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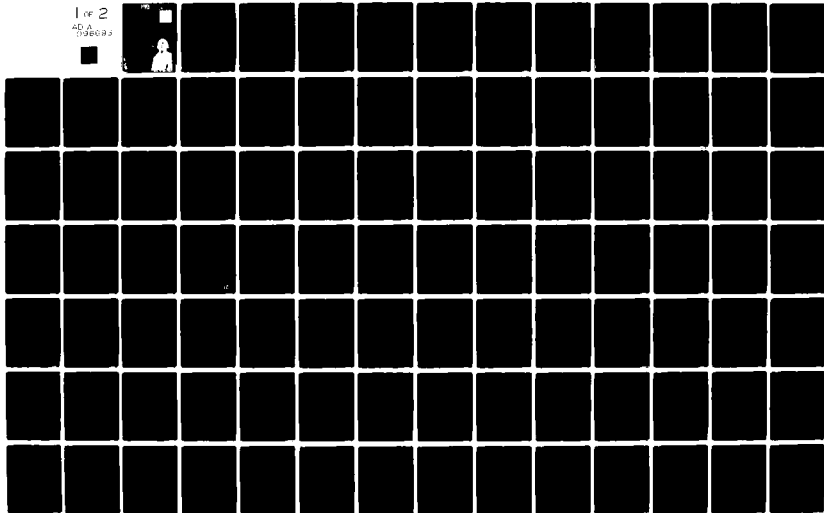
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A USER'S MANUAL FOR SENSUMT:

A Penalty Function Computer Program for Solution, Sensitivity Analysis, and Optimal Value Bound Calculation in Parametric Nonlinear Programs.

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by

Scientific Dept.

Anthony V. Fiacco
Abolfazl Ghaemi

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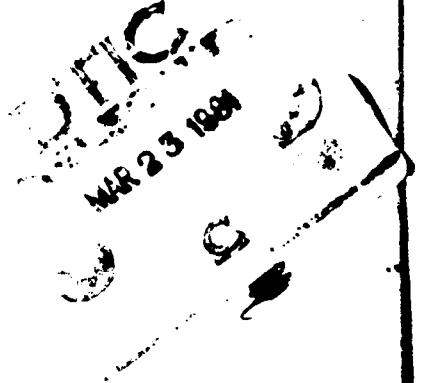
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output is included. For completeness, the computer listing and a brief description of all the subroutines comprising SENSUMT, many of which are taken intact from a previously developed program SUMT-Version 4, are also provided.

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This manual is intended to serve as a guide for coding, solving, and conducting sensitivity and optimal value bound analysis for parametric nonlinear programming problems with the computer programming model SENSUMT. The basic sensitivity results and bound calculation techniques are briefly reviewed and the algorithms implementing them are presented. A procedure for coding problems for SENSUMT and detailed illustrations of the computer output is included. For completeness, the computer listing and a brief description of all the subroutines comprising SENSUMT, many of which are taken intact from a previously developed program SUMT-Version 4, are also provided.

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This latest version of the SENSUMT program is an experimental program that is still evolving, though it has been used routinely at The George Washington University over the past several years to solve intermediate sized (up to about 100 variables and constraints) NLP problems and to conduct solution sensitivity analysis. A major part of it is the product of many theoretical, computational and software developments contributed by others. Several subroutines have been taken essentially intact from existing codes, most notably from SUMT-Version 4, coded by W. C. Mylander, R. L. Holmes, and G. P. McCormick; the first version of SENSUMT coded by R. L. Armacost and W. C. Mylander; and an extended version of SENSUMT coded subsequently by Armacost. The authors wish especially to acknowledge these contributions, along with the first program to implement some of the techniques used by this general approach for the calculation of sensitivity analysis, coded by B. Causey.

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1. Introduction

This is a manual intended to facilitate the use of the computer program SENSUMT. This program is capable of solving general parametric nonlinear programs and conducting a sensitivity analysis using the Sequential Unconstrained Minimization Technique [10]. It has recently been modified to include the calculation of piecewise linear parametric bounds on the optimal value function of classes of nonlinear programming problems for which this function is convex or concave. In addition, it can calculate similar bounds on a class of separable non-convex right-hand side programs, once appropriate convex overestimating and underestimating programs are formulated.

SENSUMT is the outgrowth of a routine [3] which first implemented a penalty function sensitivity approach. This was motivated by results given by Fiacco and McCormick [10] for a particular class of perturbations. The algorithmic procedure for this routine was suggested by Fiacco. It was later refined and recoded by Mylander [12] making use of

several subroutines from the SUMT-Version 4 computer program coded by Mylander, Holmes, and McCormick [13]. These routines were subsequently completely integrated with SUMT-Version 4 by Armacost and Mylander [7]. Armacost [1] further revised the program to implement the generalized version of the sensitivity theory developed by Fiacco [9] for general parameter perturbations. This routine, again following a procedure suggested by Fiacco, was recently modified by Ghaemi to conduct piecewise linear parametric bound calculation on the optimal value function of certain classes of nonlinear programming problems subject to large parametric changes [11].

This latter version of the model, designated "SENSUMT," is compiled at the Center for Academic and Administrative Computing of The George Washington University.

The various sections of this manual are arranged as follows. Section 2 reviews the basic sensitivity results and presents the steps of the algorithm implementing sensitivity calculation via a sequential unconstrained minimization algorithm.

Section 3 briefly reviews the procedure for calculating piecewise linear parametric bounds on the optimal value function of certain classes of parametric nonlinear programming problems. It presents the relevant algorithms that have been implemented.

Section 4 gives the details regarding the procedure for coding the problems for solution, sensitivity analysis, and bound calculation with SENSUMT.

Section 5 provides the codes and corresponding computer solution for two illustrative examples that are taken from [11]. For clarity, an annotated computer listing of the input and output for the first example is also provided.

Section 6 gives a description and listing of the various subroutines comprising SENSUMT. This section is included to make the

manual self-contained. With the exception of the subroutines BOUND, PERT, and TRANS, and the new version of program MAIN, the material in this section with a few minor modifications has been taken from [1] and [13].

2. Sensitivity Analysis in Nonlinear Programming

2.1 Basic Sensitivity Results

The parametric mathematical programming problem considered by Fiacco [9] is of the following general form:

$$\begin{aligned} & \text{minimize } f(x, \epsilon) \\ & \quad x \in E^n \\ & \text{subject to } g_i(x, \epsilon) \geq 0, \quad i=1, \dots, m, \\ & \quad \quad \quad h_j(x, \epsilon) = 0, \quad j=1, \dots, p, \end{aligned} \quad P(\epsilon)$$

where x is the usual vector of variables and ϵ is a k -component vector of numbers called "parameters." It is desired to analyze the behavior of a solution vector $x(\epsilon)$ and the optimal solution value $f^*(\epsilon) \equiv f[x(\epsilon), \epsilon]$ near some given value of ϵ . Without loss of generality, assume that the parameter vector of interest is $\epsilon=0$.

The Lagrangian for Problem $P(\epsilon)$ is defined as

$$L(x, u, w, \epsilon) \equiv f(x, \epsilon) - \sum_{i=1}^m u_i g_i(x, \epsilon) + \sum_{j=1}^p w_j h_j(x, \epsilon). \quad (2.1)$$

The sensitivity results are based on the following four assumptions:

- A1 - The functions defining Problem $P(\epsilon)$ are twice continuously differentiable in (x, ϵ) in a neighborhood of $(x^*, 0)$.
- A2 - The second order sufficient conditions for a local minimum of Problem $P(0)$ hold at x^* with associated Lagrange multipliers y^* and w^* .

A3 - The gradients $\nabla_x g_i(x^*, 0)$, for all i such that $g_i(x^*, 0) = 0$, and $\nabla_x h_j(x^*, 0)$, $j=1, \dots, p$ are linearly independent.

A4 - Strict complementary slackness holds at x^* when $\epsilon=0$ [i.e., $u_i^* > 0$ for all i such that $g_i(x^*, 0) = 0$].

Under the above assumptions, Fiacco [9] established the following generalization of Theorem 6 in [10].

Lemma 2.1 [local characterization of a Kuhn-Tucker triple]: If assumptions A1, A2, A3, and A4 hold for Problem $F(\epsilon)$ at $(x^*, 0)$, then

- (a) x^* is a local isolated minimizing point of Problem $P(0)$ and the associated Lagrange multipliers u^* and w^* are unique;
- (b) for ϵ in a neighborhood of 0, there exists a unique once continuously differentiable vector function $y(\epsilon) = (x(\epsilon), u(\epsilon), w(\epsilon))^T$ satisfying the second order sufficient conditions for a local minimum of Problem $P(\epsilon)$ such that $y(0) = (x^*, u^*, w^*)^T = y^*$, and hence, $x(\epsilon)$ is a locally unique, local minimum of Problem $P(\epsilon)$ with associated unique Lagrange multipliers $u(\epsilon)$ and $w(\epsilon)$; and
- (c) for ϵ near 0, the set of binding inequalities is unchanged, strict complementary slackness continues to hold, and the binding constraint gradients are linearly independent at $x(\epsilon)$.
- (d) (Armstrong and Fiacco [3]), for ϵ near 0, the gradient of the optimal value function is

$$\nabla_{\epsilon} f^*(\epsilon) = \nabla_{\epsilon} L(y(\epsilon), \epsilon), \quad (2.2)$$

- (e) which also means that, for ϵ near 0, the Hessian of the optimal value function is

$$\nabla_{\epsilon}^2 f^*(\epsilon) = \nabla_{\epsilon}^2 L(y(\epsilon), \epsilon). \quad (2.3)$$

The above results provide a characterization of a local solution of Problem $P(\epsilon)$ and its associated optimal Lagrange multipliers near $\epsilon=0$. They show that the Kuhn-Tucker triple $y(\epsilon)$ is unique and well behaved, under the given conditions. Since $y(\epsilon)$ is once differentiable, the partial derivatives of the components of $y(\epsilon)$ are well defined. This fact and assumption A1 also mean that the functions defining Problem $P(\epsilon)$ are once continuously differentiable functions of ϵ along the "solution trajectory" $x(\epsilon)$ near $\epsilon=0$, and the Lagrangian is a once continuously differentiable function of ϵ along the "Kuhn-Tucker point trajectory." The above results constitute the structure for numerous developments and extensions, many of which have been established by Fiacco [9] and Armacost and Fiacco [2 - 6].

The realization of this theorem for the parametric right hand side problem of special interest in the present study is treated in detail by Armacost and Fiacco [4]. The parametric right hand side problem is the following important realization of $P(\epsilon)$:

$$\begin{aligned} & \text{minimize } f(x) \\ & \text{subject to } g_i(x) \geq \epsilon_i \quad , \quad i=1, \dots, m \quad , \quad R(\epsilon) \\ & \quad \quad \quad h_j(x) = \epsilon_{j+m} \quad , \quad j=1, \dots, p \quad . \end{aligned}$$

The Lagrangian for $R(\epsilon)$ is

$$L(x, u, w) \equiv f(x) - \sum_{i=1}^m u_i (g_i(x) - \epsilon_i) + \sum_{j=1}^p w_j (h_j(x) - \epsilon_{j+m}) .$$

As is evident from the Lagrangian, the results (d) and (e) of Lemma 2.1 for Problem $R(\epsilon)$ simplify respectively to

$$(d') \quad \nabla_{\epsilon} f^*(\epsilon) = \begin{bmatrix} u(\epsilon) \\ -w(\epsilon) \end{bmatrix}^T \quad (2.4)$$

and

$$(e') \quad \nabla_{\epsilon}^2 f^*(\epsilon) = \begin{bmatrix} \nabla_{\epsilon} u(\epsilon) \\ -\nabla_{\epsilon} w(\epsilon) \end{bmatrix} . \quad (2.5)$$

Fiacco [9] has shown that the class of algorithms based on twice continuously differentiable penalty functions (specifically, using the logarithmic-quadratic loss penalty function) can be used to estimate $y(\epsilon)$ and its derivatives in a neighborhood of $\epsilon=0$, for the general problem $P(\epsilon)$. Minimization of the penalty function with penalty parameter r yields a solution of a perturbation of the Kuhn-Tucker system in a neighborhood of $(\epsilon, r) = (0, 0)$. Armacost and Fiacco [3] define an optimal value penalty function and obtain first- and second-order sensitivity estimates which converge to the corresponding sensitivities for the optimal value function for Problem $P(\epsilon)$.

The logarithmic-quadratic penalty function is

$$W(x, \epsilon, r) \equiv f(x, \epsilon) - r \sum_{i=1}^m \ln g_i(x, \epsilon) + (1/2r) \sum_{j=1}^p h_j^2(x, \epsilon). \quad (2.6)$$

Lemma 2.2 (Fiacco [9, Theorem 3.1]): If the assumptions A1 through A4 hold, then in a neighborhood of $(\epsilon, r) = (0, 0)$ there exists a unique once continuously differentiable vector function $y(\epsilon, r) = [x(\epsilon, r), u(\epsilon, r), w(\epsilon, r)]^T$ satisfying

$$\begin{aligned} \nabla_x L(x, u, w, \epsilon) &= 0, \\ u_i g_i(x, \epsilon) &= r, \quad i=1, \dots, m, \\ h_j(x, \epsilon) &= w_j r, \quad j=1, \dots, p, \end{aligned} \quad (2.7)$$

with $y(0, 0) = (x^*, u^*, w^*)$, and such that for any (ϵ, r) near $(0, 0)$ and $r > 0$, $x(\epsilon, r)$ is a locally unique unconstrained local minimizing point of $W(x, \epsilon, r)$, with $g_i[x(\epsilon, r), \epsilon] > 0$, $i=1, \dots, m$, and $\nabla_x^2 W[x(\epsilon, r), \epsilon, r]$ positive definite.

The relevance of Equations (2.7) is the fact that, under the given conditions, when $r=0$, they are necessary conditions that must hold at a local solution of $P(0)$ and, with $r > 0$, they are necessary conditions for an unconstrained minimum of $W(x, \epsilon, r)$. The latter fact can be made obvious by solving for u_i and w_j in (2.7) and obtaining

$$\begin{aligned}
\nabla_x L(x, u, w, \epsilon) &= \nabla_x f - \sum_i u_i \nabla_x g_i + \sum_j w_j \nabla_x h_j \\
&= \nabla_x f - r \sum_i (1/g_i) \nabla_x g_i + (1/r) \sum_j h_j \nabla_x h_j \\
&= \nabla_x W(x, \epsilon, r) .
\end{aligned} \tag{2.8}$$

Thus, if $y(\epsilon, r)$ is a solution of (2.7), then

$$\nabla_x W[x(\epsilon, r), \epsilon, r] = \nabla_x L[x(\epsilon, r), u(\epsilon, r), w(\epsilon, r), \epsilon] = 0 . \tag{2.9}$$

This explicit connection between the optimality conditions of local solutions of $P(\epsilon)$ and unconstrained minima of $W(x, \epsilon, r)$ makes it possible to approximate information characterizing a local solution of $P(\epsilon)$ by algorithmic calculations associated with utilizing $W(x, \epsilon, r)$ to solve $P(\epsilon)$. In particular, differentiating (2.9) with respect to ϵ yields

$$\nabla_x^2 W \nabla_\epsilon x + \nabla_{\epsilon x}^2 W = 0 ,$$

and using the fact that $\nabla_x^2 W$ is positive definite (a conclusion of Lemma 2.2) yields

$$\nabla_\epsilon x = -\nabla_x^2 W^{-1} \nabla_{\epsilon x}^2 W , \tag{2.10}$$

evaluated, of course, at $x(\epsilon, r)$. Given $\nabla_\epsilon x(\epsilon, r)$, the derivatives of the multipliers, $\nabla_\epsilon u_i(\epsilon, r)$ and $\nabla_\epsilon w_j(\epsilon, r)$, can then be calculated by differentiating the last two systems of equations of (2.7) at $x(\epsilon, r)$ with respect to ϵ .

Lemma 2.3 (Fiacco [9]): For ϵ in a neighborhood of $\epsilon=0$, it follows that

- (a) $\lim_{r \rightarrow 0^+} y(\epsilon, r) = y(\epsilon, 0) = y(\epsilon)$, the Kuhn-Tucker triple characterized in conclusion (b) of Lemma 2.1; and
- (b) $\lim_{r \rightarrow 0^+} \nabla_\epsilon y(\epsilon, r) = \nabla_\epsilon y(\epsilon, 0) = \nabla_\epsilon y(\epsilon)$.

Under the conditions of Lemma 2.1, $x(\epsilon, r)$ is a locally unique minimizing point of $W(x, \epsilon, r)$. Define the "optimal value penalty function" as

$$W^*(\epsilon, r) \equiv W(x(\epsilon, r), \epsilon, r) . \quad (2.11)$$

Armacost and Fiacco [2, Theorem 4 and Corollary 4.1] have obtained further results useful for estimating the first- and second-order sensitivity of the optimal value function $f^*(\epsilon)$ of Problem $P(\epsilon)$.

Lemma 2.4 (Armacost and Fiacco [3]): If the assumptions A1 through A4 hold for Problem $P(\epsilon)$, then in a neighborhood of $\epsilon=0$,

- (a) $\lim_{r \rightarrow 0^+} W^*(\epsilon, r) = f^*(\epsilon)$;
- (b) $\nabla_{\epsilon} W^*(\epsilon, r) = \nabla_{\epsilon} L(y, \epsilon)$ at $y = y(\epsilon, r)$;
- (c) $\lim_{r \rightarrow 0^+} \nabla_{\epsilon} W^*(\epsilon, r) = \nabla_{\epsilon} f^*(\epsilon)$;
- (d) $\nabla_{\epsilon}^2 W^*(\epsilon, r) = \nabla_{\epsilon}^2 L(y(\epsilon, r), \epsilon)$; and
- (e) $\lim_{r \rightarrow 0^+} \nabla_{\epsilon}^2 W^*(\epsilon, r) = \nabla_{\epsilon}^2 f^*(\epsilon)$;

where convergence is component by component in all cases.

Lemmas 2.1 through 2.4 enable us to calculate an estimate of $y(\epsilon)$, $\nabla_{\epsilon} y(\epsilon)$, $\nabla_{\epsilon} f^*(\epsilon)$, and $\nabla_{\epsilon}^2 f^*(\epsilon)$ when ϵ is near 0 and r is near 0, once $y(\epsilon, r)$ is available.

In the next section we briefly present the algorithmic implementation of some of the above results.

2.2 The Algorithm Implementing the Sensitivity Results

The penalty function algorithm SUMT estimates the solution of the general mathematical problem $P(\epsilon)$ by estimating the unconstrained minima

of the penalty function $W(x, \epsilon, r)$ at successively decreasing values of the penalty function parameter $r > 0$. Under conditions weaker than those assumed here, Fiacco and McCormick [10] have shown that as r approaches zero, the sequence of the unconstrained minima of $W(x, \epsilon, r)$ will approach a solution of $P(\epsilon)$. Each unconstrained penalty function minimization is thus a "subproblem" associated with a particular value of the penalty function parameter r .

The successive steps of the algorithm for first order sensitivity analysis of the Kuhn-Tucker triple with respect to the j th parameter that are implemented in SENSUMT [1] are listed below. Here the notation \bar{x} or $\bar{x}(\bar{\epsilon})$ denotes the estimate of a solution point of $P(\bar{\epsilon})$ calculated by SUMT for a given value of the penalty function parameter r , where $\bar{\epsilon}$ denotes the value of the problem parameter at which this sensitivity is estimated.

Algorithm 2.1:

Step 1. Compute a representation of $\nabla_x^2 W^{-1} = \nabla_x^2 W(\bar{x}, \bar{\epsilon}, r)^{-1}$ by L-U decomposition using the SUMT subroutines. If $\nabla_x^2 W$ is not positive definite, terminate the sensitivity analysis.

Step 2. Estimate $\partial(\nabla_x W^T)/\partial \epsilon_j$ using the central differencing formula

$$\partial(\nabla_x W^T)/\partial \epsilon_j \approx \left(\frac{1}{2\Delta}\right) \left(\nabla_x W(\bar{x}, \bar{\epsilon} + \Delta e_j, r)^T - \nabla_x W(\bar{x}, \bar{\epsilon} - \Delta e_j, r)^T \right),$$

where Δ is the differencing interval and e_j is the j th unit vector.

Step 3. Calculate

$$\partial \bar{x}(\bar{\epsilon})/\partial \epsilon_j = -\nabla_x^2 W^{-1} \partial(\nabla_x W^T)/\partial \epsilon_j.$$

Step 4. Estimate $\nabla_{\epsilon} g_1(\bar{x}, \bar{\epsilon})$ and $\nabla_{\epsilon} h_1(\bar{x}, \bar{\epsilon})$ using

$$\partial g_1(\bar{x}, \bar{\epsilon})/\partial \epsilon_j \approx \left(\frac{1}{2\Delta}\right) \left(g_1(\bar{x}, \bar{\epsilon} + \Delta e_j) - g_1(\bar{x}, \bar{\epsilon} - \Delta e_j) \right), \text{ and}$$

$$\partial h_1(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j \approx \left(\frac{1}{2\Delta} \right) (h_1(\bar{x}, \bar{\epsilon} + \Delta e_j) - h_1(\bar{x}, \bar{\epsilon} - \Delta e_j)) .$$

Step 5. Estimate the components of $\nabla_{\epsilon} u(\bar{\epsilon})$ for $i=1, \dots, m$ using

$$\partial u_i(\bar{\epsilon}) / \partial \epsilon_j \approx - \left(r / g_i(\bar{x}, \bar{\epsilon})^2 \right) \left(\nabla_x g_i(\bar{x}, \bar{\epsilon}) \partial \bar{x}(\bar{\epsilon}) / \partial \epsilon_j + \partial g_i(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j \right) .$$

Step 6. Estimate the components of $\nabla_{\epsilon} w(\bar{\epsilon})$ for $i=1, \dots, p$ using

$$\partial w_i(\bar{\epsilon}) / \partial \epsilon_j \approx (1/r) \left(\nabla_x h_i(\bar{x}, \bar{\epsilon}) \partial \bar{x}(\bar{\epsilon}) / \partial \epsilon_j + \partial h_i(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j \right) .$$

There are two methods for estimating $\nabla_{\epsilon} f^*(\bar{\epsilon})$, the first using $\nabla_{\epsilon} f^* = \nabla_x f \nabla_{\epsilon} x + \nabla_{\epsilon} f$, with $\nabla_{\epsilon} x$ obtained from Step 3, and the second method using the gradient of the penalty function W , or equivalently, the Lagrangian taken with respect to the parameters [Lemma 2.1, conclusion (d); Lemma 2.4, conclusion (b)]. Both are incorporated in the computer program but used for different purposes. The former method gives the most accurate estimate of $\nabla_{\epsilon} f^*(\bar{\epsilon})$ and is summarized in Steps 7 and 8.

Step 7. Estimate the components of $\nabla_{\epsilon} f(\bar{x}, \bar{\epsilon})$ using the central differencing formula

$$\partial f(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j \approx \left(\frac{1}{2\Delta} \right) (f(\bar{x}, \bar{\epsilon} + \Delta e_j) - f(\bar{x}, \bar{\epsilon} - \Delta e_j)) .$$

Calculate an estimate of the components of $\nabla_{\epsilon} f^*(\bar{\epsilon})$ using the results of Steps 3 and 7 as

$$\partial f^*(\bar{\epsilon}) / \partial \epsilon_j \approx \nabla_x f(\bar{x}, \bar{\epsilon}) \partial \bar{x}(\bar{\epsilon}) / \partial \epsilon_j + \partial f(\bar{x}, \bar{\epsilon}) / \partial \epsilon_j .$$

The second method, using the gradient of the Lagrangian to estimate $\nabla_{\epsilon} f^*(\bar{\epsilon})$, is computationally less expensive and is used to obtain rough estimates that single out the more crucial parameters for further analysis. This approximation is calculated as follows.

Step 9. Estimate the components of $V_{\epsilon} f^*(\bar{\epsilon})$ using the results of Steps 4 and 7 as

$$\begin{aligned} \partial f^*(\bar{\epsilon})/\partial \epsilon_j &\approx \partial f(\bar{x}, \bar{\epsilon})/\partial \epsilon_j - \sum_{i=1}^m u_i(\bar{\epsilon}, r) \partial g_i(\bar{x}, \bar{\epsilon})/\partial \epsilon_j \\ &\quad + \sum_{i=1}^p w_i(\bar{\epsilon}, r) \partial h_i(\bar{x}, \bar{\epsilon})/\partial \epsilon_j . \end{aligned}$$

3. Parametric Optimal Value Bounds

In this section we briefly review the procedure for calculating piecewise linear parametric upper and lower bounds on the optimal value function of convex problems with right hand side perturbations of the form $CR(\epsilon)$. It is important to note that the given technique is not only applicable to problems of this form, but also to any class of problems which have convex or concave optimal value functions. In addition, we show how the above technique can be extended to calculate similar bounds on the optimal value function of separable nonconvex right hand side perturbation problems $SR(\epsilon)$, once their (computable) over- and underestimating problems are available. For a more detailed treatment, readers should refer to [11].

3.1 Parametric Bounds on the Optimal Value Function of Convex Right Hand Side Problems $CR(\epsilon)$

Consider the following right hand side parametric programming problem $R(\epsilon)$, discussed in Section 2:

$$\begin{aligned} &\text{minimize } f(x) \\ &\text{subject to } g_i(x) \geq \epsilon_i, \quad i=1, m \qquad R(\epsilon) \\ &\qquad\qquad h_j(x) = \epsilon_j, \quad j=m+1, p \end{aligned}$$

where $x \in E^n$, f, g , and $h : E^n \rightarrow E^1$, and C^2 and $\epsilon \in E^p$. If $f(x)$ and $-g_i(x)$, $i=1, m$, are convex and $h_j(x)$, $j=m+1, p$ are linear, then

the problem $R(\epsilon)$ is convex and will be designated by $CR(\epsilon)$. It is well known that $f^*(\epsilon)$, the optimal value function of the problem $CR(\epsilon)$, is a convex function of ϵ .

The convexity of the optimal value function $f^*(\epsilon)$ of the problem $CR(\epsilon)$ enables one to calculate parametric upper and lower bounds on this function when any of the problem parameters is radically perturbed. This of course requires the solution and corresponding optimal value function sensitivity information for both perturbed and unperturbed problems. Before delving into the calculation procedure of these bounds, we must remind ourselves of the following two basic properties of convex functions.

- (i) Any line connecting two points on the graph of a convex function does not underestimate that function between the points.
- (ii) Any line tangent to the graph of a convex function does not overestimate that function.

These two properties lend themselves in a natural way to the calculation of parametric bounds on the optimal value function of the problem $CR(\epsilon)$ under large perturbations of any of the problem parameters, say ϵ_i .

3.2 Algorithm for Calculation of Bounds on $f^*(\epsilon)$ of the Problem $CR(\epsilon)$

In the following we list the successive steps required in calculation of bounds on $f^*(\epsilon)$, the optimal value function of $CR(\epsilon)$, as a function of ϵ_i , when ϵ_i is perturbed from $\bar{\epsilon}_{i1}$ to $\bar{\epsilon}_{i2}$ while the remaining parameters are fixed at their base values.

Algorithm 3.1:

Step 1. Solve the unperturbed problem and obtain $f^*(\epsilon_i)$ and $df^*(\epsilon_i)/d\epsilon_i$ at $\epsilon_i = \bar{\epsilon}_{i1}$. Note that

$$\frac{df^*(\epsilon_i)}{d\epsilon_i} = \begin{cases} u_i(\epsilon_i), & i=1, \dots, m \\ -w_i(\epsilon_i), & i=m+1, p. \end{cases} \quad (2.4)$$

$$(2.5)$$

Step 2. Re-solve the problem and obtain $f^*(\epsilon_i)$ and $df^*(\epsilon_i)/d\epsilon_i$ at $\epsilon_i = \bar{\epsilon}_{i2}$.

Step 3. Derive the equation of the line passing through the points $(\epsilon_i = \bar{\epsilon}_{i1}, f^*(\epsilon_i) = f^*(\bar{\epsilon}_{i1}))$ and $(\epsilon_i = \bar{\epsilon}_{i2}, f^*(\epsilon_i) = f^*(\bar{\epsilon}_{i2}))$. This line provides a parametric upper bound $\bar{f}^*(\epsilon_i)$ for $f^*(\epsilon_i)$ as a function of $\epsilon_i \in [\bar{\epsilon}_{i1}, \bar{\epsilon}_{i2}]$.

