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Linesearch Algorithms for Split Generalized Equilibrium Problems and Two Families of Strict Pseudo-Contraction Mappings

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Abstract : In this paper, we study linesearch algorithms finding a common solution of a split generalized equilibrium problems and two families of strict pseudocontraction mappings in Hilbert spaces. Weak and strong convergence theorems for such algorithms are studied. Our results improve many known recent results in the literature.

Keywords : split generalized equilibrium problem; strict pseudo-contraction; linesearch rule; weak and strong convergence.

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1 Introduction

Throughout this paper, let \mathbb{R} denote the set of all real numbers, \mathbb{N} denote the set of all positive integer numbers. Let \mathcal{H}_1 and \mathcal{H}_2 be two real Hilbert spaces with inner product $\langle \cdot, \cdot \rangle$ and norm $\|\cdot\|$. Let C and Q be nonempty closed convex subsets of \mathcal{H}_1 and \mathcal{H}_2 , respectively, and $A: \mathcal{H}_1 \to \mathcal{H}_2$ be a bounded linear operator. The split feasible problem (SFP) in the sense of Censor and Elfving [1] is to find $x^* \in C$ such that $Ax^* \in Q$. It turns out that SFP provides a unified framework for study

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of many sigificant real-world problems such as in signal processing, medical image reconstruction, intensity-modulated radiation therapy, *et cetera*; see, for example, [2]. To find a solution of SFP in finite-dimensional Hilbert spaces, a basic scheme proposed by Byrne [3], called the CQ-algorithm, is defined as follows:

$$x^{k+1} = P_C(x^k + \gamma A^T (P_Q - I)Ax^k),$$

where I is the identity, mapping, and P_C is projection mapping onto C. Xu [4] investigated the SEP setting in infinite-dimensional Hilbert spaces. In this case, the CQ-algorithm becomes

$$x^{k+1} = P_C(x^k + \gamma A^*(P_Q - I)Ax^k),$$

where A^* is the adjoint operator of A.

The split feasibility problem when C or Q are fixed points of mappings or common fixed points of mappings and solutions of variational inequality problems was considered in some recent research papers; see, for instance, [5].

In 2011, Moudafi [6] introduced the following split equilibrium problem (SEP, for short): Let $g_1 : C \times C \to \mathbb{R}$ and $g_2 : Q \times Q \to \mathbb{R}$ are two bifunctions; $A : \mathcal{H}_1 \to \mathcal{H}_2$ be a bounded linear operator, then the SEP is to find $x^* \in C$ such that

$$g_1(x^*, x) \ge 0, \forall x \in C, \tag{1.1}$$

and such that

$$y^* = Ax^* \in Q \text{ solves } g_2(y^*, y) \ge 0, \forall y \in Q.$$

$$(1.2)$$

When looked separately, (1.1) is the classical equilibrium problem EP and we denoted its solution set by EP(C, g_1). The SEP (1.1) and (1.2) constitutes a pair of equilibrium problems which have to be solved so that the image $y^* = Ax^*$ under a given bounded linear operator A, of the solution x^* of the EP (1.1) in \mathcal{H}_1 is the solution of another EP(1.2) in another space \mathcal{H}_2 , we denote the solution set of EP(1.2) by EP(Q, g_2).

The solution set of SEP (1.1) and (1.2) is denoted by

$$\Omega = \{ p \in \operatorname{EP}(C, g_1) : Ap \in \operatorname{EP}(Q, g_2) \}.$$

See [7] for more detail on equilibrium problems.

In 2013, Kazmi and Rizvi [8] proposed the split generalized equilibrium problem (SGEP, for short): SGEP is the problem of finding $x^* \in C$ such that

$$g_1(x^*, x) + h_1(x^*, x) \ge 0, \forall x \in C,$$
(1.3)

and such that

$$y^* = Ax^* \in Q$$
 solves $g_2(y^*, y) + h_2(y^*, y) \ge 0, \forall y \in Q.$ (1.4)

where $g_1, h_1 : C \times C \to \mathbb{R}$ and $g_2, h_2 : Q \times Q \to \mathbb{R}$ are nonlinear bifunctions and $A : H_1 \to H_2$ is a bounded linear operator. We denote the solution set of SGEP

(1.3) and (1.4) by $S_{\text{GEP}}(C, g_1, h_1)$ and $S_{\text{GEP}}(Q, g_2, h_2)$, respectively. The solution set of SGEP is denoted by

$$S_{\text{GEP}} := \{ z \in C : z \in S_{\text{GEP}}(C, g_1, h_1) \text{ such that } Az \in S_{\text{GEP}}(Q, g_2, h_2) \}$$

If $h_1 = 0$ and $h_2 = 0$, then SGEP reduces to SEP. If $h_1 = h_2 = 0$ and $g_2 = 0$, then SGEP reduces to EP.

On the other hand, many researchers have been proposed numerical algorithms for finding a common element of the set of solutions of monotone equilibrium problems and the set of fixed points of nonexpansive mappings;, for example, [9], [10] and the references therein.

Recently, Dinh, *et al.* [11] studied the split equilibrium problem and nonexpansive mapping involving pseudomonotone and monotone equilibrium bifunctions in real Hilbert spaces, that is, let $f: C \times C \to \mathbb{R}$ be a pseudomonotone bifunction with respect to its solution set, $g: Q \times Q \to \mathbb{R}$ be a monotone bifunction, and $S: C \to C$ and $T: Q \to Q$ be nonexpansive mappings. They stated problem as follows (SEPNM(C, Q, A, f, g, S, T)) or SEPNM for short):

Find $x^* \in C$ such that $x^* \in S_{EP}(C, f) \cap Fix(S)$ and $Ax^* \in S_{EP}(Q, g) \cap Fix(T)$,

where Fix(S) and Fix(T) are the fixed points of the mappings S and T, respectively. They combined the extragradient method incorporated with the Armijo linesearch rule for solving equilibrium problem and the Mann method for finding a fixed point of an nonexpansive mapping. In addition, they combined the proposed algorithm with hybrid cutting technique to get a strong convergence algorithm for SEPNM.

We recall that a mapping $S: C \to C$ is said to be *L*-strict pseudo-contractive (in the sense of Browder-Petryshyn) if there exists $L \in [0, 1)$ such that

$$||S(x) - S(y)||^2 \le ||x - y||^2 + L||(I - S)(x) - (I - S)(y)||^2, \forall x, y \in C,$$
(1.5)

where I is the identity mapping on \mathcal{H} . Note that the class of strict pseudocontractions includes the class of nonexpansive mappings, which are mappings Son C such that

$$||S(x) - S(y)|| \le ||x - y||, \forall x, y \in C.$$

The problem of finding fixed points of nonexpansive mappings via Mann's algorithm [12] has been widely investigated in the literature (see e.g. [13]). Mann's algorithm generates, on initializing with an arbitrary $x_1 \in C$, a sequence according to the recursive formula

$$x_1 \in C, x_{n+1} = \alpha_n x_n + (1 - \alpha_n) S x_n, \forall n \ge 1,$$

where $\{\alpha_n\}_{n\in\mathbb{N}} \subset (0,1)$. Furthermore, iterative algorithms for strict pseudocontractions are still less developed than those for nonexpansive mappings, despite the pioneering work of Browder and Petryshyn [14] dating from 1967. However, strict pseudo-contractions have many applications, due to their ties with inverse stronglymonotone operators. Indeed, if A is a strongly monotone operator, then S = I - A is a strict pseudo-contraction, and so we can redraft a problem of zeros for A in a fixed point problem for S (see e.g. [15], [16]).

This paper, we propose on a split generalized equilibrium problems and two families of strict pseudo-contraction mappings in Hilbert spaces. In detail, let $f : C \times C \to \mathbb{R}$ be a pseudomonotone bifunction with respect to its solution set, $g, h : Q \times Q \to \mathbb{R}$ be a monotone bifunction, and S_i and T_j are L_i and L'_j -strict pseudo-contractions for some $0 \leq L_i < 1$ and $0 \leq L'_j < 1$, respectively, where for each $i \in \{1, 2, \ldots, p\}$ and $j \in \{1, 2, \ldots, p'\}$. The problem considered in this paper can be stated as follows:

Find
$$\bar{x} \in C$$
 such that $\bar{x} \in S_{EP}(C, f) \cap \left(\cap_{i=1}^{p} \operatorname{Fix}(S_{i}, C) \right)$

and

$$A\bar{x} \in S_{\text{GEP}}(Q, g, h) \cap \left(\bigcap_{j=1}^{p'} \text{Fix}(T_j, Q) \right),$$

where $\operatorname{Fix}(S_i, C)$ is the set of the fixed points set of the mapping $S_i(i = 1, \ldots, p)$ and $\operatorname{Fix}(T_j, Q)$ is the set of the fixed points set of the mapping $T_j(j = 1, \ldots, p')$.

In this paper, motivated and inspired by the work of Dinh, *et al.* [11] and by research going on this area, we shall introduce a linesearch algorithms for split generalized equilibrium problems and two families of strict pseudo-contraction mappings in Hilbert space. Weak and strong convergence theorems for such algorithms are studied. Our results complement many known recent results in the literature.

2 Preliminaries

Let C be a nonempty convex subset of a Hilbert space \mathcal{H} . We write $x^k \to x$ to indicate that the sequence $\{x^k\}$ converges weakly to x as $k \to \infty$, and $x^k \to x$ to indicate that the sequence $\{x^k\}$ converges strongly to x as $k \to \infty$. Since C is closed, convex, for any $x \in \mathcal{H}$, there exists an uniquely point in C, denoted by $P_C(x)$ satisfying

$$||x - P_C(x)|| \le ||x - y||, \forall y \in C.$$

 P_C is called the metric projection of \mathcal{H} to C.

