



## Three Step Algorithm for Weighted Resolvent Average of a Finite Family of Monotone Operators

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**Abstract.** In this paper, we introduce a composite iterative method for finding a common zero point of weighted resolvent average of a finite family of monotone operators. Furthermore, the strong convergence of the proposed iterative method is established. Finally, our results are illustrated by some numerical examples.

### 1. Introduction

Let  $H$  be a real Hilbert space with the norm  $\|\cdot\|$  and the inner product  $\langle \cdot, \cdot \rangle$ . Let  $A$  be a set-valued mapping with the domain  $\text{dom } A = \{x \in H : A(x) \neq \emptyset\}$  and the range  $\text{ran } A = \{u \in H : \exists x \in \text{dom } A, u \in A(x)\}$ . The graph of  $A$  is the set  $\text{gra } A = \{(x, u) \in H \times H : x \in \text{dom } A, u \in A(x)\}$ . An operator  $A : H \rightrightarrows H$  is said to be *monotone* if

$$\langle x - y, u - v \rangle \geq 0, \quad \forall (x, u), (y, v) \in \text{gra } A.$$

A monotone operator  $A$  is called *maximal monotone* if there exists no monotone operator  $B$  such that  $\text{gra } A$  is a proper subset of  $\text{gra } B$ . The *resolvent* of  $A$  is the mapping  $J_A = (A + \text{Id})^{-1}$ .

Recall [2] that a map  $T : H \rightarrow H$  is called *nonexpansive* if

$$\|Tx - Ty\| \leq \|x - y\|, \quad \forall x, y \in H.$$

A point  $x \in H$  is said to be a *fixed point* of the operator  $T : H \rightarrow H$ , if  $Tx = x$ . The set of all fixed points of  $T$  is denoted by  $\text{Fix}(T)$ , i.e.,

$$\text{Fix}(T) = \{x \in H : Tx = x\}.$$

Let us consider the zero point problem for monotone operator  $A$  on a real Hilbert space  $H$ , i.e., finding a point  $x \in \text{dom } A$  such that  $0 \in A(x)$ . It was first introduced by Martinet [12] in 1970. Rockafellar [16] defined the proximal point algorithm of Martinet by generalizing a sequence  $\{x_n\}$  such that

$$x_{n+1} = J_{s_n A} x_n + e_n, \quad n \in \mathbb{N},$$

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2010 *Mathematics Subject Classification.* Primary 47H05; Secondary 49J40

*Keywords.* Weighted resolvent average, proximal point algorithm, projection algorithm, monotone operators

Received: 07 April 2018; Accepted: 05 October 2018

Communicated by Erdal Karapınar

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for arbitrary point  $x_0 \in H$ , where  $\{e_n\}$  is a sequence of errors and  $\{s_n\} \subseteq (0, \infty)$ . The sequence  $\{x_n\}$  is known to converge weakly to a zero of  $A$ , if  $\liminf_{n \rightarrow \infty} s_n > 0$  and  $\sum_{n=0}^{\infty} \|e_n\| < \infty$ , see [16], but fails in general to converge strongly [6]. Xu [21] investigated a modified version of the initial proximal point algorithm studied by Rockafellar with  $x_0 \in H$  chosen arbitrarily,

$$x_{n+1} = \beta_n x_0 + (1 - \beta_n) J_{s_n A} x_n + e_n, \quad n \in \mathbb{N},$$

where  $\{e_n\}$  is the errors sequence. For  $\{e_n\}$  summable, it was proved that [21]  $\{x_n\}$  is strongly convergent if  $s_n \rightarrow \infty$  and  $\{\beta_n\} \subseteq (0, 1)$  with  $\sum_{n=0}^{\infty} \beta_n = \infty$  and  $\lim_{n \rightarrow \infty} \beta_n = 0$ .

Recently, Marino and Rugiano [9] introduced the following iteration process: for arbitrary chosen  $x_0 \in C$  construct a sequence  $\{x_n\}$  by

$$x_{n+1} = \beta_n f(x_n) + (1 - \beta_n) T(\alpha_n x_n + (1 - \alpha_n) x_{n+1}), \quad n \in \mathbb{N},$$

where  $\alpha_n, \beta_n \in (0, 1)$  and  $f$  is a  $k$ -contraction mapping on  $H$ . They showed that this process converges strongly to the unique fixed point of the contraction  $P_{\text{Fix}(T)}$ .

In 2014, Mongkolkeha, Cho and Kumam [13], defined the following iterative scheme, by  $x_0 \in C$  and

$$\begin{cases} z_n = (1 - \gamma_n)x_n + \gamma_n Ux_n, \\ y_n = (1 - \beta_n)Tx_n + \beta_n Sz_n, \\ x_{n+1} = (1 - \alpha_n)x_n + \alpha_n y_n, \end{cases}$$

where  $\{\alpha_n\}, \{\beta_n\}$  and  $\{\gamma_n\}$  are sequences in  $(0, 1)$ . They show that if  $\liminf(1 - \alpha_n)\alpha_n > 0$ ,  $\liminf(1 - \beta_n)\beta_n > 0$  and  $\sum_{n \in \mathbb{N}} \gamma_n < \infty$  then  $\{x_n\}$  converges weakly to an element of  $\text{Fix}(T) \cap \text{Fix}(S)$ .

In this paper, we introduce a composite iteration of resolvent average for a finite family of monotone operators as follows:

$$\begin{cases} x_1 \in H, \\ z_n = \gamma_n x_n + (1 - \gamma_n) J_{R(A, \lambda)} x_n, \\ y_n = \beta_n x_n + (1 - \beta_n) J_{R(A, \lambda)} z_n, \\ x_{n+1} = \alpha_n \gamma f(x_n) + (\text{Id} - \alpha_n B) y_n + e_n, \end{cases} \tag{1}$$

where  $B$  is a strongly monotone linear bounded self-adjoint operator and  $f$  is a  $k$ -contraction mapping on  $H$ . We prove, under certain appropriate assumptions on sequences  $\{\alpha_n\}, \{\beta_n\} \subseteq (0, 1)$ ,  $\{\gamma_n\} \subseteq [0, 1]$  and  $\{e_n\}$ , that  $\{x_n\}$  converges strongly to a zero point of resolvent average of the family.

## 2. Preliminaries

Let  $K$  be a closed convex subset of  $H$ . Then for every point  $x \in H$ , there exists a unique *nearest point* in  $K$ , denoted by  $P_K(x)$ , such that

$$\|x - P_K(x)\| \leq \|x - u\|, \quad \forall u \in K.$$

The operator  $P_K$  is called the *metric projection* of  $H$  onto  $K$ . It is well known that  $P_K(x)$  is nonexpansive. The metric projection  $P_K(x)$  is characterized by  $P_K(x) \in K$  and

$$\langle u - P_K(x), x - P_K(x) \rangle \leq 0, \quad \forall u \in K.$$

An operator  $T : H \rightarrow H$  is said to be *firmly nonexpansive* if

$$\|Tx - Ty\|^2 + \|(\text{Id} - T)x - (\text{Id} - T)y\|^2 \leq \|x - y\|^2, \quad \forall x, y \in H.$$

A mapping  $f : H \rightarrow H$  is said to be  $k$ -contraction on  $H$  if there exists a constant  $k \in (0, 1)$  such that

$$\|f(x) - f(y)\| \leq k\|x - y\|, \quad \forall x, y \in H.$$

A sequence of points  $\{x_n\}$  in a Hilbert space  $H$  is said to converge weakly to a point  $x$  in  $H$  if

$$\langle x_n, y \rangle \rightarrow \langle x, y \rangle, \quad \forall y \in H;$$

in symbols,  $x_n \rightharpoonup x$ .

