

GENERALIZED REGULARIZED GAP FUNCTIONS AND ERROR BOUNDS FOR GENERALIZED VECTOR VARIATIONAL-LIKE INEQUALITIES

MOHD ASIF¹, ZUBAIR KHAN², FEEROZ BABU^{2,*}, ABDUL FAIZ ANSARI³

¹*Department of Mathematics and Statistics, Integral University, Lucknow 226 026, India*

²*Department of Applied Mathematics, Aligarh Muslim University, Aligarh, 202 002, India*

³*Department of Mathematics and Astronomy, University of Lucknow, Lucknow 226 007, India*

Abstract. This paper aims to study the generalized vector variational-like inequalities (in short, GVVLI). A gap function and a generalized regularized gap function which are independent of the scalarization parameter for the GVVLI. A result on error bounds for the GVVLI with generalized regularized gap functions without any Lipschitz continuity is presented. Finally, we give examples to illustrate our results. The results presented in the paper improve and generalize some known results in the literature.

Keywords. Error bounds; Gap functions; Monotone operators; Vector variational inequalities.

1. INTRODUCTION

A vector variational inequality (in short, VVI) was initially originated by Giannessi [1], who developed the concept of VVI in the setting of finite-dimensional spaces. VVI plays a fruitful role in operation research, economics, optimization, etc. Since the last three decades, many generalizations of VVIs have been considered and formulated to solve different types of problems, such as equilibrium problems, boundary value problems, vector optimization problems; see, e.g., [2, 3, 4, 5, 6] and the references therein.

In the study of variational inequalities, gap functions provide more advantages than other methods. Gap functions play an important role to reduce variational inequalities into optimization problems, which are easy to handle from viewpoint of computation. There are several known methods to construct gap functions for VVIs in the literature; see, for example, [7, 8, 9, 10, 11] and the references therein. One of the well-formed methods to establish a gap function for VVI is the scalarization method. Generated gap functions from this method depends on some scalarization parameters. In 2014, Sun and Chai [12] proposed the scalar-valued gap function for generalized vector variational inequalities in the absence of scalarization parameters. On the other hand, Fukushima [13] introduced a regularized gap function for classical variational inequalities. Further, Wu et al. [14] extend the Fukushima's work by introducing a generalized regularized gap function for classical variational inequalities. Motivated by [13, 14], we in this paper introduce a new type of generalized regularized gap functions

*Corresponding author.

E-mail addresses: asifzakir007@gmail.com (M. Asif), zubkhan403@gmail.com (Z. Khan), firoz77b@gmail.com (F. Babu), imfaizofficial@gmail.com (A.F. Ansari).

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for generalized vector variational-like inequalities (in short, GVVLI) without any scalarization parameter and any Lipschitz continuity of monotone operator.

Gap functions provide efficient methods to establish error bounds for variational inequalities, and error bounds play a key role in investigating the convergence analysis of optimization problems and variational inequalities. Error bounds give an upper estimate of the distance between the feasible point and solution set of the considered problem. For more details on error bounds, we refer to [15, 16, 17, 18, 19, 20] and the references therein.

We organized this paper as follows. In Section 2, we gather all the necessary definitions related to monotonicity and convexity. Section 3 is devoted to the formulation of the generalized vector variational-like inequalities and generalized variational-like inequalities with their special cases. In Section 4, without any scalarization techniques and any Lipschitz continuity, we propose a gap function and its generalized regularized version for the GVVLI with some examples. In Section 5, we use the generalized regularized gap function to establish error bounds for the GVVLI. In the last section, Section 6, we provide an example that helps to understand the numerical applicability of our proposed results.

2. PRELIMINARIES

Let K be a nonempty subset of \mathbb{R}^n . Let $\langle \cdot, \cdot \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ be an inner product, and let $\|\cdot\|$ be the corresponding norm. Consider the vector-valued mappings $A_j, B_j : \mathbb{R}^n \rightarrow \mathbb{R}^n$ ($j = 1, 2, \dots, m$) and a real valued mapping $h_j : \mathbb{R}^n \rightarrow \mathbb{R}$ ($j = 1, 2, \dots, m$). For abbreviation, let

$$A := (A_1, A_2, \dots, A_m), B := (B_1, B_2, \dots, B_m) \text{ and } H := (h_1, h_2, \dots, h_m),$$

and for every $p, q \in \mathbb{R}^n$, we denote

$$\langle A(p), q \rangle := (\langle A_1(p), q \rangle, \langle A_2(p), q \rangle, \dots, \langle A_m(p), q \rangle).$$

Definition 2.1. A mapping $\eta : K \times K \rightarrow \mathbb{R}^n$ is said to be skew if, for all $p, q \in K$,

$$\eta(p, q) + \eta(q, p) = 0.$$

Motivated by Verma [21], we define the pseudocontractivity in terms of the mapping $\eta : K \times K \rightarrow \mathbb{R}^n$.