Step 4. Derive the equation of the tangent lines to $f^*(\epsilon_i)$ at the above two points with the slopes

$$\left. \frac{df^*(\epsilon_i)}{d\epsilon_i} \right|_{\epsilon_i = \bar{\epsilon}_{i1}} \quad \text{and} \quad \left. \frac{df^*(\epsilon_i)}{d\epsilon_i} \right|_{\epsilon_i = \bar{\epsilon}_{i2}}$$

calculated in Steps 1 and 2, respectively. These two lines provide a parametric lower bound $\underline{f}^*(\epsilon_i)$ for $f^*(\epsilon_i)$ as a function of $\epsilon_i \in [\bar{\epsilon}_{i1}, \bar{\epsilon}_{i2}]$.

The lines obtained in Steps 3 and 4 form a triangle which encloses the optimal value function $f^*(\epsilon_i)$ over the given range of ϵ_i . An estimate of $f^*(\epsilon_i)$ can also be made by fitting a convex function that passes through the points given in Step 3 and having the corresponding slopes at these points obtained in Steps 1 and 2.

It is clear that the fundamental requirement in using the above algorithm is that the optimal value function be convex and differentiable at least at $\epsilon_i = \bar{\epsilon}_{i1}$ and $\epsilon_i = \bar{\epsilon}_{i2}$. As it was given in Section 2, Fiacco, under the assumptions A1 - A4 of Lemma 2.1, has established the differentiability of the optimal value function in a neighborhood of $\epsilon=0$ for the general parametric programming problem $P(\epsilon)$. Thus, not

only Problem CR(ϵ), but all classes of parametric nonlinear programming problems possessing a convex optimal value function and, in particular, meeting assumptions A1 - A4 at $\epsilon_i = \bar{\epsilon}_{i1}$ and $\epsilon_i = \bar{\epsilon}_{i2}$, can be analyzed for bounds according to Algorithm 3.1.

3.3 Parametric Bounds on the Optimal Value Function of the Problem SR(ϵ)

In this section we show how Algorithm 3.1 can be extended to calculate piecewise linear parametric upper and lower bounds on the optimal value function of a separable nonconvex right hand side perturbation problem, given the over- and underestimating problems.

3.3.1 Separable right hand side perturbation problems.

The separable nonconvex problem addressed here has the following structure:

$$\begin{aligned} \text{minimize}_{x \in E^n} f(x) &= \sum_{j=1}^n f_j(x_j) \\ \text{subject to } g_i(x) &= \sum_{j=1}^n g_{ij}(x_j) \geq \epsilon_i, \quad i=1, \dots, m \quad \text{SR}(\epsilon) \\ &L_j \leq x_j \leq U_j, \quad j=1, \dots, n, \end{aligned}$$

where each $f_j(x_j)$ and $g_{ij}(x_j)$ is a function of a single variable x_j and is differentiable; L_j and U_j are lower and upper bounds on the variable x_j , respectively. Note that the discussion that follows is also valid when Problem SR(ϵ) involves linear equality constraints. Let

$$\begin{aligned} G &\equiv \left\{ x : g_i(x) = \sum_{j=1}^n g_{ij}(x_j) \geq \epsilon_i, \quad j=1, \dots, m \right\}, \\ B_j &\equiv \left\{ x_j : L_j \leq x_j \leq U_j \right\}, \quad j=1, \dots, n, \\ B &\equiv \prod_{j=1}^n B_j, \end{aligned}$$

and

$$S \equiv G \cap B .$$

The set G , and therefore set S , depends on ϵ . Thus the problem $SR(\epsilon)$ will read as follows:

$$\begin{aligned} \text{minimize } f(x) &= \sum_{j=1}^n f_j(x_j) \\ \text{subject to } x &\in B \cap G \equiv S \subset E^n . \end{aligned} \quad SR(\epsilon)$$

3.3.2 Convex underestimating problem.

Let $\tilde{f}_j(x_j)$ be the convex envelope of $f_j(x_j)$, $j=1, \dots, m$ and $\tilde{g}_{ij}(x_j)$ be the concave envelope of $g_{ij}(x_j)$, $j=1, \dots, n$, $i=1, \dots, m$ over the interval B_j . Thus, for $x_j \in B_j$,

$$\tilde{f}_j(x_j) \leq f_j(x_j)$$

and

$$\tilde{g}_{ij}(x_j) \geq g_{ij}(x_j), \quad i=1, \dots, m; j=1, \dots, n .$$

Definition: Define the problem $\tilde{C}SR(\epsilon)$ as follows:

$$\begin{aligned} \text{minimize } \tilde{f}(x) &= \sum_{j=1}^n \tilde{f}_j(x_j) \\ \text{subject to } x &\in G \cap B \equiv \tilde{S} , \end{aligned}$$

where

$$\tilde{G} \equiv \left\{ x : \tilde{g}_i(x) = \sum_{j=1}^n \tilde{g}_{ij}(x) \geq \epsilon_i, \quad i=1, \dots, m \right\} .$$

It follows easily that problem $\tilde{C}SR(\epsilon)$ is a separable convex RHS perturbation problem whose optimal value function underestimates the global optimal value function of the problem $SR(\epsilon)$.

Thus the basic ingredients required to formulate the convex underestimating problem $\tilde{C}SR(\epsilon)$ are the convex envelope of each term in the

objective function and the concave envelope of each term in the constraints of the problem $SR(\epsilon)$ over the rectangular polytope B .

In the next section we will discuss the formulation of a convex overestimating problem $C\tilde{S}R(\epsilon)$ of the problem $SR(\epsilon)$.

3.3.3 Convex overestimating problem.

Let $\tilde{f}_j(x_j)$ be a convex function which does not underestimate $f_j(x_j)$, $j=1, \dots, n$, and $\tilde{g}_{ij}(x_j)$ be a concave function which does not overestimate $g_{ij}(x_j)$, $j=1, \dots, n$, $i=1, \dots, m$, over the interval B_j . Thus, for $x_j \in B_j$,

$$\tilde{f}_j(x_j) \geq f_j(x_j)$$

and

$$\tilde{g}_{ij}(x_j) \leq g_{ij}(x_j), \quad i=1, \dots, m; j=1, \dots, n.$$

Henceforth in discussion, $\tilde{f}_j(x_j)$ will be abbreviated as convex "nuf" (nonunderestimating function) of the single variable function $f_j(x_j)$, and $\tilde{g}_{ij}(x_j)$ will be abbreviated as concave "nof" (nonoverestimating function) of $g_{ij}(x_j)$ over the interval B_j , $i=1, \dots, m$, $j=1, \dots, n$.

Definition: Let us define the problem $C\tilde{S}R(\epsilon)$ as follows:

$$\begin{aligned} & \underset{x \in E^n}{\text{minimize}} \quad \tilde{f}(x) = \sum_{j=1}^n \tilde{f}_j(x_j) \\ & \text{subject to} \quad x \in (\tilde{G} \cap B) \equiv \tilde{S}, \end{aligned}$$

where

$$\tilde{G} \equiv \left\{ x : \tilde{g}_i(x) = \sum_{j=1}^n \tilde{g}_{ij}(x_j) \geq \epsilon_i, \quad i=1, \dots, m \right\}.$$

It is easy to show that problem $C\tilde{S}R(\epsilon)$ is a separable convex RHS problem whose optimal value function underestimates the global optimal value function of the problem $SR(\epsilon)$.

Thus, in order to formulate an overestimating convex problem $\tilde{C}\tilde{S}R(\epsilon)$ of the problem $SR(\epsilon)$, one must calculate convex nuf, $f_{\approx j}(x_j)$, for each component of $f(x)$ and concave nof, $\tilde{g}_{ij}(x_j)$, for each component of $g_i(x)$, $i=1, \dots, m$, over the interval B_j . It must be noted that convex nuf and concave nof of a given function, unlike its envelopes, are not unique. Thus the overestimating problem $\tilde{C}\tilde{S}R(\epsilon)$ of problem $SR(\epsilon)$ is not unique. For further treatment of convex nuf and concave nof of single variable functions, readers may refer to [11].

An important point to be noted here is that formulation of the convex problems $\tilde{C}\tilde{S}R(\epsilon)$ and $\tilde{C}\tilde{S}R(\epsilon)$ enable one to calculate parametric upper and lower bounds on the (global) optimal value function of the separable nonconvex programming problem $SR(\epsilon)$.

3.4 Algorithm for Calculation of Bounds on the (Global) Optimal Value Function of the Problem $SR(\epsilon)$

In the following an algorithm is presented for calculation of upper and lower bounds on the (global) optimal value function of the problem $SR(\epsilon)$ as a function of each RHS parameter ϵ_i , where $\epsilon_i \in [\bar{\epsilon}_{i1}, \bar{\epsilon}_{i2}]$. Implicit in this algorithm is the assumption that the overestimating problem $\tilde{C}\tilde{S}R(\epsilon)$ has a nonempty interior, which is a basic requirement of almost any nonlinear programming algorithm.

Algorithm 3.2:

Step 1: Calculate $f_{\approx j}(x_j)$, a convex nuf for each $f_j(x_j)$, $j=1, \dots, n$, and $\tilde{g}_{ij}(x_j)$, a concave nof for each $g_{ij}(x_j)$, $i=1, \dots, m$, $j=1, \dots, n$, in closed interval B_j , and construct the corresponding overestimating problem $\tilde{C}\tilde{S}R(\epsilon)$.

Step 2: Solve the problem $\tilde{C}\tilde{S}R(\epsilon)$ at $\epsilon_i = \bar{\epsilon}_{i1}$ and $\epsilon_i = \bar{\epsilon}_{i2}$ and obtain $f_{\approx}^*(\epsilon_i)$ at these values of ϵ_i .

Step 3: Derive the equation for the line passing through the points $(\epsilon_i = \bar{\epsilon}_{i1}, f^*(\epsilon_i) = f^*(\bar{\epsilon}_{i1}))$ and $(\epsilon_i = \bar{\epsilon}_{i2}, f^*(\epsilon_i) = f^*(\bar{\epsilon}_{i2}))$. This line provides a parametric upper bound $\bar{f}^*(\epsilon_i)$ on the (global) optimal value function $f^*(\epsilon_i)$ of the problem $SR(\epsilon)$ as a function of ϵ_i over the interval $[\bar{\epsilon}_{i1}, \bar{\epsilon}_{i2}]$.

Step 4: Calculate $\tilde{f}_j(x_j)$, the convex envelope for each $f_j(x_j)$, $i=1, \dots, n$ and $\tilde{g}_{ij}(x_j)$, the concave envelope for each $g_{ij}(x_j)$, $i=1, \dots, m$, $j=1, \dots, n$ in closed interval B_j , and construct the corresponding underestimating problem $\underline{CSR}(\epsilon)$.

Step 5: Solve the problem $\underline{CSR}(\epsilon)$ at $\epsilon_i = \bar{\epsilon}_{i1}$ and $\epsilon_i = \bar{\epsilon}_{i2}$ and obtain $\underline{f}^*(\epsilon_i)$ and $\underline{u}_i(\epsilon_i) = d\underline{f}^*(\epsilon_i)/d\epsilon_i$ at these values of ϵ_i .

Step 6: Derive the equation for the tangent lines to $\underline{f}^*(\epsilon_i)$ at $\epsilon_i = \bar{\epsilon}_{i1}$ and $\epsilon_i = \bar{\epsilon}_{i2}$, with respective slopes of $\underline{u}_i(\bar{\epsilon}_{i1})$ and $\underline{u}_i(\bar{\epsilon}_{i2})$ derived in Step 5. These two lines over the interval $[\bar{\epsilon}_{i1}, \bar{\epsilon}_{i2}]$ jointly provide a parametric lower bound $\underline{f}^*(\epsilon_i)$ on the (global) optimal value function $f^*(\epsilon_i)$ of the problem $SR(\epsilon)$ as a function of ϵ_i .

The implicit assumption in Steps 5 and 6 of this algorithm is that assumptions A1 - A4 of Lemma 2.1 hold at the solution points when ϵ_i takes values of $\bar{\epsilon}_{i1}$ and $\bar{\epsilon}_{i2}$.

Before delving into the next section, we must point out that if over- and underestimating problems with concave optimal value functions could be formulated, then the algorithm above, with minor alterations, can be used to calculate the desired bounds. In this instance Step 6, when applied to the overestimating problem, will provide an upper bound

while Step 3, when applied to the underestimating problem, will provide a lower bound on the optimal value function. For calculation of envelopes, convex nuf and concave nof of single variable functions, readers may refer to [11]. The entries in Table 1 summarize the concept of convex nuf and concave nof for a variety of single variable functions.

4. SENSUMT Input Specifications

The coding procedure for nonlinear programming problems for solution, sensitivity analysis, and bound calculation using SENSUMT closely follows the coding procedure for SUMT-Version 4 [13]. In order to code a given problem for the above calculations, one must have (i) a user-supplied subroutine, and (ii) user's information cards.

4.1 User-supplied Subroutines

User-supplied subroutines will generally consist of four subroutines called READIN, RESTNT, GRAD1 and MATRIX. These subroutines are the only subroutines for SENSUMT that are problem dependent. The communication between these subroutines and the rest of the subroutines of SENSUMT is mostly through the COMMON blocks, but partly via their arguments. Each of the user-supplied subroutines must contain the double precision card

```
IMPLICIT REAL*8(A-H,O-Z).
```

The subroutines RESTNT, GRAD1, and MATRIX must contain the COMMON card

```
COMMON/SHARE/X(45), DEL(45), A(45,45), N, M, MN, NP1, NM1.
```

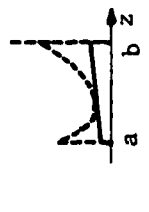
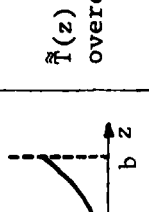
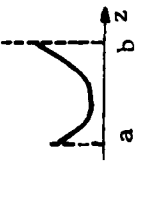
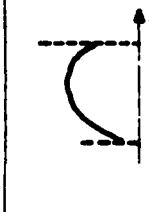
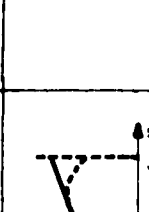
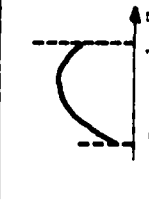
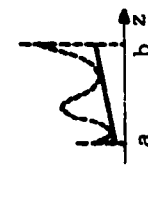
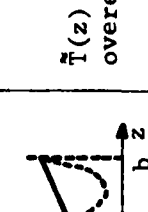
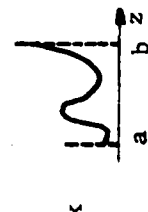
Also, depending on the problem being solved, the following COMMON card may be required in some or all of the user subroutines:

```
COMMON/SEN/PAR(45), DPAR(45), NFAR, ISENS.
```

When bound analysis is desired, subroutine READIN must contain the COMMON card

```
COMMON/ABG/LY, LZ, PER(45).
```

TABLE 1
 CONVEX nuf AND CONCAVE nuf OF THE SINGLE VARIABLE FUNCTION T(z) OVER THE CLOSED INTERVAL $a \leq z \leq b$

T(z)	T(z) [convex nuf of T(z)]	$\tilde{T}(z)$ [concave nuf of T(z)]
$a \leq z \leq b$	$a \leq z \leq b$	$a \leq z \leq b$
Graph	Graph	Graph
Convex 	$\tilde{T}(z) = T(z)$ 	$\tilde{T}(z) = \text{any line not overestimating } T(z)$ 
Concave 	$T(z) = \text{any line not underestimating } T(z)$ 	$\tilde{T}(z) = T(z)$ 
Neither convex nor concave 	$T(z) = \text{any line not underestimating } T(z)$ 	$\tilde{T}(z) = \text{any line not overestimating } T(z)$ 

In the following section we discuss the purpose and coding procedure of each of these user-supplied subroutines. The general description and listing of the rest of the subroutines comprising SENSUMT will be given in Section 6.

4.1.1 READIN

This subroutine allows the user to read in the data which is necessary for evaluation of the problem functions and convey it, through the COMMON blocks, to the related subroutines. The problem parameters for which sensitivity and bound calculation are desired must also be defined and initialized in this subroutine. The initial values of the parameters must be placed in array PAR. The corresponding perturbation values for bound calculation must be placed in array PER. NPAR in this subroutine must be initialized to the total number of problem parameters on which sensitivity is desired. *Notice that the presence of array PER in this subroutine will trigger SENSUMT to calculate the desired bound calculation.* READIN may also be used to read in and print out any pertinent information of the user's choice, such as problem title, date, and so on. This subroutine is called only once for each problem being solved. For clarity, see Example 1 and its code in Figure 2. Any input data read by subroutine READIN must be placed immediately after the first option card.

4.1.2 RESTNT (IN,VAL)

This subroutine is used to read in the problem functions. For $IN=0$, set $VAL=f(x)$ and for $IN \neq 0$ set

$$VAL = g_{IN}(x) , \quad IN=1,2,\dots,m$$

$$VAL = h_{IN}(x) , \quad IN=m+1,\dots,m+p .$$

For clarity see Example 1 and Figure 2.

4.1.3 GRAD1 (IN)

This subroutine is used to read in the gradients of the problem functions. For IN=0 set

$$\text{DEL}(J) = f(x)/\partial x_J, \quad J=1,2,\dots,n$$

and for IN≠0 set

$$\text{DEL}(J) = \begin{cases} \partial g_{\text{IN}}(x)/\partial x_J, & J=1,2,\dots,n \\ \partial h_{\text{IN}}(x)/\partial x_J, & J=1,2,\dots,n \end{cases}$$

For an illustration, see Example 1 and Figure 2.

SENSUMT can internally compute numerical approximations for some or all of the gradients. To utilize this option, instead of coding the gradients code the statements

```
CALL DIFF1(IN)
RETURN
```

for any desired value of the variable IN (the index of the problem functions, with IN=0 corresponding to the objective function $f(x)$). Because of the computational effort involved in numerical differencing, this option may be prohibitive for large problems.

4.1.4 MATRIX (IN,K)

This subroutine reads in the *upper triangle and diagonal elements* of the Hessian matrix of the problem functions. For IN=0, set

$$A(I,J) = \partial^2 f(x)/\partial x_I \partial x_J, \quad I=1,2,\dots,n; J=1,2,\dots,n; \text{ and } I \leq J.$$

For IN≠0 set

$$A(I,J) = \begin{cases} \partial^2 g_{\text{IN}}(x)/\partial x_I \partial x_J, & \text{IN}=1,2,\dots,m \\ \partial^2 h_{\text{IN}}(x)/\partial x_I \partial x_J, & \text{IN}=m+1,\dots,p \end{cases}$$

where $I=1,2,\dots,n$; $J=1,2,\dots,n$; and $I \leq J$. Before the call is made to this subroutine, all entries of the matrix $A(I,J)$ are set to zero.

The second argument K of the subroutine MATRIX is provided so that the user may communicate to SENSUMT that the Hessian of a problem function is identically zero. Set $K=1$ if the second partial derivatives of the IN th constraint are zero ($IN=0$ corresponds to the objective function). For an illustration, see Example 1 and Figure 2.

SENSUMT can internally compute numerical approximations for some or all of the Hessian matrices. To utilize this option, instead of coding the Hessian matrices, code the statements

```
CALL DIFF2(IN)
RETURN
```

for any desired value of the variable IN (the index of the problem function, with $IN=0$ designating the objective function). *Although this option can be used in calculating an optimal solution, the current sensitivity routines require explicit coding of the closed form of the Hessian matrices. Thus, this option cannot be used for sensitivity and bound calculations.* Again, because of the computational effort involved in numerical differencing, use of this option is not advised for large problems.

4.2 User's Information Cards

The other inputs required by SENSUMT are the user's information cards, i.e.,

- PARAMETER CARD
- INITIAL VECTOR CARD(S)
- FIRST OPTION CARD
- TOLERANCE CARD
- SECOND OPTION CARD.

In the following we specify the type of information, along with the corresponding format, name, and description, which must be provided by the user to SENSUMT via the above cards. All of this information is read by the program MAIN, which sets up the SENSUMT algorithm to solve and

analyze the problem under study. Tables 2, 3, and 4, which follow, are taken from [13].

4.2.1 Parameter card.

Input specifications of the parameter card are shown in Table 2.

TABLE 2
PARAMETER CARD

Columns	Format	Name	Use
01-12	E12.0	EPSI (ϵ)	Tolerance used to decide if an unconstrained minimum has been achieved for each subproblem [see Option 9]
13-24	E12.0	RHØIN (r_1)	Possible initial value of r (often set at 1.0) [see Option 1]
25-36	E12.0	THETAO (θ_0)	Tolerance used to decide if the solution to the NLP problem $P(\epsilon)$ has been approximated [see Option 5]
37-48	E12.0	RATIO (c)	Parameter (>1) used to compute consecutive values of r ; $r_{i+1} = r_i/c$ (often set at 16.0)
49-60	E12.0	TMMAX	Maximum amount of time for solving problem (in seconds)
61-64	I4	M	Number (integer) of nontrivial constraints [see Option 2] $(M+MZ) \leq 200^*$
65-68	I4	N	Number (integer) of variables, $N \leq 45$
69-72	I4	MZ	Number (integer) of equality constraints

*The limits on $M+MZ$ and N are governed by the size of the arrays in SENSUMT.

4.2.2 Initial vector card(s).

The cards designating the initial starting point immediately follow the parameter card. There are six components per card, requiring $n/6$ cards for the initial vector. Each card has the format 6E12.6 .

4.2.3 First option card.

The input specifications for the first option card are shown in Table 3. This card must immediately follow the user-supplied subroutines.

4.2.4 Tolerance card.

The input specifications for the tolerance card are given in Table 4. This card must immediately follow the first option card.

4.2.5 Second option card.

The input specifications for the second option card, which immediately follows the tolerance card, are given in Table 5.

Figure 1 shows the arrangement of the data together with JOB and JCL cards in a coded deck.

TABLE 3
FIRST OPTION CARD

Option	Column	Value	Meaning
1 (normally set to 3)	7	=1	The value for r is made by finding an approximate solution $\min\{[\nabla W(x^0, r)[\nabla^2 W(x^0, r)]^{-1}\nabla W(x^0, r)\}$ which is a good approximation only when x^0 is close to the boundary (i.e., for some i, $g_i(x)$ is close to zero) or when $\nabla^2 f(x^0)=0$ and when $MZ=0$.
		=2	r_1 is given by formula 8.65 [10, p. 191] (can only be used when $MZ=0$).
		=3	$r_1 = RH\emptyset IN$ (see parameter card).
2	14	=1	The requirements (trivial constraints) that $x_i \geq 0$ for $i=1, \dots, n$ are to be automatically included in the problem.
		=2	The only constraints on the problem are those inputted by the user.
3 (normally set to 1)	21	=1	Standard printout (this includes a call to OUTPUT after the solution of every subproblem). Also the estimates of the "Lagrange multipliers" and first and second order solution estimates are printed.
		=2	For additional printout (includes standard printout and every intermediate point, gradient of P and the vector S).
4 (normally set to 1)	28	=1	Final convergence is determined on the basis of current solution to the subproblem.
		=2	Final convergence is determined on the basis of the first order estimates. The first order estimate of the solution vector must be close to feasible. See below.
		=3	Final convergence is determined on the basis of the second order estimates. The second order estimate of the solution vector must be close to feasible before the convergence check is made. If \bar{x} is a solution

Option	Column	Value	Meaning
5 (normally set to 1) (see [10, p. 195])	35		estimate it is considered close to being feasible if $g_i(\bar{x}) + \theta_0 \geq 0$, $i=1,2,\dots,m$, where θ_0 is defined on the parameter card.
			The convergence criterion determining the NLP problem has been solved (only use =1, when NT4#1).
		=1	Quit when $\frac{G - f[x(r_k)]}{G[x(r_k), \mu(r_k), \lambda(r_k)]} < \theta_0$
		=2	Quit when $ r \sum_{j=1}^m \lambda n g_j[x(r_k)] < \theta_0$
		=3	Quit when $\frac{\text{first order estimate of } v_0 - 1}{G[x(r_k), \mu(r_k), \lambda(r_k)]} < \theta_0$
6 (normally set to 1)	42	=1	After final convergence the program reads in new data and solves the next problem.
		=2	After final convergence has been determined a call to PUNCH is made before proceeding on to the next problem.
7 (normally set to 1)	49		First move after a minimum to a subproblem is achieved
		=1	No extrapolation.
		=2	Extrapolate through last two minima.
		=3	Extrapolate through last three minima.
8	56		Not used.
9	63		Subproblem convergence criterion, or when to stop minimizing P function for fixed value of r (see parameter card).
		=1	Quit when $ \nabla_x W^T(x^i, r) \left[\frac{\partial^2 W(x, r)}{\partial x_i \partial x_j} \right]^{-1} \nabla_x W(x^i, r) < \epsilon$.
		=2	Quit when $ \nabla_x W^T(x^i, r) \left[\frac{\partial^2 W(x, r)}{\partial x_j \partial x_j} \right]^{-1} \nabla_x W(x^i, r) < \frac{W(x^{i-1}) - W(x^i)}{5}$.

Table 3--continued

Option	Column	Value	Meaning
10 ^a	70	=3	Quit when $ \nabla_x W(x^i, r) < \epsilon$
		=1	At least one nonlinear constraint
		=2	Linear constraints
		=3	Linear constraints and linear objective function (i.e., a linear programming problem)

^aWhen option 10=3, MATRIX (the user subroutine supplying the second partial derivatives) will not be called, and when option 10=2 it will be called only to get the second partials of f(x).

TABLE 4

TOLERANCE CARD

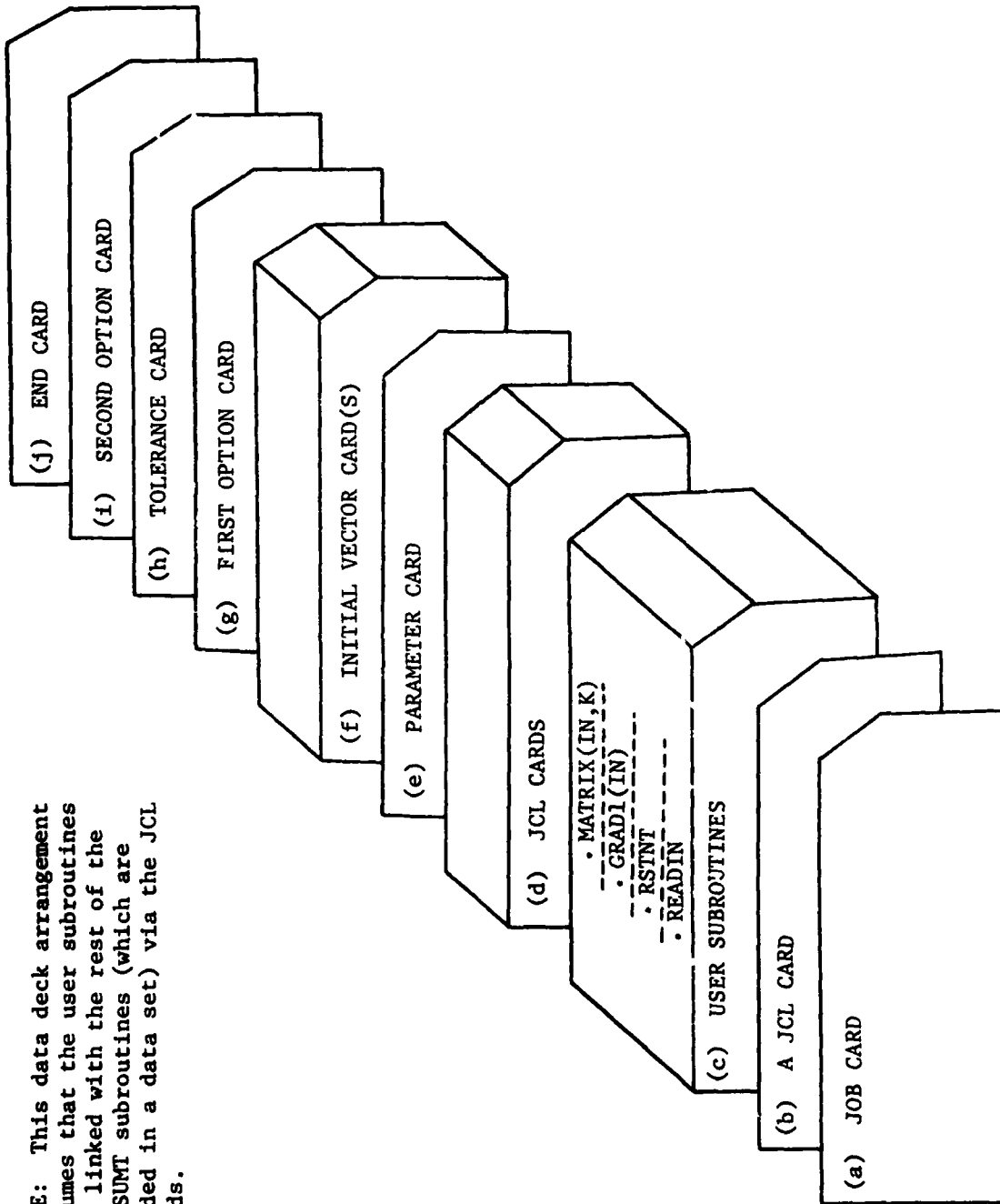
Columns	Format	Name	Description
01-12	E12.6	XEP1	If some first or second derivatives are to be gotten by numerical differencing (see addition option card) this is the value used to compute them. See the description of DIFF1. (Usually setting XEP1 equal to .0001 is satisfactory, although a good value is dependent on the scaling of the problem.)
13-24	E12.6	XEP2	When minimizing the W function for a given value of r (RH0) the value of W must decrease by an amount exceeding XEP2 for each iteration after the first. If it does not, then the code prints out the message "apparently roundoff errors prevent a more accurate determination of the minimum of this subproblem," and it is assumed a minimum has been found. (Usually we set XEP2 equal to 0.)