Lemma 2.1. Suppose that C is a nonempty closed convex subset in \mathcal{H} . Then P_C has the following properties:

- (a) $z = P_C(x)$ if and only if $\langle x z, y z \rangle \leq 0, \forall y \in C;$
- (b) $\langle x y, P_C(x) P_C(y) \rangle \ge ||P_C(x) P_C(y)||^2, \quad \forall x, y \in \mathcal{H};$
- (c) $\langle x P_C(x), P_C(x) y \rangle \ge 0, \quad \forall x \in \mathcal{H}, y \in C;$
- (d) $||x y||^2 \ge ||x P_C(x)||^2 + ||y P_C(x)||^2, \quad \forall x \in \mathcal{H}, y \in C.$

Lemma 2.2. Let \mathcal{H} be a real Hilbert space. Then, for all $x, y \in \mathcal{H}$ and $\alpha \in [0, 1]$, we have

$$\|\alpha x + (1-\alpha)y\|^2 = \alpha \|x\|^2 + (1-\alpha)\|y\|^2 - \alpha(1-\alpha)\|x-y\|^2.$$

Lemma 2.3 (Opial's condition). For any sequence $\{x^k\} \subset \mathcal{H}$ with $x^k \rightharpoonup x$, we have the inequality

$$\liminf_{k \to +\infty} \|x^k - x\| < \liminf_{k \to +\infty} \|x^k - y\|$$

hold for all $y \in \mathcal{H}$ such that $y \neq x$.

The concept of strict pseudo-contraction is considered in [17], which defined as follows.

Definition 2.4. We say that an operator $S : \mathcal{H} \to \mathcal{H}$ is *demiclosed at* 0 if, for any sequence $\{x^k\}$ such that $x^k \to x$ and $Sx^k \to 0$ as $k \to \infty$, we have Sx = 0.

The following proposition lists some useful properties of a strict pseudo-contraction mapping.

Proposition 2.5. [17] Let C be a nonempty closed convex subset of a real Hilbert space \mathcal{H} , $S: C \to C$ be a L-strict pseudo-contraction and for each $i = 1, \dots, p, S_i: C \to C$ is a L_i -strict pseudo-contraction for some $0 \leq L_i < 1$. Then:

1. S satisfies the following Lipschitz condition:

$$||S(x) - S(y)|| \le \frac{1+L}{1-L} ||x - y||, \ \forall x, y \in C;$$

- 2. I S is demiclosed at 0. That is, if the sequence $\{x^k\}$ contains in C such that $x^k \rightarrow \bar{x}$ and $(I S)(x^k) \rightarrow 0$ then $(I S)(\bar{x}) = 0$;
- 3. The set of fixed points Fix(S) is closed and convex;
- 4. If $\eta_i > 0 (i = 1, \dots, p)$ and $\sum_{i=1}^p \eta_i = 1$ then $\sum_{i=1}^p \eta_i S_i$ is a \overline{L} -strict pseudocontraction with $\overline{L} := \max\{L_i : 1 \le i \le p\};$
- 5. If η_i is chosen as in (iv) and $\{S_i : i = 1, ..., p\}$ has a common fixed point then:

$$\operatorname{Fix}\left(\sum_{i=1}^{p} \eta_i S_i\right) = \bigcap_{i=1}^{p} \operatorname{Fix}(S_i, C).$$

Lemma 2.6. [18] Suppose that $\{\alpha_k\}$ and $\{\beta_k\}$ are two sequences of nonnegative real numbers such that

$$\alpha_{k+1} \le \alpha_k + \beta_k, \quad k \ge 0,$$

where $\sum_{k=0}^{\infty} \beta_k < \infty$. Then the sequence $\{\alpha_k\}$ is convergent.

Now, we assume that the equilibrium bifunction f, g and h satisfy the following assumptions I, II and III, respectively.

Assumption I : Assume that $f : C \times C \to \mathbb{R}$, let us assume that f satisfies the following conditions:

(A1) f is pseudomonotone on C, that is, if $f(x,y) \ge 0$ implies $f(y,x) \le 0$ for all $x, y \in C$;

(A2) $f(x, \cdot)$ is convex and subdifferentiable on C for all $x \in C$;

(A3) f is jointly weakly continuous on $C \times C$ in the sense that, if $x, y \in C$ and $\{x^k\}$ and $\{y^k\} \subset C$ converge weakly to x and y, respectively, then $f(x^k, y^k) \rightarrow f(x, y)$ as $k \rightarrow \infty$.

Assumption II : Assume that $g : Q \times Q \to \mathbb{R}$, let us assume that g satisfies the following conditions:

(B1) g(x, x) = 0 for all $x \in Q$;

(B2) g is monotone, i.e, $g(x, y) + g(y, x) \le 0$ for all $x, y \in Q$;

(B3) for each $x, y, z \in Q$, $\lim_{t \to 0} g(tz + (1 - t)x, y) \le g(x, y)$;

(B4) for each $x \in Q$, $y \mapsto g(x, y)$ is convex and lower semicontinuous.

Assumption III : Let the bifunction $h: Q \times Q \to \mathbb{R}$ be satisfied

(C1) $h(x,x) \ge 0, \forall x \in Q$,

(C2) For each $y \in Q$ fixed, the function $x \mapsto h(x, y)$ is upper semicontinuous,

(C3) For each $x \in Q$ fixed, the function $y \mapsto h(x, y)$ is convex and lower semicontinuous,

Assumption IV : For fixed r > 0 and $z \in C$, there exists a nonempty compact convex subset K of H and $x \in C \cap K$ such that

$$f(x,y) + h(y,x) + \frac{1}{r} \langle y - x, x - z \rangle < 0, \forall y \in C \backslash K.$$

Let f be an equilibrium bifunction defined on $C \times C$. For $x, y \in C$, we denote by $\partial f(x, y)$ the subgradient of the convex function $f(x, \cdot)$ at y, that is,

$$\partial f(x,y) := \left\{ \hat{t} \in \mathcal{H} : f(x,z) \ge f(x,y) + \langle \hat{t}, z - y \rangle, \text{ for all } z \in C \right\}.$$

In particular,

$$\partial f(x,x) := \left\{ \hat{t} \in \mathcal{H} : f(x,z) \ge \langle \hat{t}, z - y \rangle, \text{ for all } z \in C \right\}.$$

Let Δ be an open convex set containing C. The next lemma can be considered as infinite-dimensional version of Theorem 24.5 in [19].

Lemma 2.7. [20] Let $f : \Delta \times \Delta \to \mathbb{R}$ be an equilibrium bifunction satisfying condition (B2) on Δ and (B4) on C. Let $\bar{x}, \bar{y} \in \Delta$, and let $\{x^k\}, \{y^k\}$ be two sequences in Δ converging weakly to \bar{x}, \bar{y} , respectively. Then, for any $\varepsilon > 0$, there exist $\eta > 0$ and $k_{\varepsilon} \in \mathbb{N}$ such that

$$\partial f(x^k, y^k) \subset \partial f(\bar{x}, \bar{y}) + \frac{\varepsilon}{\eta} B$$

for every $k > k_{\varepsilon}$, where B denotes the closed unit ball in \mathcal{H} .

Lemma 2.8. [11] Let the equilibrium bifunction f satisfy assumptions (B2) on Δ and (B4) on C, and $\{x^k\} \subset C, 0 < \rho' \leq \rho'', \{\rho_k\} \subset [\rho', \rho'']$. Consider the sequence $\{y^k\}$ defined as

$$y^{k} = argmin \Big\{ f(x^{k}, y) + \frac{1}{2\rho_{k}} \|y - x^{k}\|^{2} : y \in C \Big\}.$$

If $\{x^k\}$ is bounded, then $\{y^k\}$ is also bounded.

Lemma 2.9. [21] Let g satisfy Assumption II. Then, for all r > 0 and $u \in \mathcal{H}$, there exists $w \in Q$ such that

$$g(w,v) + \frac{1}{\alpha} \langle v - w, w - u \rangle \ge 0, \forall v \in Q.$$

Lemma 2.10. [22] Assume that the bifunctions $g, h : Q \times Q \to \mathbb{R}$ satisfy Assumption II, Assumption III, respectively. For $\alpha > 0$ and $x \in \mathcal{H}$, define a mapping $T_{\alpha}^{(g,h)} : \mathcal{H} \to Q$ as follows:

$$T_{\alpha}^{(g,h)}(x) = \left\{ z \in Q : g(z,y) + h(z,y) + \frac{1}{\alpha} \langle y - z, z - x \rangle \ge 0, \ \forall y \in Q \right\}.$$

Then, the following hold:

(i) $T_{\alpha}^{(g,h)}(x) \neq \emptyset$. (ii) $T_{\alpha}^{(g,h)}$ is single-valued.

(iii) $T_{\alpha}^{(g,h)}$ is firmly nonexpansive, i.e., for any $x, y \in \mathcal{H}$, $||T_{\alpha}^{(g,h)}||_{\mathcal{H}} = T_{\alpha}^{(g,h)} = T_{\alpha}^{(g,h)} = T_{\alpha}^{(g,h)} = T_{\alpha}^{(g,h)}$

$$||T_{\alpha}^{(g,n)}x - T_{\alpha}^{(g,n)}y||^{2} \le \langle T_{\alpha}^{(g,n)}x - T_{\alpha}^{(g,n)}y, x - y \rangle$$

- (iv) $\operatorname{Fix}(T_{\alpha}^{(g,h)}) = S_{\operatorname{GEP}}(Q,g,h).$
- (v) $S_{\text{GEP}}(Q, g, h)$ is compact and convex.

Lemma 2.11. [23] Let $g : Q \times Q \to \mathbb{R}$ be a bifunction satisfying Assumption II hold and let $T_{\alpha}^{(g,h)}$ be defined as in Lemma 2.10 with $\alpha, \beta > 0$. Then, for any $x, y \in \mathcal{H}$ and

$$\|T_{\alpha}^{(g,h)}x - T_{\beta}^{(g,h)}y\| \le \|x - y\| + \left|\frac{\beta - \alpha}{\beta}\right| \|T_{\beta}^{(g,h)}x - x\|.$$

Lemma 2.12. [23] Let $g: Q \times Q \to \mathbb{R}$ be a bifunction satisfying Assumption II and $T_{\alpha}^{(g,h)}, T_{\beta}^{(g,h)}$ be defined as in Lemma 2.10 with $\alpha, \beta > 0$. Then the following holds:

$$\|T_{\alpha}^{(g,h)}x - T_{\beta}^{(g,h)}x\|^{2} \le \frac{\alpha - \beta}{\alpha} \langle T_{\alpha}x - T_{\beta}x, T_{\alpha}x - x \rangle$$

for all $x \in \mathcal{H}$.

