

## EXPLICIT ITERATIVE METHODS FOR THE SPLIT FEASIBILITY PROBLEM WITH MULTIPLE OUTPUT SETS

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**Abstract.** The purpose of this paper is to introduce some new explicit iterative methods for finding a solution of the split feasibility problem with multiple output sets. These methods are established by using the Tikhonov regularization method in real Hilbert spaces.

**Keywords.** Multiple output sets; Nonexpansive mapping; Regularization; Split feasibility problem.

**2020 Mathematics Subject Classification.** 65K05, 90C30.

### 1. INTRODUCTION

Let  $H_1$  and  $H_2$  be two real Hilbert spaces. Let  $C$  and  $Q$  be nonempty, closed, and convex subsets of  $H_1$  and  $H_2$ , respectively. Let  $T : H_1 \rightarrow H_2$  be a bounded and linear operator. The *split convex feasibility problem* (SCFP, for short) is presented as follows:

$$\text{Find an element } u^* \in C \text{ such that } Tu^* \in Q. \quad (1.1)$$

The SCFP was first introduced by Censor and Elfving [4] in order to model certain inverse problems. It plays an important role in medical image reconstruction and in signal processing; see, e.g., [1, 2]. Recently, various iterative algorithms were introduced for solving (1.1); see, e.g., [1, 2, 3, 5, 6, 9, 15, 16, 17, 19, 21, 24] and the references therein.

In 2010, Xu [19] introduced the following iterative method for solving Problem (1.1). For any  $u_0 \in H$ , he defined the sequence  $\{u_n\}$  by

$$u_{n+1} = P_C[(1 - t_n \varepsilon_n)u_n - \varepsilon_n T^*(I - P_Q)Tu_n], \quad n \geq 0. \quad (1.2)$$

He proved that the sequence  $\{u_n\}$  generated by (1.2) converges strongly to the minimum-norm solution to Problem (1.1) when  $\{t_n\}$  and  $\{\varepsilon_n\}$  satisfying the conditions below:

- i)  $t_n \rightarrow 0$  and  $0 < \varepsilon_n < \frac{t_n}{\|T\|^2 + t_n}$ ;
- ii)  $\sum_{n=0}^{\infty} t_n \varepsilon_n = \infty$ ;
- iii)  $\frac{|\varepsilon_{n+1} - \varepsilon_n| + \varepsilon_n |t_{n+1} - t_n|}{t_n^2 \varepsilon_n^2} \rightarrow 0$ .

In 2012, Yao et al. [23] proved the strong convergence of iterative method (1.2) under the following conditions:

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Received 14 March 2024; Accepted 2 September 2024; Published online 20 March 2025.

- i)  $t_n \rightarrow 0$ , and  $\sum_{n=0}^{\infty} t_n = \infty$ ;
- ii)  $0 < \varepsilon_n < \frac{2}{\|T\|^2 + 2t_n}$ ,  $\inf_n \varepsilon_n > 0$  and  $|\varepsilon_{n+1} - \varepsilon_n| \rightarrow 0$ .

In 2020, Reich et al. [10] presented and studied the following split feasibility problem with multiple output sets in Hilbert spaces: Let  $H, H_i, i = 1, 2, \dots, m$ , be real Hilbert spaces, and let  $T_i : H \rightarrow H_i, i = 1, 2, \dots, m$ , be bounded linear operators. Let  $C$  and  $Q_i$  be nonempty, closed, and convex subsets of  $H$  and  $H_i, i = 1, 2, \dots, m$ , respectively. Suppose that  $\Omega^{SFPMS} = C \cap (\cap_{i=1}^m T_i^{-1}(Q_i)) \neq \emptyset$ . They considered the following problem:

$$\text{Find an element } u^* \in \Omega^{SFPMS}, \quad (1.3)$$

that is, a point  $u^* \in C$  such that  $T_i u^* \in Q_i$  for all  $i = 1, 2, \dots, m$ . In order to solve Problem (1.3), Reich et al. [10, 11] introduced some iterative methods which are based on the optimization approach. In 2022, Reich and Tuyen [12] proposed and studied the strong convergence of the following iterative scheme. Take any  $u_0 \in H$  and define the sequence  $\{u_n\}$  by

$$u_{n+1} = u_n - \varepsilon_n(F(u_n) + t_n U(u_n)), \quad n \geq 0, \quad (1.4)$$

where  $F = I - P_C + \sum_{i=1}^m T_i^*(I - P_{Q_i})T_i$  and  $U : H \rightarrow H$  is  $L_U$ -Lipschitz and  $\gamma_U$ -strongly monotone. They proved that the sequence  $\{u_n\}$  defined by (1.4) converges strongly to a solution of Problem (1.3) when the parameters control satisfy the following conditions:

- i)  $\lim_{n \rightarrow \infty} t_n = 0, \{t_n\} \subset (0, (\gamma_U - \varepsilon_n K L_U) / \varepsilon_n L_U^2)$ , where  $K = 1 + \sum_{i=1}^m \|T_i\|^2$ ;
- ii)  $\{\varepsilon_n\} \subset (0, \gamma / 2 K L_U)$  and  $\sum_{n=1}^{\infty} t_n \varepsilon_n = \infty$ ;
- iii)  $\lim_{n \rightarrow \infty} \varepsilon_n / t_n = 0$ ;
- iv)  $\lim_{n \rightarrow \infty} \frac{|t_{n+1} - t_n|}{t_n \varepsilon_n} = 0$ .

In this paper, we analyze and establish the strong convergence of iterative scheme (1.4) based on some conditions which are simpler than the conditions of Reich and Tuyen in [12]. We introduce several relaxed iterative methods for solving Problem (1.3) in the case where  $C$  and  $Q_i, i = 1, 2, \dots, m$ , are sublevel sets of convex functions. Two numerical examples are also presented to illustrate proposed methods.

## 2. PRELIMINARIES

Let  $H$  be a real Hilbert space. We denote by  $\langle u, v \rangle$  the inner product of two elements  $u, v$  in  $H$ . The induced norm is denoted by  $\|\cdot\|$ , that is,  $\|u\| = \sqrt{\langle u, u \rangle}$  for all  $u \in H$ .

Let  $C$  be a nonempty, closed, and convex subset of  $H$ . It is known that, for each  $u \in H$ , there exists a unique point  $P_C u \in C$  such that

$$\|u - P_C u\| = \inf_{v \in C} \|u - v\|. \quad (2.1)$$

The mapping  $P_C : H \rightarrow C$  defined by (2.1) is called the *metric projection* of  $H$  onto  $C$ . We also recall (see, e.g., [8, Section 3]) that

$$\langle u - P_C u, v - P_C u \rangle \leq 0, \quad \forall u \in H, \quad \forall v \in C. \quad (2.2)$$

Let  $S, A : H \rightarrow H$  are two operators from  $H$  into itself.

