

A Direct Algorithm for the Vertical Generalized Complementarity Problem Associated with *P*-Matrices

Aniekan Ebiefung¹, George Habetler², Michael Kostreva³, Bohdan Szanc⁴

¹Department of Mathematics, University of Tennessee, Chattanooga, USA ²Department of Mathematical Sciences Rensselaer Polytechnic Institute, Troy, USA ³Department of Mathematical Sciences Clemson University, Clemson, USA ⁴Department of Mathematics Maryville University, St. Louis, USA Email: aniekan-ebiefung@utc.edu

How to cite this paper: Ebiefung, A., Habetler, G., Kostreva, M. and Szanc, B. (2017) A Direct Algorithm for the Vertical Generalized Complementarity Problem Associated with *P*-Matrices. *Open Journal of Optimization*, **6**, 101-114.

https://doi.org/10.4236/ojop.2017.63008

Received: July 12, 2017 **Accepted:** August 13, 2017 **Published:** August 16, 2017

Copyright © 2017 by authors and Scientific Research Publishing Inc. This work is licensed under the Creative Commons Attribution International License (CC BY 4.0).

http://creativecommons.org/licenses/by/4.0/

Abstract

We present a direct algorithm for solving the vertical generalized linear complementarity problem, first considered by Cottle and Dantzig, when the associated matrix is a vertical block *P*-matrix. The algorithm converges to a unique solution in a finite number of steps, without an assumption of nondegeneracy on the given problem. The algorithm is simple, efficient, and easy to implement.

Keywords

Complementarity Problems, *P*-Matrix, Direct Algorithms, Linear Programming, Bi-Matrix Game

1. Introduction

The linear complementarity problem for any $n \times n$ matrix M is defined as follows:

For any given $q \in \mathbb{R}^n$, find $w, z, \in \mathbb{R}^n$ such that w = Mz + q (1) $w \ge 0, z \ge 0$ $w^t z = 0$ $(w_i z_i = 0, i = 1, \dots, n)$

The study of linear complementarity problems (LCPs) began in the 1960s. Linear programming, quadratic programming, bi-matrix games, as well as certain problems in economics and engineering, can be represented as LCPs. Murty [1] presented an algorithm that finds a unique solution to (1) in a finite number of steps, if the matrix M associated with the problem is an $n \times n$ P-matrix.

The vertical generalized linear complementarity problem for an $m \times n, m \ge n$, vertical block matrix N of type (m_1, m_2, \dots, m_n) is:

For any given $q \in \mathbb{R}^m$, find $w \in \mathbb{R}^m$, $z \in \mathbb{R}^n$ such that

$$w = Nz + q$$

$$w \ge 0, z \ge 0$$

$$z_{j} \prod_{i=1}^{m_{j}} w_{i}^{j} = 0 \quad (j = 1, \dots, n)$$
(2)

We will denote this problem by VGLCP(q, N). For a horizontal generalization of the LCP, see the paper by Chakraborty and Ebiefung [2].

In 1970 Cottle and Dantzig published the first paper to describe the VGLCP [3]. They showed that if N is a strictly positive (or copositive plus) vertical block matrix or a P-matrix, the VGLCP(q, N) has a solution, and also introduced a technique for solving the VGLCP(q, N), when N is a copositive plus vertical block matrix. Their technique could be considered as an extension of Lemke's algorithm for the LCP (with covering vector \mathbf{e}) [4].

The fact that the VGLCP(q, N) has a unique solution when N is a vertical block P-matrix was established by Habetler and Szanc [5], while existence of solutions in terms of representative submatrices was developed by Ebiefung [6]. The set of Q-matrices for the VGLCP(q, N) was characterized by Ebiefung [7], and by Ebiefung and Kostreva [8]. In [8], an algorithm for solving the GLCP for any vector $q \in \mathbb{R}^m$, irrespective of the matrix class, is also provided. As expected, the method they provided is complicated and expensive to implement. For this reason, specialized algorithms are needed when properties of the associated matrices can be exploited to have simpler algorithms. One of these specialized algorithms for the VGLCP is given in Ebiefung, Kostreva, and Ramanujam [9], where the associated matrix is a vertical block Z-matrix.

Applications of Vertical Complementarity Problems are becoming more prevalent. One engineering application is that of Mixed Lubrication which was discussed in the paper of Oh [10]. Calculations were made on a journal bearing with elastic support to illustrate the method of solution over a wide range of conditions. Regions of solid-to-solid contact, hydrodynamic lubrication, and cavitation were observed. The solutions were obtained using a version of the direct algorithm presented here. No proof of convergence was given, and the solution of the generalized complementarity problem is contained in one short paragraph.

In the area of economics, the VGLCP has been applied to the generalization of Leontief's production model and the choice of technology by Ebiefung and Kostreva [11]; and to the determination of equilibrium points in multi-unit manufacturing systems, Ebiefung and Kostreva [12]. Other economic applica-

tions can be found in Ebiefung, Kostreva, and Majumdar [13], Ebiefung and Isaac [14], Ebiefung [15], and in the references at the end of this paper.

In this paper, we modify Murty's direct algorithm to solve the LCP when the associated matrix N is a vertical block P-matrix of type (m_1, m_2, \dots, m_n) . We will show that the new algorithm finds a non-negative solution to problem VGLCP(q, N) in a finite number of steps, where $q \in \mathbb{R}^m$ and N is a vertical block P-matrix of type (m_1, m_2, \dots, m_n) .

