

Research Article

On Set-Valued Complementarity Problems

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This paper investigates the set-valued complementarity problems (SVCP) which poses rather different features from those that classical complementarity problems hold, due to the fact that the index set is not fixed, but dependent on x . While comparing the set-valued complementarity problems with the classical complementarity problems, we analyze the solution set of SVCP. Moreover, properties of merit functions for SVCP are studied, such as being level bounded and error bounded. Finally, some possible research directions are discussed.

1. Motivations and Preliminaries

The set-valued complementarity problem (SVCP) is to find $x \in \mathbb{R}^n$ such that

$$x \geq 0, \quad y \geq 0, \quad x^T y = 0, \quad \text{for some } y \in \Theta(x), \quad (1)$$

where $\Theta : \mathbb{R}_+^n \rightrightarrows \mathbb{R}^n$ is a set-valued mapping. The set-valued complementarity problem plays an important role in the sensitivity analysis of complementarity problems [1] and economic equilibrium problems [2]. However, there has been very little study on the set-valued complementarity problems compared to the classical complementarity problems. In fact, the SVCP (1) can be recast as follows, which is denoted by SVNCP(F, Ω) to find $x \in \mathbb{R}^n$, such that

$$x \geq 0, \quad F(x, w) \geq 0, \quad x^T F(x, w) = 0, \quad (2)$$

for some $w \in \Omega(x)$,

where $F : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n$ and $\Omega : \mathbb{R}^n \rightrightarrows \mathbb{R}^m$ are a set-valued mapping. To see this, if letting

$$\Theta(x) = \bigcup_{w \in \Omega(x)} \{F(x, w)\}, \quad (3)$$

then (1) reduces to (2). Conversely, if $F(x, w) = w$ and $\Omega(x) = \Theta(x)$, then (2) takes the form of (1).

The SVNCP(F, Ω) given as in (2) provides an unified framework for several interesting and important problems in optimization fields described as below.

(i) *Nonlinear complementarity problem* [1], which is to find $x \in \mathbb{R}^n$, such that

$$x \geq 0, \quad F(x) \geq 0, \quad \langle x, F(x) \rangle = 0. \quad (4)$$

This corresponds to $F(x, w) := F(x) + w$ and $\Omega(x) = \{0\}$ for all $x \in \mathbb{R}^n$. In other words, the set-valued complementarity problem reduces to the classical complementarity problem under such case.

(ii) *Extended linear complementarity problem* [3, 4], which is to find $x, w \in \mathbb{R}^n$, such that

$$x \geq 0, \quad w \geq 0, \quad x^T w = 0, \quad (5)$$

with $M_1 x - M_2 w \in P$,

where $M_1, M_2 \in \mathbb{R}^{m \times n}$ and $P \subseteq \mathbb{R}^m$ are a polyhedron. This corresponds to $F(x, w) = w$ and $\Omega(x) = \{w | M_1 x - M_2 w \in P\}$. In particular, when $P = \{q\}$, it further reduces to the horizontal linear complementarity problem and to the usual linear complementarity problem, in addition to M_2 being an identity matrix.

(iii) *Implicit complementarity problem* [5], which is to find $x, w \in \mathbb{R}^n$ and $z \in \mathbb{R}^m$, such that

$$x \geq 0, \quad w \geq 0, \quad x^T w = 0, \quad \text{with } F(x, w, z) = 0, \quad (6)$$

where $F : \mathbb{R}^{2n \times m} \rightarrow \mathbb{R}^l$. This can be rewritten as

$$x \geq 0, \quad w \geq 0, \quad x^T w = 0, \quad (7)$$

with w satisfying $F(x, w, z) = 0$ for some z .

This is clearly an SVNCP(F, Ω) where $F(x, w) = w$ and $\Omega(x) = \cup_{z \in \mathbb{R}^m} \{w \mid F(x, w, z) = 0\}$.

(iv) *Mixed nonlinear complementarity problem*, which is to find $x \in \mathbb{R}^n$, and $w \in \mathbb{R}^m$ such that

$$x \geq 0, \quad F(x, w) \geq 0, \quad \langle x, F(x, w) \rangle = 0, \quad (8)$$

with $G(x, w) = 0$.

This is an SVNCP(F, Ω) where it corresponds to $\Omega(x) = \{x \mid G(x, w) = 0\}$. Note that the mixed nonlinear complementarity problem is a natural extension of Karush-Kuhn-Tucker (KKT) conditions for the following nonlinear programming:

$$\begin{aligned} \min \quad & f(x) \\ \text{s.t.} \quad & g_i(x) \leq 0, \quad i = 1, 2, \dots, m, \\ & h_j(x) = 0, \quad j = 1, \dots, l. \end{aligned} \quad (9)$$

To see this, we first write out the KKT conditions as follows:

$$\begin{aligned} \nabla f(x) + \sum_{i=1}^m \lambda_i \nabla g_i(x) + \sum_{j=1}^l \mu_j \nabla h_j(x) &= 0, \\ h(x) &= 0, \end{aligned} \quad (10)$$

$$g(x) \leq 0, \quad \lambda \geq 0, \quad \langle \lambda, g(x) \rangle = 0,$$

where $g(x) := (g_1(x), \dots, g_m(x))$, $h(x) := (h_1(x), \dots, h_l(x))$, and $\lambda := (\lambda_1, \dots, \lambda_m)$. Then, letting $w := (\lambda, \mu)$, $F(x, w) := -g(x)$, and

$$G(x, w) := \begin{pmatrix} \nabla f(x) + \sum_{i=1}^m \lambda_i \nabla g_i(x) + \sum_{j=1}^l \mu_j \nabla h_j(x) \\ h(x) \end{pmatrix} \quad (11)$$

implies that the KKT system (10) becomes a mixed complementarity problem.

Besides the above various complementarity problems, SVNCP(F, Ω) has a close relation with the Quasi-variational inequality, a special of the extended general variational inequalities [6, 7], and min-max programming, which is elaborated as below.

(v) *Quasi-variational inequality* [2]. Given a point-to-point map F from \mathbb{R}^n to itself and a point-to-set map K from \mathbb{R}^n into subsets of \mathbb{R}^n , the Quasi-variational inequality QVI(K, F) is to find a vector $x \in K(x)$, such that

$$\langle F(x), y - x \rangle \geq 0, \quad \forall y \in K(x). \quad (12)$$

It is well-known that QVI(K, F) reduces to the classical nonlinear complementarity problem when $K(x)$ is independent of x , say, $K(x) = \mathbb{R}_+^n$ for all x . Now, let us explain why it is related to SVNCP(F, Ω). To this end, given $x \in \mathbb{R}^n$, we define $I(x) = \{i \mid F_i(x) > 0\}$, and let

$$K(x) = \{x \mid x_i \geq 0 \text{ for } i \in I \setminus I(x) \text{ and } x_i = 0 \text{ for } i \in I(x)\}. \quad (13)$$

Clearly, $0 \in K(x)$ which says $\langle x, F(x) \rangle \leq 0$ by taking $y = 0$ in (12). Note that $x \geq 0$ because $x \in K(x)$. Next, we will claim that $F_i(x) \geq 0$ for all $i = 1, 2, \dots, n$. It is enough to consider the case where $i \in I \setminus I(x)$. Under such case, by taking $y = \beta e_i$ in (12) with β being an arbitrarily positive scalar, we have $\beta F_i(x) \geq F(x)^T x$. Since β can be made sufficiently large, it implies that $F_i(x) \geq 0$. Thus, we obtain $F(x)^T x \geq 0$. In summary, under such case, QVI(K, F) becomes

$$x \geq 0, \quad F(x) \geq 0, \quad x^T F(x) = 0, \quad \text{with } x \in K(x) \quad (14)$$

which is an SVNCP(F, Ω).

