Research Article

An Implicit Extragradient Method for Hierarchical Variational Inequalities

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Received 20 September 2010; Accepted 7 November 2010

Academic Editor: Jen Chih Yao

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As a well-known numerical method, the extragradient method solves numerically the variational inequality VI(*C*, *A*) of finding $u \in C$ such that $\langle Au, v-u \rangle \ge 0$, for all $v \in C$. In this paper, we devote to solve the following hierarchical variational inequality HVI(*C*, *A*, *f*) Find $\tilde{x} \in VI(C, A)$ such that $\langle (I - f)\tilde{x}, x - \tilde{x} \rangle \ge 0$, for all $x \in VI(C, A)$. We first suggest and analyze an implicit extragradient method for solving the hierarchical variational inequality HVI(*C*, *A*, *f*). It is shown that the net defined by the suggested implicit extragradient method converges strongly to the unique solution of HVI(*C*, *A*, *f*) in Hilbert spaces. As a special case, we obtain the minimum norm solution of the variational inequality VI(*C*, *A*).

1. Introduction

The variational inequality problem is to find $u \in C$ such that

$$\langle Au, v - u \rangle \ge 0, \quad \forall v \in C.$$
 (1.1)

The set of solutions of the variational inequality problem is denoted by VI(C, A). It is well known that the variational inequality theory has emerged as an important tool in studying a wide class of obstacle, unilateral, and equilibrium problems; which arise in several branches of pure and applied sciences in a unified and general framework. Several numerical methods have been developed for solving variational inequalities and related optimization problems, see [1–24] and the references therein. In particular, Korpelevich's extragradient method which was introduced by Korpelevič [4] in 1976 generates a sequence $\{x_n\}$ via the recursion

$$y_n = P_C[x_n - \lambda A x_n],$$

$$x_{n+1} = P_C[x_n - \lambda A y_n], \quad n \ge 0,$$
(1.2)

where P_C is the metric projection from R^n onto $C, A : C \to H$ is a monotone operator, and λ is a constant. Korpelevich [4] proved that the sequence $\{x_n\}$ converges strongly to a solution of VI(C, A). Note that the setting of the space is Euclid space R^n .

Recently, hierarchical fixed point problems and hierarchical minimization problems have attracted many authors' attention due to their link with some convex programming problems. See [25–32]. Motivated and inspired by these results in the literature, in this paper we are devoted to solve the following hierarchical variational inequality HVI(C, A, f):

Find
$$z \in VI(C, A)$$
 such that $\langle (I - f)z, x^* - z \rangle \ge 0, \quad \forall x^* \in VI(C, A).$ (1.3)

For this purpose, in this paper, we first suggest and analyze an implicit extragradient method. It is shown that the net defined by this implicit extragradient method converges strongly to the unique solution of HVI(C, A, f) in Hilbert spaces. As a special case, we obtain the minimum norm solution of the variational inequality VI(C, A).

2. Preliminaries

Let *H* be a real Hilbert space with inner product $\langle \cdot, \cdot \rangle$ and norm $\|\cdot\|$, and let *C* be a closed convex subset of *H*. Recall that a mapping $A : C \to H$ is called α -inverse strongly monotone if there exists a positive real number α such that

$$\langle Au - Av, u - v \rangle \ge \alpha \|Au - Av\|^2, \quad \forall u, v \in C.$$

$$(2.1)$$

A mapping $f : C \to H$ is said to be ρ -contraction if there exists a constant $\rho \in [0, 1)$ such that

$$||f(x) - f(y)|| \le \rho ||x - y|| \quad \forall x, y \in C.$$
 (2.2)

It is well known that, for any $u \in H$, there exists a unique $u_0 \in C$ such that

$$||u - u_0|| = \inf\{||u - x|| : x \in C\}.$$
(2.3)

We denote u_0 by $P_C u$, where P_C is called the *metric projection* of H onto C. The metric projection P_C of H onto C has the following basic properties:

(i) ||P_Cx - P_Cy|| ≤ ||x - y|| for all x, y ∈ H;
(ii) ⟨x - y, P_Cx - P_Cy⟩ ≥ ||P_Cx - P_Cy||² for every x, y ∈ H;
(iii) ⟨x - P_Cx, y - P_Cx⟩ ≤ 0 for all x ∈ H, y ∈ C;
(iv) ||x - y||² ≥ ||x - P_Cx||² + ||y - P_Cx||² for all x ∈ H, y ∈ C.

