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NUMERICALLY STABLE DUAL PROJECTION METHODS FOR SOLVING
POSITIVE DEFINITE QUADRATIC PROGRAMS

City University of New York

PH.D.

1980

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**NUMERICALLY STABLE DUAL PROJECTION METHODS FOR
SOLVING POSITIVE DEFINITE QUADRATIC PROGRAMS**

by

ASHOK U. IDNANI

A dissertation submitted to the Graduate
Faculty in Computer Science, School of
Engineering in partial fulfillment of the
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1980

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Abstract

Numerically stable dual projection methods for solving positive definite Quadratic Programs

by

Ashok U. Idnani

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Powell's efficient algorithm for solving general non-linear programs (NLP) with non-linear constraints which is based upon methods due to Han and Biggs, requires solving a positive definite Quadratic Program (QP) at each iteration. An efficient dual projection (DP) method for solving such a QP is presented. The DP method for QP starts at an optimal, but, in general, infeasible point and seeks feasibility while maintaining optimality. It always takes steps in the direction of the normal of a violated constraint projected onto the currently active constraint manifold. Its finite termination is proved.

Another method which we call the primal-dual projection (P-DP) method is also developed. It incorporates ideas used in our DP method and resorts to solving subproblems by using the primal projection method. It has the advantage of being able to start from an arbitrary point.

Though the DP method is related to Van de Panne and Whinston's dual tableau method for QP, the projection method takes fewer operations per iteration even under the most unfavourable conditions.

An efficient numerically stable implementation using the orthogonal and triangular factors particularly suited for the positive definite QP is presented. Tests on the use of this numerically stable DP method in Powell's algorithm on six published NLP problems produced a 45 to 70 percent reduction in the computational effort over that required by Fletcher's QP algorithm. Rigorous testing on randomly generated problems with as many as 81 variables and 243

constraints of six different primal projection methods, and the DP method show that the DP method is far more efficient. In most cases, the feasible point routine for the primal methods required more computational effort than obtaining the solution using the DP method. When started from the unconstrained minimum, the P-DP method performed slightly worse than the DP method. Finally, a method based on our results is suggested for initiating the dual simplex method for Linear Programs (LP).

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NOMENCLATURE

symbol	meaning
a	vector of coefficients for linear term in objective function
b	r.h.s. of constraints
e_j	j-th column of identity matrix
f	quadratic objective function
g	gradient of the objective function
k	total number of inequality constraints
m	number of variables
n	a constraint normal
n^+	normal of a violated constraint to be added to the basis
n^-	normal of a constraint to be removed from the basis
n^{-*}	row corresponding to the constraint being deleted in N^* .
p	usually, index of a violated constraint with normal n^+
q	number of active linearly independent constraints
r	infeasibility multipliers IM, $r = N^* n^+$.
s	vector of slacks
t	step length in z direction
t_1	max. step length in z direction with nonnegative LM
t_2	step length in z direction when a violated constraint becomes active
u	Lagrange multipliers, LM, $u = N^* g$
u^+	$N^{+*} g$
u^-	$N^{-*} g$
x	solution variables
z	step direction, $H n^+$
A	set of indices of q linearly independent active constraints
A^+	set $A \cup \{p\}$, $p \notin A$

A^-	a proper subset of A with one fewer element than A.
B	$L^{-1}N$
B^+	$L^{-1}N^+$
B^-	$L^{-1}N^-$
C	constraint matrix
D_j	defining hyperplane corresponding to the j-th constraint
G	constant symmetric positive definite Hessian matrix of f.
H	non-Euclidean projection operator with normals N, $H = G^{-1} - G^{-1}NN^*$
H^+	H defined with N^+
H^-	H defined with N^-
I	identity matrix
J	$Q^T L^{-1}$, $m \times m$ matrix
K	set of all k constraints.
L	Cholesky factor of $G = LL^T$, lower triangular matrix
M	linear manifold formed by the intersection of constraints in A
M^+, M^-	M defined with A^+ , A^- respectively
N	$m \times q$ basis matrix of normals in set A.
N^+	N defined with A^+ .
N^-	N defined with A^-
N^*	$q \times m$ matrix of generalized pseudo-inverse of N, $(N^T G^{-1} N)^{-1} N^T G^{-1}$
N^{+*}	N^* with N^+
N^{-*}	N^* with N^-
P_{j+1}^j	Givens matrix causing rotation in (j - j+1) plane
Q	orthogonal factor of $B = Q \begin{pmatrix} R \\ 0 \end{pmatrix}$
R	upper triangular factor of B, size $q \times q$
Y	Product of Givens matrices
ϕ	Null set

LIST OF ABBREVIATIONS

D	- Dual Method
DP	- Dual Projection
D.P.1	- Primal-Dual with P.1 Primal
F	- Feasible Point Routine
FP	- Fletcher's Feasible Point Routine for QP
FQ	- Fletcher's Quadratic Programming Routine
IM	- Infeasibility Multiplier(s)
K-T	- Kuhn-Tucker
LM	- Lagrange Multiplier(s)
LP	- Linear Programming
NLP	- Non Linear Programming
P	- Primal Method
POP	- Post Office Parcel Problem
P.1	- Primal Projection Method (Fletcher's)
P.1.1	- Same as P.1 with Steepest Edge Option
P.2	- Primal Projection Method (Goldfarb's)
P.2.1	- Same as P.2 with Steepest Edge Option
P.2'	- Two Phased P.2
P.2'.1	- Two Phased P.2.1
P-DP	- Primal-Dual Projection
QP	- Quadratic Programming
QPP	- Quadratic Programming Problems
SDS	- Satisfied Dual System
SP	- Subproblem
SSP	- Solved Subproblem

- VDS** - **Violated Dual System**
- VLDS** - **Violated Linearly Dependent Dual System**
- VW** - **Van de Panne - Whinston**

1. INTRODUCTION.

Quadratic Programming (QP) is becoming prominent as a method useful for solving general Non-linear Programming (NLP) problems with non-linear constraints (see Powell (29,30,31)). Algorithms to solve the Quadratic Programming Problems (QPP) are abundant in the recent literature. A parametric approach is given in Grigoriadis and Ritter (22), Ritter (32). Methods to solve the general QPP are given in Fletcher (12), Bunch and Kaufman (5) and Murray (28). Goncalves(21) describes a primal-dual method.

Basically all the above methods and various other methods can be classified into two categories. Methods in the first category use ideas similar to those used in Rosen's (33) gradient projection method (see Fletcher (12), Goldfarb (16)). Those in the other category use simplex-like tableau. The methods of Beale (2), Dantzig(7)-Wolfe(40) and Van de Panne - Whinston (38) fall into this category. A good description of these and other methods can be found in Boot (4) and Van de Panne (39). The projection methods are more efficient because they only store and update the necessary operators rather than the whole tableaux as in the simplex-like methods. There are similarities between gradient projection methods, simplex methods for QPP and the simplex method for Linear Programming (LP) and these are dealt with in detail in the literature (see Rosen(33), Goldfarb(16), Van de Panne-Whinston(38)).

All of the above methods with the exception of parametric approach and Van de Panne-Whinston's (VW) dual tableau method require a feasible point to start the algorithm and remain feasible throughout. VW's dual tableau method requires a dual feasible, that is, a primal optimal point for for some subproblem of the original problem, to start. It was found in our computational experience that on the average, 20 to 60% of the total effort required to solve the QPP, involves finding a feasible point and the necessary operators. This suggests that significant savings can be achieved, if, while seeking feasibility, optimality conditions are also satisfied simultaneously. Dual algorithms are based upon this idea. A method to find a good (in the sense that it is close to the solution of the problem) starting point for QPP with no addi-

tional work was suggested by Idnani(26). We propose a method based upon this recommendation to solve the strictly convex QPP by a method which can be viewed as dual to the gradient projection methods. This can also be thought of as an extension of the dual simplex method for LP. Unlike the primal methods, the proposed dual projection method starts at an optimal, but, in general, an infeasible point. It remains optimal throughout while achieving feasibility. Steps are always taken in the direction of the projected normal, hence it can be viewed as constraint normal projection method. It is related to VW's dual tableau method.

The gradient projection methods and the simplex-like methods move in a direction such that the objective function improves while maintaining feasibility. Such a direction is given by projecting the gradient onto the manifold defined by the intersection of the active constraints. This leads to a further classification among the projection algorithms according to the dimension of the manifold in which the steps are taken. When a comparison of two similar gradient projection methods was done by Idnani(26), it was found that the algorithm that took steps on as high dimensional manifolds as possible was more efficient and took fewer steps. Some recent computational results of Lenard (27) support these findings. Since our proposed dual projection method starts from unconstrained point, it is expected to take steps on high dimensional manifolds and appears to be efficient.

In section 2, we define the problem, present the notation and recursion formulae for updating the operators. In addition, we present some important properties of the operators necessary to develop the algorithm. In section 3, we develop a dual and a primal-dual projection method and prove their finite termination. An efficient numerically stable implementation of the algorithms using orthogonal factors and their updating whenever a basis change takes place is shown in section 4. Section 5 shows the relationship of the dual projection algorithm to VW's dual tableau method which is based on duality theory. In section 6, we show how this dual method can be used as a subroutine in Powell's code for solving general NLP problems. We also present the results of our runs with Powell's NLP method on six published NLP problems, comparing the dual projection method with one of the primal methods, namely, Fletcher's. In

section 7, we present the steepest face or edge considerations for dropping constraints in the primal methods and a somewhat comparable rule which we call the greatest increase choice for adding a constraint in the dual methods. The performance and comparison of six primal projection methods, dual and primal-dual projection methods, along with the experimental design of our testing procedure is described in section 8. Finally, some conclusions and areas for future research along with a suggestion for dual simplex method for LP and phase 1 of primal simplex method for LP which is based on our results is made.

2. PRELIMINARIES.

Problem Definition.

The general convex QPP in m variables and k constraints is defined as

$$\text{minimize } f(x) = a^T x + 1/2 x^T G x \quad (2.1)$$

$$\text{subject to } C^T x - b = s \geq 0 \quad (2.2)$$

where x and a are m vectors,

G is an $m \times m$ symmetric positive definite matrix,

C is an $m \times k$ constraint matrix,

b is a k vector

and s is the k vector of non-negative slacks.

The matrix C can be written as $(n_1 \ n_2 \ \cdots \ n_k)$ where n_j is the m vector normal of the j -th constraint in (2.2).

Notation.

We will use matrix notation. Lower case letters x, y, \dots etc. will denote column vectors, scalars or elements of a set. Upper case letters C, G, \dots etc., will denote matrices or a set of indices of the k constraints in (2.2). Superscript T denotes transpose, so that a^T is a row vector. A vector relation applies to each component of the vector, (i.e.) $x \geq 0$, indicates that each component of x is non-negative. The identity matrix will be denoted by I and the j -th column of I will be represented as e_j . Sometimes we will use subscripted I to show the identity matrix of the size of the subscript. Normally, the components of a column vector will be indicated by subscripts. Matrix, vector and scalar zero will be represented as 0 . The null set will be denoted by ϕ . G^{-1} will represent the inverse of the matrix G . We will use \bar{A} to indicate A after an update.

We will use K to indicate the set of indices of all the k constraints in (2.2). We will denote the set of indices corresponding to linearly independent constraints that are currently active as

A. A constraint is said to be active if it is satisfied as an equality, i.e., the slack corresponding to that constraint is 0. Clearly, the set A is a subset of K . The matrix of normals of constraints whose indices are in A will be represented as $N(A)$ or N and the b vector as $b(A)$ and the s vector corresponding to the set A will be denoted as $s(A, x)$. We will use $K \setminus A$ to indicate the set of indices in K but not in A . A^+ will denote the set $(A \cup \{p\})$, where p is an element in $K \setminus A$. Similarly, A^- will represent a proper subset of A , whenever A is not null and will contain one fewer element than A . To be consistent, N^+ and $N(A^+)$ represent the set of normals in A^+ and N^- and $N(A^-)$ represent the set of normals in A^- . n^+ will be used to indicate the column vector in C , corresponding to p in the definition of A^+ . Likewise, n^- will be used to indicate the column vector to be deleted from N to get N^- . Usually, the number of elements in A will be q , $q \leq m$.

Operators and Factorization.

The matrix G as defined in (2.1) can be decomposed into the product of two more convenient matrices as follows:

$$G = LL^T \quad (2.3)$$

where L is a lower triangular square matrix of size m . Define

$$B = L^{-1}N \quad (2.4)$$

where B is an $m \times q$ matrix. The orthogonal factors of B can be written as

$$B = Q \begin{pmatrix} R \\ 0 \end{pmatrix} \quad (2.5)$$

where Q is an $m \times m$ orthogonal matrix and R is an upper triangular matrix of dimension $q \times q$.

The k inequality constraints in (2.2) represent k closed half-spaces in m dimensional Euclidean space E^m . The hyperplane corresponding to a strict equality in (2.2) is a "defining hyperplane" for the associated half-space. Let D_j represent such a defining hyperplane corresponding to the j -th constraint in (2.2). (i.e.)

$$D_j = \{ x \mid n_j^T x - b_j = s_j = 0 \} \quad (2.6)$$

Let the set of indices of q linearly independent constraints be A . Then we will use $M(A)$ or M to indicate the $m-q$ dimensional manifold, formed by the intersection of these q linearly independent defining hyperplanes. A set of hyperplanes are linearly independent if the corresponding constraint normals are linearly independent. When A is the null set, $M(A)$ is the full space E^m and when A has cardinality m , (i.e. $q=m$), $M(A)$ is a point with zero degrees of freedom.

We will frequently encounter the projection of a vector y in the m dimensional space E^m onto the linear manifold M . This projection can be accomplished by using the projection operator,

$$H(N(A)) = H(N) = H = G^{-1} - G^{-1}N(N^T G^{-1}N)^{-1}N^T G^{-1} \quad (2.7)$$

where H is an $m \times m$ positive semi-definite matrix of rank $m-q$. If $G \neq I$, H is a non-orthogonal projection operator. We will use M^+ and M^- for $M(A^+)$ and $M(A^-)$ respectively. We will use a like notation for H , (i.e.) H^+ and H^- .

Another operator we will use is the generalized pseudo-inverse of N defined as

$$N^* = (N^T G^{-1}N)^{-1}N^T G^{-1} \quad (2.8)$$

where N^* is a $q \times m$ matrix of rank q . We will use a notation similar to N for N^{+*} and N^{-*} . We will use n^{-*} to denote the row corresponding to the index of the deleted constraint in N^* .

The two operators H and N^* can be written in terms of the factors of B . Applying (2.3) through (2.5) in (2.7) we can write

$$\begin{aligned} H &= L^{-T} (I - B(B^T B)^{-1}B^T) L^{-1} \\ &= L^{-T} (I - Q \begin{bmatrix} R \\ 0 \end{bmatrix} (R^T R)^{-1} (R^T \ 0) Q^T) L^{-1} \end{aligned} \quad (2.9)$$

Partitioning Q as $(Q_1 \ Q_2)$ where Q_1 is of size $m \times q$ and using the orthogonality of Q , we can reduce (2.9) to

$$H = L^{-T} Q_2 Q_2^T L^{-1} \quad (2.10)$$

Similarly N^* can be simplified as follows.

$$\begin{aligned} N^* &= (B^T B)^{-1} B^T L^{-1} \\ &= (R^T R)^{-1} (R^T 0) Q^T L^{-1} \\ &= R^{-1} Q_1^T L^{-1} \end{aligned} \quad (2.11)$$

The following well known properties will be used frequently and their proofs can be found in Goldfarb (15,16).

$$Hw = 0 \text{ iff } w = N v. \quad (2.12)$$

$$w^T Hw = 0 \text{ iff } w = N v. \quad (2.13)$$

Using definitions (2.7) and (2.8) and algebraic manipulations, it is easy to see that

$$H G H = H \quad (2.14)$$

$$H H^+ = H^+ \quad (2.15)$$

$$N^* G H = 0 \quad (2.16)$$

In the algorithm presented in the next section, we will add a new constraint to or delete a constraint from the current basis. The number of operations (multiplications and divisions) required to calculate the two operators from scratch is prohibitively large and would make our algorithms impractical. However, updating of these operators can be achieved by using the following recursion formula which cost only $O(m^2)$ operations (see Fletcher(12)). In our actual implementation we store only the matrices $Q^T L^{-1}$ and R in addition to the original data, updating them whenever there is a change of basis. The required recursions are given in section 4.

Adding a constraint with normal n^+ .

$$H^+ = H - zz^T / (z^T n^+) \quad (2.17)$$

$$N^{+*} = \begin{bmatrix} N^* \\ 0 \end{bmatrix} + \begin{bmatrix} -N^* n^+ \\ 1 \end{bmatrix} z^T / (z^T n^+) \quad (2.18)$$

where $z = H n^+$.

Deleting a constraint.

$$H^- = H + \frac{n^{-*T} n^{-*}}{n^{-*} G n^{-*T}} \quad (2.19)$$

$$N^{-*} = N^* - \frac{N^* G n^{-*T} n^{-*}}{n^{-*} G n^{-*T}} \quad (2.20)$$

where n^{-*} is the row corresponding to the constraint being deleted in N^* . Of course, the row corresponding to the dropped constraint is zero in N^{-*} . Clearly, $H = G^{-1}$ when A is the null set and $H = 0$ when $q = m$. Also $N^* = N^{-1}$ in the latter case.

Multipliers.

For a given linearly independent set of normals $N = N(A)$, we will define a q -vector v of multipliers corresponding to the m vector y as

$$v = N^* y \quad (2.21)$$

If $N w = y$, then $v = w$ since $N^* N = I$, but (2.21) does not, in general imply that $N v = y$.

Lagrange Multipliers.(LM)

The gradient of the objective function (2.1) can be written as

$$g(x) = \nabla_x f(x) = G x + a \quad (2.22)$$

The vector of multipliers corresponding to $y = g(x)$ in (2.21) are called the Lagrange Multipliers (LM) and are represented by vector

$$u(x) = N^* g(x) \quad (2.23)$$

We will denote $N^{+*} g(x)$ as $u^+(x)$ and $N^{-*} g(x)$ as $u^-(x)$. Other names used for LM are shadow prices and dual variables.

Lemma 1.

$u(x) = N^* g(x)$ is independent of x on a given manifold M .

Proof:

Let y and z be two points on M . Then,

$$\begin{aligned} u(y) - u(z) &= N^* G (y-z) \\ &= (N^T G^{-1} N)^{-1} N^T (y-z) \end{aligned}$$

Since y and z are on M , $N^T (y-z) = 0$ and therefore $u(y) = u(z)$ ●

Infeasibility Multipliers. (IM)

Infeasibility multipliers (IM) corresponding to a particular constraint with normal n are defined by replacing y in (2.21) by the normal n . We denote this vector by r .

$$r = N^* n \tag{2.24}$$

As we will see in the next section, the IM indicate whether or not a feasible point exists. By the above definition, r is independent of x on a given manifold M .

3. ALGORITHM AND FINITENESS.

Unconstrained minimum

The unconstrained minimum of the objective function (2.1) can be obtained by using classical techniques. In particular, by setting the derivative of f with respect to x to zero, $\nabla_x f(x) = Gx + a = 0$ and solving for x , we get the unconstrained minimum

$$x^0 = -G^{-1}a \quad (3.1)$$

Minimum on a manifold.

If, at the unconstrained minimum x^0 , all the constraints (2.2) are satisfied, (i.e.) $s(K, x^0) \geq 0$, then x^0 solves the problem. Otherwise, the solution to the problem has to lie on some manifold of q linearly independent defining hyperplanes and the indices of these q constraints are a subset of K and $1 \leq q \leq \min(k, m)$. Our problem then, is to find this manifold and the minimum of $f(x)$ on it. (See figure 1 on page 25.)

The necessary condition for x^* to be the minimum on M is that the gradient at x^* lies in the subspace orthogonal to M , that is,

$$H g(x^*) = 0 \quad (3.2)$$

This implies that

$$g(x^*) = Gx^* + a = Nu \quad (3.3)$$

Basic approach

The following definitions of subproblem(SP) and solved subproblem (SSP) are essential for explaining the basic dual and primal-dual approaches developed in this section.

Subproblem (SP)

It is a QPP with the same objective function as (2.1) with only a subset of the original constraints (2.2). It can be thought of as a set of constraints for all practical purposes, (i.e.), $SP \subseteq K$.

Solved Subproblem (SSP)

Let the solution of a Subproblem SP lie on a manifold of q linearly independent active constraints. Then SSP is this set of q constraints. Clearly, $SSP \subseteq SP$.

Dual approach

Now the basic dual approach can be explained as follows:

Assume that some SSP and the solution point on the manifold corresponding to it are available.

- 1) Pick a violated constraint, if any, say p -th.
- 2a) Define $SP = (SSP \cup \{p\})$.
- 2b) Solve the SP.
- 2c) If no solution, terminate because the feasible region is empty; otherwise, define an SSP.

Repeat 1 and 2 until solution. One can use the empty set and the unconstrained minimum as a possible start. Steps 2b and 2c are developed later.

Primal-Dual approach

The primal-dual approach can be explained in the following five steps.

0) *Starting Point.*

Define an SP, some subproblem of the original problem and go to step 2.

1) *Define SP.*

If an SSP exists, then $SP = (SSP \cup \{p\})$, where p is the index of a violated constraint, if any. Otherwise, define some SP to be solved.

2) *Feasibility.*

Apply dual projection method to find a feasible point for the SP. If none can be found, terminate.

3) *Optimality.*

If the feasible point solves the subproblem SP, then define SSP and go to step 1, otherwise go to step 4.

4) *Solve subproblem SP.*

Apply primal projection method to solve SP and define SSP. Go to step 1.

Step direction

In order to develop the necessary tools for the dual algorithm, first, we will show the form of the step from the minimum on a manifold M to the minimum on a submanifold M^+ of M .

Lemma 2.

Let x and x^+ be optimum points on manifolds M and $M^+ \subset M$, respectively. Then, we can find x^+ from x in one step, z , and z is of the form $G^{-1}N^+y$, where y is a $(q+1)$ vector, that is,

$$x^+ = x + G^{-1}N^+y \quad (3.4)$$

Proof:

Since $M^+ \subset M$, $g(x)$ is orthogonal to M^+ . Therefore, $g(x^+) - g(x)$ is orthogonal to M^+ .

This implies that

$$\begin{aligned} g(x^+) - g(x) &= N^+y \\ G(x^+ - x) &= N^+y \end{aligned}$$

which results in (3.4)●

The above holds as long as $M^+ \subset M$, therefore, $A^+ \setminus A$ need not be only one element. One can apply the above lemma even when A is the null set. The following lemma gives another representation of the optimum point x^+ in (3.4).

Lemma 3.

Let x and x^+ be the optimum points on manifolds M and $M^+ \subset M$. Then, we can find x^+ from x in one step, z , (i.e.)

$$x^+ = x + z \quad (3.5)$$

where
$$z = -N^{+*T}s(A^+,x). \quad (3.6)$$

Moreover, x^+ is on M^+ even if x is an arbitrary point in E^m .

Proof:

From (3.4) it follows that

$$\begin{aligned} s(A^+,x^+) &= 0 \\ &= s(A^+,x) + N^{+T}G^{-1}N^+y \end{aligned}$$

Consequently (3.4) can be written as

$$x^+ = x - G^{-1}N^+(N^{+T}G^{-1}N^+)^{-1}s(A^+,x)$$

which is just (3.6).

When x is an arbitrary point in E^m , (3.5) and (3.6) yield a point which lies on M^+ since it follows from (3.5) and (3.6) that $s(A^+,x^+) = 0$ ●

Theorem 1.

The minimum of (2.1) on manifold M lies at

$$x = N^{*T}b(A) - H a \quad (3.7a)$$

and the LMs of active constraints are given by

$$u(x) = (N^T G^{-1} N)^{-1} b(A) + N^* a \quad (3.7b)$$

Proof:

Applying Lemma 3 with A replaced by null set and A^+ by A , we get

$$\begin{aligned} x &= -G^{-1}a - N^{*T}s(A,-G^{-1}a) \\ &= -G^{-1}a - G^{-1}N(N^T G^{-1} N)^{-1}(-N^T G^{-1}a - b(A)) \\ &= N^{*T}b(A) - H a \end{aligned}$$

which is (3.7a). (3.7b) can be shown by using the definition of LM and the fact that

$$N^* G H = 0 \bullet$$

One could enumerate subsets of k constraints and evaluate x , f and s at the minimum point for every subset. Theil-Van de Panne (37) suggest an intelligent way of picking the subsets to be evaluated. It involves enormous amount of work and is impractical for large problems.

Before we present the development of the algorithm, let us stress one well known condition for the step direction z to satisfy. In order to yield an improvement in s , that is, $s(A, \bar{x}) \geq s(A, x)$, where $\bar{x} = x + z$, the direction z must satisfy

$$N^T z \geq 0. \quad (3.8)$$

This is the consequence of the fact that $s(A, \bar{x}) - s(A, x) = N^T z$ and hence the condition (3.8).

Kuhn-Tucker conditions for optimum point.

Let the manifold M be determined uniquely by a set of active constraint indices in A , (i.e), M is not contained in the $m-1$ dimensional subspace corresponding to any of the defining hyperplanes whose indices are in $K \setminus A$. Then, the Kuhn-Tucker (K-T) necessary conditions for a point $x^* \in M$ to be the global minimum of the problem (2.1), (2.2) are (see Goldfarb (15,16)),

Optimality:

$$H g(x^*) = 0 \quad (3.9)$$

$$N^* g(x^*) = u(x^*) \geq 0 \quad (3.10)$$

Feasibility:

$$s(K \setminus A, x^*) \geq 0 \quad (3.11)$$

$$s(A, x^*) = 0 \quad (3.12)$$

The optimality conditions imply that the gradient at x^* can be written as nonnegative linear combination of the normals N .

We need to define a violated dual system (VDS) in order to make the development of the

algorithm simple. A VDS with respect to the set $A \cup \{p\}$ and a point x which we denote as $E(x, A, p)$, is one that meets the following criteria.

1) p is the index of a constraint in $K \setminus A$ with normal n^+ , such that $|Hn^+| \neq 0$, (i.e.) the set of constraints whose indices are in $A^+ = A \cup \{p\}$, are linearly independent. (3.13)

$$2) H^+ g(x) = 0. \quad (3.14)$$

$$3) s(A, x) = 0. \quad (3.15)$$

$$4) u^+(x) = N^{+*} g(x) \geq 0. \quad (3.16)$$

$$5) s(\{p\}, x) < 0. \quad (3.17)$$

We will define a satisfied dual system (SDS) with respect to the set $A \cup \{p\}$ and a point x as one which satisfies (3.13) - (3.16) and $s(\{p\}, x) = 0$. Such an SDS will be denoted as $S(x, A, p)$. Now we will present a Lemma on the rate of change of LM when we take a step from a VDS in the projected normal direction.

Lemma 4.

Let $E(x, A, p)$ be a VDS. Let $u^+(x)$ be the vector of $q+1$ LMs. Let the step direction z be,

$$z = H n^+. \quad (3.18)$$

Let \bar{x} be a point such that

$$\bar{x} = x + t z, \quad (3.19)$$

where t is a positive scalar representing the step length. The LM along the direction z is given by

$$u^+(\bar{x}) = \begin{cases} u_j^+(x) - t e_j^T N^* n^+, & j = 1, 2, \dots, q \\ u_{q+1}^+(x) + t \end{cases} \quad (3.20)$$

Proof

$$\begin{aligned} u^+(\bar{x}) &= N^{+*} g(x) + t N^{+*} G z \\ &= N^{+*} g(x) + t N^{+*} G H n^+ \end{aligned} \quad (3.21)$$

Using recursion formula (2.18) for N^{+*} and properties (2.14) and (2.16) in (3.21) gives (3.20)●

By applying Lemmas 2, 3 and 4 to a VDS we now show how a step in the direction of the projected normal results in a VDS or an SDS depending upon the step length.

Theorem 2.

Let $E(x, A, p)$ be a VDS. If the point \bar{x} is defined by

$$\bar{x} = x + t z, \quad (3.22)$$

where $z = H n^+, \quad (3.23)$

$$t = \text{minimum } (t_1, t_2) \geq 0 \quad (3.24)$$

$$t_1 = \min. \left\{ \begin{array}{l} \text{min.} \\ r_j > 0 \\ j=1, \dots, q \end{array} \left\{ \frac{u_j^+(x)}{r_j} \right\}, \alpha \right\} \quad (3.25)$$

$$t_2 = -s(\{p\}, x) / (z^T n^+) \quad (3.26)$$

and

$$r = N^* n^+$$

then

$$s(\{p\}, \bar{x}) \geq s(\{p\}, x) \quad (3.27)$$

and

$$f(\bar{x}) - f(x) = t z^T n^+ (1/2 t + u_{q+1}^+(x)) \geq 0. \quad (3.28)$$

Moreover, if $t = t_1$, we have a VDS, $E(\bar{x}, A, p)$, if $t = t_2$ we have an SDS, $S(\bar{x}, A, p)$.

Proof:

From (2.18), (3.23) and (3.26) it follows that

$$-N^{+*T} s(A^+, x) = (-s(\{p\}, x)/(z^T n^+)) z = t_2 z \quad (3.29)$$

since $E(x, A, p)$ is a VDS. Therefore, if $t = t_2 < t_1$, it follows from Lemma 3 that $s(A^+, \bar{x}) = 0$. Also

$$H^+ g(\bar{x}) = H^+ g(x) - H^+ G N^{+*T} s(A^+, x) = 0 \quad (3.30)$$

since $H^+ G N^{+*T} = 0$ by (2.16) and $u^+(\bar{x}) > 0$ by Lemma 4. Since

$$\bar{x} = \bar{x}(t) = \bar{x}(t_2),$$

we have an SDS of the form $S(\bar{x}, A, p)$.

If $t = t_1 < t_2$, then, we have a VDS of the form $E(\bar{x}, A, p)$ by the above and by linearity of $S(A^+, \bar{x}(t))$, $H^+ g(\bar{x}(t))$ and $u^+(\bar{x}(t))$ with respect to the step length t .

The change in function value can be calculated by using Taylor's theorem as

$$f(\bar{x}) - f(x) = t z^T g(x) + 1/2 t^2 z^T G z \quad (3.31)$$

where

$$z^T G z = n^{+T} H G H n^+ = n^{+T} H n^+ = z^T n^+ > 0 \quad (3.32)$$

and

$$z^T g(x) = n^{+T} H g(x) \quad (3.33)$$

From the recursion formulae (2.17), (2.18) and the fact that $H^+ g(x) = 0$,

$$u_{q+1}^+(x) = \frac{n^{+T} H g(x)}{n^{+T} H n^+} \quad (3.34)$$

and

$$H g(x) = H n^+ u_{q+1}^+(x) \quad (3.35)$$

$$\begin{aligned} n^{+T} H g(x) &= n^{+T} H n^+ u_{q+1}^+(x) \\ &= z^T n^+ u_{q+1}^+(x) \end{aligned} \quad (3.36)$$

Substituting (3.36) into (3.33) and (3.31), and (3.32) into (3.31) yields (3.28). Since all elements of (3.28) are nonnegative, $f(\bar{x}) \geq f(x)$. Moreover, as long as $t > 0$, $f(\bar{x}) > f(x)$ ●

The above theorem suggests that if one starts with a VDS, one can move to another VDS or an SDS with improvement in f and s . At an SDS one can update N^* and H by adding the constraint and check if another VDS can be obtained by picking a violated constraint. In that case, one can reapply the above theorem. If there is no violated constraint, we can terminate. If the violated constraint is linearly dependent, then, theorem 2 breaks down. This situation is discussed in theorem 3. However, if a VDS is arrived after the step, then the constraint corresponding to the minimum in t_1 can be dropped from the basis and A , N^* and H updated. We only need to show that this new point with respect to the new set of constraints forms a VDS by showing that the projection of the gradient on the new manifold is zero. This can be seen from (2.19) since the LM of the constraint dropped is zero. An SDS is guaranteed to be obtained in no more than $q'+1$ applications of theorem 2 where q' is the number of active constraints at the most recent SDS.

When A is a full set, (i.e. $q = m$) or when $|H n^+| = 0$, we no longer have a VDS and hence theorem 2 is not applicable. At this juncture, one could drop a constraint to form a VDS and apply theorem 2. The condition to be met for the choice of the constraint to be dropped can be found from the requirement for VDS and Lemma 4 and is shown in the next theorem.

Dropping criteria

We will define a violated linearly dependent system VLDS, with respect to the set $A \cup \{p\}$, denoted as $V(x, A, p)$ the one that meets the following criteria.

1) p is the index of a constraint in $K \setminus A$ with normal n^+ such that

$$|H n^+| = 0.$$

2) $|H g(x)| = 0$

3) $s(A, x) = 0$

$$4) u(x) = N^* g(x) \geq 0$$

$$5) s(\{p\}, x) < 0$$

Theorem 3.

Let $V(x, A, p)$ be a VLDS. Compute

$$t_1 = \min. \left\{ \infty, \min_{j=1, \dots, q} \left\{ \frac{u_j(x)}{r_j} \right\} \right\} \quad (3.37)$$

Then,

a) if $t_1 = \infty$, then there is no allowable direction z such that $s(\{p\}, x) < s(\{p\}, \bar{x})$ and $s(A, \bar{x}) \geq s(A, x) = 0$, where $\bar{x} = x + z$.

b) if $t_1 < \infty$, then the constraint corresponding to the minimum in (3.37) can be dropped from the set A to give A^- and $E(x, A^-, p)$ is a VDS

Proof:

Since $V(x, A, p)$ is a VLDS, we have $|H n^+| = 0$ which implies

$$N r = n^+ \quad (3.38)$$

and

$$r = N^* n^+$$

Now there are two possibilities.

Case 1 $r \leq 0$.

From (3.38) we can write

$$n^{+T} z = r^T N^T z \quad (3.39)$$

where z is the step direction. By (3.8), $s(A, \bar{x}) \geq s(A, x)$ only if $N^T z \geq 0$ and $s(\{p\}, \bar{x}) > s(\{p\}, x)$ only if $n^{+T} z > 0$. If $r \leq 0$, then, it follows from (3.39) that no allowable direction exists such that $s(A, \bar{x}) \geq s(A, x)$ and $s(\{p\}, \bar{x}) > s(\{p\}, x)$.

Case 2 At least one component of r (say r_q) is positive.

