AN EXPERIMENTAL STUDY ON SOLVING LINEAR PROGRAMS

by

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UNIVERSITY OF CALIFORNIA - BERKELEY
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ABSTRACT

An experimental study to compare the simplex method and the Lemke's method to solve linear programs is made. The M3 code for simplex method and the author's code for the Lemke method were used in the study. Comparison was made only with regard to the number of iterations each method takes and our little study shows encouraging results about the superiority of Lemke method, but no general recommendation is made by the author due to size of the study and data. A by-product of our study is a complementary pivot algorithm to solve linear programs which is a modification of the Lemke's method and which saves a considerable storage and time of computation.
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1. INTRODUCTION

From the beginning of the era of linear programming a lot of work has been done to find a better and more efficient method to solve linear programs than the simplex method of G. B. Dantzig [1] but without much success. Quite a few papers have been published about the performance of the simplex method in terms of the number of iterations or pivot steps. Kuhn and Quandt made a study to test different pivot choices for the simplex code [4]; but strangely, the firm theoretical bounds given in the papers are much higher than the actual number encountered in practice. In spite of extensive computational experience, not very much is known precisely about the performance of the simplex method, and it is still considered to be the best known method to solve linear programs by many workers in the field.

To add to this huge amount of literature on the study of the simplex method, we have made a small scale study to compare it with the "complementary pivotal method" of Lemke and Howson. Our main object has been to compare the number of iterations in both cases. Our study revealed that generally Lemke method is as efficient as simplex and in special class of $A$ matrices (viz $A > 0$) Lemke method takes as little as $1/2$ to $1/3$ the number of iterations of simplex! This opens the door to further study of the Lemke method on a large scale to solve linear programs. A by-product of our study is a modified algorithm to that of Lemke-Howson to solve linear programs which will save a considerable amount of storage and computation time.
To make this account self-contained, an exposition of the Lemke's method in the form we have used (i.e., programmed\(^\dagger\)) is presented in Section 2. The details of the statistical experiment and the results of our study are described in Section 3. Our modified Lemke's algorithm is explained in Section 4.

The fact that led us to make this study is that the simplex method always starts with a basic feasible solution \([1]\). In many problems in fact we do not have an initial basic feasible solution readily available and to find one, one uses the so called Phase 1 technique of Dantzig which means solving another linear program with additional artificial variables. This Phase 1 is avoided when one uses Lemke method as it is easy to find one "almost complementary solution."

\(^\dagger\)An independent computer program was written in FORTRAN IV for the Lemke-Howson's algorithm by the author.
2. THE COMPLEMENTARY PROBLEM AND LEMKE’S ITERATIVE TECHNIQUE

The following complementary problem was posed by Dantzig and Cottle in 1963 [2]

\[ \begin{align*}
  w &= Mz + q \\
  w &> 0, \ z &> 0 \\
  w \cdot z &= 0 \quad \text{(i.e., } w_i z_i = 0) 
\end{align*} \]  

(1)

where \( M \) is \((n \times n)\) square matrix.

\( w, z, q \) are \( n \)-vectors.

It can be seen that linear programming, quadratic programming and bi-matrix (2-person, nonzero-sum) games can be transformed to the above complementary problem form.

Dantzig and Cottle in their paper [2] also proposed an algorithm which is applicable to matrices \( M \) that have positive principal minors (in particular to positive definite matrices) and after some modification (which uses the notion of "block pivot") to positive semi-definite matrices. But we will be concerned with the iterative technique of Lemke and Howson [4] for finding equilibrium points of bi-matrix games which was later extended by Lemke [5] to the complementary problem (1). We will essentially give the step by step procedure of the Lemke’s algorithm as presented by Gale [3] which is also used as the basis for the computer code written.

Consider the following linear program:

\[ \begin{align*}
  \text{Minimize} \quad & cx \\
  \text{Subject to} \quad & Ax \geq b \\
  \quad & x \geq 0
\end{align*} \]  

(2)

*No special notation to denote the transpose of a vector is used throughout this paper.*
Let $A$ be an $m \times n$ matrix, $b$ is an $m$-vector, $c$ and $x$ are $n$-vectors.

The above problem can be transformed into a complementary problem very easily as follows:

$$
\begin{bmatrix}
u \\
v
\end{bmatrix} = \begin{bmatrix} 0 & A \\ -A^T & 0 \\
\end{bmatrix} \begin{bmatrix} y \\
x
\end{bmatrix} + \begin{bmatrix} -b \\
c
\end{bmatrix}
$$

$$
uy + vx = 0
$$

where $u$-denotes the primal slack variables

$y$-denotes the dual variables

$v$-denotes the dual slack variables.

The reader will immediately note that system (3) represents primal and dual feasibility and (4) represents the complementary slackness condition.

Comparing these to system (1) will show

$$
M = \begin{bmatrix} 0 & A \\ -A^T & 0 \\
\end{bmatrix},
u = \begin{bmatrix} u \\
v
\end{bmatrix},z = \begin{bmatrix} y \\
x
\end{bmatrix},q = \begin{bmatrix} -b \\
c
\end{bmatrix}
$$

It should be remarked here that $M$ is a square nonsymmetric positive semidefinite matrix of order $m + n$.

We shall now describe Lemke's method to solve the original complementary problem (1) which guarantees to terminate with a solution if it exists for the following cases: (i) $M > 0$ (all elements), (ii) $M$ is positive definite, (iii) $M$ having positive principal determinants, (iv) $M$ is positive semidefinite, (v) $M$ co-positive. We shall first start with some definitions.

Let $a_1, ..., a_n$ denote the first $n$-column vectors of $M$ matrix. $e_1, ..., e_n$ are the first $n$ unit vectors.
Definition 1: Complementary Solution

A solution \((w,z) > 0\) satisfying \(w = Mz + q\) and for each \(i = 1, \ldots, n\), either \(e_i\) (variable \(w_i\)) or \(-a_i\) (variable \(z_i\)) is in the basis and not both, i.e., a solution to system (1).