- i)  $S$  is  $L_S$  Lipschitz if there exists a positive real number  $L_S > 0$  such that

$$\|S(u) - S(v)\| \leq L_S \|u - v\|$$

- for all  $u, v \in H$ . If  $L_S = 1$ , then we say that  $S$  is *nonexpansive*. In addition, if  $L_U \in [0, 1)$ , then  $S$  is called a *strict contraction*;
- ii)  $S$  is *firmly nonexpansive* if  $2S - I$  is nonexpansive, which is equivalent to  $S = (I + U)/2$ , where  $U : H \rightarrow H$  is nonexpansive;
  - iii)  $S$  is *averaged* if  $S = (1 - t)I + tU$ , where  $t \in (0, 1)$  and  $U : H \rightarrow H$  is nonexpansive. In this case, we say that  $S$  is  $t$ -averaged.
  - iv)  $A$  is *monotone* if  $\langle u - v, A(u) - A(v) \rangle \geq 0$  for all  $u, v \in H$ ;
  - v)  $A$  is  $\beta_A$ -*strongly monotone* with  $\beta_A > 0$  if  $\langle u - v, A(u) - A(v) \rangle \geq \beta_A \|u - v\|^2$  for all  $u, v \in H$ ;
  - vi)  $A$  is  $\gamma_A$ -*co-coercive* if  $\langle u - v, A(u) - A(v) \rangle \geq \gamma_A \|A(u) - A(v)\|^2$  for all  $u, v \in H$ .

We also need the following lemmas for our main results of this paper.

**Lemma 2.1.** (see [10]) *Let  $H$  be a real Hilbert space. Let  $C$  be a nonempty, closed, and convex subset of  $H$ . Then, for all  $u, v \in H$ ,*

- i)  $\langle u - v, P_C u - P_C v \rangle \geq \|P_C u - P_C v\|^2$ ;
- ii)  $\langle u - v, (I - P_C)u - (I - P_C)v \rangle \geq \|(I - P_C)u - (I - P_C)v\|^2$ .

**Remark 2.1.** It follows from Lemma 2.1 that  $I - P_C$  is a nonexpansive mapping.

**Lemma 2.2.** (see [2, 20]) *The following statements hold:*

- i) *If  $A$  is  $\gamma_A$ -co-coercive, then  $\varepsilon A$  is  $\gamma_A/\varepsilon$ -co-coercive.*
- ii)  *$S$  is averaged if and only if the component  $I - S$  is  $\gamma$ -co-coercive with  $\gamma > 1/2$ . More precisely, for  $t \in (0, 1)$ ,  $S$  is  $t$ -averaged if and only if  $I - S$  is  $1/2t$ -co-coercive.*

**Lemma 2.3.** [7] *Let  $T$  be a nonexpansive self-mapping of a closed and convex subset  $C$  of a Hilbert space  $H$ . Then  $I - T$  is demiclosed, that is, whenever  $\{u_n\}$  is a sequence in  $C$  which weakly converges to some  $u \in C$  and the sequence  $\{(I - T)(u_n)\}$  strongly converges to some  $v$ , it follows that  $(I - T)(u) = v$ .*

**Lemma 2.4.** [14] *Let  $\{a_n\}$  and  $\{b_n\}$  be bounded sequences in a Hilbert space  $H$ , and let  $\{t_n\}$  be a sequence in  $[0, 1]$  with  $0 < \liminf_{n \rightarrow \infty} \beta_n \leq \limsup_{n \rightarrow \infty} \beta_n < 1$ . Let  $a_{n+1} = (1 - \beta_n)b_n + \beta_n a_n$  for all  $n \geq 0$  and  $\limsup_{n \rightarrow \infty} (\|b_{n+1} - b_n\| - \|a_{n+1} - a_n\|) \leq 0$ . Then  $\lim_{n \rightarrow \infty} \|a_n - b_n\| = 0$ .*

**Lemma 2.5.** [18] *Let  $\{\Gamma_n\}$  be a sequence of nonnegative numbers,  $\{b_n\}$  be a sequence in  $(0, 1)$ , and  $\{c_n\}$  a sequence of real numbers satisfying the following two conditions:*

- i)  $\Gamma_{n+1} \leq (1 - b_n)\Gamma_n + b_n c_n$ ;
- ii)  $\sum_{n=0}^{\infty} b_n = \infty$ ,  $\limsup_{n \rightarrow \infty} c_n \leq 0$ .

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

### 3. MAIN RESULTS

Consider Problem (1.3), and let  $\Psi : H \rightarrow \mathbb{R}$  be defined by

$$\Psi(u) := \frac{1}{2} \|(I - P_C)u\|^2 + \frac{1}{2} \sum_{i=1}^m \|(I - P_{Q_i})T_i u\|^2$$

for all  $u \in H$ .

It is not difficult to see that  $\Psi$  is a convex, continuous, and proper function. Indeed, it is easy to see that  $\Psi$  is a continuous and proper function. We now prove that  $\Psi$  is a convex function. To

do this, we first show that function  $d_C(u) = \|(I - P_C)u\|$  for all  $u \in H$ , is a convex function. For every  $x, y \in H$ , it follows from the definition of  $d_C(x)$  and  $d_C(y)$  that there exist two sequences  $\{x_n\}$  and  $\{y_n\}$  in  $C$  such that  $\|x - x_n\| \rightarrow d_C(x)$  and  $\|y - y_n\| \rightarrow d_C(y)$ , as  $n \rightarrow \infty$ . Since  $C$  is a convex set,  $\lambda x_n + (1 - \lambda)y_n \in C$  for all  $\lambda \in [0, 1]$  and for all  $n \geq 1$ . Thus

$$\begin{aligned} d_C(\lambda x + (1 - \lambda)y) &= \inf_{z \in C} \|\lambda x + (1 - \lambda)y - z\| \\ &\leq \|\lambda x + (1 - \lambda)y - [\lambda x_n + (1 - \lambda)y_n]\| \\ &\leq \lambda \|x - x_n\| + (1 - \lambda)\|y - y_n\| \end{aligned}$$

for all  $n \geq 1$ . Letting  $n \rightarrow \infty$ , we obtain

$$d_C(\lambda x + (1 - \lambda)y) \leq \lambda d_C(x) + (1 - \lambda)d_C(y).$$

This shows that  $d_C(u)$  is a convex function. Hence,  $f(u) = d^2(u, C)/2$  is also a convex function (note that, the square of a nonnegative convex function is a convex function).