(vi) *Min-max programming* [8], which is to solve the following problem:

$$\min_{x \in \mathbb{R}_+^n} \max_{w \in \Omega} f(x, w), \quad (15)$$

where $f : \mathbb{R}^n \times \Omega \rightarrow \mathbb{R}$ is a continuously differentiable function, and Ω is a compact subset in \mathbb{R}^m . First, we define $\psi(x) := \max_{w \in \Omega} f(x, w)$. Although ψ is not necessarily Frechet differentiable, it is directional differentiable (even semi-smooth), see [9]. Now, let us check the first-order necessary conditions for problem (15). In fact, if x^* is a local minimizer of (15), then

$$\begin{aligned} \psi'(x^*; x - x^*) &= \max_{w \in \Omega(x^*)} \langle \nabla_x f(x^*, w), x - x^* \rangle \geq 0, \\ \forall x &\in \mathbb{R}_+^n, \end{aligned} \quad (16)$$

which is equivalent to

$$\inf_{x \in \mathbb{R}_+^n} \max_{w \in \Omega(x^*)} \langle \nabla_x f(x^*, w), x - x^* \rangle = 0, \quad (17)$$

where $\Omega(x)$ means the active set at x , that is, $\Omega(x) := \{w \in \Omega \mid \psi(x) = f(x, w)\}$. At our first glance, the formula (17) is not related to SVNCP(F, Ω). Nonetheless, we will show that if Ω is convex and the function $f(x, \cdot)$ is concave over Ω , then the first-order necessary conditions form an SVNCP(F, Ω), see below proposition.

Proposition 1. *Let Ω be nonempty, compact, and convex set in \mathbb{R}^m . Suppose that, for each x , the function $f(x, \cdot)$ is concave over Ω . If x^* is a local optimal solution of (15), then there exists $w^* \in \Omega(x^*)$, such that*

$$x^* \geq 0, \quad \nabla_x f(x^*, w^*) \geq 0, \quad \langle \nabla_x f(x^*, w^*), x^* \rangle = 0. \quad (18)$$

Proof. Note first that for each x the inner problem

$$\psi(x) := \max_{w \in \Omega} f(x, w) \tag{19}$$

is a concave optimization problem, since $f(x, \cdot)$ is concave and Ω is convex. This ensures that $\Omega(x)$, which denotes the optimal solution set of (19), is convex as well. Now we claim that the function

$$h(w) := \langle \nabla_x f(x^*, w), x - x^* \rangle \tag{20}$$

is concave over $\Omega(x^*)$. Indeed, for $w_1, w_2 \in \Omega(x^*)$ and $\alpha \in [0, 1]$, we have

$$\begin{aligned} & h(\alpha w_1 + (1 - \alpha) w_2) \\ &= \langle \nabla_x f(x^*, \alpha w_1 + (1 - \alpha) w_2), x - x^* \rangle \\ &= \lim_{t \downarrow 0} (f(x^* + t(x - x^*), \alpha w_1 + (1 - \alpha) w_2) \\ &\quad - f(x^*, \alpha w_1 + (1 - \alpha) w_2)) (t)^{-1} \\ &= \lim_{t \downarrow 0} \frac{f(x^* + t(x - x^*), \alpha w_1 + (1 - \alpha) w_2) - \psi(x^*)}{t} \\ &\geq \lim_{t \downarrow 0} (\alpha f(x^* + t(x - x^*), w_1) + (1 - \alpha) \\ &\quad \times f(x^* + t(x - x^*), w_2) - \psi(x^*)) (t)^{-1} \\ &= \lim_{t \downarrow 0} \frac{\alpha [f(x^* + t(x - x^*), w_1) - f(x^*, w_1)]}{t} \\ &\quad + \lim_{t \downarrow 0} \frac{(1 - \alpha) [f(x^* + t(x - x^*), w_2) - f(x^*, w_2)]}{t} \\ &= \alpha \langle \nabla_x f(x^*, w_1), x - x^* \rangle + (1 - \alpha) \\ &\quad \times \langle \nabla_x f(x^*, w_2), x - x^* \rangle \\ &= \alpha h(w_1) + (1 - \alpha) h(w_2), \end{aligned} \tag{21}$$

where we use the fact that $\alpha w_1 + (1 - \alpha) w_2 \in \Omega(x^*)$ (since $\Omega(x^*)$ is convex) and $f(x^*, w) = \psi(x^*)$ for all $w \in \Omega(x^*)$. On the other hand, applying the Min-Max Theorem [10, Corollary 37.3.2] to (17) yields

$$\max_{w \in \Omega(x^*)} \inf_{x \in \mathbb{R}_+^n} \langle \nabla_x f(x^*, w), x - x^* \rangle = 0. \tag{22}$$

Hence, for arbitrary $\varepsilon > 0$, we can find $w_\varepsilon \in \Omega(x^*)$, such that

$$\inf_{x \in \mathbb{R}_+^n} \langle \nabla_x f(x^*, w_\varepsilon), x - x^* \rangle \geq -\varepsilon, \tag{23}$$

that is,

$$\langle \nabla_x f(x^*, w_\varepsilon), x - x^* \rangle \geq -\varepsilon, \quad \forall x \in \mathbb{R}_+^n. \tag{24}$$

In particular, plugging in $x = 0$ in (24) implies

$$\langle \nabla_x f(x^*, w_\varepsilon), x^* \rangle \leq \varepsilon. \tag{25}$$

Since Ω is bounded and $\Omega(x^*)$ is closed, we can assume, without loss of generality, that $w_\varepsilon \rightarrow w^* \in \Omega(x^*)$ as $\varepsilon \rightarrow 0$. Thus, taking the limit in (25) gives

$$\langle \nabla_x f(x^*, w^*), x^* \rangle \leq 0. \tag{26}$$

Now, let $x = x^* + ke_i \in \mathbb{R}_+^n$. It follows from (24) that

$$(\nabla_x f(x^*, w_\varepsilon))_i \geq -\frac{\varepsilon}{k}, \tag{27}$$

which implies that $(\nabla_x f(x^*, w_\varepsilon))_i \geq 0$ by letting $k \rightarrow \infty$, and hence $(\nabla_x f(x^*, w^*))_i \geq 0$ for all $i = 1, 2, \dots, n$, that is, $\nabla_x f(x^*, w^*) \geq 0$. This together with (26) means that $\langle \nabla_x f(x^*, w^*), x^* \rangle = 0$. Thus, (18) holds. \square

From all the above, we have seen that $\text{SVNCP}(F, \Omega)$ given as in (2) covers a range of optimization problems. Therefore, in this paper, we mainly focus on $\text{SVNCP}(F, \Omega)$. Due to its equivalence to SVCP (1), our analysis and results for $\text{SVNCP}(F, \Omega)$ can be carried over to SVCP (1). This paper is organized as follows. In Section 1, connection between $\text{SVNCP}(F, \Omega)$ and various optimization problems is introduced. We recall some background materials in Section 2. Besides comparing the set-valued complementarity problems with the classical complementarity problems, we analyze the solution set of SVCP in Section 3. Moreover, properties of merit functions for SVCP are studied in Section 4, such as level bounded and error bound. Finally, some possible research directions are discussed.