Such properties of P_C will be crucial in the proof of our main results. Let A be a monotone mapping of C into H. In the context of the variational inequality problem, it is easy to see from property (iii) that

$$x^* \in \operatorname{VI}(C, A) \Longleftrightarrow x^* = P_C(x^* - \mu A x^*), \quad \forall \mu > 0.$$

$$(2.4)$$

We need the following lemmas for proving our main result.

Lemma 2.1 (see [13]). Let *C* be a nonempty closed convex subset of a real Hilbert space *H*. Let the mapping $A : C \rightarrow H$ be α -inverse strongly monotone, and let $\lambda > 0$ be a constant. Then, one has

$$\|(I - \lambda A)x - (I - \lambda A)y\|^{2} \le \|x - y\|^{2} + \lambda(\lambda - 2\alpha)\|Ax - Ay\|^{2}, \quad \forall x, y \in C.$$
(2.5)

In particular, if $0 \le \lambda \le 2\alpha$, then $I - \lambda A$ is nonexpansive.

Lemma 2.2 (see [32]). Let *C* be a nonempty closed convex subset of a real Hilbert space *H*. Assume that the mapping $F : C \to H$ is monotone and weakly continuous along segments, that is, $F(x + ty) \to F(x)$ weakly as $t \to 0$. Then, the variational inequality

$$x^* \in C, \quad \langle Fx^*, x - x^* \rangle \ge 0, \quad \forall x \in C$$
(2.6)

is equivalent to the dual variational inequality

$$x^* \in C, \quad \langle Fx, x - x^* \rangle \ge 0, \quad \forall x \in C.$$
 (2.7)

3. Main Result

In this section, we will introduce our implicit extragradient algorithm and show its strong convergence to the unique solution of HVI(C, A, f).

Algorithm 1. Let C be a closed convex subset of a real Hilbert space H. Let $A : C \to H$ be an α -inverse strongly monotone mapping. Let $f : C \to H$ be a (nonself) contraction with coefficient $\rho \in [0, 1)$. For any $t \in (0, 1)$, define a net $\{x_t\}$ as follows:

$$y_{t} = P_{C}[tf(x_{t}) + (1-t)(x_{t} - \lambda A x_{t})], x_{t} = P_{C}[y_{t} - \lambda A y_{t}], \quad t \in (0, 1),$$
(3.1)

where $\lambda \in [a, b] \subset (0, 2\alpha)$ is a constant.

Note the fact that f is a possible nonself mapping. Hence, if we take f = 0, then (3.1) reduces to

$$y_t = P_C[(1-t)(x_t - \lambda A x_t)], x_t = P_C[y_t - \lambda A y_t], \quad t \in (0, 1),$$
(3.2)

Remark 3.1. We notice that the net $\{x_t\}$ defined by (3.1) is well defined. In fact, we can define a self-mapping $W_t : C \to C$ as follows:

$$W_t x := P_C [I - \lambda A] P_C [tf(x) + (1 - t)(I - \lambda A)x], \quad \forall x \in C.$$
(3.3)

From Lemma 2.1, we know that if $\lambda \in (0, 2\alpha)$, the mapping $I - \lambda A$ is nonexpansive. For any $x, y \in C$, we have

$$\|W_{t}x - W_{t}y\| = \|P_{C}[I - \lambda A]P_{C}[tf(x) + (1 - t)(I - \lambda A)x] -P_{C}[I - \lambda A]P_{C}[tf(y) + (1 - t)(I - \lambda A)y]\| \leq \|tf(x) - tf(y) + (1 - t)(I - \lambda A)x - (1 - t)(I - \lambda A)y\| \leq t\|f(x) - f(y)\| + (1 - t)\|(I - \lambda A)x - (I - \lambda A)y\| \leq t\rho\|x - y\| + (1 - t)\|x - y\| = [1 - (1 - \rho)t]\|x - y\|.$$
(3.4)

This shows that the mapping W_t is a contraction. By Banach contractive mapping principle, we immediately deduce that the net (3.1) is well defined.