Definition 2: Almost Complementary Solution

A solution \((w,z) > 0, Z_o > 0\) to the system \(w = Mz + e Z_o + q\) where \(e = (1, \ldots, 1)\) such that exactly for one \(i\) both \(e_i\) and \(-a_i\) are out of the basis; and of course \(Z_o\) is in the basis as \(Z_o > 0\).

Lemke's algorithm starts with an almost complementary solution by adding the vector \(e Z_o\) to \(w = Mz + q\) and using a complementary pivot rule (described later) moves from one almost complementary solution to the other until it terminates which is given by the termination conditions described later.

The initial almost complementary solution is found by augmenting the vector \(e Z_o\) to the system (1) as follows:

\[
\begin{align*}
  w - Mz &= e Z_o + q \\
  w, z, Z_o &> 0 \\
  w z &= 0
\end{align*}
\]

The initial tableau with respect to the unit basis will be as follows:

<table>
<thead>
<tr>
<th>Basis Vectors</th>
<th>(e_1, e_2, \ldots, e_n, -a_1, \ldots, -a_n, -e)</th>
<th>(q)</th>
</tr>
</thead>
<tbody>
<tr>
<td>(e_1)</td>
<td>1</td>
<td>-1</td>
</tr>
<tr>
<td>(e_2)</td>
<td>1</td>
<td>-1</td>
</tr>
<tr>
<td>(e_3)</td>
<td>1</td>
<td>-1</td>
</tr>
<tr>
<td>(e_n)</td>
<td>1</td>
<td>-1</td>
</tr>
</tbody>
</table>

TABLE 1
Note that the initial basis though complementary need not be feasible as some \( q_i < 0 \).

**Step 1:**

- e vector (variable \( Z_0 \)) is brought into the basis so as to replace one of e_i's, namely, e_s from the basis where \( q_s = \min_{i=1,2,\ldots,n} (q_i) \). If \( q_s > 0 \) then terminate. We have a complementary solution to (1) given by \( w_i = q_i \forall i \) and \( z_i = 0 \). Otherwise perform the pivot as shown below with circled element as pivot element.

The pivot operation yields the following tableau

<table>
<thead>
<tr>
<th>1</th>
<th>-1</th>
<th>0</th>
<th>( m_{11} )</th>
<th>( m_{1n} )</th>
<th>0</th>
</tr>
</thead>
<tbody>
<tr>
<td>e_1</td>
<td>1</td>
<td>-1</td>
<td>0</td>
<td>( m_{11} )</td>
<td>( m_{1n} )</td>
</tr>
<tr>
<td>e_2</td>
<td>0</td>
<td>1</td>
<td>-1</td>
<td>( m_{22} )</td>
<td>( m_{2n} )</td>
</tr>
<tr>
<td>e_3</td>
<td>0</td>
<td>-1</td>
<td>0</td>
<td>( m_{33} )</td>
<td>( m_{3n} )</td>
</tr>
<tr>
<td>-e</td>
<td>-1</td>
<td>0</td>
<td>( m_{s1} )</td>
<td>( m_{sn} )</td>
<td>1</td>
</tr>
<tr>
<td>e_{s+1}</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>e_n</td>
<td>0</td>
<td>-1</td>
<td>0</td>
<td>( m_{n1} )</td>
<td>( m_{nn} )</td>
</tr>
</tbody>
</table>

where

\[ q_s = -q_s, \quad q_i = q_i - q_s \forall i \neq s \]

\[ m_{sj} = \frac{-m_{s1}}{-1} = m_{sj} \quad \forall j = 1,2,\ldots,n \]

\[ m_{ij} = -m_{ij} + m_{sj} \quad \forall j = 1,2,\ldots,n \]

\( \forall i \neq s \).

Note

\[ q_i \geq 0 \forall i. \]
The corresponding basic solution \( w_1 = q_1', \ldots, w_{s-1} = q_{s-1}', z_0 = q_s', w_{s+1} = q_{s+1}', \ldots \), \( w_n = q_n' \), is an almost complementary solution as both variables \( w_s \) and \( z_s \) are out of the basis, \( Z_0 > 0 \) and \( w, z > 0 \).

**Step 2:**

Since both the vectors \( -a_s \) and \( e_s \) are out of the basis, we bring in \( -a_s \) into the basis which will still maintain almost complementarity. To find the vector leaving the basis, form the ratios

\[
\frac{q_i'}{m_{is}} \quad \text{for } i = 1, \ldots, n \quad \text{and} \quad m_{is} > 0 \quad \forall \ i = 1, \ldots, n.
\]

Let

\[
\frac{q_k'}{m_{ks}} = \min \frac{q_i'}{m_{is}} \quad m_{is} > 0
\]

then \( w_k \) leaves the basis. Obtain the new tableau by performing the pivot with \( a_{ks} \) as the pivot element.

**Step 3:**

Now bring in \( -a_k \) into the basis and continue as before until one of the two things happen which indicates termination.

(i) Min ratio happens at the \( s^{th} \) row and \( Z_0 \) leaves the basis. The resulting basic solution (after the pivot) is the complementary solution to (1).

(ii) All \( m_{1k} \) for \( i = 1, 2, \ldots, n \) are \( \leq 0 \) \( \Rightarrow \) there exists no feasible solution to the original problem (1).

In the coding of the algorithm there was a slight modification done. Instead of doing the pivot operation to the whole tableau only the inverse of the basis
is kept in the memory and the pivot operations were done only to the inverse. Similar ideas that are common in revised simplex method with explicit inverse are employed.
3. DESIGN, DATA AND RESULTS OF THE EXPERIMENT

In order to evaluate the statistical efficiency of the simplex method and
Lemke's method in solving linear programs, a number of linear programs were
generated randomly and the solutions computed by both methods. The linear
programs generated were of the form: \( Ax \geq b, x \geq 0 \) minimize \( c^T x \). Elements of
matrix \( A \), vectors \( b \) and \( c \) were chosen randomly from a uniform distribution
from 0 and 1. Matrices of size 5 x 5, 10 x 10, 15 x 15 were tried for the \( A \)
matrix. Initially, 10 problems for each size were generated and for each problem
and method the number of iterations and the means of iteration count for each
size matrix are noted. They are shown in the following tables.

