

A non-interior-point algorithm predictor-corrector algorithm for Variational Inequality Problem with equality linear constraints

Abstract. We present a noninterior-point predictor-corrector algorithm for variational inequality Problem with equality linear constraints based on Chen-Kanzow-Smale smoothing techniques. This method is based upon a modified predictor-corrector interior-point algorithm. It is established the global linear convergence.

Streszczenie. W artykule przedstawiono metodę punktów nie-wewnętrznych predyktorkorektor, uwzględniający nierówności wariacyjne z liniowym ograniczeniem liniowości. Jego działanie oparto na technice Chen-Kanzow-Smale oraz zmodyfikowanej metodzie punktów wewnętrznych predyktorkorektor. Analizie poddano liniową zbieżność algorytmu. (**Metoda punktów nie-wewnętrznych predyktorkorektor w rozwiązywaniu nierówności wariacyjnych z liniowym ograniczeniem równościowym.**)

Keywords: Variational inequality problem; Predictor-corrector; Non-interior-point algorithm; Equality linear constraints.

Słowa kluczowe: nierówność wariacyjna, predyktorkorektor, metoda punktu nie wewnętrznego, liniowe ograniczenie równościowe.

Introduction

The finite-dimensional variational inequality (VI) provides a broad unifying setting for the study of optimization and equilibrium problems and serves as the main computational framework for the practical solution of a host of continuum problems in the mathematical sciences. Extensive studies on theoretical aspects and computational of VI have been done. For a detailed review, we refer to the monographs[1-4].

In this paper, we consider the following variational inequalities problem(VIP):

Problem1.1 Find $x \in K$ such that

$$(1) \quad (t - x)^T f(x) \geq 0 \quad \forall t \in K$$

Where $K = \{Ax=b, x \in R^n\}$, $A \in R^{m \times n}$ and $R(A) = m$, $f(x) : R^n \rightarrow R^n$ is a vector-valued functions and is ξ -monotone on R^n .

The KKT system of the problem1.1 can be cased as NCPs as follows
Problem1.2

$$(2) \quad \begin{cases} Ax=b \\ f(x)-A^T y-z=0 \\ x^T z=0 \end{cases}$$

Note $S = \{(x, y, z) | Ax=b, f(x)-A^T y-z=0, x \geq 0, z \geq 0\}$, we assume the set $S^+ = \{(x, y, z) | (x, y, z) \in S, x > 0, z > 0\}$ is nonempty. The path to be followed is the central path

$$C = \{(x, y, z) | Ax=b, f(x)-A^T y-z=0, Xz=\mu^2 e, x>0, z>0, \mu>0\}$$

We denote by $e \in R^n$ the vector each of whose components is 1 and by X the diagonal matrix whose diagonal entries are given by the vector $x \in R^n$. The algorithm is based on Chen--Harker---Kanzow---Smale smoothing techniques [5-7] and as such relies on the function

$$(3) \quad \phi(a, b, \mu) = a + b - \sqrt{(a-b)^2 + 4\mu^2}$$

This function is a member of the Chen—Mangasarian class of smoothing functions for the problem LCP(q, M). It is easily verified that for $\mu \geq 0$:

$$(4) \quad \phi(a, b, \mu) = 0 \text{ if and only if } a \geq 0, b \geq 0, ab = \mu^2$$

We develop a predictor-corrector path following algorithm for solving Problem1.2, based on the idea which is first non-interior predictor—corrector strategy proposed for following the central path. The central idea is to apply Newton's method to equations of the form $F(x, y, z, \mu) = v$ for appropriate values for v where the function $F : R^n \times R^m \times R^n \times R_{++} \rightarrow R^m \times R^n \times R^n \times R_{++}$ is given by:

$$(5) \quad F(x, y, z, \mu) = \begin{bmatrix} Ax-b \\ f(x)-A^T y-z \\ \phi(x, z, \mu) \\ \mu \end{bmatrix}$$

with

$$(6) \quad \phi(x, z, \mu) := \begin{bmatrix} \phi(x_1, z_1, \mu) \\ \vdots \\ \phi(x_n, z_n, \mu) \end{bmatrix}$$