TABLE 5
SECOND OPTION CARD

Option	Column	Value	Meaning
1 (normally set to 4)	7	=1	Solve problems without checking derivatives
		=2	Solve problem after checking only first derivatives
		=3	Do not solve problem after checking only first derivatives
		=4	Solve problem after checking first and second derivatives
		=5	Do not solve problem after checking first and second derivatives
2	14	=1	The method for minimizing the unconstrained penalty function is to be the generalized Newton-Raphson method as modified to handle indefinite Hessian matrices. This method requires function values, first and second derivatives.
		=2	Same as 1, except that when an "orthogonal move" is made because of an indefinite Hessian matrix, $-\nabla P$ is added to the orthogonal move vector.
		=3	Steepest descent is used to minimize P-function.
		=4	The method for minimizing the unconstrained penalty function is McCormick's modification of the Fletcher-Powell method as reported in [10]. This requires function values and first derivatives.
3	21	=0	Do not conduct a sensitivity analysis.
		=1	Conduct a sensitivity analysis at the final subproblem with a fixed value for the differencing interval.
		=2	Conduct a sensitivity analysis at each subproblem along the minimizing trajectory with the value of the differencing interval depending on the particular subproblem.
		=3	Conduct a sensitivity analysis at the final subproblem for a range of differencing intervals.

Table 5--continued

Option	Column	Value	Meaning
4	28	=0	Do not estimate the partial derivatives of the estimates of the Lagrange multipliers.
		=1	Estimate the partial derivatives of the estimates of the Lagrange multipliers whenever a sensitivity analysis of the solution point is conducted.
5	35	=0	Estimate the partial derivatives of the optimal value function and eliminate those parameters which do not affect the optimal value functions from subsequent sensitivity calculations.
		=1	Estimate the partial derivatives of the optimal value function with respect to all parameters, but continue all subsequent sensitivity calculations with respect to all parameters.
		=2	Do not estimate the partial derivatives of the optimal value function first. Conduct the sensitivity analysis with respect to all parameters.
6	42	=0	Do not transform the results.
		=1	The problem being solved, $P(x)$, is a convex equivalent of a geometric programming problem $G(t)$ obtained by transformation $t_i = e^{-x_i}$. Thus back transform the results to t space.
7	49	=1	$f^*(\epsilon)$ is convex. Derive parametric upper and lower bounds on $f^*(\epsilon)$.
		=2	$f^*(\epsilon)$ is convex. Derive parametric upper bound on $f^*(\epsilon)$.
		=3	$f^*(\epsilon)$ is convex. Derive parametric lower bound on $f^*(\epsilon)$.
		=4	$f^*(\epsilon)$ is concave. Derive parametric upper and lower bounds on $f^*(\epsilon)$.
		=5	$f^*(\epsilon)$ is concave. Derive parametric upper bound on $f^*(\epsilon)$.
		=6	$f^*(\epsilon)$ is concave. Derive parametric lower bound on $f^*(\epsilon)$.



NOTE: This data deck arrangement assumes that the user subroutines are linked with the rest of the SENSUMT subroutines (which are loaded in a data set) via the JCL cards.

Figure 1.--Data deck structure for SENSUMT.

5. Coded Examples and Input/Output Illustrations

In this section we provide the input listing and output description of two illustrative examples which are taken from [11].

5.1 Example 1

Consider the following convex RHS programming problem:

$$\begin{aligned} \text{minimize } f(x) &= (x_1 - 4)^2 + (x_2 - 2)^2 \\ \text{subject to } g_1(x) &= -x_1^2 + x_2 \geq \epsilon_1 \\ g_2(x) &= -x_1 - x_2 \geq \epsilon_2 . \end{aligned}$$

It is desired to solve and analyze this problem for sensitivity when $\epsilon_1 = 0$ and $\epsilon_2 = -3$. Moreover, it is desired to derive parametric upper and lower bounds on the optimal value function of this problem when

- (i) $-1 \leq \epsilon_1 \leq 0$ while $\epsilon_2 = -3$, and
- (ii) $-3 \leq \epsilon_2 \leq -1$ while $\epsilon_1 = 0$.

5.1.1 Computer listing of the code deck.

Figure 2 shows the computer listing of the code of Example 1. The letters in circles categorizing the input data correspond to those indicated in the code deck structure depicted in Figure 1. The format of the listed data is given in Section 4.

5.1.2 Selected pages from the computer output.

Figure 3 lists an annotated computer output for Example 1. An explanation of the meaning of the corresponding steps follows.

<pre> J=GHAEHL T=3 I=3 // EXEC FORG1 SUBROUTINE READIN IMPLICIT REAL*8(A-H,O-Z) COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS COMMON/ABG/LY,LZ,PER(45) NPAR=2 PAR(1)=0. PAR(2)=-3.0 PER(1)=-1. PER(2)=2. RETURN END </pre>	<p>① Job card</p>
<pre> SUBROUTINE RSTNT(IN,VAL) IMPLICIT REAL*8(A-H,O-Z) COMMON/SHARE/X(45),DEL(45), A(45,45),N,M,MN,NPI,NMI COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS COMMON/ABG/LY,LZ,PER(45) I=IN+1 GO TO (20,1,2),I 20 VAL=(X(I)-4.)**2+(X(2)-2.)**2 RETURN 1 VAL=-X(I)+2+X(2)-PAR(1) RETURN 2 VAL=-X(I)-X(2)-PAR(2) RETURN END </pre>	<p>② A JCL card</p> <p>③ User subroutines</p> <p>Subroutine READIN</p>
<pre> SUBROUTINE GRAD(IN) IMPLICIT REAL*8(A-H,O-Z) COMMON/SHARE/X(45),DEL(45), A(45,45),N,M,MN,NPI,NMI COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS J=IN+1 DO 5 I=1,N DEL(I)=0.0 5 CONTINUE GO TO (20,1,2),J 20 DEL(1)=2.*(X(1)-4.) DEL(2)=2.*(X(2)-2.) RETURN 1 DEL(1)=-2.*X(1) DEL(2)=1.C RETURN 2 DEL(1)=-1.0 DEL(2)=-1.0 RETURN END </pre>	<p>Subroutine RSTNT (IN,VAL)</p>
<pre> SUBROUTINE MATRIX(IN,K) IMPLICIT REAL*8(A-H,O-Z) COMMON/SHARE/X(45),DEL(45), A(45,45),N,M,MN,NPI,NMI COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS L=IN+1 GO TO (20,1,2),L 20 A(1,1)=2.0 A(1,2)=0. A(2,2)=2.0 RETURN 1 A(1,1)=-2.0 A(1,2)=0. A(2,2)=0. RETURN 2 K=1 RETURN END </pre>	<p>Subroutine GRAD (IN)</p>
<pre> // EXEC FORG6,DSN=OR799661.BOUNDS45,PRGM=MAIN //GO,SYSLIB DD // DD // DD DSN=GNU.FORTLIB,DISP=SHR //GO,SYSLIN DD // DD DSN=&TEMPUNCH,DISP=(OLD,DELETE) //F107F001 DD SYS.WT=A 1.0E-08 10.0 1.0E-06 4.0 900.0 2 2 0 3 2 1 1 1 1 1 1 0.1E-10 0.00 1 1 0 1 // </pre>	<p>Subroutine MATRIX (IN,K)</p>
<pre> // EXEC FORG6,DSN=OR799661.BOUNDS45,PRGM=MAIN //GO,SYSLIB DD // DD // DD DSN=GNU.FORTLIB,DISP=SHR //GO,SYSLIN DD // DD DSN=&TEMPUNCH,DISP=(OLD,DELETE) //F107F001 DD SYS.WT=A 1.0E-08 10.0 1.0E-06 4.0 900.0 2 2 0 3 2 1 1 1 1 1 1 0.1E-10 0.00 1 1 0 1 // </pre>	<p>④ JCL cards</p>
<pre> 1.0E-08 10.0 1.0E-06 4.0 900.0 2 2 0 3 2 1 1 1 1 1 1 0.1E-10 0.00 1 1 0 1 // </pre>	<p>⑤ Parameter card</p>
<pre> 3 2 1 1 1 1 1 1 0.1E-10 0.00 1 1 0 1 // </pre>	<p>⑥ Initial vector card</p>
<pre> 0.1E-10 0.00 1 1 0 1 // </pre>	<p>⑦ First option card</p>
<pre> 0.1E-10 0.00 1 1 0 1 // </pre>	<p>⑧ Tolerance card</p>
<pre> 0.1E-10 0.00 1 1 0 1 // </pre>	<p>⑨ Second option card</p>
<pre> 0.1E-10 0.00 1 1 0 1 // </pre>	<p>⑩ End card</p>

Figure 2.--Computer listing of the code for Example 1.

```

1  NONLINEAR PROGRAMMING SUBROUTINE-SUMT VERSION 4 03/01/73
   N= 2  M= 2  NZ= 0
   MAX. TIME= 3.900000E 03  R= 0.1000000E 02  KATIO= 0.4000000E 01  EPSILON= 0.999997E-08  THETA= 0.999999E-06
   OPTIUMS SELECTED 1 1 1 1 1 0 1 1 1
   TOLERANCES 0.0 0.999999E-03
   SECOND SET OF OPTIONS
   5 TIME= 0.059 SECONDS 1 1 0 1
   P= 0.9250000 01  P= 0.0  G= 0.0  RSIGMA= 0.0  H= 0.0
   THE CURRENT VALUE OF X IS
   0.1000000 01
   THE CONSTRAINT VALUES
   0.5000000 00 0.5000000 00
   TIME= 0.102 SECONDS

7 *****THE FEASIBLE STARTING POINT TO BE USED IS ***
   F= 0.9250000 01  P= 0.0  G= 0.0  RSIGMA= 0.0  H= 0.0
   THE CURRENT VALUE OF X IS
   0.1000000 01 0.1500000 01
   THE CONSTRAINT VALUES
   0.5000000 00 0.5000000 00
   APPARENTLY ROUNDOFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
   TIME= 0.219 SECONDS

8 *****
   POINT= 4  ODI= 0.40671930-06  RHO= 0.1000000 02  MAGNITUDE= 0.19361530-02  PHASE= 2
   THE CURRENT VALUE OF X IS  P= 0.80717330 01  G= -0.42197120 01  RSIGMA= -0.77085550 01  H= 0.0
   0.51007220-01 0.15690190 01
   THE CONSTRAINT VALUES
   0.15666170 01 0.13799740 01

9 LAGRANGE MULTIPLIERS
   F= 0.15780290 02  P= 0.80717330 01  G= -0.42197120 01  RSIGMA= -0.77085550 01  H= 0.0
   THE CURRENT VALUE OF X IS
   0.51007220-01 0.15690190 01
   THE CONSTRAINT VALUES
   0.62839040 01 0.72465150 01
   APPARENTLY ROUNDOFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
   TIME= 0.340 SECONDS

10 *****
   POINT= 7  ODI= 0.23087850-04  RHO= 0.2500000 01  MAGNITUDE= 0.18934470-01  PHASE= 2
   THE CURRENT VALUE OF X IS  P= 0.11665690 02  G= 0.61115940 01  RSIGMA= 0.55609590 00  H= 0.0
   0.69826920 00 0.15615400 01
   THE CONSTRAINT VALUES
   0.10539800 01 0.76017110 00

11 1ST ORDER ESTIMATES
   F= 0.97618970 01  P= 0.12863680 02  G= 0.95533620 01  RSIGMA= 0.0  H= 0.0
   THE CURRENT VALUE OF X IS
   0.91402310 00 0.15124070 01
   THE CONSTRAINT VALUES
   0.00000000 00 0.00000000 00

```

Figure 3.--Annotated computer output for Example 1.

LAGRANGE MULTIPLIERS
 P= 0.9741897D 01 P= 0.1286368D 02 G= 0.9555362D 01 ASIGNA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.9140231D 00 0.1532407D 01
 THE CONSTRAINT VALUES
 0.2371962D 01 0.3288733D 01
 APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
 TIME= 0.469 SECONDS

12 POINT= 10 D017= 0.7459573D-04 RHO= 0.4250000D 00 MAGNITUDE= 0.4996642D-01 PHASE= 2
 G= 0.7261950D 01 ASIGNA= 0.1394476D 01 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1106836D 01 0.1623767D 01
 THE CONSTRAINT VALUES
 0.3986811D 00 0.2693975D 00

13 2ND ORDER ESTIMATES
 F= 0.7596676D 01 P= 0.9083760D 01 G= 0.7645402D 01 RSIGNA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1264958D 01 0.1659086D 01
 THE CONSTRAINT VALUES
 0.5896728D-01 0.7595534D-01

1ST ORDER ESTIMATES
 F= 0.7722596D 01 P= 0.9320005D 01 G= 0.7645402D 01 RSIGNA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1243025D 01 0.1651169D 01
 THE CONSTRAINT VALUES
 0.1060589D 00 0.1058063D 00

LAGRANGE MULTIPLIERS
 F= 0.7722596D 01 P= 0.9320005D 01 G= 0.7645402D 01 RSIGNA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1243025D 01 0.1651169D 01
 THE CONSTRAINT VALUES
 0.1567669D 01 0.2319992D 01
 APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
 TIME= 0.750 SECONDS

14 POINT= 13 D017= 0.1185880D-03 RHO= 0.1562500D 00 MAGNITUDE= 0.1157642D 00 PHASE= 2
 G= 0.7671997D 01 P= 0.8614190D 01 ASIGNA= 0.7421929D 00 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1249429D 01 0.1673076D 01
 THE CONSTRAINT VALUES
 0.1128031D 00 0.7669464D-01

2ND ORDER ESTIMATES
 F= 0.7381120D 01 P= 0.7823229D 01 G= 0.7392012D 01 RSIGNA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1300556D 01 0.1693207D 01
 THE CONSTRAINT VALUES
 0.1761123D-02 0.6223730D-02

1ST ORDER ESTIMATES
 F= 0.7462169D 01 P= 0.7916780D 01 G= 0.7392012D 01 RSIGNA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1300556D 01 0.1693207D 01

Figure 3.--continued

LAGRANGE MULTIPLIER VALUES
J.12460360-01

LAGRANGE MULTIPLIERS
F= 0.74221650 01 P= 0.79167780 01 G= 0.73920120 01 R= 0.000000 00 H= 0.000000 00
THE CURRENT VALUE OF X IS
0.12968633 01 P= 0.16905790 01
THE CONSTRAINT VALUES
J.13851570 01 0.20473000 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 0.879 SECONDS

POINT= 15 DOTT= 0.75072110-03 RHO= 0.39062500-01 MAGNITUDE= 0.34600050 00 PHASE= 2
F= 0.74436570 01 P= 0.77354740 01 G= 0.73655320 01 R= 0.000000 00 H= 0.000000 00
THE CURRENT VALUE OF X IS
0.12892490 01 P= 0.16909900 01
THE CONSTRAINT VALUES
0.28827073-01 0.19760510-01

2ND ORDER ESTIMATES
F= 0.73681310 01 P= 0.74820660 01 G= 0.73675440 01 R= 0.000000 00 H= 0.000000 00
THE CURRENT VALUE OF X IS
0.13028930 01 P= 0.16971030 01
THE CONSTRAINT VALUES
-0.42809230-03 0.39366340-05

1ST ORDER ESTIMATES
F= 0.73683790 01 P= 0.75092360 01 G= 0.73675440 01 R= 0.000000 00 H= 0.000000 00
THE CURRENT VALUE OF X IS
0.13025220 01 P= 0.16966950 01
THE CONSTRAINT VALUES
0.13032880-03 0.78246310-03

LAGRANGE MULTIPLIERS
F= 0.73683790 01 P= 0.75092360 01 G= 0.73675440 01 R= 0.000000 00 H= 0.000000 00
THE CURRENT VALUE OF X IS
0.13025220 01 P= 0.16966950 01
THE CONSTRAINT VALUES
J.13550630 01 0.19767970 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 1.000 SECONDS

POINT= 17 DOTT= 0.18547470-04 RHO= 0.97656250-02 MAGNITUDE= 0.10391300 00 PHASE= 2
F= 0.73861610 01 P= 0.74858740 01 G= 0.73666300 01 R= 0.000000 00 H= 0.000000 00
THE CURRENT VALUE OF X IS
0.12393440 01 P= 0.16956500 01
THE CONSTRAINT VALUES
0.73495540-02 0.50044530-02

2ND ORDER ESTIMATES
F= 0.73669620 01 P= 0.73955690 01 G= 0.73669960 01 R= 0.000000 00 H= 0.000000 00
THE CURRENT VALUE OF X IS
0.13027240 01 P= 0.16972360 01
THE CONSTRAINT VALUES
0.16605230-03 0.39322420-04

1ST ORDER ESTIMATES
F= 0.73669620 01 P= 0.74000000 01 G= 0.73669960 01 R= 0.000000 00 H= 0.000000 00

Figure 3.--continued

THE CURRENT VALUE OF X IS
0.13027173 01 0.16972030 01
THE CONSTRAINT VALUES
0.16507197-03 0.05766720-04

LAGRANGE MULTIPLIERS
F= 0.73670510 01 P= 0.74026730 01 M= 0.0
THE CURRENT VALUE OF X IS
0.13027173 01 0.16972030 01
THE CONSTRAINT VALUES
0.13287370 01 0.19513870 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 1.102 SECONDS

17
POINT= 19 DOTT= 0.20282430-05 PHASE= 2
F= 0.73715670 01 P= 0.74032310 01 M= 0.0
THE CURRENT VALUE OF X IS
0.13019160 01 0.16968320 01
THE CONSTRAINT VALUES
0.18663760-02 0.12518430-02
RHO= 0.24414060-02 MAGNITUDE= 0.70276280-01
G= 0.73666950 01 RSIGMA= 0.31683760-01

2ND ORDER ESTIMATES
F= 0.73666830 01 P= 0.73739130 01 M= 0.0
THE CURRENT VALUE OF X IS
0.13027770 01 0.16972280 01
THE CONSTRAINT VALUES
-0.21372170-07 -0.46807200-05
RHO= 0.73667030 01
G= 0.73667030 01 RSIGMA= 0.0

1ST ORDER ESTIMATES
F= 0.73667060 01 P= 0.73757100 01 M= 0.0
THE CURRENT VALUE OF X IS
0.13027730 01 0.16972260 01
THE CONSTRAINT VALUES
0.13222670 01 0.19502500 01
0.90672090-05 0.97236950-06
RHO= 0.73667030 01
G= 0.73667030 01 RSIGMA= 0.0

LAGRANGE MULTIPLIERS
F= 0.73667060 01 P= 0.73757100 01 M= 0.0
THE CURRENT VALUE OF X IS
0.13027730 01 0.16972260 01
THE CONSTRAINT VALUES
0.13222670 01 0.19502500 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 1.219 SECONDS

18
POINT= 21 DOTT= 0.55388060-06 PHASE= 2
F= 0.73679130 01 P= 0.73775240 01 M= 0.0
THE CURRENT VALUE OF X IS
0.13025600 01 0.16971260 01
THE CONSTRAINT VALUES
0.46235060-03 0.31344850-03
RHO= 0.61035160-03 MAGNITUDE= 0.80343090-01
G= 0.73666920 01 RSIGMA= 0.96112440-02

2ND ORDER ESTIMATES
F= 0.73666940 01 P= 0.73684970 01 M= 0.0
THE CURRENT VALUE OF X IS
0.13027793 01 0.16972240 01
THE CONSTRAINT VALUES
0.27612570-06 0.62494790-06
RHO= 0.73666940 01
G= 0.73666940 01 RSIGMA= 0.0

Figure 3.--continued

```

LAGRANGE MULTIPLIERS
  F = 0.7366990 01  P = 0.7366990 01  RHO = 0.15258790-03  MAGNITUDE = 0.20081460 00  PHASE = 2
  THE CURRENT VALUE OF X IS  G = 0.73666880 01  RSIGMA = 0.0  M = 0.0
  0.13027750 01  0.16972250 01
  THE CONSTRAINT VALUES
  0.13201063 01  0.19672150 01
  APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
  TIME = 1.371 SECONDS

```

```

LAGRANGE MULTIPLIERS
  F = 0.7366990 01  P = 0.7366990 01  G = 0.73666940 01  RHO = 0.15258790-03  MAGNITUDE = 0.20081460 00  PHASE = 2
  THE CURRENT VALUE OF X IS  G = 0.73666880 01  RSIGMA = 0.0  M = 0.0
  0.13027750 01  0.16972250 01
  THE CONSTRAINT VALUES
  0.13201063 01  0.19672150 01
  APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
  TIME = 1.371 SECONDS

```

19

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*****
POINT = 23  DOIT = 0.62757270-06  RHO = 0.15258790-03  MAGNITUDE = 0.20081460 00  PHASE = 2
  F = 0.7366990 01  P = 0.7366990 01  G = 0.73666880 01  RSIGMA = 0.28304970-02  M = 0.0
  THE CURRENT VALUE OF X IS
  0.13027230 01  0.16972000 01
  THE CONSTRAINT VALUES
  0.11404160-03  0.77054420-04

```

```

2ND ORDER ESTIMATES
  F = 0.73666860 01  P = 0.73671440 01  G = 0.73666860 01  RSIGMA = 0.0  M = 0.0
  THE CURRENT VALUE OF X IS
  0.13027770 01  0.16972250 01
  THE CONSTRAINT VALUES
  -0.22661800-05  -0.19032050-05

```

```

1ST ORDER ESTIMATES
  F = 0.73666860 01  P = 0.73672560 01  G = 0.73666860 01  RSIGMA = 0.0  M = 0.0
  THE CURRENT VALUE OF X IS
  0.13027770 01  0.16972250 01
  THE CONSTRAINT VALUES
  -0.20730320-05  -0.17436010-05

```

```

LAGRANGE MULTIPLIERS
  F = 0.73666860 01  P = 0.73672560 01  G = 0.73666860 01  RSIGMA = 0.0  M = 0.0
  THE CURRENT VALUE OF X IS
  0.13027770 01  0.16972250 01
  THE CONSTRAINT VALUES
  0.13300270 01  0.13302610 01
  APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
  TIME = 1.461 SECONDS

```

20

```

*****
POINT = 25  DOIT = 0.68064420-07  RHO = 0.16146970-04  MAGNITUDE = 0.81242480-01  PHASE = 2
  F = 0.73667630 01  P = 0.73675810 01  G = 0.73666920 01  RSIGMA = 0.81243640-03  M = 0.0
  THE CURRENT VALUE OF X IS
  0.13027620 01  0.16972180 01
  THE CONSTRAINT VALUES
  0.28863580-04  0.19511670-04

```