The calculations were done on the IBM-7094 at the Computer Center, University of
California, Berkeley. The M3 code was used for the simplex method and the author's
code for the Lemke method.
Results of the Study

<table>
<thead>
<tr>
<th>Prob. No.</th>
<th>SIZE 5 x 5 (A &gt; 0)</th>
<th>SIZE 10 x 10 (A &gt; 0)</th>
<th>SIZE 15 x 15 (50% of A &gt; 0)</th>
<th>SIZE 15 x 15 (50% of A &lt; 0)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>No. of Iterations</td>
<td>Comment</td>
<td>No. of Iterations</td>
<td>Comment</td>
</tr>
<tr>
<td></td>
<td>Simplex Method</td>
<td>Lemke Method</td>
<td>Simplex</td>
<td>Lemke</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>1</td>
<td>14</td>
<td>3</td>
<td>Optimal Sol.</td>
<td>1</td>
</tr>
<tr>
<td>2</td>
<td>10</td>
<td>7</td>
<td>&quot;</td>
<td>2</td>
</tr>
<tr>
<td>3</td>
<td>10</td>
<td>11</td>
<td>&quot;</td>
<td>3</td>
</tr>
<tr>
<td>4</td>
<td>14</td>
<td>7</td>
<td>&quot;</td>
<td>4</td>
</tr>
<tr>
<td>5</td>
<td>11</td>
<td>5</td>
<td>&quot;</td>
<td>5</td>
</tr>
<tr>
<td>6</td>
<td>11</td>
<td>7</td>
<td>&quot;</td>
<td>6</td>
</tr>
<tr>
<td>7</td>
<td>9</td>
<td>5</td>
<td>&quot;</td>
<td>7</td>
</tr>
<tr>
<td>8</td>
<td>12</td>
<td>5</td>
<td>&quot;</td>
<td>8</td>
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<td>9</td>
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<td>9</td>
<td>&quot;</td>
<td>9</td>
</tr>
<tr>
<td>10</td>
<td>12</td>
<td>15</td>
<td>&quot;</td>
<td>10</td>
</tr>
<tr>
<td>Average</td>
<td>11.3</td>
<td>7.4</td>
<td>1.52 Simplex</td>
<td>Average</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Prob. No.</th>
<th>Simplex</th>
<th>Lemke</th>
<th>Comment</th>
<th>Prob. No.</th>
<th>Simplex</th>
<th>Lemke</th>
<th>Comment</th>
</tr>
</thead>
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<td>1</td>
<td>48</td>
<td>13</td>
<td>Optimal</td>
<td>1</td>
<td>22</td>
<td>41</td>
<td>Optimal</td>
</tr>
<tr>
<td>2</td>
<td>48</td>
<td>15</td>
<td>&quot;</td>
<td>2</td>
<td>33</td>
<td>27</td>
<td>Optimal</td>
</tr>
<tr>
<td>3</td>
<td>52</td>
<td>17</td>
<td>&quot;</td>
<td>3</td>
<td>22</td>
<td>13</td>
<td>Infeasible</td>
</tr>
<tr>
<td>4</td>
<td>47</td>
<td>15</td>
<td>&quot;</td>
<td>4</td>
<td>30</td>
<td>1</td>
<td>Optimal</td>
</tr>
<tr>
<td>5</td>
<td>57</td>
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<td>&quot;</td>
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<td>35</td>
<td>33</td>
<td>Optimal</td>
</tr>
<tr>
<td>6</td>
<td>43</td>
<td>11</td>
<td>&quot;</td>
<td>6</td>
<td>21</td>
<td>18</td>
<td>Infeasible</td>
</tr>
<tr>
<td>7</td>
<td>49</td>
<td>17</td>
<td>&quot;</td>
<td>7</td>
<td>14</td>
<td>17</td>
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</tr>
<tr>
<td>8</td>
<td>46</td>
<td>23</td>
<td>&quot;</td>
<td>8</td>
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</tr>
<tr>
<td>9</td>
<td>45</td>
<td>21</td>
<td>&quot;</td>
<td>9</td>
<td>24</td>
<td>27</td>
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</tr>
<tr>
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<td>46</td>
<td>21</td>
<td>&quot;</td>
<td>10</td>
<td>33</td>
<td>35</td>
<td>Infeasible</td>
</tr>
<tr>
<td>Average</td>
<td>48.1</td>
<td>16.6</td>
<td>48.1 = 2.89</td>
<td>Average</td>
<td>26.4</td>
<td>27.3</td>
<td>26.4 = .96</td>
</tr>
<tr>
<td>Prob. No.</td>
<td>No. of Iterations</td>
<td>Comment</td>
<td>Prob. No.</td>
<td>No. of Iterations</td>
<td>Comment</td>
<td></td>
<td></td>
</tr>
<tr>
<td>----------</td>
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<td>---------</td>
<td>----------</td>
<td>-------------------</td>
<td>---------</td>
<td></td>
<td></td>
</tr>
<tr>
<td>SIZE (15 x 15) 80% of A = 0</td>
<td>SIZE (15 x 15) 10% of A &gt; 0</td>
<td>SIZE (15 x 15) 10% of A &lt; 0</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Simplex</td>
<td>Lemke</td>
<td></td>
<td>Simplex</td>
<td>Lemke</td>
<td></td>
<td></td>
</tr>
<tr>
<td>1</td>
<td>24</td>
<td>43</td>
<td>Optimal</td>
<td>1</td>
<td>19</td>
<td>17</td>
<td>Infeasible</td>
</tr>
<tr>
<td>2</td>
<td>22</td>
<td>1</td>
<td>Infeasible</td>
<td>2</td>
<td>26</td>
<td>11</td>
<td>Infeasible</td>
</tr>
<tr>
<td>3</td>
<td>26</td>
<td>15</td>
<td>Optimal</td>
<td>3</td>
<td>11</td>
<td>15</td>
<td>Infeasible</td>
</tr>
<tr>
<td>4</td>
<td>30</td>
<td>19</td>
<td>Optimal</td>
<td>4</td>
<td>9</td>
<td>3</td>
<td>Infeasible</td>
</tr>
<tr>
<td>5</td>
<td>36</td>
<td>19</td>
<td>Optimal</td>
<td>5</td>
<td>20</td>
<td>7</td>
<td>Infeasible</td>
</tr>
<tr>
<td>6</td>
<td>22</td>
<td>7</td>
<td>Infeasible</td>
<td>6</td>
<td>17</td>
<td>13</td>
<td>Infeasible</td>
</tr>
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<td>7</td>
<td>18</td>
<td>25</td>
<td>Infeasible</td>
<td>7</td>
<td>20</td>
<td>19</td>
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</tr>
<tr>
<td>8</td>
<td>19</td>
<td>17</td>
<td>Infeasible</td>
<td>8</td>
<td>9</td>
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<td>Infeasible</td>
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<td>9</td>
<td>24</td>
<td>1</td>
<td>Infeasible</td>
<td>9</td>
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<td>Infeasible</td>
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<tr>
<td>10</td>
<td>50</td>
<td>24</td>
<td>Optimal</td>
<td>10</td>
<td>13</td>
<td>1</td>
<td>Infeasible</td>
</tr>
<tr>
<td><strong>Average</strong></td>
<td>27.1</td>
<td>17.1</td>
<td><strong>1.58</strong></td>
<td><strong>Average</strong></td>
<td>15.2</td>
<td>10.4</td>
<td><strong>Simplex</strong> 1.46</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td><strong>Simplex</strong> 1.46</td>
</tr>
</tbody>
</table>