A few words about the notations used throughout the paper. For any $x, y \in \mathbb{R}^n$, the inner product is denoted by $x^T y$ or $\langle x, y \rangle$. We write $x \geq y$ (or $x > y$) if $x_i \geq y_i$ (or $x_i > y_i$) for all $i = 1, 2, \dots, n$. Let e be the vector with all components being 1, and let e_i be the i -row of identity matrix. Denote $N_\infty := \cup_{n=1}^\infty \{\{n, n + 1, \dots\}\}$. While $\text{SVNCP}(F, \Omega)$ meaning the set-valued nonlinear complementary problem (2), $\text{SVLCP}(M, q, \Omega)$ denotes the linear case, that is, $F(x, w) = M(w)x + q(w)$, where $M : \mathbb{R}^m \rightarrow \mathbb{R}^{n \times n}$ and $q : \mathbb{R}^m \rightarrow \mathbb{R}^n$. For a continuously differentiable function $F : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^l$, we denote the $l \times n$ Jacobian matrix of partial derivatives of F at (\bar{x}, \bar{w}) with respect to x by $J_x F(\bar{x}, \bar{w})$, whereas the transposed Jacobian is denoted by $\nabla_x F(\bar{x}, \bar{w})$. For a mapping $H : \mathbb{R}^n \rightarrow \mathbb{R}^m$, define

$$\liminf_{x \rightarrow \bar{x}} H(x) := \begin{pmatrix} \liminf_{x \rightarrow \bar{x}} H_1(x) \\ \liminf_{x \rightarrow \bar{x}} H_2(x) \\ \vdots \\ \liminf_{x \rightarrow \bar{x}} H_m(x) \end{pmatrix}. \tag{28}$$

Given a set-valued mapping $M : \mathbb{R}^n \rightrightarrows \mathbb{R}^m$, define

$$\begin{aligned} & \limsup_{x \rightarrow \bar{x}} M(x) \\ &:= \{u \mid \exists x^n \rightarrow \bar{x}, \exists u^n \rightarrow u \text{ with } u^n \in M(x^n)\}, \\ & \liminf_{x \rightarrow \bar{x}} M(x) \\ &:= \{u \mid \forall x^n \rightarrow \bar{x}, \exists u^n \rightarrow u \text{ with } u^n \in M(x^n)\}. \end{aligned} \tag{29}$$

We say that M is outer semicontinuous at \bar{x} , if

$$\limsup_{x \rightarrow \bar{x}} M(x) \subset M(\bar{x}), \tag{30}$$

and inner semicontinuous at \bar{x} if

$$\liminf_{x \rightarrow \bar{x}} M(x) \supset M(\bar{x}). \tag{31}$$

We say that M is continuous at \bar{x} if it is both outer semicontinuous and inner semicontinuous at \bar{x} . For more details about these functions, please refer to [9, 11]. Throughout this paper, we always assume that the set-valued mapping $\Omega : \mathbb{R}^n \rightrightarrows \mathbb{R}^m$ is closed valued; that is, $\Omega(x)$ is closed for all $x \in \mathbb{R}^n$ [11, Chapter 1].

2. Focus on SVLCP(M, q, Ω)

It is well known that various matrix classes play different roles in the theory of linear complementarity problem, such as P -matrix, S -matrix, Q -matrix, and Z -matrix, see [1, 12] for more details. Here we recall some of them which will be needed in the subsequent analysis.

Definition 2. A matrix $M \in \mathbb{R}^{n \times n}$ is said to be an S -matrix if there exists $x \in \mathbb{R}^n$, such that

$$x > 0, \quad Mx > 0. \tag{32}$$

Note that $M \in \mathbb{R}^{n \times n}$ is an S -matrix, if and only if the classical linear complementarity problem $LCP(M, q)$ is feasible for all $q \in \mathbb{R}^n$, see [12, Prop. 3.1.5]. Moreover, the above condition in Definition 2 is equivalent to

$$x \geq 0, \quad Mx > 0, \tag{33}$$

see [13, Remark 2.2]. However, such equivalence fails to hold for its corresponding cases in set-valued complementarity problem. In other words,

$$x > 0, \quad M(w)x > 0, \quad \text{for some } w \in \Omega(x) \tag{34}$$

is not equivalent to

$$x \geq 0, \quad M(w)x > 0, \quad \text{for some } w \in \Omega(x). \tag{35}$$

It is clear that (34) implies (35). But, the converse implication does not hold, which is illustrated in Example 3.

Example 3. Let

$$M(w) = \begin{pmatrix} w & 0 \\ 0 & w \end{pmatrix}, \quad \Omega(x) = \begin{cases} \{0, 1\}, & x = (1, 0) \in \mathbb{R}^2; \\ \{0\}, & \text{otherwise.} \end{cases} \tag{36}$$

If $M(w)x > 0$, then $w = 1$, and such case holds only when $x = (1, 0)$. Therefore, (35) is satisfied, but (34) is not.

We point out that the set-valued mapping $\Omega(x)$ in Example 3 is indeed outer semi-continuous. A natural question arises: *what happens if $\Omega(x)$ is inner semi-continuous?* The answer is given in Theorem 4 as below.

Theorem 4. *If $\Omega(x)$ is inner semicontinuous, and $M(w)$ is continuous, then (34) and (35) are equivalent.*

Proof. We only need to show (35) \Rightarrow (34). Let $H(x) = \max_{w \in \Omega(x)} M(w)x$, and denote by $a_i(x)$ the i th row of $M(w)$. Hence $H_i(x) = \max_{w \in \Omega(x)} a_i(x)^T x$. With this, suppose x_0 is an arbitrary but fixed point, we know that for any $\varepsilon > 0$, there exists $w_0 \in \Omega(x_0)$ such that $a_i(w_0)^T x_0 > H_i(x_0) - \varepsilon$. Since $\Omega(x)$ is inner semi-continuous, for any $x_n \rightarrow x_0$, there exists $w_n \in \Omega(x_n)$ satisfying $w_n \rightarrow w_0$. This implies

$$H_i(x_n) = \max_{w \in \Omega(x_n)} a_i(w)^T x_n \geq a_i(w_n)^T x_n. \tag{37}$$

Then, taking the lower limit yields

$$\liminf_{n \rightarrow \infty} H_i(x_n) \geq \lim_{n \rightarrow \infty} a_i(w_n)^T x_n = a_i(w_0)^T x_0 > H_i(x_0) - \varepsilon, \tag{38}$$

where the equality follows from the continuity of $a_i(w)$, which is ensured by the continuity of $M(w)$. Because $\varepsilon > 0$ is arbitrary, and $\{x_n\}$ is an arbitrary sequence converging to x_0 , we obtain

$$\liminf_{x \rightarrow x_0} H_i(x) \geq H_i(x_0), \tag{39}$$

which says H_i is lower semi-continuous. This further implies

$$\begin{aligned} \liminf_{x \rightarrow x_0} H(x) &= \begin{pmatrix} \liminf_{x \rightarrow x_0} H_1(x) \\ \vdots \\ \liminf_{x \rightarrow x_0} H_n(x) \end{pmatrix} \\ &\geq \begin{pmatrix} H_1(x_0) \\ \vdots \\ H_n(x_0) \end{pmatrix} = H(x_0), \end{aligned} \tag{40}$$

that is,

$$\liminf_{x \rightarrow x_0} \max_{w \in \Omega(x)} M(w)x \geq \max_{w \in \Omega(x_0)} M(w)x_0. \tag{41}$$

If \bar{x} satisfies (35), then

$$\bar{x} \geq 0, \quad M(\bar{w})\bar{x} > 0, \quad \text{for some } \bar{w} \in \Omega(\bar{x}) \tag{42}$$

which is equivalent to

$$\bar{x} \geq 0, \quad H(\bar{x}) > 0. \tag{43}$$

On the other hand, $\liminf_{\lambda \rightarrow 0^+} H(\bar{x} + \lambda e) \geq H(\bar{x}) > 0$, and $\bar{x} + \lambda e > 0$ for $\lambda > 0$. By taking $\lambda > 0$ small enough, we know $\bar{x} + \lambda e$ satisfies (34). Thus, the proof is complete. \square

There is another point worthy of pointing out. We mentioned that the classical linear complementarity problem $LCP(M, q)$ is feasible for all $q \in \mathbb{R}^n$ if and only if $M \in \mathbb{R}^{n \times n}$ is a S -matrix; that is, there exists $x \in \mathbb{R}^n$, such that

$$x > 0, \quad Mx > 0. \tag{44}$$

Is there an analogous result in the set-valued set? Yes, we have an answer for it in Theorem 5 below.

Theorem 5. Consider the set-valued linear complementarity problem $SVLCP(M, q, \Omega)$. If there exists $x \in \mathbb{R}^n$, such that

$$x \geq 0, \quad M(w)x > 0, \quad \text{for some } w \in \bigcap_{\bar{N} \in N_\infty} \bigcup_{n \in \bar{N}} \Omega(nx), \quad (45)$$

then $SVLCP(M, q, \Omega)$ is feasible for all $q : \mathbb{R}^m \rightarrow \mathbb{R}^n$ being bounded from below.