```

2ND ORDER ESTIMATES
  F = 0.73666940 01  P = 0.7366990 01  G = 0.73666940 01  RSIGMA = 0.0  M = 0.0
  THE CURRENT VALUE OF X IS
  0.13027750 01  0.16972250 01
  THE CONSTRAINT VALUES
  0.46824440-04

```

Figure 3.--continued

1ST ORDER ESTIMATES
 F= 0.7366690D 01 P= 0.7366833D 01 G= 0.7366694D 01 RSIGMA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1302775D 01
 THE CONSTRAINT VALUES
 0.4661930D-06 0.3307564D-06

LAGRANGE MULTIPLIERS
 F= 0.7366690D 01 P= 0.7366833D 01 G= 0.7366694D 01 RSIGMA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1302775D 01
 THE CONSTRAINT VALUES
 0.1321768D 01 0.1955085D 01
 APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
 TIME= 1.621 SECONDS

21
 POINT= 27 DOT= 0.5385601D-07 PHASE= 2
 F= 0.7366711D 01 P= 0.7366691D 01 RHO= 0.9536743D-05 MAGNITUDE= 0.2332791D 00
 THE CURRENT VALUE OF X IS G= 0.7366692D 01 RSIGMA= 0.2298500D-03 H= 0.0
 0.1302776D 01
 THE CONSTRAINT VALUES
 0.1302772D 01 0.1697223D 01
 THE CONSTRAINT VALUES
 0.7118220D-05 0.4791486D-05

2ND ORDER ESTIMATES
 F= 0.7366692D 01 P= 0.7366721D 01 G= 0.7366692D 01 RSIGMA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1302776D 01
 THE CONSTRAINT VALUES
 -0.1689720D-06 -0.1449764D-06

1ST ORDER ESTIMATES
 F= 0.7366692D 01 P= 0.7366728D 01 G= 0.7366692D 01 RSIGMA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1302776D 01
 THE CONSTRAINT VALUES
 -0.1292761D-06 -0.1152431D-06

LAGRANGE MULTIPLIERS
 F= 0.7366692D 01 P= 0.7366728D 01 G= 0.7366692D 01 RSIGMA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1302776D 01
 THE CONSTRAINT VALUES
 0.1339765D 01 0.1990392D 01
 APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
 TIME= 1.711 SECONDS

22
 POINT= 29 DOT= 0.5868017D-07 PHASE= 2
 F= 0.7366697D 01 P= 0.7366761D 01 RHO= 0.2384186D-05 MAGNITUDE= 0.7124525D 00
 THE CURRENT VALUE OF X IS G= 0.7366692D 01 RSIGMA= 0.6390440D-04 H= 0.0
 0.1302775D 01 0.1697224D 01
 THE CONSTRAINT VALUES
 0.1826825D-05 0.1252313D-05

2ND ORDER ESTIMATES
 F= 0.7366693D 01 P= 0.7366699D 01 G= 0.7366693D 01 RSIGMA= 0.0 H= 0.0
 THE CURRENT VALUE OF X IS
 0.1302775D 01

Figure 3.--continued

J. 736693D-01

L.V. OBJECTIVE FUNCTION VALUE OF X IS
 P = 0.736693D 01 P = 0.7366701D 01 G = 0.7366693D 01 H = 0.0
 THE CURRENT VALUE OF X IS
 J.1302776J 01 J.1697224D 01
 THE CONSTRAINT VALUES
 J.0302841J-07 J.725R8310-07

LAGRANGE MULTIPLIERS
 P = 0.736693D 01 P = 0.7366701D 01 G = 0.7366693D 01 H = 0.0
 THE CURRENT VALUE OF X IS
 0.1302776J 01 0.1697224D 01
 THE CONSTRAINT VALUES
 J.1305098D 01 0.1903026D 01

SENSITIVITY ANALYSIS

(23) THE VALUE OF R(RHO) IS 0.23842D-05
 THE POINT AT WHICH THE ESTIMATE OF SENSITIVITY WILL BE MADE IS
 X(1) = 1.302775 X(2) = 1.697224

PARAMETER VALUE DIFFERENCING INTERVAL
 1 0.0 0.10000D-09
 2 -3.00000 0.10000D-09

OPTIMAL VALUE FUNCTION SENSITIVITY

(24) DF/DAI 1) = 1.305098 DF/DAI 2) = 1.903824 DF/DAI

(25) DETAILED SENSITIVITY RESULTS FOLLOW FOR PARAMETERS
1 2

(26) X-DERIVATIVES ARE WITH RESPECT TO PARAMETER , 1
DX(1) = -0.2773500 DX(2) = 0.2773500 DX(

(27) U-DERIVATIVES WITH RESPECT TO PARAMETER 1
DU(1) = 0.5063357 DU(2) = -0.4622335D-01 DU(

(28) DFIX(R)/DA = 1.32820

(29) X-DERIVATIVES ARE WITH RESPECT TO PARAMETER , 2
DX(1) = -0.2773497 DX(2) = -0.7226484 DX(

U-DERIVATIVES WITH RESPECT TO PARAMETER 2
DU(1) = -0.4622325D-01 DU(2) = 1.308571 DU(

DFIX(R)/DA = 1.93375

Figure 3.--continued

MULTIPLER PROGRAMMING ROUTINE-SUMT VERSION 4 03/01/73

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
MAX. TIME= 0.900000E 03 R= 0.100000E 02 KATIO= 0.400000E 01 EPSILON= 0.999997E-08 PHETA= 0.999994E-06

OPTIONS SELECTED 1 1 1 1 1 1 0 1 1 1

TOLERANCES 0.0 0.999999E-03

SECOND SET OF OPTION, 1 1 1 0 1

TIME= 0.012 SECONDS
F= 0.9250000 01 P= 0.0
THE CURRENT VALUE OF X IS
0.1000000 01 0.1500000 01
THE CONSTRAINT VALUES
0.1500000 01 0.5000000 00
TIME= 0.090 SECONDS

*****THE FEASIBLE STARTING POINT TO BE USED IS ... G= 0.0 RSIGMA= 0.0 H= 0.0
THE CURRENT VALUE OF X IS
0.1000000 01 0.1500000 01
THE CONSTRAINT VALUES
0.1500000 01 0.5000000 00
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 0.250 SECONDS

POINT= 3 DOUT= 0.1649188D-02 RHO= 0.1000000 02 MAGNITUDE= 0.1642964D 00 PHASE= 2
F= 0.1512942D 02 P= 0.2670273D 01 G= -0.4670584D 01 RSIGMA= -0.1265914D 02 H= 0.0
THE CURRENT VALUE OF X IS
0.1650633D 00 0.1210901D 01
THE CONSTRAINT VALUES
0.2183655D 01 0.1624035D 01

LAGRANGE MULTIPLIERS
F= 0.1512942D 02 P= 0.2670273D 01 G= -0.4670584D 01 RSIGMA= -0.1265914D 02 H= 0.0
0.1650633D 00 0.1210901D 01
THE CONSTRAINT VALUES
0.4579477D 01 0.6157501D 01
APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
TIME= 0.449 SECONDS

POINT= 5 DOUT= 0.7491886D-03 RHO= 0.2500000 01 MAGNITUDE= 0.7059479D-01 PHASE= 2
F= 0.1007773D 02 P= 0.9660579D 01 G= 0.5077728D 01 RSIGMA= -0.4171488D 00 H= 0.0
THE CURRENT VALUE OF X IS
0.9053927D 00 0.1292092D 01
THE CONSTRAINT VALUES
0.1672350D 01 0.4025152D 00

LIST ORDER ESTIMATES
F= 0.8975690 01 P= 0.1199069 02 G= 0.4327165 01 RSIGMA= 0.0 H= 0.0
THE CURRENT VALUE OF X IS
0.1152160 01 0.1319156 01
THE CONSTRAINT VALUES
0.4280751 01 0.4280751 01

Figure 3.--continued

1ST ORDER ESTIMATES
 F = 0.62613650 01 P = 0.62615070 01 R SIGMA = 0.0 M = 0.0
 THE CURRENT VALUE OF X IS
 0.15615530 01
 THE CONSTRAINT VALUES
 -0.15033151-06 -0.352286570-06

LAGRANGE MULTIPLIERS
 F = 0.62613650 01 P = 0.62615070 01 R SIGMA = 0.0 M = 0.0
 THE CURRENT VALUE OF X IS
 0.15615530 01 0.14384470 01
 THE CONSTRAINT VALUES
 0.91698533 00 0.20602070 01
 APPARENTLY ROUND OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
 TIME = 2.391 SECONDS

 POINT = 23 DOTT = 0.90250670-07 RHO = 0.95367430-05 MAGNITUDE = 0.31949610 00 PHASE = 2
 F = 0.62613950 01 P = 0.62616120 01 R SIGMA = 0.22689850-03 M = 0.0
 THE CURRENT VALUE OF X IS
 0.15615490 01 0.14384460 01
 THE CONSTRAINT VALUES
 0.10230960-04 0.65429340-05

2ND ORDER ESTIMATES
 F = 0.62613660 01 P = 0.62613940 01 R SIGMA = 0.0 M = 0.0
 THE CURRENT VALUE OF X IS
 0.15615530 01 0.14384470 01
 THE CONSTRAINT VALUES
 -0.19056200-06 -0.98911230-07

1ST ORDER ESTIMATES
 F = 0.62613660 01 P = 0.62614020 01 R SIGMA = 0.0 M = 0.0
 THE CURRENT VALUE OF X IS
 0.15615530 01 0.14384470 01
 THE CONSTRAINT VALUES
 -0.22537890-06 -0.11478340-06

LAGRANGE MULTIPLIERS
 F = 0.62613660 01 P = 0.62614020 01 R SIGMA = 0.0 M = 0.0
 THE CURRENT VALUE OF X IS
 0.15615530 01 0.14384470 01
 THE CONSTRAINT VALUES
 0.93214500 00 0.20992480 01
 APPARENTLY ROUND OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.
 TIME = 2.500 SECONDS

 POINT = 25 DOTT = 0.29651740-07 RHO = 0.23941860-05 MAGNITUDE = 0.42775980 00 PHASE = 2
 F = 0.62613710 01 P = 0.62616340 01 R SIGMA = 0.63127680-04 M = 0.0
 THE CURRENT VALUE OF X IS
 0.15615570 01 0.14384470 01
 THE CONSTRAINT VALUES
 0.26639090-05 0.11895300-05

31

2ND ORDER ESTIMATES
 F = 0.62613670 01 P = 0.62613730 01 R SIGMA = 0.0 M = 0.0
 THE CURRENT VALUE OF X IS
 0.15615530 01 0.14384470 01

Figure 3.--continued

THE CURRENT VALUE OF X IS
 0.15615533 01

LAGRANGE MULTIPLIERS
 F= 0.62613670 01 P= 0.62613750 01 G= 0.62613670 01 H= 0.0
 THE CURRENT VALUE OF X IS
 0.15615533 01 Q= 0.14944470 01 R= 0.0
 THE CONSTRAINT VALUES
 0.16155533-06 0.71729250-07

LAGRANGE MULTIPLIERS
 F= 0.62613670 01 P= 0.62613750 01 G= 0.62613670 01 H= 0.0
 THE CURRENT VALUE OF X IS
 0.15615533 01 Q= 0.14944470 01 R= 0.0
 THE CONSTRAINT VALUES
 0.89499520 00 0.20043080 01

32

SENSITIVITY ANALYSIS

THE VALUE OF R(RHO) IS 0.238420-05
 THE POINT AT WHICH THE ESTIMATE OF SENSITIVITY WILL BE MADE IS
 X(1)= 1.561552 X(2)= 1.438447 X(

PARAMETER VALUE DIFFERENCING INTERVAL
 1 -1.00000 0.100000-09
 2 -3.00000 0.100000-09

OPTIMAL VALUE FUNCTION SENSITIVITY

33 DF/DA(1)= 0.8949943 DF/DA(2)= 2.004307 DF/DA(

Figure 3.--continued

OPTIMAL VALUE FUNCTION, BOUNDS AND, PAR(I) IS PERTINENT

POINT 1 (UNPERTURBED SOLUTION) :
 PAR(I) = 0.231663D-15 F(PAR(I)) = 0.7366693D 01
 POINT 2 (PERTURBED SOLUTION) :
 PAR(I) = -0.100000D 01 F(PAR(I)) = 0.6261367D 01

34

LINE PASSING THROUGH POINTS 1 AND 2 AND OVER ESTIMATING F*
 F = 0.1105326D 01 * PAR(I) + 0.7366693D 01

35

LINE UNDER ESTIMATING F* AT POINT 1

F = 0.1335099D 01 * PAR(I) + 0.7366693D 01

36

LINE UNDER ESTIMATING F* AT POINT 2

F = 0.6949943D 00 * PAR(I) + 0.7156361D 01

37

QUADRATIC ESTIMATION OF F* THROUGH POINTS 1 AND 2

F = 0.1997719D 00 * PAR(I)**2 + 0.1305098D 01 * PAR(I) + 0.7366693D 01

38

F* BOUND EVALUATION AT TEN EQUIDISTANCE POINTS BETWEEN POINTS 1 AND 2

PAR(I)	LOWER BOUND	UPPER BOUND	QUAD ESTIMATE
0.000	0.7366693D 01	0.7366693D 01	0.7366693D 01
-0.100	0.7238183D 01	0.7256160D 01	0.7238180D 01
-0.200	0.7105673D 01	0.7149627D 01	0.7113664D 01
-0.300	0.6975163D 01	0.7035095D 01	0.6993143D 01
-0.400	0.6846653D 01	0.6924562D 01	0.6876617D 01
-0.500	0.6716144D 01	0.6814030D 01	0.6764087D 01
-0.600	0.6619364D 01	0.6703497D 01	0.6655552D 01
-0.700	0.6529865D 01	0.6592964D 01	0.6551012D 01
-0.800	0.6440365D 01	0.6482432D 01	0.6450468D 01
-0.900	0.6350866D 01	0.6371899D 01	0.6353920D 01
-1.000	0.6261367D 01	0.6261367D 01	0.6261367D 01

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Figure 3.--continued



MULTI-CR PROGRAMMING ROUTINE-SUMT VERSION 4 03/01/73

N= 2 M= 2 NZ= 0
NAT. TIME= 0.900000E 03 R= 0.100000E 02 RATIO= 0.400000E 01 EPSILON= 0.9999997E-08 THETA= 0.9999994E-06

OPTIMIS SELECTED 1 1 1 1 1 1 0 1 1 1
TOLERANCES 0.0 0.9999999E-03

SECOND SET OF OPTIONS
TIME= 0.012 SECONDS
F= 0.9250300 01 P= 0.0
THE CURRENT VALUE OF X IS
0.1000000 01 0.1500000 01
THE CONSTRAINT VALUES
0.5000000 00 -0.1500000 01
TIME= 0.219 SECONDS

POINT= 3 DOTT= 0.1492879D-03 RHO= 0.1000000 02 MAGNITUDE= 0.1215422D-01 PHASE= 1
F= 0.8750897D 01 P= -0.1427585D 02 G= -0.1124910D 02 RSIGMA= -0.2302675D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.5003024D 00 0.1025120D 02
THE CONSTRAINT VALUES
0.1000090D 02 -0.8750897D 01

LAGRANGE MULTIPLIERS
F= 0.8750897D 01 P= -0.1427585D 02 G= -0.1124910D 02 RSIGMA= -0.2302675D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.5003024D 00 0.1025120D 02
THE CONSTRAINT VALUES
0.9999103D 00 -0.1142740D 01
TIME= 0.379 SECONDS

THE FEASIBLE STARTING POINT TO BE USED IS ...
F= 0.2050444D 02 P= 0.1279110D 02 G= 0.0 RSIGMA= 0.1279110D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.4999494D 00 0.1495133D 01
THE CONSTRAINT VALUES
0.1245184D 01 0.4916430D-02
TIME= 0.060 SECONDS

APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.

POINT= 7 DOTT= 0.8961926D-05 RHO= 0.1000000 02 MAGNITUDE= 0.1689025D-01 PHASE= 2
F= 0.2038678D 02 P= 0.3018354D 02 G= 0.4677850D-01 RSIGMA= 0.1009676D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.4016757D 00 0.7420775D 00
THE CONSTRAINT VALUES
0.6510695D 00 0.5595977D 00

LAGRANGE MULTIPLIERS
F= 0.2038678D 02 P= 0.3018354D 02 G= 0.4677850D-01 RSIGMA= 0.1009676D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.4016757D 00 0.7420775D 00
THE CONSTRAINT VALUES
0.1445037D 00 0.1744498D 02

APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.

LAGRANGE MULTIPLIERS
F= 0.2038678D 02 P= 0.3018354D 02 G= 0.4677850D-01 RSIGMA= 0.1009676D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.4016757D 00 0.7420775D 00
THE CONSTRAINT VALUES
0.1445037D 00 0.1744498D 02

APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.

LAGRANGE MULTIPLIERS
F= 0.2038678D 02 P= 0.3018354D 02 G= 0.4677850D-01 RSIGMA= 0.1009676D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.4016757D 00 0.7420775D 00
THE CONSTRAINT VALUES
0.1445037D 00 0.1744498D 02

APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.

LAGRANGE MULTIPLIERS
F= 0.2038678D 02 P= 0.3018354D 02 G= 0.4677850D-01 RSIGMA= 0.1009676D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.4016757D 00 0.7420775D 00
THE CONSTRAINT VALUES
0.1445037D 00 0.1744498D 02

APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.

LAGRANGE MULTIPLIERS
F= 0.2038678D 02 P= 0.3018354D 02 G= 0.4677850D-01 RSIGMA= 0.1009676D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.4016757D 00 0.7420775D 00
THE CONSTRAINT VALUES
0.1445037D 00 0.1744498D 02

APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.

LAGRANGE MULTIPLIERS
F= 0.2038678D 02 P= 0.3018354D 02 G= 0.4677850D-01 RSIGMA= 0.1009676D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.4016757D 00 0.7420775D 00
THE CONSTRAINT VALUES
0.1445037D 00 0.1744498D 02

APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.

LAGRANGE MULTIPLIERS
F= 0.2038678D 02 P= 0.3018354D 02 G= 0.4677850D-01 RSIGMA= 0.1009676D 02 H= 0.0
THE CURRENT VALUE OF X IS
-0.4016757D 00 0.7420775D 00
THE CONSTRAINT VALUES
0.1445037D 00 0.1744498D 02

APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.

Figure 3.--continued



OPTIMAL VALUE FUNCTION ROUNDS WHEN PAR(2) IS PERTURBED

POINT 1 (UNPERTURBED SOLUTION) : F(PAR(2))= 0.7366693D 01
 PAR(2)= -0.3000000D 01
 POINT 2 (PERTURBED SOLUTION) : F(PAR(2))= 0.1405573D 02
 PAR(2)= -0.1000000D 01

LINE PASSING THROUGH POINTS 1 AND 2 AND OVER ESTIMATING F*

$$F = 0.3366518D 01 * PAR(2) + 0.1740025D 02$$

LINE UNDER ESTIMATING F* AT POINT 1

$$F = 0.1903024D 01 * PAR(2) + 0.1307017D 02$$

LINE UNDER ESTIMATING F* AT POINT 2

$$F = 0.6796184D 01 * PAR(2) + 0.1005191D 02$$

QUADRATIC ESTIMATION OF F* THROUGH POINTS 1 AND 2

$$F = 0.720367D 00 * PAR(2)^2 + 0.6225905D 01 * PAR(2) + 0.1956129D 02$$

F* ROUND EVALUATION AT TEN EQUIDISTANCE POINTS BETWEEN POINTS 1 AND 2

PAR(2)	LOWER BOUND	UPPER BOUND	QUAD ESTIMATE
-2.000	0.7766593D 01	0.7466693D 01	0.7366693D 01
-1.800	0.7747457D 01	0.4035596D 01	0.7776271D 01
-1.600	0.8128222D 01	0.8707500D 01	0.8243478D 01
-1.400	0.8508987D 01	0.9373403D 01	0.8768312D 01
-1.200	0.8889752D 01	0.1004231D 02	0.9350774D 01
-1.000	0.9270517D 01	0.1071171D 02	0.9990864D 01
-0.800	0.1021878D 02	0.1136911D 02	0.1068854D 02
-0.600	0.1117802D 02	0.1204902D 02	0.1144393D 02
-0.400	0.1213725D 02	0.1271792D 02	0.1225690D 02
-0.200	0.1309649D 02	0.1338682D 02	0.1312750D 02
-0.000	0.1405573D 02	0.1405573D 02	0.1405573D 02

Figure 3.--continued

- I Output for solution of the unperturbed problem, i.e., $\epsilon_1 = 0$, $\epsilon_2 = -3$ (Step 1 of Algorithm 3.1).
- ① Printout of parameter card data.
 - ② Printout of first option card data.
 - ③ Printout of tolerance card data.
 - ④ Printout of second option card data.
 - ⑤ Elapsed time since the start of the problem. (This information is not accurate. Subroutine TIMEC, which monitors the elapsed time, has to be modified.)
 - ⑥ Initial starting point x^0 and corresponding problem function values.
 - ⑦ Feasibility of x^0 is verified. Note that if x^0 is not feasible, then SENSUMT invokes the subroutines FEAS and BODY to find a feasible starting point.
 - ⑧ The solution to the first subproblem. Program took four inverse product moves to minimize W for

$$RH0 = 10, (r=10)$$

$$D0TT = \nabla_x W[x(r)]^T \left[\frac{\partial^2 W[x(r)]}{\partial x_i \partial x_j} \right]^{-1} \nabla_x W[x(r)] < 10^{-7}$$

$$\text{MAGNITUDE} = \|\nabla_x W[x(r)]\| \quad (\text{the magnitude of the gradient of } W)$$

$$F = f[x(r)] \quad (\text{the value of the objective function})$$

$$W = f[x(r)] - r \sum_{j=1}^M \ln g_j[x(r)] + \sum_{j=M+1}^{M+MZ} [h_j[x(r)]]^2 / r$$

$$(\text{the value of the P-function})$$

Note: The W -function coded in SENSUMT differs from that given in (2.6) in the last term by a factor of 2.

$$\text{RESIGMA} = -r \sum_{j=1}^M \ln g_j[x(r)] \quad (\text{note that this may be negative})$$

$$H = \sum_{j=M+1}^{M+MZ} h_j^2[x(r)]/r$$

$$G = \text{dual value} = f[x(r)] - r^*M + 2.*H$$

(in a convex problem $F \geq v^* \geq G$)

where the current value of x is $x(r)$ and the current constraint values are $g_1[x(r)]$ and $g_2[x(r)]$.

- ⑨ Values of F , W (designated by P in the computer output), G , RSIGMA , and x are repeated. Current constraint values (i.e., u_j and w_j) are r/g_j , $j=1, M$, and $2n_j(x)/r$, $j=M+1, M+MZ$. Note: Since Example 1 has no equality constraint, the entries relating to h_j in 8 and 9 are zero.
- ⑩ The solution to the second subproblem ($r = 10/4 = 2.5$).
- ⑪ The current first order estimates of x (obtained by first order extrapolation, using the solution of the first and second subproblems) and the corresponding problem functions.
- ⑫ The solution to the third subproblem ($r = 2.5/4 = .625$).
- ⑬ The current second order estimates of x (obtained by second order extrapolation using solutions of the first, second, and third subproblems) and the corresponding problem functions.
- ⑭ The solution to the fourth subproblem.
- ⑮ The solution to the fifth subproblem.
- ⑯ The solution to the sixth subproblem.
- ⑰ The solution to the seventh subproblem.
- ⑱ The solution to the eighth subproblem.
- ⑲ The solution to the ninth subproblem.
- ⑳ The solution to the tenth subproblem.

- ⑳ The solution to the eleventh subproblem.
 - ㉑ The solution to the twelfth (final) subproblem.
 - ㉒ The general information about the sensitivity analysis including the value of r , the estimated final solution point and the parameter values and associated differencing intervals.
 - ㉓ Estimates of the sensitivity of the optimal value function obtained by taking the gradient of the Lagrangian with respect to the parameters as described in Step 9 of Algorithm 2.1.
 - ㉔ The parameters for which detailed sensitivity results follow. Here it is both parameters since the optimal value function is sensitive to both.
 - ㉕ Estimates of the partial derivatives of the solution point taken with respect to parameter one as in Steps 1-3 of Algorithm 2.1.
 - ㉖ Estimates of the partial derivatives of the Lagrange multipliers taken with respect to parameter one as in Step 5 of Algorithm 2.1.
 - ㉗ Estimate of the partial derivative of the optimal value function taken with respect to parameter one and obtained by using the chain rule as in Step 8 of Algorithm 2.1.
 - ㉘ Same as ㉕, ㉖, and ㉗ but with respect to parameter two.
- II Output for solution of the first perturbed problem, i.e., $\epsilon_1 = -1$, $\epsilon_2 = -3$ (Step 2 of Algorithm 3.1).
- ㉙ Printout of the input data and subproblem solutions for the first perturbed problem.
 - ㉚ Printout of the final subproblem for the first perturbed problem.
 - ㉛ Same as ㉒ but for the first perturbed problem.
 - ㉜ Same as ㉓ but for the first perturbed problem.

III

Output for optimal value function bounds as a function of first parameter (Steps 3 and 4 of Algorithm 3.1).

- ③④ $(f^*(\epsilon_1), \epsilon_1)$ of the unperturbed and first perturbed problems [see ②②, ②③, and ③①, ③②, respectively].
- ③⑤ Equations of parametric upper bound on $f^*(\epsilon_1)$.
- ③⑥ Equations of parametric lower bounds on $f^*(\epsilon_1)$.
- ③⑦ Maximum of these equations over the range of $-1 \leq \epsilon_1 \leq 0$ provides a lower bound on $f^*(\epsilon_1)$.
- ③⑧ Equation of a quadratic estimate of $f^*(\epsilon_1)$ [see the description of subroutine BOUND in Section 6 for the method of deriving this equation].
- ③⑨ Value of bounds and the quadratic estimate of $f^*(\epsilon_1)$ at eleven equidistant points over the perturbation range of the first parameter ϵ_1 .

IV

Output for solution of the second perturbed problem, i.e., $\epsilon_1 = 0$, $\epsilon_2 = -1$ (Step 2 of Algorithm 3.1).

Description of the output for the second perturbed problem is similar to that of the first perturbed problem, i.e., ③⑩ - ③⑬.

V

Output for optimal value function bounds as a function of the second parameter (Steps 3 and 4 of Algorithm 3.1).

The description here again closely parallels that of the first parameter, i.e., ③⑭ - ③⑱.

The graphical depiction of the bounds derived for $f^*(\epsilon)$ as a function of ϵ_2 , together with a plot of the analytical solution and quadratic estimates of $f^*(\epsilon)$ at different values of ϵ_2 is in Figure

4.

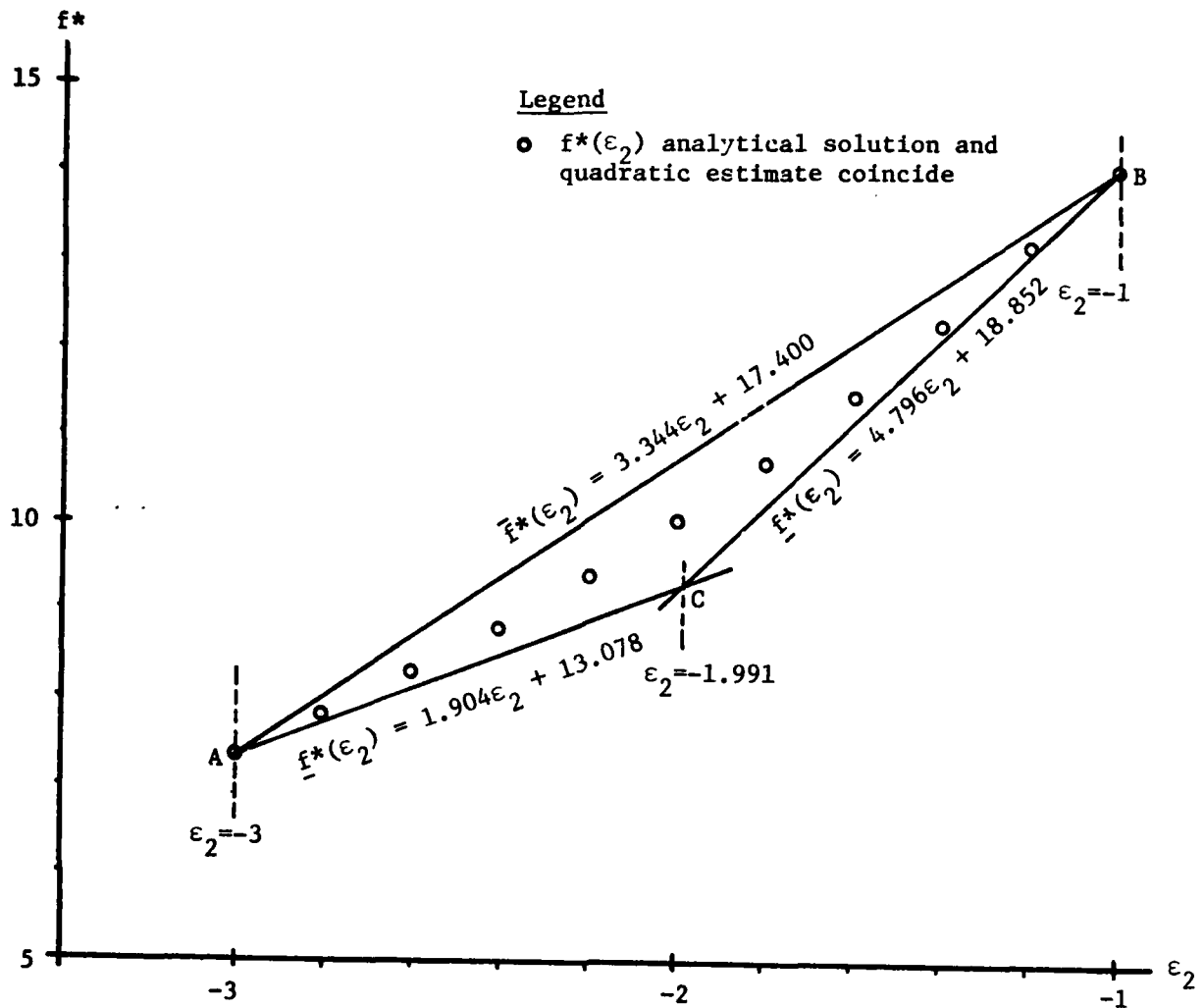


Figure 4.--Graph of bounds on $f^*(\epsilon_2)$ of Example 1 (computer solution).

5.2 Example 2

$$\text{Minimize } f(x) = x_1^2 + .5e^{x_2} - .125x_3^3 + .25(x_4 - 40)^2 - .1e^{x_5}$$

x_1, \dots, x_5

$$\text{Subject to } g_1(x) = -.5x_1^2 + 6x_2 - 5e^{x_3} - .05x_4^2 - .5/x_5 \geq \epsilon_1$$

$$g_2(x) = -5e^{-x_1} - 2x_2^2 + 3x_3 + x_4 + 3x_5 \geq -12$$

$$g_3(x) = 3x_1 + x_2 - x_3^2 + x_4 - x_5^2 \geq 2$$

$$0 \leq x_3 \leq 5, \quad 0 \leq x_5 \leq 5, \quad 0 \leq x_j \leq \infty, \quad j=1,2,4.$$

It is desired to calculate bounds on $f^*(\epsilon_1)$ of this problem for $\epsilon_1 \in [-10, 5]$.

(i) Problem $SR(\epsilon)$.

• Standard format:

$$\begin{aligned} \text{Min}_{x_1, \dots, x_5} f(x) &= \sum_{j=1}^5 f_j(x_j) = x_1^2 + .5e^{x_2} - .125x_3^3 \\ &\quad + .25(x_4 - 40)^2 - .1e^{x_5} \end{aligned}$$

S.t. $x \in G \cap B$, where

$$G = \{x : g_1(x) \geq \epsilon_1, g_2(x) \geq -12, g_3(x) \geq 2\}, \quad -10 \leq \epsilon_1 \leq 5$$

and

$$B = \{0 \leq x_3 \leq 5, 0 \leq x_5 \leq 5, 0 \leq x_i \leq \infty, i=1, 2, 4\}.$$

(ii) Problem $\tilde{C}SR(\epsilon)$, (convex overestimating problem of $SR(\epsilon)$).

• Problem formulation:

$$\tilde{f}_3(x_3) = -3.125x_3 + 6.014$$

$$\tilde{f}_5(x_5) = -2.948x_5 + 7.027$$

Remaining terms are identical to those of problem $SR(\epsilon)$; thus problem $\tilde{C}SR(\epsilon)$ reads

$$\begin{aligned} \text{Min}_{x_1, \dots, x_5} \tilde{f}(x) &= x_1^2 + .5e^{x_2} - 3.125x_3 + .25(x_4 - 40)^2 \\ &\quad - 2.948x_5 + 13.041 \end{aligned}$$

S.t. $x \in \tilde{G} \cap B$, $-10 \leq \epsilon_1 \leq 5$,

where $\tilde{G} \supseteq G$.

Note: $\tilde{f}_3(x_3)$ and $\tilde{f}_5(x_5)$ were chosen here to be the lowest nonunderestimating lines parallel to the convex envelopes of $f_3(x_3)$ and $f_5(x_5)$, respectively.

(iii) Problem $\tilde{C}\tilde{S}R(\epsilon)$, (convex underestimating problem).

• Problem formulation:

$$\tilde{f}_3(x_3) = -3.125x_3$$

$$\tilde{f}_5(x_5) = -2.948x_5$$

Remaining terms are identical to those of problem $SR(\epsilon)$; thus problem $\tilde{C}\tilde{S}R(\epsilon)$ reads

$$\begin{aligned} \text{Min}_{x_1, \dots, x_5} \quad f(x) = & x_1^2 + .5e^{x_2} - 3.12x_3 + .25(x_4 - 40)^2 \\ & - 2.948x_5 - .1 \end{aligned}$$

$$\text{S.t.} \quad x \in \tilde{G} \cap B, \quad -10 \leq \epsilon_1 \leq 5$$

where $\tilde{G} \equiv G$.

The computer listing of the code for problem $\tilde{C}\tilde{S}R(\epsilon)$ is shown in Figure 5. The listing of the code for $\tilde{C}\tilde{S}R(\epsilon)$ is identical to that of the problem $\tilde{C}\tilde{S}R(\epsilon)$ except for the constant term in the objective function, which is + 13.041, rather than - .1.

(iv) Bounds.

Figures 6 and 7 depict the calculated upper and lower bounds via the analysis of the problems $\tilde{C}\tilde{S}R(\epsilon)$ and $\tilde{C}\tilde{S}R(\epsilon)$, respectively. These results are summarized below.

Upper bound:

$$\bar{f}^*(\epsilon_1) = 4.825\epsilon_1 + 136.628 .$$

Lower bound:

$$\underline{f}^*(\epsilon_1) = \max[1.880\epsilon_1 + 94.032, 5.466\epsilon_1 + 120.284] .$$

The graphical depiction of these bounds is in Figure 8.

