



A Computer Routine for Quadratic and Linear Programming Problems [H]

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Abstract. A computer program based on Lemke's complementary pivot algorithm is presented. This can be used to solve linear and quadratic programming problems. The program has been extensively tested on a wide range of problems and the results have been extremely satisfactory.

Key Words and Phrases: linear program, quadratic program, complementary problem, Lemke's algorithm, simplex method

CR Categories: 5.41

Language: Fortran

Description

Introduction. The computer routine given below is based on Lemke's complementary pivot algorithm [2] to solve the complementary problem of the form:

$$\begin{aligned} &\text{Find } w, z \geq 0 \\ &\text{such that } w = Mz + q \\ &w'z = 0 \end{aligned} \quad (1)$$

where M is an $(N \times N)$ square matrix; w, z and q are $(N \times 1)$ column vectors. ("Prime" denotes the transpose of a vector or matrix.)

A solution to the above problem will be called a complementary solution, and Lemke's algorithm is guaranteed to find a complementary solution to system (1) only if the matrix M satisfies one of the following:

1. M has all positive elements.
2. M is a positive semidefinite matrix or $x'Mx \geq 0$ for all x .
3. M has positive principal determinants.

Applications. The two important applications of the complementary problem (1) are to solve linear and quadratic programming problems by converting them to an equivalent complementary problem.

Quadratic Programming. Consider the quadratic program:

$$\begin{aligned} &\text{Minimize } Z = c'x + x'Qx \\ &\text{subject to } Ax \geq b \\ &x \geq 0 \end{aligned}$$

where A is an $(m \times n)$ matrix, Q is an $(n \times n)$ matrix of the quadratic form, c and x are $(n \times 1)$ column vectors, and b is an $(m \times 1)$ column vector

Submission of an algorithm for consideration for publication in Communications of the ACM implies unrestricted use of the algorithm within a computer is permissible.

An optimum solution to the above problem may be obtained by solving a complementary problem of the form:

$$\begin{pmatrix} v \\ u \end{pmatrix} = \begin{pmatrix} Q + Q' & -A' \\ A & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} + \begin{pmatrix} c \\ -b \end{pmatrix} \quad (2)$$

$$u, v, x, y \geq 0$$

$$v'x + u'y = 0$$

where u denotes the slack variables of the given quadratic program and (y, v) denotes the variables of the dual problem. Comparing the above system (2) with the original complementary problem (1), we note that

$$w = \begin{pmatrix} v \\ u \end{pmatrix}, z = \begin{pmatrix} x \\ y \end{pmatrix}, M = \begin{pmatrix} Q + Q' & A' \\ A & 0 \end{pmatrix} \text{ and } q = \begin{pmatrix} c \\ -b \end{pmatrix}.$$

System (2) can be solved by the given computer routine and then an optimum solution to the given quadratic program may be obtained by reading off the values of $(z_1, z_2, \dots, z_n, w_{n+1}, \dots, w_{n+m})$ from the complementary solution. It should be remarked here that the matrix M in this case is positive semidefinite if and only if the matrix Q is positive semidefinite. Hence, the computer routine is guaranteed to find an optimum solution to the given quadratic program only if the objective function Z is a convex function.

Linear Programming. Consider the linear program:

$$\begin{aligned} &\text{Minimize } Z = c'x \\ &\text{subject to } Ax \geq b \\ &x \geq 0. \end{aligned}$$

The only difference between a linear program and a quadratic program is in the objective function. Hence, by setting $Q = 0$ in system (2), we get the equivalent complementary problem for a linear program.

Program. A detailed description of Lemke's algorithm to solve the complementary problem, on which the computer routine is based, is given in [3]. The program consists of six subroutines and a main program which calls these subroutines in proper order. The various input data to the program are the number of problems to be solved in succession, the size of the problem and the elements of matrix M and vector q . The original Lemke's algorithm [2] was modified by the author along the lines of the revised simplex method [1] for a linear program to take advantage of the fact that for solving linear and quadratic programs, the M matrix in system (1) has many zero entries. This led to a greater efficiency of the computer routine.