Lemma 2.13. [24] Let C be a convex subset of a real Hilbert space \mathcal{H} and $g : C \to \mathbb{R}$ be subdifferentiable on C. Then x^* is a solution to the following convex problem:

$$\min\{g(x): x \in C\}$$

if and only if $0 \in \partial g(x^*) + N_C(x^*)$, where $\partial g(\cdot)$ denotes the subdifferential of g and $N_C(x^*)$ is the (outward) normal cone of C at $x^* \in C$.

3 Main Results

3.1 A Weak Convergence Algorithm

Algorithm I : Initialization.

- Pick $x^0 \in C$ and choose the parameters $\beta, \eta, \theta \in (0, 1), 0 < \rho' \leq \rho'', \{\rho_k\} \subset [\rho', \rho''], 0 < \gamma' \leq \gamma'' < 2, \{\gamma_k\} \subset [\gamma', \gamma''], 0 < \alpha, \{\alpha_k\} \subset [\alpha, \infty), \mu \in (0, \frac{1}{\|A\|}).$
- For each i = 1, 2, ..., p, $\{\eta_{k,i}\}$ is a real sequence of nonnegative numbers satisfying $\sum_{i=1}^{p} \eta_{k,i} = 1$ for all $k \ge 1$.
- For each $i \in \{1, 2, ..., p\}$ and $j \in \{1, 2, ..., p'\}$, $S_i : C \to C$ and $T_j : Q \to Q$ are L_i and L'_j -strict pseudo-contractions for some $0 \le L_i < 1$ and $0 \le L'_j < 1$, respectively.
- $\{\beta_k\}$ is a nonnegative real sequence satisfying $0 < \bar{L} < \beta_k < 1$ and $\beta_k \rightarrow \frac{1}{2}$ as $k \rightarrow \infty$, where $\bar{L} := \max\{L_i : i = 1, 2, ..., p\}$ and $L' := \max\{L'_j : j = 1, 2, ..., p'\}$.

For each k, (k = 0, 1, 2, ...), the sequence $\{x_k\}$ is generated by the following steps: Step I : Solve the strongly convex program :

$$y^k := \operatorname{argmin} \left\{ f(x^k, y) + \frac{1}{2\rho_k} \|y - x^k\|^2 : y \in C \right\}$$

to obtain its unique solution y^k . If $y^k = x^k$, then set $u^k = x^k$ and go to Step III. Otherwise, go to Step II.

Step II : (Armijo linesearch rule) Find m_k as the smallest positive integer number m such that

$$\begin{cases} z^{k,m} = (1 - \eta^m) x^k + \eta^m y^m, \\ f(z^{k,m}, x^k) - f(z^{k,m}, y^k) \ge \frac{\theta}{2\rho_k} \|x^k - y^k\|^2. \end{cases}$$
(3.1)

Set $\eta_k = \eta^{m_k}, z^k = z^{k, m_k}$.

Step III : Take $t^k \in \partial f(z^k, x^k)$, $\sigma_k = \frac{f(z^k, x^k)}{\|t^k\|^2}$, and denote $\begin{cases}
u^k = P_C(x^k - \gamma_k \sigma_k t^k), \\
v^k = \beta_k u^k + (1 - \beta_k) \sum_{i=1}^p \eta_{k,i} S_i(u^k), \\
w^k = T_{\alpha_k}^{(g,h)} A v^k, \\
x^{k+1} := P_C(v^k + \mu A^*(\sum_{j=1}^{p'} \eta'_{k,j} T_j(w^k) - A v^k))) \\
\text{and go to iteration } k \text{ with } k \text{ replaced by } k+1.
\end{cases}$

Applying Lemma 4.1, Lemma 4.2 and Lemma 4.3 obtained in [25], we obtain the following Lemma immediately.

Lemma 3.1. Suppose that $p \in EP(C, f)$, $f(x, \cdot)$ is convex subdifferentiable on C for all $x \in C$ and that f is pseudomonotone on C. Then, we have:

(a) The Armijo linesearch rule (3.1) is well defined;

(b)
$$f(z^k, x^k) > 0;$$

(c)
$$0 \notin \partial_2 f(z^k, x^k);$$

(d) $||u^k - p||^2 \le ||x^k - p||^2 - \gamma_k (2 - \gamma_k) (\sigma_k ||t^k||)^2$.

Now, we are in a position to state and prove the main weak convergence theorem for the given iterative scheme.

Theorem 3.2. Let $\mathcal{H}_1, \mathcal{H}_2$ be two real Hilbert spaces and $C \subset \mathcal{H}_1, Q \subset \mathcal{H}_2$ be nonempty closed convex subsets. For each $i \in \{1, 2, \ldots, p\}$ and $j \in \{1, 2, \ldots, p'\}$, $S_i : C \to C$ and $T_j : Q \to Q$ are L_i and L'_j -strict pseudo-contractions for some $0 \leq L_i < 1$ and $0 \leq L'_j < 1$, respectively. Let the bifunctions f, g and h satisfy Assumptions I, II and III, respectively. Let $A : \mathcal{H}_1 \to \mathcal{H}_2$ be a bounded linear operator with its adjoint A^* . If

$$\Omega := \left\{ x^* \in S_{\rm EP}(C, f) \cap \left(\bigcap_{i=1}^p \operatorname{Fix}(S_i, C)\right) : Ax^* \in S_{\rm GEP}(Q, g, h) \cap \left(\bigcap_{j=1}^{p'} \operatorname{Fix}(T_j, Q)\right) \right\}$$

is nonempty set, then the sequences $\{x^k\}, \{u^k\}$ and $\{v^k\}$ generated by Algorithm I converge weakly to an element $\bar{x} \in \Omega$, and $\{w^k\}$ converges weakly to $A\bar{x}$.

Proof. Let $x^* \in \Omega$. Then $x^* \in S_{\text{EP}}(C, f) \cap \left(\bigcap_{i=1}^p \text{Fix}(S_i, C)\right)$ and $Ax^* \in S_{\text{GEP}}(Q, g, h) \cap \left(\bigcap_{j=1}^{p'} \text{Fix}(T_j, Q)\right)$. From Lemma 3.1(d), we have

$$\|u^{k} - x^{*}\|^{2} \leq \|x^{k} - x^{*}\|^{2} - \gamma_{k}(2 - \gamma_{k})(\sigma_{k}\|t^{k}\|)^{2} \leq \|x^{k} - x^{*}\|^{2}.$$
 (3.2)

For each $k \geq 1$, let the mapping \bar{S}_k be given by

$$\bar{S}_k := \sum_{i=1}^p \eta_{k,i} S_i.$$

By Proposition 2.5, we see that \bar{S}_k is a \bar{L} -strict pseudocontraction on C. Then, for all $k \geq 1$, we have

$$\begin{aligned} \|v^{k} - x^{*}\|^{2} &= \|\beta_{k}u^{k} + (1 - \beta_{k})\bar{S}_{k}(u^{k}) - x^{*}\|^{2} \\ &= \|\beta_{k}(u^{k} - x^{*}) + (1 - \beta_{k})(\bar{S}_{k}(u^{k}) - x^{*})\|^{2} \\ &= \beta_{k}\|u^{k} - x^{*}\|^{2} + (1 - \beta_{k})\|\bar{S}_{k}(u^{k}) - x^{*}\|^{2} \\ &- \beta_{k}(1 - \beta_{k})\|\bar{S}_{k}(u^{k}) - u^{k}\|^{2} \\ &= \beta_{k}\|u^{k} - x^{*}\|^{2} + (1 - \beta_{k})\|\bar{S}_{k}(u^{k}) - \bar{S}_{k}(x^{*})\|^{2} \\ &- \beta_{k}(1 - \beta_{k})\|\bar{S}_{k}(u^{k}) - u^{k}\|^{2} \\ &\leq \beta_{k}\|u^{k} - x^{*}\| \\ &+ (1 - \beta_{k})\Big(\|u^{k} - x^{*}\|^{2} + \bar{L}\|(I - \bar{S}_{k})(u^{k}) - (I - \bar{S}_{k})(x^{*})\|^{2}\Big) \end{aligned}$$

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$$\begin{aligned} &-\beta_k (1-\beta_k) \|\bar{S}_k(u^k) - u^k\|^2 \\ &= \|u^k - x^*\|^2 \\ &+ (1-\beta_k) \left(\bar{L} \|\bar{S}_k(u^k) - u^k\|^2\right) - \beta_k (1-\beta_k) \|\bar{S}_k(u^k) - u^k\|^2 \\ &= \|u^k - x^*\|^2 + (1-\beta_k) (\bar{L} - \beta_k) \|\bar{S}_k(u^k) - u^k\|^2. \end{aligned}$$
(3.3)

Since $0 < \overline{L} < \beta_k < 1$, it follows from (3.3) that

$$\|v^{k} - x^{*}\|^{2} \le \|u^{k} - x^{*}\|^{2} \le \|x^{k} - x^{*}\|^{2}.$$
(3.4)

By Lemma 2.10, we have

$$\begin{aligned} \left\| T_{\alpha_{k}}^{(g,h)} A v^{k} - A x^{*} \right\|^{2} &= \left\| T_{\alpha_{k}}^{(g,h)} A v^{k} - T_{\alpha_{k}}^{(g,h)} A x^{*} \right\|^{2} \\ &\leq \left\langle T_{\alpha_{k}}^{(g,h)} A v^{k} - T_{\alpha_{k}}^{(g,h)} A x^{*}, A v^{k} - A x^{*} \right\rangle \\ &= \left\langle T_{\alpha_{k}}^{(g,h)} A v^{k} - A x^{*}, A v^{k} - A x^{*} \right\rangle \\ &= \frac{1}{2} \Big[\left\| T_{\alpha_{k}}^{(g,h)} A v^{k} - A x^{*} \right\|^{2} + \left\| A v^{k} - A x^{*} \right\|^{2} \Big] \\ &- \frac{1}{2} \left\| T_{\alpha_{k}}^{(g,h)} A v^{k} - A v^{k} \right\|^{2}. \end{aligned}$$

