

# COMPLEMENTARY PIVOT THEORY OF MATHEMATICAL PROGRAMMING 

by

Richard W. Cottle ${ }^{1}$ and George B. Dantzig ${ }^{2}$

TECHNICAL REPORT NO. 67-2

April 1967

Operations Research House
Stanford University
Stanford, California
$1_{\text {Research partially supported by National Science Foundation }}$ Grant GP-3739.
${ }^{2}$ Research partially supported by U. S. Army Research Office Contract No. DAHC04-67-C-0028, Office of Naval Research, Contract ONR-N-00014-67-A-0112-0011, U. S. Atomic Energy Commission, Contract No. AT(04-3)-326 PA \#18, and National Science Foundation Grant GP 6431.

Reproduction in whole or in part for any purpose of the United States Government is permitted.

# COMPLEAENTAKY PIVOT THEORY OF MATHEMATICAL PROCRAMMING 

by

## Rlchard W. Cottle and George B. Dantzis

1. Formulation. Linear programaing, quadratic programing, and bimatrix (two-person, non zero-sum) games lead to the consideration of the followins Fundamental Problem: Given a real p-vector $q$ and a real $p \times p$ matrix $M$, find vectors $w$ and $z$ which satisfy the conditions*

$$
w=q+M z, \quad w \geq 0, z \geq 0
$$

$2 w=0$

The remainder of this section is devoted to an explanation of why this is so. (There are other fields in which this fundamental problem arises -- see fr, example [6] and [13] -- but we do not treat them here.) Sections 2 and 3 are concerned with constructive procedures for solving the fundamental problem under various asoumptions on the data $q$ and $M$.

[^0]Consider first linear programs in the symetric primal-dual form due to J. von Neumann (20).

Prial linear program: Find a vector $x$ and ainimum $\bar{z}$ such that (3)

$$
A x \geq b, \quad x \geq 0, \quad \bar{z} \bullet c x
$$

Dusl innear prosram: Find a vector $y$ and aaximum $z$ such that (4)

$$
y A \leq c, \quad y \geq 0, z^{-} y b
$$

The duality theorem of linear programming [3] states that $\min \bar{z}=\max \underline{z}$ when the primal and dual syatems (3) and (4), respectively, are consistent or -- in mathematical programming parlance -- "feasible." Since

$$
\underline{z}=y b \leq y A x \leq c x=\bar{z}
$$

for all primal-feasible $x$ and dual-feasible $y$, one seeks such solutions for which
(5)

$$
y t=c x
$$

The inequality constraints of the primal and dual problems can be converted to equivalent systems of equations in non-negative variables through the introduction of non-negative "slack" variables. Jointly, the systems (3) and (4) are equivalent to
(6)

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

and the linear programing problem becomes one of finding vectore $u, v, x, y$ auch that
(7)

$$
\binom{u}{v}=\binom{c}{-b}+\left(\begin{array}{cc}
0 & -A^{T} \\
A & 0
\end{array}\right)\binom{x}{y} \quad x \geq 0, v \geq 0
$$

and (by (5) )
(8)

$$
x u+y v=0
$$

The definitions
(9)

$$
w=\binom{u}{v}, \quad q=\binom{c}{-b}, \quad M=\left(\begin{array}{cc}
0 & -A^{T} \\
A & 0
\end{array}\right), z=\binom{x}{y}
$$

establish the correspondence between (1), (2) and (3), (4).
The quadratic programming problem is typically stated in the following manner: Find a vector $x$ and uinimum $\bar{z}$ such that

$$
\begin{equation*}
A x \geq b, \quad x \geq 0, \quad \bar{z}=c x+\frac{1}{2} x D x \tag{10}
\end{equation*}
$$

In this formulation, the matrix $D$ may be assumed to be symmetric. The minimand $\bar{z}$ is a globally convex function of $x$ if and only if the quadratic form $x D x$ (or matrix $D$ ) is positive semi-definite, and when this is the case, (10) is called the convex quadratic programming problem. It is immediate that when $D$ is the zero matrix, (10) reduces to the linear program (3). In this sense, the linear programming problem is a special case of the quadratic programming problem.

For any quadratic programming problem (10), dofine $u$ and $v$ by

$$
\begin{equation*}
u=D x-A^{I} y+c{ }^{\prime} v=A x-b \tag{11}
\end{equation*}
$$

A vector $x^{0}$ yields minimum $\bar{z}$ only if there exists a vector $y^{0}$ and vectors $u^{0}, v^{0}$ given by (11) fur $x=x^{0}$ satiafying

$$
\begin{gather*}
x^{0} \geq 0, u^{0} \geq 0, y^{0} \geq 0, v^{0} \geq 0  \tag{12}\\
x^{0} u^{0}=0, \quad y^{0} v^{0}=0
\end{gather*}
$$

These necessary conditions for a minimum in (10) are a direct
consequence of a theorem of H. W. Kuhn and A. W. Tucker [14]. It is well known -- and not difficult to prove from first principles -that (12), known as the Kuhn-Tucker conditions, are also sufficient In the case of convex quadratic programming. By direct substitution, we have for any feasible vector $x$,

$$
\begin{aligned}
\bar{z}-\bar{z}^{0} & =c\left(x-x^{0}\right)+\frac{1}{2} x D x-\frac{1}{2} x^{0} D x^{0} \\
& =u^{0}\left(x-x^{0}\right)+y^{0}\left(v-v^{0}\right)+\frac{1}{2}\left(x-x^{0}\right) D\left(x-x^{0}\right) \\
& =u^{0} x+y^{0} v+\frac{1}{2}\left(x-x^{0}\right) D\left(x-x^{0}\right) \geq 0
\end{aligned}
$$

which proves the sufficiency of conditions (12) for a minimum in the convex case.

Thus, the problem of solving a quadratic program leads to a search for solution of the system

$$
\begin{array}{rll}
u=D x-A^{T} y+c & x \geq 0, y \geq 0  \tag{13}\\
v=A x & -b & u \geq 0, v \geq 0
\end{array}
$$

(14)
$x u+y v=0$

## The definitiuns

$$
w=\binom{u}{v}, \quad q=\binom{c}{-b}, \quad M=\left(\begin{array}{cc}
D & -A^{T}  \tag{15}\\
A & 0
\end{array}\right), \quad z=\binom{x}{y}
$$

establish (13), (14) as a problem of the form (1). (2).
Dual of a convex quadratic proaram. From (15) one is led naturally to the consideration of matrix $M=\left(\begin{array}{cc}D & -A^{T} \\ A & E\end{array}\right)$ wherain $E$, like $D$. is positive semi-definite. It is shown in [1] that the

Primal quadratic program: Find $x$ and minimum $\bar{z}$ such that

$$
\begin{equation*}
A x+E y \geq b, \quad x \geq 0, \quad \bar{z}=c x+\frac{1}{2}(x D x+y E y) \tag{16}
\end{equation*}
$$

## has the associated

Dual quadratic program: Find $y$ and maximum 2 such that

$$
\begin{equation*}
-D x+A^{T} y \leq c, y \geq 0, z=b y-\frac{1}{2}(x D x-y E y) \tag{17}
\end{equation*}
$$

All the results of duality in linear programing extend to these problems, and indeed they are jointly solvable if either is solvable. When $E=0$, the primal problem is just (10) for which W. S. Dorn [5] first established the duality theory later extended in [1]. When both $D$ and $E$ are zero matrices, this dual pair (16), (17) reduces to the dual pair of linear programs (3), (4).

