

## GENERAL NONCONVEX SPLIT VARIATIONAL INEQUALITY PROBLEMS

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ABSTRACT. In this paper, we established a *general nonconvex split variational inequality problem*, this is, an extension of *general convex split variational inequality problems* in two different Hilbert spaces. By using the concepts of prox-regularity, we proved the convergence of the iterative schemes for the *general nonconvex split variational inequality problems*. Further, we also discussed the iterative method for the *general convex split variational inequality problems*.

### 1. Introduction

Let  $\mathcal{H}_1$  and  $\mathcal{H}_2$  be two real Hilbert spaces with inner product and norm denoted by  $\langle \cdot, \cdot \rangle$  and  $\| \cdot \|$ , respectively. Let  $\mathcal{C}$  and  $\mathcal{Q}$  be nonempty closed convex subsets of  $\mathcal{H}_1$  and  $\mathcal{H}_2$ , respectively. For  $i \in \{1, 2\}$ , let  $f_i : \mathcal{H}_i \rightarrow \mathcal{H}_i$  and  $g_i : \mathcal{H}_i \rightarrow \mathcal{H}_i$  be nonlinear mappings and  $A : \mathcal{H}_1 \rightarrow \mathcal{H}_2$  be a bounded linear operator with its adjoint operator  $A^*$ . Consider a problem for finding  $x^* \in \mathcal{H}_1$  such that  $g_1(x^*) \in \mathcal{C}$  and

$$(1.1) \quad \langle f_1(x^*), x - g_1(x^*) \rangle \geq 0, \quad \forall x \in \mathcal{C},$$

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and  $y^* = Ax^* \in \mathcal{H}_2$  such that  $g_2(y^*) \in \mathcal{Q}$  solves

$$(1.2) \quad \langle f_2(y^*), y - g_2(y^*) \rangle \geq 0, \quad \forall y \in \mathcal{Q}.$$

The problem (1.1)-(1.2) are called *general convex split variational inequality problems*. The split convex variational inequality problem is introduced and studied by Censor *et al.* [6–8]. It is worth mentioning that split convex variational inequality problem is quite general and permits split minimization between two spaces, so the image of a minimizer of a given function, under a bounded linear operator, is a minimizer of another function.

The *general convex split variational inequality problems* (1.1)-(1.2), to take into account of non convexity of subsets  $\mathcal{C}$  and  $\mathcal{Q}$ . This new non-convex problem is called *general nonconvex split variational inequality problems*.

Poliquin and Rockafellor [17] and Clarke *et al.* [10] have introduced and studied a class of nonconvex sets which are called uniformly prox-regular sets. This class of uniformly prox-regular sets has played an important role in many nonconvex applications such as optimization, dynamic systems and differential inclusions.

Inspired by the recent works going in this fields [1–5,9,12,13,16,21,22], we established the *general nonconvex split variational inequality problems*. By using the concepts of prox-regularity, we proved the convergence of an iterative schemes for the *general nonconvex split variational inequality problems*. Further we also discussed the convergence of an iterative schemes for the *general convex split variational inequality problems*.

Let  $\mathcal{C}$  be a nonempty closed subsets of a Hilbert space  $\mathcal{H}$ , not necessarily convex. Then we have the following:

DEFINITION 1.1. The proximal normal cone of  $\mathcal{C}$  at a point  $x \in \mathcal{H}$  is given by

$$N_{\mathcal{C}}^P(x) = \{\zeta \in \mathcal{H} : x \in P_{\mathcal{C}}(x + \alpha\zeta)\},$$

where  $\alpha > 0$  is a constant and  $P_{\mathcal{C}}$  is projection of operator of  $\mathcal{H}$  onto  $\mathcal{C}$ , that is

$$P_{\mathcal{C}}(x) = \{x^* \in \mathcal{C} : d_{\mathcal{C}}(x) = \|x - x^*\| \},$$

where  $d_{\mathcal{C}}(x)$  or  $d(\cdot, \mathcal{C})$  is the usual distance function to the subset of  $\mathcal{C}$ , that is

$$d_{\mathcal{C}}(x) = \inf_{\hat{x} \in \mathcal{C}} \|\hat{x} - x\|.$$

LEMMA 1.2. [10] *Let  $\mathcal{C}$  be a nonempty closed subset in  $\mathcal{H}$ . Then  $\zeta \in N_{\mathcal{C}}^P(x)$  if and only if there exists a constant  $\alpha = \alpha(\zeta, x) > 0$  such that*

$$\langle \zeta, \hat{x} - x \rangle \leq \alpha \|\hat{x} - x\|^2, \quad \forall \hat{x} \in \mathcal{C}.$$

DEFINITION 1.3. [19] *The Clarke normal cone denoted by*

$$N_{\mathcal{C}}^{cl}(x) = \bar{co}[N_{\mathcal{C}}^P(x)],$$

where  $\bar{co}A$  means the closure of the convex hull of  $A$ .

