

# Iterative Entropy Method for Linear Complementarity Problem

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**Abstract.** Linear complementarity problem can be formulated as a fix point problem. The paper deals with this problem by means of taking the place of maximum entropy function. By constructing an interval extension of entropy function, a new interval algorithm is presented. The relevant properties are proven.

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**Keywords:** linear complementarity problem, maximum entropy function, interval extension

## 1 Introduction

Complementarity theory, which has been studied in the last several decades<sup>[1,2,3]</sup>, is generally considered to be a domain of applied mathematics. The complementarity problem arises in a variety of contexts such as optimization, game theory, economics, classical mechanics, stochastic optimal control, etc.

Let  $M$  be an  $n \times n$  matrix and  $x \in R^n$ . The linear complementarity problem, denoted by  $LCP(M, q)$ , is to find a vector  $x$  such that

$$x \geq 0, (Mx + q) \geq 0, x^T (Mx + q) = 0. \quad (1.1)$$

The following notations is used throughout our paper. We define the absolute value of a vector  $x$  by

$$|x| = (|x_1|, |x_2|, \dots, |x_n|)^T$$

$X = (X_1, X_2, \dots, X_n)^T$  is a  $n$ -dimensional interval vector, which  $X_i = [\underline{x}_i, \overline{x}_i]$  is a closed bounded set.

Let  $I(R_+^n)$  be the set of all interval vector in  $R_+^n$ .

Let  $f$  be a real valued function of  $n$  real variables  $x_1, x_2, \dots, x_n$ . By an interval extension of  $f$ , we mean an interval valued function  $F$  of  $n$  interval variables  $X_1, X_2, \dots, X_n$ , with the property  $F(x_1, x_2, \dots, x_n) = f(x_1, x_2, \dots, x_n)$  for real variable. The further knowledge about interval mathematics can be found in [4], [5], etc.

## 2 Maximum Entropy Function for LCP

The problem (1.1) can be formulated as a fix point problem<sup>[4]</sup>. Just as following :

**Lemma 2.1**<sup>[4]</sup>  $x^*$  is the solution of (1.1) if and only if  $x^*$  is a fixed point of the map

$$h(x) = \max\{0, x - (Mx + q)\}. \quad (1.2)$$

Let  $N = I - M, r = q$ , so (1.2) can be transformed as following

$$g(x) = \max\{0, Nx - r\} \quad (1.3)$$

Let  $F(x, p)$  be maximum entropy function<sup>[6]</sup> of  $g(x)$ , which

$$F(x, p) = (F_1(x, p), \dots, F_n(x, p))^T,$$

$$F_i(x, p) = \frac{1}{p} \ln\{1 + \exp(p(\sum_{j=1}^n n_{ij}x_j - r_i))\}, \quad (i = 1, 2, \dots, n)$$

**Lemma 2.2** When  $p \rightarrow +\infty$ ,  $F_i(x, p)$  is uniform convergence to  $g_i(x)$  in its domain with the following inequality

$$g_i(x) \leq F_i(x, p) \leq g_i(x) + \frac{1}{p} \ln 2 \quad (i = 1, 2, \dots, n)$$

**Proof** With the expression of  $F_i(x, p)$ , we get

$$\begin{aligned} F_i(x, p) - g_i(x) &= \frac{1}{p} \ln \{1 + \exp(p(\sum_{j=1}^n n_{ij}x_j - r_i))\} - g_i(x) \\ &= \frac{1}{p} \ln \{ \exp(p(0 - g_i(x))) + \exp(p(\sum_{i=1}^n n_{ij}x_j - r_i - g_i(x))) \} \end{aligned}$$

From  $g_i(x) = \max\{0, \sum_{i=1}^n n_{ij}x_j - r_i\}$ , we have

$-g_i(x) \leq 0$  and  $\sum_{i=1}^n n_{ij}x_j - r_i - g_i(x) \leq 0$ , also one of two inequality is

zero .So we have

$$1 \leq \exp(p(-g_i(x))) + \exp(p(\sum_{i=1}^n n_{ij}x_j - r_i - g_i(x))) \leq 2$$

then , we get

$$g_i(x) \leq F_i(x, p) \leq g_i(x) + \frac{1}{p} \ln 2 .$$

This completes the proof of the lemma2.2.