Note: The table compares the performance of the Simplex and Lemke methods for solving linear programming problems with constraints on the matrix A. The table shows the number of iterations required for each method, along with comments on whether the solution was optimal or infeasible. The average number of iterations is also provided for each method, along with a ratio of Simplex to Lemke iterations.
Discussion of the Results

The previous tables show when the matrix $A$ was positive, the simplex method takes 1.52 times for $5 \times 5$, 2.27 times for $10 \times 10$ and 2.89 times for $15 \times 15$ more number of iterations compared to Lemke method and hence Lemke method seems more efficient in the number of iterations. The best reason may be that the Lemke method avoids the Phase 1 procedure of simplex method and for $A > 0$ it is possible to get an easy feasible solution to the primal even if we do not use simplex Phase 1. During our discussions with Dantzig [7] it was suggested that we should try some matrices with a lot of zero entries. And even in that excepting 50% positive and negative entries of $A$ matrix, in all the other cases, Lemke method appears to be more efficient even in finding a basic feasible solution compared to Phase 1 procedure. Since the research was directed at one aspect of the methods (viz. No. of iterations) and the data are based on the statistics of one class of linear programs, we do not make any conclusive recommendation about the efficiency of either method for solving linear programs. But our little study has evidently shown that the results we have are very encouraging to do a large scale study about these two methods for future investigation.
4. A MODIFIED APPROACH TO SOLVE LINEAR PROGRAMS BY LEMKE'S METHOD

We observed in Section 3 that when the $A$ matrix is strictly positive, it is worthwhile solving the linear program by Lemke method and in other classes of $A$ matrices both methods are equally good. The only question that can be raised is obviously the fact that we are working with a larger size of the matrix (actually twice the size of $A$ matrix when $A$ is square) when using Lemke's method. The larger size will definitely create problems of storage and time of computation. So further investigations were carried out to see whether it is possible to work with a reduced $M$ matrix where $M = \begin{bmatrix} 0 & A \\ -A^T & 0 \end{bmatrix}$.

The first observation was that the "nice" structure of the $M$ matrix is destroyed immediately after the first pivot when $-e$ vector enters the basis (refer to Figures 1 and 2 of Section 2). So the first occurred idea was to take the $-e$ vector to the right-hand side, perturbate it with the $q$ vector and consider $Z_0$ as a "parameter."

The second observation was that if we work with the "perturbed" problem, the "nice" property of $M$ matrix (i.e., the two submatrices below and above main diagonal are skew symmetric) can be retained by rearranging the columns such that the basis columns (identity matrix) appear first. These observations led to the following modified algorithm which works only with the reduced matrix of $M$ which results in a greater saving in storage and time and answers the question raised about the Lemke method at the beginning of the section. We shall present the modified algorithm completely independent of the complementary pivot theory. But those who are familiar with the complementary pivot theory can see that it is nothing but the Lemke method with a slight modification in the pivot rule and is applied only to the reduced $M$ matrix (viz the transpose of $A$ is never carried in the pivots.)
A Complementary Pivot Algorithm

Consider

\[ \begin{align*}
Ax & \geq b \\
x & \geq 0 \\
\min & \quad cx
\end{align*} \]

where \( A \) is \((m \times n)\) matrix \( c \) and \( x \) are \( n \) vectors \( b \) is \( m \) vector

\[ A = \{a_{ij}\} \]

Let

\[ u_1, \ldots, u_m \text{ denote primal slacks} \]
\[ y_1, \ldots, y_m \text{ denote dual variables} \]
\[ v_1, \ldots, v_n \text{ denote dual slacks} \]

the vectors \( u \) and \( y \) (\( \geq 0 \)) are complementary pair of vectors, i.e., in the solution \( u_i y_i = 0 \quad \forall \ i = 1, \ldots, n \) \( (u_i \text{ and } y_i \text{ are called complementary variables}) \). \( x \) and \( v \) are complementary in the sense \( x_j v_j = 0 \quad \forall \ j = 1, \ldots, n \).