The defining property of the adjoint of a bounded operator  $L$  on a Hilbert space, denoted by  $L^*$ , is that

$$\langle x, Ly \rangle = \langle L^*x, y \rangle, \quad \forall x, y \in H.$$

A bounded linear operator  $L : H \rightarrow H$  on a Hilbert space  $H$  is called self-adjoint if  $L^* = L$ .

An operator  $B : H \rightarrow H$  is called strongly monotone with constant  $\bar{\gamma} > 0$  if

$$\langle Bx - By, x - y \rangle \geq \bar{\gamma}\|x - y\|^2, \quad \forall x, y \in H.$$

These basic definitions are also have presented in various parts of the book [2]. Now, we recall some properties of monotone operators.

**Proposition 2.1.** [2, Proposition 23.7] Suppose that  $A : H \rightarrow H$  is a set-valued mapping. Then

- (i) if  $A$  is monotone, then  $J_A$  is single-valued and firmly nonexpansive.
- (ii) if  $A$  is maximal monotone, then  $J_A$  is single-valued and firmly nonexpansive and its domain is all of  $H$ .
- (iii)  $0 \in A(x)$  if and only if  $x \in \text{Fix}(J_A)$ . Since the fixed point set of nonexpansive operators is closed and convex, the projection onto  $Z = A^{-1}(0)$  is well defined whenever  $Z \neq \emptyset$  (see [16]).

We recall (see [1]) the definition of the proximal average and resolvent average. To this end, we assume that  $m \in \mathbb{N}$  and  $I = \{1, 2, \dots, m\}$ . For every  $i \in I$ , let  $A_i : H \rightarrow H$  be a set-valued mapping and let  $\lambda_i > 0$  be such that  $\sum_{i \in I} \lambda_i = 1$ . We set  $A = (A_1, \dots, A_m)$  and  $\lambda = (\lambda_1, \dots, \lambda_m)$ .

**Definition 2.2.** [1, Definition 1.4] The  $\lambda$ -weighted resolvent average of  $A$  is defined by

$$R(A, \lambda) = \left( \sum_{i \in I} \lambda_i (A_i + \text{Id})^{-1} \right)^{-1} - \text{Id}. \tag{2}$$

The equation (2) is equivalent to the following equation (see [1]):

$$J_{R(A, \lambda)} = \sum_{i \in I} \lambda_i J_{A_i}. \tag{3}$$

**Proposition 2.3.** [1, Theorem 2.5] Suppose that for each  $i \in I$ ,  $A_i : H \rightarrow H$  is monotone and  $x \in H$ . If  $\bigcap_{i \in I} A_i^{-1}(\{0\}) \neq \emptyset$ , then

$$(R(A, \lambda))^{-1}(\{0\}) = \bigcap_{i \in I} A_i^{-1}(\{0\}).$$

**Proposition 2.4.** [1, Theorem 2.2] Suppose that for each  $i \in I$ ,  $A_i : H \rightarrow H$  is a set-valued mapping. Then

$$(R(A, \lambda))^{-1} = R(A^{-1}, \lambda).$$

**Lemma 2.5.** [1, Theorem 2.11] Let  $A_i : H \rightarrow H$  be monotone for each  $i \in I$ . Then  $R(A, \lambda)$  is monotone and

$$\text{dom } J_{R(A, \lambda)} = \bigcap_{i \in I} \text{dom } J_{A_i}.$$

### 3. Main Results

In this section, we introduce a composite iteration for a finite family of monotone operators and its convergence analysis is given. First we present some useful lemmas.

**Lemma 3.1.** [11, Lemma 2.5] Assume that  $B$  is a strongly monotone linear bounded self-adjoint operator on Hilbert space  $H$  with coefficient  $\bar{\gamma} > 0$  and  $0 < \rho \leq \|B\|^{-1}$ . Then  $\|\text{Id} - \rho B\| \leq 1 - \rho\bar{\gamma}$ .

**Lemma 3.2.** Suppose that for each  $i \in I, A_i : H \rightarrow H$  is a monotone operator. Then  $(R(A, \lambda))^{-1}(0) = \text{Fix}(J_{R(A, \lambda)})$

**Lemma 3.3.** Let  $\{A_i : H \rightarrow H\}_{i \in I}$  be a finite family of monotone operators with  $(R(A, \lambda))^{-1}(\{0\}) \neq \emptyset$ , where  $\lambda_i > 0$  and  $\sum_{i \in I} \lambda_i = 1$ . Let  $B$  be a strongly monotone linear bounded self-adjoint operator with coefficient  $\bar{\gamma} > 0$ . Assume that  $f$  is a  $k$ -contraction mapping on  $H$  and  $0 < \gamma < \bar{\gamma}/k$ . Let  $\{x_n\}$  be the sequence generated by (1). Assume that the following conditions hold:

- (i)  $e_n \in H$  and  $\sum_{n \in \mathbb{N}} \|e_n\| < \infty$ ,
- (ii)  $\lim_{n \rightarrow \infty} \alpha_n = 0$ .

Then  $\{\|x_n - z\| : n \in \mathbb{N}\}$  is bounded for each  $z \in (R(A, \lambda))^{-1}(\{0\})$ . Consequently,  $\{x_n\}$  and  $\{\|J_{R(A, \lambda)}x_n - x_n\| : n \in \mathbb{N}\}$  are bounded.

*Proof.* By using the Proposition 2.1, Lemma 3.2 and triangle inequality for any  $z \in (R(A, \lambda))^{-1}(\{0\})$ , we have:

$$\begin{aligned} \|z_n - z\| &= \|\gamma_n x_n + (1 - \gamma_n)J_{R(A, \lambda)}x_n - z\| \\ &= \|\gamma_n(x_n - z) + (1 - \gamma_n)(J_{R(A, \lambda)}x_n - z)\| \\ &\leq \gamma_n\|x_n - z\| + (1 - \gamma_n)\|J_{R(A, \lambda)}x_n - J_{R(A, \lambda)}z\| \\ &\leq \gamma_n\|x_n - z\| + (1 - \gamma_n)\|x_n - z\| \\ &\leq \|x_n - z\|. \end{aligned} \tag{4}$$

By our assumption and (4), we obtain:

$$\begin{aligned} \|y_n - z\| &= \|\beta_n x_n + (1 - \beta_n)J_{R(A, \lambda)}z_n - z\| \\ &= \|\beta_n(x_n - z) + (1 - \beta_n)(J_{R(A, \lambda)}z_n - z)\| \\ &\leq \beta_n\|x_n - z\| + (1 - \beta_n)\|J_{R(A, \lambda)}z_n - J_{R(A, \lambda)}z\| \\ &\leq \beta_n\|x_n - z\| + (1 - \beta_n)\|z_n - z\| \\ &\leq \|x_n - z\|. \end{aligned} \tag{5}$$