Now we will show the condition required to drop a constraint so that $E(x, A^-, p)$ is a VDS. Consider dropping a constraint whose component in r is positive, say q -th. Since $V(x, A, p)$ is a VLDS, $|H g(x)| = 0$ and $|H n^+| = 0$. Let $A^-U\{q\} = A$. Consequently by (2.19) and the definition of $u(x)$ and r ,

$$H^-g(x) = \frac{n^{-*T}u_q(x)}{n^{-*}G n^{-*T}} \quad (3.40)$$

$$H^-n^+ = \frac{n^{-*T}r_q}{n^{-*}G n^{-*T}} \quad (3.41)$$

$$\frac{n^{+T}H^-g(x)}{n^{+T}H^-n^+} = \frac{u_q(x)}{r_q} \quad (3.42)$$

Let H^+ and N^{+*} denote the operators H and N^* with respect to the set $A^+ = A^-U\{p\}$ for the rest of this theorem. Applying the recursion (2.17) we find

$$H^+g(x) = H^-g(x) - \frac{H^-n^+n^{+T}H^-g(x)}{n^{+T}H^-n^+} \quad (3.43)$$

By substituting (3.40) - (3.42) in (3.43) we find that $H^+g(x) = 0$. Also from (3.41), $|H^-n^+| > 0$.

The other condition required for $E(x, A^-, p)$ to be a VDS is that $N^{+*}g(x) \geq 0$. Using the recursion formulae (2.20) and (2.18) we can write

$$N^{-*}g(x) = N^*g(x) - \frac{N^*G n^{-*T}u_q(x)}{n^{-*}G n^{-*T}} \quad (3.44)$$

$$N^{-*}n^+ = N^*n^+ - \frac{N^*G n^{-*T}r_q}{n^{-*}G n^{-*T}} \quad (3.45)$$

and using (3.42)

$$N^{+*}g(x) = \begin{bmatrix} N^{-*}g(x) \\ 0 \end{bmatrix} + \begin{bmatrix} N^{-*}n^+ \\ 1 \end{bmatrix} \frac{u_q(x)}{r_q} \quad (3.46)$$

Substituting (3.44), (3.45) in (3.46) we find

$$N^{+*}g(x) = \begin{bmatrix} u(x) - r \frac{u_q(x)}{r_q} \\ \frac{u_q(x)}{r_q} \end{bmatrix} \quad (3.47)$$

The requirement of nonnegativity of $N^{+*}g(x)$ suggests that the constraint to be dropped be the one corresponding to

$$\begin{array}{l} \min. \\ r_j > 0, \\ j=1, \dots, q \end{array} \left\{ \frac{u_j(x)}{r_j} \right\} \bullet$$

The Theorems 2 and 3 provide a constructive procedure which results in improvement of the function value and reduction of infeasibility of a violated constraint, ultimately reaching an SDS. Note that in case 2 of theorem 3, x is not changed; only the set A is. Successive applications of the above theorems as necessary, starting from a VDS will converge to the global solution of the problem. Now we have the necessary tools to present the algorithm and show its finiteness.

Dual Algorithm:

Step 1. Compute G^{-1} , the unconstrained minimum, $x = -G^{-1}a$ and $f = -1/2a^T x$. Set $H = G^{-1}$ and $A = \{\phi\}$, $q = 0$, the LM, $u = 0$.

Step 2. Compute $s(K \setminus A, x)$. Find $s(\{p\}, x) = \min. (s(K \setminus A, x))$. If $s(\{p\}, x) \geq 0$, then terminate; x solves the problem; otherwise, let the normal corresponding to p be n^+ and set $u^+ = \begin{bmatrix} u \\ 0 \end{bmatrix}$.

Step 3. Compute the step direction $z = H n^+$.

If $q > 0$, compute the IM, $r = N^* n^+$. If $|z| > 0$, go to step 4.

Compute t_1 as in (3.37). If $t_1 = \infty$, then there is no feasible point that solves (2.2); terminate. Otherwise, drop the constraint corresponding to the minimum in (3.37), update u as in (3.47), A, H, N^* , set $q = q-1$ and repeat this step.

Step 4. Compute t_1 as in (3.25), t_2 as in (3.26) and $t = \min.(t_1, t_2)$. Take step $\bar{x} = x + t z$ and update \bar{f} as in (3.28) and $u^+(\bar{x})$ as in (3.20). Calculate $s(\{p\}, \bar{x})$. Set $x \leftarrow \bar{x}$ and $f \leftarrow \bar{f}$. If $t = t_2$, then the p -th constraint is added, (i.e.), set $A \leftarrow AU\{p\}$, update H, N^* and q . Set $u \leftarrow u^+$. Go to step 2. If $t = t_1$, drop the constraint corresponding to the minimum value of t_1 from A , set $A \leftarrow A^-$, update H, N^* and q . Go to step 3●

Finiteness of Dual Algorithm

Theorem 4.

The given dual algorithm will produce the optimum point in a finite number of steps or terminate when there is no feasible point.

Proof:

It is clear from the algorithm that when step 2 is executed, x solves some subproblem of the original problem due to the fact that the optimality conditions with respect to that subproblem are satisfied. The subproblem may be thought of just the q active constraints. Every time a step is taken, the function value increases and we must return to step 2 in finite number of steps or terminate with no feasible point. This suggests that we will never solve the same subproblem again. Since there are a finite number of subproblems, the algorithm must terminate with the global solution, if there is one, in a finite number of steps or else indicate that the feasible region is empty●

One advantage of the dual algorithm as described is that the gradient of the objective function and its projection are never computed. The LM can be updated every time a basis change takes place. This results in a significant savings in the amount of work per iteration. Starting

from the unconstrained optimum does not require any additional work than most primal projection methods because the operators H and N^* have to be calculated for the first time (in feasible point routine) which directly or indirectly use G^{-1} or when it is a full basis, N^{-1} . Now we will present an example and geometric interpretation of the method. In order to illustrate the various parts of the algorithm, we will not require that the most violated constraint be the one to be added.

Example:

Consider the following problem.

$$\text{minimize } f(x) = (6 \ 0) \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + 1/2 (x_1 \ x_2) \begin{bmatrix} 4 & -2 \\ -2 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

subject to

- | | | |
|------------------|-----------|--------------------|
| 1) x_1 | ≥ 0 | normal $(1,0)^T$ |
| 2) x_2 | ≥ 0 | normal $(0,1)^T$ |
| 3) $x_1 + x_2$ | ≥ 2 | normal $(1,1)^T$ |
| 4) $-2x_1 - x_2$ | ≥ -4 | normal $(-2,-1)^T$ |

Solution:

$$G^{-1} = \begin{bmatrix} 2/6 & 1/6 \\ 1/6 & 2/6 \end{bmatrix}$$

The unconstrained minimum occurs at $x = (-2 \ -1)^T$ with function value -6. The slack vector s is $(-2 \ -1 \ -5 \ 9)^T$ and $H = G^{-1}$, set $q = 0$.

Iteration 1.

Among the first three violated constraints, let us pick the second constraint to be added, (i.e.) $p = 2$, thus making $E(x, \{\phi\}, 2)$ a VDS. We can now take a full step as follows.

$$z^T = (1/6 \ 2/6), z^T n^+ = 1/3, t_2 = 3, t = t_2.$$

$$x^T = (-3/2 \ 0), u = 3 \text{ and } f = -9/2.$$

Adding constraint 2 to the basis and updating H , N^* , A and q , we obtain

$$H = \begin{pmatrix} 1/4 & 0 \\ 0 & 0 \end{pmatrix}, N^* = (1/2 \ 1), A = \{2\} \text{ and } q = 1.$$

Iteration 2.

The s vector at this new x is $(-3/2 \ 0 \ -7/2 \ 7)^T$. Let us choose constraint 3 among the violated constraints to be added, (i.e.) $p = 3$. This makes $E(x, \{2\}, 3)$ a VDS.

We can now calculate

$z^T = (1/4 \ 0)$, $z^T n^+ = 1/4$, $r = 3/2$, $t_1 = 2$, $t_2 = 14$ and $t = t_1$. This is a partial step. After the step,

$$x^T = (-1 \ 0), u = (0 \ 2)^T \text{ and } f = -4.$$

Therefore, we can drop constraint 2 from the basis and update H , N^* , A and q .

$$H = \begin{pmatrix} 2/6 & 1/6 \\ 1/6 & 2/6 \end{pmatrix}, A = \{\phi\} \text{ and } q = 0.$$

The slack corresponding to the constraint 3 after the step is -3.

Iteration 3.

$E(x, \{\phi\}, 3)$ is a VDS. We can take a full step as follows:

$$z^T = (1/2 \ 1/2), z^T n^+ = 1, t_2 = 3 \text{ and } t = t_2.$$

$$x^T = (1/2 \ 3/2) \quad u = 5 \text{ and } f = 13/2.$$

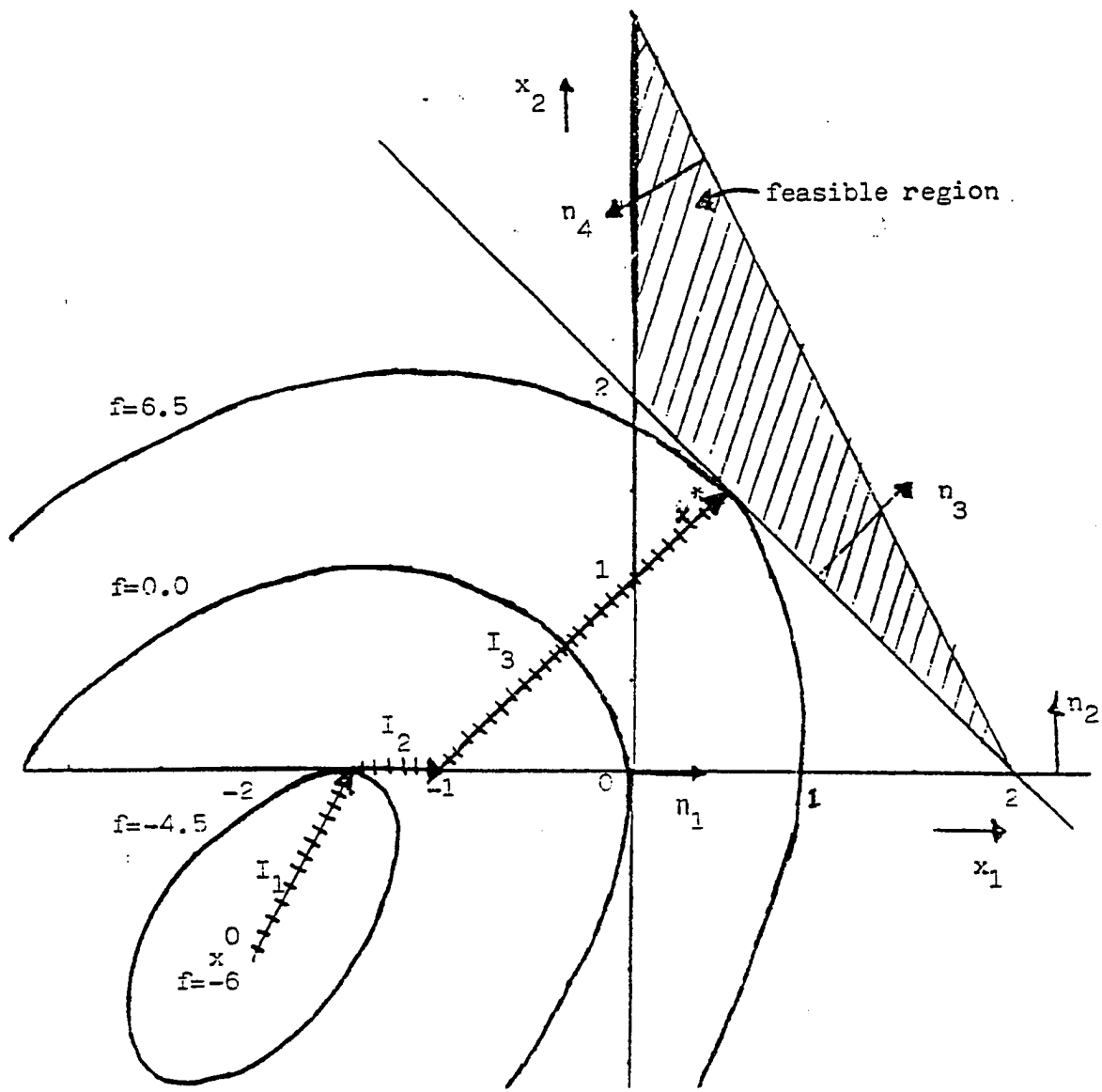
Adding constraint 3 to the basis and updating H , N^* , A and q we obtain

$$H = \begin{pmatrix} 1/12 & -1/12 \\ -1/12 & 1/12 \end{pmatrix}, N^* = (1/2 \ 1/2), A = \{3\} \text{ and } q = 1.$$

Iteration 4.

The s vector at this new x is $(1 \ 3 \ 0 \ 3)^T$. Since $s \geq 0$, $x = \begin{pmatrix} 1/2 \\ 3/2 \end{pmatrix}$ solves the problem with function value $13/2$.

See figure 1 on page 25 for geometric interpretation of these iterations.



---> A solution path

n_1, n_2, n_3, n_4 inward directed constraint normals

x^0 - unconstrained minimum

I_1 - full step, iteration 1

x^* - constrained minimum

I_2 - partial step, iteration 2

x_1, x_2 variables, axes

I_3 - full step, iteration 3

Figure 1.

Geometric interpretation of the example.

Primal-dual Method

In order to derive the primal-dual (P-D) method, we need to prove the following theorem on active constraints.

Theorem 5.

Let x solve a subproblem SP_1 with respect to q active linearly independent constraints whose indices are in set A . Let p be the index of a violated constraint in $K \setminus A$, (i.e.) $s(\{p\}, x) < 0$. Then, \bar{x} , the solution to the new subproblem SP_2 , comprising of the constraints whose indices are in $A \cup \{p\}$, must have the p -th constraint active, that is, $s(\{p\}, \bar{x}) = 0$ and $f(\bar{x}) > f(x)$. (assuming SP_2 has a solution)

Proof:

Since $s(\{p\}, x) < 0$, the feasible region of SP_2 is a subset of the feasible region of SP_1 . Moreover, since f takes the lowest value among all the feasible solutions of SP_1 at x , the function value at any of the feasible solutions of SP_2 , say x' , must satisfy $f(x') > f(x)$. This implies $f(\bar{x}) > f(x)$. Furthermore, since the feasible region of SP_1 is convex, all the points on the line (x, \bar{x}) are in the feasible set of SP_1 .

Let us assume that p is not active at \bar{x} , (i.e.) $s(\{p\}, \bar{x}) > 0$. Let x' be a point on the p -th constraint and on the line (x, \bar{x}) . Then, x' can be written as a linear combination of the two extreme points x and \bar{x} and $x' \neq x \neq \bar{x}$. Since the function f is strictly convex, we have

$$f(x) < f(x') < f(\bar{x}).$$

Moreover, x' is a feasible point for SP_2 because it is in the feasible set of SP_1 and satisfies the p -th constraint. Then, x' must solve SP_2 and not \bar{x} , hence our assumption that p is not active at the solution is invalid●

The dual projection algorithm presented earlier solves a sequence of subproblems of the original problem. The primal methods due to Fletcher (12) and Goldfarb (16) are capable of solving a QP problem provided a feasible point to that problem is available. We can combine the

two methods as follows.

Let us assume that the operators H and N^* w.r.t. some active set A are available. Then, starting from an arbitrary point and applying Lemma 3, we can get to the manifold M . Now, any primal method can be invoked to solve this subproblem of q constraints in A , ignoring the other $k-q$ constraints. Defining a new subproblem as the currently solved subproblem and a violated constraint, if any, one can take a full dual step to arrive at a feasible point for the new subproblem or apply theorem 3 and take a full dual step if possible. (A subproblem may be of only the active constraints at the solution.) If all the LMs are of correct sign and the projected gradient is zero, we can repeat this dual step. Otherwise, we switch to the primal algorithm to solve this subproblem before returning to the dual.

Primal-Dual Algorithm

- Step 1. Let x solve a subproblem with q active linearly independent constraints in set A . H , N^* and u are available. (apply Lemma 3 and/or primal method if needed)
- Step 2. Same as step 2 of dual projection method.
- Step 3. Same as step 3 of dual projection method.
- Step 4. Compute $t = t_2 = -s(\{p\}, x)/(z^T n^+)$, $\bar{x} = x + tz$, \bar{f} as in (3.28) and $u^+(\bar{x})$ as in (3.20). Compute $s(\{p\}, \bar{x})$.
Set $x \leftarrow \bar{x}$, $f \leftarrow \bar{f}$, $u \leftarrow u^+$, $A \leftarrow AU\{p\}$, $q = q+1$. Update H and N^* . If $u \geq 0$, then go to step 2. Otherwise, invoke primal to solve the subproblem whose indices are in A and go to step 2●

Finiteness of Primal-Dual Algorithm

Theorem 6.

The primal-dual algorithm converges to the global optimum, if any, in a finite number of steps or indicates that the feasible region is empty.

Proof:

At the start of the dual step, we have the solution of some subproblem of the original problem with the function value as f . The subproblem may be thought of as only the q active constraints in the set A . At this point, some constraint, say p , is violated. Consider the new subproblem as to be comprised of the constraints whose indices are in A^+ . The solution to this new subproblem can be found in either one full dual step, or if it is not optimal, by a primal algorithm which is invoked. Suppose that the optimal value of the new subproblem be \bar{f} . Since the new subproblem is more constrained, $\bar{f} > f$ as shown in theorem 5. Since the objective function is monotonically increasing, we will never consider the same subproblem again at the beginning of a dual step. Moreover, since there are finite number of possible subproblems, the method must converge to the optimum point in a finite number of steps or must terminate with no solution in the dual step●

Linear Dependence

Unlike the primal algorithms, the dual method has a choice of the constraint to be added into the basis among the violated constraints. If possible, one could postpone adding a nearly linearly dependent constraint until a vertex is obtained, or apply the theorem 3 in order to avoid the problems of loss of accuracy in the projection or multiplier operators.

Equalities

The problem as presented in (2.1), (2.2) has only inequalities. In most practical problems, it is possible that some of the constraints may be equalities. This does not pose any additional problem to the dual methods. One could add the equality constraints first and never allow them to leave the basis. The LMs associated with these constraint equalities must be unrestricted in sign.

4. NUMERICALLY STABLE IMPLEMENTATION

It has been shown in the literature (see Bartels et al.(1)) that the methods that use the orthogonal factors described in section 2 are numerically more stable than the methods that use the projection operators H and N^* . We will now show a way to update the orthogonal factors of B introduced in section 2, whenever a basis change takes place. First, we will present Givens rotations and the structure of the product of Givens rotation matrices. Later, we will show how a matrix can be multiplied with the Givens matrices in an efficient manner.

Givens Rotations

Let us consider the problem of rotating a given m vector y into a multiple of the first column of the identity matrix, e_1 , while maintaining its length. That is, determine a matrix P such that

$$P y = d e_1 \quad (4.1)$$

where d is a scalar equal to the length of the vector y , (i.e.),

$$d = \|y\|_2 \quad (4.2)$$

and P is an $m \times m$ orthogonal matrix. This reduction can be achieved by using a sequence of Givens rotation matrices P_j^i which rotate vectors only with respect to the $(i - j)$ th plane. A Givens matrix P_j^i is defined as an identity matrix, except that the i -th diagonal element is replaced by c , the j -th diagonal element replaced by $-c$, and the (i,j) and (j,i) elements replaced by h , where c and h are scalars such that $c^2 + h^2 = 1$. Note $P_j^i = (P_j^i)^T$. The scalars c and h can be determined from the fact that when P_j^i is applied to the vector y , it forms an m vector \bar{y} , which is the same as y , except that $\bar{y}_j = 0$ and \bar{y}_i is replaced with $\pm((y_i)^2 + (y_j)^2)^{1/2}$, so that \bar{y} has the same Euclidean length as y . Daniel, Gragg et al (6) describe a way to compute the scalars c and h of a Givens matrix which avoids the artificial problems of overflow and underflow along with the determination of the sign of \bar{y}_i .

There are several sequences of Givens matrices which will achieve the reduction (4.1). For

example,

$$P_2^1 P_3^2 \cdots P_{m-1}^{m-2} P_m^{m-1} y = d e_1 \quad (4.3)$$

Due to the structure of P_{j+1}^j , we only need to store the two scalars c and h of that Givens matrix rather than the whole matrix. Moreover, the product of the Givens matrices in (4.3) is an upper Hessenberg matrix. The special structure of such product matrices is shown in Gill, Golub, et al (13). They also provide the necessary recurrence relations, so that, one could calculate the product of the Givens matrices in (4.3) from the coefficients. Later we present an alternate implementation of the multiplication of a vector or a matrix with the Givens matrices which is more efficient and avoids the zero divide problems. Now we will present a method that uses Givens matrices to update the factors Q and R , similar to the one discussed in Goldfarb(17).

Adding a constraint.

We will assume that we know the factors of B as defined in (2.5). That is, we have

$$B = Q \begin{pmatrix} R \\ 0 \end{pmatrix} = L^{-1}N \quad (4.4)$$

Adding a constraint p , with normal n^+ , to the active set A , we have the factorization

$$B^+ = (B \ b^+) = Q^+ \begin{pmatrix} R^+ \\ 0 \end{pmatrix} \quad (4.5)$$

where $b^+ = L^{-1}n^+$. We can write

$$Q^T B^+ = (Q^T B \ Q^T b^+) = \begin{pmatrix} R \\ 0 \end{pmatrix} Q^T L^{-1}n^+ \quad (4.6)$$

where $Q^T L^{-1}n^+$ is an m vector, necessary for calculation of $N^* n^+$ and $H n^+$ as shown in section 2. Let Y be the product of the Givens matrices required to make $Q^T L^{-1}n^+$ the $q+1$ st column in the upper triangular matrix R^+ , by reducing the last $m-q-1$ elements to zero. The first q rows and columns of Y are identical to those of an identity matrix and the submatrix consisting of the rows and columns $q+1$ through m denoted as \bar{Y} is upper Hessenberg. i.e.,

$$Y = \begin{pmatrix} I_q & 0 \\ 0 & \bar{Y} \end{pmatrix} \quad (4.7)$$

Premultiplying both sides of (4.6) by Y we obtain,

$$Y Q^T B^+ = Y \begin{pmatrix} R \\ 0 \end{pmatrix} Q^T L^{-1} n^+ = \begin{pmatrix} R^+ \\ 0 \end{pmatrix} \quad (4.8)$$

and
$$Y Q^T = Q^{+T} \quad (4.9)$$

We can easily verify (4.5) from (4.8) and (4.9).

One can store and update $J = Q^T L^{-1}$ instead of Q. This can be accomplished as

$$Y J = J^+ \quad (4.10)$$

and
$$Y \begin{pmatrix} R & J n^+ \\ 0 & \end{pmatrix} = \begin{pmatrix} R^+ \\ 0 \end{pmatrix} \quad (4.11)$$

We can verify that $J^+ N^+ = \begin{pmatrix} R^+ \\ 0 \end{pmatrix}$. Updating of J results in fewer operations in the calculation of $Q^T L^{-1} n^+$ and the computation of the projected normal and the multipliers. Therefore, in our computer program we store J and R rather than Q and R.

Deleting p-th constraint, $1 \leq p \leq q$.

Removing the p-th column of N to form N^- is equivalent to removing the p-th column from R in (4.4). This results in a matrix with a subdiagonal in columns p through q-1. We will denote this matrix of R with p-th column removed as \hat{R} . The subdiagonals of size q-p of \hat{R} can be reduced to zero by multiplying \hat{R} by appropriate Givens rotation matrices. This multiplication will only affect the rows p through q-1 of \hat{R} . Let Y be the product of the Givens matrices required to produce the desired reduction. It can be shown that Y has the form

$$Y = \begin{pmatrix} I_{p-1} & 0 & 0 \\ 0 & \bar{Y} & 0 \\ 0 & 0 & I_{m-q} \end{pmatrix} \quad (4.12)$$

where \bar{Y} is a $(q-p+1) \times (q-p+1)$ lower Hessenberg matrix. We can write

$$Y Q^T L^{-1} N^- = Y \begin{bmatrix} \hat{R} \\ 0 \end{bmatrix} = \begin{bmatrix} R^- \\ 0 \end{bmatrix} \quad (4.13)$$

and

$$Y Q^T = (Q^-)^T \quad (4.14)$$

Again, we can verify that $(Q^-)^T B^- = \begin{bmatrix} R^- \\ 0 \end{bmatrix}$.

One can update $J = Q^T L^{-1}$ instead of Q^T in a like manner by replacing Q^T in (4.14) by J , $(Q^-)^T$ by J^- and $Q^T L^{-1}$ in (4.13) by J .

Multiplication with Givens Matrices

The updating of the factors is the most time consuming part of any projection algorithm and hence it is essential that these routines be very efficient. The upper Hessenberg matrix of the product of the Givens matrices shown in (4.3) can be obtained by using the recurrence relations shown in Gill, Golub, et al (13). These recurrences determine two vectors and any element of the matrix \bar{Y} can be determined from these two vectors and/or the Givens matrix coefficients. This is convenient when one wants to know the product matrix explicitly. The recurrence relations however, break down when implemented in a computer program when one of the Givens matrix coefficient is zero, resulting in zero-divide problems, even though they are algebraically correct. We will present an efficient implementation of updating the factors, without calculating the product of Givens matrices. Since there is no division with possible zero in the denominator involved, it can be implemented in a computer program with no checking for zero-divide. To illustrate, let us consider the problem of reducing a vector of two elements, w_1 and w_2 into w and 0. The Givens matrix coefficients required to produce this reduction are c and h given by

$$w = (w_1^2 + w_2^2)^{1/2}$$

$$c = w_1/w$$

$$h = w_2/w.$$

Define
$$y = w_2 / (w_1 + w) \tag{4.15}$$

(If $w_1 = w_2 = 0$, then there is no reduction necessary.) The above has no zero-divide possibility. Assuming c , h and y are available, then consider the problem of determining the product of a two element vector $(v_1 \ v_2)^T$ with the Givens matrix. This product can be accomplished in 3 multiplications rather than 4 and is shown in Gill, et al (13) as follows:

$$\begin{pmatrix} c & h \\ h & -c \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \begin{pmatrix} \bar{v}_1 \\ \bar{v}_2 \end{pmatrix} \tag{4.16}$$

$$\bar{v}_1 = v_1 c + v_2 h \tag{4.17}$$

$$\begin{aligned} \bar{v}_2 &= v_1 h - v_2 c \\ &= (v_1 + \bar{v}_1) y - v_2 \end{aligned} \tag{4.18}$$

The series of products as shown in (4.3) can be calculated from right to left in the above manner at the cost of $3(m-1)$ multiplications. The additional work required to calculate the y vector is $(m-1)$ multiplications. Thus, one can update m column vectors in the above manner at the cost of $3m(m-1) + m-1$, excluding the cost of calculating the Givens matrix coefficients, c and h . Another approach is described in Gill, Murray, et al (14) for finding the product with Givens matrices using formulae similar to the ones described here. However, rotations are always between j -th and m -th variables as opposed to j -th and $j+1$ -st.

X-correction and Reinversion

In our implementation of the algorithms, whenever a constraint is added to the basis, a check is made to make sure that all the basis constraints are in fact active to some tolerance. A correction procedure may be needed to make sure x is on the desired manifold. This procedure is called as x -correction and is achieved by implementing lemma 3. This procedure will work as long as the errors in the factors are not significant. When they are significant, it may be necessary to recompute the factors from scratch. This is quite expensive and is called reinversion. No x -correction or reinversion was invoked for all problems tested with our numerically stable implementation using J and double precision arithmetic on IBM 370/168. This was true for all the primal and dual methods.

5. RELATIONSHIP TO OTHER METHODS AND LINEAR PROGRAMMING (LP)

Associated with every primal problem as stated in (2.1), (2.2) there exists a corresponding dual problem as shown in Dorn (9), with both problems possessing the same solution, if there is one. This is true for LP as well. There are two methods of solving the above problems and they are called primal and dual methods. It has been shown in the literature that for LP, the primal method on the primal problem and the dual method on the dual problem are equivalent. Similarly, for LP, the primal method on the dual problem and the dual method on the primal problem are equivalent. Van de Panne and Whinston (38) VW, have shown that the primal method (simplex-like) for QPP is a straightforward extension of the simplex method for LP. VW have also shown that applying the primal method on the dual problem of the original QP, is straightforward extension of the dual simplex method for LP. Goldfarb (16) shows that the Dantzig-Wolfe simplex-like method for QP takes the same path as one of the primal projection methods when started from same point basis. This primal projection method is also identical to Fletcher's (12) QP algorithm for positive definite case. The computational requirements are also shown in (16) for the tableau and projection methods. The simplex-like methods of Dantzig-Wolfe and VW are identical. We have solved the primal problem with a dual projection method. We will show that this approach takes the same path as the dual tableau method of VW (See Appendix C) which solves the dual problem. We will also present the computational requirements for both the methods.

Dual Problem

The dual problem of the problem (2.1), (2.2) can be written as shown in Dorn (9) as

$$\text{maximize} \quad b^T u - 1/2 x^T G x \quad (5.1)$$

$$\text{subject to} \quad G x - C u = -a \quad (5.2)$$

$$u \geq 0 \quad (5.3)$$

The Lagrangean of the above problem can be written as

$$L(u,x,w) = b^T u - 1/2 x^T G x - w^T (Gx - Cu + a) \quad (5.4)$$

where w is the LM of the constraints in (5.2). Setting the partial derivative of $L(u,x,w)$ w.r.t. x , to zero, we find that

$$x = -w \quad (5.5)$$

The partial derivative of $L(u,x,w)$ w.r.t. w when set to zero yields (5.2) and the partial derivative w.r.t. u gives, using (5.5),

$$b - C^T x = -s \quad (5.6)$$

Since $u \geq 0$, the K-T optimality conditions for the above dual problem should satisfy (5.2), (5.3), (5.6) and (5.7).

$$s \geq 0 \quad (5.7)$$

In addition, the complementarity conditions (5.8) must also be satisfied.

$$u^T s = 0 \quad (5.8)$$

It is important to note that the above constraints for the dual problem are a restatement of the feasibility and optimality conditions of the primal problem. It can be shown (see Dorn(9)) that the function values are the same for the primal and dual problems at the solution.

Relationship to dual tableau method

We can write the conditions (5.2), (5.6) in tableau form as in T_0 . The matrix C^T and the vectors s , u and b are partitioned into two parts, with the first one containing q rows and the second part containing the remaining $k-q$ rows. They are indicated by superscripts for the vectors s , b and u , whereas the two parts of the matrix C are denoted by N_1 and N_2 . Tableau T_0 is the setup tableau in VW. A basic solution to T_0 can be obtained by making x , s^1 , and s^2 basic. That is, by defining B_1 to be the basic matrix, the basic feasible solution can be accomplished by premultiplying T_0 by B_1^{-1} . This basic solution is shown in $T_1 = B_1^{-1}T_0$.

variables	x	s ¹	s ²	u ¹	u ²	=	value
	G	0	0	-N ₁	-N ₂		-a
	N ₁ ^T	-I	0	0	0		b ¹
	N ₂ ^T	0	-I	0	0		b ²

Tableau T₀.

$$B_1 = \begin{pmatrix} G & 0 & 0 \\ N_1^T & -I & 0 \\ N_2^T & 0 & -I \end{pmatrix} \quad (5.9)$$

$$B_1^{-1} = \begin{pmatrix} G^{-1} & 0 & 0 \\ N_1^T G^{-1} & -I & 0 \\ N_2^T G^{-1} & 0 & -I \end{pmatrix} \quad (5.10)$$

variable	x	s ¹	s ²	u ¹	u ²	=	value
x	I	0	0	-G ⁻¹ N ₁	-G ⁻¹ N ₂		-G ⁻¹ a
s ¹	0	I	0	-N ₁ ^T G ⁻¹ N ₁	-N ₁ ^T G ⁻¹ N ₂		-N ₁ ^T G ⁻¹ a-b ¹
s ²	0	0	I	-N ₂ ^T G ⁻¹ N ₁	-N ₂ ^T G ⁻¹ N ₂		-N ₂ ^T G ⁻¹ a-b ²

Tableau T₁.

variable	x	s^1	s^2	u^1	u^2	=	value
x	I	$-D^T$	0	0	$-EN_2$		$D^T b^1 - Ea$
s^2	0	$-N_2^T D^T$	I	0	$-N_2^T EN_2$		$N_2^T (D^T b^1 - Ea) - b^2$
u^1	0	$-(N_1^T G^{-1} N_1)^{-1}$	0	I	DN_2		$(N_1^T G^{-1} N_1)^{-1} b^1 + Da$

Tableau T_2 .

Tableau method	Projection method
E	H
D	N^*
u^1	u
s^2	$s(K \setminus A, x)$
x	x
s^1	$s(A, x) = 0$
u^2	LM of inactive = 0 constraints

Table 5.1

Relationship between the tableau and projection methods.

Tableau T_1 is the initial tableau (i) of VW and corresponds to the unconstrained minimum in our projection method. We can arrive at a standard tableau as defined by VW with x, s^2, u^1 basic and s^1, u^2 as nonbasic from tableau T_0 or T_1 by a sequence of simplex pivots or by premultiplying T_0 by the inverse of the basis matrix. We will call this tableau T_2 . The inverse of this basis B_2 is shown in (5.11).

$$B_2^{-1} = \begin{pmatrix} E & D^T & 0 \\ N_2^T E & N_2^T D^T & -I \\ -D & (N_1^T G^{-1} N_1)^{-1} & 0 \end{pmatrix} \quad (5.11)$$

where $D = (N_1^T G^{-1} N_1)^{-1} N_1^T G^{-1}$, and $E = G^{-1} - G^{-1} N_1 D$. Tableau T_2 is the initial tableau (ii) in VW. In the basic solution corresponding to T_2 , $s^1 = 0$, (i.e.) $N_1^T x = b^1$, therefore, $N_1 = N$, the normals of q linearly independent active constraints in our case. x corresponds to the minimum of the objective function on the manifold of the intersection of the q active constraints as shown in theorem 1. It is easy to see that the standard tableau in VW corresponds to minimum on a manifold and the relationships between the tableau method and our projection method are shown in table 5.1.