Complementary Rule:

When a variable leaves the basis bring the complement of that variable into basis. Define:

\[ a'_{ij} = -a_{ij} \quad b'_i = -b_i \quad \forall \ i \]

The basic variables are denoted by the symbol "\( * \). The initial tableau is shown below.

The initial basis contains \( u_1, \ldots, u_m, v_1, \ldots, v_n \) (primal slacks and dual
slacks) as basic variables. In the tableau,

\[ d_i = 1 \text{ if } b_i < 0 \quad \text{and} \quad b_i > 0 \]

\[ = 0 \text{ otherwise } \quad i = 1, \ldots, m \]

\[ c_{n+1} = \ldots = c_{n+m} = 0 \]

\[ f_i = 1 \text{ if } c_j < 0 \quad i = 1, 2, \ldots, n, n+1, \ldots, n+m \]

\[ = 0 \text{ otherwise} \]

<table>
<thead>
<tr>
<th>Basic Variables</th>
<th>( u_1 ) \ldots ( u_m ) ( x_1 ) \ldots ( x_n )</th>
<th>( b' )</th>
<th>( d )</th>
</tr>
</thead>
<tbody>
<tr>
<td>* ( u_1 )</td>
<td>1 ( a_{11} ) ( a_{1n} )</td>
<td>( b_1 )</td>
<td>( d_1 )</td>
</tr>
<tr>
<td>* ( u_n )</td>
<td>1 ( a_{m1} ) ( a_{mn} )</td>
<td>( b_m )</td>
<td>( d_m )</td>
</tr>
<tr>
<td>( c_{n+1} ) \ldots ( c_{n+m} ) ( c_1 ) \ldots ( c_n )</td>
<td>( c )</td>
<td></td>
<td></td>
</tr>
<tr>
<td>( f_{n+1} ) \ldots ( f_{n+m} ) ( f_1 ) \ldots ( f_n )</td>
<td>( f )</td>
<td></td>
<td></td>
</tr>
<tr>
<td>( y_1 ) \ldots ( y_m ) ( v_1 ) \ldots ( v_n )</td>
<td>Basic Variables</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Step 1:**

\[ M_1 = \max_{d_i > 0} \left( \frac{-b'_i}{d_i} \right) = \max_{b_i < 0} \left( -b_i \right) \quad i = 1, \ldots, m \]

\[ M_2 = \max_{f_j > 0} \left( \frac{-c_j}{f_j} \right) = \max_{c_j < 0} \left( -c_j \right) \quad j = 1, \ldots, n+m \]
Let $Z_o(1) = \text{Max} \ (M_1, M_2)$ \hfill (4.1)

**Note:**

\[(4.1) \Rightarrow b_i' + Z_o(1)d_i \geq 0 \ \text{and} \ c_i + f_iZ_o(1) \geq 0\]

If $Z_o(1) < 0$ then terminate. Now there are two cases to consider. The max value $Z_o(1)$ may occur

(i) Corresponding to some $j = 1, ..., n + m$.

(ii) Corresponding to some $i = 1, ..., m$.

**Case (i):**

$Z_o(1)$ occurs corresponding to $j = 1 \ (\text{i.e.,} \ Z_o(1) = \frac{-c_1}{f_1})$ so $v_1$ leaves the basis. (Note: max-ratio cannot occur corresponding to any nonbasic "j" as their $f_j = 0$.) Since a dual variable leaves the basis, we apply the following dual complementary pivot.

By our complementary rule, the complementary variable $x_1$ (look up at the top row of the tableau corresponding to $v_1$) enters basis. (Note $x_1$ cannot replace $v_1$.) By the minimum ratio rule find the basic variable being replaced by $x_1$. For this find ratios $\frac{b_i' + Z_o(1)d_i}{a_{1i}}$ for $i = 1, ..., n$ and $a_{1i} > 0$. \hfill (4.2)

Let the minimum of 4.2 occur corresponding to $i = m$. (If no such minimum exists, terminate as we have an unbounded solution (Ray Solution) as $a_{1i}' < 0 \ \forall \ i$.) So from min ratio rule, we find $x_1$ replaces $u_m$ and $y_m$, the complementary variable of $u_m$ replaces $v_1$ from the basis.
Using \( a'_{m1} \) as pivot element, perform the pivot operation. Note that the new basis is shown again by symbol "*" and this concludes a dual complementary pivot.

\[
\begin{array}{cccccccc}
\star \ u_1 & \ldots & u_{m-1} & u_m & x_1 & x_2 & \ldots & x_n \\
\hline
1 & \frac{-a_{11}'}{a_{m1}'} & 0 & \bar{a}_{12} & \bar{a}_{1n} & \bar{b}_1 & \bar{d}_1 \\
\vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\
1 & \frac{-a_{m-1,1}'}{a_{m1}'} & 0 & \bar{a}_{m-1,2} & \bar{a}_{m-1,n} & \bar{b}_{m-1} & \bar{d}_{m-1} \\
\star \ x_1 & 0 & 1/a_{m1}' & 1 & \bar{a}_{m2} & \bar{a}_{m,n} & \bar{b}_m & \bar{d}_m \\
\hline
\bar{c}_{n+1} & \bar{c}_{n+m-1} & \bar{c}_{n+m} & 0 & \bar{c}_2 & \bar{c}_n \\
\hline
\bar{f}_{n+1} & \bar{f}_{n+m-1} & \bar{f}_{n+m} & 0 & \bar{f}_2 & \bar{f}_n \\
\hline
v_1 & y_m & v_1 & v_2 & v_n & \\
\end{array}
\]

where

\[
\bar{a}_{mj} = \frac{a_{mj}'}{a_{m1}'} \\
\bar{a}_{ij} = a_{ij}' - a_{i1}' \left( \frac{a_{m1}'}{a_{m1}'} \right) \\
\bar{b}_m = \frac{b_m'}{a_{m1}'}; \quad \bar{d}_m = \frac{d_m}{a_{m1}'} \\
\bar{b}_i = b_i' - a_{i1}' \frac{b_m'}{a_{m1}'} \\
\bar{d}_i = d_i' - a_{i1}' \frac{d_m}{a_{m1}'}
\]

\((4.3)\)
\[ c_j = c_j - c_1 \left( \frac{a'_{ml}}{a_{ml}} \right) \quad j = 2, \ldots, n + m \]  
\[ f_j = f_j - f_1 \left( \frac{a'_{ml}}{a_{ml}} \right) \quad j = 2, \ldots, n + m \]  
(Note: \( f_1 = 1 \))

Case (ii):

Let max-ratio 4.1 occur corresponding to some \( i = 1 \Rightarrow u_1 \) leaves the basis. Since a primal variable leaves the basis due to 4.1, we apply primal complementary pivot which runs as follows.