The theory needed to prove finite convergence of the direct algorithm is given in detail in [5], and will be covered briefly here. Finite convergence of the algorithm is essential. As we pointed out before, Cottle and Dantzig's algorithm is an extension of Lemke's algorithm which could cycle (or loop) as pointed out by Kostreva [16]. In fact, the example given by Kostreva can be easily modified to show that their algorithm can cycle. Such behavior in a computation would cause a failure in any implementation. Thus, another approach is motivated. It should be noted that the computer routine of Ravindran [17] does cycle when applied to the example of a 3×3 *P*-matrix given in [16].

The rest of the paper is organized as follows. In Section 2, we give notation and definitions that are needed for the rest of the paper. Section 3 is devoted to the description of the new algorithm and proof of convergence. We summarize our results in Section 4.

2. Notation and Definitions

The following notation and definitions are needed for the development of the algorithm.

Definition 1. Let N be an $m \times n$ rectangular matrix with $m \ge n$. We call N a vertical block matrix of type (m_1, \dots, m_n) if N can be particular row-wise as

$$N = \begin{bmatrix} N^1 \\ \vdots \\ N^n \end{bmatrix}$$

where the jth block, N^{j} , is of dimension $m_{j} \times n$ and $m = \sum_{j=1}^{n} m_{j}$. The vectors $w \in \mathbb{R}^{m}$ and $q \in \mathbb{R}^{m}$ are also partitioned to conform to the entries in the block, N^{j} of N:

$$w = \begin{bmatrix} w^{1} \\ \vdots \\ w^{n} \end{bmatrix}, q = \begin{bmatrix} q^{1} \\ \vdots \\ q^{n} \end{bmatrix}$$

where w^{j} and q^{j} are $m_{i} \times 1$ column vectors.

Associated with problem (2) is the related problem: given $q \in \mathbb{R}^m$, find $w \in \mathbb{R}^m, z \in \mathbb{R}^n$ such that

$$w - Nz = q$$

$$z_j \prod_{i=1}^{m_j} w_i^j = 0 \quad (j = 1, \cdots, n)$$
(3)

Definition 2. By a $m \times k$ horizontal matrix A of type (m_1, m_2, \dots, m_n) , we shall mean a matrix

$$A = \left(A^1, \cdots, A^n\right)$$

where the jth block of A, A^{j} , is a $(m \times m_{j})$ matrix and $\sum_{j=1}^{n} m_{j} = k$.

The linear system in (3) can be re-written in the form

$$(I,-N)\binom{w}{z} = q$$

where *I* is the $m \times m$ identity matrix. We shall represent *I* as a $m \times m$ horizontal block matrix by

$$I = \left(I^1, \cdots, I^n\right)$$

where I^{j} is the jth block of columns associated with the variable w_{i}^{j} , $(i=1,\dots,m_{j})$. Thus I^{j} is an $m \times m_{j}$ matrix and $\sum_{j=1}^{n} m_{j} = m$. The column vector I_{i}^{j} is associated with the variable w_{i}^{j} and the column vector $-N_{j}$ is associated with the variable z_{i} .

Definition 3. Let N be an $m \times n, (m \ge n)$, vertical block matrix of type (m_1, \dots, m_n) . Let I be an $m \times m$ horizontal block identity matrix of type (m_1, \dots, m_n) . Define B as the $m \times m$ horizontal block matrix of type (m_1, \dots, m_n) given by

$$B = \left(B^1, \cdots, B^n\right)$$

such that the ith column of the jth block of columns, B_{i}^{j} , is either I_{i} or $-N_{j}$, and if for any $k \in \{1, \dots, m_{j}\}, (j = 1, \dots, n)$, we have

$$B_{.k}^{j} = -N_{.j}$$

then for that specific *j*

$$B_{i}^{j} = I_{i}^{j}$$

for all $i \in \{\{1, \dots, m_j\} - \{k\}\}$. We call *B* a basic matrix. The vertical block matrix *N* of type (m_1, \dots, m_n) is said to be nondegenerate if and only if for each such basic matrix *B*, taking all possible combinations, *B* is nonsingular.

Definition 4. The vector $q \in \mathbb{R}^m$ in system (2) is nondegenerate with respect to N if for each $z \in \mathbb{R}^n$ at most *n* of the (m+n) variables (w, z) are zero.

Definition 5. For each j, $(j = 1, 2, \dots, n)$, the variables $(w_1^j, \dots, w_{m_j}^j, z_j)^t$ constitute the jth ordered related $(m_j + 1)$ -tuple. The set of $m_j + 1$ columns

 $\{I_{.1}, \dots, I_{.m_j}, -N_{.j}\}$ will be known as the jth ordered related set of column vectors in system (2).

Definition 6. A related basic matrix associated with system (3) is the $m \times m$ horizontal block matrix B of type (m_1, \dots, m_n) defined in Definition 3. The set of variables corresponding to the jth block of columns of B, B^j , are called the jth related set of basic variables. The variables that are excluded from the jth set of basic variables are called nonbasic. There are $\prod_{i=1}^{n} (m_j + 1)$ possible related

basic matrices associated with system (3) and we shall denote these basic related matrices by $B_{\lambda}, \lambda \in \{1, \dots, \prod_{j=1}^{n} (m_j + 1)\}$.