The rule 1 in the tableau method picks a new basic variable u corresponding to a negative s^2 to enter the basis. In the projection method, this corresponds to picking a violated constraint. Rule 2 selects the variable to leave the basis from the set of basic u variables, u^1 and the primal member of the basic pair corresponding to the entering variable according to

$$\begin{aligned} \min. & \quad (d_j/w_j) \\ & \quad j \in \hat{J} \\ & \quad w_j > 0 \end{aligned} \quad (5.12)$$

where d_j is the value of the basic variable,

w_j is the value corresponding to the entering variable column and

\hat{J} is the set of rows associated with the basic u variables, u^1 and the primal member of the basic pair. In the projection method, this is equivalent to calculating the step length t . Since D is N^* , a column of DN_2 corresponds to the IM w.r.t. that normal in our case. Moreover, the element of $N_2^T E N_2$ corresponding to the most negative s variable and its basic pair is $n^{+T} H n^+$ in our notation. The ratios (5.12) correspond to (3.24), (3.25) and (3.26) in the projection method. Thus, both methods result in the same step length and the same step direction. If the leaving variable is the s variable corresponding to the entering u variable, then another standard tableau is obtained. This in our method is equivalent to taking a full step,

i.e., $t = t_2$ and adding the constraint into the basis.

If the leaving variable is a u variable, this corresponds to taking a partial step in our case, (i.e.) $t = t_1$ and updating the tableau corresponds to dropping the constraint from the basis. The resultant tableau is termed a non-standard tableau in VW. At some step, therefore, suppose that we have a non-standard tableau, with $x, s^2, u_1^1, \dots, u_{q-1}^1$ and u_1^2 basic. This tableau T_3 can be obtained by pivoting on T_2 with the last element in u_1^2 column as the pivot. The x variables corresponding to this step get changed as follows:

$$x = x + u_q^1 E d / (e_q^T D d) \quad (5.13)$$

where d is the first column of N_2 and e_q is the q-th column vector of identity matrix of size q . This in our notation is

$$x = x + t_1 H n^+ \quad (5.14)$$

where $t_1 = u_q^+(x)/r_q$. u_1^2 becomes $u_q^1/e_q^T D d$, which corresponds in our method to the updating of the LM of the constraint to be added into the basis. The updating of the other u variables is the same as updating LM in our case. Thus, the step direction and length are identical and updating of the tableau is equivalent to dropping the q-th constraint from the basis. Defining \bar{u}^1 as u^1 with u_q^1 replaced by u_1^2 and \bar{u}^2 as u_2 with u_1^2 replaced by u_q^1 , we find that

variable	x	s^1	s^2	\bar{u}^1	\bar{u}^2
x	1	$-\bar{D}^T$	0	0	$-\bar{E}\bar{N}_2$
s^2	0	$-\bar{N}_2\bar{D}^T$	1	0	$-\bar{N}_2^T\bar{E}\bar{N}_2$
\bar{u}^1	0	$-(\bar{N}_1^T G^{-1} \bar{N}_1)^{-1}$	0	1	$-\bar{D}\bar{N}_2$

Tableau T_3 .

where \bar{N}_1 and \bar{N}_2 are different from N_1 and N_2 respectively in the same fashion as \bar{u}^1 and \bar{u}^2 differ from u^1 and u^2 and \bar{D} and \bar{E} are defined as D and E respectively with the appropriate change in N_1 (that is, N_1 in the definition is replaced by \bar{N}_1). (s_1^2, u_1^2) form the basic pair and

(s_q^1, u_q^1) form the non-basic pair in this non-standard tableau. The choice of entering variable is the primal member of the nonbasic pair, namely, s_1^2 is in \hat{J} . This is same as calculating the step length as shown before. Therefore, the two methods will take the same path in this case. Since there can be only standard and non-standard tableau, our demonstration is complete.

Although the two algorithms follow identical paths when started from the same dual feasible point, they differ considerably in terms of the amount of work per iteration. The dual method in tableau form costs $(m+k+1)(k+1)+q+1$ multiplications and/or divisions per iteration. If one were to take advantage of the symmetry, the above number is reduced by $k(k-1)/2$. At the same time, if one were to implement the dual projection method using the projection operators, H and N^* , assuming symmetry, the number of operations are given by the following formulae.

a) iteration where a constraint is dropped

$$\frac{3}{2} m (m+1) + 2mq + 4m + q + 2$$

b) iteration where a constraint is added

$$\frac{3}{2} m (m+1) + 2mq + q + 2 + (k-q)m$$

Under the most unfavourable conditions for the projection method, (i.e.), $q=m$ when dropping a constraint and $q=0$ when adding a constraint, the projection method takes fewer operations, for k slightly greater than m while dropping and for k greater than $1.73m$ when adding a constraint. It is interesting to note that when dropping a constraint, the number of operations do not involve k . It is because of the fact that the slack of the constraint to be added is the only one required to be known at this stage. Thus we have shown that even under the most unfavourable conditions for the dual projection method, it costs less per iteration than the dual tableau method of VW for k slightly greater than m and $1.73 m$ for an iteration with a drop and an iteration with an add respectively.

6. QUADRATIC PROGRAMS IN NON-LINEAR PROGRAMS (NLP)

Powell (31) presents a fast algorithm for solving general NLP based upon a variable metric approach due to Han (23,24) and Biggs (3). This NLP method solves a series of positive definite QPPs. The solution of the QPP is used to determine the step direction for the original NLP and the LMs at the solution of the QPP are used to update the positive definite quadratic form for the next invocation of the QPP. It has been shown by Powell (30,31) that the method achieves superlinear convergence. Schittkowsky (35,36) in a recent report has compared 13 different NLP codes with randomly generated test problems and comments favourably on Powell's algorithm which uses Fletcher's QP algorithm to solve the QPPs. He also notes on Powell's code: "the calculation time seems to be correlated to the number of variables and might not be most effective for solving large problems". Since our dual method is efficient for large problems (see section 8.2), we expect Powell's algorithm with our dual method for QPPs to be more effective.

Description of Powell's NLP algorithm

Consider the following general NLP.

$$\text{minimize } F(x) \quad (6.1)$$

$$\text{subject to } c_i(x) = 0, \quad i=1,2,\dots,k' \quad (6.2a)$$

$$c_i(x) \geq 0, \quad i=k'+1,\dots,k. \quad (6.2b)$$

where x is a vector of m real variables and $F(x)$ is a real function. The objective and constraint functions are differentiable and the first derivatives can be calculated. Let $g = g(x)$ be the gradient vector

$$g = \nabla_x F(x). \quad (6.3)$$

At each iteration of the NLP, a starting point x and g are computed. A positive definite matrix, G say, which is set to the unit matrix initially, defines the current metric. The vector d that minimizes

$$Q(d) = d^T g + 1/2 d^T G d, \quad (6.4)$$

the quadratic function (6.4) subject to constraints obtained by linearizing the constraints in (6.2) about x , is computed. It is used as a search direction in the x variable space, x being replaced by the vector x' ,

$$x' = x + y d \quad (6.5)$$

where y is a positive multiplier whose value depends on the form of the function of one variable,

$$\psi(y) = F(x + y d) \quad (6.6)$$

The matrix G is revised, using gradients g and $g' = \nabla F(x')$ and another iteration is begun.

The constraints for the QPP at each iteration can be formulated as follows:

$$\nabla c_i^T d + c_i = 0, i = 1, 2, \dots, k'$$

$$\nabla c_i^T d + c_i \geq 0, i = k' + 1, \dots, k \quad (6.7)$$

where $c_i = c_i(x)$. It may happen that the above linear constraints (6.7) on d are inconsistent. This does not necessarily imply that the constraints in (6.2) are inconsistent. Therefore, an extra variable v , is introduced into the QP calculations and replace the constraints (6.7) by the conditions

$$\nabla c_i^T d + c_i v = 0, i = 1, 2, \dots, k'$$

$$\nabla c_i^T d + c_i v_i \geq 0, i = k' + 1, \dots, k \quad (6.8)$$

where v_i has the value

$$v_i = 1, c_i > 0$$

$$v_i = v, c_i \leq 0 \quad (6.9)$$

and v is made as large as possible subject to the condition

$$0 \leq v \leq 1. \quad (6.10)$$

Any positive value of v allows a helpful correction to x of the form (6.5). For more detailed description of the method see Powell (29,30,31).

This new variable v appears only in the linear part of the objective function of the QP in Powell's original code which uses Fletcher's QP algorithm. This linear term is given a very high penalty so that the v variable will take a value as close to 1 as possible. Since the dual method requires a positive definite quadratic form, the objective function is modified in both the quadratic and linear terms in the form of $1/2 p v^2 - 2 p v$, rather than $- p v$, as in Powell's original code, where p is a very large positive number, say 10^6 . This modification does not affect any iterations of the NLP algorithm or Fletcher's QP.

Numerical results

Before we present various results, it is important to note some major differences in the implementation of Fletcher's QP algorithm and our dual projection algorithm used in Powell's algorithm described earlier. The NLP iterations however, were essentially the same for both.

Differences in implementation

The interface between the NLP code and the QP code is different in the sense that the original code using Fletcher's QP requires upper and lower bounds on all the variables of the QPP, whereas, for the dual QP, no such bounds are necessary except for the new variable v whose bounds are treated as constraints.

The LM is readily available at the solution of the dual method and is calculated in Powell's original code after the problem is solved.

Fletcher's QP method was implemented using the projection operators H and N^* . The dual method used the numerically stable implementation discussed in section 4. Number of operations (multiplications and/or divisions) required to update the operators H and N^* when there is a change in the basis equals (assuming symmetry)

$$(3/2) m (m+1) + m + 2 m q,$$

where q is the number of active constraints in the current basis. The corresponding formulae for updating the factors in the dual method are

$$4 m^2 - 3 m q + m - 4 q - 4,$$

when adding a constraint and

$$(q - p) (3 m + 3 (q-p-1)/2 + 4)$$

when dropping the p -th constraint. The formulae for the factors assume that $Q^T L^{-1}$ is stored and updated rather than Q .

Fletcher's QP was used with $\text{mode} = 1$, which means that a general QP is being solved and the inverse of $\begin{pmatrix} G & N \\ N^T & 0 \end{pmatrix}$ is computed to check the feasibility and optimality conditions at the solution, where N is the matrix of active normals. Dual method does not compute such inverse at the solution.

An LP routine is used in Fletcher's QP algorithm to find an initial feasible vertex. At a vertex, the projection operators are just $H = 0$ and $N^* = N^{-1}$. Moreover, the old basis is included in the infeasible basis at the start of the feasible point routine. The dual method was implemented in two different ways. In the first implementation preference was given to the active constraints of the previous iteration when adding constraints to the basis, while the second implementation ignored any such information from the previous iteration. The first type was found to be approximately 20% less expensive than the second and resulted in fewer drops. The results presented are for the dual method of the first type. It was observed that the intersection of the active constraints at the successive iterations of the QPP was very high and should help Fletcher's QP algorithm substantially. In fact, it was 100% for four of the six problems and over 90% for the other two problems.

Problems tested

We solved six NLP problems with the necessary modifications to Powell's code which uses

Fletcher's QP algorithm and our dual projection method. We report here on our experience with a problem suggested by Powell (1969) (31), Rosenbrock's post office parcel problem (POP), Powell's triangle problem (29) and three of Colville's test problems (see Dembo(8) and Himmelblau(25)). Their characteristics are as shown in table 6.1 in appendix D.

Discussion of results

The last column in table 6.1 shows the average cardinality of the active basis at the solution of the QPPs and the column of the variables under the QP characteristics indicates the maximum possible active constraints at the QP solution (in other words, for a vertex to be the solution). The manifold on which the QP solutions occur is always a vertex in two of the six problems and no more than 2 dimensional manifolds in the remaining four problems.

The dual method outperformed the primal method in all cases in terms of the total number of operations. Fletcher's (10,11) QP and LP codes do not require any square root evaluations. In most cases, the dual method required only a little more effort to solve the problems than the feasible point routine (FP) in the primal method. The ratio of the number of operations for FP to the dual (including square roots at 10 operations) ranges from 0.38 to 1.19 with an average of 0.77. This ratio is the lowest for Powell's problem which has only equality constraints. The range of the same ratio for FQ (Fletcher's QP routine) to dual is 1.44 to 2.57 with an average of 1.91. This range is 1.85 to 3.33 for (FP+FQ)/D with an average of 2.7. It is the lowest for Powell's problem (1.85) and is the highest for Colville 2 (3.33). For triangle problem it is 3.21. It is interesting to note that this ratio is high for large variable problems indicating that the dual is more efficient for large problems.

The number of steps in the dual is more than the primal method, mainly because the series of QPPs generated are such that they favour Fletcher's QP method and the feasible point routine. Even under such circumstances, the dual is more efficient. The primal hardly ever adds while the dual hardly ever drops.

7. STEEPEST EDGE OR FACE CONSIDERATIONS

In primal algorithms one must, at certain points, decide which constraint(s) in the active set to drop. The usual practice is that the constraint corresponding to the most negative LM is dropped. This choice is also the simplest computationally. One could drop a constraint among the eligible constraints in such a way that the rate of decrease of the objective function in an appropriate metric is the greatest. If that metric is G , it has been shown in Goldfarb (19) that the constraint to be dropped can be found for primal algorithms as the one that maximizes the following function.

$$\max_{u_j > 0} \left\{ u_j^2 / e_j^T (N^T G^{-1} N)^{-1} e_j \right\} \quad (7.1)$$

Since $(N^T G^{-1} N)^{-1}$ is not stored or updated, it can be calculated from N^* as shown in (7.2).

$$(N^T G^{-1} N)^{-1} = N^* G N^{*T} \quad (7.2)$$

Similar method for LP is described in Goldfarb (18). Even though we only require the diagonal elements of (7.2), the above process is quite expensive. It is therefore convenient to update only the required elements of (7.2) whenever a basis change takes place. This can be done at the cost of $O(n^2)$ operations as shown in Goldfarb (20).

Adding a constraint to the basis

Let h_i be the i -th diagonal element of (7.2). The recurrences for updating h when a constraint whose normal is n^+ is added to the basis is given by

$$\begin{aligned} \bar{h}^+ &= 1 / n^{+T} H n^+ \\ \bar{h}_i &= h_i + r_i^2 \bar{h}^+, \quad i = 1, 2, \dots, q \end{aligned} \quad (7.3)$$

where r_i is the i -th infeasibility multiplier w.r.t. n^+ and \bar{h} is the updated h . This updating can be accomplished at the cost of $m^2 + mq + m + 2q + 1$ multiplications.

Dropping a constraint from the basis

Let us define a q vector v as follows:

$$v = N^* G n^{-*T} \quad (7.4)$$

Then

$$\bar{h}_i = h_i - v_i^2 / h^- \quad (7.5)$$

where h^- is the value of h corresponding to the constraint to be dropped. This updating costs $m^2 + mq + 2q - 2$ operations.

The above updating formulae use the projection operators H and N^* . However, if one were to use the orthogonal factors discussed in section 4, the corresponding updating can be done as shown below.

Adding a constraint with factors.

Compute $w = Q^T L^{-1} n^+$. Let us denote the first q elements of w as w_1 and let w_2 be the length of the vector comprising of the last $m-q$ elements of w . Then the vector h can be updated as

$$\bar{h}^+ = (1/w_2)^2 \quad (7.6)$$

$$\bar{h}_i = h_i + e_i^T R^{-1} w_1 w_1^T R^{-T} e_i \bar{h}^+, \quad i = 1, 2, \dots, q$$

This costs $q(q+1)/2 + 2q + 2$ multiplications assuming that w_1 and w_2 are readily available.

Dropping a constraint with factors

If p -th constraint is being dropped, where $p \leq q$, the vector

$$v = R^{-1} R^{-T} e_p \quad (7.7)$$

can be computed and formula (7.5) is now applicable. This costs $q(q+1)/2 + 2q - 2 + (q-p+1)(q-p+2)/2$ operations.

Greatest increase choice in the dual method

There is no choice to be made of the constraint to be dropped from the basis in the dual method. However, it is possible to choose the constraint to be added. One possibility is to choose the one that will produce the greatest increase in the objective function if a full step is taken. This choice of adding a constraint is made after a full step is taken or at the beginning of the algorithm. As in the primal methods, this does not necessarily, theoretically decrease the number of steps or basis changes needed to obtain the solution.

If a full step is taken, the change in the objective function is given by

$$(s(\{p\},x))^2 / n^{+T} H n^+ \quad (7.9)$$

when $|Hn^+| \neq 0$. Using the projection operators, this will cost $m^2 + m + 2$ operations, assuming the s vector is available for every violated constraint. This adds considerably to the cost in the dual method. Therefore, one could apply this method of picking the constraint after the basis is almost full.

When $|Hn^+| = 0$, the corresponding formula to (7.9) is given by

$$1/2 (s(\{p\},x))^2 / n^{+T} H^- n^+ - s(\{p\},x) u^- / r^-. \quad (7.10)$$

where u^- and r^- are the LM and IM corresponding to the constraint to be dropped as shown in theorem 3.

8. PERFORMANCE COMPARISON WITH OTHER METHODS

In this section, six primal methods and the dual method are compared with respect to the number of steps, basis changes and operations. (multiplications and divisions and square root calculations, with 1 square root = 10 multiplications) All of the primal methods were started from the same feasible point for each given problem. The feasible point routine started with a random starting point, and had no control of the first feasible basis. One could also control the basis for the first feasible point by applying Lemma 3, but, this was not done. The average intersection of initial and final bases was found to be 0.5 times the number of constraints active at the solution; (i.e.) on an average, half of the final basis appeared in the initial basis. The feasible point routine used the algorithm suggested in Goldfarb (16), which is described in detail in Idnani (26). It is a generalization of a feasible point routine given by Rosen (33) applied to QPP. The dual method always used the unconstrained minimum as the initial point. The computer program was capable of solving the QPP with six primal-dual methods as well, one for each of the primal method. Thus, the program could run 14 different algorithms. The program listing is in appendix E.

The six primal methods are variations of the two primal methods of Goldfarb (16), one of which is identical to the one by Fletcher (12). (See Appendix A) We will denote these two primal methods as P.1 and P.2. (P.1 is identical to Fletcher's QP.) The difference between P.1 and P.2 is that, P.2 allows more than one constraint to be dropped at a point if certain additional conditions on the LM holds. The sufficient condition on the additional requirements is that the LM of a constraint to be dropped should be monotonically non-increasing as constraints are dropped at the same point. Using the steepest face considerations for primal methods, one could drop the constraint among eligible set as the one suggested in section 7.(see 7.1) This produces two more primal methods and we call them P.1.1 and P.2.1. Methods of type P.2 take steps on higher dimensional manifolds than methods of type P.1. The remaining two primal methods are variations of the primal method of type P.2 and are based on the necessary and sufficient conditions for dropping more than one constraint at a point. This

ensures that by dropping one of the currently active constraints, the step direction is such that any previously dropped constraint is not violated. These and the negativity of the LM are the necessary conditions for dropping more than one constraint. Since the step direction in the primal methods is the projected gradient, this can be calculated by updating recurrences as shown in Goldfarb (20). This uses similar information as the steepest face algorithms, namely $R^{-1}R^{-T}$. So, one could derive a two phased primal algorithm for the dropping part of methods of type P.2 as follows.

Drop phase 1: Use the sufficient conditions for dropping more than one constraint in addition to LM being negative. When none can be dropped further, apply drop phase 2.

Drop phase 2: Among eligible constraints (all negative LM constraints are eligible), drop the one that would not result in a step direction which would violate a previously dropped constraint in both drop phases so far. If none can be dropped, take step, otherwise repeat drop phase 2.

Since this two phase approach can be applied to P.2 and P.2.1 only, we will denote these as P.2' and P.2'.1 respectively. Table 8.1 summarizes all the six primal methods.

The letter D will stand for the dual method and the prefix D. followed by any one of the primals, for the primal-dual method corresponding to the primal used. The letter F will denote the feasible point routine. The primal-dual methods were only verified for convergence. Rigorous testing was done for all the primal and dual methods.

Problem generation

For performance comparison, 24 different problem types with known solutions were constructed. These problems were randomly generated as described in Idnani (26) and in appendix B. The method used was the one by Rosen and Suzuki (34) modified for QPP. The problems generated had 9, 27 or 81 variables. The positive definite matrix G was generated in two different ways, such that G was either well conditioned or ill-conditioned. In the well

conditioned case, the eigenvalues were close and in the ill-conditioned case, they were spread. In either case, the off diagonal elements are generated first using random numbers between -1 and +1. Then the i-th diagonal element is calculated as follows:

Well conditioned G: $G_{ii} = RS(i) + R(0,1) + 1$

Ill conditioned G: $G_{ii} = RS(i) + R(0,1)$ for $i = 1$.

$$G_{ii} = RS(i) + R(0,1) + G_{i-1,i-1} + RS(i-1), i \neq 1$$

where $RS(i)$ is sum of absolute values of the off diagonal elements of the i-th row of G, and $R(a,b)$ is a random number between a and b.

	Author	symbol	remarks
1.	Fletcher & Goldfarb	P.1	at most one drop before taking step. Drop most negative LM constraint.
2.	Goldfarb	P.1.1	same as P.1, except, use steepest face considerations for drop.
3.	Goldfarb	P.2	Apply sufficient conditions for multidropping, drop most negative LM constraint.
4.	Goldfarb	P.2.1	same as P.2, except use steepest face considerations also.
5.	Goldfarb	P.2'	same as P.2. Once sufficient conditions exhausted for drop, apply necessary conditions.
6.	Goldfarb	P.2'.1	same as remarks for P.2', replace P.2 by P.2.1.

Table 8.1 Various Primal Methods.

The largest possible diagonal element in the well conditioned case is $m+2$. In the ill conditioned case, it is 13202 for the 81 variable problem. But, the expected value of these numbers is only half of the maximum. The maximum value of i-th diagonal element in the ill

conditioned case is given by $(2i-1)(m+1)$.

The number of constraints for an m variable problem were tested at two levels, $k = m$ and $k = 3m$. Thus the 81 variable problems had a set of problems with 81 constraints and another with 243 constraints. The number of active constraints at the solution was also kept at two levels, $k/9$ and $k/3$. The problems at $k/9$ level had close to an interior point as the solution, whereas, the problems with $k/3$ level could have a vertex as the solution. Thus, in all, 24 problem types as shown in table 8.2 were generated. (See appendix D.)

The coefficients in the constraint normals were generated between -1 and $+1$. The components of the solution vector x were generated between -5 and $+5$. The constraint normals were normalized. The slacks of inactive constraints lie between 0 and 1 . The LMs of active constraints at the solution were set randomly between 0 and 30 for run 1, for types of problems 1 to 16. For run 2, for the same set of problems, the LMs for active constraints were randomly generated between 0 and $30k$, where k is the number of constraints for the problem. The difference between runs 1 and 2 is that the run 2 has more linear objective function than run 1. Run 3 comprises of 81 variable problems with LMs between 0 and $81k$. The K-T conditions are used to calculate the vector a . Runs 1 and 2 were replicated five times, whereas, run 3 was replicated only once. (i.e.) 5 different seeds were used to generate a problem type which was then solved. Totally, 168 problems were generated. All the primal methods described earlier, the dual method and a primal-dual method using primal of type P.1 were tested in all the runs except in run 3, which did not test the primal-dual method.

Observation

The dual method required fewer operations in most problems than the feasible point routine used for the primal methods (see appendix D). In addition, the primal methods take at least as many operations as the feasible point routine. The only cases where the feasible point routine took fewer operations than the dual were for problems of type 7 and 8 for runs 1 and 2, and type 4, 15 for run 2 and 23 for run 3. The problems of type 7, 8, 15 and 23 have a vertex as

the optimum solution. Problem types 7, 15 and 23 also have well conditioned G , whereas, 4 and 8 have ill conditioned G . The condition of G does not have any effect on the feasible point routine. Surprisingly, the dual method is sensitive to the condition of G in runs 2 and 3 only. In these runs, where the objective function is more linear, the problems with ill conditioned G take fewer operations than those with well conditioned G . In the problems with a vertex as the solution, the difference is quite significant. This strange behaviour of the dual method can be explained as follows.

Problems of runs 2 and 3 have a heavily weighted linear function. The gradient of the function is written as $Gx + a$, can be thought of as the sum of two vectors. One of them is the constant vector a and the other is the variable vector Gx . For well conditioned G , the constant vector dominates over the variable vectors in runs 2 and 3. For the same runs when G is ill conditioned, the contribution by the variable part Gx could be significant, because G has positive diagonals with much higher value than the off diagonals. In our dual method, all along the step direction, $H^+g = 0$ which implies that the gradient can be written as a linear combination of the LMs, of the normals. When the change in g is significant, the LMs vary significantly. The rate of change of LM with IM is what determines t_1 , the partial step length. In other words, when there is little change in g , the LMs remain high permitting full steps, whereas, when the gradient changes rapidly, the LMs change rapidly and this results in partial steps. That is, the contribution of the curvature information in the well conditioned case is not as high as in the ill-conditioned case. Another explanation due to Goldfarb can be described as follows: Let x^* and x^0 be the constrained and unconstrained optimums for (2.1), (2.2). Let $d = x^* - x^0$ and $y = \|d\|_2$. Then, $d = G^{-1}\bar{N}\bar{u}$, where \bar{N} is the matrix of active normals at x^* and \bar{u} , the associated LMs. y is largest when G is well conditioned, \bar{N} is full and \bar{u} is large and these precisely are the problems which were most difficult to solve for the dual method. This is because the dual method is more likely to take full steps when y is large. Run 2 has a higher \bar{u} for the same problem type than run 1, and therefore, y is greater.

An interesting property of the dual method is that it remains on as high dimensional manifolds as possible. Of the 168 problems tested, it picked the correct basis without dropping in 110 problems (see tables 8.6 and 8.7). This was also confirmed in the NLP runs (see table 6.2). These tables appear in appendix D.

Table 8.8 shows the number of active constraints at the first feasible point for the primal methods and its intersection with the final basis. On an average half of the final basis appeared in the initial one.

Table 8.9 shows the average of the total number of basis changes per step for all the runs. To some extent, this ratio indicates the dimension of the manifold on which steps are taken. When this ratio is 1, for the dual, primal-dual and feasible point methods, it indicates that steps are taken on high dimensional manifolds. For primal methods, the higher the ratio, the more interior or the higher dimensional manifold, on which steps are taken. Methods of type P.2 have higher ratio than methods of type P.1. the highest ratio is for P.2'.1, but, there is an additional computational requirement to achieve this increase over P.2 as shown in tables 8.3 - 8.5. As expected, we see that this ratio for the dual method is less than that for the feasible point routine and is slightly greater than 1.

Table 8.10 shows the effect of the objective function's weighting on the linear term. When the function is made more linear, as in runs 2 and 3, the dual and primal-dual methods experience a much greater increase in the number of steps and basis changes than the primal methods. A substantial part of the increase comes from the problems where the solution is a vertex with a well conditioned G. Among the primal methods, P.1.1 decreased in number of steps, while the number of basis changes increased. Generally speaking, the linearity of the function had some effect on all the primal methods, but not to the same degree as in the dual methods.

Steepest face considerations

As shown in section 7, the steepest face considerations cost a little more at the time of updating the basis. Since only the diagonal elements of $(R^T R)^{-1}$, corresponding to the negative LMs or some other criterion which is even more stringent, are required, the recurrence relations as described in section 7 were not used in our computer program. Instead, only the necessary diagonal element was computed by first finding the appropriate row of R^{-1} and taking the inner product of it with itself. This can be accomplished for the p -th diagonal at the cost of $y(y+3)/2$ operations, where $y = q-p+1$. It is not clear which method is better. If one has to calculate even a few diagonal elements, it is possible that it can become expensive. The cost mainly depends upon $q-p$. If one were to recur, either, one has to recur during the feasible point routine as well or can calculate it after the feasible point is found.

An alternative to recurring the diagonal elements for methods of type P.2, is that, after the first drop takes place, one could recur only the diagonals of the eligible set of constraints with the idea that the eligibility set is very small. The steepest face directions do result in fewer steps and basis changes for all primals. This is not so significant in the primals of type P.2 and P.2' as in P.1. Method P.2 produced on the average the best results overall among all the six different primal methods tested.

Primal-dual methods

The primal-dual methods as implemented in our program must be modified a little in order to repeat the multiplier computations in the primal or dual part. These methods always start from the unconstrained minimum, although in general, they do not have to. The computer program must be modified accordingly for general use. Since the various primal-dual methods behaved very similarly when started from the unconstrained minimum, only D.P.1 was run. A peculiar phenomenon of dropping and adding occurred in the primal part of the D.P.1 method. It can be explained as follows:

Assume that x solves a subproblem of q active constraints, say, $1, 2, \dots, q$. Let $q+1$ be a linearly independent violated constraint. The D.P methods will take a full step and add the $q+1$ -st constraint to the basis in the dual part of the method. Now let the LMs of constraint 1 and 2 become negative. Therefore, the primal part is invoked to solve this problem of $q+1$ constraints. Assuming LM of constraint 2 to be the most negative, the primal method would drop the constraint 2 and take a full step to the minimum on the new manifold. Now, if LM of constraint 1 is negative, it would drop 1 and take a step. However, if constraint 2 will be violated in the new step direction, only a partial step can be taken and constraint 2 added back into the basis. The dual method unlike the primal-dual method, under the same circumstances, would have taken a partial step and dropped constraint 1 and then taken a full step to add constraint $q+1$. The difference between the dual and the primal-dual can be explained by the fact that the dual does not drop constraints according to the most negative LM, but, by the order in which they become negative in the step direction.

Summary

As far as the number of operations are concerned, the dual method required fewer operations overall for all problems. It is not surprising that the ratio of the number of operations for the primal methods to the dual method is quite high even under the most unfavourable conditions for the dual method. If one were to add the cost of finding a feasible point, the dual is 1.69 to 8.55 times faster than P.1 and 1.49 to 4.98 times faster than P.2. On the average, these numbers are of the order 4.8 and 3.2 respectively. There are other methods of finding a feasible point, such as the one used in Fletcher's QP method. This method favours problems with low dimensional manifold solutions and may be more efficient than the feasible point routine used in our program. Moreover, methods of type P.2 which drop more than one constraint at a point and move across the interior, are expected to benefit when the solution occurs on a high dimensional manifold. The dual method performs quite well for large and small problems, and, it is the starting point, in our belief, that reduces the number of iterations and results in fewer drops.

9. CONCLUSION

For positive definite QP, it is clear that the dual method is quite efficient compared with primal methods even under the most unfavorable conditions for the dual algorithm. More importantly, if the solution occurs on some high dimensional manifold, it is able to pick the correct basis without dropping constraints in most cases. If the operators, H , N^* or the factors are readily available, then, in our opinion, a primal-dual method is strongly recommended. Even though any one of the primal methods is suitable, P.2 is recommended for large problems.

The dual method is more efficient in terms of the number of operations than primal methods even when it takes more steps because:

- a) the gradient is never calculated if started from the unconstrained minimum.
- b) the gradient is never projected in the dual, whereas it is in the primal. Projection of a constraint normal takes place in both primal and dual methods in order to update the operators or factors. This projected normal is also used as the step direction in the dual method.
- c) starting at the unconstrained minimum, it remains on the infeasible side of the solution, i.e. it approaches the solution from the infeasible side of the manifold of the active constraints at the solution (whose LM are of the correct sign).
- d) starting at the unconstrained minimum does not require much more additional effort.
- e) step length calculations are not as expensive as in the primal methods where one has to calculate one or two inner products for the constraints not in the basis.

Another advantage of the dual method is that it has a choice of which constraint to add to the basis unlike the primal methods. Thus, it is possible to maintain a linearly independent basis to a great extent.

For our implementation, we store the matrices $Q^T L^{-1}$ and R , the vectors LM , IM and z in addition to the matrix of constraint normals, the matrix G and the vectors x , a and b . In all, it requires $3/2 m(m+1) + 2m$ locations in addition to the storage of the problem.

For applications in Powell's NLP algorithm, it is not clear if a primal-dual method is more suitable than the dual method. Since the solution basis of the previous iteration is very much identical to the solution basis of the current iteration, a primal-dual method may be useful. But, the quadratic form is different and the constraint normals may be different at each iteration, thus the projection operators may not readily be available and hence they have to be calculated. In such a case, a matrix inversion has to take place and application of the dual as described in section 6 seems appropriate.

Choice of constraint to add

The dual method as implemented in our programs always picked the most violated constraint to add to the basis among all the violated constraints. An alternative rule was presented in section 7 as the greatest increase choice. This involves much more computational effort and there is no guarantee that it will produce a significant improvement. We will present two more alternatives in this regard for future research and experimentation. Both methods choose the constraint to be added to the basis among the violated constraints as the one whose normal makes the least angle with another "base" vector. One choice for the base vector is the sum of all the violated constraint normals. In this, the infeasibility is used as penalty term for the choice of constraint to enter the basis. Another alternate strategy which can be used, except, at the unconstrained minimum, is to choose the gradient vector as the base vector. Hybrid strategies could also be used.

Linear Programming (LP)

It is the starting point, in our belief, that resulted in an efficient dual projection method for the QPPs. A similar method for initiating the LP may be efficient. Consider the LP:

$$\text{minimize } a^T x \quad (9.1)$$

$$\text{subject to } C^T x - b = s \geq 0 \quad (9.2)$$

$$l_i \leq x \leq h_i \quad (9.3)$$

where a, x, l, h are m vectors,

C is the $m \times k$ matrix of k constraint normals,

b is a k vector and

s is the k vector of nonnegative slacks.

The constraints in (9.3) are the simple bounds on the m variables.

The above LP is just a special case of the QP (2.1), (2.2). In fact, the dual projection algorithm developed in section 3 can be applied to this LP. It is well known that the solution of an LP has to occur at a vertex. We know that at a vertex $H = 0$ and $N^* = N^{-1}$, where N is the matrix of m linearly independent constraint normals active at the vertex. Notice that the operators are independent of the matrix G . If one starts at a dual feasible vertex and picks a violated constraint to add to the basis (say normal n^+), one can then apply theorem 3 and obtain the constraint to drop according to the least nonnegative ratio of LM/IM , where $LM = N^{-1}a$ and $IM = N^{-1}n^+$. Since the gradient is constant, a full dual step can be taken arriving at another vertex, that is, an exchange of the constraint to drop and add is sufficient and the new x can be computed. This is the dual simplex method for LP.

One can start at a vertex which can be thought of as comparable to the unconstrained optimum in the positive definite QPP. One such vertex can be obtained by solving (9.1) w.r.t. (9.3), in other words, by picking either the lower or the upper bound of x_i depending upon the sign of a_i . In the case when a_i is zero, one of the bounds is chosen. It can be verified that the LMs at such a starting point are of the correct LM sign and hence, the x so chosen is a dual feasible solution. The inverse of this basis matrix is available trivially.