Definition 2.2. A vector-valued mapping $F : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is said to be

- (a) [21] (ρ, η) -strongly monotone if, for all $p, q \in \mathbb{R}^n$, there exists a positive integer ρ such that

$$\langle F(p) - F(q), \eta(p, q) \rangle \geq \rho \|p - q\|^2.$$

- (b) (ς, η) -strongly pseudocontractive if, for all $p, q \in \mathbb{R}^n$, there exists a positive integer ς such that

$$\langle F(p) - F(q), \eta(p, q) \rangle \leq \varsigma \|p - q\|^2.$$

Definition 2.3. [22] A real-valued mapping $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is said to be convex if, for all $p, q \in \mathbb{R}^n$ and all $0 \leq \lambda \leq 1$,

$$f(\lambda p + (1 - \lambda)q) \leq \lambda f(p) + (1 - \lambda)f(q).$$

If $-f$ is convex, then f is said to be concave. Moreover, if f is both convex and concave, then f is said to be affine.

The following definition of the convexity can be adapted to vector-valued mappings.

Definition 2.4. A vector-valued mapping $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$ defined by $F = (f_1, f_2, \dots, f_m)$ is said to be convex if, for each $j = 1, 2, \dots, m$, mapping $f_j : \mathbb{R}^n \rightarrow \mathbb{R}$ is convex on \mathbb{R}^n .

3. GENERALIZED VECTOR VARIATIONAL-LIKE INEQUALITIES

Let K be a nonempty, closed, and convex subset of \mathbb{R}^n . Let $A_j, B_j : K \rightarrow \mathbb{R}^n$, $\eta : K \times K \rightarrow \mathbb{R}^n$ be vector-valued mappings, and let $h_j : K \rightarrow \mathbb{R}$ ($j = 1, 2, \dots, m$) be real valued mappings.

In this section, we propose the following generalized vector variational-like inequality, which is to

$$(GVVLI) \begin{cases} \text{find } p^* \in K \text{ such that} \\ \langle A(p^*) - B(p^*), \eta(p, p^*) \rangle + H(p) - H(p^*) \notin -\text{int} \mathbb{R}_+^m, \forall p \in K, \end{cases}$$

where \mathbb{R}_+^m is the nonnegative orthant of \mathbb{R}^m , and $\text{int} \mathbb{R}_+^m$ is an interior of \mathbb{R}_+^m . We denote by $\text{sol}(GVVLI)$ the solution set of the GVVLI. If $\eta(x, y) = x - y$ for all $x, y \in K$ and $A = B$, then the GVVLI reduces to the generalized vector variational inequality, which was considered and studied by Sun and Chai [12]. Moreover, if $\eta(x, y) = x - y$ for all $x, y \in K$, $A = B$, and $h_j = 0$ for each $j = 1, 2, \dots, m$, then the GVVLI reduces to the vector variational inequality, which was proposed and studied by Giannessi [1].

For $m = 1$, the GVVLI turns to the following generalized variational-like inequality (GVLI):

$$(GVLI) \begin{cases} \text{Find } p^* \in K \text{ such that} \\ \langle A_1(p^*) - B_1(p^*), \eta(p, p^*) \rangle + h_1(p) - h_1(p^*) \geq 0, \forall p \in K. \end{cases}$$

For $j = 1, 2, \dots, m$, $(GVLI)^j$ denotes the GVLI with A_j and h_j , and $\text{sol}(GVLI)^j$ denotes the set of solutions of $(GVLI)^j$.

If $H = B = 0$, the origin of the nonnegative orthant, then the GVVLI reduces to the generalized vector variational-like inequality, considered and discussed by Yang and Yang [23].

Finally, we end this section by the following definition.

Definition 3.1. Let K be a nonempty subset of \mathbb{R}^n . A real-valued mapping $\mathcal{G} : K \rightarrow \mathbb{R}$ is said to be a gap function for (GVVLI) if it satisfies the following assertions:

- (a) $\mathcal{G}(p) \geq 0$ for every $p \in K$,
- (b) $\mathcal{G}(p^*) = 0 \Leftrightarrow p^* \in K \cap \text{sol}(GVVLI)$.

4. GAP FUNCTIONS FOR GVVLI

In this section, we investigate a gap function and its generalized regularized version for the GVVLI.

4.1. Gap function. We propose the following gap function $\mathcal{G} : K \rightarrow \mathbb{R}$, which is independent of the scalarization parameter for GVVLI:

$$\mathcal{G}(p) = \sup_{q \in K} \min_{1 \leq j \leq m} \{ \langle A_j(p) - B_j(p), \eta(p, q) \rangle + h_j(p) - h_j(q) \}. \quad (4.1)$$

Proposition 4.1. Let K be a nonempty, closed, and convex subset of \mathbb{R}^n , and let $\eta : K \times K \rightarrow \mathbb{R}^n$ be a skew mapping. Let $\eta(p, \cdot)$ be affine for every $p \in K$, and let $h_j : \mathbb{R}^n \rightarrow \mathbb{R}$ be convex for each $j = 1, 2, \dots, m$. Then the function \mathcal{G} given by (4.1) is well-defined.