**Lemma 2.3**  $F_i(x, p)$  is monotone decreasing about  $p$ , i.e.

$$F_i(x, p_2) > F_i(x, p_1) \quad \text{when } p_1 > p_2 > 1 .(i = 1, 2, \dots, n)$$

**Proof** From Jensen inequality

when  $p_1 > p_2 > 1$ ,

$$\begin{aligned} & \left\{ \exp(p_1 \bullet 0) + \exp\left(p_1 \left(\sum_{j=1}^n n_{ij} x_j - r_i\right)\right) \right\}^{\frac{1}{p_1}} \\ & < \left\{ \exp(p_2 \bullet 0) + \exp\left(p_2 \left(\sum_{j=1}^n n_{ij} x_j - r_i\right)\right) \right\}^{\frac{1}{p_2}} \end{aligned}$$

then ,we have

$$\begin{aligned} & \frac{1}{p_1} \ln \left\{ 1 + \exp\left(p_1 \left(\sum_{j=1}^n n_{ij} x_j - r_i\right)\right) \right\} \\ & < \frac{1}{p_2} \ln \left\{ 1 + \exp\left(p_2 \left(\sum_{j=1}^n n_{ij} x_j - r_i\right)\right) \right\} \end{aligned}$$

so  $F_i(x, p_2) > F_i(x, p_1)$ .

**Lemma 2.4** If  $x_p$  is a solution of the equation  $F(x, p) = x$ , then  $x_p$  is the approximation solution of ( 1.2 )

**Proof** Now we consider

$$\begin{aligned} & \left| g(x_p) - x_p \right| \\ & = \left| g(x_p) - F(x_p, p) + F(x_p, p) - x_p \right| \\ & = \left| g(x_p) - F(x_p, p) \right| \end{aligned}$$

from lemma 2.2 , about  $\forall \varepsilon > 0$  ,  $\exists M > 0$  , satisfies

$$\left| g(x_p) - F(x_p, p) \right| < \varepsilon$$

i.e.  $\left| g(x_p) - x_p \right| \leq \varepsilon$  , then we have for any given  $\varepsilon > 0$  ,  $x_p$  is the approximation solution of ( 1.2 ) .

From the proceeding result, we know that (1.2) equal the equation

$$F(x, p) = x . \quad (1.4)$$

### 3 Interval Extension of Entropy Function

Denote  $[\underline{s}_i, \bar{s}_i] = \sum_{j=1}^n n_{ij} X_j - r_i \quad (i = 1, 2, \dots, n)$

Define interval function

$$T(X, p) = (T_1(X, p), \dots, T_n(X, p))^T$$

which

$$T_i(X, p) = \frac{1}{p} \ln \{1 + \exp(p(\sum_{j=1}^n n_{ij} X_j - r_i))\} \quad (i = 1, 2, \dots, n)$$

**Theorem 3.1**  $T(X, p)$  is a interval extension of  $F(x, p)$  , and exceed width is zero.

**Proof** For  $i = 1, 2, \dots, n$  ,

$$T_i(x, p) = \frac{1}{p} \ln \{1 + \exp(p(\sum_{j=1}^n n_{ij} x_j - r_i))\} = F_i(x, p)$$

with inclusion monotonic of interval arithmetic, we

know  $\sum_{j=1}^n n_{ij} x_j - r_i \in \sum_{j=1}^n n_{ij} X_j - r_i$  ,

So we get

$$\begin{aligned} F_i(x, p) &= \frac{1}{p} \ln \{1 + \exp(p(\sum_{j=1}^n n_{ij} x_j - r_i))\} \in \frac{1}{p} \ln \{1 + \exp(p(\sum_{j=1}^n n_{ij} X_j - r_i))\} \\ &= T_i(X, p) , \end{aligned}$$

then  $F(x, p) \in T(X, p)$ , so  $T(X, p)$  is an interval extension of  $F(x, p)$  .