Theorem 3.2. Suppose the solution set Ω of HVI(*C*, *A*, *f*) is nonempty. Then the net {*x*_t} generated by the implicit extragradient method (3.1) converges in norm, as $t \to 0^+$, to the unique solution *z* of the hierarchical variational inequality HVI(*C*, *A*, *f*). In particular, if one takes that *f* = 0, then the net {*x*_t} defined by (3.2) converges in norm, as $t \to 0^+$, to the minimum-norm solution of the variational inequality VI(*C*, *A*).

Proof. Take that $x^* \in \Omega$. Since $x^* \in VI(C, A)$, using the relation (2.4), we have $x^* = P_C[x^* - \mu Ax^*]$, for all $\mu > 0$. In particular, if we take $\mu = \lambda(1 - t)$, we obtain

$$x^* = P_C[x^* - \lambda(1 - t)Ax^*] = P_C[tx^* + (1 - t)(x^* - \lambda Ax^*)], \quad \forall t \in (0, 1).$$
(3.5)

From (3.1), we have

$$\begin{aligned} \|y_t - x^*\| &= \|P_C[tf(x_t) + (1-t)(x_t - \lambda A x_t)] - P_C[tx^* + (1-t)(x^* - \lambda A x^*)]\| \\ &\leq \|t(f(x_t) - x^*) + (1-t)[(x_t - \lambda A x_t) - (x^* - \lambda A x^*)]\| \\ &\leq t\|f(x_t) - f(x^*)\| + t\|f(x^*) - x^*\| + (1-t)\|(I - \lambda A)x_t - (I - \lambda A)x^*\| \\ &\leq t\rho\|x_t - x^*\| + t\|f(x^*) - x^*\| + (1-t)\|x_t - x^*\| \\ &= [1 - (1-\rho)t]\|x_t - x^*\| + t\|f(x^*) - x^*\|. \end{aligned}$$
(3.6)

Noting that $I - \lambda A$ is nonexpansive, thus,

$$\|x_{t} - x^{*}\| = \|P_{C}[y_{t} - \lambda Ay_{t}] - P_{C}[x^{*} - \lambda Ax^{*}]\|$$

$$\leq \|(y_{t} - \lambda Ay_{t}) - (x^{*} - \lambda Ax^{*})\|$$

$$\leq \|y_{t} - x^{*}\|$$

$$\leq [1 - (1 - \rho)t]\|x_{t} - x^{*}\| + t\|f(x^{*}) - x^{*}\|.$$
(3.7)

That is,

$$\|x_t - x^*\| \le \frac{1}{1 - \rho} \|f(x^*) - x^*\|.$$
(3.8)

Therefore, $\{x_t\}$ is bounded and so are $\{f(x_t)\}$, $\{y_t\}$. Since *A* is α -inverse strongly monotone, it is $1/\alpha$ -Lipschitz continuous. Consequently, $\{Ax_t\}$ and $\{Ay_t\}$ are also bounded.

From (3.6),(2.5), and the convexity of the norm, we deduce

$$\|x_{t} - x^{*}\|^{2} \leq \|t(f(x_{t}) - x^{*}) + (1 - t)[(x_{t} - \lambda Ax_{t}) - (x^{*} - \lambda Ax^{*})]\|^{2}$$

$$\leq t\|f(x_{t}) - x^{*}\|^{2} + (1 - t)\|(I - \lambda A)x_{t} - (I - \lambda A)x^{*}\|^{2}$$

$$\leq t\|f(x_{t}) - x^{*}\|^{2} + (1 - t)[\|x_{t} - x^{*}\|^{2} + \lambda(\lambda - 2\alpha)\|Ax_{t} - Ax^{*}\|^{2}]$$

$$\leq t\|f(x_{t}) - x^{*}\|^{2} + \|x_{t} - x^{*}\|^{2} + (1 - t)a(b - 2\alpha)\|Ax_{t} - Ax^{*}\|^{2}.$$
(3.9)

Therefore, we have

$$(1-t)a(2\alpha - b)\|Ax_t - Ax^*\|^2 \le t \|f(x_t) - x^*\|^2.$$
(3.10)

Hence

$$\lim_{t \to 0} \|Ax_t - Ax^*\| = 0.$$
(3.11)