By the condition (ii), without loss of generality, we can assume that  $\alpha_n < \|B\|^{-1}$  for all  $n \in \mathbb{N}$ . It follows from Lemma 3.1 that  $\|\text{Id} - \alpha_n B\| \leq 1 - \alpha_n \bar{\gamma}$ . Hence, from triangle inequality and (5), we have

$$\begin{aligned} \|x_{n+1} - z\| &= \|\alpha_n \gamma f(x_n) + (\text{Id} - \alpha_n B)y_n + e_n - z\| \\ &= \|\alpha_n \gamma f(x_n) + (\text{Id} - \alpha_n B)y_n + e_n - z + \alpha_n Bz - \alpha_n Bz\| \\ &= \|\alpha_n(\gamma f(x_n) - Bz) + (\text{Id} - \alpha_n B)(y_n - z) + e_n\| \\ &\leq \alpha_n \gamma \|f(x_n) - f(z)\| + \alpha_n \|\gamma f(z) - Bz\| + (1 - \alpha_n \bar{\gamma})\|y_n - z\| + \|e_n\| \\ &\leq \alpha_n k \gamma \|x_n - z\| + \alpha_n \|\gamma f(z) - Bz\| + (1 - \alpha_n \bar{\gamma})\|x_n - z\| + \|e_n\| \\ &\leq (1 - \alpha_n(\bar{\gamma} - k\gamma))\|x_n - z\| + \alpha_n \|\gamma f(z) - Bz\| + \|e_n\| \\ &\leq (1 - \alpha_n(\bar{\gamma} - k\gamma))\|x_n - z\| + \alpha_n(\bar{\gamma} - k\gamma) \frac{\|\gamma f(z) - Bz\|}{\bar{\gamma} - k\gamma} + \|e_n\| \\ &\leq \max \left\{ \|x_n - z\|, \frac{\|\gamma f(z) - Bz\|}{\bar{\gamma} - k\gamma} \right\} + \|e_n\|. \end{aligned}$$

This shows by induction that

$$\|x_{n+1} - z\| \leq \max \left\{ \|x_1 - z\|, \frac{\|\gamma f(z) - Bz\|}{\bar{\gamma} - k\gamma} \right\} + \sum_{i=1}^n \|e_i\|.$$

Therefore,  $\{\|x_n - z\| : n \in \mathbb{N}\}$  is bounded for each  $z \in (R(A, \lambda))^{-1}(\{0\})$ . Hence  $\{x_n\}$  is bounded.

Finally, it follows from nonexpansivity of resolvent of  $R(A, \lambda)$  that

$$\|J_{R(A, \lambda)}x_n - x_n\| \leq \|J_{R(A, \lambda)}x_n - z\| + \|x_n - z\| \leq 2\|x_n - z\|.$$

Therefore,  $\{\|J_{R(A, \lambda)}x_n - x_n\| : n \in \mathbb{N}\}$  is bounded.  $\square$

**Lemma 3.4.** [18, Lemma 2.1] *Let  $\{x_n\}$  and  $\{y_n\}$  be bounded sequences in Banach space  $X$  and let  $\{\beta_n\}$  be a sequence in  $(0, 1]$  with  $0 < \liminf_{n \rightarrow \infty} \beta_n \leq \limsup_{n \rightarrow \infty} \beta_n < 1$ . Suppose that  $x_{n+1} = (1 - \beta_n)y_n + \beta_nx_n$  for all integers  $n \geq 0$  and*

$$\limsup_{n \rightarrow \infty} (\|y_{n+1} - y_n\| - \|x_{n+1} - x_n\|) \leq 0.$$

Then  $\lim_{n \rightarrow \infty} \|y_n - x_n\| = 0$ .

**Lemma 3.5.** *Let  $\{A_i : H \rightarrow H\}_{i \in I}$  be a finite family of monotone operators with  $(R(A, \lambda))^{-1}(\{0\}) \neq \emptyset$ , where  $\lambda_i > 0$  and  $\sum_{i \in I} \lambda_i = 1$ . Let  $B$  be a strongly monotone linear bounded self-adjoint operator with coefficient  $\bar{\gamma} > 0$ . Assume that  $f$  is a  $k$ -contraction mapping on  $H$  and  $0 < \gamma < \bar{\gamma}/k$ . Let  $\{x_n\}$  be the sequence generated by (1). Assume that the following conditions hold for all  $n \in \mathbb{N}$ :*

- (i)  $e_n \in H$  and  $\sum_{n \in \mathbb{N}} \|e_n\| < \infty$ ,
- (ii)  $\lim_{n \rightarrow \infty} \alpha_n = 0$ ,
- (iii)  $0 < \liminf_{n \rightarrow \infty} \beta_n \leq \limsup_{n \rightarrow \infty} \beta_n < 1$ ,
- (iv)  $\lim_{n \rightarrow \infty} |\gamma_{n+1} - \gamma_n| = 0$ ,
- (v)  $\gamma_n - \beta_n > \epsilon$ , for some  $\epsilon \in (0, 1)$ .

Then  $\lim_{n \rightarrow \infty} \|x_n - J_{R(A, \lambda)}x_n\| = 0$ .

*Proof.* It follows from Lemma 3.3 that  $\{x_n\}$  and  $\{\|J_{R(A, \lambda)}x_n - x_n\| : n \in \mathbb{N}\}$  are bounded. First, we claim that

$$\|x_{n+1} - x_n\| \rightarrow 0. \tag{6}$$

We observe from (1) that

$$\begin{cases} z_{n+1} = \gamma_{n+1}x_{n+1} + (1 - \gamma_{n+1})J_{R(A, \lambda)}x_{n+1}, \\ z_n = \gamma_nx_n + (1 - \gamma_n)J_{R(A, \lambda)}x_n, \end{cases}$$

Then

$$z_{n+1} - z_n = (1 - \gamma_{n+1})(J_{R(A, \lambda)}x_{n+1} - J_{R(A, \lambda)}x_n) + \gamma_{n+1}(x_{n+1} - x_n) + (\gamma_n - \gamma_{n+1})(J_{R(A, \lambda)}x_n - x_n).$$