Also  $\sum_{j=1}^n n_{ij} X_j - r_i$  is a natural interval extension of  $\sum_{j=1}^n n_{ij} x_j - r_i$  , and each

of the variable occurs only once in  $\sum_{j=1}^n n_{ij} x_j - r_i$ , from lemma 1.1 , the exceed

width of  $\sum_{j=1}^n n_{ij} X_j - r_i$  is zero , i.e.  $\underline{s}_i = \min_{x \in X} (\sum_{j=1}^n n_{ij} x_j - r_i)$  ,

$$\overline{s}_i = \max_{x \in X} (\sum_{j=1}^n n_{ij} x_j - r_i) .$$

Further,  $\ln x$  and  $\exp x$  are the monotone function , we know

$$\overline{F}_i(X, p) = \frac{1}{p} \ln(1 + \exp(p\overline{s}_i)) = \overline{T}_i(X, p)$$

$$\underline{F}_i(X, p) = \frac{1}{p} \ln(1 + \exp(p\underline{s}_i)) = \underline{T}_i(X, p)$$

so we get the exceed width of  $T(X, p)$  is zero.

**Theorem 3.2** For  $\forall X \in I(R_+^n)$

( 1 ) If  $x^* \in X$  is a solution of  $F(x, p) = x$ , then  $x^* \in T(X, p)$ .

( 2 ) If  $T(X, p) \cap X = \emptyset$ , then the solution of (1.4) is not exist in  $X$ .

( 3 ) If  $T(X, p) \subseteq X$ , then there is a solution of (1.4) in  $X$ .

**Proof** ( 1 ) Let  $x^*$  be a solution of (1.4), i.e.  $x^* = F(x^*, p)$ , from  $T(X, p)$  is a interval extension of  $F(x, p)$  in  $X$ ,  $x^* \in X$ , so  $F(x^*, p) \in T(X, p)$ , i.e.  $x^* \in T(X, p)$ .

( 2 ) If there is not a solution of (1.4) in  $X$ , from the statement of (1), we get  $x^* \in T(X, p)$ , this contradict with  $T(X, p) \cap X = \emptyset$ .

( 3 ) For  $\forall x \in X$ ,  $F(x, p) \in X$ , and  $X$  is a bounded convex set, from Brouwer theorem, we know there is a fixed point  $x^* \in X$  satisfied  $x^* = F(x^*, p)$ .

## 4 Algorithm and Convergence

With the above ideas, we suggest the following algorithm for solving problem (1.1)

$$\begin{cases} T_i(X^{(k)}, p) = \frac{1}{p} \ln\{1 + \exp(p(\sum_{j=1}^n n_{ij} X_j^{(k)} - r_i))\} \quad (i = 1, 2, \dots, n) \\ X^{(k+1)} = T(X^{(k)}, p) \cap X^{(k)} \quad (k = 0, 1, 2, \dots) \end{cases}$$

(1.5)

**Theorem 4.1**  $W(T(X, p)) \leq W(NX - r)$  .

**Proof** From  $\ln x$  and  $\exp x$  are the monotone function, we get

$$\begin{aligned} & \bar{T}_i(X, p) - \underline{T}_i(X, p) \\ &= \frac{1}{p} \ln\{1 + \exp(p\bar{s}_i)\} - \frac{1}{p} \ln\{1 + \exp(p\underline{s}_i)\} \\ &= \frac{1}{p} \frac{\exp(p\xi)}{1 + \exp(p\xi)} p(\bar{s}_i - \underline{s}_i) \\ &\leq \bar{s}_i - \underline{s}_i \end{aligned}$$

which  $\xi \in [\bar{s}_i - \underline{s}_i]$  , so  $W(T_i(X, p)) \leq W((\sum_{j=1}^n n_{ij} X_j - r_i))$  , i.e.

$$W(T(X, p)) \leq W(NX - r).$$

**Theorem 4.2** Let  $M$  be a  $H$ -matrix, where  $m_{ii} > 0, i = 1, 2, \dots, n$  , if there is a solution  $x_p$  of (1.4) in  $X^{(0)} \in I(R_+^n)$  , so (3.7) can get a serial  $\{X^k\}$  , satisfy  $\lim_{k \rightarrow \infty} X^{(k)} = x_p$  .