For each  $i = 1, 2, \dots, m$  and for every  $x, y \in H$ , and for any  $\lambda \in [0, 1]$ , it follows from the convexity of  $\|(I - P_{Q_i})v\|^2$  on  $H_i$  that

$$\begin{aligned} \|(I - P_{Q_i})T_i[\lambda x + (1 - \lambda)y]\|^2 &= \|(I - P_{Q_i})[\lambda T_i x + (1 - \lambda)T_i y]\|^2 \\ &\leq \lambda \|(I - P_{Q_i})T_i x\|^2 + (1 - \lambda)\|(I - P_{Q_i})T_i y\|^2, \end{aligned}$$

which implies that  $\|(I - P_{Q_i})T_i u\|^2$  is a convex function on  $H$ . Thus we conclude that  $\Psi$  is a convex function.

Let  $f_i(u) = \|(I - P_{Q_i})T_i u\|^2/2$  for all  $u \in H$ . We next prove that  $\nabla f_i(u) = T_i^*(I - P_{Q_i})T_i u$ . Indeed, we take any point  $x_0 \in H$  and letting  $v = T_i^*(I - P_{Q_i})T_i x_0$ . For every  $h \in H$ , we have

$$\begin{aligned} f_i(x_0 + h) - f_i(x_0) - \langle v, h \rangle &= \frac{1}{2} \|(I - P_{Q_i})T_i(x_0 + h)\|^2 - \frac{1}{2} \|(I - P_{Q_i})T_i x_0\|^2 - \langle v, h \rangle \\ &= \frac{1}{2} (\|T_i(x_0 + h) - P_{Q_i}T_i(x_0 + h)\|^2 - \|(I - P_{Q_i})T_i x_0\|^2) - \langle v, h \rangle \\ &\leq \frac{1}{2} (\|T_i(x_0 + h) - P_{Q_i}T_i x_0\|^2 - \|(I - P_{Q_i})T_i x_0\|^2) - \langle v, h \rangle \\ &= \frac{1}{2} (\|(I - P_{Q_i}T_i)x_0 + T_i h\|^2 - \|(I - P_{Q_i})T_i x_0\|^2) - \langle v, h \rangle \\ &= \frac{1}{2} (\|(I - P_{Q_i}T_i)x_0\|^2 + \|T_i h\|^2 - \|(I - P_{Q_i})T_i x_0\|^2) \\ &\quad + \langle (I - P_{Q_i}T_i)x_0, T_i h \rangle - \langle v, h \rangle \\ &= \frac{1}{2} \|T_i h\|^2 + \langle T_i^*(I - P_{Q_i}T_i)x_0, h \rangle - \langle v, h \rangle \\ &\leq \frac{1}{2} \|T_i\|^2 \|h\|^2. \end{aligned}$$

Similarly, we also have

$$f_i(x_0) - f_i(x_0 + h) + \langle v, h \rangle \leq \frac{1}{2} \|T_i\|^2 \|h\|^2.$$

Combining the two above inequalities, we see that

$$\frac{|f_i(x_0 + h) - f_i(x_0) - \langle v, h \rangle|}{\|h\|} \leq \frac{1}{2} \|T_i\|^2 \|h\| \rightarrow 0,$$

as  $\|h\| \rightarrow 0$ . Hence,  $\nabla f_i(x_0) = T_i^*(I - P_{Q_i})T_i x_0$ . Then we infer that  $\Psi$  is a Fréchet differentiable function and

$$\nabla \Psi(u) = (I - P_C)u + \sum_{i=1}^m T_i^*(I - P_{Q_i})T_i u$$

for all  $u \in H$ .

By Rockafellar's theorem [13],  $F := \nabla \Psi$  is a maximal monotone operator. Moreover, a point  $u^* \in H$  is a solution to Problem (1.3) if and only if  $u^*$  is a minimum point of  $\Psi$ . This is equivalent to

$$F(u^*) = (I - P_C)u^* + \sum_{i=1}^m T_i^*(I - P_{Q_i})T_i u^* = 0. \quad (3.1)$$

We first consider the following Tikhonov regularization method

$$\min_{u \in H} \left\{ \Psi(u) + \frac{t}{2} \|u\|^2 \right\},$$

where  $t > 0$ . We see that

$$\nabla \left( \Psi(u) + \frac{t}{2} \|u\|^2 \right) = F(u) + tu,$$

for all  $u \in H$ . Thus, in this case, we study and establish the convergence of the following explicit iterative method: For any  $u_0 \in H$ , construct the sequence  $\{u_n\}$  by

$$u_{n+1} = u_n - \varepsilon_n [F(u_n) + t_n u_n], \quad n \geq 0, \quad (3.2)$$

where  $\{t_n\} \subset (0, 1)$  and  $\{\varepsilon_n\}$  is a sequence of real numbers. Note that the sequence  $\{u_n\}$  generated by (3.2) can be rewritten in the following form

$$u_{n+1} = (1 - t_n \varepsilon_n) u_n - \varepsilon_n F(u_n), \quad n \geq 0. \quad (3.3)$$

In order to establish the strong convergence of the iterative method (3.3), we first introduce the following proposition.

**Proposition 3.1.** *The mapping  $F$  is  $\gamma_F$ -co-coercive with  $\gamma_F = 1/(1 + \sum_{i=1}^m \|T_i\|^2)$ .*

*Proof.* For any  $u, v \in H$ , it follows from Lemma 2.1 ii) that

$$\begin{aligned} \langle u - v, F(u) - F(v) \rangle &= \langle u - v, (I - P_C)u - (I - P_C)v \rangle \\ &\quad + \sum_{i=1}^m \langle u - v, T_i^*(I - P_{Q_i})T_i u - T_i^*(I - P_{Q_i})T_i v \rangle \\ &= \langle u - v, (I - P_C)u - (I - P_C)v \rangle \\ &\quad + \sum_{i=1}^m \langle T_i u - T_i v, (I - P_{Q_i})T_i u - (I - P_{Q_i})T_i v \rangle \\ &\geq \|(I - P_C)u - (I - P_C)v\|^2 \\ &\quad + \sum_{i=1}^m \|(I - P_{Q_i})T_i u - (I - P_{Q_i})T_i v\|^2. \end{aligned} \quad (3.4)$$

We also have

$$\begin{aligned}
& \|F(u) - F(v)\|^2 \\
&= \|(I - P_C)u - (I - P_C)v + \sum_{i=1}^m T_i^*(I - P_{Q_i})T_i u - T_i^*(I - P_{Q_i})T_i v\|^2 \\
&\leq [\|(I - P_C)u - (I - P_C)v\| + \sum_{i=1}^m \|T_i^*(I - P_{Q_i})T_i u - T_i^*(I - P_{Q_i})T_i v\|]^2 \\
&\leq [\|(I - P_C)u - (I - P_C)v\| + \sum_{i=1}^m \|T_i\| \|(I - P_{Q_i})T_i u - (I - P_{Q_i})T_i v\|]^2 \\
&\leq (1 + \sum_{i=1}^m \|T_i\|^2) [\|(I - P_C)u - (I - P_C)v\|]^2 \\
&\quad + \sum_{i=1}^m \|(I - P_{Q_i})T_i u - (I - P_{Q_i})T_i v\|^2.
\end{aligned} \tag{3.5}$$

From (3.4) and (3.5), we obtain that

$$\langle u - v, F(u) - F(v) \rangle \geq \frac{1}{1 + \sum_{i=1}^m \|T_i\|^2} \|F(u) - F(v)\|^2,$$

for all  $u, v \in H$ , that is,  $F$  is  $\gamma_F$ -co-coercive with  $\gamma_F = 1/(1 + \sum_{i=1}^m \|T_i\|^2)$ . This completes the proof.  $\square$

The following proposition is an important result that is needed to prove the strong convergence of iterative method (3.2).