Reques. (a) The minimand in (10) is atrictly convex if and only If the quadratic form $x D x$ is positive definite. Any feantble atrictly convex quadratic program has anique minimising solution $x^{0}$.
(b) When $D$ and $E$ are positive semi-definite (the case of convex quadratic programing), so is

$$
M=\left(\begin{array}{cc}
D & -A^{T} \\
A & E
\end{array}\right)
$$

A biatrix (or two-person nonzero-sun) gene, $\Gamma(A, B)$, is given by a pair of man atrices $A$ and $B$. One party, called the gou pleyg, has pure strategies which are identified with the rows of A . The other party, called the colum player, has $n$ pure strategies which correapond to the columes of B. If the row player uses his ith pure strategy and the column player uses his jth pure strategy, then their respective losese are defined as $a_{i j}$ and $b_{i j}$, respectively. Using mixed strateaies

$$
\begin{aligned}
& x=\left(x_{1}, \ldots x_{m}\right) \geq 0, \sum_{1=1}^{m} x_{1}=1 \\
& y=\left(y_{1}, \ldots, y_{n}\right) \geq 0, \sum_{j=1}^{n} y_{j}=1
\end{aligned}
$$

their expected losses are $x A y$ and $x B y$, respectively. (A component in a mixed atrategy is interpreted as the probability with which the player usea the corresponding pure strategy.)

A pair ( $x^{0}, y^{c}$ ) of mixed strategies is a Nash [19] equilibrium point of $\Gamma(A, B)$ if

$$
\begin{aligned}
& x^{0} A y^{0} \leqslant x A y^{0} \text { al1 nixed atrategiee } x \\
& x^{0} B y^{0} \leqslant x^{0} B y \text { all nixed serategiee } y
\end{aligned}
$$

It is evident (see for example (15)) that if ( $x^{0}, y^{0}$ ) is an equilibrium point of $\Gamma(A, B)$, then it is also an equilibrium point for the game $r\left(A^{\circ} B^{\circ}\right)$ in which

$$
A^{\prime}=\left[a_{i j}+K\right], B^{c}=\left[b_{i j}+L\right]
$$

where $K$ and $L$ are arbitrary scelars. Hence there is no lose of generality in ascuming that $A>0$ and $B>0$, and we shall aake this assumption hereafter.

Next, by letting $e_{k}$ denote the $k$-vector all of whose components are unity, it is easily shown that $\left(x^{0}, y^{0}\right)$ is an equilibrium point of $\Gamma(A, B)$ if and only if

$$
\begin{align*}
& \left(x^{0} A y^{0}\right) e_{m} \leq A y^{0} \quad(A>0)  \tag{18}\\
& \left(x^{0} B y^{0}\right) e_{n} \leq B^{T} x^{0} \quad(B>0)
\end{align*}
$$

This characterization of an equilibrium point leads to a theore which relates the equilibrium-point problem to a system of the form (1), (2). For $A>0$ and $B>0$, if $u^{*}, v^{*}, x^{*}, y^{*}$ is a solution of the system
(20)

$$
\begin{array}{ll}
u=A y-e_{m} & u \geq 0, y \geq 0 \\
v=B^{T} x-e_{n} & v \geq 0, x \geq 0
\end{array}
$$

## (21)

$$
x u+y v=0
$$

then

$$
\left(x^{0}, y^{0}\right)=\left(\frac{x^{*}}{x^{\dagger} e^{m}}, \frac{y^{\star}}{y^{\star} e_{n}}\right)
$$

is an equilibrium point of $\Gamma(A, B)$. Conversely, if $\left(x^{0}, y^{0}\right)$ is an equilibrius point of $\Gamma(A, B)$ then

$$
\left(x^{*}, y^{*}\right)=\left(\frac{x^{0}}{x^{0} B y^{0}}, \frac{y^{0}}{x^{0} A y^{0}}\right)
$$

is a solution of (20), (21). The latter system is clearly of the form (1), (2), where

$$
w=\binom{u}{v}, \quad q=\binom{-e_{m}}{-e_{n}}, \quad M=\left(\begin{array}{ll}
0 & A \\
B & 0
\end{array}\right), \quad z=\binom{x}{y}
$$

Notice that the assumption $A>0, B>0$ precludes the possibility of the matrix $M$ above belonging to the positive semi-definite class.

The existence of an equilibrium point for $\Gamma(A, B)$ was established by J. Nash [19] whose proof employs the Brouwer FixedPoint Theorem. Recently, an elementary constructive proof was discovered by C. E. Lemke and J. T. Howson, Jr. [15].
2. Lemke's iterative solution of the fundamental problem. This section is concerned with the iterative technique of Lemke and Howson for finding equilibrium points of bimatrix games which was later extended by Lemke to the fundamental problem (1), (2). We introduce
first some terminology common to the subject of this section and the next. Consider the syatem of linear equations

$$
\begin{equation*}
w=q+M z \tag{22}
\end{equation*}
$$

where, for the moment, the $p$-vector $q$ and the $p \times p$ matrix $M$ are arbitrary. Both $w$ and $z$ are p-vectors.

For $1=1, \ldots, p$ the corresponding variables $z_{1}$ and $w_{i}$ are called complementary and each is the complement of the other. A complementary solution of (22) is a pair of vectors atisfying (22) and

$$
\begin{equation*}
z_{i} w_{i}=0, \quad 1=1, \ldots, p \tag{23}
\end{equation*}
$$

Notice that a solution $(w ; z)$ of (1), (2) is a nonnegative complementary solution of (22). Finally, a solution of (22) will be called almost-complementary if it satisfies (23) except for one value of 1 , say $1=B$. That is, $z_{B} \neq 0, w_{B} \neq 0$.

In general, the procedure assumes as given an extreme point of the convex set

$$
z=\{z \mid w=q+M z \geq 0, z \geq 0\}
$$

which also happens to be the endpoint of an almost-complementary ray (unbounded edge) of $Z$. Each point of this ray satisfies (23) but for one value of 1 , say $B$. It is not always easy to find such a starting point for an arbitrary $M$. Yet there are two important realizations of the fundamental problem which can be so initiated.

The first is the bimatrix game case to be discussed soon; the second is the case where an entire column of $M$ is positive. The latter property can always be artifically induced by augmenting $M$ with an additional positive column; as we shall see, this turns out to be a useful device for initiating the procedure with a general $M$.

Each iteration corresponds to motion from an extreme point $P_{i}$ along an edge of $Z$ all points of which are almost-complementary solutions of (22). If this edge is bounded, an adjacent extreme point $P_{i+1}$ is reached which is either complementary or almostcomplementary. The process terminates if (1) the edge is unbounded (a ray), (1i) $P_{i+1}$ is a previously generated extreme point, or (iii) $P_{i+1}$ is a complementary extreme point.

Under the assumption of nondegeneracy, the extreme points of $Z$ are in one-to-one correspondence with the basic feasible solutions of (22) (See [3]). Still under this assumption, a complementary basic feasible solution is one in which the complement of each basic variable is nonbasic. The goal is to obtain a basic feasible solution with such a property. In an almost-complementary basic feasible of (23), there will be exactly one index, say $B$, such that both $w_{B}$ and $z_{B}$ are basic variables. Likewise, there will be exactly one index, say $v$, such that both $w_{v}$ and $z_{v}$ are nonbasic variables*.