Note that $F(x, y, z, \mu) = 0$ if and only if x solves Problem1.2. We take the set:

$$(7) \quad N(\beta) = \left\{ (x, y, z) \mid \begin{array}{l} Ax-b=0, f(x)-A^T y-z=0 \\ \phi(x, z, \mu) \leq 0, \|\phi(x, z, \mu)\| \leq \beta \mu \text{ for some } \mu > 0 \end{array} \right\}$$

as our neighborhood for the central path, where $\beta > 0$ is given. For $\mu > 0$, this neighborhood can be viewed as the union of the slices

$$(8) \quad N(\beta, \mu) = \left\{ (x, y, z) \mid \begin{array}{l} Ax-b=0, f(x)-A^T y-z=0 \\ \phi(x, z, \mu) \leq 0, \|\phi(x, z, \mu)\| \leq \beta \mu \end{array} \right\}$$

A Predictor-Corrector Algorithm

In this section, we state our predictor—corrector algorithm and show that it is well-defined. The algorithm:

Step0 (Initialization) Choose $(x^0, y^0, z^0) \in R^n \times R^m \times R^n$,

set $k = 0$, $Ax^0 = b$, $\nabla f(x^0)^T x^0 - A^T y^0 - z^0 = 0$ and let

$\mu_0 > 0$ be such that $\phi(x^0, y^0, z^0, \mu_0) < 0$. choose β so that

$\|\phi(x^0, y^0, z^0, \mu_0)\| \leq \beta \mu_0$. we now have

$(x^0, y^0, z^0) \in N(\beta, \mu_0)$. choose γ, α_1 and α_2 from (0,1).

Step1(the predictor step) Let $(\Delta x^k, \Delta y^k, \Delta z^k)$ solve the equation:

$$(9) \quad F(x^k, y^k, z^k, \mu_k) + \nabla F(x^k, y^k, z^k, \mu_k)^T \begin{bmatrix} \Delta x^k \\ \Delta y^k \\ \Delta z^k \\ \Delta \mu_k \end{bmatrix} = 0$$

If $\|\phi(x^k + \Delta x^k, z^k + \Delta z^k, 0)\| = 0$, then stop and

$(x^k + \Delta x^k, y^k + \Delta y^k, z^k + \Delta z^k)$ solves Problem1.1.

else let τ_k be the first element in the sequence $\{1, \alpha_1, \alpha_1^2, \dots\}$ such that

$$(10) \quad \|\phi(x^k + \tau_k \Delta x^k, z^k + \tau_k \Delta z^k, \tau_k \mu_k)\| \leq \tau_k \beta \mu_k$$

and set $\hat{x} := x^k + \tau_k \Delta x^k$, $\hat{y}^k = y^k + \tau_k \Delta y^k$,

$$(11) \quad \hat{z}^k = z^k + \tau_k \Delta z^k, \hat{\mu}_k = \tau_k \mu_k$$

Step2 (the corrector step) Let $(\Delta \hat{x}^k, \Delta \hat{y}^k, \Delta \hat{z}^k, \Delta \hat{\mu}_k)$ solve the equation

$$(12) \quad F(\hat{x}^k, \hat{y}^k, \hat{z}^k, \hat{\mu}_k) + \nabla F(\hat{x}^k, \hat{y}^k, \hat{z}^k, \hat{\mu}_k)^T \begin{bmatrix} \Delta \hat{x}^k \\ \Delta \hat{y}^k \\ \Delta \hat{z}^k \\ \Delta \hat{\mu}_k \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ (1-\gamma) \hat{\mu}_k \end{bmatrix}$$

and let λ_k be the maximum of the values $1, \alpha_2, \alpha_2^2, \dots$ such that

$$(13) \quad \|\phi(\hat{x}^k + \lambda_k \Delta x^k, \hat{z}^k + \lambda_k \Delta z^k, (1-\gamma \lambda_k) \hat{\mu}_k)\| \leq (1-\gamma \lambda_k) \beta \hat{\mu}_k$$

Set $x^{k+1} = \hat{x}^k + \lambda_k \Delta x^k$, $y^{k+1} = \hat{y}^k + \lambda_k \Delta y^k$,

$$z^{k+1} = \hat{z}^k + \lambda_k \Delta z^k, \mu_{k+1} = (1-\gamma \lambda_k) \hat{\mu}_k \quad (14)$$

and set $k=k+1$ return to step 1.