In an experimental study conducted by the author [4], this computer routine was extensively used to compare the relative efficiencies of the simplex method [1] and Lemke's algorithm to solve linear programs. The study revealed the superiority of Lemke's algorithm over the simplex method in a number of problems both with regard to the number of iterations and computation time. Also in [3], another modification of Lemke's algorithm for solving linear programs has been proposed which may save a considerable storage and computation time.

References

1. Dantzig, G.B. *Linear Programming and Extensions*. Princeton U. Press, Princeton, N.J. 1963.
2. Lemke, C.E. Bimatrix equilibrium points and mathematical programming. *Management Sci.* 11 (1965), 681-689.
3. Ravindran, A. Computational aspects of Lemke's complementary algorithm applied to linear programs. *Opsearch* 7 (1970), 241-262.

Algorithm

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C REMARKS
C SINCE THIS PROGRAM IS COMPLETE IN ALL RESPECTS, IT CAN BE
C RUN AS IT IS WITHOUT ANY ADDITIONAL MODIFICATION OR
C INSTRUCTION. IN SUCH CASE FOLLOW THE INPUT FORMAT AS GIVEN
C
C PROGRAM FOR SOLVING LINEAR AND QUADRATIC PROGRAMMING
C PROBLEMS IN THE FORM  $W = Z^T U$ ,  $W = Z^T U$ ,  $W$  AND  $Z$  NONNEGATIVE
C BY LEMKE'S ALGORITHM.
C
C MAIN PROGRAM WHICH CALLS THE SIX SUBROUTINES-MATRIX,
C INITIA, NEWBAS, SORT, PIVOT AND PRINT IN PROPER ORDER.
C
COMMON AM(50,50), Q(50), H(50,50), A(50)
DIMENSION AM(50,50), Q(50), H(50,50), A(50)
DIMENSION W(50), Z(50), MBASIS(100)
C DESCRIPTION OF PARAMETERS IN COMMON
C AM A TWO DIMENSIONAL ARRAY CONTAINING THE
C ELEMENTS OF MATRIX M.
C Q A SINGLY SUBSCRIPTED ARRAY CONTAINING THE
C ELEMENTS OF VECTOR Q.
C LI AN INTEGER VARIABLE INDICATING THE NUMBER OF
C ITERATIONS TAKEN FOR EACH PROBLEM.
C R A TWO DIMENSIONAL ARRAY CONTAINING THE
C ELEMENTS OF THE INVERSE OF THE CURRENT BASIS.
C W A SINGLY SUBSCRIPTED ARRAY CONTAINING THE VALUES
C OF W VARIABLES IN EACH SOLUTION.
C Z A SINGLY SUBSCRIPTED ARRAY CONTAINING THE VALUES
C OF Z VARIABLES IN EACH SOLUTION.
C NL1 AN INTEGER VARIABLE TAKING VALUE 1 OR 2 DEPEND-
C ING ON WHETHER VARIABLE W OR Z LEAVES THE BASIS.
C NE1 SIMILAR TO NL1 BUT INDICATES VARIABLE ENTERING
C THE BASIS.
C NL2 AN INTEGER VARIABLE INDICATING WHAT COMPONENT
C OF W OR Z VARIABLE LEAVES THE BASIS.
C NE2 SIMILAR TO NL2 BUT INDICATES VARIABLE ENTERING
C THE BASIS.
C A A SINGLY SUBSCRIPTED ARRAY CONTAINING THE
C ELEMENTS OF THE TRANSFORMED COLUMN THAT IS
C ENTERING THE BASIS.
C IN AN INTEGER VARIABLE DENOTING THE PIVOT ROW AT
C EACH ITERATION. ALSO USED TO INDICATE TERMINA-
C TION OF A PROBLEM BY GIVING IT A VALUE OF 1000.
C MBASIS A SINGLY SUBSCRIPTED ARRAY-INDICATOR FOR THE
C BASIC VARIABLES. TWO INDICATORS ARE USED FOR
C EACH BASIC VARIABLE-ONE INDICATING WHETHER
C IT IS A W OR Z AND ANOTHER INDICATING WHAT
C COMPONENT OF W OR Z.
C
C READ IN THE VALUE OF VARIABLE IP INDICATING THE
C NUMBER OF PROBLEMS TO BE SOLVED.
HEAD(5,3) IP
C VARIABLE NO INDICATES THE CURRENT PROBLEM BEING SOLVED
NO=0
1 NO=NO+1
IF (NO.GT.IP) GO TO 5
WRITE(6,2) NO
2 FORMAT (1H1,10X,11HPROBLEM NO.,I2)
C
C READ IN THE SIZE OF THE MATRIX M
HEAD(5,3) N
3 FORMAT (I2)
C PROGRAM CALLING SEQUENCE
CALL MATRIX (N)
C PARAMETER N INDICATES THE PROBLEM SIZE
CALL INITIA (N)
C SINCE FOR ANY PROBLEM TERMINATION CAN OCCUR IN INITIA,
C NEWBAS OR SORT SUBROUTINE THE VALUE OF IN IS MATCHED WITH
C 1000 TO CHECK WHETHER TO CONTINUE OR GO TO NEXT PROBLEM.
IF (IN.EQ.1000) GO TO 1
4 CALL NEWBAS (N)
IF (IN.EQ.1000) GO TO 1
CALL SORT (N)
IF (IN.EQ.1000) GO TO 1
CALL PIVOT (N)
GO TO 4
5 STOP
END
SUBROUTINE MATRIX (N)
C PURPOSE - TO INITIALIZE AND READ IN THE VARIOUS INPUT DATA
C
COMMON AM(50,50), Q(50), H(50,50), A(50)
DIMENSION AM(50,50), Q(50), H(50,50), A(50)
DIMENSION W(50), Z(50), MBASIS(100)
C READ THE ELEMENTS OF M MATRIX COLUMN BY COLUMN
DO 1 J=1,N
1 HEAD(5,2) (AM(I,J),I=1,N)
2 FORMAT (7F10.5)
C READ THE ELEMENTS OF Q VECTOR
HEAD(5,2) (Q(I),I=1,N)
C IN ITERATION 1, BASIS INVERSE IS AN IDENTITY MATRIX.
DO 3 J=1,N
DO 4 I=1,N
IF (I.EQ.J) GO TO 3
B(I,J)=0.0
GO TO 4
3 B(I,J)=1.0
4 CONTINUE
5 CONTINUE
RETURN
END
SUBROUTINE INITIA (N)
C PURPOSE-TO FIND THE INITIAL ALMOST COMPLEMENTARY SOLUTION
C BY ADDING AN ARTIFICIAL VARIABLE Z0.
C
COMMON AM(50,50), Q(50), H(50,50), A(50)
DIMENSION AM(50,50), Q(50), H(50,50), A(50)
DIMENSION W(50), Z(50), MBASIS(100)