[^1]An almost-complementary edge is generated by holding all nonbasic variables at value zero and increasing either $z_{v}$ or $w_{v}$ of the nonbasic pair $z_{v}, w_{v}$. There are consequently exactly two almostcomplementary edges associated with an almost-complementary extreme point (corresponding to an almost-complementary basic feasible solution).

Suppose that $z_{v}$ is the nonbasic variable to be increased. The values of the basic variables will change linearly with the changes in $z_{v}$. For sufficiently small positive values of $z_{v}$, the almost-complementary solution remains feasible. This is a consequence of the nondegeneracy assumption. But in order to retain feasibility, the values of the basic variables must be prevented from becoming negative.

If the value of $z_{v}$ can be made arbitrarily large without forcing any basic variable to become negative, then a ray is generated. In this event, the process terminates. However, if some basic variable blocks the increase of $z_{v}$ (i.e. vanishes for a positive value of $z_{v}$ ), then a new basic solution is obtained which is either complementary or almost-complementary. A complementary solution occurs only if a member of the basic pair blocks $z_{V}$. A new almost-complementary extreme point solution is obtained if the blocking occurs otherwise. In the complementary case, we have the desired result: a complementary basic feasible solution. In the almost-complementary case, the nondegeneracy assumption guarantees the uniqueness of the blocking variable. It will become nonbasic in
place of $z_{v}$ and its index becomes the new value of $v$.
The complementary rule. The complement of the (now nonbasic) blocking variable -- or equivalently put, the other member of the "new" nonbasic pair -- is the next nonbasic variable to be increased. The procedure consists of the iteration of these steps. The generated sequence of almost-complementary extreme points and edges is called an almost-complementary path.

THEOREM 1. Along an almost-complementary path, the only almostcomplementary basic feasible solution which can re-occur is the initial one.

PROOF: We assume that all basic feasible solutions of (22) are nondegenerate. (This can be assured by any of the standard lexicographic techniques [3] for resolving the ambiguities of degeneracy.) Suppose, contrary to the assertion of the theorem, that the procedure generates a sequence of almost-complementary basic feasible solutions in which a term other than the first one ( $P_{0}$ in the figure below) is repeated (say $P_{1}$ ). By the nondegeneracy assumption, the extreme points of $Z$ are in one-one correspondence with basic feasible solutions of (22). Let $P_{2}$ denote the successor of $P_{1}$ and let $P_{k}$ denote the second predecessor to $P_{1}$ namely the one along the path just before the return to $P_{1}$.


The extreme points $P_{0}, P_{2}, P_{k}$ are distinct and each is adjacent to $P_{1}$ along an almost-complementary edge. But there are only two such edges at $P_{1}$. This contradiction completes the proof.

We can immediately state the
COROLLARY. If the almost-complementary path is initiated at the endpoint of an almost-complementary ray, the procedure must terminate either in a different ray or a complenentary basic feasible solution.

It is easy to show by examples that starting from an almostcomplementary basic feasible solution which is not the endpoint of an almost-complementary ray, the procedure can return to the initial point regardless of the existence or non-existence of a solution to (1), (2).

EXAMPLE 1. The set $Z$ associated with

$$
q=\left(\begin{array}{r}
1 \\
-1 \\
3
\end{array}\right) \quad M=\left(\begin{array}{rrr}
0 & 0 & 0 \\
1 & 0 & 0 \\
-1 & -1 & -1
\end{array}\right)
$$

is nonempty and bounded. It is clear that no solution of (1) can also satisfy (2) since $z_{1} w_{1}>0$. Let the extreme point corresponding to the solution $w=(1,0,0), z=(1,0,2)$ be the initial point of a path which begins by increasing $z_{2}$. This will return to the initial extreme point after 4 iterations. EXAMPLE 2. The set $Z$ associated with

$$
q=\left(\begin{array}{r}
1 \\
-1 \\
3 \\
1
\end{array}\right) \quad, \quad M=\left(\begin{array}{rrrr}
0 & 0 & 0 & 0 \\
1 & 0 & 0 & 1 \\
-1 & -1 & -1 & -1 \\
0 & 0 & 0 & -1
\end{array}\right)
$$

is likewise nonempty and bounded. The corresponding fundamental problem (1), (2) has a complementary solution $w=(1,0,1,0)$, $z=(0,1,0,1) . \quad$ Yet by starting at $w=(1,2,0,1), \quad z=(3,0,0,0)$ and increasing $2_{3}$, the method generates a path which returns to its starting point after 4 iterations.

Furthermore, even if the procedure is initiated from an extreme point at the end of an almost-complementary ray, termination in a ray is possible whether or not the fundamental problem has a solution. EXAMPLE 3. Given the data

$$
q=\left(\begin{array}{r}
1 \\
-1 \\
3 \\
1
\end{array}\right) \quad, \quad M=\left(\begin{array}{rrrr}
0 & 0 & 0 & 1 \\
1 & 0 & 0 & 1 \\
1 & -1 & -1 & 1 \\
0 & 0 & 0 & -1
\end{array}\right)
$$

the point of $Z$ corresponding to $w=(1,0,4,1), z=(1,0,0,0)$ is at the end of an almost-complementary ray, $w=\left(1, w_{2}, 4+w_{2}, 1\right)$, $z=\left(1+w_{2}, 0,0,0\right)$. Moving along the edge generated by increasing $z_{2}$ leads to a new almost-complementary extreme point at which the required increase of $z_{3}$ is unblocked, so that the process terminates
in a ray, and yet the fundamental probelm is solved by $w=(2,0,1,0), \quad 2=(0,1,0,1)$.

EXAMPLE 4. In the problem with

$$
q=\binom{1}{-1} \quad, \quad M=\left(\begin{array}{cc}
0 & 0 \\
1 & -1
\end{array}\right)
$$

the inequalities (1) have solutions, but none of them satisfy (2). The point corresponding to $(w ; z)=(1,0 ; 1,0)$ is at the end of an almost-complementary ray $w=\left(1, w_{2}\right), z=\left(w_{2}, 0\right)$. When $z_{2}$ is increased, it is not blocked, and the process terminates in a ray. Consequences of termination in a ray. In this geometrical approach to the fundamental problem, it is useful to interpret algebraically the meaning of termination in an almost-complementary ray. This can be achieved by use of a standard result in linear inequality theory [11], [3].

LEMMA. If $\left(w^{*} ; z^{*}\right)$ is an almost-complementary basic feasible solution of (22), and $\left(w^{*} ; z^{*}\right)$ is incident to an almost-complementary ray, there exist $p$-vectors $w^{h}, z^{h}$ such that

$$
\begin{equation*}
w^{h}=M z^{h}, \quad w^{h} \geq 0, \quad z^{h} \geq 0, \quad z^{h} \neq 0 \tag{24}
\end{equation*}
$$

and points along the almost-complementary ray are of the form

$$
\begin{equation*}
\left(w^{*}+\lambda w^{h}, z^{*}+\lambda z^{h}\right) \quad \lambda \geq 0 \tag{25}
\end{equation*}
$$

and satisfy

$$
\begin{equation*}
\left(w_{1}^{\star}+\lambda w_{1}^{h}\right)\left(z^{\star}+\lambda z_{1}^{h}\right)=0 \quad \text { for all } \lambda \geq 0, \text { and all } 1 \notin \beta \tag{26}
\end{equation*}
$$

THEOREM 2. If $M>0$, (22) has a complementary basic feasible solution for any vector $q$.