**Proposition 3.2.** *If  $t \in (0, 1)$  and  $\varepsilon \in (0, \frac{2\gamma_F}{1 + 2t\gamma_F})$ , then  $G^{t,\varepsilon} = (1 - t\varepsilon)I - \varepsilon F$  is a strict contraction mapping with the coefficient  $k = 1 - t\varepsilon$ .*

*Proof.* For any  $u, v \in H$ , using Proposition 3.1, we have

$$\begin{aligned}
\|G^{t,\varepsilon}(u) - G^{t,\varepsilon}(v)\|^2 &= \|(1 - t\varepsilon)(u - v) - \varepsilon(F(u) - F(v))\|^2 \\
&= (1 - t\varepsilon)^2 \|u - v\|^2 - 2(1 - t\varepsilon)\varepsilon \langle u - v, F(u) - F(v) \rangle \\
&\quad + \varepsilon^2 \|F(u) - F(v)\|^2 \\
&\leq (1 - t\varepsilon)^2 \|u - v\|^2 - \varepsilon[2(1 - t\varepsilon)\gamma_F - \varepsilon] \|F(u) - F(v)\|^2
\end{aligned}$$

It follows from  $\varepsilon \in (0, \frac{2\gamma_F}{1 + 2t\gamma_F})$  that  $2(1 - t\varepsilon)\gamma_F - \varepsilon > 0$ . Thus

$$\|G^{t,\varepsilon}(u) - G^{t,\varepsilon}(v)\| \leq (1 - t\varepsilon) \|u - v\|,$$

for all  $u, v \in H$ . This completes the proof.  $\square$

**Theorem 3.1.** *Let  $\{t_n\}$  and  $\{\varepsilon_n\}$  be two positive sequences such that  $\{t_n\} \subset (0, 1)$  and  $\{\varepsilon_n\} \subset [a, b] \subset (0, \frac{2\gamma_F}{1 + 2t_n\gamma_F})$ . Let  $\{t_n\}$  satisfy the following conditions*

$$t_n \rightarrow 0, \sum_{n=0}^{\infty} t_n = \infty.$$

Then the sequence  $\{u_n\}$  generated by (3.2) converges strongly to an element  $u^* = P_{\Omega_{SFPMOS}}0$ , as  $n \rightarrow \infty$ .

*Proof.* We first show that  $\{u_n\}$  is bounded. Put  $u^* = P_{\Omega_{SFPMOS}}0$ . It follows from (3.1) and Proposition 3.2 that

$$\begin{aligned} \|u_{n+1} - u^*\| &= \|G^{t_n \varepsilon_n}(u_n) - G^{t_n \varepsilon_n}(u^*) - t_n \varepsilon_n u^*\| \\ &\leq (1 - t_n \varepsilon_n) \|u_n - u^*\| + t_n \varepsilon_n \|u^*\| \\ &\leq \max\{\|u_n - u^*\|, \|u^*\|\} \\ &\vdots \\ &\leq \max\{\|u_0 - u^*\|, \|u^*\|\}, \end{aligned}$$

which implies that  $\{u_n\}$  is bounded.

We now prove that  $\|u_{n+1} - u_n\| \rightarrow 0$ . Indeed, since  $F$  is  $\gamma_F$ -co-coercive, then  $\varepsilon F$  is  $\gamma_F/\varepsilon_n$ -co-coercive. We note that  $\gamma_F/\varepsilon_n > 1/2$ . Thus, from Lemma 2.2, we deduce that  $I - \varepsilon_n F$  is  $\varepsilon_n/2\gamma_F$  averaged, which implies that there exists a nonexpansive mapping  $S$  such that

$$I - \varepsilon_n F = (1 - \frac{\varepsilon_n}{2\gamma_F})I + \frac{\varepsilon_n}{2\gamma_F}S.$$

Thus, we can rewrite (3.3) in the following form

$$\begin{aligned} u_{n+1} &= (I - \varepsilon_n F)(u_n) - t_n \varepsilon_n u_n \\ &= [(1 - \frac{\varepsilon_n}{2\gamma_F})I + \frac{\varepsilon_n}{2\gamma_F}S](u_n) - t_n \varepsilon_n u_n \\ &= \beta_n u_n + (1 - \beta_n)v_n, \end{aligned}$$

where  $v_n = S(u_n) - 2\gamma_F t_n u_n$  and  $\beta_n = 1 - \frac{\varepsilon_n}{2\gamma_F}$ .

Next, we have

$$\begin{aligned} \|v_{n+1} - v_n\| &= \|S(u_{n+1}) - 2\gamma_F t_{n+1} u_{n+1} - S(u_n) + 2\gamma_F t_n u_n\| \\ &\leq \|S(u_{n+1}) - S(u_n)\| + 2\gamma_F \|t_{n+1} u_{n+1} - t_n u_n\| \\ &\leq \|u_{n+1} - u_n\| + 2\gamma_F (t_{n+1} \|u_{n+1} - u_n\| + |t_{n+1} - t_n| \|u_n\|). \end{aligned}$$

It follows from the boundedness of the sequence  $\{u_n\}$  and  $t_n \rightarrow 0$  that

$$\limsup_{n \rightarrow \infty} (\|v_{n+1} - v_n\| - \|u_{n+1} - u_n\|) \leq 0.$$

Since  $\{\varepsilon_n\} \subset [a, b] \subset (0, \frac{2\gamma_F}{1 + 2t_n \gamma_F})$ , then  $0 < \liminf_{n \rightarrow \infty} \beta_n \leq \limsup_{n \rightarrow \infty} \beta_n < 1$ . Thus, from Lemma 2.4, we see that  $\|u_n - v_n\| \rightarrow 0$ . Hence,

$$\|u_{n+1} - u_n\| = \frac{\varepsilon_n}{2\gamma_F} \|u_n - v_n\| \rightarrow 0. \quad (3.6)$$

This together with (3.3) and the condition  $t_n \rightarrow 0$  obtains  $\|F(u_n)\| \rightarrow 0$ . Observe that

$$\begin{aligned} \langle u_n - u^*, F(u_n) \rangle &= \langle u_n - u^*, F(u_n) - F(u^*) \rangle \\ &\geq \|(I - P_C)u_n\|^2 + \sum_{i=1}^m \|(I - P_{Q_i})T_i u_n\|^2, \end{aligned}$$

which together with the boundedness of  $\{u_n\}$  and  $\|F(u_n)\| \rightarrow 0$  deduces that

$$\|(I - P_C)u_n\| \rightarrow 0, \quad (3.7)$$

$$\|(I - P_{Q_i})T_i u_n\| \rightarrow 0, \quad (3.8)$$

for all  $i = 1, 2, \dots, m$ .