 B_1 will always denote the $m \times m$ horizontal block identity matrix. This is equivalent to requiring that z_j be the nonbasic variable for $\lambda = 1, j = 1, \dots, n$. We now consider the solutions corresponding to the $\prod_{j=1}^{n} (m_j + 1)$ related basic matrices associated with the system (3).

If a solution exists to the following system

$$B_{\lambda}r_{\lambda} = q, \quad \lambda \in \left\{1, \cdots, \prod_{j=1}^{n} \left(m_{j}+1\right)\right\}$$

we will call the solution a basic related point, and denote it by r_{λ} . The vector $r_{\lambda} \in \mathbb{R}^{m}$ is subdivided in accordance with B_{λ} , *i.e.* $r_{\lambda} = (r_{\lambda}^{1}, \dots, r_{\lambda}^{n})^{t}$ and $r_{\lambda}^{j} \in \mathbb{R}^{m_{j}}, (j = 1, \dots, n)$.

Definition 7. A basic related point is called degenerate if at least one of its components is zero.

Consider the related problem (3). To satisfy the related condition,

$$z_{\lambda,j}\prod_{i=1}^{m_j}w_{\lambda,i}^j=0$$

at least one component of the jth ordered related $(m_j + 1)$ -tuple, s_{λ}^j , must be assigned the value zero.

We call this component the related nonbasic variable associated with s_{λ}^{j} , and we will denote it by $t_{\lambda,j}$. If each s_{λ} is an ordered related vector, *i.e.* satisfies the related condition for each $\lambda \in (1, \dots, \prod_{j=1}^{n} (m_j + 1))$, we must have for each $j = 1, \dots, n$ and some $i \in \{1, \dots, m_j\}$

$$t_{\lambda,j} = s_{\lambda,i}^j = w_{\lambda,i}^j = 0$$

or

$$t_{\lambda,j} = s_{\lambda,m_j+1}^j = z_{\lambda,i} = 0$$

We denote the related nonbasic vector associated with λ by

$$t_{\lambda} = \left(t_{\lambda,1}, \cdots, t_{\lambda,n}\right)$$

The components of s_{λ}^{j} that are left if we eliminate $t_{\lambda,j}$ are called the related basic variable of s_{λ}^{j} and denoted by r_{λ}^{j} . The jth ordered block of related columns associated with s_{λ}^{j} is denoted by:

$$C^{j} = \left(I_{.1}^{j}, \cdots, I_{.m_{i}}^{j}\right)$$

For each λ and each $j \in \{1, \dots, n\}$, we denote the columns associated with r_{λ}^{j} by B_{λ}^{j} . If we eliminate the columns of B_{λ}^{j} from the columns of C^{j} , we are left with the column that is considered nonbasic. We will denote this column by $D_{\lambda,i}$, and this column will be associated with the nonbasic variable $t_{\lambda,i}$.

Using this notation, we can rewrite Equations (2) as follows: For any given $q \in \mathbb{R}^m$, find $\lambda \in \{1, \dots, \prod_{i=1}^n (m_i + 1)\}$ such that

$$B_{\lambda}r_{\lambda} = q, \ r_{\lambda} \ge 0, t_{\lambda} = 0 \tag{4}$$

If B_{λ} is nonsingular, we can find an explicit expression for r_{λ} and a solution to Equation (4) would have to satisfy

$$r_{\lambda} = B_{\lambda}^{-1}q \ge 0, t_{\lambda} = 0$$

3. The Algorithm

The algorithm that we prepose is described as follows:

Step 1: If $q \ge 0$, then $B_1 = I$, $r_1 = w = q \ge 0$, $t_1 = z = 0$ is the related solution to Equations (2). Terminate.

Step 2: Suppose $q \ge 0$. Start the scheme by picking *w* as the initial related basic vector, r_1 . Then $B_1 = I$, $r_1 = w = q$, $t_1 = z = 0$. Go to Step 3.

Step 3: All the matrices obtained during the scheme will be basic related matrices, $B_{\lambda}, \lambda \in \{1, \dots, \prod_{j=1}^{n} (m_j + 1)\}$. In a general stage of the scheme, suppose that $B_{\overline{\lambda}}$ is the present basic related matrix. If $r_{\overline{\lambda}} \ge 0, t_{\overline{\lambda}} = 0$, we are done. The basic related vector $r_{\overline{\lambda}}$ represents the basic related variables and $t_{\overline{\lambda}}$ represents the nonbasic related variables of the solution of Equations (2). Terminate.

Step 4: If $r_{\overline{\lambda}} \geq 0$ or $t_{\overline{\lambda}} \neq 0$, find

$$k = \min\left\{j : r_{\bar{\lambda},i}^{j} < 0, i \in \{1, \cdots, m_{j}\}, j = 1, \cdots, n\right\}$$
$$l = \min\left\{i : r_{\bar{\lambda},i}^{k} < 0, i \in \{1, \cdots, m_{k}\}\right\}$$

Interchange the related basic variable represented by $r_{\bar{\lambda},l}^k$ and the related nonbasic variable represented by $t_{\bar{\lambda},k}$. This is equivalent to interchanging the columns $B_{\bar{\lambda},l}^k$ and $D_{\bar{\lambda},k}$. After rearranging the columns of the result matrix, if necessary, to conform with Definition 3, denote the resulting related basic matrix by $B_{\bar{\lambda}+1}$, that is, increase $\bar{\lambda}$ by 1. Continue the scheme as above until there exits a related basic matrix $B_{\mu}, \mu \geq \bar{\lambda} + 1$, such that

$$B_{\mu}r_{\mu} = q$$
$$r_{\mu} \ge 0, \quad t_{\mu} = 0$$

Definition 8. If N is a vertical block matrix of type (m_1, \dots, m_n) , then H is defined to be the $(m-m_n) \times (n-1)$ submatrix of type (m_1, \dots, m_{n-1}) if H is the matrix that results from eliminating the nth block and the nth column of N.