It is easy to see that the increase in the objective function at every step is given by $-s(\{p\}, x) u^-/r^-$, which is nonnegative, and zero when u^- is zero, where $s(\{p\}, x)$ is the

violation of the constraint to be added and u^- , r^- are the LM and IM of the constraint to be dropped. The greatest increase of the objective function can be obtained by maximizing the above function with respect to all the constraints that are violated.

The same technique for the starting vertex can be used for phase 1 of the primal simplex method for LP as well and is expected to reduce computational effort during phase 2. Notice that this technique requires no additional computation and only further experimentation can show the gain by such an approach. Furthermore, the recommendations made earlier for the choice of the constraint to add for the QPP are applicable for LP. One can also use the same numerically stable computer implementation used for the positive definite QPP with a minor change, that is, by setting $G = L = I$ initially.

Dual Projection Method for Convex NLP

In the QPP, the matrix G was constant. However, for convex NLP, the corresponding matrix is the Hessian which is updated at every iteration, just like in the case of Powell's algorithm for NLP described in section 6. Moreover, the updated Hessian is always positive definite. Using ideas similar to the dual projection method for QPP, it is possible that one could develop a dual projection algorithm for solving convex NLP with linear constraints using quasi-Newton, modified Newton or Newton like methods. This is an area for future research.

General QPP

It is not clear at this time of writing how this method can be applied to QPP with positive semi-definite G or an indefinite G . One of the major problems is the starting point in these cases. This is also an area of future study.

Summary

An efficient dual projection algorithm and its implementation in a numerically stable fashion is presented for solving positive definite QPP applicable in solving general NLP problems. A primal-dual projection method is also presented. These methods derived from the primal problem has the advantage of geometric interpretation.

Appendix A

Primal projection methods.

Method P.1

Step 1) Compute some x in the feasible region, $g = Gx + a$. Compute H and N^* defined by the linearly independent defining hyperplanes on which x lies. Let A be the set of these constraints.

Step 2) Compute $z = Hg$; if $|z| = 0$ go to step 4.

Step 3) Set $x = x - tz$, compute $g = Gx + a$ and s where $t = \min. (1, y)$

$$\text{and} \quad y = \min. \begin{matrix} s_j / (n_j^T z) \\ n_j^T z > 0 \\ j \notin A \end{matrix} \quad (\text{A.1})$$

If $t < 1$, add constraint to the basis corresponding to the minimum value of y .

Update A, H, N^* . Go to step 2.

If $t = 1$, go to step 4.

Step 4) Compute $LM, u = N^*g$. Determine $\min. u_j = u_p$. If $u_p \geq 0$, terminate; x solves the problem. Otherwise, $u_p < 0$, drop constraint p from the basis, update A, H, N^* and go to step 2●

Method P.2

Step 1) Same as in P.1.

Step 2) Compute $LM, u = N^*g$. Set $i = 0, J_1 = A$. Determine $\min_{j \in J_1} u_j = u_p$. If $u_p \geq 0$, go to step 4. Otherwise, $u_p < 0$, go to step 3.

Step 3) Drop p -th constraint from the basis, update A, H, N^* and J_1 . Set $i = i + 1$. Compute $LM, u' = N^*g$. Determine $J_2 = \{j | u'_j \leq u_j\}$, set $J_1 = J_1 \cap J_2$ and $u \leftarrow u'$. Determine $\min_{j \in J_1} u_j = u_p$. If $u_p \geq 0$, go to step 4. Otherwise, repeat this step.

Step 4) Compute $z = Hg$. If $|z| = 0$ terminate; x solves the problem.

If $|z| \neq 0$, set $x = x - tz$, compute $g = Gx + a$ and s where $t = \min. (1, y)$ and y is given in (A.1).

If $t = 1$ and $i = 0$ terminate; x solves the problem.

If $t = 1$ and $i \neq 0$, go to step 2.

If $t < 1$, add constraint corresponding to the minimum value of y to the basis, update A, H, N^* and go to step 2●

Appendix B

Generation of Random positive definite QPPs with known solution.

For a given number of variables m , number of constraints k , number of active constraints at the solution q^* , the positive definite QPP can be generated using the Rosen-Suzuki(34) approach as follows. (We use the expression ' $\epsilon R(a,b)$ ' to mean element of a set of pseudo random numbers uniformly distributed between a and b .)

Step 1) *Solution*

- a) Generate the components of an arbitrary optimum point $x^* \in R(-5,5)$.
- b) Specify the index set A of q^* active constraints at the solution.

Step 2) *Positive definite matrix G*

- a) Generate the off-diagonal elements $\epsilon R(-1,1)$.
- b) Generate the diagonal elements of G as described in section 8 for well and ill conditioned cases of G .

Step 3) *Constraint matrix C*

- a) Generate the elements of $C \in R(-1,1)$.
- b) Normalize the columns of C to unit length.

Step 4) *Complementary Slackness*

- a) For $j \in A$, set $s_j = 0$ and generate LM, $u_j \in R(0,y)$ where $y = (30, 30k, 81k)$ for runs 1, 2, 3 respectively.
- b) For $j \notin A$, set LM, $u_j = 0$ and generate $s_j \in R(0,1)$.

Step 5) Feasibility: Compute $b = s - c^T x^*$

Step 6) Optimality: Compute $a = C u - G x^*$ ●

Appendix C

Van de Panne - Whinston's Dual Tableau Method

Definitions

- Primal member, s_i : slack or primal variable corresponding to constraint i .
- Dual member, u_i : LM or dual variable corresponding to constraint i .
- Basic pair: (s_i, u_i) are both basic.
- Non-basic pair: (s_i, u_i) are both non-basic.
- Standard Tableau: A tableau with no basic pair.
- Non-Standard tableau: A tableau that is not standard.

Dual Tableau Method

Repeat rules 1 and 2 to select the pivot and

- a) terminate with optimal solution if rule 1 cannot be applied.
- b) terminate with no feasible solution if rule 2 cannot be applied.

Rule 1 (Pick an entering variable)

- a) *Standard Tableau*
Dual variable corresponding to the most negative primal variable.
- b) *Non-Standard Tableau*
Primal member of the non-basic pair.

Rule 2 (choice of leaving variable)

Choose the basic variable to leave the basis according to the least nonnegative value of the ratio.

$$\min_{j \in \hat{J}} \left(d_j / w_j \right)$$

where d_j is the value of the basic variable,

w_j is the value corresponding to the entering variable column for that variable and \hat{J} is the set as described below.

a) *Standard Tableau*

\hat{J} is the set of basic dual variables and the basic primal variable corresponding to the entering dual variable.

b) *Non-Standard Tableau*

\hat{J} is the set of basic dual variables excluding the dual variable of the basic pair, and the primal member of the basic pair●

Note: Start with a standard tableau which is dual feasible (i.e. the dual variables are non-negative).

Appendix D

Tables

Problem	NLP characteristics			QP characteristics					average cardinality of basis at QP Soln.
	var.	cons.	eqns.	var.	Fletcher QP			dual QP	
					cons.	low bd	hi bd	cons.	
Powell	5	3	3	6	3	6	6	5	4.0
POP	3	7	0	4	7	4	4	9	2.2
Triangle	7	9	0	8	9	8	8	11	6.1
Colville 1	5	15	0	6	15	6	6	17	6.0
Colville 2	15	20	0	16	20	16	16	22	14.0
Colville 3	5	16	0	6	16	6	6	18	6.0

Table 6.1. Problem characteristics.

Problem	Total of all iterations.												Ratio of Operations**			
	operations			sqrts.	steps			adds			drops			to Dual		
	FP	FQ*	D	D	FP	FQ	D	FP	FQ	D	FP	FQ	D	FP	FQ	FP+FQ
Powell	2353	9237	4830	144	2	12	24	2	0	24	2	12	0	.38	1.47	1.85
POP	2228	3509	1384	49	15	11	13	15	0	13	15	11	0	1.19	1.87	3.06
Triangle	21394	70474	24236	440	33	40	83	33	6	81	33	31	2	.75	2.46	3.21
Colville 1	5896	12077	6060	119	14	7	32	14	0	30	14	4	4	.81	1.66	2.47
Colville 2	196142	665394	237983	2056	59	46	239	59	4	224	59	36	15	.76	2.57	3.33
Colville 3	4147	8012	4711	85	8	0	27	8	0	26	8	0	2	.75	1.44	2.19
Average														.77	1.91	2.69

Table 6.2. Numerical Results.

FP - feasible point routine for Fletcher's QP (phase 1 only)

FQ - Fletcher's QP algorithm (phase 2 only)

D - Dual projection algorithm

Note: See differences in implementation on page 43
Operations do not include square root calculations.

FP, FQ do not require square root calculations.

* includes effort for $\begin{pmatrix} G & N \\ N^T & 0 \end{pmatrix}^{-1}$ at the solution of each QPP, i.e. final reinversion

** includes square root calculation for the dual (10 operations/sqrt.)

Problem type	variables	constraints	constraints in final basis	conditon of G
1	9	9	1	well
2	9	9	1	ill
3	9	9	3	well
4	9	9	3	ill
5	9	27	3	well
6	9	27	3	ill
7	9	27	9	well
8	9	27	9	ill
9	27	27	3	well
10	27	27	3	ill
11	27	27	9	well
12	27	27	9	ill
13	27	81	9	well
14	27	81	9	ill
15	27	81	27	well
16	27	81	27	ill
17	81	81	9	well
18	81	81	9	ill
19	81	81	27	well
20	81	81	27	ill
21	81	243	27	well
22	81	243	27	ill
23	81	243	81	well
24	81	243	81	ill

Table 8.2 Problem Types

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1	D.P.1
1	19	5	37	35	12	11	10	10	5
2	21	5	41	41	15	16	12	12	5
3	24	15	22	22	11	11	9	8	15
4	24	17	33	33	17	17	16	17	17
5	47	15	50	45	22	23	13	13	15
6	44	17	74	79	21	20	18	18	17
7	55	60	44	33	39	37	38	37	64
8	50	54	45	35	46	42	47	41	54
9	73	15	103	101	30	31	16	16	15
10	67	15	95	97	31	34	20	21	15
11	67	45	75	75	30	30	28	28	45
12	74	45	114	106	44	37	40	36	45
13	147	47	186	165	43	43	27	28	47
14	191	47	197	211	50	50	46	48	47
15	172	141	155	108	107	121	115	117	141
16	211	144	140	115	128	109	127	122	144
Totals	1286	687	1411	1301	646	632	582	572	691

Table 8.3(a)* Number of steps run 1.

* See notes on page 79

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1	D.P.1
1	19	5	32	30	24	24	24	24	5
2	21	5	36	36	30	30	28	28	5
3	24	15	17	17	15	15	15	15	15
4	24	17	29	29	29	31	29	31	17
5	52	15	61	53	45	43	43	41	15
6	49	17	72	78	56	54	50	50	17
7	65	61	86	64	78	74	76	74	65
8	56	55	83	65	89	81	91	81	55
9	73	15	98	96	86	82	76	76	15
10	67	15	90	92	94	94	78	78	15
11	67	45	70	70	64	64	68	68	45
12	74	45	109	101	103	91	99	91	45
13	164	47	215	193	145	139	127	129	47
14	247	47	224	246	178	174	166	172	47
15	209	141	292	212	214	242	230	234	141
16	287	144	202	168	256	218	254	244	144
Totals	1498	689	1716	1550	1506	1466	1454	1436	693

Table 8.3(b)* Number of basis changes, run 1.

* See notes on page 79

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1	D.P.1
1	10702	5975	24768	24444	12369	11979	11990	11951	5490
2	11510	5975	27229	28239	14589	15549	13194	13257	5490
3	12365	11369	15276	15814	9835	10263	9338	8999	13075
4	12155	12173	23081	23805	16163	17028	16241	17275	15173
5	28414	14615	56518	54170	31176	33816	26529	26808	16315
6	27487	15595	74385	83368	31403	31567	29878	30260	19231
7	32700	36414	63647	52333	49423	49190	50816	50269	62935
8	29550	34363	63500	55132	57757	55861	62564	57132	51832
9	282262	114658	508384	529042	236616	243620	221812	224109	105235
10	266428	114665	482830	513580	251333	273137	227738	235871	105235
11	267196	237742	386032	400781	228353	234753	244528	247748	280460
12	283173	237749	576078	580236	344399	310554	395827	365435	280460
13	698267	317597	1537625	1558327	547619	609889	719114	738303	376857
14	932444	318186	1640739	2021097	659355	723717	1019411	1079672	377432
15	828776	714197	1684396	1420029	1086798	1277296	1359581	1379179	992003
16	1035779	722888	1291100	1252035	1290434	1176664	1574555	1586534	1031804

Table 8.3(c)* Number of operations, run 1

* See notes on page 79

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1	D.P.1
1	1.79	1.00	4.15	4.09	2.07	2.00	2.01	2.00	0.92
2	1.93	1.00	4.56	4.73	2.44	2.60	2.20	2.22	0.92
3	1.09	1.00	1.34	1.39	0.87	0.90	0.82	0.79	1.15
4	1.00	1.00	1.90	1.96	1.33	1.40	1.33	1.42	1.25
5	1.94	1.00	3.87	3.71	2.13	2.31	1.82	1.83	1.12
6	1.76	1.00	4.77	5.35	2.01	2.02	1.92	1.94	1.23
7	0.90	1.00	1.75	1.44	1.36	1.35	1.40	1.38	1.73
8	0.86	1.00	1.85	1.60	1.68	1.63	1.82	1.66	1.51
9	2.46	1.00	4.43	4.61	2.06	2.12	1.93	1.95	0.92
10	2.32	1.00	4.21	4.48	2.19	2.38	1.99	2.06	0.92
11	1.12	1.00	1.62	1.69	0.96	0.99	1.03	1.04	1.18
12	1.19	1.00	2.42	2.44	1.45	1.31	1.66	1.54	1.18
13	2.20	1.00	4.84	4.91	1.72	1.92	2.26	2.32	1.19
14	2.93	1.00	5.16	6.35	2.07	2.27	3.20	3.39	1.19
15	1.16	1.00	2.36	1.99	1.52	1.79	1.90	1.93	1.39
16	1.43	1.00	1.79	1.73	1.79	1.63	2.18	2.19	1.43
Aver.	1.63	1.00	3.19	3.28	1.73	1.79	1.84	1.86	1.20

Table 8.3(d)* Ratio of number of operations to dual, run 1

* See notes on page 79

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1	D.P.1
1	19	5	37	37	13	13	11	11	5
2	21	5	47	45	19	19	13	13	5
3	24	15	22	22	11	11	9	8	15
4	24	19	37	37	21	21	19	20	19
5	47	17	46	45	22	23	13	13	17
6	44	19	70	64	24	25	23	23	19
7	55	84	44	32	39	38	36	36	88
8	50	69	44	38	43	42	52	44	71
9	73	15	111	111	32	33	20	20	15
10	67	17	103	99	28	26	17	18	17
11	67	45	85	87	41	41	34	34	45
12	74	47	112	110	37	37	36	36	47
13	147	73	205	170	63	55	46	39	73
14	191	73	204	179	50	53	44	46	73
15	172	350	144	118	133	127	142	146	380
16	211	191	134	92	102	104	117	122	195
Totals	1286	1094	1445	1286	678	668	634	631	1084

Table 8.4(a)* Number of steps, run 2

* See notes on page 79

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1	D.P.1
1	19	5	32	32	28	28	26	26	5
2	21	5	42	40	38	36	32	32	5
3	24	15	17	17	15	15	15	15	15
4	24	19	33	33	35	37	35	37	19
5	52	17	57	53	45	43	43	41	17
6	49	19	84	78	60	64	60	60	19
7	65	113	88	64	78	76	72	72	117
8	56	77	87	75	85	83	103	87	79
9	73	15	106	106	88	86	82	82	15
10	67	17	98	94	78	78	74	74	17
11	67	45	80	82	82	82	80	80	45
12	74	47	107	105	83	83	91	91	47
13	164	73	251	211	189	167	165	151	73
14	247	73	256	222	164	160	164	168	73
15	209	397	288	236	266	254	284	292	427
16	287	191	256	180	204	208	234	244	195
Totals	1498	1178	1882	1628	1538	1500	1560	1552	1218

Table 8.4(b)* Number of basis changes, run 2.

* See notes on page 79

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1	D.P.1
1	10702	5975	24510	25526	13594	14047	12646	12964	5490
2	11510	5975	30483	30402	18270	18341	14573	14805	5490
3	12365	11369	15274	15742	10079	10420	9587	9156	13075
4	12155	12923	25215	26055	19431	20735	19231	20540	16478
5	28414	15227	52793	53805	31435	34051	26453	27032	19174
6	27487	16483	77016	74989	35385	38551	37124	37769	22080
7	32700	54823	64544	52524	50062	50452	48658	49611	84104
8	29550	42137	64600	60359	54194	56896	69842	60894	69728
9	282262	114658	544452	579394	251260	261776	245874	249555	105235
10	266428	120922	516523	519380	218950	219228	206170	215599	122147
11	267196	237742	429467	456927	301953	308302	282255	284612	280460
12	283173	243963	555094	585850	295400	305927	331246	337180	297334
13	698267	403156	1747620	1633473	763232	715243	784594	722874	559393
14	932444	413383	1761174	1754503	647635	730399	894491	942016	644029
15	828776	1442354	1610623	1513305	1323407	1332172	1572056	1599277	3147916
16	1035779	874863	1455188	1195073	1028918	1114449	1365317	1449225	1563452

Table 8.4(c)* Number of operations, run 2.

* See notes on page 79

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1	D.P.1
1	1.79	1.00	4.10	4.27	2.28	2.35	2.12	2.17	0.92
2	1.93	1.00	5.10	5.09	3.06	3.07	2.44	2.48	0.92
3	1.09	1.00	1.34	1.38	0.89	0.92	0.84	0.81	1.15
4	0.94	1.00	1.95	2.02	1.50	1.60	1.49	1.59	1.28
5	1.83	1.00	3.40	3.47	2.02	2.19	1.70	1.74	1.23
6	1.67	1.00	4.67	4.55	2.15	2.34	2.25	2.29	1.34
7	0.60	1.00	1.18	0.96	0.91	0.92	0.89	0.90	1.53
8	0.70	1.00	1.53	1.43	1.29	1.35	1.66	1.45	1.65
9	2.46	1.00	4.75	5.05	2.19	2.28	2.14	2.18	0.92
10	2.20	1.00	4.27	4.30	1.81	1.81	1.71	1.78	1.01
11	1.12	1.00	1.81	1.92	1.27	1.30	1.19	1.20	1.18
12	1.16	1.00	2.28	2.40	1.21	1.25	1.36	1.38	1.22
13	1.73	1.00	4.33	4.05	1.89	1.77	1.95	1.79	1.39
14	2.26	1.00	4.26	4.24	1.57	1.77	2.16	2.28	1.56
15	0.57	1.00	1.12	1.05	0.92	0.92	1.09	1.11	2.18
16	1.18	1.00	1.66	1.37	1.18	1.27	1.56	1.66	1.79
Aver.	1.45	1.00	2.98	2.97	1.63	1.70	1.66	1.68	1.33

Table 8.4(d)* Ratio of number of operations to dual, run 2.

* See notes on page 79

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1
17	44	9	72	70	13	13	8	7
18	38	9	62	60	13	12	10	10
19	32	29	68	64	40	42	42	42
20	38	29	72	72	26	33	32	31
21	105	103	189	161	60	58	75	68
22	93	35	155	128	31	36	48	35
23	107	252	135	91	129	119	128	131
24	118	113	133	90	105	111	120	111
Totals	575	579	886	736	417	424	463	435

Table 8.5(a)* Number of steps, run 3.

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1
17	44	9	71	69	53	49	49	47
18	38	9	61	59	49	49	47	47
19	32	29	67	63	83	85	87	87
20	38	29	71	71	61	75	73	71
21	129	103	278	222	168	166	202	188
22	105	35	202	164	114	124	148	122
23	133	267	270	182	258	238	256	262
24	155	113	264	180	210	222	240	222
Totals	674	594	1284	1010	996	1008	1102	1046

Table 8.5(b)* Number of basis changes, run 3

* See notes on page 79

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1
17	1436003	550164	2991116	3230184	944420	959405	1204284	1065050
18	1295627	550163	2645193	2753002	922019	947848	1323632	1350157
19	1145963	1225579	2900305	2980464	2500936	2618295	2795222	2827916
20	1295627	1227923	3048899	3466566	1717564	2188260	2379792	2347479
21	4289510	3638952	14982501	16152724	5984002	6367082	10624157	10538954
22	3801085	1799584	11593327	12852311	3545462	4558379	8783870	8276938
23	4447039	8016384	12897024	12534633	11008810	10722045	14526959	14366066
24	4912031	4421230	12800635	12638596	9015265	9937241	13307823	12832151

Table 8.5(c) * Number of operations, run 3.

Problem type	F	D	P.1	P.1.1	P.2	P.2.1	P.2'	P.2'.1
17	2.61	1.00	5.44	5.87	1.72	1.74	2.19	1.94
18	2.36	1.00	4.81	5.00	1.68	1.72	2.41	2.45
19	0.94	1.00	2.37	2.43	2.04	2.14	2.28	2.31
20	1.06	1.00	2.48	2.82	1.40	1.78	1.94	1.91
21	1.18	1.00	4.12	4.44	1.64	1.75	2.92	2.90
22	2.11	1.00	6.44	7.14	1.97	2.53	4.88	4.60
23	0.55	1.00	1.61	1.56	1.37	1.34	1.81	1.79
24	1.11	1.00	2.90	2.86	2.04	2.25	3.01	2.90
Ave.	1.48	1.00	3.76	4.01	1.73	1.90	2.67	2.59

Table 8.5(d) * Ratio of Number of operations to dual, run 3.

* See notes on page 79

Notes on Tables 8.3 through 8.5

The following remarks apply to the tables 8.3, 8.4 and 8.5.

The numbers under six primal method columns assume that the feasible point is readily available. Thus for comparison purposes with column D, one should add the appropriate columns labelled by F.

The number of operations include square roots (10 operations).

In tables 8.3(a,b,c), 8.4(a,b,c) the entries for given algorithm and given problem type are summed over the five replicates for that run. In tables 8.3(d), 8.4(d) the averages are column averages.

N_d = number of drops	number of problems with N_d drops		
	run 1	run 2	run 3
0	62	46	2
1	12	8	2
2	5	3	0
3 and up	1	23	4
total	80	80	8
Total drops	25	244	153

Table 8.6 Frequency of drops for the dual.

Problem Type	run 1. LM = (0,30) repetition					run 2. LM = (0,30k) repetition					run 3. LM = (0,81k)
	1	2	3	4	5	1	2	3	4	5	
1	0	0	0	0	0	0	0	0	0	0	
2	0	0	0	0	0	0	0	0	0	0	
3	0	0	0	0	0	0	0	0	0	0	
4	0	0	0	0	1	0	0	0	0	2	
5	0	0	0	0	0	0	1	0	0	0	
6	0	0	0	0	1	0	0	1	0	1	
7	3	2	1	1	1	12	7	5	6	4	
8	1	2	1	0	1	4	3	3	2	4	
9	0	0	0	0	0	0	0	0	0	0	
10	0	0	0	0	0	0	1	0	0	0	
11	0	0	0	0	0	0	0	0	0	0	
12	0	0	0	0	0	0	1	0	0	0	
13	0	0	0	1	0	0	0	7	6	1	
14	0	0	1	0	0	1	5	5	2	1	
15	0	1	0	2	0	24	27	20	35	25	
16	1	0	2	2	0	4	7	5	3	9	
17											0
18											0
19											1
20											1
21											38
22											4
23											93
24											16

Table 8.7. Number of drops in dual

Problem Type	initial basis. repetition					final basis.	initial \cap final basis repetition				
	1	2	3	4	5		1	2	3	4	5
1	5	2	3	3	6	1	0	1	0	0	0
2	3	6	3	4	5	1	0	1	0	0	0
3	6	7	4	5	2	3	3	3	3	1	2
4	4	9	3	3	5	3	1	3	0	1	2
5	9	7	8	9	9	3	2	1	2	1	2
6	8	8	6	7	9	3	2	2	0	1	1
7	9	9	9	9	9	9	4	6	2	6	2
8	9	9	8	9	9	9	2	3	4	6	2
9	16	17	16	9	15	3	1	1	3	1	3
10	13	9	13	17	15	3	1	1	2	3	1
11	14	14	16	10	13	9	7	5	6	5	4
12	11	11	13	20	19	9	3	6	3	6	6
13	24	27	27	27	25	9	5	6	7	4	8
14	27	27	27	27	27	9	6	1	5	9	7
15	27	27	27	27	27	27	15	18	14	13	12
16	27	27	27	27	27	27	17	11	13	19	15
17	44					9	6				
18	38					9	4				
19	32					27	10				
20	38					27	11				
21	81					27	14				
22	81					27	13				
23	81					81	40				
24	81					81	38				

Table 8.8 Intersection of initial and final bases for primal methods.

Method	Run 1	Run 2	Run 3
F	1.16	1.16	1.17
D	1.00	1.08	1.03
D.P.1	1.00	1.12	not run
P.1	1.21	1.30	1.45
P.1.1	1.19	1.27	1.37
P.2	2.33	2.27	2.39
P.2.1	2.32	2.25	2.38
P.2'	2.50	2.46	2.38
P.2'.1	2.51	2.46	2.40

Table 8.9. Ratio of basis changes to step

Method	Run 2 - Run 1 steps	Run 2 - Run 1 basis changes
D	407	489
D.P.1	393	525
P.1	34	166
P.1.1	-15	78
P.2	32	32
P.2.1	36	34
P.2'	52	104
P.2'.1	59	116

Table 8.10. Effect of linearity of function on algorithms.

Appendix E
Program Listing

Following are the names and functions of various routines used in the experimental run.

<i>NAME</i>	-	<i>FUNCTION</i>
MAIN	-	It is the main program, sets up I/O and summary.
GENER	-	Generate a random problem with known solution.
DECOMP	-	Decompose G into LL^T
INVL	-	Invert L
SOLVL	-	Solve linear system $Ly = w$
SOLVLT	-	Solve linear system $L^T y = w$
SOLVR	-	Solve linear system $Ry = w$
SOLVRT	-	Solve linear system $R^T y = w$
UNCONM	-	Find unconstrained minimum
RAND	-	Generate random number
GRA	-	Calculate gradient
PRTMAT	-	Print matrix
PROJ	-	Project vector onto the current manifold
RBAR	-	Calculate $Q^T L^{-1} y$
MXMIN	-	Find max. or min. of a vector
ADD	-	Add a constraint to the basis
GIVENM	-	Calculate coefficients of Givens Matrix
DROP	-	Remove a constraint from basis
INTLQL	-	Set $Q^T L^{-1} = L^{-1}$
GVNUPV	-	Multiply a vector of two elements with Givens Matrix
LAMDA	-	Calculate Slack vector

XCOR	-	Perform X-correction
REIN	-	Perform reinversion
FEAS	-	Feasible point routine
STPLEN	-	Calculate step length for primal methods
FLET	-	Primal algorithm P.1, P.1.1
NSGNST	-	Calculate a diagonal element of $R^{-1}R^{-T}$
STPEDG	-	Calculate steepest edge
DUAL	-	Dual Projection method D
DSTPLN	-	Dual step length
UPRIN	-	Update logging print array
PRIMAL	-	Primal routine initializer
ACTCHK	-	Check active constraints on manifold
RENCHK	-	Check if reinversion necessary
PRDUAL	-	Primal-dual initializer
GOLD	-	Primal algorithm P.2, P.2.1, P.2', P.2'.1
CHECK	-	Checks with known solution
SINNER	-	Calculate inner product of two vectors (Assembly routine)

```

C      THIS IS THE MAIN PROGRAM
0001      IMPLICIT REAL*8(A-H,O-Z)
COMMON C1
0002      COMMON /C01/PROJV,LENY,QLTZ
0003      REAL * 8 PROJV(30),LENY,QLTZ(30)
COMMON C1 END
COMMON C2
0004      COMMON /C02/LG
0005      REAL*8 LG(30,31),L(30,30),G(30,30)
0006      EQUIVALENCE(LG(1,1),L(1,1))
0007      EQUIVALENCE(LG(1,2),G(1,1))
COMMON C2 END
COMMON C3
0008      COMMON /C03/QTLINV
0009      REAL*8 QTLINV(30,30)
COMMON C3 END
COMMON C4
0010      COMMON /C04/Q1R
0011      REAL*8 Q1R(30,32),Q1(30,30),R(30,30)
0012      EQUIVALENCE(Q1R(1,1),Q1(1,1))
0013      EQUIVALENCE(Q1R(1,3),R(1,1))
COMMON C4 END
COMMON C5
0014      COMMON /C05/C,B,XLAMDA,ICONOK
0015      REAL*8 C(30,90),B(90),XLAMDA(90)
0016      INTEGER*2 ICONOK(90)
COMMON C5 END
COMMON C6
0017      COMMON /C06/FF,FNQ,FQ
0018      INTEGER FNQ(30),FQ
0019      REAL*8 FF
COMMON C6 END
COMMON C7
0020      COMMON /C07/ FQTLIN
0021      REAL*8 FQTLIN(30,30)
COMMON C7 END
COMMON C8
0022      COMMON /C08/ FXLINR
0023      REAL*8 FXLINR(30,32),FX(30),LINV(30,30),FR(30,30)
0024      EQUIVALENCE(FXLINR(1,1),FX(1))
0025      EQUIVALENCE(FXLINR(1,2),LINV(1,1))
0026      EQUIVALENCE(FXLINR(1,3),FR(1,1))
COMMON C8 END
COMMON C9
0027      COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,F5OKFG,TYPG
0028      1, ISEED,ISOPT,IAOPT,IUOPT
      INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,F5OKFG,TYPG
COMMON C9 END
COMMON C10
0029      COMMON /C10/UNCONX,UNCONF
0030      REAL*8 UNCONX(30),UNCONF
COMMON C10 END
COMMON C11
0031      COMMON /C11/X
0032      REAL*8 X(30)
COMMON C11 END

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COMMON 12
0033     COMMON /C12/NQ
0034     INTEGER NQ(30)
COMMON 12 END
COMMON 13
0035     COMMON /C13/LM,IM
0036     REAL*8 LM(30),IM(30)
COMMON 13 END
COMMON 14
0037     COMMON /C14/F,GRAD
0038     REAL*8 F,GRAD(30)
COMMON 14 END
COMMON 15
0039     COMMON /C15/TOLDCM, TOLDEP, TOLINV, TOLPR, TOLAM, TOLMUL
COMMON 15 END
COMMON 16
0040     COMMON /C16/HIX,LOX, HIXLOX, HIG,LOG, HIGLOG, HILM, HIC, LOC, HICLOC, HIB
0041     REAL*8 HIX,LOX, HIXLOX, HIG,LOG, HIGLOG, HILM, HIC, LOC, HICLOC, HIB
COMMON 16 END
COMMON 17
0042     COMMON /C17/IUNIT5, IUNIT6, IUNIT8
COMMON 17 END
COMMON 18
0043     COMMON /C18/SOLNX, SOLNF, SOLNQ
0044     REAL*8 SOLNX(30), SOLNF
0045     INTEGER SOLNQ
COMMON 18 END
COMMON 19
0046     COMMON /C19/IPRIN, IPTR, IPMAX, NBRK
0047     INTEGER*2 IPRIN(5000), IPTR, IPMAX, NBRK(15)
COMMON 19 END
COMMON 20
0048     COMMON /C20/A
0049     REAL*8 A(30)
COMMON 20 END
COMMON 21
0050     COMMON /C21/IAG, NSTEPS, NADDS, NDROPS, NOPS, NSORT, NREIN, NXCDR
0051     1, N1STEP, NPSTEP, ICC, NTIME, HXSTPS, HXREIN
COMMON 21 END
COMMON 22
0051     COMMON /C22/ TITLE, NPAGE, NLIN, MXLINE, PRTOPT
0052     REAL*8 TITLE(15)
0053     INTEGER PRTOPT
COMMON 22 END
COMMON 23
0054     COMMON /C23/GC, GH, GP, GB
0055     REAL*8 GC(30), GH(30), GP(30), GB(30)
COMMON 23 END
0056     INTEGER IF11(12)
0057     EQUIVALENCE(IF11( 1), IAG )
0058     EQUIVALENCE(IF11( 2), NSTEPS)
0059     EQUIVALENCE(IF11( 3), NADDS )
0060     EQUIVALENCE(IF11( 4), NDROPS)
0061     EQUIVALENCE(IF11( 5), NOPS )
0062     EQUIVALENCE(IF11( 6), NSORT )
0063     EQUIVALENCE(IF11( 7), NREIN )

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0064      EQUIVALENCE(IF11( 8),NXCOR )
0065      EQUIVALENCE(IF11( 9),NFSTEP)
0066      EQUIVALENCE(IF11(10),NPSTEP)
0067      EQUIVALENCE(IF11(11),ICC )
0068      EQUIVALENCE(IF11(12),NTIME)
0069      INTEGER * 2 IJPRIN(14)
0070      REAL * 8 BLANK, TYPRIN(12), APRIN, TYPALG(14), APROB
0071      DATA APROB/'PROB DEF'/
0072      REAL * 8 DAT
0073      DATA BLANK, TYPRIN/'      ', 'ALG. # ', 'STEPS ',
1' * ADDS ', 'DROPS ', 'MLT/100', 'SQRT ', 'REINV ',
1' * XCOR ', 'F STPS', 'P STPS', 'CJMPCODE', 'TIM*10|3'/
0074      DATA TYPALG/'FEAS PT ', 'DUAL ', 'FLET- 1 ', 'FLET-STP', 'GOLD- 1 ',
1' * GOLD-STP', 'G-1+VIDL', 'GSTP+VID', 'PD-F-1 ', 'PD-F-STP',
1' * PD-G-1 ', 'PD-G-STP', 'PDG1VIDL', 'PDGSTPVI' /
0075      INTEGER ITOTS(16,14,5)
0076      DATA ITOTS / 1120 * 0/
0077      REAL * 8 TYPTOT(6)
0078      REAL * 8 FTOTS(17,14)
0079      EQUIVALENCE(FTOTS, IPRIN)
0080      DATA TYPTCT/'STEPS ', 'BAS.CHG', 'DPS.', 'TIME ', 'PRDBS ',
1' * RATIO D' /
C      THIS IS THE END OF DECLARATIONS
C
C
C
CCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCC
C
C      DECLARATIONS FOR PROBLEM RUN ETC.
0081      INTEGER * 2 IPMRUN(80), NPRDB, IAGRUN(14), PRDEF(16,5), ICPT(14,3)
0082      INTEGER * 4 PSEED(60)
C      NPRDB IS NUMBER OF PROBLEMS
C      IPMRUN IS PROBLEM NUMBERS TO BE RUN (NPRDB IS ITS INDEX)
C      IAGRUN IS ALGORITHMS TO BE RUN
C      -1 - FEASIBLE PT. ROUTINE
C      -2 - DUAL ROUTINE
C      -3 - PRIMAL FLETCHER
C      -4 - PRIMAL FLETCHER STEEPEST EDGE
C      -5 - PRIMAL GOLDFARB
C      -6 - PRIMAL GOLDFARB STEEPEST EDGE
C      -7 - PRIMAL GOLDFARB WITHOUT VIOLATING ALREADY DROPPED BY STEP
C      -8 - PRIMAL GOLDFARB (COMBO OF 5 & 7)
C      -9 - DUAL PRIMAL(263)
C      -10 - " " (264)
C      -11 - " " (265)
C      -12 - " " (266)
C      -13 - " " (267)
C      -14 - " " (268)
C
C      PRDEF IS PROBLEM DEFINITION OF ALL 16 PROBLEMS (ROWS)
C      COL. 1 IS M - NUMBER OF VARIABLES
C      2 IS K - NUMBER OF CONSTRAINTS
C      3 IS TYPG - TYPE OF G (IN OBJ. FN. 1 - WELL, 2 - ILL)
C      4 IS SOLNQ - NUMBER OF CONSTRAINTS IN FINAL BASIS(FIRST)
C      5 IS IMAGE (IN BINARY REP)
C

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C
C NOW THE MAIN PROGRAM STARTS
C READ THE TITLE OF RUN
0083      ISIZ = 30
0084      IUNIT5 = 5
0085      IUNIT6 = 6
0086      IUNIT8 = 8
0087      IPMAX = 5000
0088      CN = 1
0089      OFF = 0
0090      CALL DATE(DAT)
0091      WRITE(IUNIT6,4011)DAT
0092      4011  FORMAT('1',//,5X,'RUN DATE = ',A8)
0093      WRITE(IUNIT6,4012)DAT
0094      4012  FORMAT(5X,'RUN DATE = ',A8)
C
0095      WRITE(IUNIT6,18)
0096      18    FORMAT(////////)
0097      17    READ(IUNIT5,19)IOMIT,(TITLE(J),J=1,20)
0098      19    FORMAT(11,19A4,A3)
0099      WRITE(IUNIT6,16)IOMIT,(TITLE(J),J=1,20)
0100      WRITE(IUNIT8,19)IOMIT,(TITLE(J),J=1,20)
0101      16    FORMAT(' ',5X,11,19A4,A3)
0102      IF(IOMIT.EQ.C)GOTO 17
0103      NLINE = 0
0104      NPAGE = 1
0105      READ(IUNIT5,50)TITLE
0106      50    FORMAT(1CA8/5A8)
C
C READ THE NUMBERS OF THE PROBLEMS TO BE RUN
C      NRPT IS THE NO. OF REPETITION OF A PROBLEM (5)
C      NDALG IS NUMBER OF DIFFERENT ALGS = 14
C      NPROB IS NUMBER OF DIFFERENT TYPES OF PROBLEMS (16)
C      NTPROB IS TOTAL NO OF PROBLEMS
0107      NDALG = 14
0108      NRPT = 5
0109      WRITE(IUNIT6,31) TITLE,NPAGE
0110      WRITE(IUNIT8,51)TITLE
0111      51    FORMAT(15A8)
0112      31    FORMAT('1',1X,15A8, 'PAGE=',14,//,5X,' PROBLEMS TO BE RUN
0113      NPROB = 16
0114      NPRJB = 0
0115      DO 60 I=1,1000
0116      READ(IUNIT5,65)IOMIT,(IPRIN(J),J=1,15)
0117      65    FORMAT(16(I2,1X))
0118      IF(IOMIT.LT.0) GOTO 60
0119      WRITE(IUNIT6,62)(IPRIN(J),J=1,15)
0120      62    FORMAT(' ',15I4)
0121      DO 70 J=1,15
0122      IF(IPRIN(J))90,70,08
0123      68    NPROB = NPROB + 1
0124      IPHRUN(NPROB) = IPRIN(J)
0125      70    CONTINUE
0126      60    CONTINUE
C
C READ THE ALGORITHMS TO BE RUN