LEMMA 1.4. [10] *Let  $\mathcal{C}$  be a nonempty closed convex subset in  $\mathcal{H}$ . Then  $\zeta \in N_{\mathcal{C}}^P(x)$  if and only if*

$$\langle \zeta, \hat{x} - x \rangle \leq 0, \quad \forall \hat{x} \in \mathcal{C}.$$

DEFINITION 1.5. *For any  $r \in (0, +\infty]$ , a subset  $\mathcal{C}_r$  of  $\mathcal{H}$  is called the normalized uniformly prox-regular (or uniformly  $r$ -prox-regular) if and only if every nonzero proximal normal to  $\mathcal{C}_r$  can be realized by an  $r$ -ball that is, for all  $x \in \mathcal{C}_r$  and  $0 \neq \zeta \in N_{\mathcal{C}_r}^P(x)$  with  $\|\zeta\| = 1$ , one has*

$$\left\langle \frac{\zeta}{\|\zeta\|}, \hat{x} - x \right\rangle \leq \frac{1}{2r} \|\hat{x} - x\|^2, \quad \forall \hat{x} \in \mathcal{C}_r.$$

It is known that if  $\mathcal{C}_r$  is a uniformly  $r$ -prox-regular set, the proximal normal cone  $N_{\mathcal{C}_r}^P(x)$  is closed as a set valued mapping. Thus, we have  $N_{\mathcal{C}_r}^{cl}(x) = N_{\mathcal{C}_r}^P(x)$ . We make the conversion  $\frac{1}{r} = 0$  for  $r \rightarrow +\infty$ . If  $r = +\infty$  then uniformly  $r$ -prox-regularity of  $\mathcal{C}_r$  reduces to its convexity, see, [11, 14, 20].

LEMMA 1.6. [11] *A closed set  $\mathcal{C} \subseteq \mathcal{H}$  is convex if and only if it is proximally smooth of radius  $r$  for every  $r > 0$ .*

PROPOSITION 1.7. [18] *For each  $r > 0$  and let  $\mathcal{C}_r$  be a nonempty closed and uniformly  $r$ -prox-regular subset of  $\mathcal{H}$ . Set*

$$\mathcal{U}(r) = \{x \in \mathcal{H} : 0 \leq d_{\mathcal{C}_r}(x) < r\}.$$

*Then the following statements are hold:*

- (a) *for all  $x \in \mathcal{U}(r)$ ,  $P_{\mathcal{C}_r}(x) \neq \emptyset$ ;*
- (b) *for all  $r' \in (0, r)$ ,  $P_{\mathcal{C}_r}$  is Lipschitz continuous mapping with constant  $\frac{r}{r-r'}$  on*

$$\mathcal{U}(r') = \{x \in \mathcal{H} : 0 \leq d_{\mathcal{C}_r}(x) < r'\};$$

- (c) *the proximal normal cone is closed as a set valued mapping.*

LEMMA 1.8. [15]

- (i)  $\|x + y\|^2 \leq \|x\|^2 + 2\langle y, x + y \rangle, \forall x, y \in \mathcal{H}$   
(ii)  $\|(1-t)x + ty\|^2 = (1-t)\|x\|^2 + t\|y\|^2 - (1-t)t\|x - y\|^2, \forall x, y \in \mathcal{H}$   
and for any fixed  $t \in [0, 1]$ .

## 2. General nonconvex split variational inequality problems

Throughout this paper, we assume that for given  $r, s \in (0, +\infty)$ ,  $\mathcal{C}_r, \mathcal{Q}_s$  are uniformly prox regular subsets of  $\mathcal{H}_1$  and  $\mathcal{H}_2$ , respectively. For each  $i = \{1, 2\}$ , let  $f_i : \mathcal{H}_i \rightarrow \mathcal{H}_i$  and  $g_i : \mathcal{H}_i \rightarrow \mathcal{H}_i$  be nonlinear mappings and  $A : \mathcal{H}_1 \rightarrow \mathcal{H}_2$  be a bounded linear operator with its adjoint operator  $A^*$ . The *general nonconvex split variational inequality problems* is formulated as follows: find  $x^* \in \mathcal{H}_1, g_1(x^*) \in \mathcal{C}_r$  such that

$$(2.1) \quad \langle f_1(x^*), x - g_1(x^*) \rangle + \left( \frac{\|f_1(x^*)\|}{2r} \right) \|x - g_1(x^*)\|^2 \geq 0, \forall x \in \mathcal{C}_r,$$

and  $y^* = Ax^* \in \mathcal{H}_2$  such that  $g_2(y^*) \in \mathcal{Q}_s$  solves

$$(2.2) \quad \langle f_2(y^*), y - g_2(y^*) \rangle + \left( \frac{\|f_2(y^*)\|}{2s} \right) \|y - g_2(y^*)\|^2 \geq 0, \forall y \in \mathcal{Q}_s.$$

By making use of Definition 1.5 and Lemma 1.2, the *general nonconvex split variational inequality problems* can be reformulated as follows: finding  $(x^*, y^*) \in \mathcal{C}_r \times \mathcal{Q}_s$  with  $y^* = Ax^*, g_1(x^*) \in \mathcal{C}_r, g_2(y^*) \in \mathcal{Q}_s$  such that

$$(2.3) \quad \begin{aligned} 0 &\in \rho f_1(x^*) + N_{\mathcal{C}_r}^P(g_1(x^*)), \\ 0 &\in \lambda f_2(y^*) + N_{\mathcal{Q}_s}^P(g_2(y^*)) \end{aligned}$$