Proof. Let q be any mapping from \mathbb{R}^m to \mathbb{R}^n being bounded from below, that is, there exists $\beta \in \mathbb{R}$, such that $q(w) \geq \beta e$. Suppose that x_0 and w_0 satisfy (45), which means

$$x_0 \geq 0, \quad M(w_0)x_0 > 0, \quad w_0 \in \bigcap_{\bar{N} \in N_\infty} \bigcup_{n \in \bar{N}} \Omega(nx_0). \quad (46)$$

Then, for any $\bar{N} \in N_\infty$, we have $w_0 \in \bigcup_{n \in \bar{N}} \Omega(nx_0)$. In particular, we observe the following:

- (1) if taking $\bar{N} = \{1, 2, \dots\}$, then there exists n_1 , such that $w_0 \in \Omega(n_1x_0)$;
- (2) if taking $\bar{N} = \{n_1 + 1, \dots\}$, then there exists n_2 with $n_2 > n_1$, such that $w_0 \in \Omega(n_2x_0)$.

Repeating the above process yields a sequence $\{n_k\}$, such that $w_0 \in \Omega(n_kx_0)$ and $n_k \rightarrow \infty$. Since $M(w_0)x_0 > 0$, it ensures the existence of $\alpha > 0$, such that $M(w_0)x_0 > \alpha e$. Taking k large enough to satisfy $n_k > \max\{-\beta/\alpha, 0\}$ gives $\alpha n_k e > -\beta e \geq -q(w)$. Then, it implies that

$$M(w_0)n_kx_0 > \alpha n_k e \geq -q(w), \quad (47)$$

and hence

$$n_kx_0 \geq 0, \quad M(w_0)(n_kx_0) + q(w) > 0, \quad w_0 \in \Omega(n_kx_0), \quad (48)$$

which says n_kx_0 is a feasible point of $SVLCP(M, q, \Omega)$. \square

Definition 6. A matrix $M \in \mathbb{R}^{n \times n}$ is said to be a P -matrix if all its principal minors are positive, or equivalently [12, Theorem 3.3.4],

$$\forall x \neq 0, \quad \exists k \in \{1, 2, \dots, n\} \text{ such that } x_k(Mx)_k > 0. \quad (49)$$

From [12, Corollary 3.3.5], we know that every P -matrix is an S -matrix. In other words, if M satisfies (49), then the following system is solvable:

$$x \geq 0, \quad Mx > 0. \quad (50)$$

Their respective corresponding conditions in set-valued complementarity problem are

$$\forall x \neq 0, \quad \exists k \in \{1, \dots, n\} \text{ such that } x_k(M(w)x)_k > 0, \quad \text{for some } w \in \Omega(x), \quad (51)$$

$$x \geq 0, \quad M(w)x > \quad \text{for some } w \in \Omega(x). \quad (52)$$

Example 7 shows that the aforementioned implication is not valid as well in set-valued complementarity problem.

Example 7. Let

$$M(w) = \begin{pmatrix} w & 0 \\ 0 & -w \end{pmatrix}, \quad \Omega(x) = \begin{cases} -1, & x_1 = 0; \\ 1, & \text{otherwise.} \end{cases} \quad (53)$$

For $x_1 \neq 0$, we have $M(w)x = (x_1, -x_2)$, and hence $x_1(M(w)x)_1 = x_1^2 > 0$. For $x_1 = 0$, we know $x_2 \neq 0$ (since $x \neq 0$) which says $M(w)x = (-x_1, x_2)$, and hence $x_2(M(w)x)_2 = x_2^2 > 0$. Therefore, condition (51) is satisfied. But, condition (52) fails to hold because $M(w)x = (x_1, -x_2)$ or $(-x_1, x_2)$. Hence, $M(w)x > 0$ implies that $x_2 < 0$ or $x_1 < 0$, which contradicts with $x \geq 0$.

Definition 8. A matrix $M \in \mathbb{R}^{n \times n}$ is said to be semimonotone if

$$\forall 0 \neq x \geq 0 \implies \exists x_k > 0 \text{ such that } (Mx)_k \geq 0. \quad (54)$$

For the classical linear complementarity problem, we know that M is semimonotone if and only if $LCP(M, q)$ with $q > 0$ has a unique solution (zero solution), see [12, Theorem 3.9.3]. One may wonder whether such fact still holds in set-valued case. Before answering it, we need to know how to generalize concept of semi-monotonicity to its corresponding definition in the set-valued case.

Definition 9. The set of matrices $\{M(w) \mid w \in \Omega(x)\}$ is said to be

(a) strongly semimonotone if for any nonzero $x \geq 0$, $\exists x_k > 0$ such that $(M(w)x)_k \geq 0 \quad \forall w \in \Omega(x); \quad (55)$

(b) weakly semimonotone if for any nonzero $x \geq 0$, $\exists x_k > 0$ such that $(M(w)x)_k \geq 0$ for some $w \in \Omega(x). \quad (56)$

Unlike the classical linear complementarity problem case, here are parallel results regarding set-valued linear complementarity problem which strong (weak) semi-monotonicity plays in.

Theorem 10. For the $SVLCP(M, q, \Omega)$, the following statements hold.

- (a) If the set of matrices $\{M(w) \mid w \in \Omega(x)\}$ is strongly semi-monotone, then for any positive mapping q , that is, $q(w) > 0$ for all w , $SVLCP(M, q, \Omega)$ has zero as its unique solution.
- (b) If $SVLCP(M, q, \Omega)$ with $q(w) > 0$ has zero as its unique solution, then the set of matrices $\{M(w) \mid w \in \Omega(x)\}$ is weakly semi-monotone.

Proof. (a) It is clear that, for any positive mapping q , $x = 0$ is a solution of $SVLCP(M, q, \Omega)$. Suppose there is another nonzero solution \bar{x} , that is, $\exists \bar{w} \in \Omega(\bar{x})$, such that

$$\bar{x} \geq 0, \quad M(\bar{w})\bar{x} + q(\bar{w}) \geq 0, \quad \bar{x}^T (M(\bar{w})\bar{x} + q(\bar{w})) = 0. \quad (57)$$

It follows from (55) that there exists $k \in \{1, 2, \dots, n\}$, such that $\bar{x}_k > 0$ and $(M(\bar{w})\bar{x})_k \geq 0$, and hence $(M(\bar{w})\bar{x} + q(\bar{w}))_k > 0$, which contradicts condition (57).

(b) Suppose $\{M(w) \mid w \in \Omega(x)\}$ is not weakly semi-monotone. Then, there exists a nonzero $\bar{x} \geq 0$, for all $k \in I^+(\bar{x}) := \{i \mid \bar{x}_i > 0\}$, $(M(w)\bar{x})_k < 0$ for all $w \in \Omega(\bar{x})$. Choose $\bar{w} \in \Omega(\bar{x})$. Let $q(w) = 1$ for all $w \neq \bar{w}$ and

$$q_k(\bar{w}) = \begin{cases} -(M(\bar{w})\bar{x})_k, & k \in I^+(\bar{x}); \\ \max\{(M(\bar{w})\bar{x})_k, 0\} + 1, & \text{otherwise.} \end{cases} \quad (58)$$

Therefore, $q(w) > 0$ for all w . According to the above construction, we have

$$\begin{aligned} \bar{x} \geq 0, \quad M(\bar{w})\bar{x} + q(\bar{w}) \geq 0, \quad \bar{x}^T(M(\bar{w})\bar{x} + q(\bar{w})) = 0, \\ \text{with } \bar{w} \in \Omega(\bar{x}), \end{aligned} \quad (59)$$

that is, the nonzero vector \bar{x} is a solution of $\text{SVLCP}(M, q, \Omega)$, which is a contradiction. \square

Theorem 10(b) says that the weak semi-monotonicity is a necessary condition for zero being the unique solution of $\text{SVLCP}(M, q, \Omega)$. However, it is not the sufficient condition, see Example 11.

