + (1 - \beta_n - \gamma_n) \left\| \left( \frac{1 - \gamma_n}{1 - \beta_n - \gamma_n} I - \frac{\beta_n}{1 - \beta_n - \gamma_n} \rho F \right) z_n - p \right\| \\ &\leq \beta_n (\delta \|x_n - p\| + \|f(p) - p\|) + \gamma_n \|x_n - p\| \\ &\quad + (1 - \gamma_n) \left\| \left( I - \frac{\beta_n}{1 - \gamma_n} \rho F \right) z_n - \left( I - \frac{\beta_n}{1 - \gamma_n} \rho F \right) p + \frac{\beta_n}{1 - \gamma_n} (I - \rho F) p \right\| \\ &\leq \beta_n (\delta \|x_n - p\| + \|f(p) - p\|) + \gamma_n \|x_n - p\| \\ &\quad + (1 - \gamma_n - \beta_n \tau) \|z_n - p\| + \beta_n \|(I - \rho F)p\| \\ &\leq [1 - \beta_n(\tau - \delta)] \|x_n - p\| + \beta_n(\tau - \delta) \cdot \frac{M_1 + \|f(p) - p\| + \|(I - \rho F)p\|}{\tau - \delta} \\ &\leq \max \left\{ \|x_n - p\|, \frac{M_1 + \|f(p) - p\| + \|(I - \rho F)p\|}{\tau - \delta} \right\}. \end{aligned}$$

By induction, we obtain

$$\|x_n - p\| \leq \max \left\{ \|x_1 - p\|, \frac{M_1 + \|f(p) - p\| + \|(I - \rho F)p\|}{\tau - \delta} \right\} \forall n \geq 1.$$

Thus,  $\{x_n\}$  is bounded, and so are the sequences  $\{w_n\}$ ,  $\{y_n\}$ ,  $\{z_n\}$ ,  $\{f(x_n)\}$ ,  $\{Fz_n\}$ ,  $\{T_n x_n\}$ .

**Step 2.** We show that

$$(1 - \beta_n \tau - \gamma_n) \left( 1 - \mu \frac{\lambda_n}{\lambda_{n+1}} \right) [\|w_n - y_n\|^2 + \|z_n - y_n\|^2] \leq \|x_n - p\|^2 - \|x_{n+1} - p\|^2 + \beta_n M_4,$$

for some  $M_4 > 0$ . Indeed, observe that

$$\begin{aligned} x_{n+1} - p &= \beta_n (f(x_n) - p) + \gamma_n (x_n - p) + (1 - \beta_n - \gamma_n) \\ &\quad \times \left\{ \frac{1 - \gamma_n}{1 - \beta_n - \gamma_n} \left[ \left( I - \frac{\beta_n}{1 - \gamma_n} \rho F \right) z_n - \left( I - \frac{\beta_n}{1 - \gamma_n} \rho F \right) p \right] \right. \\ &\quad \left. + \frac{\beta_n}{1 - \beta_n - \gamma_n} (I - \rho F) p \right\} \\ &= \beta_n (f(x_n) - f(p)) + \gamma_n (x_n - p) + (1 - \gamma_n) \\ &\quad \times \left[ \left( I - \frac{\beta_n}{1 - \gamma_n} \rho F \right) z_n - \left( I - \frac{\beta_n}{1 - \gamma_n} \rho F \right) p \right] + \beta_n (f - \rho F) p. \end{aligned}$$