Definition 9. Let

$$\Psi = \left(w_1^1, \dots, w_{m_1}^1, \dots, w_1^{n-1}, \dots, w_{m_{n-1}}^{n-1}\right)^t$$
$$\xi = \left(z_1, \dots, z_{n-1}\right)^t$$
$$\gamma = \left(q_1^1, \dots, q_{m_1}^1, \dots, q_1^{n-1}, \dots, q_{m_{n-1}}^{n-1}\right)^t$$

The generalized vertical linear complementarity problem for the $(m-m_n) \times (n-1)$ leading submatrix *H* is: Given $\gamma \in \mathbb{R}^{m-m_n}$, find $\psi \in \mathbb{R}^{m-m_n}, \xi \in \mathbb{R}^{n-1}$ such that

(1)
$$\psi - H\xi = \gamma$$

(2) $\psi \ge 0, \xi \ge 0$
(3) $\xi_j \prod_{i=1}^{m_j} \psi_i^j = 0, j = 1, \dots, n-1$

Problem (5) is called the leading subproblem of Equations (2).

The related basic matrices B_{λ} associated with (5) are the

 $(m-m_n) \times (m-m_n)$ matrices given by Definition 2. The associated related basic and nonbasic vectors are denoted by r_{λ} and ρ_{λ} , respectively.

Lemma 1 Suppose (\hat{w}, \hat{z}) is a solution of Equations (2) and that $\hat{z}_n = 0$. Let

$$\begin{split} \hat{\psi} &= \left(\hat{w}_{1}^{1}, \cdots, \hat{w}_{m_{1}}^{1}, \cdots, \hat{w}_{1}^{n-1}, \cdots, \hat{w}_{m_{n}-1}^{n-1}\right)^{t} \\ &\hat{\xi} &= \left(\hat{z}_{1}, \cdots, \hat{z}_{n-1}\right)^{t} \end{split}$$

Then $(\hat{\psi}, \hat{\xi})$ is a solution of problem (5). **Proof:** Since $\hat{z}_n = 0$, we have that

$$\hat{\nu}^{j} = \hat{w}^{j} = N^{j}\hat{z} + q^{j} = H^{j}\hat{\xi} + \gamma^{j}, \ j = 1, \cdots, n-1$$

The positivity and satisfied related condition for $(\hat{\psi}, \hat{\xi})$ follows from the definition of $(\hat{\psi}, \hat{\xi})$.

Lemma 2 If
$$(\overline{\psi}, \overline{\xi})$$
 is a solution to problem (5), let $\overline{z} = (\overline{\xi}, 0)^t$. Suppose
 $\overline{w}^n = N^n \overline{z} + q^n \ge 0$

Then $(\overline{w}, \overline{z})$ is a solution to Equations (2), where $\overline{w} = (\overline{\psi}, \overline{w}^n)$.

Lemma 3 If N is a vertical block matrix of type (m_1, \dots, m_n) , then H is vertical block $(m-m_n) \times (n-1)$ P-matrix of type (m_1, \dots, m_{n-1}) and the leading subproblem has a unique solution.

Proof: Any representative submatrix of H is a leading submatrix of a representative submatrix of N. Every leading submatrix of N is an $(n-1)\times(n-1)$ *P*-matrix.

Therefore, *H* is a vertical block $(m-m_n) \times (n-1)$ P-matrix of type (m_1, \dots, m_{n-1}) and therefore, a unique solution exists for Equations (5).

Lemma 4 Let N be a vertical block P-matrix of type (m_1, \dots, m_n) and let (\hat{w}, \hat{z}) be the unique solution of Equation (2). Suppose $\hat{z}_n = 0$. For some $\lambda \in \{1, \dots, \prod_{j=1}^{n-1} (m_j + 1)\}$, let

$$\rho = \left(\rho_{\lambda,1}^{1}, \cdots, \rho_{\lambda,m_{1}}^{1}, \cdots, \rho_{\lambda,1}^{n-1}, \cdots, \rho_{\lambda,m_{n-1}}^{n-1}\right) = \left(\rho_{\lambda}^{1}, \cdots, \rho_{\lambda}^{n-1}\right)^{t}$$
$$r_{\lambda} = \left(r_{\lambda,1}, \cdots, r_{\lambda,n-1}\right)^{t}$$

be the related nonbasic and basic vectors associated with the unique solution to Equations (5). Then

$$r_{\lambda} = \left(\rho_{\lambda}^{1}, \cdots, \rho_{\lambda}^{n-1}, \hat{w}^{n}\right)^{t}$$
$$t_{\lambda} = \left(r_{\lambda,1}, \cdots, r_{\lambda,n-1}, \hat{z}_{n}\right)^{t}$$

are the related basic and nonbasic vectors associated with the solution (\hat{w}, \hat{z}) .

Proof: Let $(\hat{\psi}, \hat{\xi})$ be the unique solution to Equations (5), then ρ_{λ} and r_{λ} represent the related nonbasic and basic vectors, respectively of $(\hat{\psi}, \hat{\xi})$.