Theorem1 nonlinear system of equations (9) and (12) has a unique solution, i.e. the matrix

$$\nabla F(x, y, z, \mu)^T = \begin{bmatrix} A & 0 & 0 & 0 \\ \nabla f(x)^T & -A^T & -I & 0 \\ \nabla_x \phi(x, y, z, \mu) & 0 & \nabla_z \phi(x, y, z, \mu) & \nabla_\mu \phi(x, y, z, \mu) \\ 0 & 0 & 0 & I \end{bmatrix}$$

where

$$\nabla_x \phi(x, z, \mu) = \text{diag} \left(\frac{\partial \phi(a, b, \mu)}{\partial a} \Big|_{\substack{a=x \\ b=z \\ \mu=\mu}} \right) = \text{diag} \left(1 - \frac{x_i - z_i}{\sqrt{(x_i - z_i)^2 + 4\mu^2}} \right),$$

$$\nabla_z \phi(x, z, \mu) = \text{diag} \left(\frac{\partial \phi(a, b, \mu)}{\partial b} \Big|_{\substack{a=x \\ b=z \\ \mu=\mu}} \right), \quad \nabla_\mu \phi(x, z, \mu) = \text{diag} \left(\frac{\partial \phi(a, b, \mu)}{\partial \mu} \Big|_{\substack{a=x \\ b=z \\ \mu=\mu}} \right).$$

Convergence Analysis

We are now ready to establish the global linear convergence of the algorithm.

Theorem 2 (global linear convergence) Given $\beta > 0$ and $\mu_0 > 0$ We assume that $\|\nabla F(x, y, z)(\hat{x}, \hat{y}, \hat{z}, \mu)^{-1}\| \leq C$ for all

$0 < \mu \leq \mu_0$ and $(\hat{x}^k, \hat{y}^k, \hat{z}^k) \in N(\beta, \mu_k)$. Let (x^k, y^k, z^k, μ_k) be the sequence generated by the algorithm.

If the algorithm does not terminate finitely at the unique solution to Problem1.2, then for $k = 0, 1, \dots$,

$$(15) \quad (x^k, y^k, z^k) \in N(\beta, \mu_k)$$

$$(16) \quad (1-\gamma \hat{\lambda}_{k-1}) \tau_{k-1} \dots (1-\gamma \hat{\lambda}_0) \tau_0 \mu_0 = \mu_k$$

with

$$(17) \quad \lambda_k \geq \bar{\lambda} = \min \left\{ 1, \frac{(1-\gamma)\beta}{C^2 (\beta + 2\sqrt{n}\gamma)^2 + \sqrt{n}\gamma^2 + \gamma(1-\gamma)\beta} \right\}$$

where C is the constant. Therefore μ_k converges to 0 at a global linear rate. In addition, the sequence (x^k, y^k, z^k) converges to the unique solution of Problem1.2.

Proof. Since $A(x^k + \Delta x^k) = Ax^k + A\Delta x = b$

$$\begin{aligned} & \nabla f(x^k)^T (x^k + \Delta x^k) - A^T (y^k + \Delta y^k) - (z^k + \Delta z^k) \\ &= \nabla f(x^k)^T x^k + -A^T y^k - z^k + \left(\nabla f(x^k)^T \Delta x^k - A^T \Delta y^k - \Delta z^k \right) = 0 \end{aligned}$$

First note that the componentwise concavity of ϕ implies that for any $(x, z) \in R^{2n}$ with $\mu > 0$, and $(\Delta x, \Delta z) \in R^{2n}$ one has

$$\phi(x + \Delta x, z + \Delta z) \leq \phi(x, z) + \nabla \phi(x, z)^T \begin{bmatrix} \Delta x \\ \Delta z \end{bmatrix}$$

Look at the Newton equations (9) and (12). In (13), we have $\Delta \hat{\mu}_k = -(1-\gamma \lambda_k) \hat{\mu}_k$, so (23) reducing to the system

$$\begin{cases} A \Delta \hat{x}^k = 0 \\ \nabla f(\hat{x}^k)^T \Delta \hat{x}^k - A^T \Delta \hat{y}^k - \Delta \hat{z}^k = 0 \\ \nabla_x \phi(\hat{x}^k, \hat{z}^k, \hat{\mu}_k)^T \Delta x + \nabla_z \phi(\hat{x}^k, \hat{z}^k, \hat{\mu}_k)^T \Delta z \\ = -\phi(\hat{x}^k, \hat{z}^k, \hat{\mu}_k) + (1-\lambda_k \gamma) \hat{\mu}_k \nabla_\mu \phi(\hat{x}^k, \hat{z}^k, \hat{\mu}_k)^T \end{cases}$$