Proof. Assume that $\xi(p, q) = \min_{1 \leq j \leq m} \{\langle A_j(p) - B_j(p), \eta(p, q) \rangle + h_j(p) - h_j(q)\}$. Then, $\mathcal{G}(p) = \sup_{q \in K} \{\xi(p, q)\}$, $\forall p \in K$. Since $\eta(p, \cdot)$ is affine for every $p \in K$, and h_j is convex for each $j = 1, 2, \dots, m$, then, for every $p, q_1, q_2 \in K$ and $0 \leq \alpha \leq 1$,

$$\begin{aligned} & \xi(p, \alpha q_1 + (1 - \alpha)q_2) \\ &= \min_{1 \leq j \leq m} \{\langle A_j(p) - B_j(p), \eta(p, \alpha q_1 + (1 - \alpha)q_2) \rangle + h_j(p) - h_j(\alpha q_1 + (1 - \alpha)q_2)\} \\ &\geq \min_{1 \leq j \leq m} \{\alpha \langle A_j(p) - B_j(p), \eta(p, q_1) \rangle + (1 - \alpha) \langle A_j(p) - B_j(p), \eta(p, q_2) \rangle + \alpha h_j(p) \\ &\quad + (1 - \alpha)h_j(p) - \alpha h_j(q_1) - (1 - \alpha)h_j(q_2)\} \\ &= \alpha \xi(p, q_1) + (1 - \alpha)\xi(p, q_2), \end{aligned}$$

that is, $\xi(p, \cdot)$ is concave on closed and convex subset K for every $p \in K$. Then $\xi(p, \cdot)$ attains the maximum at a unique point in K . Hence, \mathcal{G} is well defined. \square

Theorem 4.1. *Let K be a nonempty closed convex subset of \mathbb{R}^n , and let $\eta : K \times K \rightarrow \mathbb{R}^n$ be a skew mapping. Let $\eta(p, \cdot)$ be affine for every $p \in K$, and let $h_j : \mathbb{R}^n \rightarrow \mathbb{R}$ be convex for each $j = 1, 2, \dots, m$. Then the function $\mathcal{G} : K \rightarrow \mathbb{R}$ defined by (4.1) is a gap function for GVLLI.*

Proof. (i) Since $\mathcal{G}(p) \geq \min_{1 \leq j \leq m} \{\langle A_j(p) - B_j(p), \eta(p, q) \rangle + h_j(p) - h_j(q)\}$, $\forall p \in K$. By setting $q = p$, it is easy to see that $\mathcal{G}(p) \geq 0$ for all $p \in K$.

(ii) Let $p^* \in K$ and $\mathcal{G}(p^*) = 0$. It follows that

$$\sup_{q \in K} \min_{1 \leq j \leq m} \{\langle A_j(p^*) - B_j(p^*), \eta(p^*, q) \rangle + h_j(p^*) - h_j(q)\} = 0.$$

This guarantees that

$$\min_{1 \leq j \leq m} \{\langle A_j(p^*) - B_j(p^*), \eta(p^*, q) \rangle + h_j(p^*) - h_j(q)\} \leq 0, \forall q \in K.$$

Then there exists a $1 \leq j_0 \leq m$ such that the inequality above reduces to

$$\langle A_{j_0}(p^*) - B_{j_0}(p^*), \eta(p^*, q) \rangle + h_{j_0}(p^*) - h_{j_0}(q) \leq 0, \forall q \in K.$$

Since η is skew, then $\langle A_{j_0}(p^*) - B_{j_0}(p^*), \eta(q, p^*) \rangle + h_{j_0}(q) - h_{j_0}(p^*) \geq 0, \forall q \in K$. It follows that

$$\langle A(p^*) - B(p^*), \eta(q, p^*) \rangle + H(q) - H(p^*) \notin -\text{int } \mathbb{R}_+^m, \forall q \in K.$$

This implies that $p^* \in \text{sol}(GVLLI)$.

Conversely, let p^* be a solution of GVLLI. Then, there exists a $1 \leq j_0 \leq m$ such that

$$\langle A_{j_0}(p^*) - B_{j_0}(p^*), \eta(q, p^*) \rangle + h_{j_0}(q) - h_{j_0}(p^*) \geq 0, \forall q \in K,$$

Since η is a skew mapping, one concludes that

$$\langle A_{j_0}(p^*) - B_{j_0}(p^*), \eta(p^*, q) \rangle + h_{j_0}(p^*) - h_{j_0}(q) \leq 0, \forall q \in K,$$

and

$$\min_{1 \leq j \leq m} \{\langle A_j(p^*) - B_j(p^*), \eta(p^*, q) \rangle + h_j(p^*) - h_j(q)\} \leq 0, \forall q \in K.$$

It follows that $\mathcal{G}(p^*) \leq 0$. In view of $\mathcal{G}(p) \geq 0, \forall p \in K$, one has $\mathcal{G}(p^*) = 0$. This completes the proof. \square

Example 4.1. Consider a closed convex subset $K = [-1, 1] \times [-1, 1]$ of \mathbb{R}^2 . For $n = 1$ and $m = 2$, define four mappings $A_1, A_2 : K \rightarrow \mathbb{R}^2$ and $B_1, B_2 : K \rightarrow \mathbb{R}^2$ by

$$A_1(p) = (-3p_1, 1 - 4p_2) \quad \text{and} \quad A_2(p) = (-3p_1 - p_1^3, -8p_2), \quad \forall p = (p_1, p_2) \in K,$$

and

$$B_1(p) = (-2p_1, 1 - 4p_2) \quad \text{and} \quad B_2(p) = (-p_1, -8p_2), \quad \forall p = (p_1, p_2) \in K.$$