Then by Lemma 2.6 and the convexity of the function  $h(t) = t^2 \forall t \in \mathbf{R}$ , we get

$$\begin{aligned} \|x_{n+1} - p\|^2 &\leq \beta_n \delta \|x_n - p\|^2 + \gamma_n \|x_n - p\|^2 + (1 - \beta_n \tau - \gamma_n) \|z_n - p\|^2 \\ &\quad + 2\beta_n \langle (f - \rho F)p, x_{n+1} - p \rangle \\ &\leq \beta_n \delta \|x_n - p\|^2 + \gamma_n \|x_n - p\|^2 \\ &\quad + (1 - \beta_n \tau - \gamma_n) \|z_n - p\|^2 + \beta_n M_2, \end{aligned} \quad (3.20)$$

where  $\sup_{n \geq 1} 2\|(f - \rho F)p\| \|x_n - p\| \leq M_2$  for some  $M_2 > 0$ . Substituting (3.15) for (3.20), we get

$$\begin{aligned} \|x_{n+1} - p\|^2 &\leq \beta_n \delta \|x_n - p\|^2 + \gamma_n \|x_n - p\|^2 + (1 - \beta_n \tau - \gamma_n) [\|w_n - p\|^2 \\ &\quad - \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) \|w_n - y_n\|^2 \\ &\quad - \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) \|z_n - y_n\|^2] + \beta_n M_2. \end{aligned} \quad (3.21)$$

Also, from (3.19) we have

$$\|w_n - p\|^2 \leq \|x_n - p\|^2 + \beta_n M_3, \quad (3.22)$$

where  $\sup_{n \geq 1} (2M_1 \|x_n - p\| + \beta_n M_1^2) \leq M_3$  for some  $M_3 > 0$ . Combining (3.21) and (3.22), we obtain

$$\begin{aligned} \|x_{n+1} - p\|^2 &\leq \beta_n \delta \|x_n - p\|^2 + \gamma_n \|x_n - p\|^2 + (1 - \beta_n \tau - \gamma_n) [\|x_n - p\|^2 + \beta_n M_3] \\ &\quad - (1 - \beta_n \tau - \gamma_n) \left[ \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) \|w_n - y_n\|^2 \right. \\ &\quad \left. + \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) \|z_n - y_n\|^2 \right] + \beta_n M_2 \\ &\leq \|x_n - p\|^2 - (1 - \beta_n \tau - \gamma_n) \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) [\|w_n - y_n\|^2 \\ &\quad + \|z_n - y_n\|^2] + \beta_n M_4, \end{aligned}$$

where  $M_4 := M_2 + M_3$ . This immediately implies that

$$(1 - \beta_n \tau - \gamma_n) \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) [\|w_n - y_n\|^2 + \|z_n - y_n\|^2] \leq \|x_n - p\|^2 - \|x_{n+1} - p\|^2 + \beta_n M_4. \quad (3.23)$$

**Step 3.** We show that

$$\begin{aligned} \|x_{n+1} - p\|^2 &\leq [1 - \beta_n (\tau - \delta)] \|x_n - p\|^2 \\ &\quad + \beta_n (\tau - \delta) \left[ \frac{2}{\tau - \delta} \langle (f - \rho F)p, x_{n+1} - p \rangle + \frac{3M}{\tau - \delta} \cdot \frac{\alpha_n}{\beta_n} \cdot \|x_n - x_{n-1}\| \right] \end{aligned}$$

for some  $M > 0$ . Indeed, we have

$$\|w_n - p\|^2 \leq \|x_n - p\|^2 + \alpha_n \|x_n - x_{n-1}\| [2\|x_n - p\| + \alpha_n \|x_n - x_{n-1}\|]. \quad (3.24)$$