We obtain

$$\begin{aligned} \|z_{n+1} - z_n\| &\leq (1 - \gamma_{n+1})\|J_{R(A, \lambda)}x_{n+1} - J_{R(A, \lambda)}x_n\| + \gamma_{n+1}\|x_{n+1} - x_n\| \\ &\quad + |\gamma_n - \gamma_{n+1}|\|J_{R(A, \lambda)}x_n - x_n\| \\ &\leq \|x_{n+1} - x_n\| + |\gamma_n - \gamma_{n+1}|M, \end{aligned} \tag{7}$$

where  $M := \sup\{\|J_{R(A,\lambda)}x_n - x_n\| : n \in \mathbb{N}\}$ . Put

$$h_n = \frac{x_{n+1} - \beta_n x_n}{1 - \beta_n}.$$

Then

$$x_{n+1} = (1 - \beta_n)h_n + \beta_n x_n, \quad n \in \mathbb{N}. \tag{8}$$

By using our assumption and (8), we have

$$\begin{aligned} h_{n+1} - h_n &= \frac{x_{n+2} - \beta_{n+1}x_{n+1}}{1 - \beta_{n+1}} - \frac{x_{n+1} - \beta_n x_n}{1 - \beta_n} \\ &= \frac{\alpha_{n+1}\gamma f(x_{n+1}) + (\text{Id} - \alpha_{n+1}B)y_{n+1} + e_{n+1} - \beta_{n+1}x_{n+1}}{1 - \beta_{n+1}} \\ &\quad - \frac{\alpha_n \gamma f(x_n) + (\text{Id} - \alpha_n B)y_n + e_n - \beta_n x_n}{1 - \beta_n} \\ &= \frac{\alpha_{n+1}(\gamma f(x_{n+1}) - By_{n+1})}{1 - \beta_{n+1}} - \frac{\alpha_n(\gamma f(x_n) - By_n)}{1 - \beta_n} \\ &\quad + \frac{y_{n+1} - \beta_{n+1}x_{n+1}}{1 - \beta_{n+1}} - \frac{y_n - \beta_n x_n}{1 - \beta_n} + \frac{e_{n+1}}{1 - \beta_{n+1}} - \frac{e_n}{1 - \beta_n} \\ &= \frac{\alpha_{n+1}(\gamma f(x_{n+1}) - By_{n+1})}{1 - \beta_{n+1}} - \frac{\alpha_n(\gamma f(x_n) - By_n)}{1 - \beta_n} \\ &\quad + J_{R(A,\lambda)}z_{n+1} - J_{R(A,\lambda)}z_n + \frac{e_{n+1}}{1 - \beta_{n+1}} - \frac{e_n}{1 - \beta_n} \end{aligned}$$

Hence,

$$\begin{aligned} \|h_{n+1} - h_n\| &\leq \frac{\alpha_{n+1}}{1 - \beta_{n+1}} \|\gamma f(x_{n+1}) - By_{n+1}\| + \frac{\alpha_n}{1 - \beta_n} \|\gamma f(x_n) - By_n\| \\ &\quad + \|z_{n+1} - z_n\| + \frac{\|e_{n+1}\|}{1 - \beta_{n+1}} + \frac{\|e_n\|}{1 - \beta_n}. \end{aligned} \tag{9}$$

Now, substitute (7) into (9) yields:

$$\begin{aligned} \|h_{n+1} - h_n\| &\leq \frac{\alpha_{n+1}}{1 - \beta_{n+1}} \|\gamma f(x_{n+1}) - By_{n+1}\| + \frac{\alpha_n}{1 - \beta_n} \|\gamma f(x_n) - By_n\| \\ &\quad + \|x_{n+1} - x_n\| + |\gamma_{n+1} - \gamma_n| M + \frac{\|e_{n+1}\|}{1 - \beta_{n+1}} + \frac{\|e_n\|}{1 - \beta_n}. \end{aligned}$$

Then

$$\begin{aligned} \|h_{n+1} - h_n\| - \|x_{n+1} - x_n\| &\leq \frac{\alpha_{n+1}}{1 - \beta_{n+1}} \|\gamma f(x_{n+1}) - By_{n+1}\| + \frac{\alpha_n}{1 - \beta_n} \|\gamma f(x_n) - By_n\| \\ &\quad + |\gamma_{n+1} - \gamma_n| M + \frac{\|e_{n+1}\|}{1 - \beta_{n+1}} + \frac{\|e_n\|}{1 - \beta_n}. \end{aligned}$$

By conditions (i)-(iv), we get:

$$\limsup_{n \rightarrow \infty} (\|h_{n+1} - h_n\| - \|x_{n+1} - x_n\|) \leq 0.$$

It follows from Lemma 3.4 that

$$\lim_{n \rightarrow \infty} \|h_n - x_n\| = 0. \tag{10}$$

From (8), we have:

$$x_{n+1} - x_n = (1 - \beta_n)(h_n - x_n),$$

so (10) yields that  $\lim_{n \rightarrow \infty} \|x_{n+1} - x_n\| = 0$ , i.e., (6) holds.

By assumption, we have  $x_{n+1} - y_n = \alpha_n(\gamma f(x_n) - B y_n) + e_n$ . Therefore

$$\lim_{n \rightarrow \infty} (x_{n+1} - y_n) = 0. \tag{11}$$

Observing

$$\|y_n - x_n\| \leq \|x_n - x_{n+1}\| + \|x_{n+1} - y_n\|,$$

using (6) and (11), we get  $\lim_{n \rightarrow \infty} \|y_n - x_n\| = 0$ .

On the other hand, by assumption and nonexpansivity of resolvent of  $R(A, \lambda)$ , we have

$$\begin{aligned} \|J_{R(A,\lambda)}x_n - x_n\| &\leq \|x_n - y_n\| + \|y_n - J_{R(A,\lambda)}x_n\| \\ &\leq \|x_n - y_n\| + \beta_n \|x_n - J_{R(A,\lambda)}x_n\| + \|J_{R(A,\lambda)}x_n - J_{R(A,\lambda)}z_n\| \\ &\leq \|x_n - y_n\| + \beta_n \|x_n - J_{R(A,\lambda)}x_n\| + \|x_n - z_n\| \\ &\leq \|x_n - y_n\| + \beta_n \|x_n - J_{R(A,\lambda)}x_n\| + (1 - \gamma_n) \|x_n - J_{R(A,\lambda)}x_n\|, \end{aligned}$$

which implies  $(\gamma_n - \beta_n) \|J_{R(A,\lambda)}x_n - x_n\| \leq \|x_n - y_n\|$ .

So, by condition (v), we obtain  $\lim_{n \rightarrow \infty} \|J_{R(A,\lambda)}x_n - x_n\| = 0$ .  $\square$

**Lemma 3.6.** [14] *There holds the following inequality:*

$$\|x + y\|^2 \leq \|x\|^2 + 2 \langle y, x + y \rangle, \quad \forall x, y \in H.$$

**Lemma 3.7.** [5, Lemma 2.2 ] *For each  $x_i \in H$ ,  $a_i \in [0, 1]$ ,  $i = 1, 2$  with  $\sum_{i=1}^2 a_i = 1$ , we have  $\|a_1x_1 + a_2x_2\|^2 \leq a_1\|x_1\|^2 + a_2\|x_2\|^2$ .*

**Lemma 3.8.** *Let  $x \in H$  and  $\{\alpha_n\}$  be a sequence in  $H$  such that  $\|\alpha_n\| \rightarrow 0$ . Then there exists a constant  $L > 0$  such that  $\|x + \alpha_n\|^2 \leq \|x\|^2 + L\|\alpha_n\|$ .*

*Proof.* By Cauchy-Schwarz inequality and for  $L \geq 2\|x\| + \sup_{n \in \mathbb{N}} \|\alpha_n\|$ , we have:

$$\begin{aligned} \|x + \alpha_n\|^2 &= \|x\|^2 + 2 \langle x, \alpha_n \rangle + \|\alpha_n\|^2 \\ &\leq \|x\|^2 + 2\|x\|\|\alpha_n\| + \|\alpha_n\|^2 \\ &\leq \|x\|^2 + \|\alpha_n\|(2\|x\| + \|\alpha_n\|) \\ &\leq \|x\|^2 + L\|\alpha_n\|. \end{aligned}$$