Example 11. Let

$$M(w) = \begin{pmatrix} -w & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix}, \quad \Omega(x) = \{0, 1\}. \quad (60)$$

For any nonzero $x = (x_1, x_2, x_3) \geq 0$, we have $M(0)x = (x_2, x_3, x_1) \geq 0$. If we plug in $q = (1, 1, 1)$, by a simple calculation, $x = (1, 0, 0)$ satisfies

$$x \geq 0, \quad M(1)x + q \geq 0, \quad x^T(M(1) + q) = 0 \quad (61)$$

which means that $\text{SVLCP}(M, q, \Omega)$ has a nonzero solution. We also notice that the set-valued mapping $\Omega(x)$ is even continuous in Example 11.

So far, we have seen some major difference between the classical complementarity problem and set-valued complementarity problem. Such phenomenon undoubtedly confirms that it is an interesting, important, and challenging task to study the set-valued complementarity problem, which, to some extent, is the main motivation of this paper.

To close this section, we introduce some other concepts which will be used later too. A function $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is level bounded, if the level set $\{x \mid f(x) \leq \alpha\}$ is bounded for all $\alpha \in \mathbb{R}$. The metric projection of x to a closed convex subset $A \subset \mathbb{R}^n$ is denoted by $\Pi_A(x)$, that is, $\Pi_A(x) := \arg \min_{y \in A} \|x - y\|$. The distance function is defined as $\text{dist}(x, A) := \|x - \Pi_A(x)\|$.

3. Properties of Solution Sets

Recently, many authors study other classes of complementarity problems, in which another type of vector $w \in \Omega$ is

involved, for example, the *stochastic complementarity problem* [14–17], to find $x \in \mathbb{R}^n$, such that

$$x \geq 0, \quad F(x, w) \geq 0, \quad x^T F(x, w) = 0, \quad \text{a.e. } w \in \Omega, \quad (62)$$

where w is a random vector in a given probability space and the *semi-infinite complementarity problem* [18] to find $x \in \mathbb{R}^n$, such that

$$x \geq 0, \quad F(x, w) \geq 0, \quad x^T F(x, w) = 0, \quad \forall w \in \Omega, \quad (63)$$

which we denote it by $\text{SINCP}(F, \Omega)$. In addition, the authors introduce the following two complementarity problems in [18] to find $x \in \mathbb{R}^n$, such that

$$\begin{aligned} x \geq 0, \quad F_{\min}(x) \geq 0, \quad x^T F_{\min}(x) = 0, \\ x \geq 0, \quad F_{\max}(x) \geq 0, \quad x^T F_{\max}(x) = 0, \end{aligned} \quad (64)$$

where

$$F_{\min}(x) := \begin{pmatrix} \min_{w \in \Omega} F_1(x, w) \\ \vdots \\ \min_{w \in \Omega} F_n(x, w) \end{pmatrix}, \quad (65)$$

$$F_{\max}(x) := \begin{pmatrix} \max_{w \in \Omega} F_1(x, w) \\ \vdots \\ \max_{w \in \Omega} F_n(x, w) \end{pmatrix}. \quad (66)$$

These two problems are denoted by $\text{NCP}(F_{\min})$ and $\text{NCP}(F_{\max})$, respectively. Is there any relationship among their solutions sets? In order to further describing such relationship, we adapt the following notations:

- (i) $\text{SOL}(F, \Omega)$ means the solution set of $\text{SVNCP}(F, \Omega)$,
- (ii) $\text{SOL}(M, q, \Omega)$ means the solution set of $\text{SVLCP}(F, \Omega)$,
- (iii) $\widehat{\text{SOL}}(F, \Omega)$ means the solution set of $\text{SINCP}(F, \Omega)$,
- (iv) $\text{SOL}(F_{\min})$ means the solution set of $\text{NCP}(F_{\min})$,
- (v) $\text{SOL}(F_{\max})$ means the solution set of $\text{NCP}(F_{\max})$.

Besides, for the purpose of comparison, we restrict that $\Omega(x)$ is fixed; that is, there exists a subset Ω in \mathbb{R}^m , such that $\Omega(x) = \Omega$ for all $x \in \mathbb{R}^n$.

It is easy to see that the solution set of $\text{SINCP}(F, \Omega)$ is $\bigcap_{w \in \Omega} \text{SOL}(F_w)$, but that of $\text{SVNCP}(f, \Omega)$ is $\bigcup_{w \in \Omega} \text{SOL}(F_w)$, where $F_w(x) := F(x, w)$. Hence, the solution set of $\text{SINCP}(F, \Omega)$ is included in that of $\text{SVNCP}(F, \Omega)$. In other words, we have

$$\widehat{\text{SOL}}(F, \Omega) \subseteq \text{SOL}(F, \Omega). \quad (67)$$

The inclusion (67) can be strict as shown in Example 12.

Example 12. Let $F(x, w) = (w, 1)$ and $\Omega(x) = [0, 1]$. Then, we can verify that $\widehat{\text{SOL}}(F, \Omega) = \{0, 0\}$, whereas $\text{SOL}(F, \Omega) = \{x \mid x_1 \geq 0, x_2 = 0\}$.

However, the solution set of $SVNCP(F, \Omega)$, $NCP(F_{\min})$, and $NCP(F_{\max})$ are not included each other. This is illustrated in Examples 13–14.

Example 13. $SOL(F_{\min}) \not\subseteq SOL(F, \Omega)$ and $SOL(F, \Omega) \not\subseteq SOL(F_{\min})$.

- (a) Let $F(x, w) = (1 - w, w)$ and $\Omega = [0, 1]$. Then, $SOL(F_{\min}) = \mathbb{R}_+^2$, $SOL(F, \Omega) = \bigcup_{w \in \Omega} SOL(F_w) = \{(x_1, x_2)^T \mid x_1 \geq 0, x_2 \geq 0, \text{ and } x_1 x_2 = 0\}$.
- (b) Let $F(x, w) = (w - 1, x_2)$ and $\Omega = [0, 1]$. Then, $SOL(F_{\min}) = \emptyset$ and $SOL(F, \Omega) = \{(x_1, x_2) \mid x_1 \geq 0, x_2 = 0\}$.

Example 14. $SOL(F_{\max}) \not\subseteq SOL(F, \Omega)$ and $SOL(F, \Omega) \not\subseteq SOL(F_{\max})$.

- (a) Let $F(x, w) = (w - 1, -w)$ and $\Omega = [0, 1]$. Then, $SOL(F_{\max}) = \mathbb{R}_+^2$ and $SOL(F, \Omega) = \emptyset$.
- (b) Let $F(x, w) = (w, -w)$ and $\Omega = [0, 1]$. Then, $SOL(F_{\max}) = \{(x_1, x_2) \mid x_1 = 0, x_2 \geq 0\}$ and $SOL(F, \Omega) = \mathbb{R}_+^2$.

Similarly, Example 15 shows that the solution set of $NCP(F_{\max})$ and $NCP(F_{\min})$ cannot be included each other.

Example 15. $SOL(F_{\max}) \not\subseteq SOL(F_{\min})$ and $SOL(F_{\min}) \not\subseteq SOL(F_{\max})$.

- (a) Let $F(x, w) = (w - 1, 0)$ and $\Omega = [0, 1]$. Then, $SOL(F_{\min}) = \emptyset$ and $SOL(F_{\max}) = \mathbb{R}_+^2$.
- (b) Let $F(x, w) = (w, w)$ and $\Omega = [0, 1]$. Then, $SOL(F_{\min}) = \mathbb{R}_+^2$ and $SOL(F_{\max}) = \{(0, 0)\}$.

In spite of these, we obtain some results which describe the relationship among them.

Theorem 16. *Let $\Omega(x) = \Omega$ for all $x \in \mathbb{R}^n$. Then, we have*

- (a) $SOL(F, \Omega) \cap \{x \mid F_{\min}(x) \geq 0\} \subseteq SOL(F_{\min})$;
- (b) $SOL(F_{\max}) \cap \{x \mid F(x, w) \geq 0 \text{ for some } w \in \Omega\} \subseteq SOL(F, \Omega)$;
- (c) $SOL(F_{\min}) \cap \{x \mid x^T F_{\max}(x) \leq 0\} = SOL(F_{\max}) \cap \{x \mid F_{\min}(x) \geq 0\} \subseteq SOL(F, \Omega)$.