```

```

C
0127      WRITE(IUNIT6,81)
0126      81  FORMAT(' ',5X,'ALGS TO BE RUNG START, ADD,DROP OPTIONS ARE ')
0129      90  DO 80 I = 1,NDALG
0130      75  FORMAT(4(I2,1X))
0131      READ(IUNIT5,75)IAGRUN(I),(IOPT(I,J),J=1,3)
0132      80  WRITE(IUNIT6,82)I,TYPALG(I),IAGRUN(I),(IOPT(I,J),J=1,3)
0133      82  FORMAT(' ',5X,15,3X,A0,4I5)
C      IOPT IS THE VARIOUS OPTIONS FOR THE ALGORITHMS
C      COL. 1 IS FOR STARTING PT FOR FR EAS PT.
C          1 - UNCONSTRAINED MIN.
C          2 - RANDOM STARTING PT.
C      COL. 2 IS FOR ADDING CONSTRAINT(IF APPLICABLE)
C          1 - ADD MOST VIOLATED CONS.
C          2 - ADD LEAST VIOLATED CONS.
C          3 - ADD RANDOM VIOLATED CONS.
C      COL. 3 IS FOR DRIPPING CONSTRAINT(IF APPLIC.)
C          1 - DRDP MOST -VE LM OR IM.
C          2 - DROP LEAST -VE LM OR IM.
C          3 - DROP RANDOM -VE LM OR IM(INFEASIBILITY MULT.)
C
C      READ THE TOLERANCES
0134      READ(IUNIT5,85)TOLDCH,TOLDEP,TOLINV,TOLPR,TOLAM,TOLMUL
0135      WRITE(IUNIT6,83)TOLDCH,TOLDEP,TOLINV,TOLPR,TOLAM,TOLMUL
0136      83  FORMAT(' ',5X,'TOLERANCES: DCH,LDEP,INV,PROJ,LAMDA,MULT',/
0137      85  FORMAT(6F10.7)
C
C      TOLDEP - LINEAR DEPENDENCE TOLERANCE
C      TOLPR  - PROJECTION TOLERANCE
C      TOLAM  - LAMDA(VIOLATION) TOLERANCE
C      TOLMUL - MULTIPLIER( LM & IM) TOLERANCE
C
0138      READ(IUNIT5,95)MXSTPS,MXREIN
0139      WRITE(IUNIT6,91)MXSTPS,MXREIN
0140      91  FORMAT(' ',5X,'MAX STEP,REIN ',2I5)
0141      95  FORMAT(2(I4,1X))
C      MXSTPS IS MAX NUMBER OF STEPS IN AN ALGORITHM
C      MXREIN IS MAX NUMBER OF REINVERSIONS IN AN ALGORITHM
C
0142      READ(IUNIT5,75)PRTOPT
0143      WRITE(IUNIT6,92)PRTOPT
0144      92  FORMAT(' ',5X,'PRINT OPTION :0 - LESS,1 - MORE ',I5)
C
C      PRTOPT IS PRINT OPTION
C      0 - LEAST PRINTING
C      1 - MORE
C
C      NOW READ THE VARIOUS PROBLEM DEFINITIONS
0145      WRITE(IUNIT6,103)
0146      103  FORMAT(' ',5X,'PROB. DEF.: M,K,G(1-WELL,2-ILL),FB,BINARY')
0147      DO 110 I = 1,NPROB
0148      REAL(IUNIT5,105)(PRODEF(I,J),J=1,5)
0149      WRITE(IUNIT6,106)I,(PRODEF(I,J),J=1,5)
0150      106  FORMAT(' ',5X,6I6)

```

```

0151      105      FORMAT(5I5)
0152      110      CONTINUE
0153      NTPROB = NDPROB * NRPT
0154      READ(IUNIT5,111)HIC,LOC,HIG,LOG,HIX,LOX,HIB,HILM
0155      111      FORMAT(4F10.5/4F10.5)
0156      HICLOC = HIC - LOC
0157      HIGLOG = HIG - LOG
0158      HIXLOX = HIX - LOX
0159      WRITE(IUNIT6,112)HIC,LOC,HIG,LOG,HIX,LOX,HIB,HILM
0160      112      1  FORMAT(' ',2X,'HIC,LOC',2F10.5,5X,'HIG,LOG',2F10.5,
1 5X,'HIX,LOX',2F10.5,5X,'HIB,HILM',2F10.5)
0161      DO 120 I=1,NTPROB
0162      115      FORMAT(110)
0163      READ(IUNIT5,115)PSEED(I)
0164      120      CONTINUE
C
C
C      RUN ALL THE PROBLEMS
C
C
0165      WRITE(IUNIT6,1017)TYP,IN
0166      1017     FORMAT(/'PROB #',5X,12(A8,1X))
0167      DO 1000 I = 1,NPROB
0168      IPRCB = IPHRUN(I)
0169      IPRTYP = IPRCB - IPROB /NDPROB *NDPROB
0170      IF(IPRTYP.EQ.0)IPRTYP = NDPROB
0171      M = PRDEF(IPRTYP,1)
0172      K = PRDEF(IPRTYP,2)
0173      TYPG = PRDEF(IPRTYP,3)
0174      SOLNQ = PRDEF(IPRTYP,4)
0175      KINDPR = PRDEF(IPRTYP,5)
0176      ISEED = PSEED(IPROB)
0177      WRITE(IUNIT6,125)TITLE
0178      125     FORMAT('1',/,2X,15A8)
0179      WRITE(IUNIT6,127)IPROB,IPRTYP,M,K,TYPG,SOLNQ,KINDPR
0180      127     FORMAT(' ',/5X,' PROB. NO.',15,' PROB. TYPE',15,
1'M = ',15,' K = ',15,' G(1 -WELL,2-ILL)',15,
2' FINAL BASIS(FIRST) ',15,' CONS. ',15,' KIND OF PROB',15)
C
C      NOW CALL GENER TO GENERATE THE PROBLEM
C
C
0181      NOPS = 0
0182      NSQRT = 0
0183      CALL GENER (IRC,III,ITIMDI)
0184      NGOPS = NOPS
0185      NGSQRT = NSQRT
0186      IF(IRC.EQ.0) GOTO 140
0187      WRITE(IUNIT6,129)IRC
0188      129     FORMAT(' ',10X,'***** GENERATION OF PROBLEM FAILED - RC =',15)
0189      GOTO 1000
C
C
C      NOW FIND THE UNCONSTRAINED MINIMUM FOR THE PROBLEM
C
C

```

```

      C
0190      140      NOPS = 0
0191              NSQRT = 0
0192              CALL CLOCK(0)
0193              CALL UNCONM
0194              CALL CLOCK(1,TIME)
0195              ITIMU = TIME * 100000
0196              NUOPS = NOPS
0197              NUSQRT = NSQRT

      C
      C
      C
      C      SET FEASOKFLAG TO BE OFF
0198              FSOKFG = OFF

      C
0199              IPTR = 0
0200              DO 900 IJK = 1,NDALG
0201              NBRK(IJK) = IPTR + 1
0202              DO 250 IJKL=1,12
0203              IPTR = IPTR + 1
0204              IF11(IJKL) = 0
0205              IPRIN(IPTR) = 0
0206              IAG = IJK
0207              IF (IAGRUN(IAG).LE.0)GOTO 900

      C
      C      SET THE COUNTERS TO ZERO
      C      COPY THE OPTIONS FOR ALG. TO BE RUN
0208              ISOPT = IOPT(IAG,1)
0209              IAOPT = IOPT(IAG,2)
0210              IDOPT = IOPT(IAG,3)
0211              GOTO(141,142,143),IAOPT
0212              141      IAOPT = 1
0213              GOTO 144
0214              142      IAOPT = -1
0215              GOTO 144
0216              143      IAOPT = 0
0217              GOTO(145,146,147),IDOPT
0218              145      IDOPT = 1
0219              GOTO 148
0220              146      IDOPT = -1
0221              GOTO 148
0222              147      IDOPT = 0
0223              148      CONTINUE
0224              WRITE(IUNIT6,153)TYPALG(IAG)
0225              153      FORMAT(' ',3X,A8)

      C
      C      NOW WE HAVE COPIED THE OPTIONS IN EFFECT
      C
0226              CALL CLOCK(0)
0227              GOTO(1,2,3,3,3,3,3,3,9,9,9,9,9,9),IAG
0228              1      CALL FEAS(IRC,III)
0229              CALL CLOCK(1,TIME)
0230              NTIME = TIME * 100000
0231              NFOPS = NOPS
0232              NFSQRT = NSQRT
0233              NOPS=NOPS+NGOPS

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0234      NSQRT=NSQRT+NGSQRT
0235      INT = 0
0236      IF (C.EQ.0.OR.SOLNQ.EQ.0)GOTO 230
0237      DO 210 I1 = 1,0
0238      IF (NQ(I1).LE.SOLNQ)INT=INT+1
0239      210  CONTINUE
0240      230  WRITE(IUNIT8,220)IPROB,IPRTYP,KINDPR,M,K,SOLNQ,TPG,INT
          1,ITIMDI,NGOPS,NGSQRT,ITIMU,NUOPS,NUSQRT
          1,APR03
0241      220  FORMAT(12,1X,12,3X,14,' M=',13,' K=',13,' F8=',
          112,'(FIRST)', ' G-',13,' INT. ',13,617,A8)
0242      ICC = INT
0243      IF (IRC.EQ.0)FSCKFC = ON
0244      GOTO 850
0245      2   CALL DUAL(IRC)
0246      CALL CLOCK(1,TIME)
0247      NOPS=NOPS+NGOPS+NUOPS
0248      NSQRT=NSQRT+NGSQRT+NUSQRT
0249      NTIME = TIME + 100000
0250      GOTO 800
0251      3   IPAG = IAG - 2
0252      IF (FSCKFC.EQ.ON)GOTO 7
0253      WRITE(IUNIT6,6)IAG
0254      6   FORMAT(' ',2X,'SINCE FEAS PT IS NOT FOUND - ALG. NOT RUN',15)
0255      GOTO 900
0256      7   CONTINUE
0257      CALL PRIMAL(IPAG,IRC,0)
0258      CALL CLOCK(1,TIME)
0259      NTIME = TIME + 100000
0260      GOTO 800
0261      9   IPDAG = IAG - 2
0262      CALL PRDUAL(IPDAG,IRC,1)
0263      CALL CLOCK(1,TIME)
0264      NTIME = TIME + 100000
0265      GOTO 800
0266      800  CONTINUE
0267      CALL CHECK
0268      850  CONTINUE
0269      WRITE(IUNIT8,852)IPROB,IF11,TYPALG(IAG)
0270      852  FORMAT(12,1X,12,318,112,719,8X,A8)
0271      ITOTS(IPRTYP,IAG,1) = ITJTS(IPRTYP,IAG,1) + NSTEPS
0272      ITOTS(IPRTYP,IAG,2) = ITJTS(IPRTYP,IAG,2) + NADDS + NDRDPS
0273      ITOTS(IPRTYP,IAG,3) = ITJTS(IPRTYP,IAG,3) + NOPS + 10*NSQRT
0274      ITOTS(IPRTYP,IAG,4) = ITJTS(IPRTYP,IAG,4) + NTIME
0275      ITJTS(IPRTYP,IAG,5) = ITOTS(IPRTYP,IAG,5) + 1
0276      WRITE(IUNIT6,5371)TYPALG
0277      5371 FORMAT(' ',12(A8,1X))
0278      WRITE(IUNIT6,853)IF11
0279      853  FORMAT(' ',5X,12,318,112,719)
0280      IF11(5) = (IF11(5)+50)/100
0281      NTIME = (NTIME + 50)/100
0282      DO 860 I1 = 1,12
0283      IPRIN(NBRK(IAG)+I1-1) = IF11(I1)
0284      860  CONTINUE
0285      NMAX = 0
0286      WRITE(IUNIT6,903)TYPALG

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0247      903      FORMAT(' ',6X,14(A8,1X)/)
0248      NBRK(NDALG+1)=IPTR*1
0249      DO 910 I1 = 1,NDALG
0290      IF(NBRK(I1+1)-NBRK(I1)).LE.NMAX)GOTO910
0291      NMAX = NBRK(I1+1) - NBRK(I1)
0292      910      CONTINUE
0293      DO 940 II = 1,NMAX
0294      IM1 = II - 1
0295      DO 930 JJ = 1,NDALG
0296      N3B = NBRK(JJ) * IM1
0297      IJPRIN(JJ) = 0
0298      IF(N3B.LT.NBRK(JJ+1))IJPRIN(JJ) = IPRIN(N3B)
0299      930      CONTINUE
0300      APRIN = BLANK
0301      IF(II.LE.12)APRIN = TYPRIN(II)
0302      WRITE(IUNIT6,931)APRIN,IJPRIN
0303      931      FORMAT(' ',A5,1X,14I9)
0304      940      CONTINUE
0305      1000     CONTINUE
0306      NDPR1=NDPROB*1
0307      DO 605 I=1,17
0308      DJ 605 J=1,14
0309      FTOTS(I,J)=0.00
0310      DO 610 I=1,NDALG
0311      SUM=0.00
0312      DO 607 J=1,NDPROB
0313      FTOTS(J,I)=ITOTS(J,I,3)/(0.00+ITOTS(J,2,3))
0314      SUM=SUM+FTOTS(J,I)
0315      607     CONTINUE
0316      FTOTS(NDPR1,I)=SUM/NDPROB
0317      610     CONTINUE
0318      DO 540 KK = 1,6
0319      WRITE(IUNIT6,573)TYPTOT(KK),TYPALG
0320      WRITE(IUNIT8,574)TYPTOT(KK),TYPALG
0321      IF(KK.NE.6)GOTO 535
0322      WRITE(IUNIT6,575)(II,(FTOTS(II,JJ),JJ=1,NDALG),II=1,NDPR1)
0323      WRITE(IUNIT8,576)(II,(FTOTS(II,JJ),JJ=1,NDALG),II=1,NDPR1)
0324      575     FORMAT(17(1X,I2,14F9.4/))
0325      576     FORMAT(17(1X,I2,3X,14F9.4/))
0326      GOTO 540
0327      535     CONTINUE
0328      573     FORMAT(' ',5X,A8/1X,'PRTP',14(1X,A8))
0329      574     FORMAT(A5/14(1X,A3))
0330      WRITE(IUNIT6,571)(II,(ITOTS(II,JJ,KK),JJ=1,NDALG),II=1,NDPROB)
0331      571     FORMAT(16(1X,I2,14I9/))
0332      WRITE(IUNIT8,572)(II,(ITOTS(II,JJ,KK),JJ=1,NDALG),II=1,NDPROB)
0333      572     FORMAT(16(1X,I2,3X,14I9/))
0334      540     CONTINUE
0335      STOP
0336      END

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0001          SUBROUTINE GENER (IRC,111,ITIMDI)
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON C2
0003          COMMON /C02/LG
0004          REAL*8 LG(30,31),L(30,30),G(30,30)
0005          EQUIVALENCE(LG(1,1),L(1,1))
0006          EQUIVALENCE(LG(1,2),G(1,1))
          COMMON C2 END
          COMMON C4
0007          COMMON /C04/Q1R
0008          REAL*8 Q1R(30,32),Q1(30,30),R(30,30)
0009          EQUIVALENCE(Q1R(1,1),Q1(1,1))
0010          EQUIVALENCE(Q1R(1,3),R(1,1))
          COMMON C4 END
          COMMON C5
0011          COMMON /C05/C,B,XLAMDA,ICONDK
0012          REAL*8 C(30,90),B(90),XLAMDA(90)
0013          INTEGER*2 ICONDK(90)
          COMMON C5 END
          COMMON C8
0014          COMMON /C08/ FXLINR
0015          REAL*8 FXLINR(30,32),FX(30),LINV(30,30),FR(30,30)
0016          EQUIVALENCE(FXLINR(1,1),FX(1))
0017          EQUIVALENCE(FXLINR(1,2),LINV(1,1))
0018          EQUIVALENCE(FXLINR(1,3),FR(1,1))
          COMMON C8 END
          COMMON C9
0019          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSDKFG,TYPG
0020          1, ISEED,ISOPT,IAOPT,IDOPT
          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSDKFG,TYPG
          COMMON C9 END
          COMMON C10
0021          COMMON /C10/UNCONX,UNCONF
0022          REAL*8 UNCONX(30),UNCONF
          COMMON C10 END
          COMMON C11
0023          COMMON /C11/X
0024          REAL*8 X(30)
          COMMON C11 END
          COMMON C13
0025          COMMON /C13/LH,IM
0026          REAL*8 LH(30),IM(30)
          COMMON C13 END
          COMMON C14
0027          COMMON /C14/F,GRAD
0028          REAL*8 F,GRAD(30)
          COMMON C14 END
          COMMON C16
0029          COMMON /C16/HIX,LOX,HIXLOX,HIG,LOG,HIGLOG,HILH,HIC,LOC,HICLOC,HIB
0030          REAL*8 HIX,LOX,HIXLOX,HIG,LOG,HIGLOG,HILH,HIC,LOC,HICLOC,HIB
          COMMON C16 END
          COMMON C17
0031          COMMON /C17/IUNITS,IUNIT6,IUNIT8
          COMMON C17 END
          COMMON C18
0032          COMMON /C18/SOLNX,SOLNF,SOLNQ

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0033          REAL*8 SOLNX(30),SOLNF
0034          INTEGER SOLNQ
          COMMON 18 END
          COMMON 19
0035          COMMON /C19/IPRIN,IPTR,IPMAX,NBRK
0036          INTEGER*2 IPRIN(5000),IPTR,IPMAX,NBRK(15)
          COMMON 19 END
          COMMON 20
0037          COMMON /C20/A
0038          REAL*8 A(30)
          COMMON 20 END
          COMMON 22
0039          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0040          REAL*8 TITLE(15)
0041          INTEGER PRTOPT
          COMMON 22 END
0042          REAL*8 TEMP(30),TEMP1(30)
          C
          C
          C      THIS SR WILL GENERATE A PROBLEM
          C
          C
          C      GENERATE G (UPPERTRIANGULAR)
          C
          C      CALCULATE RANGE OF NUMBERS FOR G
          C
0043          IRC = 0
0044          III = ISEED
          C
          DO 10 I = 2,M
0045             I1 = I-1
0046             DO 10 J=1,I1
0047                G(J,I) = RAND(III)*HIGLUG+LOG
0048          10
          C
          C      NOW FILL UP THE DIAGONALS OF G
          C
          C      DEPENDING UPON TYPE OF G (WELL OR ILL)
          C
          C
0049             DO 70 I=1,M
0050                SUM = 0.00
0051                DO 40 J= 1,M
0052                   IF (I-J)30,40,20
0053                   SUM=SUM+DABS(G(J,I))
0054                   GOTO 40
0055                   SUM=SUM+DABS(G(I,J))
0056                   CONTINUE
0057                   G(I,I) = SUM + RAND(III)
0058                   IF(TYPE.EQ.1)GOTO 60
          C      IT IS ILL CONDITIONED G
          C
          C
          IF(I.EQ.1) GOTO 50
0059             G(I,I)=G(I-1,I-1)+G(I,I)+SUM1
0060             SUM1 = SUM
0061             GOTO 70
0062             G(I,I) = G(I,I)+1.00
0063          70

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```

0064      7)      CONTINUE
          C
          C
          C
          C
          C      NOW DECOMPOSE G
          C
0065      IF(PRTOPT.GT.2)
0066      1CALL PRMAT(G,1,M,M,-1,M,ISIZ,ISIZ+1,1,1)
0067      CALL CLOCK (0)
0068      CALL DECOMP (IRC)
          IF(PRTOPT.GT.2)
0069      1CALL PRMAT(L,1,M,1,1,M,ISIZ,1,0,2)
          C
          C
          C      IF(IRC.NE.0) GOTO 1000
          C
          C      NOW CALL INVL TO INVERT L
          C
          C
          C      CALL INVL
0069      CALL CLOCK(1,TIME)
0070      ITIMDI = TIME * 100000
0071      IF(PRTOPT .GT.2)
0072      1CALL PRMAT(LINV,1,M,1,1,M,ISIZ,1,0,2)
          C
          C
0073      IF (IRC.NE.0)GOTO 1000
          C
          C
          C      GENERATE OPTIMUM PT.
          C
          C
0074      DO 30 I=1,M
0075      SOLNX(I)=RAND(III)*HIXLOX + LOX
0076      IF(PRTOPT.GT.1)WRITE(6,159)(SOLNX(I),I=1,M)
0077      159      FORMAT(' ',2X,'SOLNX =',(/9D14.7))
          C
          C
          C      GENERATE +VE LM FOR CONS. IN FINAL BASIS
          C
          C
0078      DO 90 I=1,K
0079      LM(I) = 0.00
0080      IF(I.LE.SOLNQ)LM(I) = RAND(III)*HILM
0081      DO 100 J= 1,M
          C
          C
          C      GENERATE THE CONSTRAINT I
          C
          C
0082      C(J,I) = RAND(III)*HICLOC +LOC
0083      100      CONTINUE
0084      S =DSQRT(SINNER(C(1,I),C(1,I),M))
0085      DO 110 J=1,M

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0086      110      C(J,I) = C(J,I)/S
0087      B(I) = SINNR(C(1,I),SOLNX,M)
0088      XLAMDA(I) = 0.00
0089      IF(I.GT.SOLNQ)XLAMDA(I) = RAND(III) * HIB
0090      B(I) = B(I) - XLAMDA(I)
0091      IF(PRTOPT.GT.2)WRITE(6,158)J,(C(J,I),J=1,M)
0092      158      FORMAT(' ',5X,'CONSTRAINT ',I5,'/(9D14.7)')
0093      90      CONTINUE
0094      IF(PRTOPT.GT.1)WRITE(6,164)(LM(I),I=1,SOLNQ)
0095      164      FORMAT(' ',5X,'L.M. OF ACTIVE CONS.',/(9D14.7)')
0096      IF(PRTOPT.GT.1)WRITE(6,165)(B(I),I=1,K)
0097      165      FORMAT(' ',5X,'RHS OF CONSTRAINTS',/(9D14.7)')
0098      IF(PRTOPT.GT.1)WRITE(6,166)(XLAMDA(I),I=1,K)
0099      166      FORMAT(' ',5X,'LAMNDAS OF CONSTRAINTS',/(9D14.7)')
C
C
C
C      CALCULATE A = NQ #LM
C
C
0100      DO 130 I=1,M
C
C      CALC GX
C
0101      130      TEMP(I) = SINNR(L(1,I),SOLNX(I),M-I+1)
0102      DO 140 I=1,M
0103      140      TEMP1(I) = SINNR(L(I,1),TEMP(I),I,ISIZ)
0104      IF(SOLNQ.EQ.0)GOTO200
0105      DO 150 I=1,M
0106      150      A(I) = SINNR(C(I,1),LM(I),SOLNQ,ISIZ)-TEMP1(I)
0107      DO 210 I=1,M
0108      210      X(I)=SOLNX(I)
0109      IF(PRTOPT.GT.1)WRITE(6,167) (A(J),J=1,M)
0110      167      FORMAT(' ',5X,'A VECTOR IN F ',/(9D14.7)')
C
C
C      CALC GRADIENT AND FIND THE OBJ. FN
C
C
0111      SOLNF = 0.5 * SINNR(SOLNX,TEMP1,M) + SINNR(SOLNX,A,M)
0112      1000      WRITE(IUNIT6,1005)IRC,SOLNF
0113      1005      FORMAT(' ',4X,' END GENER, IRC = ',I5,' SOLN F =',F16.7)
0114      RETURN
0115      END

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0001      SUBROUTINE DECOMP (IRC)
0002      IMPLICIT REAL*8(A-H,O-Z)
          COMMON C2
0003      COMMON /C02/LG
0004      REAL*8 LG(30,31),L(30,30),G(30,30)
0005      EQUIVALENCE(LG(1,1),L(1,1))
0006      EQUIVALENCE(LG(1,2),G(1,1))
          COMMON C2 END
          COMMON C9
0007      COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAOPT,IDOPT
0008      INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON C9 END
          COMMON C15
0009      COMMON /C15/TOLDCH,TOLLEP,TJLINV,TOLPR,TOLAM,TULMUL
          COMMON C15 END
          COMMON C17
0010      COMMON/C17/IUNIT5,IUNIT6,IUNIT8
          COMMON C17 END
          COMMON C21
0011      COMMON/C21/IAO,NSTEPS,NADDS,NDROPS,NDPS,NSORT,NKEIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON C21 END
          C      THIS SR WILL DECOMPOSE G, A +VE DEF. SYMM. MATRIX INTO ITS
          C      CHOLESKY FACTORS, LL(T).
          C      G IS ASSUMED TO BE STORED AS UPPER TRIANGULAR
          C
          C      TOLDCH IS TOLERANCE REQUIREMENT OF DECOMP.
          C      (I.E.) G - LL(T) > TOLDCH (ABS. ON BOTH SIDES)
          C
          C      FOR TESTING GENERATE G
          IRC = 0
0012      NSQRT = NSQRT + M
0013      L(1,1) = DSQRT(G(1,1))
0014      IF(M.LE.1) GOTO 100
0015      DO 20 I = 2,M
0016      DO 20 J=1,I
0017      SUM = 0.00
0018      IF(J.GT.1)SUM = SINNER(L(I,1),L(J,1),J-1,ISIZ,ISIZ)
0019      NGPS = NDPS + J - 1
0020      L(I,J) = G(J,I) - SUM
0021      IF(1.EQ.J) GOTO 10
0022      L(I,J) = L(I,J)/L(J,J)
0023      NDPS = NDPS + 1
0024      GOTO 20
0025      L(I,J) = DSQRT(L(I,J))
0026      10 CONTINUE
0027      NCK VERIFY BY FINDING LL(T)
          C
          C      DO 30 I=1,M
          C      DO 30 J= 1,I
          C      DIFF = G(J,I) - SINNER(L(I,1),L(J,1),J,ISIZ,ISIZ)
          C      IF(ABS(DIFF).LE.TOLDCH)GOTO 30
          C      IRC = IRC + 1
          C      WRITE(6,25)I,J,G(J,I),DIFF
          C25      FORMAT (' ',5X,'DECOMP',5X,' I,J,G(J,I),G-LL(T)',2I5,2F20.10)
          C30      CONTINUE
          RETURN
          END
    
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```

0001          SUBROUTINE INVL
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON 02
0003          COMMON /C02/LG
0004          REAL*8 LG(30,31),L(30,30),G(30,30)
0005          EQUIVALENCE(LG(1,1),L(1,1))
0006          EQUIVALENCE(LG(1,2),G(1,1))
          COMMON 02 END
          COMMON 08
0007          COMMON /C08/ FXLINR
0008          REAL*8 FXLINR(30,32),FX(30),LINV(30,30),FR(30,30)
0009          EQUIVALENCE(FXLINR(1,1),FX(1))
0010          EQUIVALENCE(FXLINR(1,2),LINV(1,1))
0011          EQUIVALENCE(FXLINR(1,3),FR(1,1))
          COMMON 08 END
0012          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAOPT,IOGPT
0013          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON 09 END
          COMMON 15
0014          COMMON /C15/TOLDCH,TOLDEP,TOLINV,TOLPR,TOLAH,TOLMUL
          COMMON 15 END
          COMMON 17
0015          COMMON/C17/IUNIT5,IUNIT6,IUNIT8
          COMMON 17 END
          COMMON 21
0016          COMMON/C21/IAC,NSTEPS,NADDS,NROPS,NOPS,NSORT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          C THIS SR INVERTS L AND INVERSE IS PUT IN LINV
          C LINV IS LOWER TRIANGULAR, SO IS L
          C
0017          IRC = 0
0018          LINV(1,1) = 1.00/L(1,1)
0019          NOPS = NOPS + 1
0020          IF(M.LE.1) GOTO 100
0021          DO 10 I=2,M
0022          LINV(1,I) = 1.00/L(I,I)
0023          IM1 = I-1
0024          DO 20 J1=1,IM1
0025          J = I-J1
0026          20 LINV(I,J) = -SINNER(LINV(I,J+1),L(J+1,J),J1,ISIZ)/L(J,J)
0027          NOPS = NOPS + I * (I+1)/2 - 1
0028          10 CONTINUE
          C DO 30 I = 1,M
          C DO 30 J=1,I
          C DIFF = 0.00
          C IF(I.EQ.J) DIFF=1.00
          C DIFF = DIFF - SINNER(L(I,J),LINV(J,J),I-J+1,ISIZ)
          C IF(ABS(DIFF).LE.TOLINV)GOTO 30
          C WRITE(6,25)I,J,DIFF
          C25 FORMAT(' ',5X,'INVL ',5X,'I,J,I - LL(-1)',2I5,F15.7)
          C IRC = IRC + 1
          C30 CONTINUE
          100 RETURN
          0030 ENC

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```

0001      SUBROUTINE SOLVL(Y,S)
0002      IMPLICIT REAL*8(A-H,O-Z)
          COMMON C2
0003      COMMON /C02/LG
0004      REAL*8 LG(30,31),L(30,30),G(30,30)
0005      EQUIVALENCE(LG(1,1),L(1,1))
0006      EQUIVALENCE(LG(1,2),G(1,1))
          COMMON C2 END
          COMMON C09
0007      COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISGPT,IAOPT,IDGPT
0008      INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON C09 END
          COMMON C21
0009      COMMON/C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON C21 END
          C      THIS SR WILL SOLVE FOR Y IN L Y = S
          C
          C      L IS LOWER TRIANGULAR MATRIX
0010      REAL*8 Y( 1),S( 1)
          C
          C
          C      Y IS CUTPUT FROM THIS ROUTINE
          C      P IS VECTOR INPUT
          C      M IS NO. OF VARIABLES (= M)
          C      ISIZ IS OFFSET FOR L (30)
          C
          C
          C
0011      Y(1) = S(1)/L(1,1)
0012      IF(M.LE.1)GOTO 100
0013      DO 10 I=2,M
0014      Y(I) = (S(I) - SINNER(L(I,1),Y(1),I-1,ISIZ))/L(I,I)
          C      DIFF = S(I)-SINNER(L(I,1),Y(1),I,ISIZ)
0015      10 CONTINUE
0016      100 NOPS = NOPS + M * (M+1)/2
0017      RETURN
0018      END