PROOF. Select $w_{1}, \ldots, w_{p}$ as the basic variables in (22). We may assume that $q \geq 0$ for otherwise $(w ; z)=(q ; 0)$ immediately solves the problem. A starting ray of feasible almost-complementary solutions is generated by taking a sufficiently large value of any nonbasic variable, say $z_{1}$. Reduce $z_{1}$ toward zero until it reaches a value $2_{1}^{0} \geq 0$ at which a unique basic variable (assuming non-degeneracy) becomes zero. An extreme point has then been reached.

The procedure has been initiated in the manner described by the corollary above, and consequently the procedure must terminate either in a complementary basic feasible solution or in an almostcomplementary ray after some basic feasible solution (w; $z^{*}$ ) is reached. We now show that the latter cannot happen. For if it does, conditions (24) - (26) of the lemma obtain with $\beta=1$. Since $M>0$ and $z^{h}-0$, this implies $w^{h}, 0$. Hence by (26), $z_{i}^{*}=z_{i}^{h}=0$ for all $i \neq 1$. Hence the only variables which change with $\lambda$ are ${ }^{2} 1$ and the components of $w$. Therffore the final generated ray is the same as the initiating ray, which contradicts the corollary.

THEOREM 3. A bimatrix game $\Gamma(A, B)$ has an extreme equilibrium point.

PROOF. Initiate the algorithm by choosing the smallest positive value of $x_{1}$, say $x_{1}^{0}$, such that

$$
\begin{equation*}
v=-e_{n}+B_{1}^{T} x_{1}^{0} \geq 0 \tag{27}
\end{equation*}
$$

where $B_{1}^{T}$ is the first column of $B^{T}$. With

$$
v^{o}=-e_{n}+B_{1}^{T} x_{1}^{o}
$$

it follows (assuming nondegeneracy) that $v^{0}$ has exactly one zero component, say the r-th. The ray is generated by choosing as basic variables $x_{1}$ and all the slack variables $u, v$ except for $\mathbf{v}_{\mathbf{r}}$. The complement of $v_{r}$, namely $y_{r}$, is chosen as the nonbasic variable to increase indefinitely. For sufficiently large values of $y_{r}$, the basic variables are all nonnegative and the ray so generated is complementary except possibly $x_{1} u_{1}$ might not equal 0 . Letting $y_{r}$ decrease toward zero, the initial extreme point is obtained for some positive value of $\mathrm{y}_{\mathbf{r}}$.

If the procedure does not terminate in an equilibrium point, then by the corollary, it terminates in an almost-complementary ray. The latter implies the existence of a class of almost-complementary solutions of the form *

$$
\binom{u^{*}+\lambda u^{h}}{v^{*}+\lambda v^{h}}=\binom{-e_{m}}{-e_{n}}+\left(\begin{array}{ll}
0 & A  \tag{28}\\
B^{T} & 0
\end{array}\right)\binom{x^{*}+\lambda x^{h}}{y^{*}+\lambda y^{h}}
$$

* The notational analogy with the previously studied case $M>0$ is obvious.

$$
\left.\begin{array}{ll}
\left(u_{i}^{*}+\lambda u_{i}^{h}\right)\left(x_{i}^{*}+\lambda x_{i}^{h}\right)=0 & \text { all } 1 \neq 1  \tag{29}\\
\left(v_{j}^{*}+\lambda v_{j}^{h}\right)\left(y_{j}^{*}+\lambda y_{j}^{h}\right)=0 & \text { all } j
\end{array}\right\} \text { all } \lambda \geq 0
$$

Assume first that $x^{h} \neq 0$. Then $v^{h}=B^{T} x^{h}>0$. By (30), $y_{j}^{*}+\lambda y_{j}^{h}=0$ for all $j$ and all $\lambda \geq 0$. But then $u^{*}+\lambda u^{h}=-e_{m}<0$, a contradiction. Assume next that $y^{h} \neq 0$ and $x^{h}=0$. Then $u^{h}=A y^{h}>0$. By (29), $x_{i}^{*}=0$ for all $1 \notin 1$; and $x_{1}^{h}=0$ for all 1. Hence $v^{h}=B_{x}{ }^{h}=0$ and $v^{*}$ is the same as $v$ defined by (27) since $x_{1}$ must be at the smallest value in order that ( $u^{*}, v^{*}, x^{\star}, y^{*}$ ) be an extreme-point solution. By the nondegeneracy assumption, only $v_{r}^{*}=0$, and $v_{j}^{*}>0$ for all $f * r$. Hence (30) implies $y_{j}^{*}+\lambda y_{j}^{h}=0$ for all $j \neq r$. It is now clear that the postulated terminating ray is the original ray. This furnishes the desired contradiction. The algorithm must terminate in an equilibrium point of the bimatrix game $\Gamma(A, B)$.

A modification of almost-complementary basic sets. Consider the system of equations

$$
\begin{equation*}
w=q+e_{p} z_{0}+M z \tag{31}
\end{equation*}
$$

where $z_{0}$ represents an "artifical variable" aind $e_{p}$ is a p-vector ( $1, \ldots, 1$ ). It is clear that (31) always has nonnegative solutions.

A solution of (31) is called almost-complementary if
$z_{i} W_{1}=0$ for $1=1, \ldots, p$ and is complementary if, in addition,
$z_{o}=0$. (See [16, p. 685] where a different but equivalent
definition is given.) In this case, let

$$
z_{0}=\left\{\left(z_{0}, z\right) \mid w=q+e_{p} z_{0}+M z \geq 0, z_{0} \geq 0, z \geq 0\right\}
$$

We consider the almost-complementary ray generated by sufficiently large $z_{0}$. The variables $w_{1}, \ldots, w_{p}$ are initially basic while $z_{0}, z_{1}, \ldots, z_{p}$ are nonbasic variables. For a sufficiently large value of $z_{0}$, say $z_{0}^{+}$,

$$
w^{+}=q+e_{p} z_{0}^{+}, 0
$$

As $z_{0}$ decreases toward zero, the basic variables $w_{i}$ decrease. An initial extreme point is reached when $z_{0}$ attains the minimum value $z_{0}^{0}$ for which $w=q+e_{p} z_{0} \geq 0$. If $z_{0}^{0}=0$, then $q \geq 0$; this is the trivial case for which no algorithm is required. If $z_{o}^{o}>0$, some unique basic variable, say $w_{r}$ has reached its lower bcund 0. Then $z_{o}$ becomes a basif variable in place of $w_{r}$ and we have $v=r$. Next, ${ }^{2}{ }_{r}$, the complement of $W_{r}$, is to be increased.

The remaining steps of the procedure are now identical to those in the preceding algorithm. After a blocking variable tecomes basic, its complement is increased until either a basic variable blocks the increase (by attaining its lower bound 0 ) or else an almostcomplementary ray is generated. There are precisely two forms of termination. One is in a ray as just described; the other is in the reduction of $z_{o}$ to the value 0 and hence the attainment of a complementary basic feasible solution of (31), i.e. a solution of

## (1), (2).

Interest now centers on the meaning of termination in an almostcomplementary ray solution of (31). For certain classes of matrices, the process described above terminates in an almost-complementary
ray if and only if the original system (1) has no solution. In the remainder of this section, we shall amplify the preceding statement.