Proof. Parts (a) and (b) follow immediately from the fact

$$x^T F_{\min}(x) \leq x^T F(x, w) \leq x^T F_{\max}(x) \quad \forall w \in \Omega, x \in \mathbb{R}_+^n. \tag{68}$$

Part (c) is from (67), since the two sets in the left side of (c) is $\overline{SOL(F, \Omega)}$ by [18]. \square

For further characterizing the solution sets, we recall that for a set-valued mapping $M : \mathbb{R}^n \rightrightarrows \mathbb{R}^m$, its inverse mapping (see [9, Chapter 5]) is defined as

$$M^{-1}(y) := \{x \mid y \in M(x)\}. \tag{69}$$

Theorem 17. *For $SVNCP(F, \Omega)$, we have*

$$SOL(F, \Omega) = \bigcup_{w \in \mathbb{R}^m} [SOL(F_w) \cap \Omega^{-1}(w)]. \tag{70}$$

Proof. In fact, the desired result follows from

$$\begin{aligned} SOL(F, \Omega) &= \{x \mid x \in SOL(F_w) \text{ and } w \in \Omega(x) \\ &\quad \text{for some } w \in \mathbb{R}^m\} \\ &= \{x \mid x \in SOL(F_w) \text{ and } x \in \Omega^{-1}(w) \\ &\quad \text{for some } w \in \mathbb{R}^m\} \\ &= \bigcup_{w \in \mathbb{R}^m} [SOL(F_w) \cap \Omega^{-1}(w)], \end{aligned} \tag{71}$$

where the second equality is due to the definition of inverse mapping given as above. \square

4. Merit Functions for SVNCP and SVLCP

It is well known that one of the important approaches for solving the complementarity problems is to transfer it to a system of equations or an unconstrained optimization via NCP functions or merit functions. Hence, we turn our attention in this section to address merit functions for $SVNCP(F, \Omega)$ and $SVLCP(M, q, \Omega)$.

A function $\phi : \mathbb{R}^2 \rightarrow \mathbb{R}$ is called an NCP function if it satisfies

$$\phi(a, b) = 0 \iff a \geq 0, \quad b \geq 0, \quad ab = 0. \tag{72}$$

For example, the natural residual $\phi_{NR}(a, b) = \min\{a, b\}$ and the Fischer-Burmeister function $\phi_{FB}(a, b) = \sqrt{a^2 + b^2} - (a + b)$ are popular NCP-functions. Please also refer to [19] for a detailed survey on the existing NCP-functions. In addition, a real-valued function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is called a merit (or residual) function for a complementarity problem if $f(x) \geq 0$ for all $x \in \mathbb{R}^n$ and $f(x) = 0$ if and only if x is a solution of the complementarity problem. Given an NCP-function ϕ , we define

$$r(x, w) := \|\Phi(x, F(x, w))\| \tag{73}$$

$$\text{where } \Phi(x, y) := (\phi(x_1, y_1), \dots, \phi(x_n, y_n)).$$

Then, it is not hard to verify that the function given by

$$r(x) := \min_{w \in \Omega(x)} r(x, w) \tag{74}$$

is a merit function for $SVNCP(F, \Omega)$. Note that the merit function (74) is rather different from the traditional one, because the index set is not a fixed set, but dependent on x . We say that a merit function $r(x)$ has a global error bound with a modulus $c > 0$ if

$$\text{dist}(x, SOL(F, \Omega)) \leq c \cdot r(x), \quad \forall x \in \mathbb{R}^n. \tag{75}$$

For more information about the error bound, please see [20] which is an excellent survey paper regarding the issue of error bounds.

Theorem 18. Assume that there exists a set $\Omega \subset \mathbb{R}^m$, such that $\Omega(x) = \Omega$ for all $x \in \mathbb{R}^n$, and that for each $w \in \Omega$, $r(x, w)$ is a global error bound of $NCP(F_w)$ with the modulus $\eta(w) > 0$, that is,

$$\text{dist}(x, \text{SOL}(F_w)) \leq \eta(w) r(x, w), \quad \forall x \in \mathbb{R}^n. \quad (76)$$

In addition, if

$$\eta := \max_{w \in \Omega} \eta(w) < +\infty, \quad (77)$$

then $r(x) = \min_{w \in \Omega} r(x, w)$ provides a global error bound for $\text{SVNCP}(F, \Omega)$ with the modulus η .

Proof. Noticing that if $\Omega(x) = \Omega$ for all $x \in \mathbb{R}^n$, then

$$\Omega^{-1}(w) = \begin{cases} \mathbb{R}^n, & w \in \Omega, \\ \emptyset, & w \notin \Omega. \end{cases} \quad (78)$$

It then follows from Theorem 17 that

$$\text{SOL}(F, \Omega) = \bigcup_{w \in \mathbb{R}^m} [\text{SOL}(F_w) \cap \Omega^{-1}(w)] = \bigcup_{w \in \Omega} \text{SOL}(F_w). \quad (79)$$

Therefore,

$$\begin{aligned} & \text{dist}(x, \text{SOL}(F, \Omega)) \\ &= \text{dist}\left(x, \bigcup_{w \in \Omega} \text{SOL}(F_w)\right) \\ &\leq \min_{w \in \Omega} \text{dist}(x, \text{SOL}(F_w)) \\ &\leq \min_{w \in \Omega} \eta(w) \cdot r(x, w) \\ &\leq \min_{w \in \Omega} \max_{w \in \Omega} \eta(w) \cdot r(x, w) \\ &= \max_{w \in \Omega} \eta(w) \min_{w \in \Omega} r(x, w) \\ &= \eta \cdot r(x). \end{aligned} \quad (80)$$

Thus, the proof is complete. \square

One may ask when condition (77) is satisfied? Indeed, the condition (77) is satisfied if

- (i) Ω is a finite set;
- (ii) $F(x, w) = M(w)x + q(w)$ where $M(w)$ is continuous, and for each $w \in \Omega$ the matrix $M(w)$ is a P -matrix. In this case the modulus $\eta(w)$ takes an explicitly formula, that is,

$$\eta(w) = \max_{d \in [0,1]^n} \|(I - D + DM(w))^{-1}\|, \quad (81)$$

see [21, 22]. Hence, we see that

$$\eta = \max_{\substack{d \in [0,1]^n \\ w \in \Omega}} \|(I - D + DM(w))^{-1}\| \quad (82)$$

is well defined because $M(w)$ is continuous, and Ω is compact.

For simplification of notations, we write $x \rightarrow \infty$ instead of $\|x\| \rightarrow \infty$. We now introduce the following definitions which are similar to (29):

$$\begin{aligned} & \limsup_{x \rightarrow \infty} M(x) \\ &:= \{u \mid \exists x^n \rightarrow \infty, \exists u^n \rightarrow u \text{ with } u^n \in M(x^n)\}, \\ & \liminf_{x \rightarrow \infty} M(x) \\ &:= \{u \mid \forall x^n \rightarrow \infty, \exists u^n \rightarrow u \text{ with } u^n \in M(x^n)\}. \end{aligned} \quad (83)$$

Definition 19. For $\text{SVLCP}(M, q, \Omega)$, the set of matrices $\{M(w) \mid w \in \Omega(x)\}$ is said to have the limit- R_0 property if

$$\begin{aligned} & x \geq 0, \quad M(w)x \geq 0, \quad x^T M(w)x = 0 \\ & \text{for some } w \in \limsup_{x \rightarrow \infty} \Omega(x) \implies x = 0. \end{aligned} \quad (84)$$

In the case of a linear complementarity problem, that is, $\Omega(x)$ is a fixed single-point set, Definition 19 coincides with that of R_0 -matrix.