We are done.  $\square$

**Lemma 3.9.** [21, Lemma 2.5 ] *Assume that  $\{a_n\}$  is a sequence of nonnegative real numbers such that*

$$a_{n+1} \leq (1 - \gamma_n)a_n + \gamma_n\delta_n + \beta_n, \quad n \geq 0,$$

where  $\{\gamma_n\}, \{\beta_n\}$  and  $\{\delta_n\}$  satisfy the conditions:

- (i)  $\gamma_n \subseteq [0, 1], \sum_{n=1}^{\infty} \gamma_n = \infty,$
- (ii)  $\limsup_{n \rightarrow \infty} \delta_n \leq 0$  or  $\sum_{n=1}^{\infty} |\gamma_n \delta_n| < \infty,$
- (iii)  $\beta_n \geq 0$  for all  $n \geq 0$  with  $\sum_{n=0}^{\infty} \beta_n < \infty,$

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

**Theorem 3.10.** Let  $\{A_i : H \rightarrow H\}_{i \in I}$  be a finite family of monotone operators with  $Z = (R(A, \lambda))^{-1}(\{0\}) \neq \emptyset,$  where  $\lambda_i > 0$  and  $\sum_{i \in I} \lambda_i = 1.$  Let  $B$  be a strongly monotone linear bounded self-adjoint operator with coefficient  $\bar{\gamma} = \|B\| > 0.$  Assume that  $f$  is a  $k$ -contraction mapping on  $H$  and  $0 < \gamma < \bar{\gamma}/k.$  Let  $\{x_n\}$  be the sequence generated by (1). Assume that the following conditions hold for all  $n \in \mathbb{N}:$

- (i)  $e_n \in H$  and  $\sum_{n \in \mathbb{N}} \|e_n\| < \infty,$
- (ii)  $\lim_{n \rightarrow \infty} \alpha_n = 0$  and  $\sum_{n \in \mathbb{N}} \alpha_n = \infty,$
- (iii)  $0 < \liminf_{n \rightarrow \infty} \beta_n \leq \limsup_{n \rightarrow \infty} \beta_n < 1,$
- (iv)  $\lim_{n \rightarrow \infty} |\gamma_{n+1} - \gamma_n| = 0,$
- (v)  $\gamma_n - \beta_n > \epsilon,$  for some  $\epsilon \in (0, 1).$

Then  $\{x_n\}$  converges strongly to  $z = P_Z(\gamma f + (\text{Id} - B))(z).$

*Proof.* First, we show that there exists a unique  $z \in Z$  such that  $z = P_Z(\gamma f + (\text{Id} - B))(z).$  Since,  $(R(A, \lambda))^{-1}(\{0\})$  is nonempty, closed and convex, the projection  $P_Z$  is well defined. Since  $P_Z$  is nonexpansive and  $f$  is  $k$ -contraction, for each  $x, y \in H,$  we get:

$$\begin{aligned} \|P_Z(\gamma f + (\text{Id} - B))(x) - P_Z(\gamma f + (\text{Id} - B))(y)\| &\leq \|(\gamma f + (\text{Id} - B))(x) - (\gamma f + (\text{Id} - B))(y)\| \\ &\leq \|\gamma f(x) - \gamma f(y)\| + \|\text{Id} - B\| \|x - y\| \\ &\leq \gamma k \|x - y\| + (1 - \bar{\gamma}) \|x - y\| \\ &\leq (1 - (\bar{\gamma} - \gamma k)) \|x - y\|. \end{aligned}$$

Banach’s Contraction Principle guaranties that  $P_Z(\gamma f + (\text{Id} - B))$  has a unique fixed point. That is, there exists a unique element  $z \in Z$  such that  $z = P_Z(\gamma f + (\text{Id} - B))(z).$  Now, consider the mapping  $x \rightarrow t\gamma f(x) + (\text{Id} - tB)J_{R(A, \lambda)}x.$

For each  $t \in (0, 1),$  let  $\phi_t$  on  $H$  be defined by

$$\phi_t(x) = t\gamma f(x) + (\text{Id} - tB)J_{R(A, \lambda)}x.$$

For every  $x, y \in H$  and  $t \in (0, 1),$  we have:

$$\begin{aligned} \|\phi_t(x) - \phi_t(y)\| &= \|(t\gamma f(x) + (\text{Id} - tB)J_{R(A, \lambda)}x) - (t\gamma f(y) + (\text{Id} - tB)J_{R(A, \lambda)}y)\| \\ &\leq t\gamma \|f(x) - f(y)\| + \|\text{Id} - tB\| \|J_{R(A, \lambda)}x - J_{R(A, \lambda)}y\| \\ &\leq t\gamma k \|x - y\| + (1 - t\bar{\gamma}) \|x - y\| \\ &\leq (1 - t(\bar{\gamma} - \gamma k)) \|x - y\|. \end{aligned}$$

Then  $\phi_t$  is contraction. Next, we show that

$$\limsup_{n \rightarrow \infty} \langle \gamma f(z) - Bz, x_n - z \rangle \leq 0, \tag{12}$$

where  $z = \lim_{t \rightarrow 0} x_t$  with  $x_t$  being fixed point of the contraction  $x \mapsto t\gamma f(x) + (\text{Id} - tB)J_{R(A, \lambda)}x.$  Since  $x_t$  solves the fixed point equation,

$$x_t = t\gamma f(x_t) + (\text{Id} - tB)J_{R(A, \lambda)}x_t$$



By using Lemma 3.6 and Lemma 3.1, we obtain:

$$\begin{aligned} \|x_t - x_n\|^2 &= \|(\text{Id} - tB)(J_{R(A,\lambda)}x_t - x_n) + t(\gamma f(x_t) - Bx_n)\|^2 \\ &\leq (1 - \bar{\gamma}t)^2 \|J_{R(A,\lambda)}x_t - x_n\|^2 + 2t \langle \gamma f(x_t) - Bx_n, x_t - x_n \rangle \\ &\leq (1 - \bar{\gamma}t)^2 (\|J_{R(A,\lambda)}x_t - J_{R(A,\lambda)}x_n\|^2 + \|J_{R(A,\lambda)}x_n - x_n\|^2 \\ &\quad + 2 \langle J_{R(A,\lambda)}x_t - J_{R(A,\lambda)}x_n, J_{R(A,\lambda)}x_n - x_n \rangle) + 2t \langle \gamma f(x_t) - Bx_n, x_t - x_n \rangle \\ &\leq (1 - \bar{\gamma}t)^2 (\|x_t - x_n\|^2 + \|J_{R(A,\lambda)}x_n - x_n\|^2 + 2\|J_{R(A,\lambda)}x_n - x_n\| \|x_n - x_t\|) \\ &\quad + 2t \langle \gamma f(x_t) - Bx_n, x_t - x_n \rangle + 2t \langle Bx_t - Bx_n, x_t - x_n \rangle \\ &\leq (1 - \bar{\gamma}t)^2 (\|x_t - x_n\|^2 + \|J_{R(A,\lambda)}x_n - x_n\|^2 + 2\|J_{R(A,\lambda)}x_n - x_n\| \|x_n - x_t\|) \\ &\quad + 2t \langle \gamma f(x_t) - Bx_t, x_t - x_n \rangle + 2t \|B\| \|x_n - x_t\|^2. \end{aligned}$$