If termination in an almost-complementary ray occurs after the process reaches a basic feasible solution ( $w^{*} ; z_{0}^{*}, z^{*}$ ) corresponding to an extreme point of $Z_{o}$, then there exists a nonzero vector ( $w^{h} ; z_{0}^{h}, z^{h}$ ) such that

$$
\begin{equation*}
w^{h}=e_{p} z_{0}^{h}+M z^{h}, \quad\left(w^{h} ; z_{0}^{h}, z^{h}\right) \geq 0 \tag{32}
\end{equation*}
$$

Moreover for every i $\therefore 0$,
(33)

$$
\left(w^{\star}+\lambda w^{h}\right)=q+e_{p}\left(z_{0}^{\star}+\lambda z_{0}^{h}\right)+M\left(z^{\star}+\lambda z^{h}\right)
$$

and

$$
\begin{equation*}
\left(w_{i}^{\star}+\lambda z_{i}^{h}\right)\left(z_{i}^{*}+\lambda z_{i}^{h}\right)=0 \quad 1=1, \ldots, p . \tag{34}
\end{equation*}
$$

The case $z^{h}=0$ is ruled out, for otherwise $z_{0}^{h}>0$ and then $w^{h}$. 0 because $\left(w^{h} ; z_{0}^{h}, z^{h}\right) \neq 0$. Now if $w^{h} \geqslant 0$, (34) implies $z^{*}+\lambda z^{h}-z^{*}=0$. This, in turn, implies that the ray is the original one which is not possible.

Furthermore, it follows from the almost-complementarity of solutions along the ray that

$$
\begin{equation*}
z_{i}^{*} w_{1}^{*}=z_{1}^{*} w_{1}^{h}=2_{i}^{h} w_{1}^{*}=2_{1}^{h} w_{1}^{h}=0 \quad 1=1, \ldots, p . \tag{35}
\end{equation*}
$$

The individual equations of the system (32) are of the form

$$
\begin{equation*}
w_{i}^{h}=z_{0}^{h}+\left(M z^{h}\right)_{1} \quad 1=1, \ldots, p \tag{36}
\end{equation*}
$$

Multipjication of (36) by $z_{i}^{h}$ leads, via (35), to

$$
\begin{equation*}
0=z_{i}^{h} z_{0}^{h}+z_{i}^{h}\left(M z^{h}\right)_{i} \quad i=1, \ldots, p \tag{37}
\end{equation*}
$$

from which we conclude
THEOREM 4: Termination in a ray implies there exists a nonzoro nonnegative vector $z^{h}$ such that

$$
\begin{equation*}
z_{i}^{h}\left(M z^{h}\right)_{1} \leq 0 \quad 1=1, \ldots, p \tag{38}
\end{equation*}
$$

At this functure, two large classes of matrices $M$ will be considered. For the first class, we show that termination in a ray implies the inconsistency of the system (1). For the second class, we will show that termination in a ray cannot occur, so that for this class of matrices, (1), (2) always has a solution regardless of what $q$ is.

The first class mentioned above was introduced by Lemke [16]. These matrices, which we shall refer to as copositive plus, are required to satisfy the two conditions.

$$
\begin{equation*}
u M u \quad 0 \text { for all } u \geq 0 \tag{39}
\end{equation*}
$$

$$
\begin{equation*}
\left(M+M^{T}\right) u=0 \text { if } u M u=0 \text { and } u \geq 0 \tag{40}
\end{equation*}
$$

Matrices satisfying conditions (39) alone are known in the literature as copositive (see [18], [12].) To our knowledge, there is no reference other than [16] on copositive matrices satisfying the condition (40). However, the class of such matrices is large and includes
(i) all strictly copositive matrices, i.e. those for which $u M u \geq 0$ when $0 \notin u \geq 0$
(ii) all positive semi-definite matrices, i.e. those for which $u M u$ - 0 for all $u$.

Positive matrices are obviously strictly copositive while positive definite matrices are both positive semi-definite and strictly copositive. Furthermore, it is possible to "build" matrices satisfying (39) and (40) out of smaller ones. For example, if $M_{1}$ and $M_{2}$ are matrices satisfying (39) and (40) then so is the block-diagonal matrix

$$
M=\left(\begin{array}{ll}
M_{1} & 0 \\
0 & M_{2}
\end{array}\right)
$$

Moreover, if $M$ satisfies (39) and (40) and $S$ is any skewsymmetric matrix (of its order), then $M+S$ satisfies (39) and (40). Consequently, block matrices such as

$$
M=\left(\begin{array}{cc}
M_{1} & -A^{T} \\
A & M_{2}
\end{array}\right)
$$

satisfy (39) and (40) if and only if $M_{1}$ and $M_{2}$ do too. However, as Lemke [16], [17] has pointed out, the matrices encountered in the bimatrix game problem with $A, 0$ and $B \quad 0$ need not satisfy (40). The Lemke-Howson iterative procedure for bimatrix games was given earlier in this section. If applied to bimatrix games, the modification just given always terminates in a ray after just one iteration, as can be verified by taking any example.

The second class, consisting of matrices having positive principal minors, has been studied by numerous investigators; see for example, [2], [4], [8], [9], [10], [22], [24]. In the case of symmetric matrices, those with positive principal minors are positive definite. But the equivalence breaks down in the non-symmetric situation. Nonsymmetric matrices with positive principal minors need not be positive definite. For example, the matrix

$$
\left(\begin{array}{rr}
2 & -7 \\
-1 & 4
\end{array}\right)
$$

[^2]We shall make use of the fact that $w=q+M z, \quad(w ; z) \geq 0$ has no solution if there exists a vector $v$ such that

$$
\begin{equation*}
v M \leq 0, \quad v q \cdot 0, \quad v \simeq 0 \tag{41}
\end{equation*}
$$

for otherwise, $0 \leq v w=v q+v M z<0$, a contradiction. Indeed, it is a consequence of J. Farkas' theorem [7] that (1) has no solution if and only if there exists a solution of (4).

THEOREM 5. Let $M$ be copositive plus. If the iterative procedure terminates in a ray, then (1) has no solution.

PROOF. Termination in a ray means that a basic feasible solution $\left(w^{*} ; z_{0}^{*}, z^{*}\right)$ will be reached at which conditions (32) - (34) hold and also

$$
\begin{equation*}
0=z^{h} w^{h}=z^{h} e_{p} z_{o}^{h}+z^{h} M z^{h} \tag{42}
\end{equation*}
$$

Since $M$ is copositive and $z^{h} \simeq 0$, both terms on the right side of (42) are nonnegative, hence both are zero. The scalar $z_{o}^{h}=0$ because $z^{h} e_{p}, 0$. The vanishing of the quadratic form $z^{h} \mathrm{Mz}^{h}$ means

$$
M z^{h}+M_{z^{T}}=0
$$

But by (32), $z_{0}^{h}=0$ implies that $w^{h}=M z^{h}-0$, whence $M^{T} z^{h} \leq 0$ or, what is the same thing, $z^{h} M=0$. Next, by (35),

$$
0=z^{*} W^{h}=z^{*} M z^{h}=z^{*}\left(-M^{T} z^{h}\right)=-z^{h} M z^{*}
$$

and we obtain again by (35)

$$
0=z^{h} w^{*}=z^{h} q_{q}+z^{h} e_{p} z_{0}^{*}+z^{h} M_{z}^{*}=z^{h} q^{h}+z^{h} e_{p} z_{0}^{*}
$$

It follows that $z^{h} q_{q}<0$ because $z^{h} e_{p} z_{0}^{*}, 0$. The conditions (1) are therefore inconsistent because $v=z^{h}$ satisfies (41). COROLLARY. if $M$ is strictly copositive, the process terminates in a complementary basic feasible solution of (31). PROOF. If not, the proof of theorem 5 would imply the existence of a vector $z^{h}$ satisfying $z^{h} M z^{h}=0,0 \notin z^{h} \geq 0$ which contradicts the strict copositivity of $M$.