```

J=GRAEMI T=2 L=2
// EXEC FORG1
SUBROUTINE READIN
  IMPLICIT REAL*8(A-H,O-Z)
  COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
  COMMON/ABG/LY,LZ,PER(45)
  NPAR=3
  PAR(1)=5.
  PAR(2)=-12.
  PAR(3)=2.
  PER(1)=-15.
  RETURN
END
SUBROUTINE RESTNT(IN,VAL)
  IMPLICIT REAL*8(A-H,O-Z)
  COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1
  COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
  COMMON/ABG/LY,LZ,PER(45)
  I=IN+1
  GO TO (20,1,2,3,4,5),I
20 VAL=X(1)**2+.5*DEXP(X(2))-3.125*X(3)+.25*(X(4)-40.)**2
  * -2.948*X(5)+13.041
  RETURN
1 VAL=-.5*X(1)**2+6.*X(2)-5.*DEXP(X(3))-0.5*X(4)**2-.5/X(5)
  * -PAR(1)
  RETURN
2 VAL=-5.*DEXP(-X(1))-2.*X(2)**2+3.*X(3)+X(4)+3.*X(5)-PAR(2)
  RETURN
3 VAL=3.*X(1)+X(2)-X(3)**2+X(4)-X(5)**2-PAR(3)
  RETURN
4 VAL=-X(3)+5.
  RETURN
5 VAL=-X(5)+5.
  RETURN
END
SUBROUTINE GRAD1(IN)
  IMPLICIT REAL*8(A-H,O-Z)
  COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1
  COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
  J=IN+1
  DO 15 I=1,N
  DEL(I)=0.0
15 CONTINUE
  GO TO (20,1,2,3,4,5),J
20 DEL(1)=2.*X(1)
  DEL(2)=.5*DEXP(X(2))
  DEL(3)=-3.125
  DEL(4)=.5*(X(4)-40.)
  DEL(5)=-2.948
  RETURN
1 DEL(1)=-1.*X(1)
  DEL(2)=6.
  DEL(3)=-5.*DEXP(X(3))
  DEL(4)=-.1*X(4)
  DEL(5)=.5/X(5)**2
  RETURN
2 DEL(1)=5.*DEXP(-X(1))
  DEL(2)=-4.*X(2)
  DEL(3)=3.
  DEL(4)=1.
  DEL(5)=3.

```

Figure 5.--Computer listing of the code for the convex overestimating problem of Example 2.

```

RETURN
3 DEL(1)=3.
  DEL(2)=1.
  DEL(3)=-2.*X(3)
  DEL(4)=1.
  DEL(5)=-2.*X(5)
  RETURN
4 DEL(3)=-1.
  RETURN
5 DEL(5)=-1.
  RETURN
  END
  SUBROUTINE MATRIX(IN,K)
  IMPLICIT REAL*8(A-H,O-Z)
  COMMON/SHARE/X(45),DEL(45), A(45,45),N,M,MN,NPI,NM1
  COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
  L=IN+1
  GO TO (20,1,2,3,4,4),L
20 A(1,1)=2.
  A(2,2)=.5*DEXP(X(2))
  A(4,4)=.5
  RETURN
1 A(1,1)=-1.
  A(3,3)=-5.*DEXP(X(3))
  A(4,4)=-.1
  A(5,5)=-1./X(5)**3
  RETURN
2 A(1,1)=-5.*DEXP(-X(1))
  A(2,2)=-4.
  RETURN
3 A(3,3)=-2.
  A(5,5)=-2.
  RETURN
4 K=1
  RETURN
  END
// EXEC FORG6,DSN='OR799661.BOUNDS45',PROG=MAIN
//GO.SYSLIB DD
// DD
// DD DSN=GNU.FORTLIB,DISP=SHR
//GO.SYSLIN DD
// DD DSN=STEMPUNCH,DISP=(OLD,DELETE)
//F107F001 DD SYSOUT=A
1.0E-08 10.0 1.0E-06 4.0 900.0 5 5 0
3 3. 3. 4. 1. 4. 1 1
0.1E-08 0.0001 1 1 0 2
//

```

Figure 5.--continued

OPTIMAL VALUE FUNCTION BOUNDS WHEN PARAM 1 IS PERTURBED

POINT 1 (UNPERTURBED SOLUTION) :
 PARAM 1) = 0.5000000 01 F(PARAM 1) = 0.16075470 03
 POINT 2 (PERTURBED SOLUTION) :
 PARAM 1) = -0.1000000 02 F(PARAM 1) = 0.98373770 02

LINE PASSING THROUGH POINTS 1 AND 2 AND OVER ESTIMATING F*

$$F = 0.8253980 01 * PARAM 1) + 0.13662770 03$$

F* (Q) EVALUATION AT TEN EQUIDISTANCE POINTS BETWEEN POINTS 1 AND 2

PARAM 1)	LOWER BOUND	UPPER BOUND	QUAD ESTIMATE
5.000		0.16075470 03	
3.500		0.15351660 03	
2.000		0.14627850 03	
0.500		0.13904040 03	
-1.000		0.13180230 03	
-2.500		0.12456430 03	
-4.000		0.11732620 03	
-5.500		0.11008810 03	
-7.000		0.10285000 03	
-8.500		0.95611190 02	
-10.000		0.98373770 02	

Figure 6.--Parametric upper bound on $f^*(\epsilon_1)$ via problem $\tilde{C}\tilde{S}R(\epsilon)$, Example 2 (computer solution).

OPTIMAL VALUE FUNCTION BOUNDS WHEN PARAM 11 IS PERTURBED

POINT 1 (UNPERTURBED SOLUTION) :
 PARAM 11 = 0.5000000 01 F(PARAM 11) = 0.14761370 03
 POINT 2 (PERTURBED SOLUTION) :
 PARAM 11 = -0.1000000 02 F(PARAM 11) = 0.75232770 02

LINE UNDER ESTIMATING F* AT POINT 1

$$F = 0.54659210 01 * \text{PARAM 11} + 0.12028410 03$$

LINE UNDER ESTIMATING F* AT POINT 2

$$F = 0.18799810 01 * \text{PARAM 11} + 0.94032580 02$$

F* VALUE EVALUATION AT TEN EQUIDISTANCE POINTS BETWEEN POINTS 1 AND 2
 =====

PARAM 11	LOWER BOUND	UPPER BOUND	QUAD ESTIMATE
5.000	0.14761370 03		
4.500	0.13941490 03		
4.000	0.13121600 03		
3.500	0.12301710 03		
-1.000	0.11481820 03		
-2.500	0.10661930 03		
-4.000	0.99420460 02		
-5.500	0.90221580 02		
-7.000	0.82022700 02		
-8.500	0.78052740 02		
-10.000	0.75232770 02		

Figure 7.--Parametric lower bound on $f^*(\epsilon_1)$ via problem $\text{CSR}(\epsilon)$,
 Example 2 (computer solution).

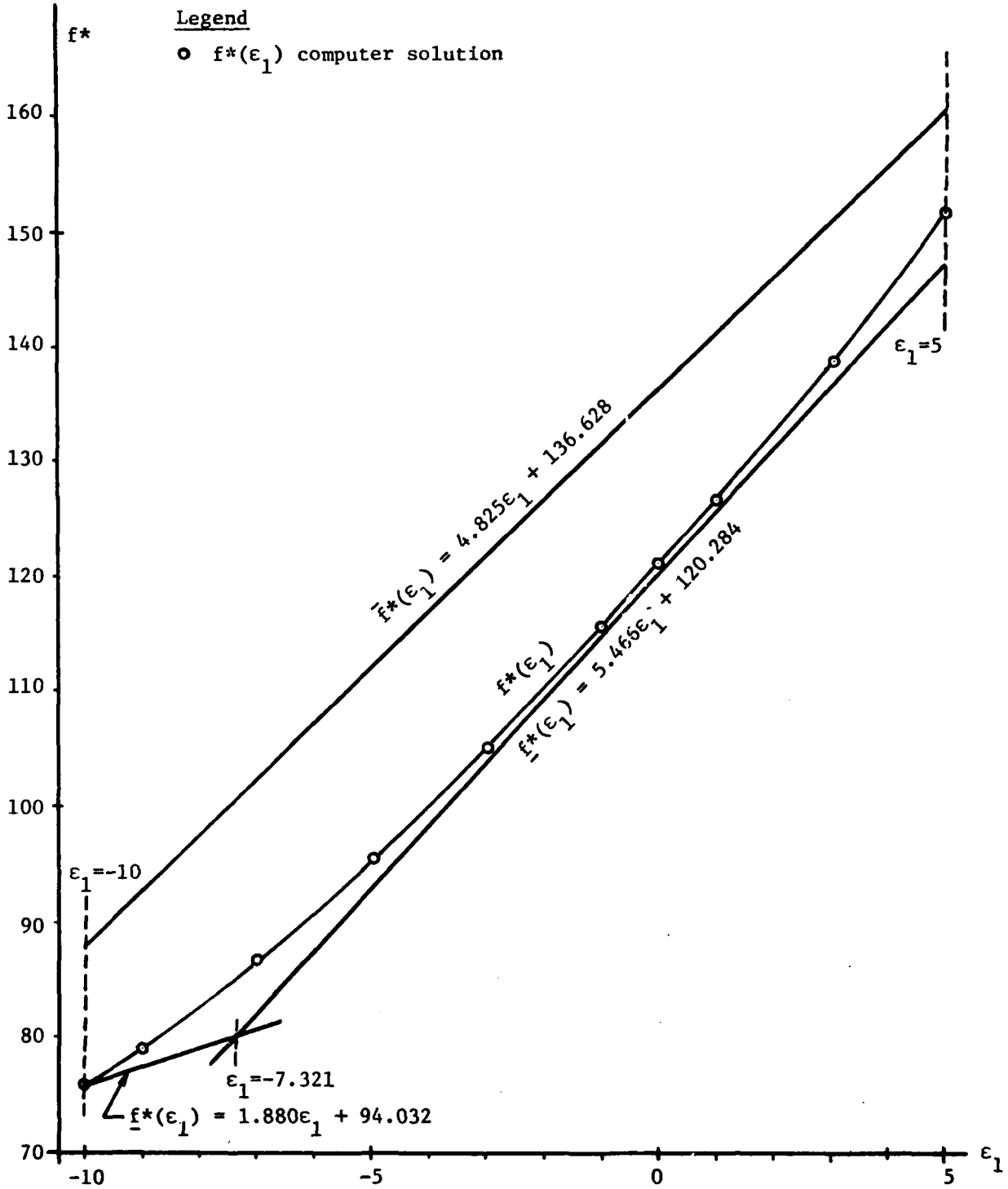


Figure 8.--Parametric bounds on $f^*(\epsilon)$, Example 2 (computer solution).

6. General Description and Listing of the SENSUMT Subroutines

The subroutines comprising SENSUMT fall into four categories:

- (i) user subroutines,
- (ii) SUMT subroutines,
- (iii) sensitivity subroutines, and
- (iv) bound subroutines.

As mentioned before, user subroutines are generally composed of four subroutines, i.e., READIN, RSTNT, GRAD1, and MATRIX. These subroutines respectively provide pertinent information about the problem, functions of the problem, their gradients and Hessian matrices. A detailed description of these subroutines, together with the corresponding codes for Examples 1 and 2, were given in Section 4 and Figures 2 and 5.

SUMT subroutines implement the Sequential Unconstrained Minimization Technique of Fiacco and McCormick [10]. The subroutines, along with program MAIN, comprising this group are:

• BØDY	• ØUTPUT
• CHCKER	• PEVALU
• CØNVRG	• PUNCH
• DIFF1	• REJECT
• DIFF2	• RHØCØM
• ESTIM	• SECØND
• EVALU	• SECØRD
• FEAS	• SET
• FINAL	• STØRE
• GRAD	• TCHECK
• INVERS	• TIMEC
• MAIN	• TRANS
• ØPT	• XMØVE

With the exception of the subroutine TRANS, the above routines have been developed by Mylander, Holmes, and McCormick [13]. Some of

these routines, in particular program MAIN and subroutines BODY and OUTPUT, have been modified for implementing sensitivity and bound calculation routines. Subroutine TRANS was recently developed to aid the user when analyzing a convex equivalent of geometric programming problems by SENSUMT. See page 104 for a more detailed description of this subroutine.

Sensitivity subroutines implement the sensitivity analysis Algorithm 2.1 and interface it with the SUMT subroutines. A brief history of the development of these subroutines was given in Section 1. The latest version of the codes comprising this group, due to Armacost [1], are subroutines SENS, PARDIF, LMULT, and PRESEN. Subroutines SENS and PRESEN have been slightly modified in implementing the bound calculation routines.

Bound subroutines, developed in [11], implement the bound calculation Algorithms 3.1 and 3.2 given in Section 3. The two subroutines in this group are BOUND and PERT.

All of the subroutines comprising SENSUMT are dimensioned to solve problems having at most 45 variables, 45 parameters, and 200 constraints. However, if the computer capacity permits, they may readily be re-dimensioned to solve problems of larger size. All of these subroutines are separately filed in the Conversational Monitor System (CMS) component of the IBM Virtual Machine facility 37C (VM/370) at The George Washington University Center for Academic and Administrative Computing under their corresponding names.

To make this manual self-contained, the computer listing and a general description of each subroutine is provided (in alphabetical order by name). The descriptions of the SUMT subroutines are largely taken from [13], and those of the sensitivity subroutines are taken from [1]. The user subroutines are problem-dependent, thus they are included in the following listings. For familiarity with the user subroutines, the reader should refer to the coding of Examples 1 and 2 provided in Figures 2 and 5, respectively.

6.1 BODY

Subroutine BODY coordinates the flow among the subroutines that actually do the calculations required by the various phases of the algorithm. The flow in this routine is slightly different when the program is in the feasibility phase (solving the entry problem) rather than the normal phase (solving the stated NLP problem). The listing appears as Figure 9.

6.2 BOUND

The subroutine BOUND, called by the program MAIN, generates the equations for the upper and lower bounds of the optimal value function as functions of the problem parameters. When analyzing a problem which is known to have a convex or concave optimal value function, BOUND also derives the equation for a quadratic function which approximates the optimal value function. If $f^*(\epsilon)$ is convex (concave), it corresponds to the higher (lower) of the two quadratic functions, in the perturbation range of the parameter of interest, which passes through the two calculated points in $f^*(\epsilon), \epsilon$ space and has the calculated slope at each of these points. Evaluation and printout of this quadratic function (when applicable), and the upper and/or lower bound at eleven equidistant points over the perturbation range of the parameters of interest are also programmed in subroutine BOUND. The listing appears as Figure 10.

6.3 CHCKER

Subroutine CHCKER evaluates the first partial derivatives for all the functions at the starting point first by calling the user-supplied subroutine GRAD1 and then by calling DIFF1. The results are printed out to aid the user in debugging the subroutines used to describe the NLP problem he wishes to solve.

The matrix of second partial derivatives of each function is also evaluated by the two methods; first by calling MATRIX and then calling

FORTAN IV G LEVEL 21	BODY	DATE = 80240	10/05/54
0001	SUBROUTINE BODY		80000000
	C		80000000
	C AUGUST 1971		80000000
	C		80000000
	C BODY COORDINATES THE FLOW AMONG THE SUBROUTINES THAT ACTUALLY DO THE		80000100
	C CALCULATIONS REQUIRED BY THE VARIOUS PARTS OF THE ALGORITHM.		80000110
0002	IMPLICIT REAL*8(A-H,O-Z)		80000120
0003	REAL*4 RHOIN,RATIO,EPSI,THETA0,XEPI,XFP2		80000130
0004	COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1		80000140
0005	COMMON /CPTS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10		80000150
0006	COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO		80000160
0007	COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0,		80000170
	IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PRI,		80000180
	ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),		80000190
	3 PRFV3,DELX, NTCTR, NUMINI, NPHASE, NSATIS		80000200
0008	COMMON /CONPAR/ NF1,NF2,NF3		80000210
0009	COMMON/EXPDP1/NEXP1,NEXP2,NEXP3,NEXP4,NEXP5,XEPI,XEPI2		80000220
0010	COMMON/SEN1/PAR(45),OPAR(45),NPAR,ISENS		80000230
0011	COMMON/ABG/LY,LZ,PER(45)		80000240
0012	COMMON/ABG1/FE1,FE2		80000250
0013	COMMON/ABG2/DF1(45),DF2(45)		80000260
0014	NF2=2		80000270
0015	NF3=2		80000280
0016	MN=0		80000290
0017	NUMINI=0		80000300
0018	ISENS = 0		80000310
	C OPTION OF GETTING INITIAL RHO		80000320
0019	CALL RHOCON		80000330
0020	CALL EVALU		80000340
0021	10 CALL XMOVE		80000350
0022	GO TO (30,20), NT3		80000360
0023	20 CALL TIMEC		80000370
0024	CALL OUTPUT (1)		80000380
0025	GO TO 40		80000390
0026	30 CALL TCHECK		80000400
	C IN FEASIBILITY PHASE 4 MEANS FEAS ACHIEVED		80000410
0027	40 GO TO (50,50,50,200), NSATIS		80000420
0028	50 CALL CONVRG (N1)		80000430
0029	GO TO (60,10,125), N1		80000440
	C MINIMUM ACHIEVED IF N1=1		80000450
0030	60 GO TO (70,80), NT3		80000460
0031	70 CALL TIMEC		80000470
0032	CALL OUTPUT (1)		80000480
	C --- NUMBER OF MINIMA ACHIEVED INCREASED BY 1		80000490
0033	80 NUMINI=NUMINI+1		80000500
0034	MN=0		80000510
0035	GO TO (190,90,90), NPHASE		80000520
0036	90 CALL ESTIM		80000530
	C FINAL MIGHT HAVE BEEN CALLED BY ESTIM-CONVERGED IF N2=1		80000540
0037	IF(NEXP3.NE.2) GO TO 27		80000550
0038	CALL SENS		80000560
0039	27 GO TO (100,110,120), NT4		80000570
	C NT4=1 FINAL CONVERGENCE ON 0 ORDER ESTIMATES, NT4=2 CONVERGE ON FIRST		80000580
	C ORDER ESTIMATES, NT4=3 CONVERGE ON SECOND ORDER ESTIMATES.		80000590
0040	100 CALL FINAL (NF1)		80000600
0041	GO TO (130,140), NF1		80000610
0042	110 GO TO (130,140), NF2		80000620
0043	120 GO TO (130,140), NF3		80000630
0044	125 NPHASE=5		80000640
0045	130 RETURN		80000650
0046	140 RHO=RHO/RATIO		80000660
	C USING PREVIOUSLY COMPUTED VALUES FOR F, AND RHO IS RECOMPUTED WITH THESE		80000670
	C NEW VALUE OF RHO.		80000680
0047	CALL PEVALU		80000690
	C A VECTOR IS LEFT IN DELX(1) BY ESTIM		80000700
0048	IF (NUMINI-2) 10,150,150		80000710
0049	150 GO TO (110,160,160), NT7		80000720
0050	160 CALL GRAD (2)		80000730
0051	CALL OPT		80000740
0052	GO TO (180,170), NT3		80000750
0053	170 WRITE (6,210)		80000760
0054	CALL OUTPUT (1)		80000770
0055	180 GO TO 50		80000780
0056	190 IF (G) 90,90,200		80000790
0057	200 RETURN		80000800
	C --- DUAL VALUE GREATER THAN 0 MEANS NO FEASIBLE POINT EXISTS		80000810
	C		80000820
0058	210 FORNAT (6X,30MOVED ON EXTRAPOLATION VECTOR)		80000830
0059	END		80000840

Figure 9.--Subroutine BODY.

```

FORTRAN IV G LEVEL 21          BOUND          DATE = 80242          12/12/03

C SUBROUTINE BOUND WAS DEVELOPED BY GHAEMII(1979) TO CALCULATE          80U00060
C UPPER AND/OR LOWER PIECEWISE LINEAR PARAMETRIC BOUNDS ON          80U00070
C OPTIMAL VALUE FUNCTIONS WHEN THESE FUNCTIONS ARE KNOWN TO BE          80U00080
C CONVEX OR CONCAVE OVER THE RANGE OF A GIVEN PARAMETER.          80U00090
0001 SUBROUTINE BOUND          80U00100
0002 IMPLICIT REAL*8(A-M,O-Z)          80U00110
0003 COMMON/SEN/PAR(45),DPA(45),NPAR,ISENS          80U00120
0004 COMMON/ABG/LY,LZ,PER(45)          80U00130
0005 COMMON/ABG1/FE1,FE2          80U00140
0006 COMMON/ABG2/DF1(45),DF2(45)          80U00150
0007 COMMON/OP7/NEXOP7          80U00160
0008 DIMENSION E(20),FQ(20),FQI(20),FU12(20),FL12(20)          80U00170
0009 E1=PAR(LZ)          80U00180
0010 E2=PAR(LZ) + PER(LZ)          80U00190
C LINE FIT THROUGH POINTS 1 AND 2          80U00200
0011 S12 = (FE1-FE2)/(E1-E2)          80U00210
0012 B12 = FE1-S12*E1          80U00220
C TANGENT LINE UNDER ESTIMATING F* AT POINT 1          80U00230
0013 S1=DF1(LZ)          80U00240
0014 B1=FE1-S1*E1          80U00250
C TANGENT LINE UNDER ESTIMATING F* AT POINT 2          80U00260
0015 S2=DF2(LZ)          80U00270
0016 B2=FE2-S2*E2          80U00280
C INTERSECTION OF THE ABOVE TWO LINES          80U00290
0017 E3=(B2-B1)/(S1-S2)          80U00300
0018 FE3=S1*(B2-B1)/(S1-S2) + B1          80U00310
C QUADRATIC FIT THROUGH POINTS 1 AND 2          80U00320
0019 A=(FE2-E2*S1-FE1+E1*S1)/((E1-E2)**2)          80U00330
0020 C=((E1**2)*(FE2-E2*S1)+(FE1-E1*S1)*(E2**2-2.*E1*E2))/((E1-E2)**2)          80U00340
0021 B=S1-2.*E1*A          80U00350
0022 AA=(FE1-E1*S2-FE2+E2*S2)/((E2-E1)**2)          80U00360
0023 CC=((E2**2)*(FE1-E1*S2)+(FE2-E2*S2)*(E1**2-2.*E2*E1))/((E2-E1)**2)          80U00370
0024 BB=S2-2.*E2*AA          80U00380
C BOUND EVALUATION AT 10 DISCRETE POINTS          80U00390
0025 E(I)=E1          80U00400
0026 DE=PER(LZ)/10.          80U00410
0027 DO 580 I=1,10          80U00420
0028 IF(I.EQ.1) GO TO 750          80U00430
0029 E(I)=E1+DE          80U00440
C QUADRATIC ESTIMATE          80U00450
0030 750 FQ(I)=A*E(I)**2+B*E(I)+C          80U00460
0031 FQI(I)=AA*E(I)**2+BB*E(I)+CC          80U00470
C UPPER LINEAR BOUND          80U00480
0032 FU12(I)=S12*E(I)+B12          80U00490
C LOWER LINEAR BOUND          80U00500
0033 IF(PER(LZ).GT.0.) GO TO 585          80U00510
0034 IF(E(I).LT.E3) GO TO 570          80U00520
0035 FL12(I)=S1*E(I)+B1          80U00530
0036 GO TO 595          80U00540
0037 570 FL12(I)=S2*E(I)+B2          80U00550
0038 GO TO 595          80U00560
0039 585 IF(E(I).GT.E3) GO TO 590          80U00570
0040 FL12(I)=S1*E(I)+B1          80U00580
0041 GO TO 595          80U00590
0042 590 FL12(I)=S2*E(I)+B2          80U00600
0043 595 CONTINUE          80U00610
0044 580 CONTINUE          80U00620
C CHOICE ON PROPER QUADRATIC FIT          80U00630
0045 IF(NEXOP7 .GE.4) GO TO 810          80U00640
0046 IF(FQ(5) .GT. FQI(5)) GO TO 800          80U00650
0047 A=AA          80U00660
0048 B=BB          80U00670
0049 C=CC          80U00680
0050 DO 100 I=1,10          80U00690
0051 FQ(I)=FQI(I)          80U00700
0052 100 CONTINUE          80U00710
0053 GO TO 800          80U00720
0054 810 IF(FQ(5) .LT. FQI(5)) GO TO 800          80U00730
0055 A=AA          80U00740
0056 B=BB          80U00750
0057 C=CC          80U00760
0058 DO 110 I=1,10          80U00770
0059 FQ(I)=FQ(I)          80U00780
0060 110 CONTINUE          80U00790
0061 800 CONTINUE          80U00800
0062 WRITE(6,600) LZ          80U00810
0063 600 FORMAT(2H1,25X,'OPTIMAL VALUE FUNCTION BOUNDS WHEN PAR',I2,          80U00820
      $) IS PERTURBED',/,24X,571(' '),//)          80U00830
0064 WRITE(6,610) LZ,E1,LZ,FE1,LZ,E2,LZ,FE2          80U00840
0065 610 FORMAT(2X,'POINT 1 (UNPERTURBED SOLUTION) :',/,2X,'PAR',I2,          80U00850
      $)=' ,E15.7,10X,'F(PAR',I2,'))=' ,E15.7,/,2X,'POINT 2 (PERTURBED'          80U00860
      $,' SOLUTION) :',/,2X,'PAR',I2,'))=' ,E15.7,10X,'F(PAR',I2,          80U00870
      $)=' ,E15.7,///)          80U00880
0066 GO TO (910,920,930,940,950,960),NEXOP7          80U00890
C OUTPUT WHEN NEXOP7=1          80U00900
0067 910 WRITE(6,620) S12,LZ,B12          80U00910
0068 620 FORMAT(7X,'LINE PASSING THROUGH POINTS 1 AND 2 AND OVER'          80U00920

```

Figure 10.--Subroutine BOUND.