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C SET Z0 EQUAL TO THE MOST NEGATIVE Q(I)
I=1
J=2
1 IF (Q(I).LT.Q(J)) GO TO 2
I=J
2 J=J+1
IF (J.LE.N) GO TO 1
C UPDATE Q VECTOR
IN=1
TI=Q(IN)
IF ((I.LE.N) AND (Q(I).GT.0.0)) GO TO 4
DO 3 I=1,N
Q(I)=Q(I)+TI
3 CONTINUE
Q(IN)=TI
C UPDATE BASIS INVERSE AND INDICATOR VECTOR
C OF BASIC VARIABLES.
DO 4 J=1,N
R(J,IN)=-1.0
W(J)=Q(J)
Z(J)=0.0
MBASIS(J)=1
L=N+J
MBASIS(L)=J
4 CONTINUE
NL1=1
L=N+IN
NL2=IN
MBASIS(IN)=3
MBASIS(L)=0
W(IN)=0.0
Z0=Q(IN)
LI=1
C PRINT THE INITIAL ALMOST COMPLEMENTARY SOLUTION
WRITE(6,5)
5 FORMAT (3(/),5X,29HINITIAL ALMOST COMPLEMENTARY ,
1 8HSOLUTION)
DO 7 I=1,N
WRITE(6,6) I,W(I)
6 FORMAT (10X,2Hw(I),10,2H)=,F15.5)
7 CONTINUE
WRITE(6,8) Z0
8 FORMAT (10X,3HZ0=,F15.5)
RETURN
9 WRITE (6,10)
10 FORMAT (5X,36HPROBLEM HAS A TRIVIAL COMPLEMENTARY ,
1 23HSOLUTION WITH W=W, Z=Z0.)
IN=1000
RETURN
END
SUBROUTINE NEWBAS (N)
C PURPOSE - TO FIND THE NEW BASIS COLUMN TO ENTER IN
C TERMS OF THE CURRENT BASIS.
C
COMMON AM(50,50), Q(50), H(50,50), A(50)
DIMENSION AM(50,50), Q(50), H(50,50), A(50)
DIMENSION W(50), Z(50), MBASIS(100)
C IF NL1 IS NEITHER 1 NOR 2 THEN THE VARIABLE Z0 LEAVES THE
C BASIS INDICATING TERMINATION WITH A COMPLEMENTARY SOLUTION
IF (NL1.EQ.1) GO TO 2
IF (NL1.EQ.2) GO TO 5
WRITE(6,1)
1 FORMAT (5X,22HCOMPLEMENTARY SOLUTION)
CALL PRINT (N)
IN=1000
RETURN
2 NE1=2
NE2=NL2
C UPDATE NEW BASIC COLUMN BY MULTIPLYING BY BASIS INVERSE.
DO 4 I=1,N
TI=0.0
DO 3 J=1,N
TI=TI-B(I,J)*AM(J,NE2)
A(I)=TI
4 CONTINUE
RETURN
5 NE1=1
NE2=NL2
DO 6 I=1,N
A(I)=B(I,NE2)
6 CONTINUE
RETURN
END
SUBROUTINE SORT (N)
C PURPOSE - TO FIND THE PIVOT ROW FOR NEXT ITERATION BY THE
C USE OF (SIMPLEX-TYPE) MINIMUM RATIO RULE.
C
COMMON AM(50,50), Q(50), H(50,50), A(50)
DIMENSION AM(50,50), Q(50), H(50,50), A(50)
DIMENSION W(50), Z(50), MBASIS(100)
I=1
1 IF (A(I).GT.0.0) GO TO 2
I=I+1
IF (I.GT.N) GO TO 6
GO TO 1
2 TI=Q(I)/A(I)
IR=1
3 I=I+1
IF (I.GT.N) GO TO 5
IF (A(I).GT.0.0) GO TO 4
GO TO 3
4 T2=Q(I)/A(I)
IF (T2.GE.TI) GO TO 3
IN=I
TI=T2
GO TO 3
5 RETURN
C FAILURE OF THE RATIO RULE INDICATES TERMINATION WITH
C NO COMPLEMENTARY SOLUTION.
6 WRITE(6,7)
7 FORMAT (5X,37HPROBLEM HAS NO COMPLEMENTARY SOLUTION)
WRITE(6,8) LI
8 FORMAT (10X,13HITERATION NO.,I4)
IN=1000
RETURN
END