Combining (3.19), (3.20) and (3.24), we have

$$\begin{aligned}
\|x_{n+1} - p\|^2 &\leq \beta_n \delta \|x_n - p\|^2 + \gamma_n \|x_n - p\|^2 + (1 - \beta_n \tau - \gamma_n) \{\|x_n - p\|^2 \\
&\quad + \alpha_n \|x_n - x_{n-1}\| [2\|x_n - p\| + \alpha_n \|x_n - x_{n-1}\|]\} \\
&\quad + 2\beta_n \langle (f - \rho F)p, x_{n+1} - p \rangle \\
&\leq [1 - \beta_n(\tau - \delta)] \|x_n - p\|^2 \\
&\quad + \alpha_n \|x_n - x_{n-1}\| [2\|x_n - p\| + \alpha_n \|x_n - x_{n-1}\|] \\
&\quad + 2\beta_n \langle (f - \rho F)p, x_{n+1} - p \rangle \\
&\leq [1 - \beta_n(\tau - \delta)] \|x_n - p\|^2 \\
&\quad + \beta_n(\tau - \delta) \left[ \frac{2\langle (f - \rho F)p, x_{n+1} - p \rangle}{\tau - \delta} + \frac{3M}{\tau - \delta} \cdot \frac{\alpha_n}{\beta_n} \cdot \|x_n - x_{n-1}\| \right],
\end{aligned} \tag{3.25}$$

where  $\sup_{n \geq 1} \{\|x_n - p\|, \alpha_n \|x_n - x_{n-1}\|\} \leq M$  for some  $M > 0$ .

**Step 4.** We show that  $\{x_n\}$  converges strongly to a unique solution  $x^* \in \Omega$  to the VIP (3.14). Indeed, putting  $p = x^*$ , we deduce from (3.25) that

$$\begin{aligned}
\|x_{n+1} - x^*\|^2 &\leq [1 - \beta_n(\tau - \delta)] \|x_n - x^*\|^2 \\
&\quad + \beta_n(\tau - \delta) \left[ \frac{2\langle (f - \rho F)x^*, x_{n+1} - x^* \rangle}{\tau - \delta} + \frac{3M}{\tau - \delta} \cdot \frac{\alpha_n}{\beta_n} \cdot \|x_n - x_{n-1}\| \right].
\end{aligned} \tag{3.26}$$

By Lemma 2.4, it suffices to show that

$$\limsup_{n \rightarrow \infty} \langle (f - \rho F)x^*, x_{n+1} - x^* \rangle \leq 0.$$

From (3.23),  $x_n - x_{n+1} \rightarrow 0$ ,  $\beta_n \rightarrow 0$  and  $\{\gamma_n\} \subset [a, b] \subset (0, 1)$ , we obtain

$$\begin{aligned}
&\limsup_{n \rightarrow \infty} (1 - \beta_n \tau - b) \left(1 - \mu \frac{\lambda_n}{\lambda_{n+1}}\right) [\|w_n - y_n\|^2 + \|z_n - y_n\|^2] \\
&\leq \limsup_{n \rightarrow \infty} [\|x_n - p\|^2 - \|x_{n+1} - p\|^2 + \beta_n M_4] \\
&\leq \limsup_{n \rightarrow \infty} (\|x_n - p\| + \|x_{n+1} - p\|) \|x_n - x_{n+1}\| = 0.
\end{aligned}$$

This immediately implies that

$$\lim_{n \rightarrow \infty} \|w_n - y_n\| = 0 \quad \text{and} \quad \lim_{n \rightarrow \infty} \|z_n - y_n\| = 0. \tag{3.27}$$

Thus, we get

$$\lim_{n \rightarrow \infty} \|z_n - w_n\| = 0. \tag{3.28}$$

Also, from Algorithm 3.1 we get

$$x_{n+1} - x_n = \beta_n (f(x_n) - \rho F z_n) + (1 - \gamma_n)(z_n - x_n),$$

which hence implies that

$$\|z_n - x_n\| \leq \frac{1}{1 - b} [\|x_{n+1} - x_n\| + \beta_n \|f(x_n) - \rho F z_n\|].$$

From  $x_n - x_{n+1} \rightarrow 0$ ,  $\beta_n \rightarrow 0$  and the boundedness of  $\{f(x_n)\}$  and  $\{Fz_n\}$  we conclude that

$$\lim_{n \rightarrow \infty} \|z_n - x_n\| = 0. \quad (3.29)$$

From the boundedness of  $\{x_n\}$ , it follows that there exists a subsequence  $\{x_{n_k}\}$  of  $\{x_n\}$  such that

$$\limsup_{n \rightarrow \infty} \langle (f - \rho F)x^*, x_n - x^* \rangle = \lim_{k \rightarrow \infty} \langle (f - \rho F)x^*, x_{n_k} - x^* \rangle. \quad (3.30)$$