By the property (ii) of the metric projection P_C , we have

$$\begin{aligned} \|y_{t} - x^{*}\|^{2} &= \|P_{C}[tf(x_{t}) + (1-t)(x_{t} - \lambda Ax_{t})] - P_{C}(x^{*} - \lambda Ax^{*})\|^{2} \\ &\leq \langle tf(x_{t}) + (1-t)(x_{t} - \lambda Ax_{t}) - (x^{*} - \lambda Ax^{*}), y_{t} - x^{*} \rangle \\ &= \frac{1}{2} \Big\{ \|(x_{t} - \lambda Ax_{t}) - (x^{*} - \lambda Ax^{*}) - t(I - \lambda A - f)x_{t}\|^{2} + \|y_{t} - x^{*}\|^{2} \\ &- \|(x_{t} - \lambda Ax_{t}) - (x^{*} - \lambda Ax^{*}) - (y_{t} - x^{*}) - t(I - \lambda A - f)x_{t}\|^{2} \Big\} \\ &\leq \frac{1}{2} \Big\{ \|(x_{t} - \lambda Ax_{t}) - (x^{*} - \lambda Ax^{*})\|^{2} + tM + \|y_{t} - x^{*}\|^{2} \\ &- \|(x_{t} - y_{t}) - \lambda (Ax_{t} - Ax^{*}) - t(I - \lambda A - f)x_{t}\|^{2} \Big\} \end{aligned}$$
(3.12)
$$&\leq \frac{1}{2} \Big\{ \|x_{t} - x^{*}\|^{2} + tM + \|y_{t} - x^{*}\|^{2} - \|x_{t} - y_{t}\|^{2} \\ &+ 2\lambda \langle x_{t} - y_{t}, Ax_{t} - Ax^{*} \rangle + 2t \langle (I - \lambda A - f)x_{t}, x_{t} - y_{t} \rangle \\ &- \|\lambda (Ax_{t} - Ax^{*}) + t(I - \lambda A - f)x_{t}\|^{2} \Big\} \\ &\leq \frac{1}{2} \Big\{ \|x_{t} - x^{*}\|^{2} + tM + \|y_{t} - x^{*}\|^{2} - \|x_{t} - y_{t}\|^{2} \\ &+ 2\lambda \|x_{t} - y_{t}\| \|Ax_{t} - Ax^{*}\| + 2t \|(I - \lambda A - f)x_{t}\| \|x_{t} - y_{t}\| \}, \end{aligned}$$

where M > 0 is some appropriate constant. It follows that

$$\|y_{t} - x^{*}\|^{2} \leq \|x_{t} - x^{*}\|^{2} + tM - \|x_{t} - y_{t}\|^{2} + 2\lambda \|x_{t} - y_{t}\| \|Ax_{t} - Ax^{*}\| + 2t\|(I - \lambda A - f)x_{t}\| \|x_{t} - y_{t}\|,$$
(3.13)

and hence (by (3.7))

$$\|x_{t} - x^{*}\|^{2} \leq \|y_{t} - x^{*}\|^{2}$$

$$\leq \|x_{t} - x^{*}\|^{2} + tM - \|x_{t} - y_{t}\|^{2} + 2\lambda \|x_{t} - y_{t}\| \|Ax_{t} - Ax^{*}\|$$

$$+ 2t \|(I - \lambda A - f)x_{t}\| \|x_{t} - y_{t}\|,$$
(3.14)

which implies that

$$\|x_t - y_t\|^2 \le tM + 2\lambda \|x_t - y_t\| \|Ax_t - Ax^*\| + 2t \| (I - \lambda A - f)x_t\| \|x_t - y_t\|.$$
(3.15)

Since $||Ax_t - Ax^*|| \rightarrow 0$, we derive

$$\lim_{t \to 0^+} \|x_t - y_t\| = 0.$$
(3.16)

Next, we show that the net $\{x_t\}$ is relatively norm-compact as $t \to 0^+$. Assume that $\{t_n\} \subset (0,1)$ is such that $t_n \to 0^+$ as $n \to \infty$. Put $x_n := x_{t_n}$ and $y_n := y_{t_n}$.