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SUBROUTINE PIVOT (N)
C PURPOSE - TO PERFORM THE PIVOT OPERATION BY UPDATING THE
C INVERSE OF THE BASIS AND Q VECTOR.
C
COMMON AM,Q,L1,B,NL1,NL2,A,NE1,NE2,IK,MBASIS,W,Z
DIMENSION AM(50,50), Q(50), B(50,50), A(50)
DIMENSION W(50), Z(50), MBASIS(100)
DO 1 I=1,N
1 B(I,I)=B(I,I)/A(I,I)
Q(I)=Q(I)/A(I,I)
DO 3 J=1,N
IF (I.EQ.IR) GO TO 3
Q(J)=Q(J)-Q(I)*A(I,J)
DO 2 J=1,N
B(I,J)=B(I,J)-B(I,I)*A(I,J)
2 CONTINUE
3 CONTINUE
C UPDATE THE INDICATOR VECTOR OF BASIC VARIABLES
NL1=MBASIS(IK)
L=N-IR
NL2=MBASIS(L)
MBASIS(IK)=NE1
MBASIS(L)=NE2
L1=L1+1
RETURN
END
SUBROUTINE PPINT (N)
C PURPOSE - TO PRINT THE CURRENT SOLUTION TO COMPLEMENTARY
C PROBLEM AND THE ITERATION NUMBER.
C
COMMON AM,Q,L1,B,NL1,NL2,A,NE1,NE2,IK,MBASIS,W,Z
DIMENSION AM(50,50), Q(50), B(50,50), A(50)
DIMENSION W(50), Z(50), MBASIS(100)
WRITE(6,1) L1
1 FORMAT (10X,13HITERATION NO.,I4)
J=N+1
J=1
2 K1=MBASIS(I)
K2=MBASIS(J)
IF (Q(J).GE.0.0) GO TO 3
Q(J)=0.0
3 IF (K2.EQ.1) GO TO 5
WRITE(6,4) K1,Q(J)
4 FORMAT (10X,2HZ(,I4,2H)=,F15.5)
GO TO 7
5 WRITE(6,6) K1,Q(J)
6 FORMAT (10X,2HW(,I4,2H)=,F15.5)
7 J=J+1
J=N+1
IF (J.LE.N) GO TO 2
RETURN
END