Since (\hat{w}, \hat{z}) is the unique solution to Equations (2) and $\hat{z}_n = 0$, we have $\hat{w}_n \ge 0$ and Lemma 2 implies that

$$\hat{w} = \left(\hat{\psi}^1, \cdots, \hat{\psi}^{n-1}, \hat{w}^n\right)^t, \quad \hat{z} = \left(\hat{\xi}_1, \cdots, \hat{\xi}_{n-1}, 0\right)^t$$

Since the first n-1 blocks of \hat{w} and $\hat{\psi}$ are the same and the first n-1 components of \hat{z} and $\hat{\xi}$ are the same, we must have

$$r_{\lambda} = \left(\rho_{\lambda}^{1}, \cdots, \rho_{\lambda}^{n-1}, \hat{w}^{n}\right)^{t}, \quad t_{\lambda} = \left(r_{\lambda,1}, \cdots, r_{\lambda,n-1}, 0\right)^{t}$$

Theorem 5 Suppose that the algorithm is applied to the problem in Equations (2), when N is a vertical block P-matrix of type (m_1, \dots, m_n) and there exists a $\lambda \in \{1, \dots, \prod_{j=1}^n (m_j + 1)\}$ such that

$$r_{\lambda} = B_{\lambda}^{-1}q, \quad t_{\lambda} = 0$$

$$r_{\lambda}^{j} \ge 0, \quad j = 1, \cdots, n-1$$
(6)

and there exists an $i \in \{1, \dots, m_n\}$ such that $r_{\lambda,i}^n < 0$.

Then in any succeeding stage of the scheme, the nonbasic related variable represented by $t_{\lambda,n}$ will always be a basic related variable.

Proof: Without loss of generality, let $\lambda = 1$. Then

$$B_1 = I, \quad r_1 = w, \quad t_1 = z = 0$$

 $w^j \ge 0, \quad j = 1, \dots, n-1$

and suppose $w_{1,1}^n < 0$.

The next step in the scheme would be to interchange z_n with $w_{1,1}^n$. Suppose in some succeeding stage of the scheme, we have for some

 $\sigma, 1 < \sigma \leq \prod_{j=1}^{n} (m_j + 1)$

$$r_{\sigma}^{j} \ge 0, \quad j = 1, \cdots, n-1, \quad t_{\sigma} = 0$$
$$z_{n} = r_{\sigma,k}^{n} < 0, \quad k \in \{1, \cdots, m_{n}\}$$

Suppose $k = \min\{i: r_{\sigma,i}^n < 0, i \in \{1, \dots, m_n\}\}$. Consider the ordered related $(m_j + 1)$ -tuples associated with $r_1, t_1, r_\sigma, t_\sigma$.

$$s_{1}^{j} = \left(w_{1,1}^{j}, \dots, w_{1,m_{j}}^{ju}, z_{1,j}\right), \quad j = 1, \dots, n$$

$$s_{\sigma}^{j} = \left(w_{\sigma,1}^{1}, \dots, w_{\sigma,m_{j}}^{j}, z_{\sigma,j}\right), \quad j = 1, \dots, n$$

$$s_{1}^{j}, s_{\sigma}^{j} \ge 0, \ j = 1, \dots, n-1, \quad s_{1,m_{j}+1}^{j} = 0, \ j = 1, \dots, n$$

$$s_{1,1}^{n} < 0, \quad s_{1,k}^{n} < 0$$

We will construct a representative submatrix R of N according to the following criteria:

Case 1: If $s_{\sigma,m_j+1}^{j} = 0$, then let $R_{j} = N_{1}^{j}$, $j = 1, \dots, n$, $h_j = 1$

Case 2: For $j = 1, \dots, n-1$, if $s^j_{\sigma,m_j+1} > 0$, find $h_j \in \{1, \dots, m_j\}$ such that $s^j_{1,h_i} \ge 0$ and $s^j_{\sigma,h_i} = 0$.

Let $R_{j} = N_{j}^{j}$. For j = n, we have $s_{1,1}^{n} < 0, s_{1,m_{n}+1}^{n} = 0$, and $s_{\sigma,m_{n}+1}^{n} = r_{\sigma,k}^{n} < 0$, for some $k \in \{1, \dots, m_{j}\}$, where k is the minimum defined above.

If k = 1, this implies that $s_{\sigma,1}^n = t_{\sigma,n} = 0$. If k > 1, then $s_{\sigma,1}^n > 0$. In either case, fix $R_n = N_1^n$, where $h_n = 1$. Let

$$y = \begin{pmatrix} s_{1,h_1}^1 \\ \vdots \\ s_{1,h_n}^n \end{pmatrix}, \quad \overline{y} = \begin{pmatrix} s_{\sigma,h_1}^1 \\ \vdots \\ s_{\sigma,h_n}^n \end{pmatrix}$$
$$x = \begin{pmatrix} s_{1,m_1+1}^1 \\ \vdots \\ s_{1,m_n+1}^n \end{pmatrix}, \quad \overline{x} = \begin{pmatrix} s_{\sigma,m_1+1}^1 \\ \vdots \\ s_{\sigma,m_n+1}^n \end{pmatrix}, \quad p = \begin{pmatrix} q_{h_1}^1 \\ \vdots \\ q_{h_n}^n \end{pmatrix}$$

Since the components of y, \overline{y} and p are chosen to correspond with the rows of R, we have

$$y - Rx = p$$
$$\overline{y} - R\overline{x} = p$$

Therefore,

 $y - Rx = \overline{y} - R\overline{x}$

or

 $(y-\overline{y}) = R(x-\overline{x})$

and

$$(y_j - \overline{y}_j)(x_j - \overline{x}_j) \le 0, \quad j = 1, \dots, n$$

We conclude that *R* cannot be a *P*-matrix by Gale and Nakaido [18]. Therefore, *k* does not equal the minimum subscript such that $r_{\sigma,i}^n < 0$, for some $i \in \{1, \dots, m_n\}$. Therefore, the basic variable z_n is not a candidate to be interchanged with $t_{\sigma,n}$ in any succeeding step of the scheme for $1 < \sigma \leq \prod_{i=1}^{n} (m_j + 1)$ such that the conditions of Equations (6) are met.