We next prove that the weak cluster point set of  $\{u_n\}$  is contained in  $\Omega^{SFPMOS}$ . Indeed, suppose that  $p$  is an arbitrary weak cluster point of  $\{u_n\}$ . There exists a subsequence  $\{u_{k_n}\}$  of  $\{u_n\}$  such that  $u_{k_n} \rightharpoonup p$ , as  $n \rightarrow \infty$ . Since  $T_i$  is a bounded linear operator, one has  $T_i u_{k_n} \rightharpoonup T_i p$  for each  $i = 1, 2, \dots, m$ . Thus, applying Lemma 2.3 and using (3.7)–(3.8), we obtain that  $p \in \Omega^{SFPMOS}$ , as claimed.

Finally, we prove that  $u_n \rightarrow u^*$ , as  $n \rightarrow \infty$ . Note that

$$\begin{aligned} \|u_{n+1} - u^*\|^2 &= \langle G^{t_n, \varepsilon_n}(u_n) - G^{t_n, \varepsilon_n}(u^*) + G^{t_n, \varepsilon_n}(u^*) - u^*, u_{n+1} - u^* \rangle \\ &= \langle G^{t_n, \varepsilon_n}(u_n) - G^{t_n, \varepsilon_n}(u^*), u_{n+1} - u^* \rangle + \langle G^{t_n, \varepsilon_n}(u^*) - u^*, u_{n+1} - u^* \rangle \\ &= \langle G^{t_n, \varepsilon_n}(u_n) - G^{t_n, \varepsilon_n}(u^*), u_{n+1} - u^* \rangle - t_n \varepsilon_n \langle u^*, u_{n+1} - u^* \rangle \\ &\leq \|G^{t_n, \varepsilon_n}(u_n) - G^{t_n, \varepsilon_n}(u^*)\| \|u_{n+1} - u^*\| - t_n \varepsilon_n \langle u^*, u_{n+1} - u^* \rangle \\ &\leq (1 - t_n \varepsilon_n) \|u_n - u^*\| \|u_{n+1} - u^*\| - t_n \varepsilon_n \langle u^*, u_{n+1} - u^* \rangle \\ &\leq (1 - t_n \varepsilon_n) \frac{\|u_n - u^*\|^2 + \|u_{n+1} - u^*\|^2}{2} - t_n \varepsilon_n \langle u^*, u_{n+1} - u^* \rangle \\ &\leq \frac{1 - t_n \varepsilon_n}{2} \|u_n - u^*\|^2 + \frac{1}{2} \|u_{n+1} - u^*\|^2 - t_n \varepsilon_n \langle u^*, u_{n+1} - u^* \rangle, \end{aligned}$$

which implies that

$$\|u_{n+1} - u^*\|^2 \leq (1 - t_n \varepsilon_n) \|u_n - u^*\|^2 - 2t_n \varepsilon_n \langle u^*, u_{n+1} - u^* \rangle. \quad (3.9)$$

Putting  $\Gamma_n = \|u_n - u^*\|^2$ ,  $b_n = t_n \varepsilon_n$  and  $c_n = 2\langle u^*, u^* - u_{n+1} \rangle$ , we can rewrite (3.9) in the following form  $\Gamma_{n+1} \leq (1 - b_n)\Gamma_n + b_n c_n$ . Suppose that  $\{u_{l_n}\}$  is a subsequence of the sequence  $\{u_n\}$  such that

$$\limsup_{n \rightarrow \infty} \langle u^*, u^* - u_n \rangle = \lim_{n \rightarrow \infty} \langle u^*, u^* - u_{l_n} \rangle. \quad (3.10)$$

Since  $\{u_n\}$  is bounded, we see that there exists a subsequence  $\{u_{k_{l_n}}\}$  of  $\{u_{l_n}\}$  such that  $u_{k_{l_n}} \rightharpoonup q$ , as  $n \rightarrow \infty$ . We may assume without loss of generality that  $u_{l_n} \rightharpoonup q$  as  $n \rightarrow \infty$ . From the proof above, we have that  $q \in \Omega^{SFPMOS}$ . It follows from (2.2) and (3.10) that

$$\limsup_{n \rightarrow \infty} \langle u^*, u^* - u_n \rangle = \langle u^*, u^* - q \rangle = \langle 0 - P_{\Omega^{SFPMOS}} 0, q - P_{\Omega^{SFPMOS}} 0 \rangle \leq 0. \quad (3.11)$$

In view of  $\|u_{n+1} - u_n\| \rightarrow 0$ , we imply that  $\limsup_{n \rightarrow \infty} c_n \leq 0$ . Furthermore, it is easy to see that  $\sum_{n=0}^{\infty} t_n \varepsilon_n = \infty$ . Hence, all the conditions of Lemma 2.5 are satisfied. Therefore, we deduce that  $\Gamma_n \rightarrow 0$ , that is,  $u_n \rightarrow u^*$  as  $n \rightarrow \infty$ , as asserted. This completes the proof.  $\square$

Next, we consider the Tikhonov regularization method

$$\min_{u \in H} \left\{ \Psi(u) + \frac{t}{2} \|u - \bar{u}\|^2 \right\}.$$

Using similar arguments as above, we can easily prove the following theorem.



**Theorem 3.2.** Let  $\{t_n\}$  and  $\{\varepsilon_n\}$  be two positive sequences such that  $\{t_n\} \subset (0, 1)$  and  $\{\varepsilon_n\} \subset [a, b] \subset (0, \frac{2\gamma_F}{1+2t_n\gamma_F})$ . For any  $\bar{u} \in H$ , let  $\{u_n\}$  be a sequence defined by  $u_0 \in H$  and

$$u_{n+1} = u_n - \varepsilon_n[F(u_n) + t_n(u_n - \bar{u})], \quad n \geq 0. \quad (3.12)$$

Let  $\{t_n\}$  satisfy  $t_n \rightarrow 0$  and  $\sum_{n=0}^{\infty} t_n = \infty$ . Then  $\{u_n\}$  converges strongly to  $P_{\Omega^{SFP MOS}} \bar{u}$ , as  $n \rightarrow \infty$ .