By our complementary rule, since \( u_1 \) leaves the basis bring \( y_1 \) into basis which replaces one of the \( v_j \)'s \( j = 1, \ldots, n \). To find the basic variable being replaced by \( y_1 \) apply max-ratio rule as follows: Form the ratios to find

\[ \text{Max} \left( \frac{c_1 + Z_0 (1)f_1}{a'_{1j}} \right) \quad \forall j = 1, \ldots, n \text{ and } a'_{1j} < 0 \]  

(As a matter of fact this should be carried out for all elements of 1st row which are \( < 0 \) but in the first pivot step we know the first \( m \) elements of Row 1 are \( \geq 0 \).)

Let the max of 4.4 occur corresponding to variable \( v_n \). (If no max exists, then terminate. We have an unbounded solution (Ray Solution as all \( a_{ij} < 0 \forall j \).) So \( y_1 \) replaces \( v_n \) and \( x_n \), the complementary variable of \( v_n \), replaces \( u_1 \). Using \( a'_{ln} \) as the pivot element perform the pivot operation as in Case (i) (Note: \( a'_{mn}, n < 0 \))

Step 2:

With new tableau at hand, form the ratios as in Step 1 to find
\[ \bar{M}_1 = \max_{\bar{d}_i > 0} \left( -\frac{b_i}{\bar{d}_i} \right) \quad i = 1, \ldots, m \]

\[ \bar{M}_2 = \max_{\bar{f}_j > 0} \left( -\frac{c_j}{\bar{f}_j} \right) \quad j = 1, \ldots, n + m \]

\[ Z_0(2) = \max(\bar{M}_1, \bar{M}_2). \] If \( Z_0(2) \leq 0 \), then terminate.

**Theorem 4.1:**

The new value of \( Z_0(2) \) (after a primal or dual complementary pivot) is less than or equal to \( Z_0(1) \).

**Proof:**

(a) Consider first Case (i) where we have a dual complementary pivot:

\[ Z_0(1) = \frac{-c_1}{f_1} \quad (4.5) \text{ and } f_1 = 1 \]

and by 4.1,

\[ \frac{-c_1}{f_1} \geq \frac{-b_i}{\bar{d}_i} \quad \forall i \]

and

\[ Z_0(1) = \frac{-c_1}{f_1} > \frac{-b_i}{\bar{d}_i} \quad \forall i \]

To prove \( Z_0(2) \leq Z_0(1) \) we have to show
\[ \tilde{M}_1 < Z_0(1) \quad (4.7) \]

and

\[ \tilde{M}_2 < Z_0(1) \quad (4.8) \]

Now

\[ \tilde{M}_1 = \max_{d_i > 0} \frac{-b_i}{d_i} \quad \forall \ i \]

\textbf{Note:} 4.7 and 4.8 \( \Rightarrow \) theorem for Case (i). From min-ratio rule (4.2)

\[ \frac{b'_i + Z_0(1)d_i}{a'_m} \leq \frac{b'_i + Z_0(1)d_i}{a'_l} \quad \forall \ i \]

\[ \frac{b'_i}{a'_m} - \frac{d}{a'_l} < Z_0(1) \leq b'_i + Z_0(1)d_i \quad \forall \ i \]

\[ (a'_m - \frac{b'_i}{a'_m} a'_l) + Z_0(1) (d_i - \frac{d}{a'_l} a'_l) \geq 0 \quad \forall \ i \]

\[ \bar{b}_i + Z_0(1) \tilde{d}_i \geq 0 \text{ by } (4.3) \]

\[ \frac{-\bar{b}_i}{\tilde{d}_i} \leq Z_0(1) \quad \forall \ i \Rightarrow \tilde{M}_1 < Z_0(1) \Rightarrow 4.7 \]

Similarly

\[ \tilde{M}_2 = \max_{f_j > 0} \left[ \begin{array}{c} -c_j \\ \bar{f}_j \end{array} \right] \]
we will show

\[ \bar{c}_j + \bar{f}_j Z_0(1) \geq 0 \ (\Rightarrow 4.8) \]

by 4.3,

\[ \bar{c}_j + \bar{f}_j Z_0(1) = \left[ c_j - c_1 \left( \frac{a_{\text{m1}}'}{a_{\text{m1}}} \right) \right] + \left[ f_j - f_1 \left( \frac{a_{\text{m1}}'}{a_{\text{m1}}} \right) \right] Z_0(1) = \]

\[ (c_j + Z_0(1)f_j) - \frac{a_{\text{m1}}'}{a_{\text{m1}}} [c_1 + Z_0(1)f_1] \geq 0 \Rightarrow 4.8 . \]

\[ \geq 0 \]

by (4.6)

\[ 0 \]

by (4.5)

Hence after a dual complementary pivot

\[ Z_0(2) \leq Z_0(1) \]

(b) Now consider Case (ii) where we have a primal complementary pivot.

\[ Z_0(1) = -\frac{-b_i'}{d_i} \]

and

\[ (4.1) \Rightarrow Z_0(1) > -\frac{-b_i'}{d_i'} \ \forall \ i \]

\[ \text{and } Z_0(1) > -\frac{-c_i}{\bar{f}_j} \ \forall \ j. \]
We will first show:

\[ \bar{N}_1 \leq Z_o(1) \]

i.e.,

\[ \frac{-\bar{b}_i}{\bar{d}_i} \leq Z_o(1) \quad \forall \ i \]

\[ \bar{b}_i + Z_o(1)\bar{d}_i \geq 0 \quad \forall \ i \]

i.e., \( \bar{b}_i + Z_o(1)\bar{d}_i = \left[ b'_i - a'_i \left( \frac{b'_m}{a'_m} \right) \right] + Z_o(1) \left[ d'_i - a'_i \left( \frac{d'_m}{a'_m} \right) \right] \]

\[ \left[ b'_i + d'_i Z_o(1) \right] - \frac{a'_i}{a'_m} \left[ b'_i + Z_o(1) d'_i \right] \geq 0 \]

by 4.10

= \( \bar{N}_1 \leq Z_o(1) \)

Q.E.D.