Hence,

$$\left\|T_{\alpha_{k}}^{(g,h)}Av^{k} - Ax^{*}\right\|^{2} \leq \left\|Av^{k} - Ax^{*}\right\|^{2} - \left\|T_{\alpha_{k}}^{(g,h)}Av^{k} - Av^{k}\right\|^{2}.$$

For each $k \geq 1$, let \overline{T}_k be a mapping defined by

$$\bar{T}_k = \sum_{j=1}^{p'} \eta'_{k,j} T_j.$$

By Proposition 2.5, we see that \overline{T}_k is a \overline{L} -strict pseudo-contraction on Q and the sequence $\{x^k\}$ generated by Algorithm 1 can be rewritten as

$$x^{k+1} = P_C(v^k + \mu A^*(\bar{T}_k w^k - Av^k)), \forall k \ge 1.$$

Then, for all $k \ge 1$, we have

$$\begin{aligned} \|\bar{T}_{k}w^{k} - Ax^{*}\|^{2} &= \|\bar{T}_{k}w^{k} - \bar{T}_{k}Ax^{*}\|^{2} \\ &\leq \|w^{k} - Ax^{*}\|^{2} + \bar{L}\|(I - \bar{T}_{k})(w^{k}) - (I - \bar{T}_{k})(Ax^{*})\|^{2} \\ &= \|w^{k} - Ax^{*}\|^{2} + \bar{L}\|\bar{T}_{k}(w^{k}) - w^{k}\|^{2} \\ &< \|w^{k} - Ax^{*}\|^{2} + \|\bar{T}_{k}(w^{k}) - w^{k}\|^{2} \\ &= \|T^{(g,h)}_{\alpha_{k}}Av^{k} - Ax^{*}\|^{2} + \|\bar{T}_{k}(w^{k}) - w^{k}\|^{2} \\ &\leq \|Av^{k} - Ax^{*}\|^{2} - \|T^{(g,h)}_{\alpha_{k}}Av^{k} - Ax^{*}\|^{2} \\ &+ \|\bar{T}_{k}(w^{k}) - w^{k}\|^{2}. \end{aligned}$$
(3.5)

$$\begin{aligned} \text{Using (3.5), we have} \\ \langle A(v^{k} - x^{*}), \bar{T}_{k}w^{k} - Av^{k} \rangle \\ &= \langle A(v^{k} - x^{*}) + \bar{T}_{k}w^{k} - Av^{k} - (\bar{T}_{k}w^{k} - Av^{k}), \bar{T}_{k}w^{k} - Av^{k} \rangle \\ &= \langle \bar{T}_{k}w^{k} - Ax^{*}, \bar{T}_{k}w^{k} - Av^{k} \rangle - \|\bar{T}_{k}w^{k} - Av^{k}\|^{2} \\ &= \frac{1}{2} \Big[\|\bar{T}_{k}w^{k} - Ax^{*}\|^{2} + \|\bar{T}_{k}w^{k} - Av^{k}\|^{2} - \|Av^{k} - Ax^{*}\|^{2} \Big] \\ &- \|\bar{T}_{k}w^{k} - Av^{k}\|^{2} \\ &= \frac{1}{2} \Big[\Big(\|\bar{T}_{k}w^{k} - Ax^{*}\|^{2} - \|Av^{k} - Ax^{*}\|^{2} \Big) - \|\bar{T}_{k}w^{k} - Av^{k}\|^{2} \Big] \\ &= \frac{1}{2} \Big[\Big(\|\bar{T}_{k}w^{k} - Ax^{*}\|^{2} - \|Av^{k} - Ax^{*}\|^{2} \Big) \\ &- \frac{1}{2} \|\bar{T}_{k}w^{k} - Av^{k}\|^{2} \\ &\leq \frac{1}{2} \Big(\|\bar{T}_{k}(w^{k}) - w^{k}\|^{2} - \|T^{(g,h)}_{\alpha_{k}}Av^{k} - Ax^{*}\|^{2} \Big) \\ &- \frac{1}{2} \|\bar{T}_{k}w^{k} - Av^{k}\|^{2} \\ &= \frac{1}{2} \|\bar{T}_{k}(w^{k}) - w^{k}\|^{2} - \frac{1}{2} \|T^{(g,h)}_{\alpha_{k}}Av^{k} - Av^{k}\|^{2} \\ &= \frac{1}{2} \|\bar{T}_{k}(w^{k}) - w^{k}\|^{2} - \frac{1}{2} \|T^{(g,h)}_{\alpha_{k}}Av^{k} - Av^{k}\|^{2} \end{aligned}$$

$$(3.6)$$

By the definition of x^{k+1} we have

$$\begin{split} \|x^{k+1} - x^*\|^2 &= \|P_C(v^k + \mu A^*(\bar{T}_k w^k - Av^k)) - P_C(x^*)\|^2 \\ &\leq \|(v^k - x^*) + \mu A^*(\bar{T}_k w^k - Av^k)\|^2 \\ &= \|v^k - x^*\|^2 + \|\mu A^*(\bar{T}_k w^k - Av^k)\|^2 \\ &+ 2\mu \langle v^k - x^*, A^*(\bar{T}_k w^k - Av^k) \rangle \\ &\leq \|v^k - x^*\|^2 + \mu^2 \|A^*\|^2 \|\bar{T}_k w^k - Av^k\|^2 \\ &+ 2\mu \langle A(v^k - x^*), \bar{T}_k w^k - Av^k \rangle. \end{split}$$

In combination with (3.6) and (3.4), the last inequality becomes

$$\begin{aligned} \|x^{k+1} - x^*\|^2 &\leq \|v^k - x^*\|^2 - \mu^2 \|A^*\|^2 \|\bar{T}_k w^k - Av^k\|^2 \\ &+ \mu \|\bar{T}_k (w^k) - w^k\|^2 - \mu \|\bar{T}_k w^k - Av^k\|^2 \\ &- \mu \|T_{\alpha_k}^{(g,h)} Av^k - Av^k\|^2 \end{aligned}$$

$$= \|v^k - x^*\|^2 - \mu (1 - \mu \|A^*\|^2) \|\bar{T}_k w^k - Av^k\|^2 \\ &+ \mu \|\bar{T}_k (w^k) - w^k\|^2 - \mu \|w^k - Av^k\|^2 \\ &\leq \|x^k - x^*\|^2 - \mu (1 - \mu \|A^*\|^2) \|\bar{T}_k w^k - Av^k\|^2 \\ &+ \mu \|\bar{T}_k (w^k) - w^k\|^2 - \mu \|w^k - Av^k\|^2. \end{aligned}$$
(3.7)

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From (3.4), (3.7), and $\mu \in \left(0, \frac{1}{\|A\|^2}\right)$, we get $\|x^{k+1} - x^*\| \le \|v^k - x^*\| \le \|u^k - x^*\| \le \|x^k - x^*\|$ (3.8)

and

$$\mu (1 - \mu \|A^*\|^2) \|\bar{T}_k w^k - A v^k\|^2 + \mu \|w^k - A v^k\|^2 - \mu \|\bar{T}_k (w^k) - w^k\|^2$$

$$\leq \|x^k - x^*\|^2 - \|x^{k+1} - x^*\|^2.$$
(3.9)

Therefore, $\lim_{k \to +\infty} ||x^k - x^*||$ exists, and we get from (3.8) and (3.9) that

$$\lim_{k \to +\infty} \|x^k - x^*\| = \lim_{k \to +\infty} \|v^k - x^*\| = \lim_{k \to +\infty} \|u^k - x^*\| \text{ and}$$
$$\lim_{k \to +\infty} \|\bar{T}_k w^k - A v^k\| = \lim_{k \to +\infty} \|w^k - A v^k\| = 0.$$
(3.10)

From (3.10) and the inequality

$$\|\bar{T}_k w^k - w^k\| \le \|\bar{T}_k w^k - A v^k\| + \|w^k - A v^k\|,$$

we get

$$\lim_{k \to +\infty} \|\bar{T}_k w^k - w^k\| = 0.$$
(3.11)

Besides, Lemma 3.1(d) implies

$$||u^{k} - x^{*}||^{2} \le ||x^{k} - x^{*}||^{2} - \gamma_{k}(2 - \gamma_{k})(\sigma_{k}||t^{k}||)^{2}.$$

Hence,

$$\gamma_k (2 - \gamma_k) (\sigma_k ||t^k||)^2 \le ||x^k - x^*||^2 - ||u^k - x^*||^2$$

= (||x^k - x^*|| - ||u^k - x^*||)(||x^k - x^*|| + ||u^k - x^*||).