where  $\rho$  and  $\lambda$  are parameters with positive values and 0 denotes the zero vector of  $\mathcal{H}_1$  and  $\mathcal{H}_2$ , respectively. Since  $P_{\mathcal{C}_r} = (I + N_{\mathcal{C}_r}^P)^{-1}$  and  $P_{\mathcal{Q}_s} = (I + N_{\mathcal{Q}_s}^P)^{-1}$  are equivalent to finding  $(x^*, y^*) \in \mathcal{C}_r \times \mathcal{Q}_s$  with  $y^* = Ax^*$  such that  $g_1(x^*) \in \mathcal{C}_r, g_2(y^*) \in \mathcal{Q}_s$  such that

$$(2.4) \quad \begin{aligned} g_1(x^*) &= P_{\mathcal{C}_r}(g_1(x^*) - \rho f_1(x^*)), \\ g_2(y^*) &= P_{\mathcal{Q}_s}(g_2(y^*) - \lambda f_2(y^*)) \end{aligned}$$

where  $0 < \rho < \frac{r}{1 + \|f_1(x^*)\|}, 0 < \lambda < \frac{s}{1 + \|f_2(y^*)\|}$  and  $P_{\mathcal{C}_r}$  and  $P_{\mathcal{Q}_s}$  are projection onto  $\mathcal{C}_r$  and  $\mathcal{Q}_s$ , respectively.

We note that, for  $r, s \rightarrow +\infty$  we have  $\mathcal{C}_r = \mathcal{C}$  and  $\mathcal{Q}_s = \mathcal{Q}$ , the closed convex subsets of  $\mathcal{H}_1$  and  $\mathcal{H}_2$ , respectively, then *general nonconvex split variational inequality problems* (2.1)-(2.2) reduces to the *general convex*

split variational inequality problems (1.1)-(1.2) for finding  $(x^*, y^*) \in \mathcal{C} \times \mathcal{Q}$  with  $y^* = Ax^*$  such that

$$(2.5) \quad \begin{aligned} g_1(x^*) &= P_{\mathcal{C}}(g_1(x^*) - \rho f_1(x^*)), \\ g_2(y^*) &= P_{\mathcal{Q}}(g_2(y^*) - \lambda f_2(y^*)), \end{aligned}$$

where  $P_{\mathcal{C}}$  and  $P_{\mathcal{Q}}$  are projection onto  $\mathcal{C}$  and  $\mathcal{Q}$ , respectively.

DEFINITION 2.1. Let  $f : \mathcal{H} \rightarrow \mathcal{H}$  be a mapping. Then  $f$  is said to be:

(i) monotone if

$$\langle f(x) - f(\hat{x}), x - \hat{x} \rangle \geq 0, \quad \forall x, \hat{x} \in \mathcal{H},$$

(ii)  $\alpha$ -strongly monotone if there exists a constant  $\alpha > 0$  such that

$$\langle f(x) - f(\hat{x}), x - \hat{x} \rangle \geq \alpha \|x - \hat{x}\|^2, \quad \forall x, \hat{x} \in \mathcal{H},$$

(iii)  $\xi$ -inverse strongly monotone if there exists a constant  $\xi > 0$  such that

$$\langle f(x) - f(\hat{x}), x - \hat{x} \rangle \geq \xi \|f(x) - f(\hat{x})\|^2, \quad \forall x, \hat{x} \in \mathcal{H},$$

(iii) relaxed  $(\kappa, v)$ -cocoercive mapping if there exist constants  $\kappa, v > 0$  such that

$$\langle f(x) - f(\hat{x}), x - \hat{x} \rangle \geq -\kappa \|f(x) - f(\hat{x})\|^2 + v \|x - \hat{x}\|^2, \quad \forall x, \hat{x} \in \mathcal{H},$$

(iv)  $\beta$ -Lipschitz continuous if there exists a constant  $\beta > 0$  such that

$$\|f(x) - f(\hat{x})\| \leq \beta \|x - \hat{x}\|, \quad \forall x, \hat{x} \in \mathcal{H}.$$

REMARK 2.2. Every  $\lambda$ -inverse strongly monotone mapping  $f$  is monotone and  $\frac{1}{\lambda}$ -Lipschitz continuous.

Based on above arguments, we suggest the following iterative algorithm for approximating a solution to (2.1)-(2.2).

ALGORITHM 2.3. Given  $x_0 \in \mathcal{C}_r$ , compute the iterative sequence  $\{x_n\}$  defined by the iterative schemes:

$$(2.6) \quad g_1(y_n) = P_{\mathcal{C}_r}[g_1(x_n) - \rho f_1(x_n)],$$

$$(2.7) \quad g_2(z_n) = P_{\mathcal{Q}_s}[g_2(Ay_n) - \lambda f_2(Ay_n)],$$

$$(2.8) \quad x_{n+1} = P_{\mathcal{C}_r}[y_n + \gamma A^*(z_n - Ay_n)]$$

for all  $n = 0, 1, 2, \dots, 0 < \rho < \frac{r}{1+\|f_1(x_n)\|}, 0 < \lambda < \frac{s}{1+\|f_2(Ay_n)\|}$  and  $0 < \gamma < \frac{r}{1+\|A^*(z_n - Ay_n)\|}$ .

As a particular case of Algorithm 2.3, we have the following algorithm for approximating a solution to (1.1)-(1.2).