By the property (ii) of metric projection P_C , we have

$$\begin{aligned} \left\|y_{t}-x^{*}\right\|^{2} &= \left\|P_{C}\left[tf(x_{t})+(1-t)(x_{t}-\lambda A x_{t})\right]-P_{C}\left[tx^{*}+(1-t)(x^{*}-\lambda A x^{*})\right]\right\|^{2} \\ &\leq \left\langle t(f(x_{t})-x^{*})+(1-t)\left[(x_{t}-\lambda A x_{t})-(x^{*}-\lambda A x^{*})\right],y_{t}-x^{*}\right\rangle \\ &\leq t\left\langle x^{*}-f(x_{t}),x^{*}-y_{t}\right\rangle+(1-t)\left\|(x_{t}-\lambda A x_{t})-(x^{*}-\lambda A x^{*})\right\|\left\|y_{t}-z_{0}\right\| \\ &\leq t\left\langle x^{*}-f(x^{*}),x^{*}-y_{t}\right\rangle+t\left\langle f(x^{*})-f(x_{t}),x^{*}-y_{t}\right\rangle+(1-t)\left\|x_{t}-x^{*}\right\|\left\|y_{t}-x^{*}\right\| \\ &\leq t\left\langle x^{*}-f(x^{*}),x^{*}-y_{t}\right\rangle+t\left\|f(x^{*})-f(x_{t})\right\|\left\|x^{*}-y_{t}\right\|+(1-t)\left\|y_{t}-x^{*}\right\|^{2} \\ &\leq t\left\langle x^{*}-f(x^{*}),x^{*}-y_{t}\right\rangle+t\rho\left\|x^{*}-y_{t}\right\|^{2}+(1-t)\left\|y_{t}-x^{*}\right\|^{2} \\ &= \left[1-(1-\rho)t\right]\left\|y_{t}-x^{*}\right\|^{2}+t\left\langle x^{*}-f(x^{*}),x^{*}-y_{t}\right\rangle. \end{aligned}$$

$$(3.17)$$

Hence

$$\|y_t - x^*\|^2 \le \frac{1}{1 - \rho} \langle x^* - f(x^*), x^* - y_t \rangle.$$
(3.18)

Therefore,

$$\|x_t - x^*\|^2 \le \|y_t - x^*\|^2 \le \frac{1}{1 - \rho} \langle x^* - f(x^*), x^* - y_t \rangle.$$
(3.19)

In particular,

$$\|x_n - x^*\|^2 \le \frac{1}{1 - \rho} \langle x^* - f(x^*), x^* - y_n \rangle.$$
(3.20)

Since $\{x_n\}$ is bounded, without loss of generality, we may assume that $\{x_n\}$ converges weakly to a point $z \in C$. Since $||x_t - y_t|| \rightarrow 0$, we have $||x_n - y_n|| \rightarrow 0$. Hence, $\{y_n\}$ also converges weakly to the same point z.

Next we show that $z \in VI(C, A)$. We define a mapping *T* by

$$Tv = \begin{cases} Av + N_C v, & v \in C, \\ \emptyset, & v \notin C. \end{cases}$$
(3.21)

Then *T* is maximal monotone (see [33]). Let $(v, w) \in G(T)$. Since $w - Av \in N_C v$ and $y_n \in C$, we have $\langle v - y_n, w - Av \rangle \ge 0$. On the other hand, from $y_n = P_C[t_n f(x_n) + (1 - t_n)(x_n - \lambda Ax_n)]$, we have

$$\langle v - y_n, y_n - [t_n f(x_n) + (1 - t_n)(x_n - \lambda A x_n)] \rangle \ge 0,$$
 (3.22)

that is,

$$\left\langle v - y_n, \frac{y_n - x_n}{\lambda} + Ax_n + \frac{t_n}{\lambda} (I - \lambda A - f) x_n \right\rangle \ge 0.$$
 (3.23)