Theorem 20. For $\text{SVLCP}(M, q, \Omega)$, suppose that there exists a bounded set Ω , such that $\Omega(x) \subset \Omega$ for all $x \in \mathbb{R}^n$, and $M(w)$ and $q(w)$ are continuous on Ω . If the set of matrices $\{M(w) \mid w \in \Omega(x)\}$ has the limit- R_0 property, then the merit function $r(x) = \min_{w \in \Omega(x)} \min\{x, M(w)x + q(w)\}$ is level bounded.

Proof. We argue this result by contradiction. Suppose there exists a sequence $\{x_n\}$ satisfying $\|x_n\| \rightarrow \infty$, and $r(x_n)$ is bounded. Then,

$$\begin{aligned} \frac{r(x_n)}{\|x_n\|} &= \min_{w \in \Omega(x_n)} \left\| \min \left\{ \frac{x_n}{\|x_n\|}, M(w) \frac{x_n}{\|x_n\|} + \frac{q(w)}{\|x_n\|} \right\} \right\| \\ &= \left\| \min \left\{ \frac{x_n}{\|x_n\|}, M(w_n) \frac{x_n}{\|x_n\|} + \frac{q(w_n)}{\|x_n\|} \right\} \right\|, \end{aligned} \quad (85)$$

where we assume the minimizer is attained at $w_n \in \Omega(x_n)$, whose existence is ensured by the compactness of $\Omega(x_n)$, since $\Omega(x)$ is closed and Ω is bounded. Taking a subsequence if necessary, we can assume that $\{x_n/\|x_n\|\}$ and $\{w_n\}$ are both convergent in which \bar{x} and \bar{w} represent their corresponding limit point. Thus, we have

$$\bar{w} \in \limsup_{n \rightarrow \infty} \Omega(x_n) \subset \limsup_{\|x\| \rightarrow \infty} \Omega(x). \quad (86)$$

Now, taking the limit in (85) yields

$$\|\min\{\bar{x}, M(\bar{w})\bar{x}\}\| = 0, \quad (87)$$

where we have used the fact that $q(w_n)/\|x_n\| \rightarrow 0$, because q is continuous, and $w_n \in \Omega$ is bounded. This contradicts (84) since \bar{x} is a nonzero vector. \square

Note that the condition (84) is equivalent to

$$\bigcup_{\substack{w \in \limsup_{x \rightarrow \infty} \Omega(x)}} \text{SOL}(M(w)) = \{0\}, \tag{88}$$

which is also equivalent to saying that each matrix $M(w)$ for $w \in \limsup_{x \rightarrow \infty} \Omega(x)$ is a R_0 -matrix.

Theorem 21. *For SVLCP(M, q, Ω), suppose that there exists a compact set Ω , such that $\Omega(x) \subset \Omega$ for all $x \in \mathbb{R}^n$, and $M(w)$ and $q(w)$ are continuous on Ω . If $r(x) = \min_{w \in \Omega(x)} \|\min\{x, M(w)x + q(w)\}\|$ is level bounded, then the following implication holds*

$$\begin{aligned} x \geq 0, \quad M(w)x \geq 0, \quad x^T M(w)x = 0 \\ \text{for some } w \in \bigcap_{\tilde{N} \in N_\infty} \bigcup_{n \in \tilde{N}} \Omega(nx) \implies x = 0. \end{aligned} \tag{89}$$

Proof. Suppose that there exist a nonzero vector x_0 , and $w_0 \in \bigcap_{\tilde{N} \in N_\infty} \bigcup_{n \in \tilde{N}} \Omega(nx_0)$, such that

$$x_0 \geq 0, \quad M(w_0)x_0 \geq 0, \quad x_0^T M(w_0)x_0 = 0. \tag{90}$$

Similar to the argument as in Theorem 5, there exists a sequence $\{n_k\}$ with $n_k \rightarrow \infty$ and $w_0 \in \Omega(n_k x_0)$. Hence,

$$\begin{aligned} r(n_k x_0) &= \min_{w \in \Omega(n_k x_0)} \|\min\{n_k x_0, n_k M(w)x_0 + q(w)\}\| \\ &\leq \|\min\{n_k x_0, n_k M(w_0)x_0 + q(w_0)\}\| \\ &\leq \sum_{i=1}^n \|\min\{n_k(x_0)_i, n_k(M(w_0)x_0)_i + q(w_0)_i\}\|. \end{aligned} \tag{91}$$

Next, we proceed the arguments by discussing the following two cases.

Case 1. For $(x_0)_i > 0$, we have $(M(w_0)x_0)_i = 0$ from (90). Since $\max_{w \in \Omega} q(w)$ is finite due to the compactness of Ω and the continuity of $q(w)$, $n_k(x_0)_i > \max_{w \in \Omega} q(w)$ for k sufficiently large. Therefore, we obtain

$$\|\min\{n_k(x_0)_i, n_k(M(w_0)x_0)_i + q_i(w_0)\}\| = \|q_i(w_0)\|. \tag{92}$$

Case 2. For $(x_0)_i = 0$, by a simple calculation, we have

$$\begin{aligned} &\|\min\{n_k(x_0)_i, n_k(M(w_0)x_0)_i + q_i(w_0)\}\| \\ &\times \begin{cases} = 0, & \text{if } n_k(M(w_0)x_0)_i + q_i(w_0) \geq 0, \\ \leq \|q_i(w_0)\|, & \text{if } n_k(M(w_0)x_0)_i + q_i(w_0) < 0, \end{cases} \end{aligned} \tag{93}$$

where the inequality in the latter case comes from the fact that $q_i(w_0) \leq n_k(M(w_0)x_0)_i + q_i(w_0) < 0$. Thus,

$$r(n_k x_0) \leq \sum_{i=1}^n \|q_i(w_0)\|. \tag{94}$$

This contradicts the level boundedness of $r(x)$ since $n_k x_0 \rightarrow \infty$. \square

The above conclusion is equivalent to say that for each $w \in \bigcap_{\tilde{N} \in N_\infty} \bigcup_{n \in \tilde{N}} \Omega(nx)$, the matrix $M(w)$ is a R_0 -matrix. Finally, let us discuss a special case where the set-valued mapping $\Omega(x)$ has an explicit form, for example, $\Omega(x) = \{w \mid H(x, w) = 0 \text{ and } G(x, w) \geq 0\}$, where $H : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^{l_1}$ and $G : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^{l_2}$. Then, the solution set can be further characterized.

Theorem 22. *If $\Omega(x) := \{w \mid G(x, w) \geq 0, H(x, w) = 0\}$, then*

$$\begin{aligned} \text{SOL}(F, \Omega) &= \bigcup_{w \in \mathbb{R}^m} \left\{ x \mid \begin{pmatrix} x \\ 0 \\ \alpha \end{pmatrix} \in \text{SOL}(\Theta_w) \right. \\ &\quad \left. \text{with } \alpha \in \mathbb{R}_{++}^{l_2} \text{ and } 0 \in \mathbb{R}^{l_1} \right\}, \end{aligned} \tag{95}$$

where $\Theta_w : \mathbb{R}^n \rightarrow \mathbb{R}^{n+l_1+l_2}$ is defined as $\Theta_w(x) := \begin{pmatrix} F(x, w) \\ G(x, w) \\ H(x, w) \end{pmatrix}$ and $\mathbb{R}_{++}^{l_2} := \{\alpha \in \mathbb{R}^{l_2} \mid \alpha_i > 0 \text{ for all } i = 1, \dots, l_2\}$.