Theorem 6. If N is a vertical block P-matrix of type (m_1, \dots, m_n) , the algorithm when applied to Equations (2) will terminate with a solution in a finite number of steps. A related basic matrix that appears once in the scheme will not reappear in any succeeding steps.

Proof: If n = 1, then N is a vertical block P-matrix of type (m_1) , and by Theorem 5, we see that a solution will be found in at most $(m_1 + 1)$ steps. Also once a nonbasic variable becomes basic, it cannot become nonbasic in subsequent steps. Hence, a related basic matrix can appear at most one time in the course of the scheme.

Suppose n > 1 and the theorem holds for all generalized linear complementarity problems such that N is an $m \times K$ vertical block *P*-matrix

of type $(m_1, \dots, m_K), 1 \le K \le n-1$. Let (\hat{w}, \overline{z}) be the unique solution of Equations (2).

Case 1: If $z_n = 0$, then by Lemma 1, $(\hat{\psi}, \hat{\xi})$ is the unique solution to Equations (5). Since *H* is a $m \times (m - m_n)$ vertical block P-matrix of type (m_1, \dots, m_{n-1}) , Theorem 5 applies and the scheme terminates in a finite number of steps with a solution. Any related basic matrix that appears once in the scheme will nor re-appear again. Let the sequence of related basic vectors that appear when the system is applied to *H* be $\rho_{\lambda}, \lambda = 1$ to *K*, where

$$B\rho_{\lambda} = \gamma, \quad r_{\lambda} = 0$$

and ρ_{K} and r_{K} represent the basic and nonbasic variables of the unique solution to (5), $(\hat{\phi}, \hat{\xi})$.

When the scheme is applied to (2), w^n is a related variable. The question of interchanging w_i^n for some $i \in \{1, \dots, m_n\}$ will not be considered in the scheme until we come upon λ such that

$$r_{\lambda}^{j} \ge 0, t_{\lambda,j} = 0, \ j = 1, \cdots, n-1$$
$$r_{\lambda}^{n} = w_{\lambda}^{n}, \quad t_{\lambda,n} = z_{n} = 0$$

The first *K* related basic and nonbasic vectors must be

$$r_{\lambda} = \left(\rho_{\lambda}^{1}, \cdots, \rho_{\lambda}^{n-1}, w_{\lambda}^{n}\right)^{t}$$
$$t_{\lambda} = \left(r_{\lambda,1}, \cdots, r_{\lambda,n-1}, z_{\lambda,n}\right)^{t}$$

for $\lambda = 1$ to K. Lemma 4 and the fact that $\hat{z}_n = 0$ imply that r_K, t_K are the basic and nonbasic variables (\hat{w}, \hat{z}) . Therefore, the theorem applies to the algorithm if $\hat{z}_n = 0$.

Case 2: If $\hat{z}_n > 0$ and $\hat{w}_i^n = 0$ for some $\hat{i} \in \{1, \dots, m_n\}$, then every related basic vector $(r_{\lambda}^1, \dots, r_{\lambda}^{n-1}, w_{\lambda}^n)$ cannot be a solution to (2) since $t_{\lambda n} = z_n = 0$.

Apply the scheme to (5). By the induction process, we have a sequence of related basic vectors $\rho_{\lambda}, \lambda = 1$ to k_1 , and if a basic matrix appears once in the scheme it does not reappear, k_1 is some finite number and ρ_{k_1}, r_{k_1} are the related basic and nonbasic vectors associated with the unique solution to (5).

When we apply the scheme to (2), the first k_1 related basic and nonbasic vectors must be

$$r_{\lambda} = \left(\rho_{\lambda}^{1}, \cdots, \rho_{\lambda}^{n-1}, w_{\lambda}^{n}\right)^{t}$$
$$t_{\lambda} = \left(r_{\lambda,1}, \cdots, r_{\lambda,n-1}, z_{\lambda,n}\right)^{t}$$

for $\lambda = 1$ to k_1 . The hypothesis that $\hat{z}_n > 0$ leads to

$$r_{k_1} = B_{k_1}^{-1}q, t_{k_1} = 0, r_{k_1}^j \ge 0, j = 1, \dots, n-1$$

and there exists an h_1 such that

$$h_1 = \min\{i : r_{k_1,i}^n < 0, i \in \{1, \cdots, m_n\}\}$$

The next related basic vector to appear in the scheme would be

$$r_{k_{1}+1} = \left(\rho_{k_{1}+1}^{1}, \cdots, \rho_{k_{1}+1}^{n-1}, r_{k_{1}+1}^{n}\right)$$

where the basic variables associated with $\rho_{k_1+1}^j$ are the same basic variables associated with $\rho_{k_1}^j$ for $j=1,\dots,n-1$, and $r_{k_1+1,\hat{i}}^n$ represents the variable z_n . The related nonbasic variables would be