```

```

0001          SUBROUTINE SOLVLT(Y,S)
0002          IMPLICIT REAL*8(A-H,O-Z)

          COMMON C2
0003          COMMON /C02/LG
0004          REAL*8 LG(30,31),L(30,30),G(30,30)
0005          EQUIVALENCE(LG(1,1),L(1,1))
0006          EQUIVALENCE(LG(1,2),G(1,1))

          COMMON C2 END
          COMMON C9
0007          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAOPT,IDOPT
0008          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG

          COMMON C9 END
          COMMON C21
0009          COMMON/C21/IAG,NSTEPS,NADDS,NOROPS,NOPS,NSQRT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN

          COMMON C21 END
          C
          C      THIS SR SOLVES FOR Y IN L(T)Y = S
          C      L IS STORED LOWER TRIANGULAR
          C      M IS NUMBER OF VARIABLES
          C
          C
0010          REAL*8 Y( 1),S( 1)
0011          IRC = 0
0012          Y(M) = S(M)/L(M,M)
0013          IF(M.LE.1)GOTO 100
0014          I = M
0015          MM1 = I - 1
0016          DO 10 J= 1,MM1
0017             I=I-1
0018             Y(I) = (S(I)- SINNER(L(I+1,I),Y(I+1),M-1))/L(I,I)
0019          10 CONTINUE
0020          100 NOPS = NOPS + M * (M+1)/2
0021          RETURN
0022          END

```

```

0001          SUBROUTINE SOLVR(Y,S,ILIM,IST)
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON 04
0003          COMMON /C04/QIR
0004          REAL*8 QIR(30,32),Q1(30,30),R(30,30)
0005          EQUIVALENCE(QIR(1,1),Q1(1,1))
0006          EQUIVALENCE(QIR(1,3),R(1,1))
          COMMON 04 END
          COMMON 09
0007          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAOPT,IOOPT
0008          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON 09 END
          COMMON 21
0009          COMMON/C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          COMMON 22
0010          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0011          REAL*8 TITLE(15)
0012          INTEGER PRTOPT
          COMMON 22 END
          C
          C      THIS SR WILL SOLVE FOR Y IN R Y = P
          C
          C      R IS UPPER TRIANGULAR AND STORED SUCH
          C      Y IS OUTPUT AND IS A VECTOR
          C      S IS INPUT AND RHS SIDE VECTOR
          C      ILIM IS NUMBER OF VARIABLES
          C      ISIZ IS OFFSET FOR R
          C
          C
          C      REAL*8 Y( 1),S(1)
          C
0013          IF(IST.LT.ILIM)GOTO 110
0014          Y(IST) = S(IST)/R(IST,IST)
0015          IF(IST.LE.ILIM)GOTO 100
0016          I=IST
0017          I1 = IST-ILIM
0018          DO 10 J= 1,I1
0019          I=I-1
0020          Y(I)=(S(I)-SINNER(R(I,I+1),Y(I+1),IST-I,ISIZ))/R(I,I)
0021          DIFF = S(I) - SINNER(R(I,I),Y(I),IST-I+1,ISIZ)
          C      IF(ABS(DIFF).LE.0.0000500)GOTO 10
          C      WRITE(6,64) I,DIFF,S(1),(R(I,JK),JK=1,IST)
          C64  FORMAT(' ',5X,'SOLVR- 1, DIFF,S(1),(R(I,JK),JK=1,IST) ',/15,(9D12.6))
0022          10  CONTINUE
          C
          C      VERIFY IF R Y = S?
          C
          C
0023          100  NOPS = NOPS + (IST-ILIM+1)*(IST-ILIM+2)/2
0024          IF(PRTOPT.GT.1)WRITE(6,105)ILIM,IST,(Y(J),J=ILIM,IST)
0025          105  FORMAT(' ',5X,'SOLVR- IFRD - 1TO',2I5,/(9D14.7))
0026          110  RETURN
0027          END

```

```

0001          SUBROUTINE SOLVRT(Y,S)
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON C4
0003          COMMON /C04/Q1R
0004          REAL*8 Q1R(30,32),Q1(30,30),R(30,30)
0005          EQUIVALENCE(Q1R(1,1),Q1(1,1))
0006          EQUIVALENCE(Q1R(1,3),R(1,1))
          COMMON C4 END
          COMMON 09
0007          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAOPT,IDOPT
0008          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON 09 END
          COMMON 21
0009          COMMON/C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          C
          C          R IS UPPER TRIANGULAR AND STORED SUCH
          C          Y IS OUTPUT AND IS A VECTOR
          C          S IS INPUT AND RHS SIDE VECTOR
          C          M IS NUMBER OF VARIABLES
          C
          C
0010          REAL*8 Y( 1),S(1)
          C
0011          Y(1) = S(1)/R(1,1)
0012          IF(Q.LE.1)GOTO 100
0013          DO 10 I=2,Q
0014          Y(I) = (S(I)-SINNER(R(1,I),Y(1),I-1))/R(I,I)
0015          10 CONTINUE
0016          100 NOPS = NOPS + Q * (Q+1)/2
0017          RETURN
0018          END

```

FORTRAN IV G LEVEL 21

UNCONM

DATE = 80105

```
0001      SUBROUTINE UNCONM
0002      IMPLICIT REAL*8(A-H,O-Z)
          COMMON 09
0003      COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSQKFG,TYPG
          1,  ISEED,ISOPT,IAOPT,IDOPT
0004      INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSQKFG,TYPG
          COMMON 09 END
          COMMON 10
0005      COMMON /C10/UNCONX,UNCONF
0006      REAL*8 UNCONX(30),UNCONF
          COMMON 10 END
          COMMON 17
0007      COMMON/C17/IUNIT5,IUNIT6,IUNIT8
          COMMON 17 END
          COMMON 20
0008      COMMON /C20/A
0009      REAL*8 A(30)
          COMMON 20 END
          COMMON 21
0010      COMMON/C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          COMMON 22
0011      COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0012      REAL*8 TITLE(15)
0013      INTEGER PRTOPT
          COMMON 22 END
          C
          C      THIS SR SOLVES FOR THE UNCONSTRAINED MIN OF THE PROBLEM
          C
0014      CALL SOLVL(UNCONX,A)
0015      CALL SOLVLT(UNCONX,UNCONX)
0016      DO 10 I=1,M
0017      UNCONX(I) = -UNCONX(I)
0018      UNCONF = 0.5 * SINNR(A,UNCONX,M)
0019      WRITE(IUNIT6,11)UNCONF
0020      11  FORMAT(' ',6X,'UNCONSTRAINED MINIMUM = ', F16.7)
0021      IF(PRTOPT.GT.1)WRITE(IUNIT6,12)(UNCONX(I),I=1,M)
0022      12  FORMAT(' ',5X,'UNCONX'/(9D14.6/))
0023      NOPS = NOPS + M
0024      RETURN
0025      END
```

FORTRAN IV G LEVEL 21

RAND

DATE = 80105

```
0001      FUNCTION RAND(1)
0002      REAL*8 RAND
0003      I=I#59049
0004      IF (I.LT.0) I=I+2147483647+1
0005      RAND=I*0.4656613D-9
0006      RETURN
0007      END
```

```

0001          SUBROUTINE GRA
0002          IMPLICIT REAL*8(A-H,O-Z)
                COMMON 02
0003          COMMON /C02/LG
0004          REAL*8 LG(30,31),L(30,30),G(30,30)
0005          EQUIVALENCE(LG(1,1),L(1,1))
0006          EQUIVALENCE(LG(1,2),G(1,1))
                COMMON 02 END
                COMMON 09
0007          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
                1, ISEED,ISOPT,IAOPT,IDOPT
0008          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
                COMMON 09 END
                COMMON 11
0009          COMMON /C11/X
0010          REAL*8 X(30)
                COMMON 11 END
                COMMON 14
0011          COMMON /C14/F,GRAD
0012          REAL*8 F,GRAD(30)
                COMMON 14 END
                COMMON 20
0013          COMMON /C20/A
0014          REAL*8 A(30)
                COMMON 20 END
                COMMON 21
0015          COMMON /C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCDR
                1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
                COMMON 21 END
                COMMON 22
0016          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0017          REAL*8 TITLE(15)
0018          INTEGER PRTOPT
                COMMON 22 END
C
C
C THIS SR CALCULATES THE GRADIENT OF THE OBJ.FN
C AND EVALUATES THE VALUE OF THE FN.
C
C INPUT - L (LOWER TRIANGULAR MATRIX)
C         X (PT. X WHERE GRADIENT REQD)
C         A LINEAR PART OF THE OBJ. FN.
C         M NO. OF VARIABLES
C         ISIZ SIZE OF THE MATRIX L (30)
C
C OUTPUT - GRAD GRADIENT VECTOR
C          F OBJ. FN.
C          IRC RETURN CODE (0)
C
C TEMP. VAR TEMP VECTOR OF LENGTH ISIZ
C
C
C METHODOLOGY:
C          TEMP = L(T) * X
C          GRAD = L * TEMP
C          F = 0.5*( A(T) * X + X(T) * GRAD)

```

```

C
C   DATE       JULY 25 1979
C
0019      REAL*8 TEMP(30)
0020      IRC = 0
0021      DO 10 I = 1 , M
0022      TEMP(I) = SINNER(L(I,1),X(I),M-I+1)
0023      10   GRAD(I) = A(I) + SINNER(L(I,1),TEMP(I),I,ISIZ)
0024      F = 0.5 * (SINNER(X,GRAD,M) + SINNER(A,X,M))
0025      NOPS = NOPS + 2*M*(M+2) + 1
0026      20   FORMAT (' ',5X,'F = ',F16.7)
0027      IF (PRTOPT.GT.0)WRITE(6,20) F
0028      IF (PRTOPT.GT.1)WRITE(6,21)(GRAD(J),J=1,M)
0029      21   FORMAT(' ',5X,'GRADIENT',(/9D14.7))
0030      RETURN
0031      END

```

```

0001          SUBROUTINE PRTMAT(ARAY,NSTART,NROWS,NCOL1R,NRTCHG,MXCOL,
0002          1      ROFSET,RPOFST,POFSET,IWHAT)
              IMPLICIT REAL*8(A-H,O-Z)
C
C
C      THIS SR WILL PRINT A MATRIX, UPPER, LOWER, UPPER HESSENBERG
C      SQUARE
C      ARAY IS THE ARRAY THAT GETS REFERNCED
C      NSTART IS THE FIRST ELEMENT IN THE MATRIX
C      NROWS TH NO. OF ROWS TO BE PRINTED
C      NCOL1R IS THE NO. OF COLUMNS IN 1ST ROW TO BE PRINTED
C      NRTCHG IS THE RATE OF CHANGE IN NO. OF COLUMNS/ROWCHANGE
C          1 FOR LOWER TRIANGULAR, UPPER HESSENBERG
C          0 FOR SQUARE
C          -1 FOR UPPER TRIANGULAR
C      MXCOL IS MAX. NO OF COLUMNS TO BE PRINTED IN ANY ROW
C      ROFSET IS ADJACENT ELEMENTS IN SAME ROW OFFSET(USUALLY ISIZ)
C      RPOFST IS ADJACENT ROW TO BE PRINTED ELEMENT OFFSET(ISIZ+1 FOR
C          UPPER TRIANGULAR
C      POFSET IS FOR PRINTING OFFSET TO SHOW (EG - UPPER TRIANGULAR
C          WILL LOOK UPPER TRIANGULAR IF IT IS 1
C      IWHAT IS 1 FOR MATRIX G
C          2 L
C          3 LINV
C          4 QTLINV
C          5 Q1
C          6 R
C
C
0003          REAL*8 ARAY(1)
0004          REAL*8 IDATA(6) / ' G ' , ' L ' ,
1      ' LINV ' , ' QTLINV ' , ' Q1 ' , ' R ' ,
0005          INTEGER * 4 RPOFST,RPOFST,POFSET,CUREL,PREL
0006          REAL*8 PARAY(30)
C
C
0007          WRITE(6,5)IDATA(IWHAT)
0008          5      FORMAT(' ',5X,'MATRIX',4X,A8)
0009          DO 40 I= 1,NROWS
0010          PREL = NSTART + (I-1) * RPOFST
0011          CUREL = (I-1)*POFSET
0012          IF(CUREL.LE.0) GOTO 20
0013          DO 10 J=1,CUREL
0014          10      PARAY(J)=0.00
0015          20      NCOL = NCOL1R+NRTCHG*(I-1)
0016          IF(NCOL.GT.MXCOL)NCOL=MXCOL
0017          DO 30 J=1,NCOL
0018          CUREL=CUREL+1
0019          PARAY(CUREL)=ARAY(PREL)
0020          PREL=PREL+ROFSET
0021          30      CONTINUE
0022          WRITE(6,35)(PARAY(J),J=1,CUREL)
0023          35      FORMAT(' ',9D14.6)
0024          40      CONTINUE
0025          RETURN
0026          END
    
```

```

0001          SUBROUTINE PROJ
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON 01
0003          COMMON /C01/PROJV,LENY,QLZ
0004          REAL * 8 PROJV(30),LENY,QLZ(30)
          COMMON 01 END
          COMMON 03
0005          COMMON /C03/QLTINV
0006          REAL*8 QLTINV(30,30)
          COMMON 03 END
          COMMON 09
0007          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISLEO,ISOPT,IAOPT,IDUPT
0008          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON 09 END
          COMMON 21
0009          COMMON/C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSORT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          COMMON 22
0010          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0011          REAL*8 TITLE(15)
0012          INTEGER PRTOPT
          COMMON 22 END
          C
          C      THIS SR WILL FIND PROJECTION VECTOR PROJV FOR A GIVEN QTLZ
          C
          C      INPUT
          C      QTLINV IS THE FACTOR OF B (Q(T)L(-1))
          C      M - NO. OF VARIABLES
          C      Q - NO. ACTIVE CONSTRAINTS
          C      ISIZ - USED TO FIND THE OFFSET (30)
          C
          C      OUTPUT
          C      PROJV - PROJECTED VECTOR
          C      LENY - LENGTH OF THE PROJECTED VECTOR
          C
          C      METHODOLOGY:
          C      IF Q = M THEN Y=LENY=0.0
          C      ELSE
          C      QTLZ=(W1 W2) WHERE W2 IS OF SIZE M-Q
          C      CALCULATE W2
          C      LET W= (0 W2)
          C      PROJV = QTLINV(T) * W
          C      CALCULATE LENGTH OF PROJV
          C
          C      DATE : JULY 28 1979
          C
          C
          C
0013          INTEGER QP1,MMQ
          C
0014          QP1=Q+1
0015          MMQ=M-Q
0016          LENY=0.00
0017          IF(Q.LT.M)GOTO 5
0018          DO 3 I=1,M

```

```
0019      3      PROJV(I)=0.DO
0020      GOTO 30
      C
      C      CALCULATE QTLINV(T) (0 W2)
      C
0021      5      DO 20 I=1,M
0022      20      PROJV(I)=SINNER(QTLINV(QP1,I),QTLZ(QP1),MMQ)
0023      LENY=DSQRT(SINNER(PROJV,PROJV,M))
0024      NSQRT = NSQRT + 1
0025      30      IF(PRTOPT.GT.1)WRITE(6,11)LENY
0026      11      FORMAT(' ',5X,'LENGTH OF PROJV = ', F16.7)
0027      IF(PRTOPT.GT.1)WRITE(6,12)(PROJV(I),I=1,M)
0028      12      FORMAT(' ',5X,'PROJV '/(9D14.6/))
0029      NQPS = NQPS + (MMQ+1)*M
0030      RETURN
0031      END
```

```

0001          SUBROUTINE RBAR(Z,RZ,IFRO,ITO)
0002          IMPLICIT REAL*8(A-H,C-Z)
          COMMON 03
0003          COMMON /C03/QTLINV
0004          REAL*8 QTLINV(30,30)
          COMMON 03 END
          COMMON 09
0005          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAOPT,IDOPT
0006          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON 09 END
          COMMON 21
0007          COMMON/C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          COMMON 22
0008          COMMON /C22/ TITLE,NPAGE,NLINE,MALINE,PRTOPT
0009          REAL*8 TITLE(15)
0010          INTEGER PRTOPT
          COMMON 22 END
          C
          C      THIS SR WILL CALCULATE QTLINV * Z = RZ FOR GIVEN Z
          C
          C      INPUT
          C      Z - VECTOR FOR WHICH RZ IS          REQD. (GRAD OR N(Q+1))
          C      QTLINV - Q(T)L(-1) MATRIX
          C      Q - SIZE OF R
          C      M - SIZE OF QTLINV
          C      IFRO,ITO- THE LOWER AND UPPER BOUNDS FOR WHICH RZ IS DESIR.
          C
          C      OUTPUT
0011          REAL*8 Z( 1),RZ( 1)
          C
          C
          C      IF (ITO.LT.IFRO)RETURN
          C      DO 10 I=IFRO,ITO
0012          C      RZ(I)=SINNER(QTLINV(I,1),Z(1),M,ISIZ)
0013          C      RZ(I)=SINNER(QTLINV(I,1),Z(1),M,ISIZ)
0014          10      RZ(I)=SINNER(QTLINV(I,1),Z(1),M,ISIZ)
0015          C      NOPS=NOPS+(ITO-IFRO+1)*M
0016          C      IF(PRTOPT.GT.1)WRITE(6,12)IFRO,ITO,(RZ(I),I=IFRO,ITO)
0017          12      FORMAT(' ',5X,'BAR,IFRO-ITO',2I5,'/(9D14.6)')
0018          C      RETURN
0019          C      END

```



```

0001          SUBROUTINE ADD(ICGNAD)
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON 01
0003          COMMON /C01/PROJV,LENY,QLZ
0004          REAL * 8 PROJV(30),LENY,QLZ(30)
          COMMON 01 END
          COMMON 03
0005          COMMON /C03/QTLINV
0006          REAL*8 QTLINV(30,30)
          COMMON 03 END
          COMMON 04
0007          COMMON /C04/Q1R
0008          REAL*8 Q1R(30,32),Q1(30,30),R(30,30)
0009          EQUIVALENCE(Q1R(1,1),Q1(1,1))
0010          EQUIVALENCE(Q1R(1,3),R(1,1))
          COMMON 04 END
          COMMON 05
0011          COMMON /C05/C,B,XLAMDA,ICONOK
0012          REAL*8 C(30,90),B(90),XLAMDA(90)
0013          INTEGER*2 ICONOK(90)
          COMMON 05 END
          COMMON 09
0014          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAUPT,IDOPT
0015          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON 09 END
          COMMON 12
0016          COMMON /C12/NQ
0017          INTEGER NQ(30)
          COMMON 12 END
          COMMON 17
0018          COMMON/C17/IUNIT5,IUNIT6,IUNIT8
          COMMON 17 END
          COMMON 21
0019          COMMON/C21/IAG,NSTEPS,NADDS,NDRGDS,NDPS,NSQRT,NREIN,NXCDR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          COMMON 22
0020          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0021          REAL*8 TITLE(15)
0022          INTEGER PRTOPT
          COMMON 22 END
          COMMON 23
0023          COMMON/C23/GC,GH,GP,GB
0024          REAL * 8 GC(30),GH(30),GP(30),GB(30)
          COMMON 23 END
0025          INTEGER QP1
C
C          THIS SR. WILL ADD A CONSTRAINT TO THE BASIS AND UPDATE
C          THE OPERATORS, QTLINV AND R.
C
C          INPUT
C          Q # OF ACTIVE CONSTRAINTS
C          M # OF VARIABLES
C          ISIZ OFFSET FOR ROW STORAGE IN AN ARRAY
C          QTLZ = QTLINV * N(Q+1)

```

```

C      R UPPER TRIANGULAR FACTOR OF B (L(-1)N(Q))
C      QTLINV SQUARE MATRIX (Q(T) L(-1))
C
C      N(Q+1) IS ASSUMED TO BE LIN. INDEPENDENT.
C
C      METHODOLOGY:
C      Q+1ST COL. OF R IS SET TO QTLZ
C      CALCULATE COEFFICIENTS IN THE GIVENS MATRIX RECD. TO
C      REDUCE THE Q+1ST COL. OF R TO 0.0 EVERYWHERE BELOW
C      ELEMENT Q+1.
C
C      THIS IS DONE BY WORKING FROM BOTTOM UP AND FINDING
C      GC(J),GH(J), J = M-1,Q+1,-1
C      THIS CALCULATION IS DONE IN A WAY DESCRIBED IN DANIEL,
C      GRAGG,KAUFMAN ET AL. PAPER SO AS TO AVOID OVERFLOW AND
C      UNDERFLOW.
C
0026      QP1 = Q+1
0027      MM1 = M-1
0028      MMQ = M - Q
C
C      NOW CALCULATE THE PRODUCT OF THE GIVENS MATRICES.
C      SINCE THE PRODUCT HAS A SPECIAL FORM AS DESCRIBED IN
C      GILL GCLUB SAUNDERS ET AL. PAPER. CALCULATE
C      THE RECD. VECTORS TO PRODUCE THE PRODUCT.
C      PRODUCT MATRIX IS UPPER TRIANGULAR AND A SUBDIAGONAL.
C
C      QTLINV IS NOW UPDATED
C
C      NEW QTLINV = ( I 0 )   OLD QTLINV
C                  ( 0 C1 )
C
C      WHERE Q1 IS THE PRODUCT OF GIVENS MATRICES AS ABOVE.
C
C      DATE JULY 29, 1979
C
C      QTLZ IS THE VECTOR Q(T)L(-1) * (N+).
C      IT MUST HAVE BEEN CALCULATED PRIOR TO CALL OF THE ROUTINE
C      CALCULATE GIVENS MATRIX COEFFICIENTS FROM M-1 THRU Q+1
C
0029      30      ICURR = M
0030              IF (CP1.EQ.M)GOTO 70
0031              DO 60 I = QP1,MM1
0032              ICURR1 = ICURR - 1
0033              CALL GIVENV(QTLZ(ICURR1),QTLZ(ICURR),GC(ICURR1),GH(ICURR1),
0034              1GP(ICURR1))
0034              ICURR = ICURR1
0035      60      CONTINUE
C
C      NOW FIND THE PRODUCT WITH GIVENS MATRICES
0036      DO 64 I=1,M
0037              IC=M
0038              DO 63 J=QP1,MM1
0039              IC1=IC-1
0040              CALL GVINUPV(GC(IC1),GH(IC1),GP(IC1),QTLINV(IC1,I),

```

```
      1QTLINV(IC,I),NOPS)
0041      IC=IC-1
0042      63      CONTINUE
0043      64      CONTINUE
      C
0044      70      DC 65 I = 1,QP1
0045      55      R(I,QP1)=QTLZ(I)
      C
      C      UPDATE Q
      C
0046      Q = QP1
0047      CALL UPRIN(ICONAD)
0048      NADDS = NADDS + 1
0049      NQ(Q) = ICONAD
0050      IF(PRTOPT.GT.0)WRITE(IUNIT6,71)Q,(NQ(I),I=1,Q)
0051      71      FORMAT(' ',I7,' - IN BASIS ',30I3)
0052      RETURN
0053      END
```

```

0001          SUBROUTINE GIVENH(R1,R2,GC1,GH1,GMU)
0002          IMPLICIT REAL*8(A-H,O-Z)
0003          COMMON 21
          COMMON/C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCDR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          C THIS SR WILL FIND THE COEFFICIENTS IN GIVENS MATRIX
          C FOR A PAIR OF NUMBERS OF THE RHS
          C
          C      ( GC1  GH1 )  R1   =   NEW R1
          C      ( GH1 -GC1 )  R2   =   0.0 = NEW R2
          C
          C ON RETURN FROM THE ROUTINE R2 WILL BE SET TO 0.0 & R1
          C WILL BE SET TO SIGN(MAX(ABS(R1),ABS(R2)) * SQRT(R1**2+R2**2))
          C GC & GH WILL BE CALCULATED AS EXPLAINED IN DANIEL,CRAGG,KAUFM
          COMMON 22
          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
          REAL*8 TITLE(15)
          INTEGER PRTOPT
          COMMON 22 END
0004          IF(R2 .NE.0.00)GOTO 40
0005          GC1 = 1.00
0006          GH1 = 0.00
          GMU=0.00
          GOTO 60
0007          40 IF(R1.NE.0.00)GOTO 45
0008          GC1=0.00
0009          GH1=1.00
          R1=R2
          R2=0.00
          GMU=1.00
          GOTO 60
0010          45 RMU1=DABS(R1)
0011          RMU2=DABS(R2)
          IF(RMU1.LT.RMU2)GOTO 47
          TAU=R1*DSQRT(1.00+(RMU2/RMU1)**2)
          GOTO 50
0012          47 TAU=R2*DSQRT(1.00+(RMU1/RMU2)**2)
0013          GC1=R1/TAU
          GH1=R2/TAU
          GMU=R2/(R1+TAU)
          R1=TAU
          R2=C.00
          NSQRT=NSQRT+1
          NOPS=NOPS+4
          IF(GC1.LT.1.00)GOTO 55
          GC1 = 1.00
          GH1=0.00
          GOTO 60
0014          55 IF(GH1.LT.1.00)GOTO 60
          GC1=0.00
          GH1=1.00
          60 IF (PRTOPT.GT.3)WRITE(6,61)R1,R2,GC1,GH1,GMU
          61 FORMAT(' ',R1,R2,GC1,GH1,GMU',5D16.8)
          RETURN
          END
0041
0042

```

```

0001          SUBROUTINE DROP(ICONDR)
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON C3
0003          COMMON /C03/QTLINV
0004          REAL*8 QTLINV(30,30)
          COMMON C3 END
          COMMON C4
0005          COMMON /C04/Q1R
0006          REAL*8 Q1R(30,32),Q1(30,30),R(30,30)
0007          EQUIVALENCE(Q1R(1,1),Q1(1,1))
0008          EQUIVALENCE(Q1R(1,3),R(1,1))
          COMMON C4 END
          COMMON C5
0009          COMMON /C05/C,B,XLAMDA,ICONDK
0010          REAL*8 C(30,90),B(90),XLAMDA(90)
0011          INTEGER*2 ICONDK(90)
          COMMON C5 END
          COMMON C9
0012          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISZED,ISOPT,IAOPT,IDOPT
0013          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON C9 END
          COMMON C12
0014          COMMON /C12/NQ
0015          INTEGER NQ(30)
          COMMON C12 END
          COMMON C13
0016          COMMON /C13/LM,IM
0017          REAL*8 LM(30),IM(30)
          COMMON C13 END
          COMMON C17
0018          COMMON /C17/IUNIT5,IUNIT6,IUNIT8
          COMMON C17 END
          COMMON C21
0019          COMMON /C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,MXCDR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON C21 END
          COMMON C22
0020          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0021          REAL*8 TITLE(15)
0022          INTEGER PRTOPT
          COMMON C22 END
          COMMON C23
0023          COMMON /C23/GC,GH,GP,GB
0024          REAL*8 GC(30),GH(30),GP(30),GB(30)
          COMMON C23 END
C          THIS SR WILL UPDATE FACTORS WHILE DRIPPING A CONS. FROM BASIS
C
C          INPUT
C          P - THE INDEX OF THE CONSTRAINT IN THE BASIS TO BE DROPP
C          Q NO. OF CURRENTLY ACTIVE CONSTRAINTS.
C          M NO. OF VARIABLES
C          ISIZ OFFSET FOR MATRICES TO FIND ADJACENT ELEMENTS OF A
C          R UPPER TRIANGULAR FACTOR
C          QTLINV SQUARE MATRIX OF SIZE M (FACTOR * L(-1))
C

```

```

C      METHODOLOGY:
C      IF (Q = 1 ) THEN QTLINV = LINV & RETURN
C      FIND GIVENS MATRIX COEFF. REQD TO REDUCE DIAGONAL
C      ELEMENTS OF R TO 0.0 FROM P+1 TO Q.
C      MOVE R TO LEFT BY 1 COL.SO THAT COLS. P+1 TO Q
C      OCCUPY COL. P TO Q-1
C
C      NOW CALCULATE THE PRODUCT OF THE GIVENS MATRICES.
C      THIS IS LOWER TRIANGULAR AND A SUPERDIAGONAL FORM
C      INSTEAD OF STORING THE MATRIX STORE VECTORS
C      GP,GB AND GH SO THAT ANY ELEMENT OF THE PRODUCT CAN BE
C      CALCULATED EASILY FROM THE ABOVE VECTORS.
C
C      NOW UPDATE QTLINV AS FOLLOWS.
C
0025      INTEGER PP1,QM1
0026      REAL*8 MU,MU1,TAU
C      NEW QTLINV = ( I 0 0 )   OLD QTLINV
C                  ( 0 Q1 0 )
C                  ( 0 0 I )
C      WPERE Q1 IS PRODUCT OF GIVENS MATRICES OF THE FORM
C      LOWER TRIANGULAR WITH SUPER DIAGONAL
C
C      DATE JULY 29, 1979
C
0027      P = ICONDR
0028      JCONDR = NJ(ICONDR)
0029      PP1 = P + 1
0030      QM1 = Q - 1
0031      IF(Q.NE.1) GOTO 40
0032      CALL INTEQL
0033      GOTO 100
0034      IF(P.EQ.Q)GOTO 100
0035      IZ=0
C
C      CALCULATE GIVENS MATRIX COEFFICIENTS, P THRU QM1
C
0036      DO 80 I=P,QM1
0037      IP1 = I+1
0038      IP2 = I+2
0039      CALL GIVENM(R(I,IP1),R(IP1,IP1),GC(I),GH(I),GP(I))
0040      IF(PRIOPT.GT.4)WRITE(6,17)I,GC(I),GH(I)
0041      17  FORMAT(' ',5X,'ICURR1,GC,GH      ',15,2F14.7)
C
C      NOW UPDATE I,IP1 ROWS OF R TO THE RIGHT OF COL IP1
C
0042      65  IF(I.EQ.QM1)GOTO 75
0043      DO 70 J=IP2,Q
0044      CALL GVNUPV(GC(I),GH(I),GP(I),R(I,J),R(IP1,J),NOPS)
0045      70  CONTINUE
0046      75  CONTINUE
0047      80  CONTINUE
C
C      FIND THE PRODUCT OF GIVENS MATRICES
C

```

```

C
C   UPDATE QTLINV
C
C
0048          DO 85 I=1,M
0049          DJ 83 J=P,QM1
0050          JP1=J+1
0051          CALL GVNUPV(GC(J),GH(J),GP(J),QTLINV(J,I),QTLINV(JP1,I),
                INCP5)
0052          83  CONTINUE
0053          85  CONTINUE
C   NOW MOVE P+1ST THRU Q TH COLS. OF R TO P THRU Q-1
C   MOVE R TO LEFT
C
0054          IF(P.EQ.Q)GOTO 100
0055          DO 90 I = P,QM1
0056          IP1 = I+1
0057          NQ(I) = NQ(IP1)
0058          LM(I)=LM(IP1)
0059          IM(I)=IM(IP1)
0060          DO 90 J = 1,I
0061          90  R(J,I) = R(J,IP1)
0062          100  Q = Q-1
0063          CALL UPRIN(-JCONDR)
0064          NDROPS = NDROPS +1
0065          IF(Q.EQ.0)NQ(1)=0
0066          IF(PRTOPT.GT.0)WRITE(IUNIT6,71)Q,(NQ(I),I=1,Q)
0067          71  FORMAT(' ',I7,' - IN BASIS ',30I3)
0068          RETURN
0069          END

```

```

J001          SUBROUTINE INTLQL
J002          IMPLICIT REAL*8(A-H,O-Z)
              COMMON C3
J003          COMMON /C03/ QTLINV
J004          REAL*8 QTLINV(30,30)
              COMMON C3 END
              COMMON C8
J005          COMMON /C08/ FXLINR
J006          REAL*8 FXLINR(30,32),FX(30),LINV(30,30),FR(30,30)
J007          EQUIVALENCE(FXLINR(1,1),FX(1))
J008          EQUIVALENCE(FXLINR(1,2),LINV(1,1))
J009          EQUIVALENCE(FXLINR(1,3),FR(1,1))
              COMMON C8 END
              COMMON C9
J010          COMMON /C09/M,P,Q,K,GN,OFF,ISIZ,KINDPR,FSOKFG,TYPC
              1, ISEED,ISOPT,IAOPT,IOOPT
J011          INTEGER M,P,Q,K,GN,OFF,ISIZ,KINDPR,FSOKFG,TYPC
              COMMON C9 END
C              SET QTLIV = LINV
C
J012          DO 30 I = 1,M
J013          DO 30 J = 1,M
J014          IF(J-I)10,20,20
J015          10  QTLINV(J,I) = 0.00
J016          GO TO 30
J017          20  QTLINV(J,I) = LINV(J,I)
J018          30  CONTINUE
J019          RETURN
J020          END
    
```

```

J001          SUBROUTINE GVNUPV(CC1,GH1,GMU,QT1,QT2,NGPS)
C
C          THIS SUBROUTINE MULTIPLIES A VECTOR OF 2 ELEMENTS WITH GIVENS
C          MATRIX IN AN EFFICIENT MANNER
C
C          ( CC1  GH1 ) ( QT1 ) = ( QT1 )
C          ( GH1 -CC1 ) ( QT2 ) = ( QT2 )
C
C          OLD QT1,QT2 ARE REPLACED BY THE NEW VALUES ON EXIT
C          GMU IS R2/(R1+TAU) AS GIVEN IN BARTELS, GILL, ETAL
C
J002          IMPLICIT REAL*8(A-H,O-Z)
J003          IF(CC1.NE.1.00)GOTO 10
J004          QT2=-QT2
J005          GO TO 30
J006          10  IF(GH1.NE.1.00)GOTO 20
J007          QTEMP=QT1
J008          QT1=QT2
J009          QT2=QTEMP
J010          GO TO 30
J011          20  QTEMP=QT1*CC1+QT2*GH1
J012          QT2=GMU*(QT1+QTEMP)-QT2
J013          QT1=QTEMP
J014          NGPS=NGPS+3
J015          30  RETURN
J016          END
    
```

```

0001          SUBROUTINE LAMDA
0002          IMPLICIT REAL*8(A-H,D-Z)
          COMMON 11
0003          COMMON /C11/X
0004          REAL*8 X(30)
          COMMON 11 END
          COMMON 05
0005          COMMON /C05/C,B,XLAMDA,ICDNOK
0006          REAL*8 C(30,90),B(90),XLAMDA(90)
0007          INTEGER*2 ICDNOK(90)
          COMMON 05 END
          COMMON 09
0008          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAOPT,ICOPT
0009          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON 09 END
          COMMON 21
0010          COMMON /C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCDR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          COMMON 22
0011          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0012          REAL*8 TITLE(15)
0013          INTEGER PRTOPT
          COMMON 22 END
C          THIS SR CALCULATES LAMDAS (I.E.) VIOLATIONS
C          INPUT
C          M      NO. OF VARIABLES
C          K      NO. OF CONSTRAINTS
C          X      POINT X
C          C(*,*) CONSTRAINT NORMALS 1/COLUMN OF C
C          B      VECTOR OF LIMITING VALUES(RHS OF C)
C          XLAMDA VECTOR C(T) * X - B
C
C          METHODOLOGY:
C          CALC. C(T)X - B
C          DATE   JULY 31 1979
C          DO 10 I=1,K
0014          XLAMDA(I) = MINER(C(1,I),X(1),M) - B(I)
0015          IF(PRTOPT.GT.1)WRITE(6,11)(XLAMDA(I),I=1,K)
0016          11  FORMAT(' ',5X,'LAMDAS',/(9D14.7))
0017          NOPS = NOPS + M * K
0018          RETURN
0019          END
0020