ALGORITHM 2.4. Given  $x_0 \in \mathcal{C}$ , compute the iterative sequence  $\{x_n\}$  defined by the iterative schemes:

$$(2.9) \quad g_1(y_n) = P_{\mathcal{C}}[g_1(x_n) - \rho f_1(x_n)],$$

$$(2.10) \quad g_2(z_n) = P_{\mathcal{Q}}[g_2(Ay_n) - \lambda f_2(Ay_n)],$$

$$(2.11) \quad x_{n+1} = P_{\mathcal{C}}[y_n + \gamma A^*(z_n - Ay_n)]$$

for all  $n = 0, 1, 2, \dots$ ,  $\rho, \lambda, \gamma > 0$ .

Let  $\{\alpha_n\} \subseteq (0, 1)$  be a sequence such that  $\sum_{n=1}^{\infty} \alpha_n = +\infty$  and  $\rho, \lambda, \gamma$  are parameters with positive values. Then we have the following algorithm for approximating a solution to (1.1)-(1.2).

ALGORITHM 2.5. Given  $x_0 \in \mathcal{H}_1$ , compute the iterative sequence  $\{x_n\}$  defined by the iterative schemes:

$$(2.12) \quad g_1(y_n) = P_{\mathcal{C}}[g_1(x_n) - \rho f_1(x_n)],$$

$$(2.13) \quad g_2(z_n) = P_{\mathcal{Q}}[g_2(Ay_n) - \lambda f_2(Ay_n)],$$

$$(2.14) \quad x_{n+1} = (1 - \alpha_n)x_n + \alpha_n[y_n + \gamma A^*(z_n - Ay_n)]$$

for all  $n = 0, 1, 2, \dots$ ,  $\rho, \lambda, \gamma > 0$ .

We note that Algorithm 2.4 and Algorithm 2.5 are different form.

### 3. Main Results

In this section, we discuss the convergence of the iterative sequence generated by algorithms.

THEOREM 3.1. For given  $r, s \in (0, +\infty)$ , we assume that  $r' \in (0, r)$ ,  $s' \in (0, s)$  and denote  $\delta = \frac{r}{r-r'}$  and  $\eta = \frac{s}{s-s'}$ . Let  $\mathcal{C}_r$  and  $\mathcal{Q}_s$  be uniformly prox regular subsets of  $\mathcal{H}_1$  and  $\mathcal{H}_2$ , respectively. For each  $i \in \{1, 2\}$ , let  $f_i : \mathcal{H}_i \rightarrow \mathcal{H}_i$  be the relaxed  $(\kappa_i, v_i)$ -cocoercive mapping with constants  $\kappa_i, v_i > 0$  and  $\beta_i$ -Lipschitz continuous with constant  $\beta_i > 0$ . Let  $g_i : \mathcal{H}_i \rightarrow \mathcal{H}_i$  be the  $\xi_i$ -inverse strongly monotone with constant  $\xi_i > 0$  and  $\sigma_i$ -Lipschitz continuous and let  $(g_i - I_i)$  be the  $\zeta_i$ -strongly monotone with constant  $\zeta_i > 0$ , where  $I_i$  is the identity operator on  $\mathcal{H}_i$  ( $i = \{1, 2\}$ ). Let  $A : \mathcal{H}_1 \rightarrow \mathcal{H}_2$  be the bounded linear operator such

that  $A(\mathcal{C}_r) \subseteq \mathcal{Q}_s$  and  $A^*$  be its adjoint operator. Suppose that  $x^* \in \mathcal{C}_r$  is a solution of general nonconvex split variational inequality problems (2.1)-(2.2). Then the iterative sequence  $\{x_n\}$  generated by Algorithm 2.3 converges strongly to  $x^*$  provided that the constants  $\rho, \lambda$  and  $\gamma$  satisfy the following conditions:

$$\begin{aligned} & \frac{v_1 - \kappa_1\beta_1^2}{\beta_1^2} - \Omega < \rho < \min \left\{ \frac{v_1 - \kappa_1\beta_1^2}{\beta_1^2} + \Omega, \frac{r'}{1 + \|f_1(x_n)\|}, \frac{r'}{1 + \|f_1(x^*)\|} \right\}, \\ 0 < \lambda < \min \left\{ \frac{s'}{1 + \|f_2(Ay_n)\|}, \frac{s'}{1 + \|f_2(Ay^*)\|} \right\}, \text{ for some } r' \in (0, r), s' \in (0, s), \\ 0 < \gamma < \min \left\{ \frac{2}{\|A\|^2}, \frac{r'}{1 + \|A^*(z_n - Ay_n)\|} \right\}, \Omega = \frac{1}{\beta_1^2} (\sqrt{(v_1 - \kappa_1\beta_1^2)^2 - \beta_1^2(1 - \varrho_1^2)}) \\ & \text{with } v_1 > \kappa_1\beta_1^2 + \beta_1\sqrt{1 - \varrho_1^2}, \theta_1 < [\delta(1 + 2\theta_2)]^{-1} = d, \\ & \varrho_1 = \frac{d\sqrt{2\zeta_1 + 1}}{\delta} - \ell_1, \ell_1 = \sqrt{1 - 2\zeta_1\sigma_1^2 + \sigma_1^2}, \theta = \delta\theta_1(1 + 2\theta_2) < 1, \end{aligned}$$

$$(3.1) \quad \theta_2 = \sqrt{1 - 2(v_2 - \kappa_2\beta_2^2)\lambda + \beta_2^2\lambda^2} + \ell_2, \ell_2 = \sqrt{1 - 2\zeta_2\sigma_2^2 + \sigma_2^2}.$$