Define three mappings $h_1, h_2 : K \rightarrow \mathbb{R}$ and $\eta : K \times K \rightarrow \mathbb{R}^2$ by

$$h_1(p) = p_1^2 \quad \text{and} \quad h_2(p) = p_1^4 \quad \eta(p, q) = q - p, \quad \forall p = (p_1, p_2), \quad \forall q = (q_1, q_2) \in K.$$

Note that h_1 and h_2 are convex on K , and η is skew and affine on the second argument. Then GVLI's are given by

$$\begin{aligned} \text{sol(GVLI)}^1 &= \{p \in K : \langle A_1(p) - B_1(p), \eta(q, p) \rangle + h_1(q) - h_1(p) \geq 0, \forall q \in K\} \\ &= \{p \in K : -p_1(p_1 - q_1) + q_1^2 - p_1^2 \geq 0, \forall q \in K\} \\ &= \{(0, 0)\}, \end{aligned}$$

and

$$\begin{aligned} \text{sol(GVLI)}^2 &= \{p \in K : \langle A_2(p) - B_2(p), \eta(q, p) \rangle + h_2(q) - h_2(p) \geq 0, \forall q \in K\} \\ &= \{p \in K : -(2p_1 + p_1^3)(p_1 - q_1) + q_1^4 - p_1^4 \geq 0, \forall q \in K\} \\ &= \{(0, 0)\}. \end{aligned}$$

Therefore, $\bigcap_{j=1}^2 \text{sol(GVLI)}^j = \{(0, 0)\} \neq \emptyset$. From the definition of function $\mathcal{G} : \mathbb{R}^2 \rightarrow \mathbb{R}$, we obtain

$$\begin{aligned} \mathcal{G}(p) &= \sup_{q \in [-1, 1]} \min_{1 \leq j \leq m} \{ \langle A_j(p) - B_j(p), \eta(p, q) \rangle + h_j(p) - h_j(q) \} \\ &= \sup_{q \in [-1, 1]} \{ p_1(p_1 - q_1) + p_1^2 - q_1^2 \} \\ &= \frac{9}{4} p_2^2. \end{aligned}$$

Hence, \mathcal{G} is a gap function for GVLI.

4.2. Generalized regularized gap. Let $\mathcal{F} : K \times K \rightarrow \mathbb{R}$ be a mapping. For $\alpha > 0$, we define a generalized regularized gap function $\mathcal{G}_\alpha : K \rightarrow \mathbb{R}$ for GVLI:

$$\mathcal{G}_\alpha(p) = \sup_{q \in K} \{ \min_{1 \leq j \leq m} \{ \langle A_j(p) - B_j(p), \eta(p, q) \rangle + h_j(p) - h_j(q) \} - \alpha \mathcal{F}(p, q) \}, \quad \forall p \in K. \quad (4.2)$$

Proposition 4.2. Let K be a nonempty, closed, and convex subset of \mathbb{R}^n . Let $\eta : K \times K \rightarrow \mathbb{R}^n$ be a skew mapping, and let $\eta(p, \cdot)$ be affine. Let $\mathcal{F}(p, \cdot)$ be convex for every $p \in K$, and let h_j be convex for each $j = 1, 2, \dots, m$. Then, for $\alpha > 0$, the \mathcal{G}_α given by (4.2) is well defined.

Proof. Let $\xi(p, q) = \min_{1 \leq j \leq m} \{ \langle A_j(p) - B_j(p), \eta(p, q) \rangle + h_j(p) - h_j(q) \}$. Then

$$\mathcal{G}_\alpha(p) = \sup_{q \in K} \{ \xi(p, q) - \alpha \mathcal{F}(p, q) \}, \quad \forall p \in K. \quad (4.3)$$

Since $\eta(p, \cdot)$ is affine for every $p \in K$, and h_j is convex for each $j = 1, 2, \dots, m$, then it follows from Proposition 4.1 that $\xi(p, \cdot)$ is concave. From the convexity of $\mathcal{F}(p, \cdot)$, $\xi(p, \cdot) - \alpha \mathcal{F}(p, \cdot)$

is concave for every $p \in K$. Since K is closed and convex, then function $\xi(p, \cdot) - \alpha \mathcal{F}(p, \cdot)$ attains its maximum at the unique point in K . Thus \mathcal{G}_α is well defined. \square

From now on, we present some assumptions for the mapping $\mathcal{F} : K \times K \rightarrow \mathbb{R}$ as follows.