Since  $H$  is reflexive and  $\{x_n\}$  is bounded, we may assume, without loss of generality, that  $x_{n_k} \rightharpoonup \tilde{x}$ . Hence from (3.30) we get

$$\begin{aligned} \limsup_{n \rightarrow \infty} \langle (f - \rho F)x^*, x_n - x^* \rangle &= \lim_{k \rightarrow \infty} \langle (f - \rho F)x^*, x_{n_k} - x^* \rangle \\ &= \langle (f - \rho F)x^*, \tilde{x} - x^* \rangle. \end{aligned} \quad (3.31)$$

Also, from (3.28) and (3.29) we have

$$\|w_n - x_n\| \leq \|w_n - z_n\| + \|z_n - x_n\| \rightarrow 0 \quad (n \rightarrow \infty),$$

which together with  $x_{n_k} \rightharpoonup \tilde{x}$ , implies that  $w_{n_k} \rightharpoonup \tilde{x}$ .

Since  $x_n - x_{n+1} \rightarrow 0$ ,  $w_n - x_n \rightarrow 0$ ,  $w_n - y_n \rightarrow 0$  and  $w_{n_k} \rightharpoonup \tilde{x}$ , by Lemma 3.3 we infer that  $\tilde{x} \in \Omega$ . Hence from (3.14) and (3.31) we get

$$\limsup_{n \rightarrow \infty} \langle (f - \rho F)x^*, x_n - x^* \rangle = \langle (f - \rho F)x^*, \tilde{x} - x^* \rangle \leq 0, \quad (3.32)$$

which immediately leads to

$$\begin{aligned} &\limsup_{n \rightarrow \infty} \langle (f - \rho F)x^*, x_{n+1} - x^* \rangle \\ &\leq \limsup_{n \rightarrow \infty} [\|(f - \rho F)x^*\| \|x_{n+1} - x_n\| + \langle (f - \rho F)x^*, x_n - x^* \rangle] \leq 0. \end{aligned} \quad (3.33)$$

Note that  $\{\beta_n(\tau - \delta)\} \subset [0, 1]$ ,  $\sum_{n=1}^{\infty} \beta_n(\tau - \delta) = \infty$ , and

$$\limsup_{n \rightarrow \infty} \left[ \frac{2\langle (f - \rho F)x^*, x_{n+1} - x^* \rangle}{\tau - \delta} + \frac{3M}{\tau - \delta} \cdot \frac{\alpha_n}{\beta_n} \cdot \|x_n - x_{n-1}\| \right] \leq 0.$$

Consequently, applying Lemma 2.4 to (3.26), we have  $\lim_{n \rightarrow 0} \|x_n - x^*\| = 0$ .

This completes the proof.

#### 4. AN APPLICATION

In this section, our main result is applied to find a common solution of the fractional programming and fixed-point problems. Since the exact solution of the problem is not known, we make use of  $\|x_{n+1} - x_n\|$  to measure the error of the  $n$ -th iteration, which also serves as the role of checking whether or not the proposed algorithm converges to the solution.