*Proof.* Since  $x^* \in \mathcal{C}_r$  is a solution of general nonconvex split variational inequality problems (2.1)-(2.2) and the parameters  $\rho, \lambda, \gamma$  satisfying the conditions (3.1), then we have

$$(3.2) \quad g_1(x^*) = P_{\mathcal{C}_r}[g_1(x^*) - \rho f_1(x^*)],$$

$$(3.3) \quad g_2(Ax^*) = P_{\mathcal{Q}_s}[g_2(Ax^*) - \lambda f_2(Ax^*)].$$

From Lemma 1.8(i) and since  $(g_1 - I_1)$  is  $\zeta_1$ -strongly monotone, then we have

$$\begin{aligned} \|y_n - x^*\|^2 & \leq \|g_1(y_n) - g_1(x^*)\|^2 - 2\langle (g_1 - I_1)y_n - (g_1 - I_1)x^*, y_n - x^* \rangle \\ & \leq \|g_1(y_n) - g_1(x^*)\|^2 - 2\zeta_1\|y_n - x^*\|^2 \end{aligned}$$

which implies that

$$(3.4) \quad \|y_n - x^*\| \leq \frac{1}{\sqrt{2\zeta_1 + 1}} \|g_1(y_n) - g_1(x^*)\|.$$

From (2.6) and conditions (3.1) on  $\rho$ , we have

$$\begin{aligned} & \|g_1(y_n) - g_1(x^*)\| = \|P_{\mathcal{C}_r}[g_1(x_n) - \rho f_1(x_n)] - P_{\mathcal{C}_r}[g_1(x^*) - \rho f_1(x^*)]\| \\ (3.5) \quad & \leq \delta[\|x_n - x^* - (g_1(x_n) - g_1(x^*))\| + \|x_n - x^* - \rho(f_1(x_n) - f_1(x^*))\|]. \end{aligned}$$

Since  $g_1$  is  $\xi_1$ -inverse strongly monotone with constant  $\xi_1 > 0$  and  $\sigma_1$ -Lipschitz continuous with constant  $\sigma_1 > 0$ , we have

$$\begin{aligned} & \|x_n - x^* - (g_1(x_n) - g_1(x^*))\|^2 \\ & \leq \|x_n - x^*\|^2 - 2\langle g_1(x_n) - g_1(x^*), x_n - x^* \rangle + \|g_1(x_n) - g_1(x^*)\|^2 \\ & \leq \|x_n - x^*\|^2 - 2\xi_1\sigma_1^2\|x_n - x^*\|^2 + \sigma_1^2\|x_n - x^*\|^2 \\ & \leq (1 - 2\xi_1\sigma_1^2 + \sigma_1^2)\|x_n - x^*\|^2 \end{aligned}$$

$$(3.6) \quad \Rightarrow \|x_n - x^* - (g_1(x_n) - g_1(x^*))\| \leq \sqrt{1 - 2\xi_1\sigma_1^2 + \sigma_1^2}\|x_n - x^*\|.$$

Again since  $f_1$  is relaxed  $(\kappa_1, \nu_1)$ -cocoercive mapping with constants  $\kappa_1, \nu_1 > 0$  and  $\beta_1$ -Lipschitz continuous with constant  $\beta_1 > 0$ , we have

$$\begin{aligned} & \|x_n - x^* - \rho(f_1(x_n) - f_1(x^*))\|^2 \\ & \leq \|x_n - x^*\|^2 - 2\rho\langle f_1(x_n) - f_1(x^*), x_n - x^* \rangle + \rho^2\|f_1(x_n) - f_1(x^*)\|^2 \\ & \leq \|x_n - x^*\|^2 - 2\rho(-\kappa_1\|f_1(x_n) - f_1(x^*)\|^2 + \nu_1\|x_n - x^*\|^2) + \rho^2\beta_1^2\|x_n - x^*\|^2 \\ & \leq (1 - 2\rho(\nu_1 - \kappa_1\beta_1^2) + \rho^2\beta_1^2)\|x_n - x^*\|^2 \end{aligned}$$

(3.7)

$$\Rightarrow \|x_n - x^* - \rho(f_1(x_n) - f_1(x^*))\| \leq \sqrt{1 - 2\rho(\nu_1 - \kappa_1\beta_1^2) + \rho^2\beta_1^2}\|x_n - x^*\|.$$

From (3.5), (3.6) and (3.7), we have

$$(3.8) \quad \|g_1(y_n) - g_1(x^*)\| \leq (\ell_1 + \sqrt{1 - 2\rho(\nu_1 - \kappa_1\beta_1^2) + \rho^2\beta_1^2})\|x_n - x^*\|,$$

where  $\ell_1 = \sqrt{1 - 2\xi_1\sigma_1^2 + \sigma_1^2}$ . Again from (3.4) and (3.8) we obtain

$$(3.9) \quad \|y_n - x^*\| \leq \theta_1\|x_n - x^*\|$$

where

$$\theta_1 = \frac{\delta}{\sqrt{2\xi_1 + 1}}\{\ell_1 + \sqrt{1 - 2\rho(\nu_1 - \kappa_1\beta_1^2) + \rho^2\beta_1^2}\} \text{ and } \ell_1 = \sqrt{1 - 2\xi_1\sigma_1^2 + \sigma_1^2}.$$