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Editor's note: Algorithm 432 described here is available on magnetic tape from the Department of Computer Science, University of Colorado, Boulder, CO 80302. The cost for the tape is \$16.00 (U.S. and Canada) or \$18.00 (elsewhere). If the user sends a small tape (wt. less than 1 lb.) the algorithm will be copied on it and returned to him at a charge of \$10.00 (U.S. only). All orders are to be prepaid with checks payable to ACM Algorithms. The algorithm is recorded as one file of BCD 80 character card images at 556 B.P.I., even parity, on seven track tape. We will supply the algorithm at a density of 800 B.P.I. if requested. The cards for the algorithm are sequenced starting at 10 and incremented by 10. The sequence number is right justified in column 80. Although we will make every attempt to insure that the algorithm conforms to the description printed here, we cannot guarantee it, nor can we guarantee that the algorithm is correct.—L.D.F.

Algorithm 432

Solution of the Matrix Equation $AX + XB = C$ [F4]

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Key Words and Phrases: linear algebra, matrices, linear equations

CR Categories: 5.14

Language: Fortran

Description

The following programs are a collection of Fortran IV subroutines to solve the matrix equation

$$AX + XB = C \quad (1)$$

where A , B , and C are real matrices of dimensions $m \times m$, $n \times n$, and $m \times n$, respectively. Additional subroutines permit the efficient solution of the equation

$$A^T X + XA = C, \quad (2)$$

where C is symmetric. Equation (1) has applications to the direct solution of discrete Poisson equations [2].

It is well known that (1) has a unique solution if and only if the eigenvalues $\alpha_1, \alpha_2, \dots, \alpha_m$ of A and $\beta_1, \beta_2, \dots, \beta_n$ of B satisfy

$$\alpha_i + \beta_j \neq 0 \quad (i = 1, 2, \dots, m; j = 1, 2, \dots, n).$$

One proof of the result amounts to constructing the solution from complete systems of eigenvalues and eigenvectors of A and B , when they exist. This technique has been proposed as a computational method (e.g. see [1]); however, it is unstable when the eigensystem is ill conditioned. The method proposed here is based on the Schur reduction to triangular form by orthogonal similarity transformations.

Equation (1) is solved as follows. The matrix A is reduced to lower real Schur form A' by an orthogonal similarity transformation U ; that is A is reduced to the real, block lower triangular form.

$$A' = U^T A U = \begin{bmatrix} A'_{11} & & & \\ A'_{21} & A'_{22} & & \\ & \ddots & \ddots & \\ & & A'_{p1} & A'_{p2} & \cdots & A'_{pp} \end{bmatrix},$$

where each matrix A'_{ij} is of order at most two. Similarly B is reduced to upper real Schur form by the orthogonal matrix V :

$$B' = V^T B V = \begin{bmatrix} B'_{11} & B'_{12} & \cdots & B'_{1q} \\ & B'_{22} & \cdots & B'_{2q} \\ & & \ddots & \\ & & & B'_{qq} \end{bmatrix},$$

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