In either case, $\phi(x^{k+1}, z^{k+1}, \mu_{k+1}) \leq 0$. For sake of simplicity, set $(x, y, z, \mu) = (\hat{x}, \hat{y}, \hat{z}, \hat{\mu})$ and $(\Delta x, \Delta y, \Delta z) = (\Delta \hat{x}, \Delta \hat{y}, \Delta \hat{z})$, then for $i \in \{1, 2, \dots, n\}$ and $\lambda \in [0, 1]$, then,

$$\begin{aligned} & \|\phi(x + \lambda \Delta x, z + \lambda \Delta z, (1-\lambda)\mu)\| \\ & \leq (1-\lambda) \|\phi(x, z, \mu)\| + \frac{\lambda^2}{(1-\lambda)\mu} \|(\Delta x, \Delta z, -\gamma\mu)\|^2 \\ & \leq (1-\lambda) \|\phi(x, z, \mu)\| + \frac{\lambda^2}{(1-\lambda)\mu} \left[\|(\Delta x, \Delta z)\|^2 + \sqrt{n}\gamma^2 \mu^2 \right] \\ & \leq (1-\lambda) \|\phi(x, z, \mu)\| + \frac{\lambda^2}{(1-\lambda)\mu} \left[\|(\Delta x, \Delta y, \Delta z)\|^2 + \sqrt{n}\gamma^2 \mu^2 \right] \\ & \leq (1-\lambda) \beta \mu + \frac{\lambda^2}{(1-\lambda)} \left[C^2 (\beta + 2\sqrt{n}\gamma)^2 + \sqrt{n}\gamma^2 \right] \mu \end{aligned}$$

Where the last inequality follows from (17) and part (iv) of Lemma 2 in [11]. This yields the bound

$$\begin{aligned} & \|(\Delta x, \Delta y, \Delta z)\| \leq \|\nabla_{(x, y, z)} F(x, y, z, \mu)\| \\ & \left[\|\phi(x, z, \mu)\| + \gamma\mu \|\nabla_\mu \phi(x, z, \mu)\| \right] \leq C (\beta + 2\sqrt{n}\gamma) \mu \end{aligned}$$

It is easily verified that

$$(1-\lambda) \beta \mu + \frac{\lambda^2}{(1-\lambda)} \left[C^2 (\beta + 2\sqrt{n}\gamma)^2 + \sqrt{n}\gamma^2 \right] \mu \leq (1-\lambda) \beta \mu$$

Whenever

$$\lambda \leq \frac{(1-\gamma)\beta}{C^2 (\beta + 2\sqrt{n}\gamma)^2 + \sqrt{n}\gamma^2 + \gamma(1-\gamma)\beta}$$

Therefore

$$\lambda_k \geq \min \left\{ 1, \frac{(1-\gamma)\beta}{C^2 (\beta + 2\sqrt{n}\gamma)^2 + \sqrt{n}\gamma^2 + \gamma(1-\gamma)\beta} \right\}$$

To conclude, just as in (15), the relations (16) and (17) combined with part (iv) of Lemma 2 in [11] yield the bounds

$$\begin{aligned} & \left\| (\Delta x^k, \Delta y^k, \Delta z^k) \right\| \leq C (\beta + 2\sqrt{n}) \mu_k \\ & \text{and} \end{aligned}$$

$$\left\| (\Delta \hat{x}^k, \Delta \hat{y}^k, \Delta \hat{z}^k) \right\| \leq C (\beta + 2\sqrt{n}) \mu_k$$

Since $0 < \gamma < 1$ and $0 < \lambda_k \leq 1$ for all k , therefore,

$$\begin{aligned} & \left\| (x^{k+1}, y^{k+1}, z^{k+1}) - (x^k, y^k, z^k) \right\| \leq \left\| (\Delta x^k, \Delta y^k, \Delta z^k) \right\| + \\ & \lambda_k \left\| (\Delta \hat{x}^k, \Delta \hat{y}^k, \Delta \hat{z}^k) \right\| \leq 2C (\beta + 2\sqrt{n}) \mu_k \leq 2C (\beta + 2\sqrt{n}) (1-\bar{\lambda})^k \mu_0 \end{aligned}$$

Hence, (x^k, y^k, z^k) is a Cauchy sequence and so must converge to the unique solution of Problem 1.2.

Acknowledgments

This work was supported by the National Natural Science Foundation of China (Grant No. 70771079).

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