Similarly from (2.7), (3.1), (3.3) and condition on parameter  $\lambda$  and using the fact that  $f_2$  is relaxed  $(\kappa_2, \nu_2)$ -cocoercive and  $\beta_2$ -Lipschitz continuous mapping; and  $g_2$  is  $\xi_2$ -inverse strongly monotone mapping with constant  $\xi_2 > 0$  and  $\sigma_2$ -Lipschitz continuous mapping and from  $(q_2 - I_2)$  is  $\zeta_2$ -strongly monotone and  $A(\mathcal{C}_r) \subseteq \mathcal{Q}_s$ , we have

$$\|g_2(z_n) - g_2(Ax^*)\| = \|P_{\mathcal{Q}_s}[g_2(Ay_n) - \lambda f_2(Ay_n)] - P_{\mathcal{Q}_s}[g_2(Ax^*) - \lambda f_2(Ax^*)]\|$$

$$(3.10) \quad \leq \eta[\|g_2(Ay_n) - g_2(Ax^*) - \lambda(f_2(Ay_n) - f_2(Ax^*))\|]$$

and

$$(3.11) \quad \|z_n - Ax^*\| \leq \theta_2 \|Ay_n - Ax^*\|$$

where

$$\theta_2 = \frac{\eta}{\sqrt{2\zeta_2 + 1}} \{ \ell_2 + \sqrt{1 - 2\lambda(v_2 - \kappa_2\beta_2^2) + \lambda^2\beta_2^2} \} \text{ and } \ell_2 = \sqrt{1 - 2\xi_2\sigma_2^2 + \sigma_2^2}.$$

Next from (2.8) and condition (3.1) on  $\gamma$ , we have

$$(3.12) \quad \begin{aligned} \|x_{n+1} - x^*\| &\leq \|P_{C_r}[y_n + \gamma A^*(z_n - Ay_n)] - P_{C_r}[x^* + \gamma A^*(Ax^* - Ax^*)]\| \\ &\leq \delta[\|y_n - x^* - \gamma A^*(Ay_n - Ax^*)\| + \gamma \|A^*(z_n - Ax^*)\|]. \end{aligned}$$

Further using the definition of  $A^*$ , the fact that  $A^*$  is a bounded operator with  $\|A^*\| = \|A\|$  and condition (3.1), we have

$$(3.13) \quad \begin{aligned} &\|y_n - x^* - \gamma A^*(Ay_n - Ax^*)\|^2 \\ &= \|y_n - x^*\|^2 - 2\gamma \langle A^*(Ay_n - Ax^*), y_n - x^* \rangle + \gamma^2 \|A^*(Ay_n - Ax^*)\|^2 \\ &\leq \|y_n - x^*\|^2 - 2\gamma \|Ay_n - Ax^*\|^2 + \gamma^2 \|A\|^2 \|Ay_n - Ax^*\|^2 \\ &\leq \|y_n - x^*\|^2 - \gamma(2 - \gamma\|A\|^2) \|Ay_n - Ax^*\|^2 \\ &\leq \|y_n - x^*\|^2. \end{aligned}$$

Using (3.11), we have

$$(3.14) \quad \begin{aligned} \|A^*(z_n - Ax^*)\| &\leq \|A\| \|z_n - Ax^*\| \\ &\leq \theta_2 \|A\| \|Ay_n - Ax^*\| \\ &\leq \theta_2 \|A\|^2 \|y_n - x^*\|. \end{aligned}$$

Combining (3.13) and (3.14) with (3.12), we obtain

$$\begin{aligned} \|x_{n+1} - x^*\| &\leq \delta[\|y_n - x^*\| + \gamma\theta_2 \|A\|^2 \|y_n - x^*\|], \\ \|x_{n+1} - x^*\| &\leq \theta \|x_n - x^*\|, \end{aligned}$$

where  $\theta = \delta\theta_1(1 + \gamma\|A\|^2\theta_2)$ .

Thus, we obtain

$$(3.15) \quad \|x_{n+1} - x^*\| \leq \theta_n \|x_0 - x^*\|.$$

Since  $\gamma\|A\|^2 < 2$ , hence the maximum value of  $(1 + \gamma\|A\|^2\theta_2)$  is  $(1 + 2\theta_2)$ . Further  $\theta \in (0, 1)$  if and only if

$$(3.16) \quad \theta_1 < [\delta(1 + 2\theta_2)]^{-1} = d.$$

Since  $d \in (0, 1)$  and  $\delta, \eta > 1$ . Finally the inequality (3.15) holds from conditions (3.1). Thus from (3.15) that  $\{x_n\}$  strongly converges to  $x^*$  as  $n \rightarrow \infty$ . Since  $A$  is continuous, hence from (3.9) that  $y_n \rightarrow x^*, g_1(y_n) \rightarrow g_1(x^*), Ay_n \rightarrow Ax^*, z_n \rightarrow Ax^*$  and  $g_2(z_n) \rightarrow g_2(Ax^*)$  as  $n \rightarrow \infty$ . This completes the proof.  $\square$

In particular case, if  $r = +\infty, s = +\infty$ , one has  $\eta = \delta = 1$ . Then we have the following convergence result for Algorithm 2.5 to solve (1.1)-(1.2).