(P1) \mathcal{F} is continuously differentiable on $K \times K$;

(P2) \mathcal{F} is non-negative on $K \times K$;

(P3) $\mathcal{F}(\tau, \cdot)$ is strongly convex on subset K for every $\tau \in K$, i.e., $\exists \ell > 0$ such that

$$\mathcal{F}(\tau, p) - \mathcal{F}(\tau, q) \geq \langle \nabla_2 \mathcal{F}(\tau, q), p - q \rangle + \ell \|p - q\|^2, \quad \forall p, q, \tau \in K;$$

where $\nabla_2 \mathcal{F}$ denotes the gradient of $\mathcal{F}(p, \cdot)$ for every $p \in K$;

(P4) $\mathcal{F}(p, q) = 0 \Leftrightarrow p = q$;

(P5) $\nabla_2 \mathcal{F}(\tau, \cdot)$ is Lipschitz continuous for every $\tau \in K$, i.e., $\exists \kappa > 0$ such that

$$\|\nabla_2 \mathcal{F}(\tau, p) - \nabla_2 \mathcal{F}(\tau, q)\| \leq \kappa \|p - q\|, \quad \forall p, q, \tau \in K.$$

Lemma 4.1. *If $\mathcal{F} : K \times K \rightarrow \mathbb{R}$ satisfies (P1) - (P5), then*

$$\mathcal{F}(p, q) \leq (\kappa - \ell) \|p - q\|^2, \quad \forall p, q \in K.$$

Proof. It follows from (P3) that

$$\mathcal{F}(p, p) - \mathcal{F}(p, q) \geq \langle \nabla_2 \mathcal{F}(p, q), p - q \rangle + \ell \|p - q\|^2, \quad \forall p, q \in K,$$

that is, $-\mathcal{F}(p, q) \geq \langle \nabla_2 \mathcal{F}(p, q), p - q \rangle + \ell \|p - q\|^2, \forall p, q \in K$. This implies

$$\mathcal{F}(p, q) \leq \|\nabla_2 \mathcal{F}(p, q)\| \|p - q\| - \ell \|p - q\|^2, \quad \forall p, q \in K. \quad (4.4)$$

From (P5), we have $\|\nabla_2 \mathcal{F}(p, p) - \nabla_2 \mathcal{F}(p, q)\| \leq \kappa \|p - q\|$. In view of (P4), we obtain

$$\|\nabla_2 \mathcal{F}(p, q)\| \leq \kappa \|p - q\|. \quad (4.5)$$

Using inequalities (4.4) and (4.5), we arrive at $\mathcal{F}(p, q) \leq (\kappa - \ell) \|p - q\|^2, \forall p, q \in K$. The proof is complete. \square

Theorem 4.2. *Let K be a nonempty, closed and convex subset of \mathbb{R}^n , and let $\eta : K \times K \rightarrow \mathbb{R}^n$ be a skew mapping. Let $\eta(p, \cdot)$ be affine for every $p \in K$, and let $h_j : \mathbb{R}^n \rightarrow \mathbb{R}$ be convex for each $j = 1, 2, \dots, m$. If $\mathcal{F} : K \times K \rightarrow \mathbb{R}$ satisfy (P1) - (P5), then the function \mathcal{G}_α ($\alpha > 0$) defined by (4.2) is a gap function for GVLLI.*

Proof. (i) By setting $q = p$ in (4.2), we find that, for all $p \in K$,

$$\mathcal{G}_\alpha(p) = \sup_{q \in K} \left\{ \min_{1 \leq j \leq m} \{ \langle A_j(p) - B_j(p), \eta(p, q) \rangle + h_j(p) - h_j(q) \} - \alpha \mathcal{F}(p, q) \right\} \geq 0.$$

(ii) Let $\xi(p, q) = \min_{1 \leq j \leq m} \{ \langle A_j(p) - B_j(p), \eta(p, q) \rangle + h_j(p) - h_j(q) \}$, $p^* \in K$ and $\mathcal{G}_\alpha(p^*) = 0$.

Then, $\xi(p^*, q) \leq \alpha \mathcal{F}(p^*, q)$ for all $q \in K$. For random $p \in K$ and every $0 < \lambda < 1$, let $q = p + \lambda(p^* - p)$ for all $p \in K$. It follows from Lemma 4.1 that

$$\begin{aligned} \xi(p^*, p + \lambda(p^* - p)) &\leq \alpha \mathcal{F}(p^*, p + \lambda(p^* - p)) \\ &\leq \alpha(\kappa - \ell) \|p + \lambda(p^* - p) - p^*\|^2 \\ &= \alpha(\kappa - \ell)(1 - \lambda)^2 \|p - p^*\|^2, \end{aligned}$$

that is,

$$\begin{aligned} & \min_{1 \leq j \leq m} \{ \langle A_j(p^*) - B_j(p^*), \eta(p^*, p + \lambda(p^* - p)) \rangle + h_j(p^*) - h_j(p + \lambda(p^* - p)) \} \\ & \leq \alpha(\kappa - \ell)(1 - \lambda)^2 \|p - p^*\|^2. \end{aligned}$$

Since, for each $j = 1, 2, \dots, m$, h_j is convex, and for every $p \in K$, $\eta(p, \cdot)$ is affine, then

$$\begin{aligned} & \min_{1 \leq j \leq m} \{ \langle A_j(p^*) - B_j(p^*), \lambda \eta(p^*, p^*) + (1 - \lambda) \eta(p^*, p) \rangle + h_j(p^*) \\ & \quad - \lambda h_j(p^*) - (1 - \lambda) h_j(p) \} \\ & \leq \alpha(\kappa - \ell)(1 - \lambda)^2 \|p - p^*\|^2. \end{aligned}$$