The initial point  $x_0$  is randomly chosen in  $\mathbf{R}^m$ . Take

$$f(x) = F(x) = \frac{1}{2}x, \quad \mu = 0.3, \quad \beta_n = \frac{1}{n+1}, \quad \alpha = 0.1, \quad \gamma_n = \frac{1}{3}, \quad \rho = 2,$$

and

$$\alpha_n = \begin{cases} \min \left\{ \frac{\beta_n^2}{\|x_n - x_{n-1}\|}, \alpha \right\} & \text{if } x_n \neq x_{n-1}, \\ \alpha & \text{otherwise.} \end{cases}$$

Then we know that  $\kappa = \eta = \frac{1}{2}$ , and

$$\tau = 1 - \sqrt{1 - \rho(2\eta - \rho\kappa^2)} = 1 - \sqrt{1 - 2 \left( 2 \cdot \frac{1}{2} - 2 \left( \frac{1}{2} \right)^2 \right)} = 1 \in (0, 1].$$

First, we set the operator  $\Gamma(x) := Mx + q$ , which comes from [11] and has been considered by many authors for applicable examples (see, for example [16]), where  $M = BB^T + D + G$ , and  $B$  is an  $m \times m$  matrix,  $D$  is an  $m \times m$  skew-symmetric matrix,  $G$  is an  $m \times m$  diagonal matrix, whose diagonal entries are nonnegative (so  $M$  is positive semidefinite),  $q$  is a vector in  $\mathbf{R}^m$ . The feasible set  $C \subset \mathbf{R}^m$  is a closed and convex subset defined by  $C := \{x \in \mathbf{R}^m : Hx \leq d\}$ , where  $H$  is an  $l \times m$  matrix and  $d$  is a nonnegative vector. It is clear that  $\Gamma$  is  $\beta$ -monotone and  $L$ -Lipschitz-continuous with  $\beta = \min\{\text{eig}(\Gamma)\}$  and  $L = \max\{\text{eig}(\Gamma)\}$ . Next we give the operator  $A$ .

Consider the following fractional programming problem:

$$\begin{aligned} \min g(x) &= \frac{x^T Qx + a^T x + a_0}{b^T x + b_0}, \\ \text{subject to } x \in X &:= \{x \in \mathbf{R}^4 : b^T x + b_0 > 0\}, \end{aligned}$$

where

$$Q = \begin{pmatrix} 5 & -1 & 2 & 0 \\ -1 & 5 & -1 & 3 \\ 2 & -1 & 3 & 0 \\ 0 & 3 & 0 & 5 \end{pmatrix}, \quad a = \begin{pmatrix} 1 \\ -2 \\ -2 \\ 1 \end{pmatrix}, \quad b = \begin{pmatrix} 2 \\ 1 \\ 1 \\ 0 \end{pmatrix}, \quad a_0 = -2, \quad b_0 = 4.$$

It is easy to verify that  $Q$  is symmetric and positive definite in  $\mathbf{R}^4$  and consequently  $g$  is pseudoconvex on  $X = \{x \in \mathbf{R}^4 : b^T x + b_0 > 0\}$ . Then

$$Ax := \nabla g(x) = \frac{(b^T x + b_0)(2Qx + a) - b(x^T Qx + a^T x + a_0)}{(b^T x + b_0)^2}.$$

It is known that  $A$  is pseudomonotone (see, e.g., [12, 15] for more details). Now, we give a nonexpansive mapping  $T_1 : H \rightarrow C$  defined by  $T_1 x = P_C x \forall x \in H$ . Thus, Algorithm 3.1 can be rewritten as follows:

$$\begin{cases} w_n = T_1 x_n + \alpha_n(T_1 x_n - T_1 x_{n-1}), \\ y_n = P_C(w_n - \lambda_n A w_n), \\ z_n = P_{C_n}(w_n - \lambda_n A y_n), \\ x_{n+1} = \frac{1}{n+1} \cdot \frac{1}{2} x_n + \frac{1}{3} x_n + \left(\frac{n}{n+1} I - \frac{1}{3} I\right) z_n \quad \forall n \geq 1, \end{cases}$$

where for each  $n \geq 1$ ,  $C_n$  and  $\lambda_n$  are chosen as in Algorithm 3.1. Therefore, utilizing Theorem 3.1, we know that  $\{x_n\}$  converges to a common solution of the fractional programming problem and the fixed-point problem of  $T_1$  provided  $\|x_n - x_{n+1}\| \rightarrow 0$ .

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