In view of (3.10), we get

$$\lim_{k \to +\infty} \sigma_k \|t^k\| = 0. \tag{3.12}$$

Moreover, by the definition of $u^k, u^k = P_C(x^k - \gamma_k \sigma_k t^k)$. We have

$$\|u^k - x^k\| \le \gamma_k \sigma_k \|t^k\|.$$

So, we get from (3.12) that

$$\lim_{k \to +\infty} \|u^k - x^k\| = 0.$$
(3.13)

From (3.3), we get

$$\|v^{k} - x^{*}\|^{2} \leq \|u^{k} - x^{*}\|^{2} + (1 - \beta_{k}) \Big(\bar{L} - \beta_{k}) \Big\|\bar{S}_{k}(u^{k}) - u^{k}\Big\|^{2}.$$
 (3.14)

Therefore,

$$(1 - \beta_k)(\beta_k - \bar{L}) \|\bar{S}_k u^k - u^k\|^2 \le \|u^k - x^*\|^2 - \|v^k - x^*\|^2.$$

Combining the last inequality with (3.10), we obtain that

$$\lim_{k \to +\infty} \|\bar{S}_k u^k - u^k\| = 0.$$
(3.15)

Moreover,

$$\begin{aligned} \|v^{k} - x^{k}\| &\leq \|v^{k} - u^{k}\| + \|u^{k} - x^{k}\| \\ &= \|\beta u^{k} + (1 - \beta)\bar{S}_{k}u^{k} - u^{k}\| + \|u^{k} - x^{k}\| \\ &= (1 - \beta)\|\bar{S}_{k}u^{k} - u^{k}\| + \|u^{k} - x^{k}\| \end{aligned}$$

Thus, we get from (3.13) and (3.15) that

$$\lim_{k \to +\infty} \|v^k - x^k\| = 0.$$
(3.16)

Since $\lim_{k\to+\infty} ||x^k - x^*||$ exists, $\{x^k\}$ is bounded. By Lemma 2.8, $\{y^k\}$ is bounded, and consequently $\{z^k\}$ is bounded. By Lemma 2.7 $\{t^k\}$ is bounded. Step III and (3.12) yield

$$\lim_{k \to \infty} f(z^k, x^k) = \lim_{k \to \infty} \left[\sigma_k \| t^k \| \right] \| t^k \| = 0.$$
(3.17)

We have

$$0 = f(z^k, z^k) = f(z^k, (1 - \eta_k)x^k + \eta_k y^k) \le (1 - \eta_k)f(z^k, x^k + \eta_k f(z^k, y^k)),$$

so,we obtain

$$\eta_k \left[f(z^k, x^k) - f(z^k, y^k) \right] \le f(z^k, x^k).$$

Thus, we get from (3.34) that

$$\frac{\theta}{2\rho_k}\eta_k \|x^k - y^k\|^2 \le \eta_k \big[f(z^k, x^k) - f(z^k, y^k) \big] \le f(z^k, x^k).$$

Combining this with (3.17), we have

$$\lim_{k \to \infty} \eta_k \|x^k - y^k\|^2 = 0.$$
(3.18)

Suppose that \bar{x} is a weak accumulation point of $\{x^k\}$, that is, there exists a subsequence $\{x^{k_j}\}$ of $\{x^k\}$ such that $\{x^{k_j}\}$ converges weakly to $\bar{x} \in C$ as $j \to +\infty$. Then, it follows from (3.13) and (3.16) that $u^{k_j} \to \bar{x}$, $v^{k_j} \to \bar{x}$, and $Av^{k_j} \to A\bar{x}$. Since $\lim_{k\to+\infty} ||w^k - Av^k|| = 0$, we deduce that $w^{k_j} \to A\bar{x}$. Because $\{w^k\} \subset Q$ and Q is closed and convex, we have that $A\bar{x} \in Q$. From (3.18), we get

$$\lim_{i \to \infty} \eta_{k_i} \| x^{k_i} - y^{k_i} \|^2 = 0.$$
(3.19)

We now consider two distinct cases.

Case I. $\limsup_{i\to\infty} \eta_{k_i} > 0$. In this case, there exist $\overline{\eta} > 0$ and a subsequence of $\{\eta_{k_i}\}$, denoted again by $\{\eta_{k_i}\}$, such that, for some $i_0 > 0$, $\eta_{k_i} > \overline{\eta}$ for all $i > i_0$. Using this fact and (3.19), we have

$$\lim_{i \to \infty} \|x^{k_i} - y^{k_i}\| = 0.$$
(3.20)

Recall that $x^k \to \bar{x}$, together with (3.20), implies that $y^{k_i} \to \bar{x}$ as $i \to \infty$. By the definition of y^{k_i} ,

$$y^{k_i} := \operatorname{argmin}\{f(x^{k_i}, y) + \frac{1}{2\rho_{k_i}} \|y - x^{k_i}\|^2 : y \in C\},\$$

so, we have

$$0 \in \partial f(x^{k_i}, y^{k_i}) + \frac{1}{\rho_{k_i}} (y^{k_i} - x^{k_i}) + N_C (y^{k_i}).$$

Thus, there exists $\hat{t}^{k_i}\in\partial f(x^{k_i},y^{k_i})$ such that

$$\langle \hat{t}^{k_i}, y - y^{k_i} \rangle + \frac{1}{\rho_{k_i}} \langle y^{k_i} - x^{k_i}, y - y^{k_i} \rangle \ge 0, \forall y \in C.$$

Combining this with

$$f(x^{k_i}, y) - f(x^{k_i}, y^{k_i}) \ge \langle \hat{t}^{k_i}, y - y^{k_i} \rangle, \forall y \in C,$$

yields

$$f(x^{k_i}, y) - f(x^{k_i}, y^{k_i}) + \frac{1}{\rho_{k_i}} \langle y^{k_i} - x^{k_i}, y - y^{k_i} \rangle \ge 0, \forall y \in C.$$
(3.21)

Since

$$\langle y^{k_i} - x^{k_i}, y - y^{k_i} \rangle \le \|y^{k_i} - x^{k_i}\| \|y - y^{k_i}\|,$$

from (3.21) we get that

$$f(x^{k_i}, y) - f(x^{k_i}, y^{k_i}) + \frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\| \|y - y^{k_i}\| \ge 0.$$
(3.22)

Letting $i \to \infty$, by the weak continuity of f and (3.20), from (3.22) we obtain in the limit that

$$f(\bar{x}, y) - f(\bar{x}, \bar{x}) \ge 0.$$

Thus,

$$f(\bar{x}, y) \ge 0, \forall y \in C$$

Hence, \bar{x} is a solution of EP(C, f).

Case II. $\lim_{i\to\infty} \eta_{k_i} = 0$. From the boundedness of $\{y^{k_i}\}$, without loss of generality, we may assume that $y^{k_i} \to \overline{y}$ as $i \to \infty$. Replacing y by x^{k_i} in (3.22), we get

$$f(x^{k_i}, y^{k_i}) \le -\frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2.$$
(3.23)

On the other hand, by the Armijo linesearch rule (3.34), for $m_{k_i} - 1$, we have

$$f(z^{k_i,m_{k_i}-1},x^{k_i}) - f(z^{k_i,m_{k_i}-1},y^{k_i}) < \frac{\theta}{2\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2.$$

Combining this with (3.23), we get

$$f(x^{k_i}, y^{k_i}) \le -\frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2 < \frac{2}{\theta} \left[f(z^{k_i, m_{k_i} - 1}, y^{k_i}) - f(z^{k_i, m_{k_i} - 1}, x^{k_i}) \right].$$
(3.24)

According to the algorithm, we have $z^{k_i,m_{k_i}-1} = (1 - \eta^{m_{k_i}-1})x^{k_i} + \eta^{m_{k_i}-1}y^{k_i}$. Since $\eta^{m_{k_i}-1} \to 0, \{x^{k_i}\}$ converges weakly to \bar{x} , and $\{y^{k_i}\}$ converges weakly to \bar{y} , this implies that $z^{k_i,m_{k_i}-1} \rightharpoonup \bar{x}$ as $i \to \infty$. Beside that, $\{\frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2\}$ is bounded, so without loss of generality we may assume that $\lim_{i\to+\infty} \frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2$ exists. Hence, in the limit, from (3.24) we get that

$$f(\bar{x}, \bar{y}) \le -\lim_{i \to +\infty} \frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2 \le \frac{2}{\theta} f(\bar{x}, \bar{y})$$

Therefore, $f(\bar{x}, \bar{y}) = 0$ and $\lim_{i \to +\infty} \frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2 = 0$. By Case I we get $\bar{x} \in S_{\text{EP}}(f)$. Next, we prove that any weakly cluster point of the sequence $\{x^k\}$ is a common fixed point of L_i -strict pseudocontraction, for each $i = 1, 2, \ldots, p$. Inparticular, $\bar{x} \in \bigcap_{i=1}^{p} \operatorname{Fix}(S_i, C)$. Let \bar{y} be any weakly cluster point of $\{x^k\}$ and let $\{x^{k_m}\}$ be a subsequence of $\{x^k\} \subset C$ weakly converging to \bar{y} . By convexity and the closedness of C, C is weakly closed. Thus, $\bar{y} \in C$. We first show that

$$\lim_{m \to \infty} \|x^{k_m} - S(x^{k_m})\| = 0.$$
(3.25)

Since,

$$\|\bar{S}_k(u^k) - x^k\| \le \|\bar{S}_k(u^k) - u^k\| + \|u^k - x^k\|.$$

Then, by (3.15) and (3.13) we obtain

$$\lim_{k \to +\infty} \|\bar{S}_k u^k - x^k\| = 0.$$
 (3.26)

Since,

$$\|\bar{S}_k(x^k) - x^k\| \le \|\bar{S}_k(x^k) - \bar{S}_k(u^k)\| + \|\bar{S}_k(u^k) - x^k\|.$$

Then, by Proprositon 2.5(i), we obtain

$$\|\bar{S}_k(x^k) - x^k\| \le \frac{1+\bar{L}}{1-\bar{L}} \|x^k - u^k\| + \|\bar{S}_k(u^k) - x^k\|.$$

So, from (3.13) and (3.26), we obtain

$$\lim_{k \to \infty} \|\bar{S}_k(x^k) - x^k\| = 0.$$
(3.27)

For each i = 1, 2, ..., p, we suppose that $\{\eta_{k_m, i}\}$ converges to η_i as $m \to \infty$ such that $\sum_{i=1}^p \eta_i = 1$. Then, for each 1, 2, ..., p and $x \in C$, we have

$$\bar{S}_{k_m}(x) := \sum_{m=1}^{p} \eta_{k_m, i} S_i(x) \to \sum_{i=1}^{p} \eta_i S_i(x) := S(x) \quad \text{as} \quad m \to \infty.$$
(3.28)

It follows from (3.27) that

$$\begin{aligned} \|x^{k_m} - S(x^{k_m})\| &\leq \|x^{k_m} - \bar{S}_{k_m}(x^{k_m})\| + \|\bar{S}_{k_m}(x^{k_m}) - S(x^{k_m})\| \\ &= \|x^{k_m} - \bar{S}_{k_m}(x^{k_m})\| + \left\|\sum_{i=1}^p \eta_{k_m,i} S_i(x^{k_m}) - \sum_{i=1}^p \eta_i S_i(x^{k_m})\right\| \\ &= \|x^{k_m} - \bar{S}_{k_m}(x^{k_m})\| + \left\|\sum_{i=1}^p (\eta_{k_m,i} - \eta_i) S_i(x^{k_m})\right\| \\ &\leq \|x^{k_m} - \bar{S}_{k_m}(x^{k_m})\| + \sum_{i=1}^p |\eta_{k_m,i} - \eta_i| \|S_i(x^{k_m})\|. \end{aligned}$$
(3.29)