**Proof** From (3.7),

$$X^{(k+1)} \subseteq \frac{1}{p} \ln\{1 + \exp(p(NX^{(k)} - r))\}$$

so 
$$r(X^{(k+1)}) \leq r\left(\frac{1}{p} \ln\{1 + \exp(p(NX^{(k)} - r))\}\right)$$

from theorem 4.1 
$$r(X^{(k+1)}) \leq r(NX^{(k)} - r)$$

$$\begin{aligned} &\leq r(NX^{(k)}) = r((I - D\bar{M})X^{(k)}) \\ &= (I - D\bar{M})r(X^{(k)}) \end{aligned}$$

where  $M$  is  $H$ -matrix, so  $\rho(I - D\bar{M}) < 1$ , from theorem 2.2<sup>[61]</sup>, we get

$r(X^{(k)}) \rightarrow 0$ , then  $x_p \in X^{(0)}$ , so  $x_p \in X^{(k)}$ .

The detailed steps of the algorithm are the following.

Step 1 Set  $X = X^{(0)}$ , let  $p$  be sufficiently large.

Step 2 Compute  $T_i(X, p) = \frac{1}{p} \ln\{1 + \exp(p(\sum_{j=1}^n n_{ij} X_j - r_i))\}$  ( $i = 1, 2, \dots, n$ )

Step 3 Let  $X^{(k+1)} = T(X^{(k)}, p) \cap X^{(k)}$ .

Step 4 If  $X^{(k+1)} = \emptyset$ , then we terminate the algorithm; Otherwise proceed with step 2.

Step 5 If  $\|W(X^{(k+1)})\|_{\infty} < \varepsilon$ , then we obtain the solution  $X^{(k+1)}$ , and terminate the algorithm. Otherwise proceed with step 2.

## 5 Numerical Results

The algorithm has been implemented using Turbo C2.0 on computer.

$X^{(0)}$  is tentative interval,  $x^*$  is exact solution,  $X$  is numerical solution,  $K$  is iterative number of times, precision  $\varepsilon = 1e-5$ , entropy factor  $p = 300$ .

Example 1:

$$M = \begin{pmatrix} 4 & -1 & 0 & 0 \\ -1 & 4 & -1 & 0 \\ 0 & -1 & 4 & -1 \\ 0 & 0 & -1 & 4 \end{pmatrix}, \quad q = \begin{pmatrix} -4 \\ 3 \\ -4 \\ 2 \end{pmatrix}$$

$$x^* = (1, 0, 1, 0)^T$$



$$X^{(0)} = \begin{bmatrix} [0, 2] \\ [0, 2] \\ [0, 2] \\ [0, 2] \end{bmatrix}, X = \begin{bmatrix} [0.9999987, 0.9999997] \\ [0.0000012, 0.0000042] \\ [0.9999998, 1.0000023] \\ [0.0000043, 0.0000053] \end{bmatrix}, K = 63.$$

Example 2:

$$M = \begin{pmatrix} 3 & 0 & -1 & 0 \\ -1 & 3 & -1 & 0 \\ 0 & -1 & 4 & -2 \\ -1 & -1 & -1 & 5 \end{pmatrix}, q = \begin{pmatrix} -2 \\ 3 \\ -4 \\ 5 \end{pmatrix}$$

$$x^* = (1, 0, 1, 0)^T$$

$$X^{(0)} = \begin{bmatrix} [0, 2] \\ [0, 2] \\ [0, 2] \\ [0, 2] \end{bmatrix}, X = \begin{bmatrix} [0.9999877, 0.9999987] \\ [0.0000012, 0.0000024] \\ [0.9999978, 0.9999988] \\ [0.0000014, 0.0000032] \end{bmatrix}, K = 52$$

The examples above and more numerical results indicate that our algorithm works reliably and efficiently.

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