```

      S, ESTIMATING F* ',/,1X,59('-''),//,11X,'F = ',E15.7,' * PAR('
      $12,') + ',E15.7,///)
0069      WRITE(6,630) S1,LZ,B1
0070      630 FORMAT(2X,'LINE UNDER ESTIMATING F* AT POINT 1 ',/1X,36('-''),
      $//,11X,'F = ',E15.7,' * PAR('',12,') + ',E15.7,///)
0071      WRITE(6,640) S2,LZ,B2
0072      640 FORMAT(2X,'LINE UNDER ESTIMATING F* AT POINT 2 ',/1X,36('-''),
      $//,11X,'F = ',E15.7,' * PAR('',12,') + ',E15.7,///)
0073      WRITE(6,650) A,LZ,B,LZ,C
0074      650 FORMAT(2X,'QUADRATIC ESTIMATION OF F* THROUGH POINTS 1 AND'
      $,' 2 ',/1X,50('-''), //,11X,'F = ',E15.7,' * PAR('',12,
      $) ** 2 + ',E15.7,' * PAR('',12,') + ', E15.7,///)
0075      WRITE(6,660) LZ
0076      660 FORMAT(2X,'F* BOUND EVALUATION AT TEN EQUIDISTANCE POINTS',
      $' BETWEEN POINTS 1 AND 2 ',/1X,71('-''),//,10X,'PAR('',12,')',
      $10X,'LOWER BOUND',10X,'UPPER BOUND',10X,'QUAD ESTIMATE',/
      $10X,7('-''),9X,13('-''),8X,13('-''),8X,15('-''),/)
0077      WRITE(6,670) (E(I),FL12(I),FU12(I),FQ(I),I=1,10)
0078      WRITE(6,670) E2,FE2,FE2,FE2
0079      670 FORMAT(10X,F7.3,7X,E15.7,6X,E15.7,6X,E15.7)
0080      RETURN
      C OUTPUT WHEN NEXOP7=2
0081      920 WRITE(6,620) S12,LZ,B12
0082      WRITE(6,660) LZ
0083      WRITE(6,921) (E(I),FU12(I), I=1,10)
0084      WRITE(6,921) E2,FE2
0085      921 FORMAT(10X,F7.3,28X,E15.7)
0086      RETURN
      C OUTPUT WHEN NEXOP7=3
0087      930 WRITE(6,630) S1, LZ,B1
0088      WRITE(6,640) S2,LZ,B2
0089      WRITE(6,660) LZ
0090      WRITE(6,931) (E(I),FL12(I),I=1,10)
0091      WRITE(6,931) E2,FE2
0092      931 FORMAT(10X,F7.3,7X,E15.7)
0093      RETURN
      C OUTPUT WHEN NEXOP7=4
0094      940 WRITE(6,941) S12,LZ,B12
0095      941 FORMAT(2X,'LINE PASSING THROUGH POINTS 1 AND 2 AND UNDER'
      $,' ESTIMATING F* ',/,1X,59('-''),//,11X,'F = ',E15.7,' * PAR('
      $12,') + ',E15.7,///)
      WRITE(6,942) S1,LZ,B1
0096      942 FORMAT(2X,'LINE OVER ESTIMATING F* AT POINT 1 ',/1X,36('-''),
0097      $//,11X,'F = ',E15.7,' * PAR('',12,') + ',E15.7,///)
      WRITE(6,943) S2,LZ,B2
0098      943 FORMAT(2X,'LINE OVER ESTIMATING F* AT POINT 2 ',/1X,36('-''),
0099      $//,11X,'F = ',E15.7,' * PAR('',12,') + ',E15.7,///)
      WRITE(6,650) A,LZ,B,LZ,C
0100      WRITE(6,660) LZ
0101      WRITE(6,670) (E(I),FU12(I),FL12(I),FQ(I),I=1,10)
0102      WRITE(6,670) E2,FE2,FE2,FE2
0103      RETURN
0104
      C OUTPUT WHEN NEXOP7=5
0105      950 WRITE(6,942) S1,LZ,B1
0106      WRITE(6,943) S2,LZ,B2
0107      WRITE(6,660) LZ
0108      WRITE(6,921) (E(I),FU12(I), I=1,10)
0109      WRITE(6,921) E2,FE2
0110      RETURN
      C OUTPUT WHEN NEXOP7=6
0111      960 WRITE(6,941) S12,LZ,B12
0112      WRITE(6,660) LZ
0113      WRITE(6,931) (E(I),FU12(I), I=1,10)
0114      WRITE(6,931) E2,FE2
0115      RETURN
0116      END

```

Figure 10.--continued

DIFF2. Both results are printed out. If it is found that MATRIX creates a nonzero element below the main diagonal of A, then a switch is set to cause the statement STOP to be executed in MAIN. Subroutine CHCKER appears as Figure 11.

6.4 CØNVRG

After each iteration of the algorithm to locate the minimum of the unconstrained function (the W function) subroutine CØNVRG is called to determine if the current point is an acceptable approximation of a point giving the minimum value of the W function. The argument N2 of this routine is given a value of 1 if the point is close enough; otherwise, it is given a value of 2. CØNVRG appears as Figure 12.

6.5 DIFF1

Subroutine DIFF1 computes the first derivatives by numerical differencing. The user-supplied subroutine RESTNT is called 2n times for each gradient evaluated by DIFF1. For the function f, DIFF1 computes the *i*th component of the gradient using the formula

$$(\nabla f(x^0))_i = \frac{f(x^0 + \theta e_i) - f(x^0 - \theta e_i)}{2\theta},$$

where e_i is a vector of zeroes with a 1 in the *i*th component and θ is a small number whose value is assigned by the user. DIFF1 is shown as Figure 13.

6.6 DIFF2

Subroutine DIFF2 computes the second derivatives by numerical differencing. The matrix of second partials is computed using central differencing. The differences are calculated using the gradients of the function. The user-supplied subroutine GRAD1 is called 2n times for each matrix of second partial derivatives evaluated by DIFF2. DIFF2 is Figure 14.

```

FURTRAN IV G LEVEL 21          CHCKER          DATE = 80242          09/21/09

0001          SUBROUTINE CHCKER          CHE00040
C          C          CHE00050
C          MARCH 1971          CHE00060
C          C          CHE00070
C CHCKER COMPUTES AND LIST THE FIRST PARTIAL DERIVATIVES USING GRAD1
C AND THEN USING NUMERICAL DIFFERENCING (DIFF1). IF REQUESTED THE
C SECOND PARTIAL DERIVATIVES ARE COMPUTED AND LISTED USING MATRIX AND
C DIFF2.          CHE00080
0002          IMPLICIT REAL*8(A-N,O-Z)          CHE00120
0003          REAL*4 XEP1,XEP2          CHE00130
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1          CHE00140
0005          COMMON /EQAL/ M, M1, M2          CHE00150
0006          COMMON/EXOPT/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEP1,XEP2          CHE00160
0007          MMZ=1+M+M2          CHE00170
0008          DO 5 J=1,N          CHE00180
0009          DEL(J)=1.2345678          CHE00190
0010          5          CONTINUE          CHE00200
0011          DO 10 I=1,MMZ          CHE00210
0012          IN=I-1          CHE00220
0013          WRITE (6,170) IN          CHE00230
0014          CALL GRAD1 (IN)          CHE00240
0015          WRITE (6,180) (DEL(J),J=1,M)          CHE00250
0016          CALL DIFF1 (IN)          CHE00260
0017          WRITE (6,180) (DEL(J),J=1,N)          CHE00270
0018          10          CONTINUE          CHE00280
C...SOMETIMES ONLY FIRST DERIVATIVES ARE TO BE CHECKED          CHE00290
0019          IF (NEXOP1.LT.4) GO TO 160          CHE00300
0020          WRITE (6,190)          CHE00310
0021          DO 150 J=1,MMZ          CHE00320
0022          IN=I-1          CHE00330
0023          WRITE (6,170) IN          CHE00340
0024          IT=2          CHE00350
0025          DO 30 K=1,N          CHE00360
0026          DO 20 J=1,N          CHE00370
0027          20          A(K,J)=0.          CHE00380
0028          30          CONTINUE          CHE00390
0029          CALL MATRIX (IN,IT)          CHE00400
0030          IF (IT.EQ.1) GO TO 150          CHE00410
0031          DO 50 K=2,N          CHE00420
0032          KM1=K-1          CHE00430
0033          DO 40 J=1,KM1          CHE00440
0034          IF (A(K,J).EQ.0.0) GO TO 40          CHE00450
0035          NEXOP1=5          CHE00460
0036          WRITE (6,210) K,J          CHE00470
0037          GO TO 60          CHE00480
0038          40          CONTINUE          CHE00490
0039          50          CONTINUE          CHE00500
0040          DO 90 K=1,N          CHE00510
0041          DO 70 J=K,N          CHE00520
0042          IF (A(K,J).NE.0.0) GO TO 80          CHE00530
0043          70          CONTINUE          CHE00540
0044          WRITE (6,220) K          CHE00550
0045          GO TO 90          CHE00560
0046          80          WRITE (6,200) K,(A(K,J),J=1,N)          CHE00570
0047          90          CONTINUE          CHE00580
0048          DO 110 K=1,N          CHE00590
0049          DO 100 J=1,N          CHE00600
0050          A(K,J)=0.          CHE00610
0051          110          CONTINUE          CHE00620
0052          WRITE(6,115) IN          CHE00630
0053          115          FORMAT (13H0 CALL DIFF2(,12,1H) )          CHE00640
0054          CALL DIFF2 (IN)          CHE00650
0055          DO 140 K=1,N          CHE00660
0056          DO 120 J=K,N          CHE00670
0057          IF (A(K,J).NE.0) GO TO 130          CHE00680
0058          120          CONTINUE          CHE00690
0059          WRITE (6,220) K          CHE00700
0060          GO TO 140          CHE00710
0061          130          WRITE (6,270) K,(A(K,J),J=1,N)          CHE00720
0062          140          CONTINUE          CHE00730
0063          150          CONTINUE          CHE00740
0064          160          CONTINUE          CHE00750
0065          RETURN          CHE00760
C          CHE00770
0066          170          FORMAT ( 20H0CHCKER.....CONSTRAINT NO. , I3)          CHE00780
0067          180          FORMAT (14H0,24HCHCKER.....1ST PARTIALS/(1X,E20.0,E20.0,E20.0,E20.0,E20.0,E20.0,E20.0)          CHE00790
0068          190          FORMAT (14H0,24HCHCKER.....2ND PARTIALS)          CHE00800
0069          200          FORMAT (14H0ROW, I3 / (1X,E20.0,E20.0,E20.0,E20.0,E20.0,E20.0,E20.0)          CHE00810
0070          210          FORMAT (3H A1,I2,1H,,I2,10H) .NE. (.0)          CHE00820
0071          220          FORMAT (4H ROW,I3,11H ALL ZEROS.)          CHE00830
0072          END          CHE00840
0073          CHE00850

```

Figure 11.--Subroutine CHCKER.

```

FORTRAN IV G LEVEL 21          CONVGR          DATE = 80242          09/21/37

0001          SUBROUTINE CONVGR (N1)          CON00040
C          CON00050
C          OCTOBER 1970          CON00060
C          CON00070
C AFTER EACH ITERATION OF THE ALGORITHM TO LOCATE THE MINIMUM OF THE          CON00080
C PENALTY FUNCTION, CONVGR DETERMINES IF THE CURRENT POINT IS CLOSE          CON00090
C ENOUGH TO THE POINT GIVING THE MINIMUM VALUE OF THE P FUNCTION.          CON00100
C          NI SET EQUAL TO 1 IF MINIMUM HAS BEEN FOUND.          CON00110
C          NI SET EQUAL TO 2 IF MINIMUM HAS NOT BEEN FOUND AND TIME IS NOT UP          CON00120
C          NI SET EQUAL TO 3 OTHERWISE          CON00130
C          DOTT SET EQUAL TO (DEL P)/(INVERSE(DEL(DEL P)))(DEL P) IN OPT          CON00140
C          IMPLICIT REAL*8(A-H,O-Z)          CON00150
0002          REAL*4 RHOIN,RATIO,EPSI,THETA0,XEP1,XEP2          CON00160
0003          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1          CON00170
0004          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10          CON00180
0005          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RMO          CON00190
0006          COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPSI,THETA0,          CON00200
0007          IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XRI(45),PR1,          CON00210
          2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),          CON00220
          3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS          CON00230
          COMMON/EXOPT/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEP1,XEP2          CON00240
          COMMON /TSW/ NSW          CON00250
          NI=2          CON00260
          IF (MN.LE.1) Q1=PO          CON00270
          GO TO (10,20,30), NT9          CON00280
          IF(DABS(DOTT).LT.EPSI) GO TO 70          CON00290
          GO TO 40          CON00300
          IF(DABS(DOTT).LT.(P1-PO)/5.U) GO TO 70          CON00310
          GO TO 40          CON00320
          IF (ADELX.LT.EPSI) GO TO 70          CON00330
          GO TO (50,60), NSW          CON00340
          IF (MN.LE.1) RETURN          CON00350
          IF (PO+XEP2 .LT. Q1) GO TO 75          CON00360
          WRITE (6,80)          CON00370
          GO TO 70          CON00380
          CALL PUNCH          CON00390
          WRITE (6,90)          CON00400
          NI=3          CON00410
          C          CON00420
          C          FOUND THE MINIMUM TO THE SUBPROBLEM.          CON00430
          RETURN          CON00440
          NI=1          CON00450
          Q1 = PO          CON00460
          RETURN          CON00470
          C          CON00480
          80          FORMAT (100H APPARENTLY ROUND OFF ERRORS PREVENT A MORE ACCURATE DE          CON00490
          TERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.)          CON00500
          90          FORMAT (48H0*** TIME IS UP, CALLING EXIT FROM CONVGR. ***)          CON00510
          END          CON00520
0030          CON00530
0031          CON00540
0032          CON00550

```

Figure 12.--Subroutine CONVGR.

```

FORTRAN IV G LEVEL 21          DIFF1          DATE = 80242          09/22/00

0001          SUBROUTINE DIFF1 (IN)                                DIF00060
C                                                                    DIF00050
C          FEBUARY 1971                                          DIF00060
C                                                                    DIF00070
C DIFF1 COMPUTES THE FIRST DERIVATIVES BY NUMERICAL DIFFERENCING. DIF00080
C                                                                    DIF00090
C--USER CAN CALL FOR DIFFERENCING OF SELECTED FUNCTIONS        DIF00100
INPLICIT REAL*8(A-H,O-Z)                                         DIF00110
0002          REAL*4 XEPI,XEP2                                     DIF00120
0003          COMMON/SHARE/X(45),DEL(45),A(45,45),N,N,MN,NP1,NM1 DIF00130
0004          COMMON/EXOPY/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEPI,XEP2 DIF00140
0005          COMMON/STIRX/XSTR(45),XSSS(45),DLL(45)             DIF00150
0006          DO 10 J=1,N                                         DIF00160
0007          XSTR(J)=X(J)                                        DIF00170
0008          DO 30 J=1,N                                         DIF00180
0009          IF (J.EQ.1) GO TO 20                                DIF00190
0010          JMI=J-1                                             DIF00200
0011          X(JMI)=XSTR(JMI)                                    DIF00210
0012          X(J)=XSTR(J)+XEPI                                   DIF00220
0013          CALL RESTNT (IN,ZZ)                                 DIF00230
0014          X(J)=XSTR(J)-XEPI                                   DIF00240
0015          CALL RESTNT (IN,ZZ1)                                DIF00250
0016          DEL(J)=(ZZ2-ZZ1)/(2.*XEPI)                          DIF00260
0017          X(N)=XSTR(N)                                        DIF00270
0018          RETURN                                             DIF00280
0019          END                                                DIF00290
0020

```

Figure 13.--Subroutine DIFF1.

```

FORTRAN IV G LEVEL 21          DIFF2          DATE = 80242          09/22/71

0001          SUBROUTINE DIFF2 (IN)          DIF00040
C          C          DIF00050
C          OCTOBER 1970          DIF00060
C          C          DIF00070
C          DIFF2 COMPUTES THE SECOND DERIVATIVES BY NUMERICAL DIFFERENCING.          DIF00080
C          C          DIF00090
0002          IMPLICIT REAL*8(A-H,D-Z)          DIF00100
0003          REAL*4 XEP1,XEP2          DIF00110
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1          DIF00120
0005          COMMON/EXPOPT/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEP1,XEP2          DIF00130
0006          COMMON/STIRX/XSTR(45),XSSS(45),DOLL(45)          DIF00140
0007          DO 10 J=1,N          DIF00150
0008          10  XSSS(J)=X(J)          DIF00160
0009          DO 50 J=1,N          DIF00170
0010          IF (J.EQ.1) GO TO 20          DIF00180
0011          JMI=J-1          DIF00190
0012          X(JMI)=XSSS(JMI)          DIF00200
0013          20  X(J)=XSSS(J)+XEP1          DIF00210
0014          CALL GRAD1 (IN)          DIF00220
0015          DO 30 I=1,N          DIF00230
0016          30  DOLL(I)=DEL(I)          DIF00240
0017          XI(J)=XSSS(J)-XEP1          DIF00250
0018          CALL GRAD1 (IN)          DIF00260
0019          DO 40 I=J,N          DIF00270
0020          40  A(I,J,I)=(DOLL(I)-DEL(I))/(2.*XEP1)          DIF00280
0021          50  CONTINUE          DIF00290
0022          X(IN)=XSSS(IN)          DIF00300
0023          RETURN          DIF00310
0024          END          DIF00320

```

Figure 14.--Subroutine DIFF2.

6.7 ESTIM

After the minimum of the W function for a given value of r has been located, ESTIM is called. When the minimum of the W function for a given value of r is determined a subproblem is said to be solved. ESTIM performs the computations to estimate the Lagrange multipliers and make the first- and second-order estimates of the final solution of the problem. ESTIM is Figure 15.

6.8 EVALU

In the normal phase, EVALU calls the user-supplied routines to evaluate the objective function and the constraint functions at the current point x . As the constraint functions are being computed, the code performs the calculations to compute the penalty terms of the penalty function. In the feasibility phase this routine puts the negative sum of the violated constraints, which is the objective function of the entry problem, in location F . In location F the current value of the objective function is stored for use by the program. After F has been computed, the value of the penalty function is computed. Also subroutine EVALU computes a value for the dual objective function. This value is only correct at the solution to a subproblem and the value is stored in the location labeled G . EVALU appears as Figure 16.

6.9 FEAS

Subroutine FEAS determines whether the starting point is an interior feasible point or not. If the variables of a problem are supposed to be nonnegative and some of the components of the starting point vector are nonpositive, then those components are changed to small positive numbers. A modification of FEAS must be made if it is desired to change this number. If the starting point does not satisfy the non-trivial constraints (the constraints other than the nonnegativity constraints), then the program goes into the feasibility phase. In this phase the negative of the sum of all the violated inequality constraints

```

FORTRAN IV G LEVEL 21          ESTIM          DATE = 80242          09/22/45

0001          SUBROUTINE ESTIM          EST00040
C          EST00050
C          OCTOBER 1970          EST00060
C          EST00070
C ESTIM PERFORMS THE COMPUTATIONS TO ESTIMATE THE LAGRANGE MULTIPLIERS
C AND MAKE THE FIRST- AND SECOND-ORDER ESTIMATES OF THE FINAL SOLUTION
C OF THE PROBLEM.          EST00080
0002          IMPLICIT REAL*8(A-H,O-Z)          EST00110
0003          REAL*4 RHOIN,RATIO,EPSI,THETA0          EST00120
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1          EST00130
0005          COMMON /EQAL/ M, M1, M2          EST00140
0006          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10          EST00150
0007          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO          EST00160
0008          COMMON/CRST/DELX(45),DELY(45),RHOIN,RATIO,EPSI,THETA0,          EST00170
          1RSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XRI(45),PRI,          EST00180
          2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),          EST00190
          3 PREV3,ADELX, NCTR, NUMINI, NPHASE, NSATIS          EST00200
0009          COMMON /COMPAR/ NFI,NF2,NF3          EST00210
0010          CALL STORE          EST00220
0011          Z10=RATIO**2          EST00230
0012          Z9=RATIO          EST00240
0013          Z1=1.0/Z9+1.0/Z10          EST00250
0014          Z2=Z1+1./Z9**3          EST00260
0015          Z3=1./Z9**3          EST00270
0016          Z4=Z10+Z9          EST00280
0017          Z5=Z3**3          EST00290
0018          Z6=1.0/((Z10-1.0)*(Z9-1.0))          EST00300
0019          Z7=1./Z9          EST00310
0020          Z8=1./(Z9-1.)          EST00320
0021          RQ=1.0/RHO          EST00330
0022          IF (NUMINI-2) 150,80,10          EST00340
0023          10 WRITE (6,330)          EST00350
0024          PO=(PR2-Z4*PR1+Z5*P1)*Z6          EST00360
0025          G=(RATIO*G1-GR1)/(RATIO-1.)          EST00370
0026          DO 20 I=1,N          EST00380
0027          20 X(I)=(XR2(I)-Z4*XRI(I)+Z5*X1(I))*Z6          EST00390
0028          NP=NPHASE          EST00400
0029          NPHASE=4          EST00410
0030          CALL EVALU          EST00420
0031          NPHASE=NP          EST00430
0032          CALL OUTPUT (2)          EST00440
C CHECK TO SEE IF ESTIMATES HAVE CONVERGED          EST00450
0033          GO TO (70,30,70), NPHASE          EST00460
0034          30 DO 50 J=1,M          EST00470
0035          IF (RJ(J)) 40,50,50          EST00480
0036          40 IF (THETA0+RJ(J)) 70,50,50          EST00490
0037          50 CONTINUE          EST00500
0038          GO TO (70,70,60), NT4          EST00510
0039          60 CALL FINAL (NF3)          EST00520
0040          70 CONTINUE          EST00530
0041          80 WRITE (6,340)          EST00540
0042          PO=(Z9*P1-PR1)*Z8          EST00550
0043          G=(RATIO*G1-GR1)/(RATIO-1.)          EST00560
0044          DO 90 I=1,N          EST00570
0045          90 X(I)=(Z9*X1(I)-XR1(I))*Z8          EST00580
0046          NP=NPHASE          EST00590
0047          NPHASE=4          EST00600
0048          CALL EVALU          EST00610
0049          NPHASE=NP          EST00620
0050          CALL OUTPUT (2)          EST00630
C CHECK TO SEE IF ESTIMATES HAVE CONVERGED          EST00640
0051          GO TO (140,100,140), NPHASE          EST00650
0052          100 DO 120 J=1,M          EST00660
0053          IF (RJ(J)) 110,120,120          EST00670
0054          110 IF (RJ(J)+THETA0) 140,120,120          EST00680
0055          120 CONTINUE          EST00690
0056          GO TO (140,130,140), NT4          EST00700
0057          130 CALL FINAL (NF2)          EST00710
0058          140 CONTINUE          EST00720
0059          150 WRITE (6,350)          EST00730
0060          IF (M) 180,180,160          EST00740
0061          160 DO 170 J=1,M          EST00750
0062          RJ(J)=RHO/RJ1(J)          EST00760
0063          170 IF (MZ) 210,210,190          EST00770
0064          190 DO 200 J=1,MZ          EST00780
0065          MNJ=M+J          EST00790
0066          200 RJ(MNJ)=2.*RJ1(MNJ)*RQ          EST00800
0067          210 GO TO (220,240), NT2          EST00810
0068          220 DO 230 I=1,N          EST00820
0069          230 X(I)=RHO/X1(I)          EST00830
0070          240 CALL OUTPUT (2)          EST00840
0071          CALL REJECT          EST00850
0072          IF (NUMINI-2) 280,300,250          EST00860
0073          250 GO TO (280,310,260), NT7          EST00870
C          SECOND ORDER MOVE FOR NEXT MINIMUM          EST00880
0074          260 DO 270 I=1,N          EST00890
0075          270 DELX(I)=Z1*X1(I)-Z2*XRI(I)+Z3*XR2(I)          EST00900

```

Figure 15.--Subroutine ESTIM.