Therefore, we have

$$\langle v - y_{n_{i}}, w \rangle \geq \langle v - y_{n_{i}}, Av \rangle$$

$$\geq \langle v - y_{n_{i}}, Av \rangle - \left\langle v - y_{n_{i}}, \frac{y_{n_{i}} - x_{n_{i}}}{\lambda} + Ax_{n_{i}} + \frac{t_{n_{i}}}{\lambda} (I - \lambda A - f) x_{n_{i}} \right\rangle$$

$$= \left\langle v - y_{n_{i}}, Av - Ax_{n_{i}} - \frac{y_{n_{i}} - x_{n_{i}}}{\lambda} - \frac{t_{n_{i}}}{\lambda} (I - \lambda A - f) x_{n_{i}} \right\rangle$$

$$= \left\langle v - y_{n_{i}}, Av - Ay_{n_{i}} \right\rangle + \left\langle v - y_{n_{i}}, Ay_{n_{i}} - Ax_{n_{i}} \right\rangle$$

$$- \left\langle v - y_{n_{i}}, \frac{y_{n_{i}} - x_{n_{i}}}{\lambda} + \frac{t_{n_{i}}}{\lambda} (I - \lambda A - f) x_{n_{i}} \right\rangle$$

$$\geq \left\langle v - y_{n_{i}}, Ay_{n_{i}} - Ax_{n_{i}} \right\rangle - \left\langle v - y_{n_{i}}, \frac{y_{n_{i}} - x_{n_{i}}}{\lambda} + \frac{t_{n_{i}}}{\lambda} (I - \lambda A - f) x_{n_{i}} \right\rangle.$$

$$(3.24)$$

Noting that $t_{n_i} \to 0$, $||y_{n_i} - x_{n_i}|| \to 0$, and A is Lipschitz continuous, we obtain $\langle v - z, w \rangle \ge 0$. Since T is maximal monotone, we have $z \in T^{-1}(0)$ and hence $z \in VI(C, A)$.

Therefore we can substitute x^* for z in (3.20) to get

$$||x_n - z||^2 \le \frac{1}{1 - \rho} \langle z - f(z), z - y_n \rangle.$$
 (3.25)

Consequently, the weak convergence of $\{x_n\}$ and $\{y_n\}$ to z actually implies that $x_n \rightarrow z$ strongly. This has proved the relative norm-compactness of the net $\{x_t\}$ as $t \rightarrow 0^+$.

Now we return to (3.20) and take the limit as $n \to \infty$ to get

$$||z - x^*||^2 \le \frac{1}{1 - \rho} \langle x^* - f(x^*), x^* - z \rangle, \quad x^* \in \operatorname{VI}(C, A).$$
(3.26)

In particular, *z* solves the following VI

$$z \in \operatorname{VI}(C, A), \quad \left\langle \left(I - f\right) x^*, x^* - z\right\rangle \ge 0, \quad x^* \in \operatorname{VI}(C, A), \tag{3.27}$$

or the equivalent dual VI (see Lemma 2.2)

$$z \in \operatorname{VI}(C, A), \quad \langle (I - f)z, x^* - z \rangle \ge 0, \quad x^* \in \operatorname{VI}(C, A).$$
 (3.28)

Therefore, $z = (P_{VI(C,A)}f)z$. That is, z is the unique solution in VI(C, A) of the contraction $P_{VI(C,A)}f$. Clearly this is sufficient to conclude that the entire net $\{x_t\}$ converges in norm to z as $t \to 0^+$.

Finally, if we take that f = 0, then VI (3.28) is reduced to

$$z \in \operatorname{VI}(C, A), \quad \langle (z, x^* - z) \rangle \ge 0, \quad x^* \in \operatorname{VI}(C, A).$$
(3.29)

Equivalently,

$$||z||^2 \le \langle z, x^* \rangle, \quad x^* \in \operatorname{VI}(C, A). \tag{3.30}$$

This clearly implies that

$$||z|| \le ||x^*||, \quad x^* \in \operatorname{VI}(C, A).$$
 (3.31)

Therefore, z is the minimum-norm solution of VI(C, A). This completes the proof.

Remark 3.3. (1) Note that our Implicit Extragradient Algorithms (3.1) and (3.2) have strong convergence in an infinite dimensional Hilbert space.

(2) In many problems, it is needed to find a solution with minimum norm; see [34–38]. Our Algorithm (3.2) solves the minimum norm solution of VI(C, A).

Acknowledgments

The authors thank the referees for their comments and suggestions which improved the presentation of this paper. The first author was supported in part by Colleges and Universities, Science and Technology Development Foundation (20091003) of Tianjin and NSFC 11071279. The second author was supported in part by NSC 99-2221-E-230-006

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