To show:

\[ \bar{N}_2 \leq Z_o(1) \]

i.e.,

\[ \frac{-\bar{c}}{\bar{r}_j} \leq Z_o(1) \quad \forall \ j \]
By max-ratio rule (4.4)

\[
\frac{c_j + Z_o(1)f_j}{a_{ij}} < \frac{c_n + Z_o(1)f_n}{a_{in}}
\]

(Note: \( a_{in}^i < 0 \) )

Note:

\[
\bar{c}_j = c_j - c_n \left( \frac{a_i^j}{a_{in}} \right)
\]

\[
\bar{f}_j = f_j - f_n \left( \frac{a_i^j}{a_{in}} \right)
\]

\[
\frac{c_j + Z_o(1)f_j}{-a_{ij}} > \frac{c_n + Z_o(1)f_n}{-a_{in}}
\]

Note \(-a_{in}^i > 0\) and \(-a_{ij}^i > 0\)

\[
\Rightarrow \left[ c_j - c_n \left( \frac{a_i^j}{a_{in}} \right) \right] + \left[ f_j - f_n \frac{a_i^j}{a_{in}} \right] Z_o(1) > 0
\]

\[
\bar{c}_j + \bar{f}_j Z_o(1) > 0
\]

i.e., \( \bar{H}_2 < Z_o(1) \)

Q.E.D.

Hence the theorem is proved.

After finding \( Z_o(2) \), according to the basic variable which leaves the basis either primal or dual complementary pivot is done as in Step 1.
Step n:

Find

\[ Z_o(n) \] .

Since we did not use any special property of first pivot, similar arguments proves

\[ Z_o(n) \leq Z_o(n - 1) . \]

Theorem 4.2:

Terminate if \( Z_o(n) \leq 0 \), and the optimal solutions of both primal and dual
are \( b_1^{(n)} \) and \( c_1^{(n)} \) \( \forall i \).

Proof:

At Step 1, 4.1 \( \Rightarrow \) that

\[ b_1' + d_1 Z_o(1) \geq 0 \quad \forall i \]

\[ c_1' + f_1 Z_o(1) \geq 0 \quad \forall i . \]

At Step \( n - 1 \), the min-ratio rule (4.2) of the dual complementary pivot, if a dual variable leaves the basis or the max-ratio rule (4.4) of the primal complementary pivot, if a primal variable leaves the basis, guarantees that

\[ b_1^{(n)} + d_1^{(n)} Z_o(n - 1) \geq 0 \quad \forall i \]

\[ c_1^{(n)} + f_1^{(n)} Z_o(n - 1) \geq 0 \quad \forall i . \]

Also, we have \( Z_o(n - 1) > 0 \). 4.1 at Step n gives \( Z_o(n) \leq 0 \) and
\[ \Rightarrow b_i^{(n)} + d_i^{(n)} Z_0(n) > 0 \quad \forall i \]
\[ c_i^{(n)} + f_i^{(n)} Z_0(n) > 0 \quad \forall i. \]

So the above must be true for all values of \( Z_0 \) between \( Z_0(n-1) \) and \( Z_0(n) \) because of linearity. Note \( Z_0(n-1) > 0 \) and \( Z_0(n) < 0 \).

Hence it is true for \( Z_0 = 0 \)

\[ \Rightarrow b_i^{(n)} > 0 \quad \forall i \]
\[ c_i^{(n)} > 0 \quad \forall i. \]

Hence the solution we obtain by setting primal basic variables to \( b_i^{(n)} \) is feasible.

Claim:

The dual solution obtained by setting

\[ y_i = c_i^{(n)} \quad i = 1, \ldots, m \]
\[ v_j = c_j^{(n)} \quad j = 1, \ldots, n \]

is feasible for \( yA \leq c, y \geq 0 \) which is the dual L.P. Problem to

\[ Ax \geq b \]
\[ x \geq 0 \]
\[ \text{Min } cx. \]

Proof:

Since \( c_n^{(n)} > 0 \Rightarrow y_1 > 0 \). Note initially at Step \( n = 0 \)

\[ c_n^{(0)} = 0. \]

So \( y_1 \)'s can be considered as multipliers corresponding to the rows of \(-A\).
matrix which gives \( c_j^{(n)} \) for \( i = 1, \ldots, n \), i.e.,

\[
c_j^{(n)} = c_j^{(0)} + y(-a_j) \quad \forall \quad j
\]

where \( a_j \) is \( j \)th column of \( A \). Since

\[
c_j^{(0)} = c_j
\]

and

\[
c_j^{(n)} \geq 0
\]

\( \Rightarrow \) that \( c_j - ya_j \geq 0 \quad \forall \quad j \) or \( ya_j \geq c_j \quad \forall \quad j \Rightarrow y \) is feasible for dual.

Q.E.D. (Claim)

Note our primal and dual complementary pivot rule maintains the complementarity between primal and dual variables, i.e., either \( y_i \) or \( u_i \) is in the basis but not both. Similarly either \( x_i \) or \( v_i \) is in the basis but not both. Hence the complementary slackness is also satisfied by the primal and dual solution got from \( b_i^{(n)} \) and \( c_i^{(n)} \). Hence they are optimal since by claim they are feasible.