Finally, we study the convergence of the sequence  $\{w_n\}$  generated by the following scheme: For any  $w_0 \in H$ , we define sequence  $\{w_n\}$  by

$$w_{n+1} = w_n - \varepsilon_n[F(w_n) + t_n U(w_n)], \quad n \geq 0, \quad (3.13)$$

where  $U : H \rightarrow H$  is  $L_U$ -Lipschitz and  $\beta_U$ -strongly monotone operator.

**Theorem 3.3.** Let  $\{t_n\}$  and  $\{\varepsilon_n\}$  be two positive sequences such that  $\{t_n\} \subset (0, 1)$  and  $\{\varepsilon_n\} \subset [a, b] \subset (0, \frac{2\gamma_F}{1+2t_n\gamma_F})$ . Let  $\{t_n\}$  satisfy the conditions:  $t_n \rightarrow 0$  and  $\sum_{n=0}^{\infty} t_n = \infty$ . Then the sequence  $\{w_n\}$  defined by (3.13) converges strongly to an element  $p^* \in \Omega^{SFP MOS}$  which is a unique solution to the following variational inequality

$$\langle y - p^*, U(p^*) \rangle \geq 0, \quad \forall y \in \Omega^{SFP MOS}. \quad (3.14)$$

*Proof.* Let  $\mu$  be a positive number with  $\mu \in (0, 2\beta_U/L_U^2)$ . We write  $t_n = \rho_n \mu$  with  $\rho_n = t_n/\mu$ . Since  $t_n \rightarrow 0$ , we may assume that  $\rho_n < 1$  for all  $n$ . Since  $\mu \in (0, 2\beta_U/L_U^2)$ , we have that  $I - \mu U$  is a strict contraction with the contraction coefficient  $\tau = \sqrt{1 - \mu(2\beta_U - \mu L_U^2)}$  (see [22]). Thus  $P_{\Omega^{SFP MOS}}(I - \mu U)$  is also a strict contraction. Banach fixed point theorem guarantees that  $P_{\Omega^{SFP MOS}}(I - \mu U)$  has a unique fixed point  $p^*$ . It follows from (2.2) that  $p^*$  is a unique solution to variational inequality (3.14).

Let  $\{u_n\}$  be a sequence defined by (3.12), where  $\bar{u} = (I - \mu U)(p^*)$  and  $t_n$  is replaced by  $\rho_n$ . From Theorem 3.2, we obtain that  $u_n \rightarrow p^* = P_{\Omega^{SFP MOS}}(I - \mu U)(p^*)$ , as  $n \rightarrow \infty$ . We now rewrite the formulas to define  $\{u_n\}$  and  $\{w_n\}$  in the following forms:

$$\begin{aligned} u_{n+1} &= G^{\rho_n, \varepsilon_n}(u_n) + \rho_n \varepsilon_n (I - \mu U)(p^*), \\ w_{n+1} &= G^{\rho_n, \varepsilon_n}(w_n) + \rho_n \varepsilon_n (I - \mu U)(w_n). \end{aligned}$$

Note that

$$\begin{aligned} \|w_{n+1} - u_{n+1}\| &\leq \|G^{\rho_n, \varepsilon_n}(w_n) - G^{\rho_n, \varepsilon_n}(u_n)\| + \rho_n \varepsilon_n \|(I - \mu U)(w_n) - (I - \mu U)(p^*)\| \\ &\leq (1 - \rho_n \varepsilon_n) \|w_n - u_n\| + \rho_n \varepsilon_n \tau \|w_n - p^*\| \\ &\leq (1 - \rho_n \varepsilon_n) \|w_n - u_n\| + \rho_n \varepsilon_n \tau (\|w_n - u_n\| + \|u_n - p^*\|) \\ &= [1 - (1 - \tau) \rho_n \varepsilon_n] \|w_n - u_n\| + \rho_n \varepsilon_n \tau \|u_n - p^*\|. \end{aligned} \quad (3.15)$$

Putting  $\Gamma_n = \|w_n - u_n\|$ ,  $b_n = (1 - \tau) \rho_n \varepsilon_n$ , and  $c_n = \frac{\tau}{1 - \tau} \|u_n - p^*\|$  we can rewrite (3.15) in the form  $\Gamma_{n+1} \leq (1 - b_n) \Gamma_n + b_n c_n$ . It is easy to see that  $\sum_{n=0}^{\infty} b_n = \infty$  and  $\lim_{n \rightarrow \infty} c_n = 0$ . Thus all the conditions of Lemma 2.5 are satisfied. Therefore, we infer that  $\Gamma_n \rightarrow 0$ , that is,  $\|w_n - u_n\| \rightarrow 0$ . From  $u_n \rightarrow p^*$ , we obtain that  $w_n \rightarrow p^*$ . This complete the proof.  $\square$

**Remark 3.1.** We see that the strong convergence of e iterative method (3.14) is established based on simpler conditions than the results in [12]. In particular, when  $m = 1$ , Problem (1.3)

becomes Problem (1.1). Thus our results are better than the result in [23] (we remove  $|\varepsilon_{n+1} - \varepsilon_n| \rightarrow 0$ ).

#### 4. RELAXED ITERATIVE METHODS

The relaxed iterative method for solving Problem (1.1) was first introduced and studied in [21]. We now study Problem (1.3) when  $C$  and  $Q_i$ ,  $i = 1, 2, \dots, m$ , are sublevel sets of the lower semicontinuous convex functions  $h : H \rightarrow \mathbb{R}$  and  $h_i : H_i \rightarrow \mathbb{R}$ ,  $i = 1, 2, \dots, m$ , respectively. Suppose that

$$C = \{u \in H : h(u) \leq 0\},$$

$$Q_i = \{u \in H : h_i(T_i u) \leq 0\}, \quad i = 1, 2, \dots, m.$$

We assume that  $h$  and  $h_i$ ,  $i = 1, 2, \dots, m$ , are subdifferentiable on  $H$  and that the subdifferentials  $\partial h$  and  $\partial h_i$ ,  $i = 1, 2, \dots, m$ , are bounded (on bounded sets). Recall that the subdifferential of a convex function  $\Xi : H \rightarrow \mathbb{R}$  is defined by

$$\partial \Xi(u) := \{\xi \in H : \Xi(w) - \Xi(u) \geq \langle w - u, \xi \rangle \quad \forall w \in H\}.$$

For a given point  $u_n \in H$ , we define the subsets  $C_n$  and  $Q_{i,n}$  by

$$C_n := \{u \in H : h(u_n) \leq \langle u_n - u, \xi_n \rangle\},$$

$$Q_{i,n} := \{v \in H_i : h_i(T_i u_n) \leq \langle T_i u_n - v, \eta_{i,n} \rangle\}, \quad i = 1, 2, \dots, m,$$

where  $\xi_n \in \partial h(u_n)$  and  $\eta_{i,n} \in \partial h_i(T_i u_n)$  for all  $i = 1, 2, \dots, m$ . The sets  $C_n$  and  $Q_{i,n}$  are called the relaxed sets of  $C$  and  $Q_i$ , respectively. It is easy to see that  $C_n$  and  $Q_{i,n}$  are half-spaces of  $H$  and  $H_i$ , respectively, and that  $C \subset C_n$  and  $Q_i \subset Q_{i,n}$  for all  $i = 1, 2, \dots, m$ .