Therefore,

$$\langle Bx_t - \gamma f(x_t), x_t - x_n \rangle \leq \frac{\bar{\gamma}^2 t}{2} \|x_t - x_n\|^2 + \frac{(1 - \bar{\gamma}t)^2}{2t} (\|J_{R(A,\lambda)}x_n - x_n\|^2 + 2\|J_{R(A,\lambda)}x_n - x_n\| \|x_n - x_t\|).$$

By letting  $n \rightarrow \infty$ , we have:

$$\limsup_{n \rightarrow \infty} \langle Bx_t - \gamma f(x_t), x_t - x_n \rangle \leq \limsup_{n \rightarrow \infty} \frac{\bar{\gamma}^2 t}{2} \|x_t - x_n\|^2.$$

Now, taking  $t \rightarrow 0$ , we obtain (12). Now, from assumption, Lemma 3.6 and Lemma 3.8, for some appropriate constant  $L > 0$ , we have:

$$\begin{aligned} \|x_{n+1} - z\|^2 &= \|\alpha_n \gamma f(x_n) + (\text{Id} - \alpha_n B)y_n + e_n - z\|^2 \\ &= \|\alpha_n \gamma f(x_n) + (\text{Id} - \alpha_n B)y_n + \alpha_n Bz - \alpha_n Bz + e_n - z\|^2 \\ &= \|(\text{Id} - \alpha_n B)(y_n - z) + \alpha_n (\gamma f(x_n) - Bz) + e_n\|^2 \\ &\leq \|(\text{Id} - \alpha_n B)(y_n - z) + e_n\|^2 + 2\alpha_n \langle \gamma f(x_n) - Bz, x_{n+1} - z \rangle \\ &\leq \|(\text{Id} - \alpha_n B)(y_n - z)\|^2 + 2\alpha_n \langle \gamma f(x_n) - Bz, x_{n+1} - z \rangle + L\|e_n\| \\ &\leq (1 - \alpha_n \bar{\gamma})^2 \|y_n - z\|^2 + 2\alpha_n \langle \gamma f(x_n) - Bz, x_{n+1} - z \rangle + L\|e_n\| \\ &= (1 - \alpha_n \bar{\gamma})^2 \|y_n - z\|^2 + 2\alpha_n \gamma \langle f(x_n) - f(z), x_{n+1} - z \rangle \\ &\quad + 2\alpha_n \langle \gamma f(z) - Bz, x_{n+1} - z \rangle + L\|e_n\| \\ &\leq (1 - \alpha_n \bar{\gamma})^2 \|x_n - z\|^2 + 2\alpha_n \gamma k \|x_n - z\| \|x_{n+1} - z\| \\ &\quad + 2\alpha_n \langle \gamma f(z) - Bz, x_{n+1} - z \rangle + L\|e_n\| \\ &\leq (1 - \alpha_n \bar{\gamma})^2 \|x_n - z\|^2 + \alpha_n \gamma k (\|x_n - z\|^2 + \|x_{n+1} - z\|^2) \\ &\quad + 2\alpha_n \langle \gamma f(z) - Bz, x_{n+1} - z \rangle + L\|e_n\|, \end{aligned}$$

This implies that

$$\|x_{n+1} - z\|^2 \leq \frac{(1 - \alpha_n \bar{\gamma})^2 + \alpha_n \gamma k}{1 - \alpha_n \gamma k} \|x_n - z\|^2 + \frac{2\alpha_n}{1 - \alpha_n \gamma k} \langle \gamma f(z) - Bz, x_{n+1} - z \rangle + \frac{L}{1 - \alpha_n \gamma k} \|e_n\|$$

$$\begin{aligned}
 &= \frac{(1 - 2\alpha_n\bar{\gamma} + \alpha_n\gamma k)}{1 - \alpha_n\gamma k} \|x_n - z\|^2 + \frac{\alpha_n^2\bar{\gamma}^2}{1 - \alpha_n\gamma k} \|x_n - z\|^2 \\
 &\quad + \frac{2\alpha_n}{1 - \alpha_n\gamma k} \langle \gamma f(z) - Bz, x_{n+1} - z \rangle + \frac{L}{1 - \alpha_n\gamma k} \|e_n\| \\
 &\leq \left(1 - \frac{2\alpha_n(\bar{\gamma} - k\gamma)}{1 - \alpha_n\gamma k}\right) \|x_n - z\|^2 \\
 &\quad + \frac{2\alpha_n(\bar{\gamma} - k\gamma)}{1 - \alpha_n\gamma k} \left(\frac{1}{\bar{\gamma} - \gamma k} \langle \gamma f(z) - Bz, x_{n+1} - z \rangle + \frac{\alpha_n\bar{\gamma}^2}{2(\bar{\gamma} - \gamma k)} M\right) + \frac{L}{1 - \alpha_n\gamma k} \|e_n\| \\
 &\leq (1 - \delta_n) \|x_n - z\|^2 + \delta_n t_n + \eta_n,
 \end{aligned}$$

where  $M = \sup\{\|x_n - z\|^2 : n \in \mathbb{N}\}$ ,  $\delta_n = \frac{2\alpha_n(\bar{\gamma} - k\gamma)}{1 - \alpha_n\gamma k}$ ,  $\eta_n = \frac{L}{1 - \alpha_n\gamma k} \|e_n\|$  and  $t_n = \frac{1}{\bar{\gamma} - \gamma k} \langle \gamma f(z) - Bz, x_{n+1} - z \rangle + \frac{\alpha_n\bar{\gamma}^2}{2(\bar{\gamma} - \gamma k)} M$ . By assumption,  $\lim_{n \rightarrow \infty} \delta_n = 0$ ,  $\sum_{n \in \mathbb{N}} \delta_n = \infty$ ,  $\limsup_{n \rightarrow \infty} t_n \leq 0$  and  $\sum_{n \in \mathbb{N}} \eta_n < \infty$ . Hence, applying Lemma 3.9, we immediately deduce that  $x_n \rightarrow z$  where  $z = P_Z(\gamma f + (\text{Id} - B))(z)$ .  $\square$

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**Algorithm 1** Iterative algorithms for resolvent average

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**Input:**  $x_1 \in H$ ,  $\{\alpha_n\}, \{\beta_n\} \subset (0, 1)$ ,  $\{\gamma_n\} \subset [0, 1]$ ,  $\{\lambda_i\}_{1 \leq i \leq m} \subset (0, 1)$ ,  $\{e_n\} \in H$ ,  $\mathbf{A} = (A_1, \dots, A_m)$

**Output:**  $x_n$

for  $i = 1$  to  $m$  do

$$J_{A_i}(x_n) := (A_i + \text{Id})^{-1}(x_n);$$

end for

Set  $J_{R(\mathbf{A}, \lambda)}(x_n) := \sum_{i=1}^m \lambda_i J_{A_i}(x_n);$

for  $n = 1$  to  $\dots$  do

$$z_n = \gamma_n x_n + (1 - \gamma_n) J_{R(\mathbf{A}, \lambda)} x_n;$$

$$y_n = \beta_n x_n + (1 - \beta_n) J_{R(\mathbf{A}, \lambda)} z_n;$$

$$x_{n+1} = \alpha_n \gamma f(x_n) + (\text{Id} - \alpha_n B) y_n + e_n;$$

end for

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**4. Numerical Examples**

In this section, we evaluated strongly convergence of three step algorithm for weighted resolvent average of a finite family of monotone operators.