$$t_{k_1+1} = (r_{k_1+1,1}, \cdots, r_{k_1+1,n-1}, w_{\hat{i}}^n)^t$$

Let $y = (\rho_{k_1+1}^1, \dots, \rho_{k_1+1}^{n-1}, r_{k_1+1}^n)^t$. Let *B* be the associated related basic matrix and let $x = (r_{k_1+1,1}, \dots, r_{k_1+1,n-1}, w_i^n)^t$. Let *D* be the associated related nonbasic matrix. That is, those columns associated with the components of *t*. Then

$$By + Dx = q$$

Multiplying both sides of the above equation on the left by B^{-1} , we have

$$y + \tilde{N}x = \tilde{q}$$

 \tilde{N} is a vertical block *P*-matrix of type (m_1, \dots, m_n) , since it is the result of a sequence of principal pivots on *N*. The generalized linear complementarity problem with respect to \tilde{N} is:

For $\tilde{q} \in \mathbb{R}^m$, find $y \in \mathbb{R}^m$, $x \in \mathbb{R}^n$ such that

$$y - \tilde{N}x = \tilde{q}$$

$$y \ge 0, \ x \ge 0$$

$$x_j \prod_{i=1}^{m_j} y_i^j = 0, \quad j = 1, \cdots, n$$
(7)

By our assumptions, (7) has a unique solution in which $w_i^n = 0$. If $h_1 = \hat{i}$, then the unique solution to (7) is (\hat{y}, \hat{x}) and $\hat{x}_n = w_i^n = 0$. The subsequent related basic vectors for solving (2) are exactly those related basic vectors which will be obtained by applying the scheme to (7) with y as the initial related basic vector.

We showed in Case 1 that a generalized linear complementarity problem like (7) with unique solution (\hat{r}, \hat{t}) and $\hat{t}_n = 0$ is solved by applying the scheme to (7) and a solution will be obtained in a finite number of steps without a related basic matrix appearing more than once. All these basic matrices will have z_n as a basic variable of the nth block of basic variables and $x_n = w_i^n$ will be nonbasic.

If $h_1 \neq \hat{i}$, we apply the scheme to (7). When applying the scheme to (7), we can argue just as we did when we first applied the scheme to (2). If our transformed system (7) has the unique solution (\hat{y}, \hat{x}) such that $\hat{x}_n > 0$, then after k_2 steps in the scheme, we arrive at a related basic and nonbasic vector r_{k_2} and t_{k_2} , respectively, where

$$t_{k_2} = 0, \ r_{k_2}^j \ge 0, \ j = 1, \cdots, n-1$$

and there exists an

$$h_2 = \min\left\{r_{k_2,i}^n < 0, i \in \{1, \cdots, m_n\} - \{h_1\}\right\}$$

By the way the scheme continues, we must eventually reach a transformed system

$$\omega^{k} - P^{k} \eta^{k} = b^{k}$$
$$\omega^{k} \ge 0, \ \eta^{k} \ge 0$$
$$\eta_{j}^{k} \prod_{i=1}^{m_{j}} (\omega^{k})_{i}^{j} = 0, \ j = 1, \cdots, n$$

that will have as its unique solution $(\hat{\omega}^k, \hat{\eta}^k)$ and $\hat{\eta}_n^k = 0$. Theorem 5 assures us that a related basic matrix that appears once in the scheme will not re-appear in the scheme. This proves that the theorem must hold if the scheme is applied to (2). Since the theorem is true for n = k, it holds for all n.

4. Discussion and Conclusions

The vertical generalized linear complementarity Problem is very general and useful; and it can be applied to many problems in engineering, science, and economics. As such, it may be compared with systems of linear equations, the eigenvalue problem, and linear programming. Thus, it is desirable to have reliable and fail-safe algorithms to obtain a solution. Under the assumption of vertical block *P*-matrix, such a solution exists and is unique. Thus the algorithm takes the input data (N,q) and returns $z \in \mathbb{R}^n$, the unique solution. When this occurs, a mapping $F: \mathbb{R}^{m \times n} \times \mathbb{R}^m$ is defined and this mapping can be used to explore the nature of the VGLP and gain many insights which the researcher desires. An example of this is illustrated in the paper by Oh [10].

Further research might involve investigating the performance of the algorithm under random input data (N,q), measuring the number of iterations taken on average, etc., and the number of arithmetic operations necessary to obtain a solution. These investigations are in lieu of knowledge of a fixed number of iterations and arithmetic operations.

An alternative approach might be to examine a limited set of solutions (say 10 - 50) using a restricted set of points in order to answer some specific questions. This may represent a set of "what if" questions and may satisfy the requirement of this type of study. It is easy to see the appeal of this type of analysis.

Finally, one may envision the algorithm in use as an embedded system, in a vehicle, an industrial machine or a video game. It is quite likely that the model mentioned in Oh [10] would be used in such applications with models of motion, contact (or collision), fluid flow, cooling, etc. In all these cases, the use of direct algorithm will provide efficient, reliable solutions which are necessary for the application of the system in which it is embedded.

The direction of future research will be motivated by all of the above instances of the applications of the vertical generalized linear complementarity problem, and by the discovery of new application areas, which seems to be increasing, as the understanding of this problem continues to develop.