It is known that, in the general case, it is not easy to calculate the projections  $P_C x$  and  $P_{Q_i} y$ . Therefore we introduce two relaxed iterative methods corresponding to the proposed iterative methods, when  $P_C$  and  $P_{Q_i}$  are replaced by  $P_{C_n}$  and  $P_{Q_{i,n}}$ , respectively, which are defined as follows:

$$P_{C_n} u := u - \max \left\{ \frac{\langle u, \xi_n \rangle - \langle u_n, \xi_n \rangle + h(u_n)}{\|\xi_n\|^2}, 0 \right\} \xi_n,$$

$$P_{Q_{i,n}} v := v - \max \left\{ \frac{\langle v, \eta_{i,n} \rangle - \langle T_i u_n, \eta_{i,n} \rangle + h_i(T_i u_n)}{\|\eta_{i,n}\|^2}, 0 \right\} \eta_{i,n}.$$

By using a similar technique as in [12], we can easily prove the following theorem.

**Theorem 4.1.** *Let  $\{t_n\}$  and  $\{\varepsilon_n\}$  be two positive sequences such that  $\{t_n\} \subset (0, 1)$  and  $\{\varepsilon_n\} \subset [a, b] \subset (0, \frac{2\gamma_F}{1 + 2t_n\gamma_F})$ . Let  $\mu$  be a positive real number with  $\mu \in (0, 2\beta_U/L_U^2)$ . Suppose that the sequence  $\{t_n\}$  satisfies the following conditions*

$$t_n \rightarrow 0, \quad \sum_{n=0}^{\infty} t_n = \infty.$$

For any  $w_0 \in H$ , let  $\{w_n\}$  be the sequence defined by

$$w_{n+1} = w_n - \varepsilon_n [F_n(w_n) + t_n U(w_n)], \quad n \geq 0,$$

where  $F_n = I - P_{C_n} + \sum_{i=1}^m T_i^*(I - P_{Q_{i,n}})T_i$ . Then  $\{w_n\}$  converges strongly to an element  $p^* \in \Omega^{SFPMOS}$  which is a unique solution to the following variational inequality (3.14).

## 5. NUMERICAL EXPERIMENTS

In this section, our algorithms are implemented in MATLAB 14a running on the DESKTOP-9RLTPS0, Intel(R) Core(TM) i5-10210U CPU @ 1.60GHz with 2.11 GHz and 8GB RAM.

**Example 5.1.** Consider the following split feasibility with multiple output sets: Let  $C$ ,  $Q_1$ ,  $Q_2$ , and  $Q_3$  be closed and convex subsets of  $\mathbb{R}^5$ ,  $\mathbb{R}^4$ ,  $\mathbb{R}^3$ , and  $\mathbb{R}^2$ , respectively, which are defined by

$$C = \{(x_1, x_2, x_3, x_4, x_5) \in \mathbb{R}^5 : x_1 - x_2 + x_3 - x_4 + 2x_5 = 2\},$$

$$Q_1 = \{(y_1, y_2, y_3, y_4) \in \mathbb{R}^4 : y_1 + y_2 - y_3 + y_4 = 1\},$$

$$Q_2 = \{(z_1, z_2, z_3) \in \mathbb{R}^3 : z_1 - 2z_2 + z_3 = 0\},$$

$$Q_3 = \{(v_1, v_2) \in \mathbb{R}^2 : v_1 - v_2 = 1\}.$$

The representing matrices of the transfer mappings  $T_1 : \mathbb{R}^5 \rightarrow \mathbb{R}^4$ ,  $T_2 : \mathbb{R}^5 \rightarrow \mathbb{R}^3$ , and  $T_3 : \mathbb{R}^5 \rightarrow \mathbb{R}^2$  are

$$T_1 = \begin{pmatrix} 1 & 2 & -1 & 1 & 0 \\ 1 & 1 & 2 & -1 & 1 \\ 2 & -1 & 1 & 2 & 0 \\ 1 & -5 & 1 & 1 & -1 \end{pmatrix},$$

$$T_2 = \begin{pmatrix} 2 & 1 & -1 & 1 & 2 \\ 1 & -1 & 2 & 1 & -1 \\ 1 & -4 & 6 & 1 & -4 \end{pmatrix},$$

$$T_3 = \begin{pmatrix} 2 & 1 & 3 & 0 & -1 \\ 1 & 2 & 3 & 0 & -1 \end{pmatrix}.$$

It is easy to check that

$$\Omega^{SFPMS} := C \cap_{i=1}^3 T_i^{-1}(Q_i) = \{(1 + \xi, \xi, -1, -1, 0.5) : \xi \in \mathbb{R}\}.$$

We now test the convergence of the iterative methods (3.2) and (3.12). The parameters  $t_n$  and  $\varepsilon_n$  are chosen as follows:

$$t_n = (n+1)^{-0.595}, \quad \varepsilon_n = \begin{cases} \frac{2\gamma_F}{1+4\gamma_F} & \text{if } n \text{ even,} \\ \frac{2\gamma_F}{1+8\gamma_F} & \text{if } n \text{ odd,} \end{cases}$$

for all  $n \geq 0$ . Note that, in this case,  $|\varepsilon_{n+1} - \varepsilon_n| \rightarrow 0$ .

a) Applying the iterative method (3.2) with the initial point  $u_0 = (1, 2, 3, 4, 5)$ .

We see that  $u^* = (0.5, -0.5, -1, -1, 0.5)$  is the minimum norm solution to the problem. We use the condition  $\sigma_n := \|u_n - u^*\|^2 < \varepsilon$  to stop the iterative process, where  $\varepsilon$  is a given error. We obtain the following table of numerical results.

b) Applying the iterative method (3.12) with the initial point  $u_0 = (1, 2, 3, 4, 5)$  and  $\bar{u} = (2, 1, -2, 1, 2)$ .

We see that  $p^* = P_{\Omega^{SFPMS}} \bar{u} = (2, 1, -1, -1, 0.5)$ . Thus, in this case, we use the condition  $\sigma_n := \|u_n - p^*\|^2 < \varepsilon$  to stop the iterative process, where  $\varepsilon$  is a given error. We obtain the following table of numerical results.