So, we get

$$\lim_{n \to \infty} \|x^{k_m} - S(x^{k_m})\| = 0.$$
(3.30)

By Proprosition 2.5(ii), we have

$$\bar{y} \in \operatorname{Fix}(S) = \operatorname{Fix}\left(\sum_{i=1}^{p} n_i S_i\right) x.$$

It follows from 2.5(v), we have

$$\bar{y} \in \bigcap_{i=1}^{p} \operatorname{Fix}(S_i, C).$$

In particular, we conclude that $\bar{x} \in \bigcap_{i=1}^{p} \operatorname{Fix}(S_i, C)$. Hence,

$$\bar{x} \in S_{\text{EP}}(C, f) \cap \Big(\bigcap_{i=1}^{p} \operatorname{Fix}(S_i, C)\Big).$$
(3.31)

Next, we need to show that $A\bar{x} \in S_{\text{GEP}}(Q, g, h) \cap \left(\bigcap_{j=1}^{p'} \text{Fix}(T_j, Q)\right)$. Indeed, we have $S_{\text{GEP}}(Q, g, h) = \text{Fix}(T_{\beta}^{(g,h)})$. So, if $T_{\beta}^{(g,h)}A\bar{x} \neq A\bar{x}$, then, using Opial's condition, we have

$$\begin{split} \liminf_{j \to +\infty} \|Av^{k_j} - A\bar{x}\| &< \liminf_{j \to +\infty} \|Av^{k_j} - T_{\beta}^{(g,h)}A\bar{x}\| \\ &= \liminf_{j \to +\infty} \|Av^{k_j} - w^{k_j} + w^{k_j} - T_{\beta}^{(g,h)}A\bar{x}\| \\ &\leq \liminf_{j \to +\infty} (\|Av^{k_j} - w^{k_j}\| + \|w^{k_j} - T_{\beta}^{(g,h)}A\bar{x}\|). \end{split}$$

So it follows from (3.12) and Lamma 3.1 that

$$\begin{split} \liminf_{j \to +\infty} \|Av^{k_{j}} - A\bar{x}\| &< \liminf_{j \to +\infty} \|T_{\beta}^{(g,h)}A\bar{x} - w^{k_{j}}\| \\ &= \liminf_{j \to +\infty} \|T_{\beta}^{(g,h)}A\bar{x} - T_{\alpha_{k_{j}}}^{(g,h)}Av^{k_{j}}\| \\ &\leq \liminf_{j \to +\infty} \left\{ \|Av^{k_{j}} - A\bar{x}\| + \frac{|\alpha_{k_{j}} - \beta|}{\alpha_{k_{j}}} \|T_{\alpha_{k_{j}}}^{(g,h)}Av^{k_{j}} - Av^{k_{j}}\| \right\} \\ &= \liminf_{j \to +\infty} \left\{ \|Av^{k_{j}} - A\bar{x}\| + \frac{|\alpha_{k_{j}} - \beta|}{\alpha_{k_{j}}} \|w^{k_{j}} - Av^{k_{j}}\| \right\} \\ &= \liminf_{j \to +\infty} \|Av^{k_{j}} - A\bar{x}\|, \end{split}$$

a contradiction. Thus, $A\bar{x} \in \operatorname{Fix}(T_{\alpha}^{(g,h)}) = S_{\operatorname{GEP}}(Q, g, h)$. Moreover, (3.11) imply that $(I - \bar{T}_k)(w^{k_j}) \to 0$ and we have $w^{k_j} \rightharpoonup Ap$, then by proposition 2.5(ii) we get $(I - \bar{T}_k)(A\bar{x}) \to 0$ is demiclosed at 0, so we obtain that $A\bar{x} \in \bigcap_{j=1}^{p'} \operatorname{Fix}(T_j, Q)$. Therefore,

$$A\bar{x} \in S_{\text{GEP}}(Q, g, h) \cap (\bigcap_{j=1}^{p'} \text{Fix}(T_j, Q)).$$
(3.32)

From (3.31) and (3.32) we obtain that $\bar{x} \in \Omega$. To complete the proof, we must show that the whole sequence $\{x^k\}$ converges weakly to \bar{x} . Indeed, if there exists a subsequence $\{x^{l_i}\}$ of $\{x^k\}$ such that $x^{l_i} \rightharpoonup q$ with $q \neq \bar{x}$, then we have $q \in \Omega$. By Opial's condition this yields

$$\lim_{i \to +\infty} \inf \|x^{l_i} - q\| < \liminf_{i \to +\infty} \|x^{l_i} - \bar{x}\|$$

$$= \liminf_{j \to +\infty} \|x^k - \bar{x}\|$$

$$= \liminf_{j \to +\infty} \|x^{k_j} - \bar{x}\|$$

$$< \liminf_{j \to +\infty} \|x^{k_j} - q\|$$

$$= \liminf_{i \to +\infty} \|x^{l_i} - q\|.$$
(3.33)

This is a contradiction. Hence, $\{x^k\}$ converges weakly to \bar{x} . Combining this with (3.13), it is immediate that $\{u^k\}, \{v^k\}$ also converge weakly to \bar{x} and $w^{k_j} \rightarrow A\bar{x} \in S_{\text{GEP}}(Q, g, h) \cap \left(\bigcap_{j=1}^{p'} \text{Fix}(T_j, Q)\right)$.

Corollary 3.3. Let $\mathcal{H}_1, \mathcal{H}_2$ be two real Hilbert spaces and $C \subset \mathcal{H}_1, Q \subset \mathcal{H}_2$ be nonempty closed convex subsets. Let $S : C \to C$ and $T : Q \to Q$ are L and L'-strict pseudo-contractions, respectively, . Let the bifunctions f, g and h satisfy Assumptions I, II and III, respectively. Let $A : \mathcal{H}_1 \to \mathcal{H}_2$ be a bounded linear operator with its adjoint A^* . If

$$\Omega_1 := \{ x^* \in S_{\text{EP}}(C, f) \cap Fix(S, C) : Ax^* \in S_{\text{GEP}}(Q, g, h) \cap Fix(T, Q) \} \neq \emptyset.$$

Let the sequences $\{x^k\}, \{u^k\}$ and $\{v^k\}$ be generated by the following :

 $\mathbf{Step \ I:} \ Solve \ the \ strongly \ convex \ program:$

$$y^{k} := \operatorname{argmin}\left\{f(x^{k}, y) + \frac{1}{2\rho_{k}}\|y - x^{k}\|^{2} : y \in C\right\}$$

to obtain its unique solution y^k . If $y^k = x^k$, then set $u^k = x^k$ and go to Step III. Otherwise, go to Step II.

Step II : (Armijo linesearch rule) Find m_k as the smallest positive integer number m such that

$$\begin{cases} z^{k,m} = (1 - \eta^m) x^k + \eta^m y^m, \\ f(z^{k,m}, x^k) - f(z^{k,m}, y^k) \ge \frac{\theta}{2\rho_k} \|x^k - y^k\|^2. \end{cases}$$
(3.34)

Set $\eta_k = \eta^{m_k}, z^k = z^{k,m_k}$. **Step III :** Take $t^k \in \partial f(z^k, x^k), \sigma_k = \frac{f(z^k, x^k)}{\|t^k\|^2}$, and denote $\begin{cases}
u^k = P_C(x^k - \gamma_k \sigma_k t^k), \\
v^k = \beta_k u^k + (1 - \beta_k) S(u^k), \\
w^k = T^{(g,h)}_{\alpha_k} Av^k, \\
x^{k+1} := P_C(v^k + \mu A^*(T(w^k) - Av^k))) \\
and go to iteration k with k replaced by k + 1. \\
Then, <math>\{x^k\}, \{u^k\} \text{ and } \{v^k\} \text{ converge weakly to an element } \bar{x} \in \Omega_1, \text{ and } \{w^k\}\end{cases}$

converges weakly to $A\bar{x}$.

3.2 A Strong Convergence Algorithm

Algorithm II: Initialization.

- Pick $x^g \in C_0 = C$ and choose the parameters $\beta, \eta, \theta \in (0, 1), \ 0 < \rho' \leq \rho'', \{\rho_k\} \subset [\rho', \rho''], \ 0 < \gamma' \leq \gamma'' < 2, \ \{\gamma_k\} \subset [\gamma', \gamma''], \ 0 < \alpha, \ \{\alpha_k\} \subset [\alpha, \infty), \ \mu \in (0, \frac{1}{\|A\|}).$
- For each i = 1, 2, ..., p, $\{\eta_{k,i}\}$ is a real sequence of nonnegative numbers satisfying $\sum_{i=1}^{p} \eta_{k,i} = 1$ for all $k \ge 1$.
- For each $i \in \{1, 2, ..., p\}$ and $j \in \{1, 2, ..., p'\}$, $S_i : C \to C$ and $T_j : Q \to Q$ are L_i and L'_j -strict pseudo-contractions for some $0 \le L_i < 1$ and $0 \le L'_j < 1$, respectively.
- $\{\beta_k\}$ is a nonnegative real sequence satisfying $0 < \bar{L} < \beta_k < 1$ and $\beta_k \to \frac{1}{2}$ as $k \to \infty$, where $\bar{L} := \max\{L_i : i = 1, 2, ..., p\}$.

Iteration : k, (k = 0, 1, 2, ...). Having x^k , do the following steps: **Step I :** Solve the strongly convex program :

$$y^k := \operatorname{argmin} \left\{ f(x^k, y) + \frac{1}{2\rho_k} \|y - x^k\|^2 : y \in C \right\}$$

to obtain its unique solution y^k .

If $y^k = x^k$, then set $u^k = x^k$ and go to Step III. Otherwise, go to Step II.