**THEOREM 3.2.** *Let  $\mathcal{C}$  and  $\mathcal{Q}$  be nonempty closed convex subsets of  $\mathcal{H}_1$  and  $\mathcal{H}_2$ , respectively. For each  $i \in \{1, 2\}$  let  $f_i : \mathcal{H}_i \rightarrow \mathcal{H}_i$  be the relaxed  $(\kappa_i, v_i)$ -cocoercive mapping with constants  $\kappa_i, v_i > 0$  and  $\beta_i$ -Lipschitz continuous with constant  $\beta_i > 0$ . Let  $g_i : \mathcal{H}_i \rightarrow \mathcal{H}_i$  be the  $\xi_i$ -inverse strongly monotone with constant  $\xi_i > 0$  and  $\sigma_i$ -Lipschitz continuous and let  $(g_i - I_i)$  be the  $\zeta_i$ -strongly monotone with constant  $\zeta_i > 0$ , where  $I_i$  is the identity operator on  $\mathcal{H}_i$  ( $i = \{1, 2\}$ ). Let  $A : \mathcal{H}_1 \rightarrow \mathcal{H}_2$  be the bounded linear operator and  $A^*$  be its adjoint operator. Suppose that  $x^* \in \mathcal{C}$  is a solution of general convex split variational inequality problems (1.1)-(1.2). Then the iterative sequence  $\{x_n\}$  generated by Algorithm 2.5 converges strongly to  $x^*$  provided the constants  $\rho, \eta$  and  $\gamma$  satisfy the following conditions:*

$$\begin{aligned} \frac{v_1 - \kappa_1\beta_1^2}{\beta_1^2} - \Omega < \rho < \frac{v_1 - \kappa_1\beta_1^2}{\beta_1^2} + \Omega, \quad \gamma \in \left(0, \frac{2}{\|A\|^2}\right), \quad \text{where} \\ \Omega = \frac{1}{\beta_1^2} \left(\sqrt{(v_1 - \kappa_1\beta_1^2)^2 - \beta_1^2(1 - \varrho_1^2)}\right) \quad \text{with } v_1 > \kappa_1\beta_1^2 + \beta_1\sqrt{1 - \varrho_1^2}, \\ \theta_1 < [\delta(1 + 2\theta_2)]^{-1} = d, \quad \varrho_1 = \frac{d\sqrt{2\zeta_1 + 1}}{\delta} - \ell_1, \\ \ell_1 = \sqrt{1 - 2\xi_1\sigma_1^2 + \sigma_1^2}, \quad \theta = \delta\theta_1(1 + 2\theta_2) < 1, \end{aligned} \tag{3.17}$$

$$\theta_2 = \sqrt{1 - 2(v_2 - \kappa_2\beta_2^2)\lambda + \beta_2^2\lambda^2} + \ell_2, \quad \ell_2 = \sqrt{1 - 2\xi_2\sigma_2^2 + \sigma_2^2}, \quad \lambda > 1.$$

*Proof.* Since  $x^* \in \mathcal{C}$  is a solution of general convex split variational inequality problems (1.1)-(1.2), then  $\rho, \lambda > 0$  such that

$$g_1(x^*) = P_{\mathcal{C}}[g_1(x^*) - \rho f_1(x^*)], \tag{3.18}$$

$$g_2(Ax^*) = P_{\mathcal{Q}}[g_2(Ax^*) - \lambda f_2(Ax^*)]. \tag{3.19}$$

Using the same argument used in proof of Theorem 3.1, we have

$$(3.20) \quad \|y_n - x^*\| \leq \theta_1 \|x_n - x^*\|$$

where

$$\theta_1 = \frac{\delta}{\sqrt{2\zeta_1 + 1}} \{ \ell_1 + \sqrt{1 - 2\rho(v_1 - \kappa_1\beta_1^2) + \rho^2\beta_1^2} \} \text{ and } \ell_1 = \sqrt{1 - 2\xi_1\sigma_1^2 + \sigma_1^2}.$$

Again

$$(3.21) \quad \|z_n - Ax^*\| \leq \theta_2 \|Ay_n - Ax^*\|$$

where

$$\theta_2 = \frac{\eta}{\sqrt{2\zeta_2 + 1}} \{ \ell_2 + \sqrt{1 - 2\lambda(v_2 - \kappa_2\beta_2^2) + \lambda^2\beta_2^2} \} \text{ and } \ell_2 = \sqrt{1 - 2\xi_2\sigma_2^2 + \sigma_2^2}.$$

Next from Algorithm 2.5 (2.14), we have

$$(3.22) \quad \|x_{n+1} - x^*\| \leq (1 - \alpha_n) \|x_n - x^*\| + \alpha_n [\|y_n - x^* - \gamma A^*(Ay_n - Ax^*)\| + \gamma \|A^*(z_n - Ax^*)\|].$$

Since  $A^*$  is a bounded linear operator with  $\|A^*\| = \|A\|$  and given condition on  $\gamma$ , we have

$$(3.23) \quad \begin{aligned} & \|y_n - x^* - \gamma A^*(Ay_n - Ax^*)\|^2 \\ &= \|y_n - x^*\|^2 - 2\gamma \langle A^*(Ay_n - Ax^*), y_n - x^* \rangle + \gamma^2 \|A^*(Ay_n - Ax^*)\|^2 \\ &\leq \|y_n - x^*\|^2. \end{aligned}$$

Using (3.21) we have

$$(3.24) \quad \begin{aligned} \|A^*(z_n - Ax^*)\| &\leq \|A\| \|z_n - Ax^*\| \\ &\leq \theta_2 \|A\| \|Ay_n - Ax^*\| \\ &\leq \theta_2 \|A\|^2 \|y_n - x^*\|. \end{aligned}$$

Combining (3.23) and (3.24) with (3.22), we obtain

$$\|x_{n+1} - x^*\| \leq [1 - \alpha_n(1 - \theta)] \|x_n - x^*\|,$$

where  $\theta = \theta_1(1 + \gamma\|A\|^2\theta_2)$ .