This corollary clearly generalizes Theorem 1 . We now turn to the matrices $M$ having positive principal minors.

THEOREM 6. If $M$ has positive principal minors, the process terminates in a complementary basic solution of (31) for any $q$. PROOF. We have seen that termination in a ray inplies the existence of a nonzero vector $z^{h}$ satisfying the inequalities (38). However, Gale and Nikaido [10, Theorem 2] have shown that matrices with positive principal minors are characterized by the impossibility of this event. Hence termination in a ray is not a possible outcome for problems in which $M$ has positive principal minors.

We can even improve upon this.
THEOREM 7. If $M$ has the property that for each of its principal submatrices $\tilde{M}$, the system

$$
\tilde{M} \tilde{z}=0, \quad 0 \neq \tilde{z}-0
$$

has no solution, then the process terminates in a complementary basic solution of (31) for any $q$.

PROOF. Suppose the process terminates in a ray. From the solution ( $w^{h} ; z_{0}^{h}, z^{h}$ ) of the homogeneous system (32), define the vector $\dot{\omega}^{h}$ of components of $w^{h}$ for which the corresponding component of $z^{*}+z^{h}$ is positive. Then by (34) $\mathfrak{w}^{h}=0$. Let $z^{h}$ be the vector of corresponding components in $z^{h}$. Clearly $0 \notin z^{h} \geq 0$, since $0 \neq z^{h}-0$ and any positive component of $z^{h}$ is a positive component of $\dot{z}^{\text {h }}$ by definition of $\dot{w}^{\text {h }}$. Let $\dot{M}$ be the corresponding principal submatrix of $M$. Since $\dot{M}$ is a matrix of order $k \geq 1$ we may write

Hence

$$
0=w^{h}=e_{k} z_{o}^{h}+\check{M}_{2}^{h}
$$

$$
\tilde{M} z^{h}-0, \quad 0 \not \equiv \tilde{z}^{h} \geq 0
$$

which is a contradiction.
3. The principal pivoting method. We shall now describe an algorithm proposed by the authors [4] which predates that of Lemke. It evolved from a quadratic programming algorithm of $P$. Wolfe [26] who was the first to use a type of complementary rule for pivot choice. Our method is applicable to matrices $M$ that have positive principal minors (in particular to positive definite matrices) and after a minor modification, to positive semi-definite matrices.

In Lemke's procedure for general $M$, an artifical variable $z_{0}$ is introduced in order to obtain feasible almost-complementary solutions for the augmented problem. In our approach, only variables of the original problem are used, but these can take on initially
negative as well as non-negative values.

A major cycle of the algorithm is initiated with the complementary basic solution $(w ; z)=(q ; 0)$. If $q \perp 0$, the procedure is immediately terminated. If $q!0$, we may assume (relabeling if necessary) that $w_{1}=q_{1}<0$. An almost-complementary path is generated by increasing $z_{1}$, the complement of the selected negative basic variable. For points along the path, $z_{i} w_{i}=0$ for $i \neq 1$.

Step I. Increase $z_{1}$ untilit is blocked by a positive basic variable decreasing to zero or by the negative $w_{1}$ increasing to zero.

Step II. Make the blocking variable nonbasic by pivoting its complement into the basic set. The major cycle is terminated if $w_{1}$ drops out of the basic set of variables. Otherwise, return to Step I.

It will be shown that during a major cycle $w_{1}$ increases to zero. At this point, a new complementary basic solution is obtained. However, the number of basic variables with negative values is at least one less than at the beginning of the major cycle. Since there are at most $p$ negative basic variables, no more than $p$ major cycles are required to obtain a complementary feasible solution of (22). The proof depends on certain properties of matrices invariant under principal pivoting.

Principal pivot transform of a matrix. Consider the homogeneous system $v=M u$ where $M$ is a square matrix. Here the variables $v_{1}, \ldots, v_{p}$ are basic and expressed in terms of the nonbasic variables
$u_{1}, \ldots, u_{p}$. Let any subset of the $v_{i}$ be made nonbasic and the corresponding $u_{i}$ basic. Relable the full set of basic variables $\bar{v}$ and the corresponding nonbasic variables $\bar{u}$. Let $\bar{v}=\bar{M} \bar{u}$ express the new basic variables $\bar{v}$ in terms of the nonbasic ones. The matrix $\bar{M}$ is called a principal pivot transform of $M$. of course, this transformation can be carried out only if the principal submatrix of $M$ corresponding to the set of variables $z_{i}$ and $w_{1}$ interchanged is nonsingular, and this will be assumed whenever the term is used.

THEOREM 8. (Tucker [24]). If a square matrix $M$ has positive principal minors, so does every principal pivot transform of $M$.

The proof of this theorem is easily obtained inductively by exchanging the roles of one complementary pair and evaluating the resulting principal minors in terms of those of $M$.

THEOREM 9. If a matrix $M$ is positive definite or positive semidefinite so is every principal pivot transform of $M$. PROOF. The original proof given by the authors was along the lines of that for the preceding theorem. P. Wolfe has suggested the following elegant proof. Consider $v=$ Mu. After the principal pivot transformation, let $\bar{v}=\bar{M} \bar{u}$, where $\bar{u}$ is the new set of nonbasic variables. We wish to show that $\overline{u M} \bar{u}=\bar{u} \bar{v}>0$ if $u M u=u v, 0$. If $M$ is positive definite, the latter is true if $u \neq 0$, and the former must hold because every pair $\left(\bar{u}_{1}, \bar{v}_{i}\right)$ is identical with $\left(u_{i}, v_{i}\right)$ except possibly in reverse order. Hence $\sum_{i} \bar{u}_{i} \bar{v}_{i}=\sum_{i} u_{i} v_{i} \quad 0$. The proof in the semi-definite case replaces the
inequality $>$ by - .
Validity of the algorithm. The proof given below for $p=3$ goes through for general p. Consider

$$
\begin{aligned}
w_{1} & =q_{1}+m_{11} z_{1}+m_{12} z_{2}+m_{13} z_{3} \\
w_{2} \quad & =q_{2}+m_{21} z_{1}+m_{22} z_{2}+m_{23} z_{3} \\
w_{3} & =q_{3}+m_{31} z_{1}+m_{32} z_{2}+m_{33} z_{3}
\end{aligned}
$$

Suppose that $M$ has positive principal mirors so that the diagonal coefficients are ali positive:

$$
m_{11}>0, \quad m_{22}>0, \quad m_{33}>0
$$

Suppose furthermore that some $q_{i}$ is negative, say $q_{1}<0$. Then the solution $(w ; z)=\left(q_{1}, q_{2}, q_{3} ; 0,0,0\right)$ is complementary, but not feasible because a particular variable, in this case $w_{1}$, which we refer to as distinguished is negative. We now initiate an almostcomplementary path by increasing the complement of the distinguished variable, in this case $z_{1}$, which we call the driving variable. Adjusting the basic variables, we have

$$
(w ; z)^{1}=\left(q_{1}+m_{11} z_{1}, q_{2}+m_{21} z_{1}, q_{3}+m_{31} z_{1} ; 0,0,0\right)
$$