Step II: (Armijo linesearch rule) Find m_k as the smallest positive integer number m such that

$$\begin{cases} z^{k,m} = (1 - \eta^m) x^k + \eta^m y^m, \\ f(z^{k,m}, x^k) - f(z^{k,m}, y^k) \ge \frac{\theta}{2\rho_k} \|x^k - y^k\|^2. \end{cases}$$
(3.35)

Set $\eta_k = \eta^{m_k}, z^k = z^{k,m_k}$. **Step III :** Take $t^k \in \partial f(z^k, x^k)$ and compute $\sigma_k = \frac{f(z^k, x^k)}{\|t^k\|^2}$, and denote $\int u^k = P_C(x^k - \gamma_k \sigma_k t^k),$ $\begin{cases} u^{a} = T_{C}(u^{a} - \gamma_{k}o_{k}u^{a}), \\ v^{k} = \beta_{k}u^{k} + (1 - \beta_{k})\sum_{i=1}^{p}\eta_{k,i}S_{i}(u^{k}), \\ w^{k} = T_{\alpha_{k}}^{(g,h)}Av^{k}, \\ d^{k} := P_{C}(v^{k} + \mu A^{*}(\sum_{i=1}^{p}\eta_{k,i}'T_{i}(w^{k}) - Av^{k})), \\ C_{k+1} = \{x \in C_{k} : \|x - d^{k}\| \leq \|x - v^{k}\| \leq \|x - x^{k}\|\}, \\ x^{k+1} := P_{C_{k+1}}(x^{g}) \end{cases}$ and go to iteration k with k replaced by k + 1.

Now, we are in a position to state and prove the main strong convergence theorem for the given iterative scheme. Throughout this section, we suppose the following :

Theorem 3.4. Let $\mathcal{H}_1, \mathcal{H}_2$ be two real Hilbert spaces and $C \subset \mathcal{H}_1, Q \subset \mathcal{H}_2$ be nonempty closed convex subsets. For each $i \in \{1, 2, \dots, p\}$ and $j \in \{1, 2, \dots, p'\}$, S_i and T_j are L_i and L'_j -strict pseudo-contractions for some $0 \leq L_i < 1$ and $0 \leq L'_i < 1$, respectively. Let the bifunctions f, g and h satisfy Assumptions I, II and III, respectively. Let $A : \mathcal{H}_1 \to \mathcal{H}_2$ be a bounded linear operator with its adjoint A^* . If

$$\Omega := \left\{ x^* \in S_{\mathrm{EP}}(C, f) \cap \left(\bigcap_{i=1}^p \mathrm{Fix}(S_i, C)\right) : Ax^* \in S_{\mathrm{GEP}}(Q, g, h) \cap \left(\bigcap_{j=1}^{p'} \mathrm{Fix}(T_j, Q)\right) \right\}$$

is nonempty set, then the sequences $\{x^k\}, \{u^k\}$ and $\{v^k\}$ generated by Algorithm II converge strongly to an element $\bar{x} \in \Omega$, and $\{w^k\}$ converges strongly to $A\bar{x} \in$ $S_{\text{GEP}}(Q, g, h) \cap (\bigcap_{j=1}^{p'} \text{Fix}(T_j, Q)).$

Proof. First, we observe that the linesearch rule (3.35) is well defined. Let $x^* \in \Omega$. From (3.7), (3.14), and (3.2) we have

$$\begin{aligned} \|d^{k} - x^{*}\|^{2} &\leq \|v^{k} - x^{*}\| - \mu(1 - \mu\|A\|^{2})\|\bar{T}_{k}w^{k} - Av^{k}\|^{2} - \mu\|w^{k} - Av^{k}\|^{2} \\ &\leq \|u^{k} - x^{*}\|^{2} + (1 - \beta_{k})(\bar{L} - \beta_{k})\|\bar{S}_{k}(u^{k}) - u^{k}\|^{2} \\ &- \mu(1 - \mu\|A\|^{2})\|\bar{T}_{k}w^{k} - Av^{k}\|^{2} - \mu\|w^{k} - Av^{k}\|^{2} \\ &\leq \|x^{k} - x^{*}\| + (1 - \beta_{k})(\bar{L} - \beta_{k})\|\bar{S}_{k}(u^{k}) - u^{k}\|^{2} \\ &- \mu(1 - \mu\|A\|^{2})\|\bar{T}_{k}w^{k} - Av^{k}\|^{2} - \mu\|w^{k} - Av^{k}\|^{2}. \end{aligned}$$
(3.36)

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Since $0 < \bar{L} < \beta_k < 1$ and $\mu \in (0, \frac{1}{\|A\|^2})$, (3.36) implies that

$$||d^{k} - x^{*}|| \le ||v^{k} - x^{*}|| \le ||u^{k} - x^{*}|| \le ||x^{k} - x^{*}||, \forall k.$$
(3.37)

Since $x^* \in C_0$, from (3.37) we get by induction that $x^* \in C_k$ for all $k \in \mathbb{N}$ and, consequently, $\Omega \subset C_k$ for all k. By setting

$$D_k = \{x \in \mathcal{H}_1 : \|x - d^k\| \le \|x - v^k\| \le \|x - x^k\|\}, k \in \mathbb{N},\$$

it is clear that D_k is closed and convex for all k. In addition, $C_0 = C$ is also closed and convex, and $C_{k+1} = C_k \cap D_k$. Hence, C_k is closed and convex for all k. From the definition of x^{k+1} we have $x^{k+1} \in C_{k+1} \subset C_k$ and $x^k = P_{C_k}(x^g)$, so

$$||x^{k} - x^{g}|| \le ||x^{k+1} - x^{g}||, \forall k$$

Since $x^* \in C^{k+1}$, this implies that

$$||x^{k+1} - x^g|| \le ||x^* - x^g||.$$

Thus,

$$||x^k - x^g|| \le ||x^{k+1} - x^g|| \le ||x^* - x^g||, \forall k$$

Consequently, $\{\|x^k - x^g\|\}$ is nondecreasing and bounded, so $\lim_{k\to+\infty} \|x^k - x^g\|$ does exist. Combining this with (3.37), we obtain that $\{d^k\}$ and $\{v^k\}$ are also bounded. For all m > n, we have that $x^m \in C_m \subset C_n$ and $x^n = P_{C_n}(x^g)$. Combining this fact with Lemma 2.1, we get

$$\begin{aligned} \|x^m - x^n\|^2 &\leq \|x^m - x^g\|^2 - \|x^n - x^g\|^2 \\ &= (\|x^m - x^g\| - \|x^n - x^g\|)(\|x^m - x^g\| + \|x^n - x^g\|). \end{aligned}$$

Since $\lim_{k\to+\infty} ||x^k - x^g||$ exists, this implies that $\lim_{m,n\to+\infty} ||x^m - x^n|| = 0$. Therefore, $\{x^k\}$ is a Cauchy sequence, so

$$\lim_{k \to \infty} x^k = \bar{x}.$$
(3.38)

By Step III we get

$$||d^k - x^{k+1}|| \le ||v^k - x^{k+1}|| \le ||x^k - x^{k+1}||.$$

Therefore,

$$\begin{aligned} |d^{k} - x^{k}|| &\leq ||d^{k} - x^{k+1}|| + ||x^{k+1} - x^{k}|| \\ &\leq ||x^{k} - x^{k+1}|| + ||x^{k} - x^{k+1}|| \\ &= 2||x^{k} - x^{k+1}|| \end{aligned}$$
(3.39)

and

$$\begin{aligned} \|v^{k} - x^{k}\| &\leq \|v^{k} - x^{k+1}\| + \|x^{k+1} - x^{k}\| \\ &\leq \|x^{k} - x^{k+1}\| + \|x^{k} - x^{k+1}\| \\ &= 2\|x^{k} - x^{k+1}\|. \end{aligned}$$
(3.40)

So, from (3.39), (3.40), and (3.38) we get that

$$\lim_{k \to \infty} \|d^k - x^k\| = \lim_{k \to \infty} \|v^k - x^k\| = 0.$$
(3.41)

In view of (3.36) and (3.41), we have

$$(1 - \beta_k)(\beta_k - \bar{L}) \|\bar{S}_k u^k - u^k\|^2 + \mu (1 - \mu \|A\|^2) \|\bar{T}_k w^k - A v^k\|^2 + \mu \|w^k - A v^k\|^2$$

$$\leq \|x^k - x^*\|^2 - \|d^k - x^*\|^2$$

$$= (\|x^k - x^*\| + \|d^k - x^*\|)(\|x^k - x^*\| - \|d^k - x^*\|)$$

$$\leq \|x^k - d^k\|(\|x^k - x^*\| + \|d^k - x^*\|) \to 0 \text{ as } k \to \infty.$$
(3.42)

Since $0 < \overline{L} < \beta_k < 1$ and $\mu \in \left(0, \frac{1}{\|A\|}\right)$, we deduce from (3.42) that

$$\lim_{k \to +\infty} \|\bar{S}_k u^k - u^k\| = 0, \lim_{k \to +\infty} \|\bar{T}_k w^k - A v^k\| = 0, \text{ and } \lim_{k \to +\infty} \|w^k - A v^k\| = 0.$$
(3.43)

In addition, from the inequality

$$\|\bar{T}_k w^k - w^k\| \le \|\bar{T}_k w^k - A v^k\| \|w^k - A v^k\|,$$

combined with (3.43), we get

$$\lim_{k \to +\infty} \|\bar{T}_k w^k - w^k\| = 0.$$
(3.44)

Besides, (3.16), (3.40), and $\lim_{k\to+\infty} x^k = \bar{x}$ it imply

$$\lim_{k \to +\infty} u^k = \bar{x}, \lim_{k \to +\infty} v^k = \bar{x}.$$
(3.45)

Since

$$\begin{split} \|\bar{S}_{k}\bar{x}-\bar{x}\|^{2} &\leq \|\bar{S}_{k}\bar{x}-\bar{S}_{k}u^{k}\|^{2}+\|\bar{S}_{k}u^{k}-u^{k}\|^{2}+\|u^{k}-\bar{x}\|^{2} \\ &\leq \left(\|u^{k}-x^{*}\|^{2}+\bar{L}\|(I-\bar{S}_{k})(u^{k})-(I-\bar{S}_{k})(x^{*})\|^{2}\right) \\ &+\|\bar{S}_{k}u^{k}-u^{k}\|+\|u^{k}-\bar{x}\| \\ &= 2\|u^{k}-\bar{x}\|^{2}+(\bar{L}+1)\|\bar{S}_{k}u^{k}-u^{k}\|^{2}, \end{split}$$

from (3.43) and (3.45) we get that $\|\bar{S}_k\bar{x}-\bar{x}\|=0$, that is, $\bar{x}\in \text{Fix}(\bar{S}_k)$. From (3.18) we have

$$\lim_{k \to +\infty} \eta_k \|x^k - y^k\|^2 = 0.$$
(3.46)

We now consider two distinct cases.