Thus, we obtain

$$(3.25) \quad \|x_{n+1} - x^*\| \leq \prod_{j=1}^n [1 - \alpha_j(1 - \theta)] \|x_0 - x^*\|.$$

It follows from condition on  $\rho, \lambda \in (0, 1)$ . Since  $\sum_{n=1}^{\infty} a_n = +\infty$  and  $\theta \in (0, 1)$  and from [23], we have

$$\lim_{n \rightarrow \infty} \prod_{j=1}^n [1 - \alpha_j(1 - \theta)] = 0.$$

The rest of the proof is same as the proof of Theorem 3.1. This completes the proof.  $\square$

### References

- [1] M. K. Ahmad and Salahuddin, *A stable perturbed algorithms for a new class of generalized nonlinear implicit quasi variational inclusions in Banach spaces*, Adv. Pure Math. **2** (2) (2012), 139–148.
- [2] C. Baiocchi and A. Capelo, *Variational and Quasi Variational Inequalities*, John Wiley and Sons, New York, 1984.
- [3] A. Bensoussan, M. Goursat and J. L. Lions, *Controle impulsinnel et inequations quasi variationnelles stationeries*, C. R. Acad. Sci. **276** (1973), 1279–1284.
- [4] A. Bensoussan and J. L. Lions, *Impulse Controle and Quasi variational Inequalities*, Gauthiers Villers, Paris, 1973.
- [5] C. Byrne, *Iterative oblique projection onto convex subsets and the split feasibility problems*, Inverse Problems **18** (2002), 441–453.
- [6] Y. Censor, A. Gibali and S. Reich, *Algorithms for the split variational inequality problems*, Numer. Algo. **59** (2) (2012), 301–323.
- [7] Y. Censor and T. Elfying, *A multiprojection algorithm using Bregman projection in a product space*, Numer. Algo. **8** (1994), 221–239.
- [8] Y. Censor, A. Motova and A. Segal, *Perturbed projection and subgradient projection for the multi sets split feasibility problems*, J. Math. Anal Appl. **327** (2007), 1244–1256.
- [9] Y. Censor and G. T. Herman, *On some minimization techniques in image reconstruction from projection*, Appl. Numer. Math. **3** (1987), 365–391.
- [10] F. H. Clarke, *Optimization and Nonsmooth Analysis*, Wiley Int. Science, New York, 1983.
- [11] F. H. Clarke, Y. S. Ledyaw, R. J. Stern and P. R. Wolenski, *Nonsmooth Analysis and Control Theory*, Springer-Verlag, New York 1998.
- [12] R. Glowinski, J. L. Lions and R. Tremolieres, *Numerical Analysis of Variational Inequalities*, North-Holland, Amsterdam, 1981.
- [13] J. K. Kim, A. Farajzadeh and Salahuddin, *New systems of extended nonlinear regularized nonconvex set valued variational inequalities*, Commun. Appl. Non-linear Anal. **21** (3) (2014), 21–40.
- [14] B. S. Lee and Salahuddin, *A general system of regularized nonconvex variational inequalities*, Appl. Computat. Math. **3** (4) (2014), dx.doi.org/10.4172/2168-9679.1000169.

- [15] G. Marino and H. K. Xu, *Weak and strong convergence theorems for strict pseudo contractions in Hilbert spaces*, J. Math. Anal. Appl. **329** (2007), 336–346.
- [16] A. Moudafi, *Split monotone variational inclusions*, J. Optim. Theo. Appl. **150** (2011), 275–283.
- [17] R. A. Poliquin and R. T. Rockafellar, *Prox-regular functions in variational analysis*, Trans. Amer. Math. Soc. **348** (1996), 1805–1838.
- [18] R. A. Poliquin, R. T. Rockafellar and L. Thibault, *Local differentiability of distance functions*, Trans. Amer. Math. Soc. **352** (2000), 5231–5249.
- [19] Salahuddin, *Regularized equilibrium problems in Banach spaces*, Korean Math. J. **24** (1) (2016), 51–63.
- [20] Salahuddin, *System of generalized nonlinear regularized nonconvex variational inequalities*, Korean Math. J. **24** (2) (2015), 181–198.
- [21] Salahuddin, *Regularized penalty method for non-stationary set valued equilibrium problems in Banach spaces*, Korean Math. J. **25** (2) (2017), 147–162.
- [22] Salahuddin and R. U. Verma, *Split feasibility quasi variational inequality problems involving cocoercive mappings in Banach spaces*, Commun. Appl. Nonlinear Anal. **22** (4) (2015), 95–101.
- [23] R. Wittmann, *Approximation of fixed points of nonexpansive mappings*, Archiv der Mathematik, **58** (1) (1992), 486–491.

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