Hence,

$$\begin{aligned} & (1 - \lambda) \min_{1 \leq j \leq m} \{ \langle A_j(p^*) - B_j(p^*), \eta(p^*, p) \rangle + h_j(p^*) - h_j(p) \} \\ & \leq \alpha(\kappa - \ell)(1 - \lambda)^2 \|p - p^*\|^2, \end{aligned}$$

and

$$\begin{aligned} & \min_{1 \leq j \leq m} \{ \langle A_j(p^*) - B_j(p^*), \eta(p^*, p) \rangle + h_j(p^*) - h_j(p) \} \\ & \leq \alpha(\kappa - \ell)(1 - \lambda) \|p - p^*\|^2. \end{aligned}$$

By taking $\lambda \rightarrow 1$, we arrive at $\min_{1 \leq j \leq m} \{ \langle A_j(p^*) - B_j(p^*), \eta(p^*, p) \rangle + h_j(p^*) - h_j(p) \} \leq 0$. Thus, there exists a $1 \leq j_0 \leq m$ such that

$$\langle A_{j_0}(p^*) - B_{j_0}(p^*), \eta(p^*, p) \rangle + h_{j_0}(p^*) - h_{j_0}(p) \leq 0.$$

It follows that $\langle A(p^*) - B(p^*), \eta(p, p^*) \rangle + H(p) - H(p^*) \notin \text{int } \mathbb{R}_+^m, \forall p \in K$. Hence, we obtain $p^* \in \text{sol}(GVVLI)$.

Conversely, we assume that p^* is a solution of GVVLI. Then, there exists a $1 \leq i_0 \leq m$ such that

$$\langle A_{j_0}(p^*) - B_{j_0}(p^*), \eta(q, p^*) \rangle + h_{j_0}(q) - h_{j_0}(p^*) \geq 0, \forall q \in K,$$

From the skew property of η , we have

$$\langle A_{j_0}(p^*) - B_{j_0}(p^*), \eta(p^*, q) \rangle + h_{j_0}(p^*) - h_{j_0}(q) \leq 0, \forall q \in K,$$

and hence,

$$\min_{1 \leq j \leq m} \{ \langle A_j(p^*) - B_j(p^*), \eta(p^*, q) \rangle + h_j(p^*) - h_j(q) \} \leq 0, \forall q \in K.$$

that is, $\xi(p^*, q) \leq 0$. This together with (4.3) yields that $\mathcal{G}_\alpha(p^*) \leq 0$. But $\mathcal{G}_\alpha(p^*) \geq 0$ for all $p \in K$. So, $\mathcal{G}_\alpha(p^*) = 0$. The proof is complete. \square

Example 4.2. Consider $K, A_1, A_2, B_1, B_2, h_1, h_2$, and η as same as in Example 4.1 and define $\mathcal{F} : \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$ by

$$\mathcal{F}(p, q) = \frac{1}{2}(p_1 - q_1)^2 + \frac{1}{2}(p_2 - q_2)^2, \quad \forall p = (p_1, p_2), \forall q = (q_1, q_2) \in K.$$

Clearly, \mathcal{F} satisfies conditions (P1) - (P5). From the definition of the function $\mathcal{G}_\alpha : \mathbb{R}^2 \rightarrow \mathbb{R}$ for $\alpha > 0$, we have

$$\begin{aligned} \mathcal{G}_\alpha(p) &= \sup_{q \in [-1,1]} \left\{ \min_{1 \leq j \leq m} \{ \langle A_j(p) - B_j(p), \eta(p, q) \rangle + h_j(p) - h_j(q) \} - \alpha \mathcal{F}(p, q) \right\} \\ &= \sup_{q \in [-1,1]} \left\{ p_1(p_1 - q_1) + p_1^2 - q_1^2 - \frac{\alpha}{2}(p_1 - q_1)^2 - \frac{\alpha}{2}(p_2 - q_2)^2 \right\} \\ &= \frac{9}{2(\alpha + 2)} p_1^2. \end{aligned}$$

Thus, \mathcal{G}_α is a gap function for GVCLI.

5. ERROR BOUNDS

In this section, we obtain the error bounds for GVCLI with the help of the gap function and the generalized regularized gap function given by (4.2).

Lemma 5.1. *If $A_j : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is (ρ_j, η) -strongly monotone, and $B_j : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is (ζ_j, η) -strongly pseudocontractive over K such that $\rho_j > \zeta_j$ for each $j = 1, 2, \dots, m$, then $A_j - B_j$ is $(\rho_j - \zeta_j, \eta)$ -strongly monotone.*

Proof. By the additivity property of inner products, for every $p, q \in \mathbb{R}^n$, we have

$$\begin{aligned} &\langle A_j(p) - B_j(p), \eta(p, q) \rangle \\ &= \langle (A_j(p) - B_j(p)) - (A_j(q) - B_j(q)) + (A_j(q) - B_j(q)), \eta(p, q) \rangle \\ &= \langle (A_j(p) - A_j(q)) - (B_j(p) - B_j(q)), \eta(p, q) \rangle + \langle (A_j(q) - B_j(q)), \eta(p, q) \rangle \\ &= \langle (A_j(p) - A_j(q)), \eta(p, q) \rangle - \langle (B_j(p) - B_j(q)), \eta(p, q) \rangle + \langle (A_j(q) - B_j(q)), \eta(p, q) \rangle. \end{aligned}$$