```

0076      280  PR2=PR1
0077          GR2=GR1
0078          PR1=PI
0079          GR1=G1
0080          DO 290 I=1,N
0081          XR2(I)=XR1(I)
0082      290  XR1(I)=X1(I)
0083          RETURN
0084      300  GO TO (280,310,310), NT7
0085      310  DO 320 I=1,N
0086      320  DELX(I)=(X1(I)-XR1(I))*Z7
0087          GO TO 280

C
0088      330  FORMAT (/26H0      2ND ORDER ESTIMATES )
0089      340  FORMAT (/26H0      1ST ORDER ESTIMATES )
0090      350  FORMAT (/25H0 LAGRANGE MULTIPLIERS )
0091          END
EST00910
EST00920
EST00930
EST00940
EST00950
EST00960
EST00970
EST00980
EST00990
EST01000
EST01010
EST01020
EST01030
EST01040
EST01050
EST01060
EST01070

```

Figure 15.--continued

```

FORTRAN IV G LEVEL 21          EVALU          DATE = 80242          09/28/17

0001          SUBROUTINE EVALU          EVA00043
C          C          EVA00050
C          OCTOBER 1970          EVA00063
C          C          EVA00079
C IN THE NORMAL PHASE EVALU CALLS THE USER-SUPPLIED ROUTINES TO EVALUATE EVA00080
C THE OBJECTIVE FUNCTION AND THE CONSTRAINT FUNCTIONS AT THE CURRENT EVA00099
C POINT. IN THE FEASIBILITY PHASE THIS ROUTINE PUTS THE NEGATIVE SUM OF EVA00100
C THE VIOLATED CONSTRAINTS IN LOCATION F.          EVA00110
0002          IMPLICIT REAL*8(A-H,O-Z)          EVA00120
0003          REAL*4 RHOIN,RATIO,EPSI,THETAO          EVA00130
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,N,MN,NP1,NMI          EVA00140
0005          COMMON /EQUAL/ H, H1, M2          EVA00150
0006          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10          EVA00160
0007          COMMON/VALUE/F,G,PO,RSIGMA,RJ(50),RHO          EVA00170
0008          COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPSI,THETAO,          EVA00180
          IRSIG1,G1,X1(45),X2(45),X3(45),J2(45),XRI(45),PRI,          EVA00190
          ZPR2,P1,F1,RJ1(90),DOTT,PGRAJ(45),DIAG(45),          EVA00200
          3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS          EVA00210
          H=0.0          EVA00220
0009          RSIGMA=0.0          EVA00230
0010          F=0.0          EVA00240
0011          NSATIS=2          EVA00250
0012          GO TO (10,100,190,200), NPHASE          EVA00260
0013          C          EVA00270
          C =1 FEASIBILITY          EVA00280
          C =2 NORMAL          EVA00290
          C =3 GUESS          EVA00300
          C =4 ALL FUNCTIONS ARE TO BE EVALUATED          EVA00310
          C FEASIBILITY          EVA00320
0014          10 GO TO (20,40), NT2          EVA00330
          C NON-NEGATIVES INCLUDED          EVA00340
0015          20 DO 30 I=1,N          EVA00350
0016          IF (X(I)) 260,260,30          EVA00360
0017          RSIGMA=RSIGMA-RHO*DLOG(X(I))          EVA00370
0018          40 IF (M.EQ.0) GO TO 90          EVA00380
0019          DO 80 J=1,M          EVA00390
0020          CALL RESTNT (J,RJ(J))          EVA00400
0021          IF (RJ(J).LE.0.0) GO TO 50          EVA00410
0022          IF (RJ(J).GT.0.0) GO TO 60          EVA00420
          C VIOLATION OF A PREVIOUSLY SATISFIED CONSTRAINT          EVA00430
0023          GO TO 260          EVA00440
0024          50 IF (RJ(J).GT.0.0) GO TO 70          EVA00450
          C ALL VIOLATED CONSTRAINTS ADDED INTO OBJECTIVE FUNCTION          EVA00460
0025          F=F-RJ(J)          EVA00470
0026          GO TO 80          EVA00480
0027          60 RSIGMA=RSIGMA-RHO*DLOG(RJ(J))          EVA00490
0028          GO TO 80          EVA00500
          C INDICATES SATISFACTION OF CONSTRAINT(I)ORMORE          EVA00510
0029          70 NSATIS=1          EVA00520
0030          RSIGMA=RSIGMA-RHO*DLOG(RJ(J))          EVA00530
0031          80 CONTINUE          EVA00540
0032          90 CONTINUE          EVA00550
          C EQUALITIES NOT COMPUTED IN FEAS. PHASE          EVA00560
0033          PO=F+RSIGMA          EVA00570
0034          G=F-RHO*DFLOAT(M)          EVA00580
0035          IF (NT2.EQ.1) G=G-RHO*DFLOAT(N)          EVA00590
0036          RETURN          EVA00600
          C REGULAR PHASE          EVA00610
0037          100 GO TO (110,130), NT2          EVA00620
          C NON NEGATIVITIES INCLUDED          EVA00630
0038          110 DO 120 I=1,N          EVA00640
0039          IF (X(I)) 260,260,120          EVA00650
0040          120 RSIGMA=RSIGMA-RHO*DLOG(X(I))          EVA00660
0041          130 IF (M.EQ.0) GO TO 150          EVA00670
0042          DO 140 J=1,M          EVA00680
0043          CALL RESTNT (J,RJ(J))          EVA00690
0044          IF (RJ(J).LE.0.0) GO TO 260          EVA00700
0045          RSIGMA=RSIGMA-RHO*DLOG(RJ(J))          EVA00710
0046          140 CONTINUE          EVA00720
          C EVALUATE AND ADD IN EQUALITY CONSTRAINTS          EVA00730
0047          150 CONTINUE          EVA00740
0048          CALL RESTNT (O,F)          EVA00750
0049          IF (M2) 180,180,160          EVA00760
0050          160 DO 170 I=1,M2          EVA00770
0051          J=I+M          EVA00780
0052          CALL RESTNT (J,RJ(J))          EVA00790
          C ADD INTO THIRD TERM OF P FUNCTION          EVA00800
0053          H=H+(RJ(J))**2          EVA00810
0054          170 CONTINUE          EVA00820
0055          H=H/RHO          EVA00830
0056          180 PO=RSIGMA+H          EVA00840
0057          PO=F+PO          EVA00850
0058          G=2.*H-RHO*DFLOAT(M)          EVA00860
0059          G=G+F          EVA00870
0060          IF (NT2.EQ.1) G=G-RHO*DFLOAT(N)          EVA00880
          C DUAL VALUE          EVA00890
0061          RETURN          EVA00900
          C GUESS PHASE NOT CODED

```

Figure 16.-- Subroutine EVALU.

0062	190	RETURN	EVA00910
	C---	STRAIGHT FUNCTION EVALUATION (MAIN+FEAS ONLY)	EVA00920
0063	200	CONTINUE	EVA00930
0064		IF (M.EQ.0) GO TO 220	EVA00940
0065		DO 210 I=1,M	EVA00950
0066		CALL RESTNT (I,RJ(I))	EVA00960
0067	210	CONTINUE	EVA00970
0068	220	CALL RESTNT (0,F)	EVA00980
	C	EQUALITY CONSTRAINTS	EVA00990
0069		IF (MZ) 250,250,230	EVA01000
0070	230	DO 240 I=1,MZ	EVA01010
0071		KZ=M+I	EVA01020
0072	240	CALL RESTNT (KZ,RJ(KZ))	EVA01030
0073	250	RETURN	EVA01040
	C	CONSTRAINTS VIOLATED NOT SO BEFORE	EVA01050
0074	260	NSATIS=3	EVA01060
0075		PO=10.E35	EVA01070
0076		RETURN	EVA01080
0077		END	EVA01090

Figure 16.--continued

is labeled for use as the objective function in the entry problem. Then the routine calls BODY to minimize this auxiliary function subject to the set of satisfied constraints. In the feasibility phase if a violated constraint is fortuitously satisfied an immediate return to FEAS is made and a new entry problem is begun, including the newly satisfied constraint in the set of constraints active for the entry problem. Thus the entry problem can result in a series of NLP problems being partly solved.

FEAS also contains tests that indicate when a problem does not contain a feasible starting point. Such information is printed out, and control is returned to MAIN with indicators set so that MAIN begins to try the next NLP problem in a stack of problems. When the entry problem is not a convex programming problem the test indicates only that the program is in a region from which it will be unable to locate a feasible starting point. The user, by supplying another starting point, may cause the algorithm to generate another sequence of points that does lead to a feasible starting point. Figure 17 shows FEAS.

6.10 FINAL

Subroutine FINAL contains the test used to determine if a point satisfies the final convergence criterion of the algorithm. If a point does satisfy the criterion chosen by use of option 5 then N2, the argument of this routine, is given a value of 1; otherwise it is given a value of 2. FINAL is called following the computation of the solution estimates, which are made after the solution of the subproblem. FINAL appears as Figure 18.

6.11 GRAD

The gradient of the W function is given by the formula

$$\nabla W(x,r) = \nabla f(x) - \sum_{i=1}^m \frac{r}{g_i(x)} \nabla g_i(x) + \sum_{i=m+1}^{m+p} \frac{2h_i(x)}{r} \nabla h_i(x) .$$

```

FORTRAN IV G LEVEL 21          FEAS          DATE = 80242          09/23/33

0001          SUBROUTINE FEAS          FEA00040
C          C          FEA00050
C          AUGUST 1971          FEA00060
C          C          FEA00070
C FEAS DETERMINES WHETHER THE STARTING POINT IS FEASIBLE. IF IT IS NOT, FEA00080
C FEAS LOOKS FOR A FEASIBLE ONE. IF NONE EXISTS, A MESSAGE IS PRINTED FEA00090
C AND CONTROL RETURNS TO MAIN. FEA00100
0002          IMPLICIT REAL*8(A-M,O-Z) FEA00110
0003          REAL*4 RHOIN,RATIO,EPSI,THETA0 FEA00120
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 FEA00130
0005          COMMON /OPTMS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 FEA00140
0006          COMMON/VALUE/F,G,PO,RSIGNA,RJ(90),RMO FEA00150
0007          COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0, FEA00160
          IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XRI(45),PRI, FEA00170
          2PR2,P1,F1,RJ1(90),DOTF,PGRAD(45),DIAS(45), FEA00180
          3 PREV3,ADELX, NYCTR, NUMINI, NPHASE, NSATIS FEA00190
0008          NPHASE=1 FEA00200
0009          GO TO (10,50), NTZ FEA00210
0010          10 NFIX=1 FEA00220
0011          DO 30 I=1,N FEA00230
0012          IF (X(I)) 20,20,30 FEA00240
0013          20 NFIX=2 FEA00250
0014          X(I)=1.E-05 FEA00260
0015          30 CONTINUE FEA00270
0016          GO TO (50,40), NFIX FEA00280
0017          40 NPHASE=4 FEA00290
0018          CALL EVALU FEA00300
C JUST GET ALL CONSTRAINTS EVALUATED FEA00310
0019          NPHASE=1 FEA00320
0020          WRITE (6,130) FEA00330
0021          CALL OUTPUT (2) FEA00340
0022          50 IF (M) 90,90,60 FEA00350
0023          60 DO 70 I=1,M FEA00360
0024          IF (RJ(I)) 100,100,70 FEA00370
0025          70 CONTINUE FEA00380
0026          80 CALL TIMEC FEA00390
0027          WRITE (6,140) FEA00400
0028          G=0.0 FEA00410
0029          CALL RESTMT (0,F) FEA00420
0030          CALL OUTPUT (2) FEA00430
0031          90 RETURN FEA00440
0032          100 CALL BODY FEA00450
0033          IF(NPHASE .EQ. 5) RETURN FEA00460
0034          DO 110 I=1,M FEA00470
0035          IF (RJ(I)) 120,120,110 FEA00480
0036          110 CONTINUE FEA00490
0037          GO TO 80 FEA00500
0038          120 WRITE (6,150) FEA00510
C TO INDICATE TO MAIN TO START ON NEXT PROBLEM. FEA00520
0039          NPHASE=5 FEA00530
0040          GO TO 90 FEA00540
C          FEA00550
0041          130 FORMAT (1H0,2X,48HMADE VIOLATED NON-NEGATIVITIES SLIGHTLY POSITIVE FEA00560
          I) FEA00570
0042          140 FORMAT (51H0****THE FEASIBLE STARTING POINT TO BE USED IS ...) FEA00580
0043          150 FORMAT (3X,89HTHIS PROBLEM POSSESSES NO FEASIBLE STARTING POINT, W FEA00590
          ILL LOOK FOR DATA FOR NEXT PROBLEM. ) FEA00600
0044          END FEA00610

```

Figure 17.--Subroutine FEAS.

FORTRAN IV G LEVEL 21

FINAL

DATE = 80242

09/28/56

0001		SUBROUTINE FINAL (N2)	FIN00040
	C		FIN00050
	C	OCTOBER 1970	FIN00060
	C		FIN00070
	C		FIN00080
	C	FINAL CONTAINS THE TESTS USED TO DETERMINE WHETHER A POINT SATISFIES	FIN00090
	C	THE FINAL CONVERGENCE CRITERION CHOSEN TO DETERMINE IF THE NLP	FIN00100
	C	PROBLEM HAS BEEN SOLVED.	FIN00110
	C	N2 SET EQUAL TO 1 IF CONVERGENCE CRITERION IS SATISFIED.	FIN00120
	C	N2 SET EQUAL TO 2 OTHERWISE.	FIN00130
0002		IMPLICIT REAL*8(A-H,O-Z)	FIN00140
0003		REAL*8 RHOIN,RATIO,EPSI,THETA0	FIN00150
0004		COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1	FIN00160
0005		COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RMD	FIN00170
0006		COMMON/CRST/DELX(45),DELK(45),RHOIN,RATIO,EPSI,THETA0,	FIN00180
		IRSIG1,G1,X1(45),X2(45),X3(45),YR2(45),XRI(45),PRI,	FIN00190
		ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),	FIN00200
		3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS	FIN00210
0007		GO TO (10,20,30), NT5	FIN00220
0008	10	EPSIL=DABS(F/G-1.)	FIN00230
0009		IF (EPSIL-THETA0) 50,50,70	FIN00240
0010	20	IF(DABS(RSIGMA)-THETA0) 50,50,70	FIN00250
0011	30	IF (NUMINI-1) 50,40,40	FIN00260
0012	40	PEST=PRI-(PRI-PO)/(1.-1./SQRT(RATIO))	FIN00270
0013		EPSIL=DABS(PEST/G-1.)	FIN00280
0014		IF (EPSIL-THETA0) 50,70,70	FIN00290
0015	50	N2=1	FIN00300
0016		GO TO (80,60), NT6	FIN00310
0017	60	CALL PUNCH	FIN00320
0018		GO TO 80	FIN00330
0019	70	N2=2	FIN00340
0020	80	RETURN	FIN00350
0021		END	FIN00360

Figure 18.--Subroutine FINAL.

Subroutine GRAD calls the user-supplied subroutine GRAD1 to compute $\nabla f(x)$, $\nabla g_i(x)$, and $\nabla h_i(x)$ and performs the computations to evaluate $\nabla W(x,r)$. The negative of the gradient of the W function (i.e., $-\nabla W$) is left in the array DELXO when GRAD returns control to the calling routine. When the argument of GRAD has a value of 2 this is all that is done. However, when it has a value of 1, GRAD also does part of the computations needed to evaluate the matrix of second partial derivatives of W at x . Subroutine GRAD is shown in Figure 19.

6.12 INVERS

When Newton's method is used to minimize the W function for a given value of r , it maps the negative of the gradient of W with the inverse of the matrix of second partial derivatives of W evaluated at x . A positive definite matrix will always give a vector along which the value of W will initially decrease. That is, for iteration i , a move is made along the vector S^i , given by the formula

$$S^i = -[\nabla^2 W(x,r)]^{-1} \nabla W(x,r).$$

This is equivalent to solving the set of simultaneous linear equations

$$[\nabla^2 W(x,r)]S^i = -\nabla W(x,r)$$

for S^{i*} , where $\nabla^2 W$ and ∇W are respectively the matrix of second derivatives and gradient vector of W evaluated at (x,r) .

Subroutine INVERS solves this set of equations for S^i using an L-U decomposition method (the Crout procedure). If it is determined that the matrix of second partials is not positive definite, a different procedure is used to obtain a direction S^i . A complete discussion of this procedure is given on page 167 of Fiacco and McCormick [10]. When

*In the program the vector S^i is called DELX.

```

0001      SUBROUTINE GRAD (IS)
C
C          OCTOBER 1970
C
C GRAD COMPUTES THE GRADIENT OF THE PENALTY FUNCTION AND THE OUTER
C PRODUCT FACTORS OF THE MATRIX OF SECOND PARTIALS OF P.
C IF (IS=1) ACCUM. MATRIX OF 2ND PARTIALS IF (IS=2) DONT
0002      IMPLICIT REAL*8(A-M,O-Z)
0003      REAL*4 RHOIN,RATIO,EPST,THETAO
0004      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1
0005      COMMON /EQUAL/ H, M1, M2
0006      COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10
0007      COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO
0008      COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPST,THETAO,
1RSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,
2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),
3 PREV3,ADELX, NCTR, NUMINI, NPHASE, NSATIS
0009      GO TO (10,30), IS
0010      10  DO 20 I=1,N
0011          DO 20 J=1,I
0012              A(I,J)=0.
0013          DO 40 I=1,N
0014              DO 40 DELXO(I)=0.
C THIS SECTION WORKS CORRECTLY IN FEASIBILITY PHASE AS WELL AS NORMAL PH
0015      GO TO (50,80), NT2
0016      50  DO 70 I=1,N
0017          DELXO(I)=-RHO/X(I)
0018      GO TO (60,70), IS
0019      60  A(I,I)=-DELXO(I)/X(I)
0020      70  CONTINUE
0021      80  CONTINUE
0022          IF (M.LE.0) GO TO 180
0023          DO 170 K=1,M
0024              CALL GRAD1 (K)
0025              IF (RJK).GT.0.0) GO TO 110
C ALL VIOLATED CONSTRAINT GRADS ADDED TO OBJ. FUNCTION
0026      DO 100 I=1,N
0027          IF (DEL(I)) 90,100,90
0028      90  DELXO(I)=DELXO(I)-DEL(I)
0029      100 CONTINUE
0030      GO TO 170
0031      110 TT=RHO/RJ(K)
0032          DO 160 I=1,N
0033              IF (DEL(I)) 120,160,120
C IF DEL(I)=0 SKIP ALL THE FOLLOWING COMPUTATION INVOLVING * BY DEL(I)
0034      120 T=TT*DEL(I)
0035          DELXO(I)=DELXO(I)-T
0036      GO TO (130,160), IS
0037      130 T=T/RJ(K)
0038          DO 150 JJ=1,I
0039              IF (DEL(JJ)) 140,150,140
0040      140 A(I,JJ)=A(I,JJ)+T*DEL(JJ)
0041      150 CONTINUE
0042      160 CONTINUE
0043      170 CONTINUE
C EQUALITY CHANGES FOR GRAD
0044      180 IF (MZ.LE.0) GO TO 250
0045          GO TO (250,190,250), NPHASE
0046      190 RQ=2./RHO
0047          DO 240 J=1,MZ
0048              K=M+J
0049              CALL GRAD1 (K)
0050              TT=RQ*RJ(K)
0051              DO 230 I=1,N
0052                  IF (DEL(I).EQ.0.0) GO TO 230
0053                  DELXO(I)=DELXO(I)+DEL(I)*TT
0054              GO TO (200,230), IS
0055      200 T=RQ*DEL(I)
0056          DO 220 JJ=1,I
0057              IF (DEL(JJ)) 210,220,210
0058      210 A(I,JJ)=A(I,JJ)+T*DEL(JJ)
0059      220 CONTINUE
0060      230 CONTINUE
0061      240 CONTINUE
0062      250 GO TO (260,280), IS
0063      260 DO 270 I=1,N
0064          270 DIAG(I)=A(I,I)
0065      280 GO TO (290,330,290), NPHASE
C LEAVES NEGATIVE GRADIENT IN DELP
0066      290 DO 300 I=1,N
0067          300 DELXO(I)=-DELXO(I)
0068      310 ADELX=0.
0069          DO 320 I=1,N
0070          320 ADELX=ADELX+DELXO(I)**2
0071          ADELX=DSQRT(ADELX)
0072      RETURN
0073      330 CALL GRAD1 (0)
0074          DO 340 I=1,N
0075          340 DELXO(I)=-DELXO(I)-DEL(I)
C LEAVES THE NEG. GRAD OF P IN DELXO
0076      GO TO 310
0077      END

```

Figure 19.--Subroutine GRAD.

this occurs the program prints out "ORTHOGONAL MOVE." This also warns the user that the problem is probably not a convex program. Subroutine INVERS appears as Figure 20.

6.13 LMULT

Subroutine LMULT, called by subroutine SENS, calculates the Lagrange multiplier sensitivities according to steps 5 and 6 of Algorithm 2.1 (Section 2). LMULT is given in Figure 21.

6.14 MAIN

MAIN is the program that initiates the SENSUMT algorithm to solve a nonlinear programming (NLP) problem; it is not a subroutine. The input of parameters and options, as well as a starting point, is done in MAIN, and the call to READIN (a user-supplied subroutine) is made to allow the user to read in data needed to evaluate his objective function and constraint functions. Starting point data or blank cards should always be supplied by the user because starting point data cards are always read in MAIN. Subroutine FEAS is called to obtain a feasible starting point by making use of the SENSUMT algorithm to solve the entry problem if the user-supplied point is not feasible. The actual solution of the NLP problem using the SENSUMT algorithm is supervised by subroutine BØDY. The calls to SENS (sensitivity subroutine) and BØUND (bound subroutine) are also done in MAIN. When calculating bounds, after the solution and sensitivity analysis of the unperturbed problem, MAIN reiterates SENSUMT to solve the subject problem with the perturbed data. Perturbations and readjustments in the problem data are done in PERT (perturbation subroutine), on call by MAIN. Subroutine MAIN is shown as Figure 22.

6.15 ØPT

The purpose of subroutine ØPT is to obtain a $\bar{\theta} > 0$ such that $W(\bar{x} + \bar{\theta}\bar{S})$ is a minimum with respect to θ along the vector $\bar{x} + \bar{\theta}\bar{S}$.

```

FORTRAN IV G LEVEL 21          INVERS          DATE = 80242          12/20/20

0001          SUBROUTINE INVERS (NSME)          INV00040
C          OCTOBER 1970          INV00050
C          INV00060
C          INV00070
C INVERS SOLVES THE SET OF EQUATIONS FOR THE MOVE-VECTOR USING THE INV00080
C CRUT PROCEDURE. IF THE MATRIX IS NOT POSITIVE DEFINITE, A DIFFERENT INV00090
C METHOD IS USED.          INV00100
C          INV00110
C****PERFORMING A L-U DECOMPOSITION OF THE MATRIX A, TAKING ADVANAGE OF INV00120
C****THE SYMMETRY OF THE A MATRIX.          INV00130
C****IF A NON-POSITIVE PIVOT CANDIDATE IS GENERATED, THEN MCCORMICK,S INV00140
C****PROCEDURE IS USED(SEE PP. 167-168 IN FIACCO AND MCCORMICK).          INV00150
C          INV00160
C****IF NSME =1 WORKING WITH A NEW A MATRIX. IF NSME= 2 USING PREVIOUS INV00170
C****A MATRIX, BUT HAVE A NEW RIGHT-HAND-SIDE.          INV00180
C          INV00190
C****NINV IS THE NUMBER OF NON-POSITIVE PIVOT CANDIDATES GENERATED. INV00200
IMPLICIT REAL*8(A-H,O-Z)          INV00210
0002          REAL*4 RHOIN,RATIO,EPSI,THETAO,XEPI,XEP2          INV00220
0003          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1          INV00230
0004          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10          INV00240
0005          COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPSI,THETAO,          INV00250
0006          IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PRI,          INV00260
          ZPR2,P1,F1,RJ1(90),DOIT,PCRAD(45),DIAG(45),          INV00270
          3 PREV3,ADELX, NYCTR, NUMINI, NPHASE, NSATIS          INV00280
          COMMON/EXPOPT/NEKOP1,NEKOP2,NEKOP3,NEKOP4,NEKOP5,XEPI,XEP2          INV00290
          DIMENSION B(45)          INV00300
          EQUIVALENCE (B,DELX)          INV00310
          GO TO(20, 170), NSME          INV00320
          NINV=0          INV00330
          IF (A(1,1)) 40,30,50          INV00340
          NINV=1          INV00350
          GO TO 70          INV00360
          NINV=1          INV00370
          A(1,1)=1./A(1,1)          INV00380
          DO 60 I=2,N          INV00390
          A(1,I)=A(1,I)*A(1,1)          INV00400
          DO 160 J=2,N          INV00410
          JM1=J-1          INV00420
          T=0.          INV00430
          DO 90 I=1,JM1          INV00440
          IF (A(I,J)) 80,90,80          INV00450
          T=T+A(J,I)*A(I,J)          INV00460
          CONTINUE          INV00470
          A(J,J)=A(J,J)-T          INV00480
          IF (A(J,J)) 110,100,120          INV00490
          NINV=NINV+1          INV00500
          GO TO 170          INV00510
          NINV=NINV+1          INV00520
          A(J,J)=1./A(J,J)          INV00530
          IF (J.EQ.N) GO TO 170          INV00540
          JP1=J+1          INV00550
          DO 150 L=JP1,N          INV00560
          T=0.          INV00570
          DO 140 I=1,JM1          INV00580
          IF (A(I,J)) 130,140,130          INV00590
          T=T+A(I,L)*A(I,J)          INV00600
          CONTINUE          INV00610
          A(L,J)=A(L,J)-T          INV00620
          A(J,L)=A(L,J)*A(J,J)          INV00630
          CONTINUE          INV00640
          CONTINUE          INV00650
          CONTINUE          INV00660
          IF (NINV) 180,180,290          INV00670
          B(1)=B(1)*A(1,1)          INV00680
          DO 210 J=2,N          INV00690
          T=0.          INV00700
          JM1=J-1          INV00710
          DO 200 I=1,JM1          INV00720
          IF (A(J,I)) 190,200,190          INV00730
          T=T+A(J,I)*B(I)          INV00740
          CONTINUE          INV00750
          B(J)=(B(J)-T)*A(J,J)          INV00760
          CONTINUE          INV00770
          DO 240 I=1,NM1          INV00780
          NMK=N-I          INV00790
          DO 230 J=1,I          INV00800
          L=NP1-J          INV00810
          IF (A(NMK,L)) 220,230,220          INV00820
          B(NMK)=B(NMK)-A(NMK,L)*B(L)          INV00830
          CONTINUE          INV00840
          CONTINUE          INV00850
          GO TO (280,260), NT3          INV00860
          WRITE (6,430)          INV00870
          WRITE (6,420) (DELXO(I),I=1,N)          INV00880
          WRITE (6,440)          INV00890
          WRITE (6,420) (DELX(I),I=1,N)          INV00900
0068

```

Figure 20.--Subroutine INVERS.

0069	280	RETURN	INVO0910
	C---	COMPUTE ORTHOGONAL MOVE	INVO0920
0070	290	CONTINUE	INVO0930
0071		DO 350 I=1,N	INVO0940
0072		I=N-I+1	INVO0950
0073		IF (A(I,I)) 310,300,320	INVO0960
0074	300	B(I)=0.0	INVO0970
0075		GO TO 350	INVO0980
0076	310	B(I)=1.0	INVO0990
0077		GO TO 330	INVO1000
0078	320	B(I)=0.0	INVO1010
0079	330	IPI=I+1	INVO1020
0080		IF (IPI.GT.N) GO TO 350	INVO1030
0081		DO 340 J=IPI,N	INVO1040
0082	340	B(I)=B(I)-A(I,J)*B(J)	INVO1050
0083	350	CONTINUE	INVO1060
0084		GO TO 360	INVO1070
	C--	CHECK MAYBE DO DIFF FOR P.S.D.	INVO1080
0085	360	ZC2=0.0	INVO1090
0086		DO 370 I=1,N	INVO1100
0087	370	ZC2=ZC2+DELXO(I)*B(I)	INVO1110
0088		IF (ZC2) 380,400,400	INVO1120
0089	380	DO 390 I=1,N	INVO1130
0090	390	B(I)=-B(I)	INVO1140
0091	400	WRITE (6,450)	INVO1150
	C MCC	ZANGWILL ONE MOD	INVO1160
0092		IF (NEXP2.NE.2) GO TO 250	INVO1170
0093		DO 410 K=1,N	INVO1180
0094	410	B(K)=B(K)+DELXO(K)	INVO1190
0095		GO TO 250	INVO1200
	C		INVO1210
0096	420	FORMAT (7E17.8)	INVO1220
0097	430	FORMAT (1H0,6X,12HDEL P VECTOR)	INVO1230
0098	440	FORMAT (1H0,6X,24HSECOND ORDER MOVE VECTOR)	INVO1240
0099	450	FORMAT (1H0,6X,15HORTHOGONAL MOVE)	INVO1250
0100		END	INVO1260

Figure 20.--continued

FORTRAN IV G LEVEL 21

LMULT

DATE = 80242

12/00/23

```

0001          SUBROUTINE LMULT(INO,DELMU,DEM,DU)
C
C          1 MARCH 1976
C
C SUBROUTINE LMULT IS USED TO ESTIMATE THE PARTIAL DERIVATIVES OF THE
C LAGRANGE MULTIPLIERS. IT IS CALLED BY SENS WHEN NEXOP4 = 1. USING
C THE DIFFERENTIABILITY OF X(I), IT TAKES ADVANTAGE OF THE RELATIONS
C U(I) = RHO/G(I)*2 AND W(J) = 2*W(J)/RHO, DIFFERENTIATING THEM WITH
C RESPECT TO THE PARAMETERS. THIS SUBROUTINE WAS CODED BY R.L. ARMACOSTI
0002          IMPLICIT REAL*8(A-H,D-Z)
0003          REAL*4 RHOIN,RATIO,EPSI,THETAO,XEP1,XEP2
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1
0005          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO
0006          COMMON/EQUAL/H, H1, MZ
0007          COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPSI,THETAO,
          INTCTR,NUMINI,X1(45),X2(45),X3(45),XR2(45),XRI(45),PRI,
          2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),
          3PREV3,ADELX,RSIG1,G1,NPHASE,NSATIS
0008          DIMENSION DELMU(90),DU(90),KTEST(45)
0009          MMZ = M + MZ
0010          GO TO (1,2,3,3), IND
C IND = 0
0011          DO 100 I=1,MMZ
0012          DELMU(I) = 0.
0013          RETURN
C IND = 1
0014          DO 50 I=1,MMZ
0015          DELMU(I) = RJ(I)
0016          RETURN
C IND = 2
0017          DO 60 I=1,MMZ
0018          DELMU(I) = (DELMU(I) - RJ(I))/DEM
0019          RETURN
0020          DO 70 I=1,MMZ
0021          CALL GRAD1(I)
0022          CALL RESTNT(I,VAL)
0023          SUM = 0.
0024          DO 71 JJ=1,N
0025          SUM = SUM + DEL(JJ)*DELX(JJ)
0026          IF(INO.EQ.4) GO TO 80
0027          DU(I) = -(SUM + DELMU(I))*RHO/(IVAL**2)
0028          GO TO 70
0029          80 DU(I) = 2.*(SUM + DELMU(I))/RHO
0030          70 CONTINUE
0031          RETURN
0032          END
LMU00040
LMU00050
LMU00060
LMU00070
LMU00080
LMU00090
LMU00100
LMU00110
LMU00120
LMU00130
LMU00140
LMU00150
LMU00160
LMU00170
LMU00180
LMU00190
LMU00200
LMU00210
LMU00220
LMU00230
LMU00240
LMU00250
LMU00260
LMU00270
LMU00280
LMU00290
LMU00300
LMU00310
LMU00320
LMU00330
LMU00340
LMU00350
LMU00360
LMU00370
LMU00380
LMU00390
LMU00400
LMU00410
LMU00420
LMU00430
LMU00440
LMU00450
LMU00460
LMU00470
LMU00480
LMU00490
LMU00500

```

Figure 21.--Subroutine LMULT.

FORTHAN IV G LEVEL 21

MAIN

DATE = 80242

12/12/30