**Example 4.1.** Let  $A_1(x) = 2x - 1$ ,  $A_2(x) = x$  and  $A_3(x) = x + 2$ . Set  $\mathbf{A} = (2x - 1, x, x + 2)$ ,  $f(x) = \frac{x}{2}$  and for every  $1 \leq i \leq 3$ ,  $\lambda_i = \frac{1}{3}$ . Assume that  $e_n = \left\{\frac{1}{n^2}\right\}$  is the sequence of errors and let  $\alpha_n = \left\{\frac{1}{n^2}\right\}$ ,  $\beta_n = \left\{\frac{1}{n+3} + \frac{1}{10}\right\}$  and  $\gamma_n = \left\{\frac{1}{n+5} + \frac{4}{5}\right\}$ . Let  $B = \text{Id}$  and  $\gamma = 1$ . First note that  $A_1^{-1}(x) = \frac{1}{2}(x + 1)$ ,  $A_2^{-1}(x) = x$  and  $A_3^{-1}(x) = x - 2$ . So,  $\mathbf{A}^{-1} = \left(\frac{1}{2}(x + 1), x, x - 2\right)$ . Then by easy calculation, we get:

$$J_{A_1^{-1}}(x_n) = (A_1^{-1} + \text{Id})^{-1}(x_n) = \left\{\frac{1}{3}(2x_n - 1)\right\}, J_{A_2^{-1}}(x_n) = \left\{\frac{1}{2}x_n\right\}, \text{ and } J_{A_3^{-1}}(x_n) = \left\{\frac{1}{2}(x_n + 2)\right\}. \tag{13}$$

By using Proposition 2.4 and (13), we obtain:

$$\begin{aligned} (R(A, \lambda))^{-1}(\{0\}) &= (R(A^{-1}, \lambda))(\{0\}) = \left( \left( \sum_{i=1}^m \lambda_i J_{A_i^{-1}} \right)^{-1} - \text{Id} \right)(\{0\}) \\ &= \left( \left( \frac{1}{3}(A_1^{-1} + \text{Id})^{-1} + \frac{1}{3}(A_2^{-1} + \text{Id})^{-1} + \left( \frac{1}{3}(A_3^{-1} + \text{Id})^{-1} \right)^{-1} - \text{Id} \right)(\{0\}) \right) \\ &= \left\{ x \in \mathbb{R} : 0 \in \left( \frac{1}{3}(A_1^{-1} + \text{Id})^{-1}(x) + \frac{1}{3}(A_2^{-1} + \text{Id})^{-1}(x) + \frac{1}{3}(A_3^{-1} + \text{Id})^{-1}(x) \right) \right\} \\ &= \left\{ x \in \mathbb{R} : 0 \in \left( \frac{x}{6} + \frac{x+2}{6} + \frac{2x-1}{9} \right) \right\} = \{-0.4\}. \end{aligned}$$

Therefore,  $Z = (R(A, \lambda))^{-1}(\{0\}) = \{-0.4\}$ . Hence, we have:

$$P_Z(f(z)) = P_{\{-0.4\}}(f(-0.4)) = P_{\{-0.4\}}(-0.2) = -0.4$$

Let  $\{x_n\}$  be the sequence generated by (1) for starting point  $x_1 \in \mathbb{R}$ . Clearly,

$$J_{A_1}(x_n) = (A_1 + \text{Id})^{-1}(x_n) = \{y \in \mathbb{R} : x_n \in (A_1 + \text{Id})(y)\} = \left\{ \frac{1}{3}(x_n + 1) \right\}. \tag{14}$$

and similarly,

$$J_{A_2}(x_n) = \left\{ \frac{1}{2}x_n \right\}, \quad J_{A_3}(x_n) = \left\{ \frac{1}{2}(x_n - 2) \right\}. \tag{15}$$

Substituting (14) and (15) into (3), we obtain:

$$J_{R(A, \lambda)}x_n = \sum_{i=1}^m \lambda_i J_{A_i}x_n = \left\{ \frac{1}{9}(x_n + 1) + \frac{1}{6}x_n + \frac{1}{6}(x_n - 2) \right\} = \left\{ \frac{1}{9}(4x_n - 2) \right\}.$$

Therefore,

$$\begin{cases} z_n = \left( \frac{1}{n+5} + \frac{4}{5} \right)x_n + 1/9 \left( \frac{1}{5} - \frac{1}{n+5} \right)(4x_n - 2), \\ y_n = \left( \frac{1}{n+3} + \frac{1}{10} \right)x_n + 1/9 \left( \frac{9}{10} - \frac{1}{n+3} \right)(4z_n - 2), \\ x_{n+1} = \frac{1}{2n^2}x_n + \left( 1 - \frac{1}{n^2} \right)y_n + \frac{1}{n^n}. \end{cases}$$

It follows from Theorem 3.10 that  $x_n$  converges, say to  $x$ . Since  $x_n$  is convergent, by letting  $n \rightarrow \infty$  in the above equalities we obtain:

$$\begin{cases} z = \frac{4}{5}x + \frac{1}{45}(4x - 2), \\ y = \frac{1}{10}x + \frac{1}{10}(4z - 2), \\ x = y. \end{cases}$$

Then,  $x = -0.4$ . The numerical results with starting point  $x_1 = 0$ , which are shown in Table 1, shows that  $x_n \rightarrow -0.4$ .

Table 1: Results for given starting point  $x_1 = 0$  in Example 4.1

$n$	1	10	20	50	100	200	500	1000	...
$x_n$	0	-0.376042	-0.398733	-0.399836	-0.399961	-0.399990	-0.399998	-0.399999	...

**Example 4.2.** Let  $A = (x^3 - 1, x - 1, (x + 1)^3)$ ,  $f(x) = \frac{4x}{5}$  and  $\lambda_i = \frac{1}{3}$  for every  $1 \leq i \leq 3$ . Let  $\{e_n\}$ ,  $\{\alpha_n\}$ ,  $\{\beta_n\}$ ,  $\{\gamma_n\}$ ,  $B$

and  $\gamma$  be the same as in Example 4.1. We have  $A^{-1} = ((1+x)^{\frac{1}{3}}, 1+x, 1+x^{\frac{1}{3}})$ . Then

$$J_{A_1^{-1}}(x_n) = \left\{ x_n + \frac{\left(\frac{2}{3}\right)^{\frac{1}{3}}}{h_1(x_n)} - \frac{h_1(x_n)}{(2)^{\frac{1}{3}}(3)^{\frac{2}{3}}} \right\}, J_{A_2^{-1}}(x_n) = \left\{ \frac{1}{2}(x_n - 1) \right\}, J_{A_3^{-1}}(x_n) = \left\{ x_n - \frac{\left(\frac{2}{3}\right)^{\frac{1}{3}}}{h_2(x_n)} + \frac{h_2(x_n)}{(2)^{\frac{1}{3}}(3)^{\frac{2}{3}}} - 1 \right\}, \tag{16}$$