From the (ρ_j, η) -strongly monotonicity and (ζ_j, η) -strongly pseudocontractivity of A_j and B_j , respectively, we conclude

$$\langle A_j(p) - B_j(p), \eta(p, q) \rangle \geq \langle A_j(q) - B_j(q), \eta(p, q) \rangle + (\rho_j - \zeta_j) \|p - q\|^2.$$

This completes the proof. \square

Theorem 5.1. *Let $\eta : K \times K \rightarrow \mathbb{R}^n$ be a skew mapping, and let $\eta(p, \cdot)$ be affine for every $p \in K$. For each $j = 1, 2, \dots, m$, let $A_j : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be (ρ_j, η) -strongly monotone, $B_j : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be (ζ_j, η) -strongly pseudocontractive with $\rho_j > \zeta_j$, respectively over K , and $h_j : \mathbb{R}^n \rightarrow \mathbb{R}$ be convex. Let $\mathcal{F} : K \times K \rightarrow \mathbb{R}$ satisfy (P1)-(P5). Further, suppose that $\bigcap_{j=1}^m \text{sol}(\text{GVLI})^j$ is nonempty, $\vartheta = \min_{1 \leq j \leq m} (\rho_j - \zeta_j)$, and $\alpha, \kappa, \ell > 0$ such that $\vartheta > \alpha(\kappa - \ell)$. Then*

$$d(p, \text{sol}(\text{GVCLI})) \leq \frac{1}{\sqrt{(\vartheta - \alpha(\kappa - \ell))}} \sqrt{\mathcal{G}_\alpha(p)}, \quad \forall p \in K,$$

where $d(p, \text{sol}(\text{GVCLI}))$ is the distance between $\text{sol}(\text{GVCLI})$ and point p .

Proof. By (4.3) and Lemma 4.1, for every $q \in K$, we have

$$\mathcal{G}_\alpha(p) \geq \xi(p, q) - \alpha \mathcal{F}(p, q) \geq \xi(p, q) - \alpha(\kappa - \ell) \|p - q\|^2.$$

Since every $(GVLI)^j$ has same solution as $\bigcap_{j=1}^m \text{sol}(GVLI)^j$, which is nonempty. If $p^* \in K$ is a solution, then p^* is the solution to GVVLI. Note that $\mathcal{G}_\alpha(p) \geq \min_{1 \leq j \leq m} \{\langle A_j(p) - B_j(p), \eta(p, p^*) \rangle + h_j(p) - h_j(p^*)\} - \alpha(\kappa - \ell) \|p - p^*\|^2$. Suppose

$$\begin{aligned} & \langle A_j(p) - B_j(p), \eta(p, p^*) \rangle + h_j(p) - h_j(p^*) \\ &= \min_{1 \leq j \leq m} \{\langle A_j(p) - B_j(p), \eta(p, p^*) \rangle + h_j(p) - h_j(p^*)\}. \end{aligned}$$

By using Lemma 5.1, we have

$$\mathcal{G}_\alpha(p) \geq \langle A_j(p^*) - B_j(p^*), \eta(p, p^*) \rangle + h_j(p) - h_j(p^*) + (\vartheta - \alpha(\kappa - \ell)) \|p - p^*\|^2. \quad (5.1)$$

Since p^* is the solution of $(GVLI)^j$, then

$$\langle A_j(p^*) - B_j(p^*), \eta(p, p^*) \rangle + h_j(p) - h_j(p^*) \geq 0. \quad (5.2)$$

Form (5.1) and (5.2), we obtain $\mathcal{G}_\alpha(p) \geq (\vartheta - \alpha(\kappa - \ell)) \|p - p^*\|^2$. It follows that

$$\|p - p^*\| \leq \frac{1}{\sqrt{(\vartheta - \alpha(\kappa - \ell))}} \sqrt{\mathcal{G}_\alpha(p)}.$$

This implies that

$$d(p, \text{sol}(GVVLI)) \leq \frac{1}{\sqrt{(\vartheta - \alpha(\kappa - \ell))}} \sqrt{\mathcal{G}_\alpha(p)}.$$

This completes the proof. \square

Similarly, by using gap function \mathcal{G} , we obtain the following error bound for (GVVLI).