The direct algorithm presented in this paper has certain features not present

in other algorithms. It converges in a finite number of steps, and each step consists of only a unique principal pivot. No anti-cycling device is necessary, even if there is degeneracy in the defining equations. Since the choice of pivot rule is discrete, rather than continuous (such as in the minimum ratio test), no ties are possible. If desired, the use of pre-existing linear algebra software enables the solution of the linear equations, which is required for each iteration of the algorithm.

Acknowledgements

The authors thank Professor James Brannan of Clemson University for his assistance with obtaining some of the resources used in this research.

References

- [1] Murty, K.G. (1974) Note on a Bard-Type Scheme for Solving the Linear Complementarity Problem. *Opsearch*, **11**, 121-130.
- [2] Chakraborty, B. and Ebiefung, A.A. (2013) The Generalized Horizontal Linear Complementarity Problem. *Journal of Mathematical Research*, **5**, 1-6.
- [3] Cottle, R.W. and Dantzig, G.B. (1970) A Generalization of the Linear Complementarity Problem. *Journal of Combinatorial Theory*, 8, 79-90. https://doi.org/10.1016/S0021-9800(70)80010-2
- [4] Lemke, C.E. (1968) On Complementary Pivot Theory. In: Dantzig, G.B. and Pinott, A.F., Eds., *Mathematics of the Decision Sciences*, America Mathematical Society, Providence, 1, 97-115.
- [5] Habetler, G.J. and Szanc, B.P. (1995) Existence and Uniqueness of Solutions for the Generalized Linear Complementarity Problem. *Journal of Optimization Theory and Applications*, 14, 103-116. <u>https://doi.org/10.1007/BF02191738</u>
- [6] Ebiefung, A.A. (1994) A Support Submatrix for the Generalized Linear Complementarity Problem. *Applied Mathematics Letters*, 7, 35-38. https://doi.org/10.1016/0893-9659(94)90090-6
- [7] Ebiefung, A.A. (2007) Existence Theory and Q-matrix Characterization for the Generalized Linear Complementarity Problem: Revisited. *Journal of Mathematical Analysis and Applications*, 329, 1421-1429. https://doi.org/10.1016/j.jmaa.2006.07.036
- [8] Ebiefung, A.A. and Kostreva, M.M. (1992) Global Solvability of Generalized Complementarity Problems and a Related Class of Polynomial Complementarity Problems. In: Floudas, C. and Pardalos, P., Eds., *Recent Advances in Global Optimization*, Princeton University Press, New Jersey, 102-124.
- [9] Ebiefung, A.A., Kostreva, M.M. and Ramanujam, V. (1997) An Algorithm to Solve the Generalized Linear Complementarity Problem with a Vertical Block Z-Matrix. *Optimization Methods and Software*, 7, 123-138. https://doi.org/10.1080/10556789708805648
- [10] Oh, K.P. (1987) The Formulation of the Mixed Lubrication Problem as a Generalized Nonlinear Complementarity Problem, Transactions of the ASME 85. *Journal* of Tribology, 47, 598-603.
- [11] Ebiefung, A.A. and Kostreva, M.M. (1993) The Generalized Leontief Input-Output Model and Its Application to the Choice of New Technology. *Annals of Operations*

Research, 44, 161-172. https://doi.org/10.1007/BF02061065

- Ebiefung, A.A. and Kostreva, M.M. (2003) Production Equilibrium Point in Multi-Unit Manufacturing Systems and the Vertical Linear Complementarity Problem. *Annals of Operations Research*, **124**, 183-192. https://doi.org/10.1023/B:ANOR.0000004768.22040.55
- [13] Ebiefung, A.A., Kostreva, M.M. and Majumdar I. (2008) Disjunctive Programming and the Generalized Leontief Input-Output Model. *Applied Mathematics and Computation*, **198**, 551-558. <u>https://doi.org/10.1016/j.amc.2007.08.055</u>
- [14] Ebiefung, A.A. and Isaac, I. (2012) An Input-Output Pollution Control Model and Product Selection. *Journal of Mathematical Research*, 4, 1-7. <u>https://doi.org/10.5539/jmr.v4n5p1</u>
- [15] Ebiefung, A.A. (2010) Choice of Technology, Industrial Pollution, and the Vertical Linear Complementarity Problem. *Global Journal of Mathematical Sciences*, 9, 113-120.
- [16] Kostreva, M.M. (1979) Cycling in Linear Complementarity Problems. *Mathematical Programming*, 16, 127-130. <u>https://doi.org/10.1007/BF01582098</u>
- [17] Ravindran, A. (1972) Algorithm 431: A Computer Routine for Linear and Quadratic Programming. *Communications of ACM*, 15, 818-820.
- [18] Gale, D. and Nikaido, H. (1965) The Jacobian Matrix and Global Univalence of Mappings. *Mathematische Annalen*, **159**, 81-93. <u>https://doi.org/10.1007/BF01360282</u>

🔆 Scientific Research Publishing

Submit or recommend next manuscript to SCIRP and we will provide best service for you:

Accepting pre-submission inquiries through Email, Facebook, LinkedIn, Twitter, etc. A wide selection of journals (inclusive of 9 subjects, more than 200 journals) Providing 24-hour high-quality service User-friendly online submission system Fair and swift peer-review system Efficient typesetting and proofreading procedure Display of the result of downloads and visits, as well as the number of cited articles Maximum dissemination of your research work

Submit your manuscript at: <u>http://papersubmission.scirp.org/</u> Or contact <u>ojop@scirp.org</u>