Case I. $\limsup_{k\to\infty} \eta_k > 0$. Then there exist $\overline{\eta} > 0$ and a subsequence $\{\eta_{k_i}\} \subset \{\eta_k\}$ such that $\eta_{k_i} > \overline{\eta}$ for all *i*. So we get from (3.46) that

$$\lim_{k \to +\infty} \|x^{k_i} - y^{k_i}\| = 0.$$
(3.47)

Since $x^k \to \bar{x}$, (3.47) implies that $y^{k_i} \to \bar{x}$ as $i \to \infty$. For each $y \in C$, we get from (3.22) that

$$f(x^{k_i}, y) - f(x^{k_i}, y^{k_i}) + \frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\| \|y - y^{k_i}\| \ge 0.$$
(3.48)

Letting $i \to \infty$, by the continuity of f, since $x^{k_i} \to \bar{x}$ and $y^{k_i} \to \bar{x}$, in the limit, from (3.48) we obtain that $f(\bar{x}, y) - f(\bar{x}, \bar{x}) \ge 0$. Hence, $f(\bar{x}, y) \ge 0$, $\forall y \in C$, so \bar{x} is a solution of

 $\operatorname{EP}(C, f).$

Case II. $\lim_{k\to\infty} \eta_k = 0$. From the boundedness of $\{y^k\}$, we deduce that there exists $\{y^{k_i}\} \subset \{y^k\}$ such that $y^{k_i} \rightharpoonup \overline{y}$ as $i \to \infty$. Replacing y by y^{k_i} in (3.21), we get

$$f(x^{k_i}, y^{k_i}) + \frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2 \le 0.$$
(3.49)

On the other hand, by the Armijo linesearch rule (3.35), for $m_{k_i} - 1$, there exists $z^{k_i, m_{k_i} - 1}$ such that

$$f(z^{k_i,m_{k_i}-1},x^{k_i}) - f(z^{k_i,m_{k_i}-1},y^{k_i}) < -\frac{\theta}{2\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2.$$

Combining this with (3.49), we get

$$f(z^{k_i,m_{k_i}-1}, y^{k_i}) - f(z^{k_i,m_{k_i}-1}, x^{k_i}) > -\frac{\theta}{2\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2 \ge \frac{2}{\theta} f(x^{k_i}, y^{k_i}).$$
(3.50)

According to the algorithm, we have $z^{k_i,m_{k_i}-1} = (1 - \eta^{m_{k_i}-1})x^{k_i} + \eta^{m_{k_i}-1}y^{k_i}$. Since $\eta^{m_{k_i}-1} \to 0$, x^{k_i} converges weakly to \bar{x} , and y^{k_i} converges weakly to \bar{y} , this implies that $z^{k_i,m_{k_i}-1} \rightharpoonup \bar{x}$ as $i \to \infty$. Beside that, $\left\{\frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2\right\}$ is bounded, so without loss of generality, we may assume that $\lim_{i\to+\infty} \frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2$ exists. Hence, we obtain in the limit (3.50) that

$$f(\bar{x}, \bar{y}) \ge -2 \lim_{i \to +\infty} \frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2 \ge \theta f(\bar{x}, \bar{y}).$$

Therefore, $f(\bar{x}, \bar{y}) = 0$ and $\lim_{i \to +\infty} \frac{1}{\rho_{k_i}} \|y^{k_i} - x^{k_i}\|^2 = 0$. By Case I we get $\bar{x} \in \bigcap_{i=1}^p \operatorname{Fix}(S_i, C)$. So

$$\bar{x} \in S_{\rm EP}(f) \cap \Big(\bigcap_{i=1}^{p} \operatorname{Fix}(S_i, C)\Big).$$
(3.51)

We get from (3.45) that $\lim_{k\to+\infty} Av^k = A\bar{x}$. Combining this with (3.43) yields

$$\lim_{k \to +\infty} w^k = A\bar{x}.$$
(3.52)

Since,

$$\|\bar{T}_k A \bar{x} - A \bar{x}\| \le \|\bar{T}_k A \bar{x} - \bar{T}_k w^k\| + \|\bar{T}_k w^k - w^k\| + \|w^k - A \bar{x}\|.$$

Then, by Proprositon 2.5(i), we obtain

$$\begin{aligned} \|\bar{T}_k A \bar{x} - A \bar{x}\| &\leq \frac{1 + \bar{L}}{1 - \bar{L}} \|A \bar{x} - w^k\| + \|\bar{T}_k w^k - w^k\| + \|w^k - A \bar{x}\| \\ &= \left(1 + \frac{1 + \bar{L}}{1 - \bar{L}}\right) \|w^k - A \bar{x}\| + \|\bar{T}_k w^k - w^k\|. \end{aligned}$$

So, from (3.52) and (3.44) we get that

$$\|\bar{T}_k A \bar{x} - A \bar{x}\| = 0$$
, for each $j \in \{1, 2, \dots, p'\}$.

Hence for each $j \in \{1, 2, ..., p'\}$, we get $A\bar{x} \in \bigcap_{j=1}^{p'} \operatorname{Fix}(T_j, Q)$. Moreover,

$$\begin{split} \|T_{\beta}^{(g,h)}A\bar{x} - A\bar{x}\| &\leq \|T_{\beta}^{(g,h)}A\bar{x} - T_{\alpha_{k}}^{(g,h)}Av^{k} - Av^{k}\| + \|T_{\alpha_{k}}^{(g,h)}Av^{k} - Av^{k}\| \\ &+ \|Av^{k} - A\bar{x}\| \\ &= \|T_{\beta}^{(g,h)}A\bar{x} - T_{\alpha_{k}}^{(g,h)}Av^{k} - Av^{k}\| + \|w^{k} - Av^{k}\| \\ &+ \|Av^{k} - A\bar{x}\| \\ &\leq \|Av^{k} - A\bar{x}\| + \frac{|\alpha_{k} - \beta|}{\alpha_{k}}\|T_{\alpha_{k}}^{(g,h)}Av^{k} - Av^{k}\| + \|w^{k} - Av^{k}\| \\ &+ \|Av^{k} - A\bar{x}\| \\ &= 2\|Av^{k} - A\bar{x}\| + \frac{|\alpha_{k} - \beta|}{\alpha_{k}}\|w^{k} - Av^{k}\| + \|w^{k} - Av^{k}\|, \end{split}$$

where the last inequality comes from Lemma 2.12. Letting $k \to \infty$ and recalling that $\lim_{k\to+\infty} Av^k = A\bar{x}$. Then, we get

$$\|T^{(g,h)}_{\alpha}A\bar{x} - A\bar{x}\| = 0$$

Thus, $A\bar{x} \in Fix(T_{\alpha}^{(g,h)}) = S_{GEP}(Q, g, h)$. Hence,

$$A\bar{x} \in S_{\text{GEP}}(g,h) \cap \Big(\bigcap_{j=1}^{p'} \operatorname{Fix}(T_j,Q)\Big).$$

So, we conclude that $\bar{x} \in \Omega$, the proof of the theorem is complete.

For each $i \in \{1, 2, ..., p\}$ and $j \in \{1, 2, ..., p'\}$ putting $S_i = S$ and $T_j = S$, where S and T are nonexpansive mappings, and h = 0 in Theorem 3.4, we have the following result.

Corollary 3.5. [11] Let $\mathcal{H}_1, \mathcal{H}_2$ be two real Hilbert spaces and $C \subset \mathcal{H}_1, Q \subset \mathcal{H}_2$ be nonempty closed convex subsets. Let $S: C \to C; T: Q \to Q$ be a nonexpansive mapping, and let bifunctions f and g satisfy Assumptions I and II, respectively. Let $A: \mathcal{H}_1 \to \mathcal{H}_2$ be a bounded linear operator with its adjoint A^* . If

 $\Omega := \{ x^* \in S_{EP}(C, f) \cap Fix(S) : Ax^* \in Fix(Q, g) \cap Fix(T) \} \neq \emptyset.$

Then the sequences $\{x^k\}, \{u^k\}$ and $\{v^k\}$ be generated by the following :

Step I: Solve the strongly convex program :

$$y^k := \operatorname{argmin} \left\{ f(x^k, y) + \frac{1}{2\rho_k} \|y - x^k\|^2 : y \in C \right\}$$

to obtain its unique solution y^k . If $y^k = x^k$, then set $u^k = x^k$ and go to Step III. Otherwise, go to Step II.

Step II : (Armijo linesearch rule) Find m_k as the smallest positive integer number m such that

$$\begin{cases} z^{k,m} = (1 - \eta^m) x^k + \eta^m y^m, \\ f(z^{k,m}, x^k) - f(z^{k,m}, y^k) \ge \frac{\theta}{2\rho_k} \|x^k - y^k\|^2. \end{cases}$$
(3.53)

Set $\eta_k = \eta^{m_k}, z^k = z^{k,m_k}$. **Step III :** Take $t^k \in \partial f(z^k, x^k), \sigma_k = \frac{f(z^k, x^k)}{\|t^k\|^2}$, and denote $\begin{cases}
u^k = P_C(x^k - \gamma_k \sigma_k t^k), \\
v^k = (1 - \beta)u^k + \beta S(u^k), \\
w^k = T^g_{\alpha_k} Av^k, \\
x^{k+1} := P_C(v^k + \mu A^*(T(w^k) - Av^k))) \\
and go to iteration k with k replaced by <math>k + 1, \\
converge strongly to an element \bar{x} \in \Omega, and \{w^k\} converges strongly to A\bar{x}.
\end{cases}$

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