Corollary 5.1. *Let $\eta : K \times K \rightarrow \mathbb{R}^n$ be a skew mapping, and let $\eta(p, \cdot)$ be affine for every $p \in K$. Let $A_1 : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be (ρ_1, η) -strongly monotone. Let $B_1 : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be (ς_1, η) -strongly pseudocontractive with $\rho_1 > \varsigma_1$, respectively over K , and let $h_1 : \mathbb{R}^n \rightarrow \mathbb{R}$ be convex. If $\text{sol}(GVLI)^1$ is nonempty and $\vartheta = \rho_1 - \varsigma_1$, then $d(p, \text{sol}(GVLI)) \leq \sqrt{\frac{\mathcal{G}(p)}{\vartheta}}$, $\forall p \in K$, where $d(p, \text{sol}(GVLI))$ is the distance between $\text{sol}(GVLI)$ and point p .*

6. NUMERICAL RESULT

Example 6.1. Let $K = [-1, 1] \subseteq \mathbb{R}$ be a closed and convex subset, and let $\eta : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$, be a mapping given by

$$\eta(p, q) = q - p, \quad \forall p, q \in K.$$

For $n = 1, m = 2$, define four mappings $A_1, A_2 : K \rightarrow \mathbb{R}$ and $B_1, B_2 : K \rightarrow \mathbb{R}$ by

$$A_1(p) = -3p \quad \text{and} \quad A_2(p) = -3p - p^3 \quad \forall p \in K$$

and

$$B_1(p) = -2p \quad \text{and} \quad B_2(p) = -p \quad \forall p \in K.$$

Then

$$A(p) = (-3p, -3p - p^3) \quad \text{and} \quad B(p) = (-2p, -p), \quad \forall p \in K.$$

Note that A_1 and A_2 are $(3, \eta)$ -strongly monotone and $(3, \eta)$ -strongly monotone, respectively, and B_1 and B_2 are $(2, \eta)$ -strongly pseudocontractive and $(1, \eta)$ -strongly pseudocontractive, respectively. Let $h_1, h_2 : \mathbb{R} \rightarrow \mathbb{R}$, $\eta : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$ and $\mathcal{F} : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$ be mappings defined by

$$h_1(p) = p^2 \quad \text{and} \quad h_2(p) = p^4, \quad \forall p \in K$$

and

$$\mathcal{F}(p, q) = (p - q)^2, \quad \forall p, q \in K.$$

Notice that h_1 and h_2 are convex, η is skew and affine on the second argument, and \mathcal{F} satisfies (P1) - (P5). Clearly, $l = 1$, $\kappa = 2$, and the GVLI are given by

$$\begin{aligned} \text{sol}(\text{GVLI})^1 &= \{p \in K : \langle A_1(p) - B_1(p), \eta(q, p) \rangle + h_1(q) - h_1(p) \geq 0, \forall q \in K\} \\ &= \{p \in K : -p(p - q) + q^2 - p^2 \geq 0, \forall q \in K\} = \{0\} \end{aligned}$$

and

$$\begin{aligned} \text{sol}(\text{GVLI})^2 &= \{p \in K : \langle A_2(p) - B_2(p), \eta(q, p) \rangle + h_2(q) - h_2(p) \geq 0, \forall q \in K\} \\ &= \{p \in K : -(2p + p^3)(p - q) + q^4 - p^4 \geq 0, \forall q \in K\} = \{0\}. \end{aligned}$$

Therefore, we have $\bigcap_{j=1}^2 \text{sol}(\text{GVLI})^j = \{0\} \neq \emptyset$. From the definition of the function $\mathcal{G} : \mathbb{R} \rightarrow \mathbb{R}$, we obtain

$$\begin{aligned} \mathcal{G}(p) &= \sup_{q \in [-1, 1]} \min_{1 \leq j \leq m} \{\langle A_j(p) - B_j(p), \eta(p, q) \rangle + h_j(p) - h_j(q)\} \\ &= \sup_{q \in [-1, 1]} \min_{1 \leq j \leq m} \{p(p - q) + p^2 - q^2, (2p + p^3)(p - q) + p^4 - q^4\} \\ &= \sup_{q \in [-1, 1]} \{p(p - q) + p^2 - q^2\} \\ &= \frac{9}{4}p^2. \end{aligned}$$

Hence, $\mathcal{G}(p)$ is a gap function for GVLI. By the definition of the function $\mathcal{G}_\alpha : \mathbb{R} \rightarrow \mathbb{R}$ for $0 < \alpha < 1$, we have

$$\begin{aligned} \mathcal{G}_\alpha(p) &= \sup_{q \in [-1, 1]} \left\{ \min_{1 \leq j \leq m} \{\langle A_j(p) - B_j(p), \eta(p, q) \rangle + h_j(p) - h_j(q)\} - \alpha \mathcal{F}(p, q) \right\} \\ &= \sup_{q \in [-1, 1]} \{p(p - q) + p^2 - q^2 - \alpha(p - q)^2\} \\ &= \frac{9}{4(1 + \alpha)}p^2. \end{aligned}$$

Thus, $\mathcal{G}_\alpha(p)$ is a gap function for GVVLI. Therefore, all the conditions of Theorem 5.1 are satisfied. Since $0 < \alpha < 1$, $\vartheta = 1$, $\kappa = 2$, and $l = 1$. Then

$$\begin{aligned} \frac{1}{\sqrt{(\vartheta - \alpha(\kappa - l))}} \sqrt{\mathcal{G}_\alpha(p)} &= \frac{1}{\sqrt{(1 - \alpha(2 - 1))}} \sqrt{\frac{9p^2}{4(1 + \alpha)}} \\ &= \frac{3p}{2\sqrt{1 - \alpha^2}} > p. \end{aligned}$$

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