Finiteness of the Algorithm:

Under nondegeneracy assumption (i.e., no ties when forming max-ratio or min-ratio rule) \( Z_0(n) \) decreases with \( n \) (proved already), and since we terminate when \( Z_0(n) \leq 0 \) the algorithm clearly ends in finite number of steps.
Comments:

Those who are familiar with the complementary pivot theory, can easily see that in our algorithm we are essentially applying the Lemke's method but to a reduced tableau, taking advantage of the structure of M-matrix.

Dantzig [1] has claimed that the Lemke's complementary pivot method to solve linear programs is identical with respect to the pivot steps to his self-dual parametric algorithm [1]. Though this fact is not very obvious to see, using our algorithm we can see that the pivot steps are identical to those in the self-dual parametric algorithm though Dantzig uses the idea of primal and dual simplex method while we make use of the complementarity between the variables. Since our method is a condensed form of Lemke's method in some sense, we have also shown that Dantzig's claim is true.

Illustration:

Consider the following linear program: [1]

Minimize \( 3x_1 - 3x_2 \)

Subject to \( -2x_1 - 2x_2 \geq -10 \)
\( x_1 - x_2 \geq 1 \)
\( -x_1 + 2x_2 \geq 1 \)
\( x_1, x_2 \geq 0 \)

Using the algorithm of Section 4, we start with the following initial tableau.
Basic variables are designated by the symbol "*"

Step 1

$$M_1 = \max_{d_i > 0} \{+1, +1\} = 1$$

$$M_2 = \max_{f_j > 0} (5) = 3$$

$$Z_0(1) = 3$$ and $$\Rightarrow v_2$$ a dual variable leaves the basis. So we apply dual complementary pivot rule. The complementary variable (looking up the tableau corresponding to $$v_2$$) $$x_2$$ enters the basis and replaces $$u_2$$ by the min. ratio rule as $$\min_{a_{i5} > 0} \left\{ \frac{10}{2}, \frac{-1+1}{1} \right\} = 1$$. Hence $$y_2$$, the complementary variable of $$u_2$$ replaces $$v_2$$ from the basis. The new tableau after the pivot operation is as shown below

<table>
<thead>
<tr>
<th>basis</th>
<th>$$u_1$$</th>
<th>$$u_2$$</th>
<th>$$u_3$$</th>
<th>$$x_1$$</th>
<th>$$x_2$$</th>
<th>$$b'$$</th>
<th>$$d$$</th>
</tr>
</thead>
<tbody>
<tr>
<td>* $$u_1$$</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>2</td>
<td>2</td>
<td>10</td>
<td>0</td>
</tr>
<tr>
<td>* $$u_2$$</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>-1</td>
<td>1</td>
<td>-1</td>
<td>1</td>
</tr>
<tr>
<td>* $$u_3$$</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>-2</td>
<td>-1</td>
<td>1</td>
</tr>
<tr>
<td></td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>3</td>
<td>-3</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Step 2:

\[ M_1 = \text{Max} \{1, 1\} = 1 \]

\[ M_2 = \text{Max} \{0\} = 0 \]

Since we have ties here we just pick one, say, \( u_3 \). So \( u_3 \), a primal variable leaves the basis and we use primal complementary pivot rule by which \( y_3 \) enters the basis and replaces \( v_1 \) as \( \frac{0 + 1}{-1} = -1 \) so \( x_1 \) replaces \( u_3 \). The pivot element is circled. The new tableau after the pivot is as shown below:

<table>
<thead>
<tr>
<th>( u_1 )</th>
<th>( u_2 )</th>
<th>( u_3 )</th>
<th>( x_1 )</th>
<th>( x_2 )</th>
<th>( b )</th>
<th>( d )</th>
</tr>
</thead>
<tbody>
<tr>
<td>* ( u_1 )</td>
<td>1</td>
<td>-2</td>
<td>0</td>
<td>4</td>
<td>0</td>
<td>12</td>
</tr>
<tr>
<td>* ( x_2 )</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>-1</td>
<td>1</td>
<td>-1</td>
</tr>
<tr>
<td>* ( u_3 )</td>
<td>0</td>
<td>2</td>
<td>1</td>
<td>0</td>
<td></td>
<td>3</td>
</tr>
<tr>
<td></td>
<td>0</td>
<td>3</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>c</td>
</tr>
<tr>
<td></td>
<td>0</td>
<td>-1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>f</td>
</tr>
<tr>
<td>( y_1 )</td>
<td>( y_2 )</td>
<td>( y_3 )</td>
<td>( v_1 )</td>
<td>( v_2 )</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Step 3:

\[ M_1 = \begin{bmatrix} -0 \\ 10 \end{bmatrix} = 0 \]

\[ M_2 = \text{Max} \begin{bmatrix} -3 \\ 1 \\ 1 \end{bmatrix} = 0 \]

\[ Z_0(3) = 0 \] terminate optimal basis is shown by * and their values by looking up at the corresponding \( b \) and \( c \) Rows, i.e.,

\[ u_1 = 0 \quad \text{and} \quad y_2 = 3 \]
\[ x_2 = 2 \quad \text{and} \quad y_3 = 0 \]
\[ x_1 = 3 \]

(Primal basis) (Dual basis)

The others \quad The others

\[ x_3 = 0 \quad \text{and} \quad y_1 = 0 \]
\[ u_2 = 0 \quad \text{and} \quad v_1 = 0 \]
\[ v_2 = 0 \]

Minimum = 3.3 - 3.2 = 9.6 = 3
REFERENCES


An experimental study to compare the simplex method and the Lemke's method to solve linear programs is made. The M3 code for simplex method and the author's code for the Lemke method were used in the study. Comparison was made only with regard to the number of iterations each method takes and our little study shows encouraging results about the superiority of Lemke method, but no general recommendation is made by the author due to size of the study and data. A by-product of our study is a complementary pivot algorithm to solve linear programs which is a modification of the Lemke's method and which saves a considerable storage and time of computation.
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