Note that the distinguished variable $w_{1}$ increases strictly with the increase of the driving variable $z_{1}$ because $m_{11}>0$. Assuming nondegeneracy, we can increase $z_{1}$ by a positive smount before it is blocked either by $w_{1}$ reaching zero or by a basic
variable that was positive and is now turning negative.
In the former case, for some positive value $z_{1}^{*}$ of the driving variable $z_{1}$, we have $w_{1}=q_{1}+m_{11} z_{1}^{*}=0$. The solution

$$
(w ; z)^{2}=\left(0, q_{2}+m_{21} z_{1}^{\star}, q_{3}+m_{31} z_{1}^{*} ; 0,0,0\right)
$$

18 complementary and has one less negative component. Pivoting on $m_{11}$, replaces $w_{1}$ by $z_{1}$ as a basic variable. By Theorem 8 , the matrix $\vec{M}$ in the new canonical system relabeled $\bar{w}=\bar{q}+\vec{M} \bar{z}$ has positive principal minors, allowing the entire major cycle to be repeated.

In the latter case, we have some other basic variable, say $w_{2}=q_{2}+m_{21} z_{1}$ blocking when $z_{1}=z_{1}^{*} ; 0$. Then clearly $m_{21}<0$ and $q_{2}, 0$. In this case,

$$
(w ; z)^{2}=\left(m_{11} z_{1}^{*}+q_{1}, 0, m_{31} z_{1}^{*}+q_{3} ; z_{1}^{*}, 0,0\right)
$$

THEOREM 10. If the driving variable is blocked by a basic variable other than its complement, a principal pivot exchanging the blocking variable with its complement will pezmit the further increase of the driving variable.

PROOF: Pivoting on $m_{22}$ generates the canonical system

$$
\begin{aligned}
w_{1} & =\bar{q}_{1}+\bar{m}_{11} z_{1}+\bar{m}_{12} w_{2}+\bar{m}_{13} z_{3} \\
z_{2} & =\bar{q}_{2}+\bar{m}_{21} z_{1}+\bar{m}_{22} w_{2}+\bar{m}_{23} z_{3} \\
& =\bar{w}_{3}+\bar{m}_{31} z_{1}+\bar{m}_{32} w_{2}+\bar{m}_{33} z_{3}
\end{aligned}
$$

The solution $(w ; z)^{2}$ must satisfy the above since it is an equivalent system. Therefore setting $z_{1}=z_{1}^{*}, w_{2}=0, z_{3}=0$ yields

$$
(w ; z)^{2}=\left(q_{1}+\bar{m}_{11} z_{1}^{\star}, 0, \quad u_{3}+\bar{m}_{31} z_{1}^{\star} ; z_{1}^{\star}, 0,0\right)
$$

1.e., the same almost-complementary solution. Increasing $\mathbf{2}_{1}$ beyond $z_{1}^{*}$ yields

$$
\left(\bar{q}_{1}+\bar{m}_{11} z_{1}, 0, \bar{q}_{3}+m_{31} z_{1} ; z_{1}, 0,0\right)
$$

which is also almost-complementary. The sign of $\bar{m}_{21}$ is the reverse of $m_{21}$, since $\bar{m}_{21}=-m_{21} / m_{22}>0$. Hence ${ }^{2} 2$ increases with increasing $z_{1} ; z_{1}^{*}$; 1.e., the new basic variable replacing $w_{2}$ is not blocking. Since $\bar{M}$ has positive principal minors, $\bar{m}_{11}>0$. Hence $w_{1}$ continues to increase with increasing $z_{1}>z_{1}^{*}$. THEOREM 11. The number of iterations within a major cycle is finite. PROOF: There are only finitely many possible bases. No basis can be repeated with a larger value of ${ }^{2} 1^{\text {. }}$. To see this, suppose it did for ${ }_{2}^{* *}, z_{1}^{*}$. This would imply that some component of the solution turns negative at $z_{1}=z_{1}^{*}$ and yet is nonnegative when $z_{1}=z_{1}^{* *}$. Since the value of a component is linear in $z_{1}$ we nave a contradiction.

Paraphrase of the principal pivoting method. Along the almostcomplementary path there is only one degree of freedom. In the proof of the validity of the algorithm, $z_{1}$ was increasing and $z_{2}$ was shown to increase. The same class of solutions can be generated
by regarding ${ }^{2}{ }_{2}$ as the driving variable and the other variables as adjusting. Hence within each major cycle, the same almostcomplementary path can be generated as follows. The first edge is obtained by using the complement of the distinguished variable as the driving variable. As soon as the driving variable is blocked, the following steps are iterated:
a) replace the blocking variable by the driving variable and terminate the major cycle if the blocking variable is distinguished; if the blocking variable is not distinguished.
b) let the complement of the blocking variable be the new driving variable and increase it until a new blocking variable is identified; return to a).

The paraphrase form is used in practice.
THEOREM 12. The principal pivoting method terminates in a solution of (1), (2) if $M$ has positive principal minors (and, in particular, if $M$ is positive definite).

PROOF. We have shown that the completion of a major cycle occurs In a finite number of steps, and each one reduces the total number of variables with negative values. Hence in a finite number of steps, this total is reduced to zero and a solution of the fundamental problem (1), (2) is obtained. Since a positive definite matrix has positive principal minors, the method applies to such matrices. As indicated earlier, the positive semidefinite case can be handled by using the paraphrase form of the algorithm with a minor modification. The reader will find details in [4].