```

```

0001          SUBROUTINE XCOR
0002          IMPLICIT REAL*8(A-H,G-Z)
0003          REAL*8 TEMP(30),TEMP1(30)

C
C      THIS SR WILL DO A X-CORRECTION SO THAT X MAY LIE ON THE
C      DESIRED MANIFOLD
C      ASSUMES NQ(*) IS REASONABLY ACCURATE.
C      SEE ROSENS PAPER
C
C      METHODOLOGY:
C          X = X - NQ(*) (T) * LAMDA
C
COMMON 03
0004          COMMON /C03/QTLINV
0005          REAL*8 QTLINV(30,30)
COMMON 03 END
COMMON 05
0006          COMMON /C05/C,B,XLAMDA,ICONDK
0007          REAL*8 C(30,90),B(90),XLAMDA(90)
0008          INTEGER*2 ICONDK(90)
COMMON 05 END
COMMON 09
0009          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
0010          1, ISIED,ISOPT,IAOPT,IDOPT
0011          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
COMMON 09 END
COMMON 11
0011          COMMON /C11/X
0012          REAL*8 X(30)
COMMON 11 END
COMMON 12
0013          COMMON /C12/NQ
0014          INTEGER NQ(30)
COMMON 12 END
COMMON 21
0015          COMMON /C21 /IAG,NSTEPS,NADDS,NOROPS,NOPS,N5JRT,NREIN,NXCOR
0016          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
COMMON 21 END
COMMON 22
0016          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0017          REAL*8 TITLE(15)
0018          INTEGER PRTOPT

C
COMMON 22 END
0019          IF (Q.EQ.0) RETURN
0020          DO 10 I=1,Q
0021          10      TEMP(I) = XLAMDA(NQ(I))

C
C      SOLVE FOR Y IN R(T) * Y = LAMDA
C
0022          CALL SOLVRT(TEMP1,TEMP)

C
C      UPDATE X
C
0023          DO 20 I=1,M
0024          20      X(I) = X(I) - SINER(QTLINV(1,I),TEMP1(1),Q)
0025          IF(PRTOPT.GT.2)WRITE(6,21)(X(J),J=1,M)
0026          21      FORMAT(' ',5X,'X AFTER X-CORRECTION',(/,9D14.7))
0027          NOPS = NOPS + M * Q
0028          NXCOR = NXCOR + 1
0029          CALL UPRIN(-500-NXCOR)

C      INDICATE THAT AN XCORRECTION WAS JUST MADE
C
0030          RETURN
0031          END

```

```

0001      SUBROUTINE REIN (IRC)
0002      IMPLICIT REAL*8(A-H,O-Z)

COMMON 01
0003      COMMON /C01/PROJV,LENY,QLTZ
0004      REAL * 8 PROJV(30),LENY,QLTZ(30)
COMMON 01 END
COMMON 05
0005      COMMON /C05/C,B,XLAMDA,ICDNOK
0006      REAL*8 C(30,90),B(9C),XLAMDA(90)
0007      INTEGER*2 ICDNOK(90)
COMMON 05 END
COMMON 09
0008      COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSQKFG,TPG
0009      1, ISEED,ISOPT,IAOPT,IDOPT
COMMON 09 END
COMMON 12
0010      COMMON /C12/NQ
0011      INTEGER NQ(30)
COMMON 12 END
COMMON 15
0012      COMMON /C15/TOLDCH,TOLDEP,TOLINV,TOLPR,TOLAM,TOLMUL
COMMON 15 END
COMMON 17
0013      C
COMMON/C17/IUNIT5,IUNIT6,IUNIT8
COMMON 17 END
COMMON 21
0014      COMMON/C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCOR
1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
COMMON 21 END
C
C      THIS SR WILL REINVERT, (I.E.) WILL RECOMPUTE THE
C      OPERATORS , QTLINV AND R FROM SCRATCH
C
C      METHODOLOGY:
C      QTLINV = LINV
C      IF Q = C RETURN
C      SAVE Q AND NQ INTO COPYQ AND COPYNQ
C      ADD THE FIRST CONS TO THE BASIS, UPDATE OPS.
C      DO OVER THE REST OF THE CONSTRAINTS
C      PROJECT NORMAL
C      IF L. INDEP THEN ADD TO THE BASIS & UPDATE
C      END DO
C      CALCULATE LAMDA
C      DO AN XCORRECTION
C      CHECK IF ACTIVE CONSTRAINTS ARE SATISFIED
C      IF YES, RETURN ELSE SET IRC = CONS. # THAT IS VIOL.
C
C      DATE JULY 30 1979
C
C
0015      INTEGER COPYNQ(30),COPYQ
0016      IRC = 0
0017      IF (NREIN.LT.MXREIN)GOTO 5

```

```

0018      WRITE(6,6)MXREIN
0019      IRC = 1
0020      6      FORMAT(' ',5X,'NO. OF REINVERSIONS EXCEED MAX OF ',I5)
0021      GOTO 110
0022      5      CONTINUE
0023      NREIN = NREIN + 1
0024      CALL UPRIN(-600-NREIN)
0025      CALL INTLQL
      C
0026      IF(Q.EQ.0) GOTO 100
      C      COPYQ = Q
0027      COPYQ = Q
0028      DO 40 I=1,Q
0029      40      COPYNQ(I) = NQ(I)
0030      Q=0
0031      CALL RBAR(C(1,COPYNQ(1)),QTLZ,1,M)
0032      CALL ADD(COPYNQ(1))
0033      IF(COPYQ.LE.Q) GOTO 100
0034      DO 50 I=2,COPYQ
0035      CALL RBAR(C(1,COPYNQ(I)),QTLZ,Q+1,M)
0036      LENY = SINER(QTLZ(Q+1),QTLZ(Q+1),M-Q)
0037      NQPS=NQPS+M-Q
0038      IF(LENY.LE.TOLDEP)GOTO 50
0039      CALL RBAR(C(1,COPYNQ(I)),QTLZ,1,Q)
0040      CALL ADD(COPYNQ(I))
0041      50      CONTINUE
0042      100     CALL UPRIN(-600-NREIN)
0043      110     RETURN
0044      END

```

```

0001          SUBROUTINE FEAS(IRC,III)
0002          IMPLICIT REAL*8(A-H,O-Z)

          COMMON C1
0003          COMMON /C01/PROJV,LENY,QLTZ
0004          REAL * 8 PROJ V(30),LENY,QLTZ(30)
          COMMON 01 END
          COMMON 02
0005          COMMON /C02/LG
0006          REAL*8 LG(30,31),L(30,30),G(30,30)
0007          EQUIVALENCE(LG(1,1),L(1,1))
0008          EQUIVALENCE(LG(1,2),G(1,1))
          COMMON 02 END
          COMMON 03
0009          COMMON /C03/QLTINV
0010          REAL*8 QLTINV(30,30)
          COMMON 03 END
          COMMON 04
0011          COMMON /C04/Q1R
0012          REAL*8 Q1R(30,32),Q1(30,30),R(30,30)
0013          EQUIVALENCE(Q1R(1,1),Q1(1,1))
0014          EQUIVALENCE(Q1R(1,3),R(1,1))
          COMMON 04 END
          COMMON 05
0015          COMMON /C05/C,B,XLAMDA,ICONDK
0016          REAL*8 C(30,90),B(90),XLAMDA(30)
0017          INTEGER*2 ICONDK(90)
          COMMON 05 END
          COMMON 06
0018          COMMON /C06/FF, FNQ, FQ
0019          INTEGER FNQ(30),FQ
0020          REAL*8 FF
          COMMON 06 END
          COMMON 07
0021          COMMON /C07/ FQTLIN
0022          REAL*8 FQTLIN(30,30)
          COMMON 07 END
          COMMON 08
0023          COMMON /C08/ FXLINR
0024          REAL*8 FXLINR(30,32),FX(30),LINV(30,30),FR(30,30)
0025          EQUIVALENCE(FXLINR(1,1),FX(1))
0026          EQUIVALENCE(FXLINR(1,2),LINV(1,1))
0027          EQUIVALENCE(FXLINR(1,3),FR(1,1))
          COMMON 08 END
          COMMON 09
0028          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,F5OKFG,TYPG
          1, ISEED,ISOPT,IAOPT,IDOPT
0029          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,F5OKFG,TYPG
          COMMON 09 END
          COMMON 10
0030          COMMON /C10/UNCONX,UNCONF
0031          REAL*8 UNCONX(30),UNCONF
          COMMON 10 END
          COMMON 11
0032          COMMON /C11/X
0033          REAL*8 X(30)
          COMMON 11 END

```

FORTRAN IV G LEVEL 21

FEAS

DATE = 80105

```
COMMON 12
0034 COMMON /C12/NQ
0035 INTEGER NQ(30)
COMMON 12 END
COMMON 13
0036 COMMON /C13/LM,IM
0037 REAL*8 LM(30),IM(30)
COMMON 13 END
COMMON 14
0038 COMMON /C14/F,GRAD
0039 REAL*8 F,GRAD(30)
COMMON 14 END
COMMON 15
0040 COMMON /C15/TOLDCH,TOLDEP,TOLINV,TOLPR,TOLAM,TOLMUL
COMMON 15 END
COMMON 16
0041 COMMON/C16/HIX,LOX,HIXLOX,HIG,LOG,HIGLOG,HILM,HIC,LOC,HICLOC,HIB
0042 REAL*3 HIX,LOX,HIXLOX,HIG,LOG,HIGLOG,HILM,HIC,LOC,HICLOC,HIB
COMMON 16 END
COMMON 17
0043 COMMON/C17/IUNIT5,IUNIT6,IUNIT8
COMMON 17 END
COMMON 19
0044 COMMON /C19/IPRIN,IPTR,IPMAX,NBRK
0045 INTEGER*2 IPRIN(5000),IPTR,IPMAX,NBRK(15)
COMMON 19 END
COMMON 21
0046 COMMON/C21/IAG,NSTEPS,NAGDS,NDRDPS,NOPS,NSORT,HREIN,NXCJR
1,NFSTEP,NPSTEP,ICC,RTIME,MXSTPS,MXREIN
COMMON 21 END
COMMON 22
0047 COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0048 REAL*8 TITLE(15)
0049 INTEGER PRTOPT
COMMON 22 END
C
C THIS SR WILL FIND A FEASIBLE POINT, IF ANY.
C ELSE, IT WILL PRINT OUT AND SET IRC = 1
C
C METHODOLOGY:
C DEPENDING UPON THE START OPTION, SET STARTING PT. TO
C BE UNCONS. MIN OR RANDOM PT.
C
C SET JTLINV = LINV
C
C CALL LAMDA
C
C ADD ALL THE CONS. X IS CURRENTLY ON
C
C FOLLOW ROSEN START METHOD, I.E
C A) ADD A VIOLATED CONS. IF LIN. INDEP.
C B) IF L. DEP. THEN FIND A CONS TO DROP BY CALCULATING
C INFEASIBILITY MULTIPLIERS.
C IF ALL IM -VE THEN NO FEAS PT. , ELSE FIND A -VE LM
C CONSTRAINT AND DROP IT AND CONTINUE (A)
C C) UPDATE X = X - NL(*) (T) * LAMDA
```

```

C          GOTC A
C
C          DATE JULY 30 1979
C
C
0050          GOTO(5,6),ISOPT
C          HERE IF WE HAVE UNCONS. MIN. TO START
C
C
0051          5          DO 10 I=1,M
0052          10          X(I)=UNCONX(I)
0053          GOTO 30
C
C          RANDOM START
C
0054          6          DO 20 I=1,M
0055          20          X(I) = RAND(III)*HIXLOX + LUX
C
C          COPY LINV TO QTLINV
C
0056          30          CALL INTLQL
0057          Q=J
0058          CALL LAMDA
0059          DO 70 I=1,K
0060          IF(DABS(XLAMDA(I)).GT.TOLAM)GOTO 70
C
C          WE HAVE A CONSTRAINT X IS ON
C
0061          CALL RBAR(C(1,I),QTLZ,1,M)
0062          CALL PR CJ
0063          IF(LENY.LE.TOLDEP)GOTO 70
0064          CALL ADD(I)
0065          70          CONTINUE
C
C
0066          GOTO 90
C
C
0067          80          CALL RENCHK(IRC)
0068          IF(IRC.EQ.0)GOTO 90
0069          IRC=1
C          NO FEAS PT. FOUND BECAUSE OF REINVERSION WONT HELP
C
0070          GOTO 1000
C
C
C          EVERYTHIN OK
C
C
0071          90          CONTINUE
0072          4          CALL MXMIN(XLAMDA,K,IAOPT,-1,TOLAM,ICONAD,VALUE)
0073          IF(ICONAD.EQ.0)GOTO 130
0074          95          CALL RBAR(C(1,ICONAD),QTLZ,Q+1,M)
0075          IF(Q.EQ.M)GOTO 110
0076          IF(Q.EQ.0)GOTO 98
0077          LENY = SINNER(QTLZ(Q+1),QTLZ(Q+1),M-Q)

```

```

0078          NOPS=NOPS+M-Q
0079          IF (LENY.LE.TOLDEP)GOTO 110
0080          97      CALL RBAR(C(1,ICONAD),QTLZ,1,Q)
0081          98      NSTEPS = NSTEPS + 1
0082          CALL UPRIN(NSTEPS+1000)
C            CALL RBAR(C(1,ICONAD),QTLZ,1,Q)
0083          CALL ADD(ICONAD)
0084          VAL = XLAMDA(ICONAD)/R(Q,Q)
0085          DO 100 I=1,M
0086          100     X(I) = X(I) - QTLINV(Q,I) * VAL
0087          NOPS = NOPS + M + 1
0088          IF(PRTGPT.GT.1)WRITE(6,1202)(X(I),I=1,M)
0089          1202    FORMAT(' ', ' VECTOR X AFTER STEP IS', (/9D14.7))
0090          GOTO 80

C
C      NOW WE MUST DROP A CONSTRAINT
C
0091          110     CALL RBAR(C(1,ICONAD),QTLZ,1,Q)
0092          CALL SOLVR(IM,QTLZ,1,Q)
0093          14      CALL MXMIN(IM,Q,IDOPT ,+1,TOLMUL,ICONDR,VALUE)
0094          IF(ICONDR.EQ.0)GOTO 120
C      THERE IS SOME MULTIPLIER THAT IS +VE
C
0095          CALL DROP(ICONDR)
0096          CALL RBAR(C(1,ICONAD),QTLZ,ICONDR,M)
0097          LENY=SINNER(QTLZ(Q+1),QTLZ(Q+1),M-Q)
0098          NOPS=NOPS+M-Q
0099          IF (LENY.LT.TOLDEP)GOTO 110
0100          GOTO 98
0101          120     WRITE(6,125)
0102          125     FJRMAT(' ',5X, 'NO FEASIBLE PT. FOUND - ALL IM < 0.0')
0103          WRITE(6,126)(IM(J),J=1,Q)
0104          126     FORMAT(' ',5X, 'INFEASIBILITY MULTIPLIERS ARE', (/9D14.7))
0105          IRC = 1
0106          GOTO 1000

C
C      FEAS PT IS FOUND & HENCE STORE EVERYTHING
C
0107          130     DO 140 I = 1,M
0108          FX(I) = X(I)
0109          DO 140 J= 1,M
0110          140     FJTLIN(J,I) = QTLINV(J,I)
0111          FQ = Q
0112          IF(Q.EQ.0) GOTO 160
0113          DO 150 I=1,Q
0114          FNQ(I) = NQ(I)
0115          DO 150 J=1,I
0116          150     FR(J,I) = R(J,I)
0117          160     IRC = 0
0118          1000    CONTINUE
0119          WRITE(IUNIT6,71)Q,(NQ(I),I=1,Q)
0120          71      FORMAT(' ',I7, ' - IN BASIS ',30I3)
0121          RETURN
0122          END

```

```

0001          SUBROUTINE STPLEN(TAU,ICONAD)
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON C1
0003          COMMON /C1/PROJV,LENY,QLZ
0004          REAL * 8 PROJV(50),LENY,QLZ(30)
          COMMON C1 END
          COMMON C5
0005          COMMON /C5/C,8,XLAMDA,ICONOK
0006          REAL*8 C(30,90),3(90),XLAMDA(90)
0007          INTEGER*2 ICONOK(90)
          COMMON C5 END
          COMMON C9
0008          COMMON /C9/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSDKFG,TYPE
0009          1, ISEED,ISOPT,IAOPT,IDOPT
          INTEGER N,P,Q,K,ON,OFF,ISIZ,KINDPR,FSDKFG,TYPE
          COMMON C9 END
          COMMON C12
0010          COMMON /C12/NQ
0011          INTEGER NQ(30)
          COMMON C12 END
          COMMON C15
0012          COMMON /C15/TOLDCH,TOLDEP,TOLINV,TOLPR,TOLAM,TOLML
          COMMON C15 END
          COMMON C21
0013          COMMON /C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NQRT,NRF IN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON C21 END
          COMMON C22
0014          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0015          REAL*8 TITLE(15)
0016          INTEGER PRTOPT
          COMMON C22 END
0017          TAU = 1.D70
0018          ICONAD = 0
0019          DO 20 I=1,K
0020          IF(ICONOK(I).EQ.0)GOTO 20
0021          IF(Q.EQ.0)GOTO 15
          C   CHECK IF IT IS IN THE BASIS
0022          DO 10 J = 1,Q
0023          IF(NQ(J).EQ.1)GOTO 20
0024          10  CONTINUE
0025          15  DENOM = SINER(C(1,I),PROJV,M)
0026          NOPS = NOPS + M
0027          IF(DENOM.LE.TOLDEP)GOTO 20
0028          RAT = XLAMDA(I)/DENOM
0029          IF(XLAMDA(I).GE.-TOLAM.AND.XLAMDA(I).LT.0.D0)RAT=0.D0
0030          NOPS = NOPS + 1
0031          IF(RAT.GE.TAU)GOTO 20
0032          TAU = RAT
0033          ICONAD = I
0034          20  CONTINUE
          C
0035          IF(PRTOPT.GT.1)WRITE(6,125)TAU,ICONAD
0036          125  FJRMAT(' ',5X,'TAU =',E14.9,'  ICONAD = ',I5)
0037          RETURN
0038          END

```

```

0001          SUBROUTINE FLET(IPALG,IRC,IPDPRI)
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON 01
0003          COMMON /C01/PROJV,LENY,QLZ
0004          REAL * 8 PROJV(30),LENY,QLZ(30)
          COMMON 01 END
          COMMON 04
0005          COMMON /C04/Q1R
0006          REAL*8 Q1R(30,32),Q1(30,30),R(30,30)
0007          EQUIVALENCE(Q1R(1,1),Q1(1,1))
0008          EQUIVALENCE(Q1R(1,3),R(1,1))
          COMMON 04 END
          COMMON 05
0009          COMMON /C05/C,B,XLAMDA,ICLNOK
0010          REAL*8 C(30,90),B(90),XLAMDA(90)
0011          INTEGER*2 ICONDK(90)
          COMMON 05 END
          COMMON 09
0012          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,F5OKFG,TYPG
          1, ISEED,ISOPT,IAGPT,IOUPT
0013          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,F5OKFG,TYPG
          COMMON 09 END
          COMMON 11
0014          COMMON /C11/X
0015          REAL*8 X(30)
          COMMON 11 END
          COMMON 13
0016          COMMON /C13/LM,IM
0017          REAL*8 LM(30),IM(30)
          COMMON 13 END
          COMMON 14
0018          COMMON /C14/F,GRAD
0019          REAL*8 F,GRAD(30)
          COMMON 14 END
          COMMON 15
0020          COMMON /C15/TOLDCH,TOLDEP,TOLINV,TOLPR,TOLAM,TOLMUL
          COMMON 15 END
          COMMON 17
0021          COMMON/C17 /IUNIT5,IUNIT6,IUNIT8
          COMMON 17 END
          COMMON 21
0022          COMMON/C21/IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCDR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          COMMON 22
0023          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0024          REAL*8 TITLE(15)
0025          INTEGER PRTOPT
          COMMON 22 END
0026          IRC = 0
0027          CALL LAMDA
0028          7 CALL GRA
0029          9 CALL RBAR(GRAD,QLZ,Q+1,M)
0030          CALL PROJ
0031          IF(LENY.LE.TOLPR )GOTO 70
0032          CALL STPLEN(TAU,ICONAD)

```

```

0033          IF(ICONAD.EQ.0)TAU = 1.00
C          ISTEP = 1 - FULL STEP, 0-PARTIAL STEP
C
0034          ISTEP = 0
0035          IF(TAU.GT.1.0)TAU = 1.00
0036          IF(TAU .EQ.1.00) ISTEP = 1
0037          IF(PRTOPT.GT.1)WRITE(6,121)TAU
0038          121  FORMAT(' ',5X,'STEP LENGTH,TAU=',F14.7)
0039          DO 30 I=1,M
0040          30   X(I) = X(I) - TAU * PROJN(I)
0041          NOPS=NOPS+M
0042          IF(PRTOPT.GT.1)WRITE(6,1202)(X(I),I=1,M)
0043          1202  FORMAT(' ', ' VECTOR X AFTER STEP IS',(/9D14.7))
0044          NSTEPS = NSTEPS + 1
0045          IF(NSTEPS.LT.MXSTPS)GOTO 35
0046          WRITE(6,31)MXSTPS
0047          31   FORMAT(' ',5X,'NO. OF STEPS EXCEED MAX OF',15)
0048          IRC = 1
0049          GOTO 100
0050          35   CONTINUE
0051          IF(ISTEP.EQ.1)GOTO 40
0052          NPSTEP = NPSTEP + 1
0053          CALL UPRIN(-1000-NSTEPS)
0054          CALL RBAR(C(1,ICONAD),QTLZ,1,M)
0055          CALL ADD(ICONAD)
0056          5    CALL RENCHK(IRC)
0057          IF(IRC.EQ.0) GOTO 7
0058          WRITE(IUNIT6,8) IRC
0059          8    FORMAT(' ','*** REINVERSION FAILED, IRC = ',15)
0060          GOTO 100
0061          40   NFSTEP = NFSTEP + 1
0062          CALL UPRIN(1000+NSTEPS)
0063          CALL GRA
0064          70   CONTINUE
0065          CALL RBAR(GRAD,QTLZ,1,Q)
0066          CALL SOLVR(LM,QTLZ,1,Q)
0067          4    CALL MXMIN(LM,Q,IDOPT ,-1,TOLMUL,ICONDR,VALUE)
0068          IF(ICJNDR.EQ.0)GOTO 100
0069          IF(IPALG.EQ.1)GOTO 80
0070          50   CALL STPEDG(ICONDR,DIAG)
0071          IF(ICONDR.EQ.0)GOTO 100
0072          80   CONTINUE
0073          CALL DRCP(ICONDR)
0074          CALL LAMDA
0075          GOTO 9
0076          100  CALL LAMDA
0077          CALL GRA
0078          RETURN
0079          END

```

```

0001          SUBROUTINE NSGNST(I,DIAG)
0002          IMPLICIT REAL*8(A-H,O-Z)
0003          REAL*8 TEMP(30)

COMMON C4
0004          COMMON /C04/Q1R
0005          REAL*8 Q1R(30,32),Q1(30,30),R(30,30)
0006          EQUIVALENCE(Q1R(1,1),Q1(1,1))
0007          EQUIVALENCE(Q1R(1,3),R(1,1))

COMMON 09
0008          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
0009          1, ISEED,ISOPT,IAOPT,IDOPT
          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG

COMMON 09 END
COMMON 21
0010          COMMON/C21/IAG,NSTEPS,NAUDS,NDROPS,NOPS,NSORT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
COMMON 21 END
COMMON 22
0011          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0012          REAL*8 TITLE(15)
0013          INTEGER PRTOPT

COMMON 22 END
0014          TEMP(I) = 1.00/R(I,I)
0015          IP1 = I+1
0016          IF(I.EQ.Q)GOTO 65
0017          DO 60 J=IP1,Q
0018              NOPS = NOPS + J - I + 1
0019          60 TEMP(J) = -SINNER(R(I,J),TEMP(I),J-I)/R(J,J)
0020          65 DIAG = SINNER(TEMP(I),TEMP(I),Q-I+1)
0021              NOPS = NOPS + Q - I + 2
0022              IF(PRTOPT.GT.1)WRITE(6,67)I,DIAG
0023          67 FORMAT(' ',5X,'NSGNST - I,DIAG',I5,D20.10)
0024          RETURN
0025          END

```

```

0001          SUBROUTINE STPEDG(ICONDR,DIAGMX)
0002          IMPLICIT REAL*8(A-H,O-Z)
C
C THIS SR CAN BE USED TO FIND MAX LM(I)**2/STPEDG
C FOR STEEPEST EDGE
C I.E. FIND CONSTRAINT TO DROP FOR STEEPEST EDGE
C ICONDNR = 0 NONE AVAILABLE TO DROP
C ELSE, DROP ICONDNR CONSTRAINT
C
C METHOD:
C CHOOSE AMONG -VE LM
C FOR THE CORRESPONDING CONS. FIND CORRESPONDING ROW IN
C R(-1)
C FIND INNER PRODUCT OF THAT TO GET STPEDG
C NOW FIND THE MAX OF LM(I)**2/STPEDG
COMMON C9
0003          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
0004          1, ISEED,ISOPT,IAOPT,IDOPT
          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
COMMON C9 END
COMMON 13
0005          COMMON /C13/LM,IM
0006          REAL*8 LM(30),IM(30)
COMMON 13 END
COMMON 15
0007          COMMON /C15/TOLDCM,TOLDEP,TOLINV,TOLPR,TOLAM,TOLMUL
COMMON 15 END
COMMON 21
0008          COMMON/C21/IAG,NSTEPS,NAGDS,NDROPS,NOPS,NSQRT,NREIN,NXCDR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
COMMON 21 END
COMMON 22
0009          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0010          REAL*8 TITLE(15)
0011          INTEGER PRTOPT
COMMON 22 END
0012          50 RATMAX = -.070
0013          ICONDNR = 0
0014          DIAGMX = 0.00
0015          IF(J.EQ.0)GOTO 1000
0016          DO 70 I=1,J
0017          IF(LM(I).GT.TOLMUL)GOTO 70
0018          CALL NSGNST(I,DIAG)
0019          RAT = LM(I)*LM(I)/DIAG
0020          NOPS = NOPS + 2
0021          IF(RAT.LT.RATMAX)GOTO 70
0022          RATMAX = RAT
0023          DIAGMX = DIAG
0024          ICONDNR = I
0025          70 CONTINUE
0026          IF (PRTOPT.GT.1)WRITE (6,67)ICONDNR,DIAGMX
0027          67 FORMAT(' ',5X,'STPEDG - ICONDNR,DIAGMX',15,D20.10)
0028          1000 RETURN
0029          END

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0001          SUBROUTINE DUAL(IRC)
0002          IMPLICIT REAL*8(A-H,O-Z)
0003          REAL*8 UPF

COMMON C1
0004          COMMON /C01/PROJV,LENY,QLZ
0005          REAL * 8 PROJV(30),LENY,QLZ(30)
COMMON C1 END
COMMON C5
0006          COMMON /C05/C,B,XLAMDA,ICONOK
0007          REAL*8 C(30,90),3(90),XLAMDA(90)
0008          INTEGER*2 ICONOK(90)
COMMON C5 END
COMMON C8
0009          COMMON /C08/ FXLINR
0010          REAL*8 FXLINR(30,32),FX(30),LINV(30,30),FR(30,30)
0011          EQUIVALENCE(FXLINR(1,1),FX(1))
0012          EQUIVALENCE(FXLINR(1,2),LINV(1,1))
0013          EQUIVALENCE(FXLINR(1,3),FR(1,1))
COMMON C8 END
COMMON C9
0014          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
0015          1, ISEED,ISOPT,IAOPT,IDOPT
          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
COMMON C9 END
COMMON C10
0016          COMMON /C10 /JNCONX,UNCONF
0017          REAL*8 UNCONX(30),UNCONF
COMMON C10 END
COMMON C11
0018          COMMON /C11/X
0019          REAL*8 X(30)
COMMON C11 END
COMMON C13
0020          COMMON /C13/LM,IM
0021          REAL*8 LM(30),IM(30)
COMMON C13 END
COMMON C14
0022          COMMON /C14/F,GRAD
0023          REAL*8 F,GRAD(30)
COMMON C14 END
COMMON C15
0024          COMMON /C15/TOLDCH,TOLDEP,TOLINV,TOLPR,TOLAM,TOLMUL
COMMON C15 END
COMMON C17
0025          COMMON /C17 /IUNIT5,IUNIT6,IUNIT8
COMMON C17 END
COMMON C21
0026          COMMON /C21/ IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
COMMON C21 END
COMMON C22
0027          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0028          REAL*8 TITLE(15)
0029          INTEGER PRTOPT
COMMON C22 END
C THIS SR IS FOR DUAL ALGORITHM.