Proof. Noting that the problem (2) is to find $w \in \mathbb{R}^m$ and $x \in \mathbb{R}^n$, such that

$$\begin{aligned} x \geq 0, \quad F(x, w) \geq 0, \quad \langle F(x, w), x \rangle = 0, \\ G(x, w) \geq 0, \quad H(x, w) = 0, \end{aligned} \tag{96}$$

namely, to find $w \in \mathbb{R}^m$ and $x \in \mathbb{R}^n$ satisfying

$$\begin{aligned} x \geq 0, \quad F(x, w) \geq 0, \quad \langle F(x, w), x \rangle = 0, \\ 0 \geq 0, \quad G(x, w) \geq 0, \quad 0 \cdot G(x, w) = 0, \\ \alpha > 0, \quad H(x, w) \geq 0, \quad \langle \alpha, H(x, w) \rangle = 0. \end{aligned} \tag{97}$$

In other words,

$$\begin{pmatrix} x \\ 0 \\ \alpha \end{pmatrix} \geq 0, \quad \begin{pmatrix} F(x, w) \\ G(x, w) \\ H(x, w) \end{pmatrix} \geq 0, \quad \begin{pmatrix} x \\ 0 \\ \alpha \end{pmatrix}^T \begin{pmatrix} F(x, w) \\ G(x, w) \\ H(x, w) \end{pmatrix} = 0. \tag{98}$$

Then, the desired result follows. \square

The foregoing result indicates that the set-valued complementarity problem is different from the classical complementarity problem, since it restricts that some components of the solution must be positive or zero, which is not required in the classical complementarity problems.

Moreover, the set-valued complementarity problem can be further reformulated to be an equation, that is, finding $x \in \mathbb{R}^n$ and $w \in \mathbb{R}^m$ to satisfy the following equation

$$\Gamma(x, w) = \begin{pmatrix} \xi(x, w) \\ H(x, w) \\ \text{dist}^2(G(x, w) \mid \mathbb{R}_+^{l_2}) \end{pmatrix} = 0, \tag{99}$$

where $\xi(x, w) = (1/2)\|\Phi_{\text{FB}}(x, F(x, w))\|^2$. Note that when A is a closed convex set, then $\theta(x) := \text{dist}^2(x, A)$ is continuously

differentiable, and $\nabla\theta(x) = 2(x - \Pi_A(x))$. This fact together with $\|\phi_{FB}\|^2$ being continuously differentiable implies the following immediately.

Theorem 23. *Suppose that G and H are continuously differentiable, and ϕ is the Fischer-Burmeister function, then Γ is continuously differentiable and*

$$JT(x, w) = \begin{pmatrix} J_x \xi(x, w) \\ J_x H(x, w) \\ 2 \left(G(x, w) - \prod_{\mathbb{R}_+^2} (G(x, w)) \right)^T J_x G(x, w) \\ J_w \xi(x, w) \\ J_w H(x, w) \\ 2 \left(G(x, w) - \prod_{\mathbb{R}_+^2} (G(x, w)) \right)^T J_w G(x, w) \end{pmatrix}, \tag{100}$$

where

$$\begin{aligned} J_x \xi(x, w) &= \Phi_{FB}(x, F(x, w))^T \\ &\quad \times [\mathcal{D}_a(x, F(x, w)) \\ &\quad \quad + \mathcal{D}_b(x, F(x, w)) J_x F(x, F(x, w))], \\ J_w \xi(x, w) &= \Phi_{FB}(x, F(x, w))^T \mathcal{D}_b(x, F(x, w)) \\ &\quad \times J_w F(x, w). \end{aligned} \tag{101}$$

Here $\mathcal{D}_a(x, F(x, w))$ and $\mathcal{D}_b(x, F(x, w))$ mean the sets of $n \times n$ diagonal matrices $\text{diag}(a_1(x, F(x, w)), \dots, a_n(x, F(x, w)))$ and $\text{diag}(b_1(x, F(x, w)), \dots, b_n(x, F(x, w)))$, respectively, and

$$\begin{aligned} &(a_i(x, F(x, w)), b_i(x, F(x, w))) \\ &\begin{cases} = \frac{(x_i, F_i(x, w))}{\sqrt{x_i^2 + F_i^2(x, w)}} - (1, 1), & \text{if } (x_i, F_i(x, w)) \neq 0, \\ \in \bigcup_{\theta \in [0, 2\pi]} \{\cos \theta, \sin \theta\} - (1, 1), & \text{if } (x_i, F_i(x, w)) = 0. \end{cases} \end{aligned} \tag{102}$$

5. Further Discussions

In this paper, we have paid much attention to the set-valued complementarity problems which possess rather different features from those of classical complementarity problems. As suggested by one referee, we here briefly discuss the relation between stochastic variational inequalities and the set-valued complementarity problems. Given $F : \mathbb{R}^n \times \Xi \rightarrow \mathbb{R}, X_\xi \subset \mathbb{R}^n$ and $\Xi \subset \mathbb{R}^l$, a set representing future states

of knowledge, the *stochastic variational inequalities* is to find $x \in X_\xi$, such that

$$(y - x)^T F(x, \xi) \geq 0, \quad \forall y \in X_\xi, \xi \in \Xi. \tag{103}$$

If $X_\xi = \mathbb{R}_+^n$, then the stochastic variational inequalities reduce to the *stochastic complementarity problem* as follows:

$$x \geq 0, \quad F(x, \xi) \geq 0, \quad x^T F(x, \xi) = 0, \quad \xi \in \Xi. \tag{104}$$

The optimization problem corresponding to stochastic complementarity problem is

$$\min_{x \in \mathbb{R}_+^n} \mathbb{E} \{ \|\Phi(x, F(x, \xi))\| \}. \tag{105}$$

When Ξ is a discrete set, say $\Xi := \{\xi_1, \xi_2, \dots, \xi_v\}$, then

$$\mathbb{E} \{ \|\Phi(x, F(x, \xi))\| \} = \sum_{i=1}^v P(\xi_i) \|\Phi(x, F(x, \xi_i))\|, \tag{106}$$

where $P(\xi_i)$ is the probability of ξ_i . If the optimal value of (105) is zero, then it follows from (106) that (104) coincides with

$$\begin{aligned} x \geq 0, \quad F(x, \xi_i) \geq 0, \quad x^T F(x, \xi_i) = 0, \\ \forall \xi_i \in \Xi \text{ satisfying } P(\xi_i) > 0. \end{aligned} \tag{107}$$

When Ξ is a continuous set, then

$$\mathbb{E} \{ \|\Phi(x, F(x, \xi))\| \} = \int_{\Omega} \|\Phi(x, F(x, \xi_i))\| P(x) dx, \tag{108}$$

where $P(x)$ is the density function. In this case, (104) takes the form of

$$x \geq 0, \quad F(x, \xi) \geq 0, \quad x^T F(x, \xi) = 0, \quad \text{a.e. } \xi \in \Xi, \tag{109}$$

or equivalently there exists a subset $\Xi_0 \subset \Xi$ with $P(\Xi_0) = 0$, such that

$$x \geq 0, \quad F(x, \xi) \geq 0, \quad x^T F(x, \xi) = 0, \quad \forall \xi \in \Xi \setminus \Xi_0. \tag{110}$$

Hence the stochastic complementarity problem is, in certain extent, a semi-infinite complementarity problem (SICP).

Due to some major difference between set-valued complementarity problems and classical complementarity problems, there are still many interesting, important, and challenging questions for further investigation as below, to name a few.

- (i) How to extend other important concepts used in classical linear complementarity problems to set-valued cases (like $P_0, P_*, Z, Q, Q_0, S, \bar{S}$, copositive, column sufficient matrix, etc.)?
- (ii) How to propose an effective algorithm to solve (99)?
- (iii) Can we provide some sufficient conditions to ensure the existence of solutions? One possible direction is to use *fixed-point theory*. In fact, the set-valued complementarity problem is to find $x \in \mathbb{R}^n$, such that

$$x = \max \{0, x - F(x, w)\} = \Pi_{\mathbb{R}_+^n} (x - F(x, w)) \tag{111}$$

for some $w \in \Omega(x)$,

that is,

$$x \in \Pi_{\mathbb{R}_+^n} (x - \tilde{F}(x)), \quad (112)$$

where $\tilde{F}(x) := \bigcup_{w \in \Omega(x)} F(x, w)$. Note that (112) is a fixed point of a set-valued mapping $\Pi_{\mathbb{R}_+^n}(I - \tilde{F})$, where I denotes the identity mapping.

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