## REFERENCES

[1] Cottle, R. W., Symmetric dual quadratic programs, Quart. Appl. Math. 21 (1963), 237-243.
[2] Cottle, R. W., Nonlinear programs with positively bounded Jacobians, J. SIAM Appl. Math. 14 (1966), 147-158.
[3] Dantzig, G. B., Linear programming and extensions, Princeton University Press, Princeton, 1963.
[4] Dantzig, G. B. and Cottle, R. W., Positive (semi-) definite programming, ORC 63-18 (RR), May 1963, Operations Research Center, University of California, Berkeley. Revised in Nonlinear programming (J. Ahadie, ed.), North-Holland, Amsterdam, 1967, 55-73.
[5] Dorn, W. S., Duality in quadratif programming, Quart. App1. Math. 18 (1960), 155-162.
[6] Du Val, P., The unloading problem for plane curves, Amer. J. Math. 62 (1940), 307-311.
[7] Farkas, J., Theorie der einfachen Ungleichungen, J. Reine Angew. Math. 124 (1902), 1-27.
[8] Fiedler, M. and Ptak, V., On matrices with non-positive offdiagonal elements and positive principal mınors, Czech. Math. Journal 12 (1962), 382-400
[9] Fiedler, M. and Ptak, V., Some generalizations of positive definiteness and monotonicity, Numerische Math. 9 (1966), 163-172.
[10] Gale, D. and Nikaido, H., The Jacobian matrix and global univalence of mappings, Math. Ann. 159 (1965), 81-93.
[11] Goldman, A. J., "Resolution and separation theorems for polyhedral convex sets," in Linear inequalities and related systems, (H. W. Kuhn and A. W. Tucker, eds.) Princeton University Press, Princeton, 1956.
[12] Hall, M., Jr. and Newman, M., Copositive and completely positive quadratic forms, Proc. Camb. Phil. Soc. 59 (1963), 329-339.
[13] Kilmister, C. W. and Reeve, J. E., Kational mechanics, American Elsevier, New York, 1966, § 5.4.
[14] Kuhn, H. W. and Tucker, A. W., "Nonlinear programming" in Second Berkeley symposium on mathematical statistics and probability (J. Neyman, ed.) University of California Press, Berkeley, 1951.
[15] Lemke, C. E. and Howson, J. T., Jr. Equilibrium points of bimatrix games, J. Soc. Indust. App1. Math. 12 (1964), 413-423.
[16] Lemke, C. E., Bimatrix equilibrium points and mathematical programming, Management Sci. 11 (1965), 681-689.
[17] Lemke, C. E., Private communication.
[18] Motzkin, T. S., Copositive quadratic forms, Nat. Bur. Standards Report 1818 (1952), 11-12.
[19] Nash, J. F., Noncooperative games, Ann. Math. 54 (1951), 286-295.
[20] von Neumann, J., "Discussion of a maximum problem," Collected Works VI (A. H. Taub, ed.), Pergamon Press, New York, 1963.
[21] van de Panne, C. and Whinston, A., A comparison of two methods
for quadratic programming, Operations Res. 14 (1966), 422-441.
[22] Parson, T. D., A combinatorial approach to convex quadratic programming, Doctoral Dissertation, Department of Mathematics, Princeton University, May 1966.
[23] Tucker, A. W., "A combinatorial equivalence of matrices," Proceedings of Symposia in Applied Mathematics, Vol. 10
(R. Bellman and M. Hall, eds.) American Mathematical Society, 1960.
[24] Tucker, A. W., Principal pivotal transforms of square matrices, SIAM Review 5 (1963), p. 305.
[25] Tucker, A. W., Pivotal Algebra, Lecture notes (by T. D. Parsons), Department of Mathematics, Princeton University, 1965.
[26] Wolfe, P., The simplex method for quadratic programing, Econometrica 27 (1959), 382-398.

play a fundamental role in mathematical programming. This paper describes the role of such problems in linear programming, quadratic programming and bimatrix game theory and reviews the computational procedures of lemke and Howson, Lemke, and Dantzig and Cottle.

Unclassified
Security Claseification

| 14. KEY wonos | LINKA |  | LINK |  | CINK C |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | note | Wr | nols | ш ${ }^{\top}$ | HOL | WT |
| Mathematical Programming <br> Linear inequalities <br> Matrix Theory <br> Pivot Theory <br> Complementary slackness |  |  |  |  |  |  |

## INETRUCTIONS

 of the contractor, eubcoatracter, fremee, Depertment of iv fense activity or other organdsation (corporefe suthor) iseming the report.
2a. REPORT SECURTY CLASSIFICATION: Enter the overall eecurlty claesification of the report. Indicate wherher "Restricted Date" If included Marking ie to be in accordence with appropriste security resulatlone
26. GROUP: Automatlc downerading is epecified in DOD DFrective 5200.10 and Armed Forces Industrial Mamual. Enter the group number. Also, when applicable, show that optional markings have been used for Group 3 and Group 4 as avhor* marki
3. REPORT TITLE: Enter the complete report itie in all capltal lettera. Tiues in all cases ahould be unclaselfied. If a meaningful title cnnnot be melected without classificer tlon, show title classification in ". capitale in parenthesis Immediately following the tille.
4. DESCRIPTIVE NOTES: If appropriate, enter the type of report, e. E., Interlm, procrese, summary, annual, or final. Give the inclusive dates when a epecific reporting period Is covered.
5. AUTHOR(S): Enter the name( $(0)$ of author( () as shown on or in the zeport. Enter last name, firat name, midde Initial. If millitary, ahow rank and branch of errvice. The name of It military, thow rank and branch of asrvice. The name of
the principal author is an absolute minimum requirement
6. REPORT DATI: Enter the date of the zeport ac day, month, year, or month, year. If more then one dete eppears on the report, use dete of publication
7e. TOTAL NUMBER OF PAGES: The total page count chould follow normal pagination procedures, be., erter the number of pages containing Information
7b. NUMIBER OF REFERENCES Enter the total number of reforences cited in the report.
8a. CONTRACT OR GRANT NUMBER: If eppropilate, enter the applicable number of the contract or grant under which the report wae writter
86, 8c, \& 8d. PROJECT NUMBER: Enter the approppiate military department identification, ouch as project number, ubbproject number, ayatem numbere, taek number, otc.
9a. ORIGINATOR'S REPORT NUMBER(8): Ent er the off.
clal report number by which the document will be idenified and controlled by the orielnating activity. Tha mumber muet be unique to this report.
9b. OTHER REPORT NUMBER(8): If the report has been asiened any other report numbers (olther by the ort tinator or by the aponeor), aleo enter this number(s).
10. AVAILABILITY/LIMTTATION NOTICEE Eater any Ils ltations on further diseemination of the report, other than those
(1) "QualMbind meroesert may obtala coples of the repert frem breio
(2) "Porelon annomecenert and diescalantion of thit report by DDC it mot anthorised"
(3) "U. \& Coverent metoles may obtain ceples of this report exventy in PRC. Sther qual uled DDC users ohall reareet theont
(4) "U. 8. milltary agencies any obeain coples of this report directly from DDC. Other qualified veere chall sequest through
5) "All diatribution of thise report te controlied Oecl. ifled DDC ustere thell request through

If the repent has been furntehed to the Onsee of Tectateel ervices, Department of Commerce, for eate to tie pitilc, ind cate thie fact and ent the price, if teeom
11. SUPPLEMENTARY NOTES: Une for anmeal enpleac tory notes.
12 SPONEORING MILITARY ACITVIE: PHer the mame of the depertmental project office or laborchery epenmortan (pay Ind for) the research and development incluate adirese.
13. ABSTRACT: Enter an abstract efving abref ad lactual cummary of the docusent indicalive of the repert, even themen it may alco appear elsewhere in the body of the techalcel on. port. If addjional apace to requisod, a coatinuation abect efmeit be attached.

It is highly desirsble thet the abetract of clacelfied reperte be uncleseified. Each paragraph of the abetract shall end with en indication of the millitary securlty clayeification of the taformation in the paragraph, represeated as (TI), (S). (C). or (U).

There is no Liatitation on the leagth of the abetract. How. ever, the ruccented length is from 150 to 225 words.
14. KEY wORDS: Key worda are techaically meanagful torme of short phrases that charactarize a regort and any be maed at Lider entries for cataloging the report. Key words eust be eelected so that no esecurity claesilicetion is mequirel. Lemathfiere, such an equipment model desimation, trede mate, flitery project code name, geographic location, may be uned ae tey words but will be followed by an indicetion of techalcal cee. text. The assignment of linke, rales, and welehte le eptlemal.


[^0]:    * In general, capital roman letters denote matrices while vectors are denoted by lower case roman letters. Whether a vector is a row or a column will always be clear from the context, and consequently we dispense with transpose signs on vectors. In (2), for example, $2 w$ represents the scalar product of $z$ (row) and $w$ (column). The superscript ${ }^{T}$ indicates the transpose of the matrix to which it is affixed.

[^1]:    *C. van de Panne and A. Whinston [21] have used the appropriate
    

[^2]:    has positive principal minors but is indefinite and not copositive. However, positive definite matrices are a subset of those with positive principal minors. (See, e.g. [2].)