```

```

C   IRC IS THE RETURN CODE - 0 -EVERYTHING IS OK
C   1 MEANS COULD NOT SOLVE THE DUAL PROBLEM OR NO FEASIBLE
C   POINT EXISTS.
C   METHODOLOGY:
C   1) FIND A STARTING OPTIMAL POINT (UNCONS MIN)
C   2) FIND A VIOLATED CONS TO ADD( IAOPT)
C   3) CHECK IF THE CONS IS LIN. DEP.
C   4) IF IT IS L.DEP. FIND A CONS TO DROP
C   5) GO TO 3
C   6) IF IT IS L.IND. FIND STEP LENGTH- FULL AND PARTIAL
C   7) TAKE STEP
C   8) IF IT IS FULL STEP, ADD THE CONS. AND GOTO 2)
C   9) IF PARTIAL, DROP THE CONS. WHICH FORCED A PARTIAL STEP
C   10) GOTO 6.
C
C   DATE AUGUST 2,1979
C
0030      IRC = 0
C
C   COPY THE UNCONS. MIN
C
0031      F = UNCONF
0032      DO 10 I=1,M
0033      10  X(I) = UNCONX(I)
0034          CALL INTLQL
C   SET Q TO 0
0035          Q = 0
0036      40  CALL RENCHK(IRC)
0037          IF(IRC.NE.0)GOTO 1000
0038      45  CALL MXMIN(XLAMDA,K,IAOPT,-1,TOLAM,ICONAD,VALUE)
0039          IF(Q.NE.M)LM(Q+1)=C.DO
0040      50  CONTINUE
0041          IF(ICONAD.EQ.0)GOTO 900
0042      60  CALL RBAR(C(1,ICONAD),QTLZ,1,M)
0043          CALL PROJ
0044          IF(Q.EQ.0)GOTO 110
C   CALL MULT(GRAD,LM,1)
C   CALL MULT(C(1,ICONAD),1M,1)
0045          CALL SOLVR(IM,QTLZ,1,Q)
0046          IF(LENY.GT.TOLDEP)GOTO 90
0047          CALL DSTPLN(T1,ICONDR)
0048          IF(ICONDR.NE.0)GOTO 80
0049          WRITE(IUNIT6,71)
0050      71  FORMAT(' ','***** NO FEASIBLE POINT EXISTS *****')
0051          IRC = 1
0052          GOTO 1000
0053      80  CONTINUE
0054          CALL DROP(ICONDR)
0055          DO 85 I=1,Q
0056      85  LM(I) = LM(I) - T1 * IM(I)
0057          LM(Q+1) = T1
0058          NDPS = NDPS + Q
0059          GOTO 60
0060      90  IF(Q.EQ.0)GOTO 110
0061          CALL DSTPLN(T1,ICONDR)
0062      110  T3 = SINNER(QTLZ(Q+1),QTLZ(Q+1),M-Q)

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0063      T2 = -XLAMDA(ICONAD)/T3
0064      NQPS = NQPS + M - Q + 1
0065      T=T2
0066      ISTEP = 1
0067      IF(Q.EQ.0.OR.ICONDR.EQ.0)GOTO 115
0068      IF(T1.GE.T)GOTO 115
0069      T=T1
0070      ISTEP = 0
C      ISTEP = 1 FULL STEP, ADD
C      ISTEP = 0 PARTIAL STEP, DROP
C
0071      115      CONTINUE
0072      DO 120 I=1,M
0073      120      X(I) = X(I) + T * FROJV(I)
0074      F = F + T * T3 *(LM(Q+1) + 0.5 * T)
0075      IF(PRTOPT.GT.1)WRITE(6,701)T1,T2,T
0076      701      FORMAT(' ',5X,'T1,T2,T=',3E16.9)
0077      IF(PRTOPT.GT.1) WRITE(6,702)(X(I),I=1,M)
0078      702      FORMAT(' ',5X,'X VECTOR',(7D14.7))
0079      IF(Q.EQ.0)GOTO 705
0080      DO 703 I=1,Q
0081      703      LM(I)=LM(I) - IM(I) * T
0082      IF(PRTOPT.GT.2)WRITE(6,704)(LM(I),I=1,Q)
0083      704      FORMAT(' ',5X,'PREDICTED LM',(7D14.7))
0084      705      CONTINUE
0085      LM(Q+1) = LM(Q+1) + T
0086      NQPS = NQPS + M + Q + 3
0087      NSTEPS = NSTEPS + 1
0088      IF(NSTEPS.LT.MXSTPS)GOTO 135
0089      WRITE(6,131)MXSTPS
0090      131      FORMAT(' ',5X,'NO. OF STEPS EXCEED MAX OF',I5)
0091      IRC = 1
0092      GOTO 1000
0093      135      CONTINUE
0094      IF(ISTEP.EQ.0)GOTO 130
0095      CALL UPRIN(1000+NSTEPS)
0096      CALL ADD(ICONAD)
0097      NFSTEP = NFSTEP + 1
0098      GOTO 40
0099      130      CALL UPRIN(-1000-NSTEPS)
0100      CALL DROP(ICONDR)
0101      LM(Q+1) = LM(Q+2)
0102      XLAMDA(ICONAD) = SINNER(C(1,ICONAD),X,M) - B(ICONAD)
0103      NQPS = NQPS + M
0104      NPSTEP = NPSTEP + 1
0105      GOTO 60
0106      900      UPF = F
0107      CALL GRA
0108      NQPS = NQPS - 2*M*(M+2)-1
0109      IF(DABS(UPF-F).GT.TOLAM)WRITE(6,909)F,UPF
0110      909      FORMAT(' ',5X,'F , UPF ARE DIFFERENT',2F16.7)
0111      1000     RETURN
0112      END

```

```

0001          SUBROUTINE DSTPLN(T1,ICONDR)
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON 09
0003          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAOPT,IDOPT
0004          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON 09 END
          COMMON 13
0005          COMMON /C13/LM,IM
0006          REAL*8 LM(30),IM(30)
          COMMON 13 END
          COMMON 15
0007          COMMON /C15/TOLDCM,TOLDEP,TOLINV,TOLPR,TOLAM,TOLMUL
          COMMON 15 END
          COMMON 21
0008          COMMON /C21/IAG,NSTEPS,NADDS,NDRDPS,NOPS,NSQRT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          COMMON 22
0009          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTGPT
0010          REAL*8 TITLE(15)
0011          INTEGER PRTOPT
          COMMON 22 END
0012          ICONDR = 0
0013          T1 = 1.070
0014          IF(N.EQ.0)          RETURN
0015          DO 70 I=1,Q
0016          IF(IM(I).LE.TOLMUL)GOTO 70
0017          IF(LM(I).LT.0.00)LM(I)=0.00
0018          RAT = LM(I)/IM(I)
0019          NOPS = NOPS + 1
0020          IF(RAT.GT.T1)GOTO 70
0021          T1 = RAT
0022          ICONDR = I
0023          70 CONTINUE
0024          IF(PRTOPT.GT.1)WRITE(6,71)T1,ICONDR
0025          71 FORMAT(' ',5X,'DSTPLN - T1,ICONDR',E20.10,I9)
0026          RETURN
0027          END

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```

0001          SUBROUTINE UPRIN(ICON)
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON 19
0003          COMMON /C19/IPRIN,IPTR,IPMAX,NBRK
0004          INTEGER*2 IPRIN(5000),IPTR,IPMAX,NBRK(15)
          COMMON 19 END
0005          IF(IPTR.GE.IPMAX)GOTO 10
0006          IPTR=IPTR+1
0007          IPRIN(IPTR)=ICON
0008          RETURN
0009          10 WRITE(6,11)IPMAX
0010          11 FORMAT(' ',5X,'PTR HAS REACHED MAX OF ',I5)
0011          RETURN
0012          END

```

```

0001          SUBROUTINE PRIMAL (IALG, IRC, IPDPRI)
0002          IMPLICIT REAL*8 (A-H, O-Z)

COMMON 03
0003          COMMON /C03 /QTLINV
0004          REAL*8 QTLINV(30,30)
COMMON 03 END
COMMON 04
0005          COMMON /C04/Q1R
0006          REAL*8 Q1R(30,32), Q1(30,30), R(30,30)
0007          EQUIVALENCE(Q1R(1,1), Q1(1,1))
0008          EQUIVALENCE(Q1R(1,3), R(1,1))
COMMON 04 END
COMMON 05
0009          COMMON /C05/C, B, XLAMDA, ICONOK
0010          REAL*8 C(30,90), B(90), XLAMDA(90)
0011          INTEGER*2 ICONOK(90)
COMMON 05 END
COMMON 06
0012          COMMON /C06/FF, FNQ, FQ
0013          INTEGER FNQ(30), FQ
0014          REAL*8 FF
COMMON 06 END
COMMON 07
0015          COMMON /C07 /FQTLIN
0016          REAL*8 FQTLIN(30,30)
COMMON 07 END
COMMON 08
0017          COMMON /C08 /FXLINR
0018          REAL*8 FXLINR(30,32), FX(30), LINV(30,30), FR(30,30)
0019          EQUIVALENCE(FXLINR(1,1), FX(1))
0020          EQUIVALENCE(FXLINR(1,2), LINV(1,1))
0021          EQUIVALENCE(FXLINR(1,3), FR(1,1))
COMMON 08 END
COMMON 09
0022          COMMON /C09/M, P, Q, K, DN, OFF, ISIZ, KINDPR, FSOCKFG, TYPG
0023          1, ISEED, ISOPT, IAOPT, IDOPT
          INTEGER M, P, Q, K, DN, OFF, ISIZ, KINDPR, FSOCKFG, TYPG
COMMON 09 END
COMMON 11
0024          COMMON /C11/X
0025          REAL*8 X(30)
COMMON 11 END
COMMON 12
0026          COMMON /C12/NQ
0027          INTEGER NQ(30)
COMMON 12 END
C          THIS SR IS FOR PRIMAL AND PRIMAL PART OF PRIMAL DUAL
C          THE FUNCTION OF THIS SR IS TO INVOKE APPROPRIATE PRIMAL
C          ROUTINE AND SET UP NECESSARY STUFF FOR PRIMAL DUAL
C
C          IALG IS THE PRIMAL ALGORITHM TO BE RUN
C          1 - FLETCHER, 2 - FLETCHER WITH STEEPEST EDGE
C          3 - GOLDFARB, 4 - GOLDFARB WITH STEEPEST EDGE
C          5 - GOLDFARB S.T. IT WONT VIOLATE ANY CONS. ALREDY DROPPED
C          6 - GOLDFARB COMBO OF 4 & 5
C

```

```

C   IRC IS THE RETURN CODE , 0 - EVERYTHING OK,1 -OTHERWISE
C   IPDPRI IS 0 - PRIMAL, 1 - PRIMAL DUAL
C
0026   IRC = 0
0029   IF(IPDPRI.EQ.1)GOTO 40
0030   DO 10 I=1,K
0031     10   ICONDK(I) = 1
0032     DO 20 I = 1,M
0033     X(I) = FX(I)
0034     DO 20 J=1,M
0035     20   QTLINV(J,I) = FQTLIN(J,I)
0036     F = FF
0037     Q = FQ
0038     IF(Q.EQ.0)GOTO 70
0039     DO 30 I=1,Q
0040     NQ(I)=FNQ(I)
0041     DO 30 J=1,I
0042     R(J,I) = FR(J,I)
0043     30   CONTINUE
0044     GOTO 70
0045     40   CALL UPRIN(2000)
0046     DO 50 I=1,K
0047     50   ICONDK(I) = 0
0048     IF(Q.EQ.0)GOTO 70
0049     DO 60 I=1,Q
0050     60   ICONDK(NQ(I)) = 1
0051     70   GOTO(1,1,3,3,3,3),IALG
0052     1   CALL FLET(IALG,IRC,IPDPRI)
0053     GOTO 100
0054     3   IPALG = IALG - 2
0055     CALL GOLD(IPALG,IRC,IPLPRI)
0056     100  IF(IPDPRI.EQ.1)CALL UPRIN(-2000)
0057     RETURN
0058     END

```

```

0001      SUBROUTINE ACTCHK(IRC)
0002      IMPLICIT REAL*8(A-H,O-Z)
          COMMON 05
0003      COMMON /C05/C,B,XLAMDA,ICONOK
0004      REAL*8 C(30,90),B(90),XLAMDA(90)
0005      INTEGER*2 ICONOK(90)
          COMMON 05 END
          COMMON 09
0006      COMMON /C09/H,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAOPT,IDOPT
0007      INTEGER H,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON 09 END
          COMMON 15
0008      COMMON /C15/TOLDCK,TOLDEP,TOLINV,TOLPR,TOLAM,TOLMUL
          COMMON 15 END
          COMMON 12
0009      COMMON /C12/NQ
0010      INTEGER NQ(30)
          COMMON 12 END
          COMMON 22
0011      COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0012      REAL*8 TITLE(15)
0013      INTEGER PRTOPT
          COMMON 22 END
C        THIS SR WILL CHECK IF ACTIVE CONSTRAINTS ARE SATISFIED
C        IRC IS THE RETURN CODE , SET TO 0 IF ALL ACTIVE CONS. ARE
C        SATISFIED AT X, ELSE IT IS SET TO THE CONSTRAINT NUMBER
C        THAT IS NOT SATISFIED
C
0014      IRC = 0
0015      IF(Q.EQ.0)RETURN
0016      DO 10 I=1,Q
0017      IF(DABS(XLAMDA(NQ(I))).LE.TOLAM)GOTO 10
0018      IRC = I
0019      WRITE(6,11)NQ(I),XLAMDA(NQ(I)),TOLAM
0020      11  FORMAT(' ',5X,'ACT. CONS',15,' LAMDA - ',F16.7,' TOL',F16.7)
0021      RETURN
0022      10  CONTINUE
0023      RETURN
0024      END

```

```

0001          SUBROUTINE RENCHK(IRC)
0002          IMPLICIT REAL*8(A-H,O-Z)
0003          COMMON 17
          COMMON/C17/IUNIT5,IUNIT6,IUNIT8
          COMMON 17 END
C          IRC = 0 RETURN CODE 0 - EVERYTHING OK
C          != 0 THEN A CONSTRAINT IS STILL VIOLATED AND REINVERSION
C          CANNOT HELP
C
C          METHOD:
C          CALL ACTCHK
C          IF EVERYTHING OK RETURN
C          ELSE, PERFORM XCOR
C          CALL LAMDA
C          CALL ACTCHK
C          IF EVERYTHING OK RETURN
C          ELSE, PERFORM REIN
C          CALL XCOR
C          CALL ACTCHK
C          IF EVERYTHING OK, RETURN
C          ELSE, RETURN, REINVERSION FAILED
C
0004          CALL LAMDA
0005          CALL ACTCHK(IRC)
0006          IF(IRC.EQ.0)GOTO 45
0007          CALL XCOR
0008          CALL LAMDA
0009          CALL ACTCHK(IRC)
0010          IF(IRC.EQ.0)GOTO 45
0011          CALL REIN (IRC)
0012          IF(IRC.NE.0)GOTO 45
0013          CALL XCOR
0014          CALL LAMDA
0015          CALL ACTCHK(IRC)
0016          IF(IRC.EQ.0) GOTO 45
0017          WRITE(IUNIT6,201)IRC
0018          201 FORMAT(' ', '****REINVERSION FAILED -IRC = ',I5)
0019          45 RETURN
0020          END

```

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0001          SUBROUTINE PRDUAL (IALGO,IRC,IPDPRI)
0002          IMPLICIT REAL*8(A-H,U-Z)
          COMMON C1
0003          COMMON /C01/PROJV,LENY,QLZ
0004          REAL * 8 PROJV(30),LENY,QLZ(30)
          COMMON 01 END
          COMMON 05
0005          COMMON /C05/C,B,XLAMDA,ICONOK
0006          REAL*8 C(30,90),B(90),XLAMDA(90)
0007          INTEGER*2 ICONOK(90)
          COMMON 05 END
          COMMON 09
0008          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAOPT,IDOPT
0009          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON 09 END
          COMMON 10
0010          COMMON /C10/UNCONX,UNCONF
0011          REAL*8 UNCONX(30),UNCONF
          COMMON 10 END
          COMMON 11
0012          COMMON /C11/X
0013          REAL*8 X(30)
          COMMON 11 END
          COMMON 13
0014          COMMON /C13/LM,IM
0015          REAL*8 LM(30),IM(30)
          COMMON 13 END
          COMMON 14
0016          COMMON /C14/F,GRAD
0017          REAL*8 F,GRAD(30)
          COMMON 14 END
          COMMON 15
0018          COMMON /C15/TJLOCH,TOLDEP,TOLINV,TOLPR,TOLAM,TOLMUL
          COMMON 15 END
          COMMON 17
0019          COMMON/C17/IUNIT5,IUNIT6,IUNIT8
          COMMON 17 END
          COMMON 21
0020          COMMON/C21/IAG,NSTEPS,NAUDD,NDROPS,NOP,NSQRT,NREIN,NXCCR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          COMMON 22
0021          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0022          REAL*8 TITLE(15)
0023          INTEGER PRTOPT
          COMMON 22 END
0024          IRC = 0
0025          F = UNCONF
0026          DO 10 I=1,M
0027          X(I) = UNCONX(I)
0028          CALL INTLQL
0029          Q = 0
0030          40 CALL LAMDA
0031          45 CALL MXMIN(XLAMDA,K,IAOPT,-1,TOLAM,ICONAD,VALUE)
0032          IF (ICONAD.EQ.0)GOTO 1000

```

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0033          IF(Q.EQ.M)GOTO 105
C           50  CALC PROJ. OF NORMAL FOR ICONAD
0034          CALL RBAR(C(1,ICONAD),QTLZ,Q+1,M)
0035          CALL PROJ
0036          IF(LENY.LE.TOLPR)GOTO 100
C           60  CALC FULL DUAL STEP LENGTH
0037          T3 = SINNER(QTLZ(Q+1),QTLZ(Q+1),M-Q)
0038          NOPS=NOPS+M-Q
0039          T2 = -XLAMDA(ICONAD)/T3
0040          IF(PRTOPT.GT.1)WRITE(6,121)T2
0041          121  FORMAT(' ',5X,'FULL STEP , T2 =',F14.7)
0042          DO 60 I=1,M
0043          60  X(I)=X(I) + T2 * PROJV(I)
0044          IF(PRTOPT.GT.1)WRITE(6,1202)(X(I),I=1,M)
0045          1202 FORMAT(' ', ' VECTOR X AFTER STEP IS',(/9D14.7))
0046          NUPS = NOPS + M + 1
0047          NSTEPS = NSTEPS + 1
0048          NFSTEP = NFSTEP + 1
0049          CALL UPRIN(1000+NSTEPS)
0050          CALL RBAR(C(1,ICONAD),QTLZ,1,Q)
0051          CALL ADD(ICONAD)
0052          CALL RENCHK(IRC)
0053          IF(IRC.NE.0)GOTO 1000
0054          CALL GRA
C           70  CALC LAGRANGE MULTIPLIERS TO SEE IF PRIMAL TO BE CALLED
0055          CALL RBAR(GRAD,QTLZ,1,Q)
0056          CALL SOLVR(LM,QTLZ,1,Q)
0057          CALL MXMIN(LM,Q,IDOPT,-1,TOLMUL,ICONDR,VALUE)
0058          IF(ICONDR.EQ.0)GOTO 45
0059          CALL PRIMAL(IALGL,IRC,IPDPRI)
0060          IF(IRC.EQ.0)GOTO 40
0061          WRITE(6,70)IRC
0062          70  FORMAT(' ',5X,'*** PRIMAL FAILED - IRC=',I5)
0063          GOTO 1000
0064          100  CALL GRA
C           105  CALC LM & IM - DROP THE ONE CORRESPONDING TO THE MIN RATIO
0065          CALL RBAR(GRAD,QTLZ,1,Q)
0066          CALL SOLVR(LM,QTLZ,1,Q)
0067          105  CALL RBAR(C(1,ICONAD),QTLZ,1,Q)
0068          CALL SOLVR(IM,QTLZ,1,Q)
0069          CALL DSTPLN(T1,ICONDR)
0070          IF(ICONDR.EQ.0)GOTO 120
0071          CALL DROP(ICONDR)
0072          GOTO 50
0073          120  WRITE(6,130)
0074          130  FORMAT(' ',5X,'NO FEAS PT EXISTS -PD')
0075          1300  CALL GRA
0076          RETURN
0077          END

```

```

0001          SUBROUTINE GOLD(JPRIM,IRC,IPDPRI)
0002          IMPLICIT REAL*8(A-H,O-Z)
          COMMON 01
0003          COMMON /C01/PROJV,LENY,QLZ
0004          REAL * 8 PROJV(30),LENY,QLZ(30)
          COMMON 02 END
          COMMON 05
0005          COMMON /C05/C,B,XLAMDA,ICONDK
0006          REAL*8 C(30,90),B(90),XLAMDA(90)
0007          INTEGER*2 ICONDK(90)
          COMMON 05 END
          COMMON 09
0008          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          1, ISEED,ISOPT,IAOPT,IDOPT
0009          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,FSOKFG,TYPG
          COMMON 09 END
          COMMON 11
0010          COMMON /C11/X
0011          REAL*8 X(30)
          COMMON 11 END
          COMMON 12
0012          COMMON /C12/NQ
0013          INTEGER NQ(30)
          COMMON 12 END
          COMMON 13
0014          COMMON /C13/LM,IM
0015          REAL*8 LM(30),IM(30)
          COMMON 13 END
          COMMON 14
0016          COMMON /C14/F,GRAD
0017          REAL*8 F,GRAD(30)
          COMMON 14 END
          COMMON 15
0018          COMMON /C15/TOLDCH,TOLEDP,TOLINV,TOLPR,TOLAH,TOLMUL
          COMMON 15 END
          COMMON 17
0019          COMMON/C17 /IUNIT5,IUNIT6,IUNIT8
          COMMON 17 END
          COMMON 21
0020          COMMON/C21/IAG,NSTEPS,KADDS,NDROPS,NDPS,NSQRT,NREIN,NXCOR
          1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
          COMMON 21 END
          COMMON 22
0021          COMMON /C22/ TITLE,NPAGE,NLINE,MXLINE,PRTOPT
0022          REAL*8 TITLE(15)
0023          INTEGER PRTOPT
          COMMON 22 END
0024          REAL*8 B1(30),LM1(30),B2(30),LM2(30)
0025          INTEGER CM1,NNG(30),NG(30),QP1
0026          1 CALL LAMDA
0027          2 CALL GRA
0028          4 NDROPH = 0
0029          IF(Q.EQ.0)GOTO 200
0030          DO 5 I=1,Q
0031          5 NG(I) = 1
0032          CALL RBAR(GRAD,QLZ,1,Q)

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0033          CALL SOLVR(LM, QTLZ, 1, Q)
0034          6  CALL MXMIN(LM, Q, IDOPT, -1, TOLMUL, ICONDR, VALUE)
0035          IF(ICONDR.EQ.0)GOTO 200
0036          15  GOTO(40,10,20,10),JPRIM
0037          10  CALL STPEDG(ICONDR,DIAG)
0038             IF(ICONDR.EQ.0)GOTO 95
0039          17  GOTO(40,40,20,30),JPRIM
0040          20  CALL NSGNST(ICONDR,DIAG)
0041          30  BEE = LM(ICONDR) / DIAG
0042             NOPS = NOPS + 1
0043             B1(NDROPH + 1) = BEE
0044             IF(NDROPH.EQ.0)GOTO 40

C
C  UPDATE BEES
C
0045          DO 35 I=1,NDROPH
C          CALCULATE IM FOR DROPPED CONSTRAINTS
0046             CALL RBAR(C(1,NNG(I)),QTLZ,ICONDR,Q)
0047             CALL SOLVR(IM,QTLZ,ICONDR,Q)
C          FIND N(ICONDR)≠ N(NNG(I))
0048          35  B1(I) = B1(I) + IM(ICONDR) * BEE
0049             NOPS = NOPS + NDROPH
0050          40  NDROPH = NDROPH + 1
0051             B1(NDROPH) = BEE
0052             NNG(NDROPH) = NQ(ICONDR)
0053             QM1 = Q-1
0054             IF(QM1.EQ.0)GOTO 65
0055             DO 60 I=1,QM1
0056             IF(I.LT.ICONDR)GOTO 50
0057             LM1(I)=LM(I+1)
0058             NG(I) = NG(I+1)
0059             GOTO 60
0060          50  LM1(I) = LM(I)
0061          60  CONTINUE
0062          65  CONTINUE
0063             CALL DRUP(ICONDR)
0064             IF(Q.EQ.0)GOTO 200
0065             CALL RBAR(GRAD,QTLZ,1,Q)
0066             CALL SOLVR(LM,QTLZ,1,Q)
0067             DO 70 I=1,Q
0068             LM2(I)=LM(I)
0069             IF(NG(I).EQ.0)GOTO 67
C          CHECK IF LM IS LESS THAN BEFORE
0070          IF(LM(I).LT.LM1(I))GOTO 70
0071          63  NG(I)=0
0072          67  LM(I)=1.D70
0073          70  CONTINUE
0074             CALL MXMIN(LM,Q, IDOPT, -1, TOLMUL, ICONDR, VALUE)
0075             IF(ICONDR.GT.0)GOTO 15
0076          95  GOTO(200,200,100,100),JPRIM
0077          100 IF(Q.EQ.0) GOTO 200
0078             DO 103 I=1,Q
0079             LM(I)=LM2(I)
0080             GOTO 110
0081          105 CALL RBAR(GRAD,QTLZ,1,Q)
0082             CALL SOLVR(LM,QTLZ,1,Q)

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FORTRAN IV C LEVEL 21                GOLD                DATE = 80105

0083      110  CALL MXMIN(LM,Q,IDOPT,-1,TOLMUL,ICONDR,VALUE)
0084      IF(ICONDR.EQ.0)GOTO 200
0085      115  CALL NSGNST(ICONDR,DIAG)
0086      BEE = LM(ICONDR) / DIAG
0087      NUPS = NOPS + 1
0088      DO 120 I=1,NDROPH
0089      CALL RBAR(C(1,NNG(I)),QTLZ,ICONDR,Q)
0090      CALL SGLVR(IM,QTLZ,ICONDR,Q)
0091      B2(I) = B1(I)*BEE + IM(ICONDR)
0092      NQPS = NOPS + 1
0093      IF(B2(I).LE.TOLMUL)GOTO 120
0094      IF(PRTOPT.GT.2)WRITE(6,112)ICONDR,NNG(I),IM(ICONDR),B2(I),B1(I)
          1,BEE
0095      112  FORMAT(' ', 'CANT DROP', I5, 'TH INDEX CONS-WILL VIOL', I5,
          1*IM,B2,B1,BEE',4D16.7)
C
C THIS WILL VIOLATE ITH CONSTKAIN, THEREFORE DONT DROP THIS ONE
C
0096      LM(ICONDR) = 1.070
0097      GOTO 110
0098      120  CONTINUE
C
C DRCPING THIS WILL NOT VIOLATE A PREVIOUSLY DROPPED CONS,
C HENCE, IT IS OK TO DROP
C
0099      DO 130 I=1,NDROPH
0100      130  B1(I)=B2(I)
0101      NDROPH = NDROPH + 1
0102      NNG(NDROPH)=NNG(ICONDR)
0103      CALL DROP(ICONDR)
0104      IF(Q.GT.0)GOTO 105
0105      200  QP1=Q+1
0106      CALL RBAR(GRAD,QTLZ,QP1,M)
0107      CALL PROJ
0108      IF(LENY.LE.TOLPR)GOTO 275
0109      CALL STPLEN(TAU,ICONAD)
0110      ISTEP = 1
0111      IF(TAU.GT.1.00)TAU=1.00
0112      IF(TAU.EQ.1.00)ISTEP = 0
0113      IF(PRTOPT.GT.1)WRITE(6,121)TAU
0114      121  FORMAT(' ', 5X, 'STEP LENGTH,TAU=', F14.7)
0115      DO 210 I=1,M
0116      210  X(I)=X(I) - TAU * PROJV(I)
0117      IF(PRTOPT.GT.1)WRITE(6,1202)(X(I),I=1,M)
0118      1202  FORMAT(' ', ' VECTOR X AFTER STEP IS', (/9D14.7))
0119      NJPS = NOPS + M
0120      NSTEPS=NSTEPS + 1
0121      IF(NSTEPS.LT.MXSTPS)GOTO 215
0122      WRITE(IUNIT6,212)MXSTPS
0123      212  FORMAT(' ', 5X, 'MAX STEPS EXCEEDED', I7)
0124      IRC = 1
0125      GOTO 300
0126      215  IF(ISTEP.EQ.1)GOTO 220
0127      NFSTEP = NFSTEP + 1
0128      CALL UPRIN(1000*NSTEPS)
0129      IF(NDROPH.GT.0)GOTO 1

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```
0130      CALL GRA
0131      275      CONTINUE
0132      GOTO 300
0133      220      NPSTEP = NPSTEP + 1
0134      CALL UPRIN(-1000-NSTEPS)
0135      CALL RBAR(C(1,ICONAD),GTLZ,1,M)
0136      CALL ADD(ICONAD)
0137      235      CALL RENCHK(IRC)
0138      IF(IRC.EQ.0)GOTO 2
0139      300      CONTINUE
0140      RETURN
0141      END
```

```

0001          SUBROUTINE CHECK
0002          IMPLICIT REAL*8(A-H,O-Z)
C THIS ROUTINE IS MODIFIED ON 8/9/79 .
C MAJOR CHANGES ARE THAT IT RETURNS MFLAG AS A SIGNED 5 DIGIT
C AS S D1 D2 D3 D4 D5
C S = - IF THE ALG. DID NOT GET THE CORRECT FUNC. VAL. AT DPT
C = BLANK OR + OTHERWISE
C D1 = 1 ALWAYS
C D2 = 0 X IS CORRECT AT OPT. 1 OTHERWISE
C D3 = 0, NO. OF CONS. IN FINAL BASIS CORRECT, 1- INCORRECT
C D4 = 0, FINAL BASIS IS DIFFERENT FROM CORRECT ONE, 1-INCORRE
C D5 = 0, ABS(LAMDA) IS < EPS FOR CORRECT FINAL BASIS, 1-OTHERW
0003          INTEGER*2 LOGIC(90)
C LOGIC ARRAY IS TEMPORARY FOR THIS RTN. ONLY
COMMON 05
0004          COMMON /C05/C,B,XLAMDA,ICONOK
0005          REAL*8 C(30,90),B(90),XLAMDA(90)
0006          INTEGER*2 ICONOK(90)
COMMON 05 END
COMMON 09
0007          COMMON /C09/M,P,Q,K,ON,OFF,ISIZ,KINDPR,F5OKFG,TYPG
1, ISEED,ISOPT,IAOPT,IDOPT
0008          INTEGER M,P,Q,K,ON,OFF,ISIZ,KINDPR,F5OKFG,TYPG
COMMON 09 END
COMMON 11
0009          COMMON /C11/X
0010          REAL*8 X(30)
COMMON 11 END
COMMON 12
0011          COMMON /C12/NQ
0012          INTEGER NQ(30)
COMMON 12 END
COMMON 14
0013          COMMON /C14/F,GRAD
0014          REAL*8 F,GRAD(30)
COMMON 14 END
COMMON 15
0015          COMMON /C15/TOLDCM,TOLDEP,TOLINV,TOLPR,TOLAM,TOLMUL
COMMON 15 END
COMMON 17
0016          COMMON/C17 /IUNIT5,IUNIT6,IUNIT8
COMMON 17 END
COMMON 18
0017          COMMON /C18 /SOLNX,SOLNF,SOLNQ
0018          REAL*8 SOLNX(30),SOLNF
0019          INTEGER SOLNQ
COMMON 18 END
COMMON 21
0020          COMMON/C21 /IAG,NSTEPS,NADDS,NDROPS,NOPS,NSQRT,NREIN,NXCDR
1,NFSTEP,NPSTEP,ICC,NTIME,MXSTPS,MXREIN
COMMON 21 END
0021          IMF=0
0022          NMISS=0
0023          SUM=0.
0024          SUMX=0.
0025          DO 5 I=1,SOLNQ

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0026      5      LOGIC(I)=1
0027      CALL UPRIN( 9000)
0028      MFLAG=10000
0029      -10     DO 20 I=1,M
0030      SUM=SUM+(X(I)-SOLNX(I))*#2
0031      SUMX=SUMX+SOLNX(I)*#2
0032      IF(DABS(X(I)-SOLNX(I)).LE.1.D-2*DABS(SOLNX(I))) GOTO 20
0033      IMF=1
0034      CALL JPRIN(I )
0035      ICON = - DABS(X(I)-SOLNX(I)) * 100000
0036      CALL UPRIN( ICON)
0037      20     CONTINUE
0038      IF(IMF.NE.1)GOTO21
0039      CALL UPRIN( 9001)
0040      21     CONTINUE
0041      SUM=DSQRT(SUM)
0042      SUMX=DSQRT(SUMX)
0043      RAT=SUM/SUMX
0044      IRAT=RAT*100000
0045      CALL UPRIN( IRAT)
0046      CALL UPRIN( 9002)
0047      IF(RAT.GT.TOLPR)WRITE(IUNIT6, 30 )SUM,SUMX,RAT
0048      30     FORMAT(' ',5X,'RELATIVE ERROR= ',F10.4,' / ',F10.4,' = ',F10.4)
0049      IF (IMF.EQ.1) GOTO 80
0050      45     IF (Q.EQ.0.AND.SOLNQ.EQ.0) GOTO 249
0051      IF(Q.EQ.0) GOTO 155
0052      DO 48 I = 1,Q
0053      LOGIC(NQ(I))=0
0054      IF(NQ(I).LE.SOLNQ)GOTO 48
0055      WRITE(6,53)NQ(I)
0056      53     FORMAT('0',10X,'CONSTRAINT # ',I5,' MUST NOT BE IN FINAL BASIS')
0057      NMISS = NMISS + 1
0058      CALL UPRIN( NQ(I))
0059      48     CONTINUE
0060      155    IF(SOLNQ.EQ.0) GOTO 169
0061      DO 47 I=1,SOLNQ
0062      IF(LOGIC(I).NE.1)GOTO 47
0063      159    FORMAT('0',10X,'CONS.# ',I5,' LAMDA',F16.7,' MISSING IN FB')
0064      NMISS = NMISS + 1
0065      CALL UPRIN( -I)
0066      WRITE(6,159)I,XLAMDA(I)
0067      47     CONTINUE
0068      169    IF(NMISS.LE.0)GOTO 158
0069      MFLAG=MFLAG+10
0070      CALL UPRIN( 9003)
0071      158    IMF=J
0072      IF(SOLNQ.EQ.0)GOTO 249
0073      DO 50 I=1,SOLNQ
0074      IF (DABS(XLAMDA(I)).LE.TOLAM)GOTO 50
0075      140    WRITE(6,150)I,XLAMDA(I)
0076      ICON = -DABS(XLAMDA(I)) * 100000
0077      CALL UPRIN( ICON)
0078      IMF=1
0079      50     CONTINUE
0080      IF(IMF.EQ.0)GOTO 249
0081      CALL UPRIN( 9004)

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0082      MFLAG=MFLAG+1
0083      150  FORMAT(/15X,'BASIS CONSTRAINT ',I4,' IS NOT ACTIVE -LAMBDA',F10.6)
0084      GOTO 249
0085      80   WRITE (6,90)
0086      90   FORMAT('0',/15X,'SORRY THE VECTOR X IS NOT WHAT IT SHOULD BE',/
1,15X,'IT SHOULD BE ',5X,'IT IS' )
0087      DO 100 I=1,M
0088      NLINE=NLINE+1
C        IF (NLINE.GT.50) CALL HEADER
0089      100  WRITE(IUNIT6, 110)SOLNX(I),X(I)
0090      110  FORMAT(' ',5X,F15.4,F15.4)
0091      MFLAG=MFLAG+1000
0092      GOTO 45
0093      249  IF (Q.EQ.SOLNQ) GOTO 248
0094      120  WRITE (6,130) SOLNQ,Q
0095      130  FORMAT('0',/15X,'NUMBER OF CONS. IN BASIS IS NOT SAME',/5X,'SHOULD
1D BE ',5X,'AND IS',/2110)
0096      MFLAG = MFLAG + 100
0097      CALL UPRIN( SOLNQ)
0098      CALL UPRIN( -Q)
0099      CALL UPRIN( 9000)
0100      248  IF (DABS(F-SOLNF).GT.1.D-5 *DABS(SOLNF)) MFLAG = -MFLAG
0101      IF(DABS(F-SOLNF).GT.1.D-5 *DABS(SOLNF)) WRITE(6,70) SOLNF,F
0102      70   FORMAT('0',/15X,'SORRY, OPTIMUM VALUES ARE DIFF.',/15X,'SHOULD BE'
1,F15.6,5X,'BUT IS',F15.6)
0103      ICC = MFLAG
0104      RETURN
0105      END

```

ADDR1	ADDR2	STMT	SOURCE	STATEMENT	
		1	SINNER	CSECT	
	00000	2		USING *,15	
	0000C	3		ST 14,12(13)	
		4		SAVE (1,5)	
		5*		DS OH	
	00018	6*		STM 1,5,24(13)	SAVE REGISTERS
	00070	7	N 14,=X'F0FFFFFF'		
		8	SPM 14		
	00001	9		LA 4,1	
	00001	10		LA 5,1	
00008		11		TM 8(1),128	
	00036	12		BD START	
	0000C	13		L 4,12(,1)	
	00000	14		L 4,0(,4)	
0000C		15		TM 12(1),128	
	00036	16		BD START	
	00010	17		L 5,16(1)	
	00000	18		L 5,0(,5)	
	00003	19	START	SLA 4,3	
	00003	20		SLA 5,3	
		21		SDR 0,0	
		22		SDR 2,2	
		23		SDR 4,4	
	00000	24		LM 1,3,0(1)	
	00000	25		L 3,0(3)	
	00000	26	LOOP	LD 2,0(,1)	
	00000	27		LD 4,0(,2)	
		28		MDR 2,4	
		29		ADR 0,2	
		30		AR 1,4	
		31		AR 2,5	
	0004C	32		BCT 3,LOOP	
	0000C	33	L 14,12(13)		
		34	SPM 14		
		35		RETURN (1,5)	
	00018	36*		LM 1,5,24(13)	RESTORE THE REG
		37*		BR 14	RETURN
		38		END	
		39		=X'F0FFFFFF'	

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