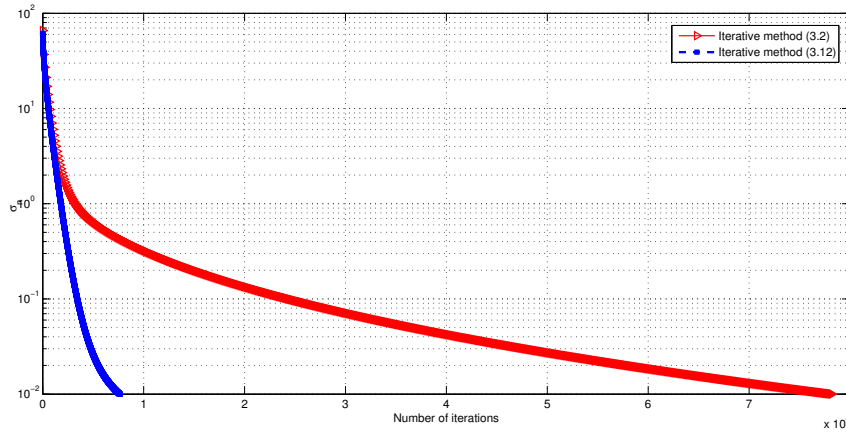
$\varepsilon$	$\sigma_n$	$n$	Time (s)
$10^{-2}$	$9.999885 \times 10^{-3}$	78256	2.510100
$10^{-3}$	$9.999951 \times 10^{-4}$	180996	5.883531
$10^{-4}$	$9.999973 \times 10^{-5}$	448854	14.459758

TABLE 1. Table of numerical results for the iterative method (3.2)

$\varepsilon$	$\sigma_n$	$n$	Time (s)
$10^{-2}$	$9.998124 \times 10^{-3}$	7639	0.284367
$10^{-3}$	$9.999802 \times 10^{-4}$	48086	1.781236
$10^{-4}$	$9.999975 \times 10^{-5}$	335068	11.933642

TABLE 2. Table of numerical results for the iterative method (3.12)

The behavior of the function  $\sigma_n$  in Table 1 and Table 2 is presented in the figure below.

FIGURE 1. The behavior of  $\sigma_n$  with  $\varepsilon = 10^{-2}$ 

**Example 5.2.** Suppose that  $H = L^2[0, 1]$  with the inner product  $\langle x, y \rangle = \int_0^1 x(t)y(t)dt$  for all  $x, y \in L^2[0, 1]$  and the norm  $\|x\| = \left( \int_0^1 x^2(t)dt \right)^{1/2}$  for all  $x \in L^2[0, 1]$ . Consider the split feasibility problem with multiple output sets with the following data:

$$C = \{x \in L^2[0, 1] : \langle a, x \rangle \leq b\},$$

$$Q_i = \{y \in L^2[0, 1] : \langle a_i, y \rangle \leq b_i\}, \quad i = 1, 2, \dots, 100,$$

where  $a(t) = t^2$ ,  $b = 0.5$ ,  $a_i(t) = \cos(it) + t$ , and  $b_i = 1/i$ , for all  $i = 1, 2, \dots, 100$ . For each  $i = 1, 2, \dots, 100$ , let  $T_i : L^2[0, 1] \rightarrow L^2[0, 1]$ , be linear operator which is defined by  $T_i x = ix$  for all  $x \in L^2[0, 1]$ .

It is easy to see that the solution set  $\Omega^{SFPMOS} \neq \emptyset$  because  $u(t) = 0$  belongs to  $\Omega^{SFPMOS}$ .

We now test the convergence of the iterative method (3.2) with the initial point  $x_0(t) = \exp(t)$  and the parameter  $\varepsilon_n$  is chosen as in Example 5.1. In this case, we use the condition  $\sigma_n :=$

$\|u_{n+1} - u_n\| < \varepsilon$  to stop the iterative process, where  $\varepsilon$  is a given error. Moreover, at  $n$ th iteration step, we also define the number  $D_n$ , which is determined by

$$D_n = \max \left\{ \langle a, u_n \rangle - b, \max_{i=1,2,\dots,100} \{ \langle a_i, T_i u_n \rangle - b_i \} \right\}.$$

Note that if  $D_n \leq 0$ , then  $u_n \in \Omega$ . We obtain the following table of numerical results.

$t_n$	$\varepsilon$	$\sigma_n$	$n$	$D_n$	Time (s)
$t_n = 1/n^{0.25}$					
	$10^{-4}$	$9.998781 \times 10^{-5}$	240	0.646375	0.225530
	$10^{-5}$	$9.999578 \times 10^{-6}$	7680	0.194301	5.278357
	$10^{-6}$	$9.999913 \times 10^{-7}$	29673	0.021703	19.577540
$t_n = 1/n^{0.85}$					
	$10^{-4}$	$9.991735 \times 10^{-5}$	239	0.646896	0.222085
	$10^{-5}$	$9.999562 \times 10^{-6}$	7674	0.196354	5.248964
	$10^{-6}$	$9.999901 \times 10^{-7}$	29542	0.023546	19.285241

TABLE 3. Table of numerical results for the iterative method (3.2) for Example 5.2

The behavior of the function  $\sigma_n$  in Table 3 is presented in the figure below.

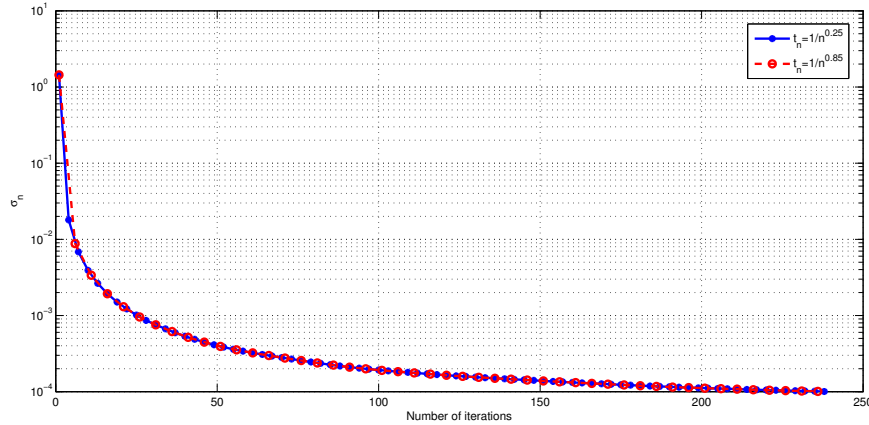


FIGURE 2. The behavior of  $\sigma_n$  with  $\varepsilon = 10^{-4}$

### Acknowledgements

The author is grateful to the reviewers for useful suggestions which improved the contents of this paper. The Science and Technology Fund of Thai Nguyen University of Sciences supports this work.

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