where  $h_1(x_n) = (9 + 9x_n - \sqrt{3} \sqrt{31 + 54x_n + 27x_n^2})^{\frac{1}{3}}$  and  $h_2(x_n) = (9 - 9x_n + \sqrt{3} \sqrt{31 - 54x_n + 27x_n^2})^{\frac{1}{3}}$ . By using Proposition 2.4 and (16), we obtain  $(R(A, \lambda))^{-1}(\{0\}) = \{1\}$ . Therefore,  $Z = (R(A, \lambda))^{-1}(\{0\}) = \{1\}$ . Hence, we have:

$$P_Z(f(z)) = P_{\{1\}}(f(1)) = P_{\{1\}}\left(\frac{4}{5}\right) = 1.$$

Let  $\{x_n\}$  be the sequence generated by (1) with starting point  $x_1 \in \mathbb{R}$ . We have:

$$J_{A_1}(x_n) = \left\{ \frac{h_1(x_n)}{(2)^{\frac{1}{3}}(3)^{\frac{2}{3}}} - \frac{\left(\frac{2}{3}\right)^{\frac{1}{3}}}{h_1(x_n)} \right\}, J_{A_2}(x_n) = \left\{ \frac{1}{2}(1 + x_n) \right\}, J_{A_3}(x_n) = \left\{ \frac{\left(\frac{2}{3}\right)^{\frac{1}{3}}}{h_2(x_n)} - \frac{h_2(x_n)}{(2)^{\frac{1}{3}}(3)^{\frac{2}{3}}} + 1 \right\}. \tag{17}$$

Substituting (17) into (3), we obtain:

$$J_{R(A,\lambda)}x_n = \sum_{i=1}^m \lambda_i J_{A_i}x_n = \frac{1}{3} \left( \frac{3}{2} + \frac{1}{2}x_n + \frac{\left(\frac{2}{3}\right)^{\frac{1}{3}}}{h_2(x_n)} - \frac{h_2(x_n)}{(2)^{\frac{1}{3}}(3)^{\frac{2}{3}}} + \frac{h_1(x_n)}{(2)^{\frac{1}{3}}(3)^{\frac{2}{3}}} - \frac{\left(\frac{2}{3}\right)^{\frac{1}{3}}}{h_1(x_n)} \right). \tag{18}$$

Therefore,

$$\begin{cases} z_n = \left(\frac{1}{n+5} + \frac{4}{5}\right)x_n + \left(\frac{1}{5} - \frac{1}{n+5}\right)J_{R(A,\lambda)}x_n, \\ y_n = \left(\frac{1}{n+3} + \frac{1}{10}\right)x_n + \left(\frac{9}{10} - \frac{1}{n+3}\right)J_{R(A,\lambda)}z_n, \\ x_{n+1} = \frac{4}{5n^2}x_n + \left(1 - \frac{1}{n^2}\right)y_n + \frac{1}{n^n}. \end{cases} \tag{19}$$

The numerical results which are shown in Table 2 shows that  $x_n \rightarrow 1$ .

Table 2: Results for given starting point  $x_1 = 0$  in Example 4.2

$n$	1	10	20	50	100	200	500	1000	...
$x_n$	0	1.001696	0.998254	0.999780	0.999948	0.999987	0.999998	0.999999	...

### References

[1] S. Bartz, H. H. Bauschke, S. M. Moffat, and X. Wang, *The resolvent average of monotone operators: dominant and recessive properties*, SIAM J. Optimiz. **26**(2016), 602–634.  
 [2] H. H. Bauschke and P. L. Combettes, *Convex Analysis and Monotone Operator Theory in Hilbert Spaces*, Springer, 2011.  
 [3] H. H. Bauschke, R. Goebel, Y. Lucet and X. Wang, *The proximal average: basic theory*, SIAM J. Optimiz. **19**(2008), 766–785.  
 [4] H. Brezis, *Operateurs Maximaux Monotones et Semi-Groups de Contractions dans les Espaces de Hilbert*, North-Holland, Amsterdam, 1973.  
 [5] M. Eslamian, *Rockafellar’s proximal point algorithm for a finite family of monotone operators*, Sci. Bull. **76**(2014), 43–50.  
 [6] O. Guler, *On the convergence of proximal point algorithm for convex minimization*, SIAM J. Control Optim. **29**(1991), 403–419.  
 [7] B. Halpern, *Fixed points of nonexpanding maps*, Bull. Amer. Math. Soc. **73** (1967), 957–961.  
 [8] P. E. Mainge, *Strong convergence of projected subgradient methods for nonsmooth and nonstrictly convex minimization*, Set-Valued Anal. **16**(2008), 899–912.  
 [9] G. Marino and A. Rugiano, *Strong convergence of a generalized viscosity implicit midpoint rule for nonexpansive mappings and equilibrium problems*, J. Nonlinear. Convex. Anal. **17**(2016), 2255–2275.  
 [10] G. Marino B. Scardamaglia and E. Karapinar, *Strong convergence theorem for strict pseudo-contractions in Hilbert spaces*, J. Inequal. Appl. (2016) 2016:134.  
 [11] G. Marino and H. K. Xu, *A general iterative method for nonexpansive mappings in Hilbert spaces*, J. Math. Anal. Appl. **318**(2006), 43–52.  
 [12] B. Martinet, *Regularisation d’inequations variationnelles par approximations successives*, Recherche Operationnelle. **4**(1970), 154–158.

- [13] C. Mongkolkeha, Y.J. Cho and P. Kumam, *Weak convergence theorems of iterative sequences in Hilbert spaces*, *J. Nonlinear. Convex. Anal.* **15**(2014), 1303–1318.
- [14] A. Moudafi, *Viscosity approximation methods for fixed-points problems*, *J. Math. Anal. Appl.* **241**(2000), 46–55.
- [15] C.I. Podilchuk and R.J. Mammone, *Image recovery by convex projections using a least-squares constraint*, *J. Optical.* **7**(1990), 517–521.
- [16] R.T. Rockafellar, *Monotone operators and the proximal point algorithm*, *SIAM J. Control Optim.* **14**(1976), 877–898.
- [17] T. Suzuki, *Strong convergence of Krasnoselskii and Mann's type sequence for one parameter nonexpansive semigroup without Bochner integrals*, *J. Math. Anal. Appl.* **305**(2005), 227–239.
- [18] T. Suzuki, *Strong convergence theorems for infinite families of nonexpansive mappings in general Banach spaces*, *Fixed Point Theory Appl.* **2005**(2005), 21 pages.
- [19] C. A. Tian and Y. Song, *Strong convergence of a regularization method for Rockafellar's proximal point algorithm*, *J. Global Optim.* **55**(4)(2013), 831–837.
- [20] H.K. Xu, *A regularization method for the proximal point algorithm*, *J. Global Optim.* **36**(2006), 115–125.
- [21] H.K. Xu, *Iterative algorithms for nonlinear operators*, *J. London Math. Soc.* **66**(2002), 240–256.
- [22] H.K. Xu, *Viscosity approximation methods for nonexpansive mappings*, *J. Math. Anal. Appl.* **298**(2004), 279–291.