```

C
C
C
C
C MAIN IS THE PROGRAM THAT INITIATES THE SUMT ALGORITHM. THE INPUT OF
C PARAMETERS, OPTIONS, AND STARTING POINT IS DONE IN MAIN. AFTER THE
C SOLUTION OF ONE NLP PROBLEM MAIN LOCKS FOR DATA FOR ANOTHER NLP PROB.
C MAIN HAS BEEN MODIFIED BY AHMACOST(1976) TO INCLUDE SENSITIVITY
C ROUTINES AND BY GHAEMI(1979) TO INCLUDE BOUND CALCULATIONS
0001      IMPLICIT REAL*8(A-H,O-Z)
0002      REAL*4 RHOIN,RATIO,EPSI,THETA0,XEP1,XEP2,TMMAX
0003      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1
0004      COMMON /EQUAL/ H, M1, MZ
0005      COMMON /OPTNS/ NT1,NT2,NT3,NT4,N,5,NT6,NT7,NT8,NT9,NT10
0006      COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO
0007      COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0,
      IRSIG1,G1,X1(45),X2(45),X3(45),XP2(45),XRI(45),PRI,
      ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),
      3 PREV3,ADDELX, NTCTR, NUMINI, NPHASE, NSATIS
0008      COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
0009      COMMON/EXOPT/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEP1,XEP2
0010      COMMON/ABG/LY,LZ,PER(45)
0011      COMMON/ABG1/FE1,FE2
0012      COMMON/ABG2/DF1(45),DF2(45)
0013      COMMON/OP6/NEXOP6
0014      COMMON/OP7/NEXOP7
0015      COMMON/ABG4/XY(45)
0016      COMMON/ABG6/XX(45)
0017      DIMENSION XZ(45)
C
C PARAMETER CARD
0018      10 READ(5,50,END=40) EPSI,RHOIN,THEIA0,RATIO,TMMAX,M,N,MZ
0019          LY=0.
0020          LZ=0.
0021          RRR=RHOIN
0022      20 CONTINUE
0023          IF(LY .EQ. 0.) GO TO 15
0024          IF(NEXOP3 .EQ. 0.) GO TO 40
0025      15 CALL SET (TMMAX)
C
C INITIAL X VECTOR CARD FORMAT
0026          IF(LY .GT. 0.) GO TO 200
0027          READ (5,60) (X(I),I=1,N)
0028          DO 4000 I=1,N
0029              XZ(I)=X(I)
0030      4000 CONTINUE
0031      200 RHOIN=RRR
0032          NTCTR=0
0033          NP1=N+1
0034          NM1=N-1
C
C SUBROUTINE READIN IS UNDER PROGRAMMER CONTROL
0035          IF(LY .GT. 0.) GO TO 210
0036          CALL READIN
C
C OPTION CARD FOLLOWS PROGRAMMERS DATA
0037          READ (5,80) NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10
0038          GO TO 215
0039      210 CALL READIN
0040          CALL PERT
0041          IF(LZ .GT. NPAR) GO TO 40
0042      215 WRITE (6,110)
0043          WRITE (6,120) N,M,MZ,TMMAX,RHOIN,RATIO,EPSI,THETA0
0044          WRITE (6,130)
0045          WRITE (6,80) NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10
C---READ TOLERANCES
0046          IF(LY .GT. 0.) GO TO 220
0047          READ (5,60) XEP1, XEP2
0048      220 WRITE (6,90)
0049          WRITE (6,70) XEP1, XEP2
C---READ SECOND OPTION CARD
0050          IF(LY .GT. 0.) GO TO 230
0051          READ (5,80) NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,NEXOP6, NEXOP7
0052      230 WRITE (6,100)
0053          WRITE (6,80) NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,NEXOP6,NEXOP7
0054          CALL TIMEC
0055          NPHASE=4
C
C --- JUST TO GET AN INITIAL PRINTOUT
0056          CALL EVALU
0057          PO=0.0
0058          G=0.0
0059          H=0.0
0060          RSIGMA=0.0
0061          CALL OUTPUT (2)
0062          CALL STORE
0063          IF (NEXOP1.GT.1) CALL CMCKER
0064          IF (NEXOP1.EQ.3) STOP 01072
0065          IF (NEXOP1.EQ.5) STOP 01104
0066          CALL FEAS
C
C NPHASE=5 INDICATES EITHER TIME RAN OUT OR NO FEASIBLE POINT WAS FOUND
0067          GO TO (30,30,30,30,10), NPHASE
0068      30 NPHASE=7

```

Figure 22.--MAIN

```

0069          NTCTR=0                                MAI00930
0070          CALL BUDY                               MAI00940
0071          IF(NPHASE.EQ.5) GO TO 17                MAI00950
0072          IF(INEXUP3.EQ.1) GO TO 18                MAI00960
0073          IF(INEXUP3.EQ.3) GO TO 19                MAI00970
C FOLLOWING STATEMENT IS INCLUDED FOR WATER POLLUTION PROJECT ONLY MAI00980
C IT HAS TO BE REMOVED LATER (INT8=5 IMPLIES POLLUTION PROJECT) MAI00990
0074          IF(INT8 .EQ. 5) CALL PROF               MAI01000
0075          GO TO 17                                  MAI01010
0076          18 ISENS = 4                             MAI01020
0077          CALL SENS                                 MAI01030
0078          IF(LY.GT.0.) CALL BOUND                  MAI01040
0079          LY=LY+1                                   MAI01050
0080          GO TO 17                                  MAI01060
0081          19 ISENS = 2                             MAI01070
0082          DO 21 I=1,12                             MAI01080
0083          CALL SENS                                 MAI01090
0084          IF(LY.GT.0.) CALL BOUND                  MAI01100
0085          LY=LY+1                                   MAI01110
0086          21 CONTINUE                               MAI01120
0087          17 DO 300 I=1,N                           MAI01130
0088          X(I)=XZ(I)                               MAI01140
0089          300 CONTINUE                             MAI01150
0090          GO TO 20                                  MAI01160
0091          40 STOP                                  MAI01170
C                                                     MAI01180
C PARAMETER CARD                                     MAI01190
0092          50 FORMAT (5E12.0,3I4)                   MAI01200
C INITIAL X VECTOR CARD FORMAT                     MAI01210
0093          60 FORMAT (6E12.6)                      MAI01220
0094          70 FORMAT (6E20.7)                      MAI01230
C OPTION CARD FORMAT                                MAI01240
C OPTION CARD FORMAT                                MAI01250
0095          80 FORMAT (10I7)                        MAI01260
0096          90 FORMAT(13H0 TOLERANCES )             MAI01270
0097          100 FORMAT (26H0 SECOND SET OF OPTIONS ) MAI01280
0098          110 FORMAT (56H1 NONLINEAR PROGRAMMING ROUTINE-SUMT VERSION 4 03/01/73MAI01290
1 ]                                                  MAI01300
0099          120 FORMAT (1H0,5X,2HN=13.6X,2HM=13.6X,3MNZ=13//8X,10HMAX. TIME=E14.7,MAI01310
14X,2HR=E14.7,4X,6HRATIO=E14.7,6X,8HEPSILON=E14.7,4X,6HMETAS=E14.7)MAI01320
0100          130 FORMAT (18H0 OPTIONS SELECTED)      MAI01330
0101          END                                     MAI01340

```

Figure 22.--continued

AD-A096 693

GEORGE WASHINGTON UNIV WASHINGTON DC INST FOR MANAGE--ETC F/6 9/2

A USER'S MANUAL FOR SENSUMT: A PENALTY FUNCTION COMPUTER PROGRA--ETC(U)

OCT 80 A V FIACCO, A GHAEMI

DAAG29-79-C-0062

UNCLASSIFIED

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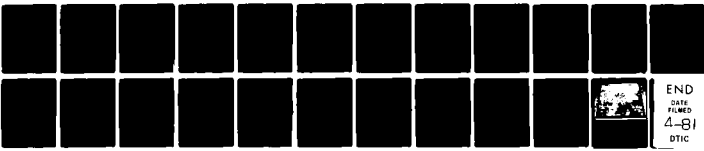
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NL

2 of 2

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The vector \bar{S} , along which the search for the minimum is made, is determined in subroutine XMOVE. The method used to locate the minimum along the vector is the Golden Section search method, which is based on the Fibonacci search method (see [10, p. 193]).

The Golden Section method is a one-dimensional search method to find the minimum of the function that does not require the computation of derivatives. Figure 23 shows \emptyset PT.

6.16 \emptyset UTPUT

Subroutine \emptyset UTPUT contains most of the "write" statements used to print out information on the results of solving an NLP problem. It is used to print out information after each iteration and also to print out the solution estimates and the estimates of the Lagrange multipliers. \emptyset UTPUT is given in Figure 24.

6.17 PARDIF

The subroutine PARDIF is called by subroutine SENS to assign a differencing increment for use in the central differencing formulas indicated in Steps 2 and 4 of Algorithm 2.1. It assigns a fixed value to the differencing interval when the sensitivity analysis is conducted at the final subproblem, or assigns a value over a specified interval for sensitivity analysis at the final subproblem. When a trajectory sensitivity analysis is performed, PARDIF returns a differencing interval which is dependent on the particular subproblem involved, refining the value as the subproblem approaches the final one. Subroutine PARDIF is shown in Figure 25.

6.18 PERT

Subroutine PERT, called by the program MAIN, performs the book-keeping regarding various parametric changes and adjustments required in the process of bound calculation. Figure 26 is subroutine PERT.

```

FORTRAN IV G LEVEL 21          OPT          DATE = 80243          10/19/39

0001          SUBROUTINE OPT          OPT00040
C          OPT00050
C          MARCH 1971          OPT00060
C          OPT00070
C OPT LOOKS FOR A MINIMUM ALONG THE SEARCH VECTOR USING THE GOLDEN          OPT00080
C SECTION SEARCH METHOD.          OPT00090
0002          IMPLICIT REAL*(A-H,O-Z)          OPT00100
0003          REAL*4 RHOIN,RATIO,EPST,THETA0          OPT00110
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1          OPT00120
0005          COMMON/VALUE/F,G,P0,RSIGMA,RJ(90),RHO          OPT00130
0006          COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPST,THETA0,          OPT00140
          IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,          OPT00150
          ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),          OPT00160
          3 PREV3,ADELX, NTCTR, NUMINE, NPHASE, NSATIS          OPT00170
0007          KSW=1          OPT00180
0008          N405=1          OPT00190
0009          P31=PU          OPT00200
0010          ISW=1          OPT00210
0011          DOTT=0.          OPT00220
0012          DO 10 J=1,N          OPT00230
0013          DOTT=DOTT+DELX(J)*DELX0(J)          OPT00240
0014          GO TO 40          OPT00250
0015          20 DO 30 I=1,N          OPT00260
0016          30 DELX(I)=-DELX(I)          OPT00270
0017          40 CONTINUE          OPT00280
0018          N404=0          OPT00290
0019          MN=MN+1          OPT00300
C MN IS NOW NUMB. OF POINTS AFTER MIN ACHIEVED          OPT00310
0020          NTCTR=NTCTR+1          OPT00320
0021          DO 50 I=1,N          OPT00330
0022          50 X2(I)=X(I)          OPT00340
0023          PX1=PO          OPT00350
0024          N401=0          OPT00360
0025          60 N401=N401+1          OPT00370
0026          DO 70 I=1,N          OPT00380
0027          70 X(I)=X2(I)+DELX(I)          OPT00390
0028          CALL EVALU          OPT00400
C 1 MEANS SATIS.OF CONSTRAINT NT.PPEV. 2MEANS NOCHANGE 3MEANS VIOLATION          OPT00410
C IF POINT IS NOT FEASIBLE GIVE IT AN ARBITRARILY HIGH VALUE          OPT00420
0029          GO TO (540,90,80), NSATIS          OPT00430
0030          80 PX2=10.E35          OPT00440
0031          PU=10.E35          OPT00450
0032          GO TO 100          OPT00460
0033          90 CONTINUE          OPT00470
0034          PX2=PO          OPT00480
0035          IF (PX1-PX2) 100,100,150          OPT00490
0036          100 IF (N401-2) 130,110,110          OPT00500
0037          110 DO 120 I=1,N          OPT00510
0038          120 X1(I)=X(I)          OPT00520
0039          P1=PX2          OPT00530
0040          GO TO 430          OPT00540
C ONLY ONE POINT SO FAR COMPUTED          OPT00550
0041          130 DO 140 I=1,N          OPT00560
0042          140 X3(I)=X2(I)          OPT00570
0043          PREV3=PX1          OPT00580
0044          GO TO 180          OPT00590
0045          150 DO 160 I=1,N          OPT00600
0046          160 X3(I)=X2(I)          OPT00610
0047          X2(I)=X(I)          OPT00620
0048          160 DELX(I)=1.61803399*DELX(I)          OPT00630
0049          PREV3=PX1          OPT00640
0050          PX1=PX2          OPT00650
0051          GO TO 60          OPT00660
C GOLDEN SECTION SEARCH METHOD.          OPT00670
C B VECTOR GOES TO X1(I)          OPT00680
0052          170 PO=1.E36          OPT00690
0053          N404=N404+1          OPT00700
0054          180 DO 190 I=1,N          OPT00710
0055          190 X1(I)=X(I)          OPT00720
0056          P1=PO          OPT00730
0057          DO 200 I=1,N          OPT00740
0058          X1(I)=.38196601*(X1(I)-X3(I))+X3(I)          OPT00750
0059          200 X2(I)=X(I)          OPT00760
0060          CALL EVALU          OPT00770
0061          GO TO (540,270,210), NSATIS          OPT00780
0062          210 IF (N404.LT.30) GO TO 170          OPT00790
0063          CONTINUE          OPT00800
C THERE IS NO REFERENCE TO 211, THE ABOVE STATEMENT IS A DUMMY STATEMENT          OPT00810
C--IT IS POSSIBLE NO FEASIBLE POINT EXIST, IF NOT TRY MOVING ON DELX0.          OPT00820
C-- IF IT IS NOT POSSIBLE TO MOVE ON DELX0 THEN WE MUST BE AT A          OPT00830
C-- SOLUTION OF NLP PROBLEM.          OPT00840
0064          IF (N404.GT.100) GO TO 240          OPT00850
0065          220 DO 230 I=1,N          OPT00860
0066          IF (DABS(DABS(X3(I))/X1(I))-1.).GT.1.E-7) GO TO 170          OPT00870
0067          230 CONTINUE          OPT00880
0068          240 GO TO (250,260), N405          OPT00890

```

Figure 23.--Subroutine OPT.

```

0069      250  N405=2
          C---TRY TO MOVE ON GRADIENT.
0070      NTCTR=NTCTR-1
0071      M4=M4-1
0072      GO TO 20
0073      260  WRITE (6,5HJJ)
0074      CALL TIMEC
0075      CALL JUTPUT (1)
0076      CALL REJECT
0077      STOP 22042

          C
0078      270  CONTINUE
0079      N404=0
0080      PX1=PX
0081      DO 280 I=1,N
0082      280  X(I)=0.38196601*(X1(I)-X2(I))+X2(I)
0083      CALL EVALJ
0084      GO TO (540,290,220), NSATIS
0085      290  PX2=PX
0086      N401=1
0087      300  N401=N401+1
0088      IF (N401-25) 340,310,310
0089      310  K3=2
0090      IF (N401-40) 320,460,460
0091      320  DO 330 I=1,N
0092      IF (DABS(X2(I)/X1(I)-1.0).GE.1.E-7) GO TO 340
0093      330  CONTINUE
0094      GO TO 460
0095      340  IF (DABS(PX1/PX2-1.).LE.1.E-7) GO TO 460
0096      IF (PX1-PX2) 350,460,400
0097      C FROM LEFTORRIGHT X3(I)=(PREV3)X2(I)+(PX1)X1(I)PX2 X1(I)P1
0098      350  DO 360 I=1,N
0099      360  X1(I)=X(I)
          C THRU AWAY RIGHT PART
0099      P1=PX2
0100      DO 370 I=1,N
          C POINTXP1 BECOMES XP2
0101      370  X(I)=.38196601*(X1(I)-X3(I))+X3(I)
          C TEMPORARILY IN X STORAGE
0102      CALL EVALJ
0103      GO TO (540,380,170), NSATIS
0104      380  CONTINUE
0105      PX2=PX1
          C SWITCH VECTORS TO PROPER POSITION
0106      PX1=PX
0107      DO 390 I=1,N
0108      XX=X2(I)
0109      X2(I)=X1(I)
0110      390  X1(I)=XX
0111      GO TO 300
          C LEFT SIDE TOSSED AWAY
          C--- CHANGES FOR NONUNIMODAL FN
          C--- GO TO THRU AWAY RIGHT IN THIS CASE INIT VAL LT FIB PT
0112      400  IF (PREV3-PX2) 350,350,410
0113      410  DO 420 I=1,N
0114      X3(I)=X2(I)
0115      420  X2(I)=X(I)
0116      PREV3=PX1
0117      PX1=PX2
0118      430  DO 440 I=1,N
0119      440  X(I)=0.38196601*(X1(I)-X2(I))+X2(I)
0120      CALL EVALU
0121      GO TO (540,450,170), NSATIS
0122      450  CONTINUE
0123      PX2=PX
0124      GO TO 300
          C THE INTERIOR POINTS NOW GIVE EQUAL VALUE FOR P. COMPUTE MIDPOINT.
0125      460  DO 470 I=1,N
0126      DELXO(I)=X(I)
0127      X(I)=(DELXO(I)+X2(I))*0.5
0128      470  CONTINUE
0129      CALL EVALU
0130      GO TO (480,490), KSW
0131      480  IF (DABS(P0/PX1-1.).GT.1.E-7) GO TO 520
0132      490  GO TO (500,510), ISW
0133      500  IF (P0.LT.P31) GO TO 510
0134      510  ISW=2
          C IF P-FUNCTION DIDN'T GO DOWN TRY NEG VECT.
0135      GO TO 20
0136      510  RETURN
0137      520  DO 530 I=1,N
0138      530  X(I)=DELXO(I)
0139      GO TO 350
          C ARE WE NOW IN FEASIBILITY PHASE
0140      540  DO 550 I=1,M
0141      IF (KJ(I)) 560,560,550
0142      550  CONTINUE
0143      NSATIS=4
0144      RETURN
          C--- PROBLEM HAS BECOME FEASIBLE
          C --- P - FUNCTION CHANGES IF A CONSTRAINT BECOMES FEASIBLE
0145      560  MN=0
0146      DO 570 I=1,M
0147      RJ(I)=XJ(I)
0148      RETURN
          C
0149      580  FORMAT ( 80H OPT CAN'T FIND A FEASIBLE POINT, THAT GIVES A LOWER
0150      IALUL OF THE P-FUNCTION. )
          ENU

```

Figure 23.--
continued

```

FORTRAN IV G LEVEL 21          OUTPUT          DATE = 80242          12/12/50

0001          SUBROUTINE OUTPUT (K)                                OUT00060
C                                                                OUT00070
C          OCTOBER 1970                                          OUT00080
C                                                                OUT00090
C OUTPUT PRINTS OUT INFORMATION ON THE RESULTS OF EACH ITERATION AND THE OUT00100
C SOLUTION ESTIMATES AND THE ESTIMATES OF THE LAGRANGE MULTIPLIERS OUT00110
C THIS ROUTINE WAS MODIFIED BY GHAMEI(1979) WHILE INCORPORATING OUT00120
C BOUND CALCULATION TECHNIQUE.                                  OUT00130
0002          IMPLICIT REAL*8(A-H,O-Z)                            OUT00140
0003          REAL*4 RHOIN,RATIO,EPSI,THETAO                      OUT00150
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 OUT00160
0005          COMMON /EQAL/ M, M1, MZ                             OUT00170
0006          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 OUT00180
0007          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO              OUT00190
0008          COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPSI,THETAO, OUT00200
          INSG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PRI,
          ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),
          3 PREV3,ADELX, NTCTR, NUMEN1, NPHASE, NSATIS
          COMMON/ABC/LY,LZ,PER(45)                                OUT00230
0009          COMMON/ABG1/FE1,FE2                                OUT00240
0010          COMMON/ABG4/XY(45)                                  OUT00250
0011          COMMON/ABG2/DF1(45),DF2(45)                        OUT00260
0012          COMMON/OP6/NEXOP6                                   OUT00270
0013          COMMON/ABG5/TX(45),TOELX(45)                       OUT00280
0014          COMMON/ABG6/XX(45)                                  OUT00290
0015          NZ=M+MZ                                             OUT00300
0016          GO TO (10,20), K                                    OUT00310
0017          10 WRITE (6,60)                                     OUT00320
0018          WRITE (6,70) NTCTR,DOTT,RHO,ADELX,NPHASE           OUT00330
0019          20 WRITE (6,80) F,PO,G,RSIGMA,M                    OUT00340
0020          IF(LY .GT. 0.) GO TO 200                            OUT00350
0021          DO 210 I=1,N                                         OUT00360
0022          XY(I)=X(I)                                           OUT00370
0023          210 CONTINUE                                         OUT00380
0024          FE1=F                                                 OUT00390
0025          GO TO 220                                             OUT00400
0026          200 FE2=F                                             OUT00410
0027          220 WRITE (6,90) (X(I),I=1,N)                        OUT00420
0028          C CALCULATION OF GP VARIABLES                        OUT00430
          IF(NEXOP6.NE.1) GO TO 315                                OUT00440
          INOEX=0.                                                 OUT00450
          CALL TRANS(INDEX)                                         OUT00460
          WRITE(6,300)(X(I),I=1,N)                                  OUT00470
          DO 410 I=1,N                                              OUT00480
          410 X(I)=TX(I)                                           OUT00490
          315 WRITE (6,110)                                         OUT00500
          GO TO (30,40), NT2                                        OUT00510
          30 WRITE (6,120)                                          OUT00520
          WRITE (6,100) (RJ(I),I=1,NZ)                              OUT00530
          GO TO 50                                                  OUT00540
          40 WRITE (6,100) (RJ(I),I=1,NZ)                            OUT00550
          300 FORMAT(/,6X,'CORRESPONDING VALUE OF GP VARIABLE IS',/(6E20.7)) OUT00560
          50 RETURN                                                OUT00570
          C                                                                OUT00580
          60 FORMAT (50H0***** )                                  OUT00590
          70 FORMAT (10X,6HPDINT=14,6X,6H DOTT=E15.7,6X,6HRHO=E15.7,6X,10HMAGNI OUT00600
          TUDE=E15.7,6X,6HMPHASE=12)                                OUT00610
          80 FORMAT (8X,2HF=E15.7,5X,2HP=E15.7,5X,2HG=E15.7,5X,7HRSIGMA=E15.7,5X OUT00620
          1X,2HM=E15.7)                                             OUT00630
          90 FORMAT (6X,25HTHE CURRENT VALUE OF X IS/(6E20.7))    OUT00640
          100 FORMAT (6E20.7)                                       OUT00650
          110 FORMAT (6X,21HTHE CONSTRAINT VALUES)                OUT00660
          120 FORMAT (28X,34HNOT INCLUDING THE NON-NEGATIVITIES)  OUT00670
          END                                                         OUT00680
0043          60 FORMAT (50H0***** )                                OUT00690
0044          70 FORMAT (10X,6HPDINT=14,6X,6H DOTT=E15.7,6X,6HRHO=E15.7,6X,10HMAGNI OUT00610
          TUDE=E15.7,6X,6HMPHASE=12)                                OUT00620
0045          80 FORMAT (8X,2HF=E15.7,5X,2HP=E15.7,5X,2HG=E15.7,5X,7HRSIGMA=E15.7,5X OUT00630
          1X,2HM=E15.7)                                             OUT00640
0046          90 FORMAT (6X,25HTHE CURRENT VALUE OF X IS/(6E20.7))    OUT00650
0047          100 FORMAT (6E20.7)                                       OUT00660
0048          110 FORMAT (6X,21HTHE CONSTRAINT VALUES)                OUT00670
0049          120 FORMAT (28X,34HNOT INCLUDING THE NON-NEGATIVITIES)  OUT00680
0050          END                                                         OUT00690

```

Figure 24.--Subroutine OUTPUT.

```

FORTRAN IV G LEVEL 21          PARDIF          DATE = 80242          12/09/35
0001          SUBROUTINE PARDIF          PAR00340
C          C          PAR00050
C          C          15 MARCH 1972          PAR00060
C          C          PAR00070
C THIS SUBROUTINE RETURNS A DIFFERENCING INTERVAL FOR DPAR(20) WHEN          PAR00080
C CALLED BY SUBROUTINE SENS. ISENS CONTROLS THE VALUE OF THE INTERVAL          PAR00090
C WITH A MINIMUM VALUE IN THE NEIGHBORHOOD OF 1.E-15 DEPENDING ON XEPI.          PAR00100
C ISENS DEPENDS ON THE KIND OF SENSITIVITY ANALYSIS BEING CONDUCTED.          PAR00110
C THE SUBROUTINE WAS CODED BY R. L. ARMACOST.          PAR00120
C          PAR00130
0002          IMPLICIT REAL*8(A-H,O-Z)          PAR00140
0003          REAL*4 XEPI          PAR00150
0004          COMMON/EXOPT/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEPI,XEPI2          PAR00160
0005          COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS          PAR00170
0006          MULT = MIN0(ISENS,8)          PAR00180
0007          XEPI = XEPI / (10.**MULT)          PAR00190
0008          IF(XEPI.LE.0.) XEPI = 1.E-10          PAR00200
0009          DO 10 I=1,NPAR          PAR00210
0010          10 DPAR(I)=XEPI          PAR00220
0011          RETURN          PAR00230
0012          END          PAR00240

```

Figure 25.--Subroutine PARDIF.

```

FORTRAN IV G LEVEL 21          PERT          DATE = 80242          12/13/33
C SUBROUTINE PERT IS CODED BY GHAENI(1979) TO PERFORM THE PARAMETER          PER00060
C ADJUSTMENTS REQUIRED AT VARIOUS STAGES OF BOUND CALCULATIONS.          PER00070
C          SUBROUTINE PERT          PER00080
0001          IMPLICIT REAL*8(A-H,O-Z)          PER00090
0002          COMMON/ABG/LY,LZ,PER(45)          PER00100
0003          COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS          PER00110
0004          20 LZ=LZ+1          PER00120
0005          IF(LZ.GT. NPAR) GO TO 30          PER00130
0006          IF(PER(LZ).NE. 0.) GO TO 10          PER00140
0007          GO TO 20          PER00150
0008          10 PAR(LZ)=PAR(LZ)+PER(LZ)          PER00160
0009          30 RETURN          PER00170
0010          END          PER00180
0011

```

Figure 26.--Subroutine PERT.

6.19 PEVALU

Subroutine PEVALU is used to compute the value of the w function and the G function. It makes use of the values of the objective function and the constraint function, which must have been computed before PEVALU is called. The G function is interpreted as the value of the dual objective function when the W function has been minimized for a given value of r ($RH\emptyset$). PEVALU is given as Figure 27.

```

FORTRAN IV G LEVEL 21          PEVALU          DATE = 80242          12/11/79

0001          SUBROUTINE PEVALU                                PEV00040
C                                                              PEV00050
C          OCTOBER 1970                                        PEV00060
C                                                              PEV00070
C PEVALU COMPUTES THE VALUE OF THE PENALTY FUNCTION AND THE VALUE OF THE PEV00080
C DUAL USING PREVIOUSLY COMPUTED VALUES FOR F, AND RJ.      PEV00090
0002          IMPLICIT REAL*8(A-M,D-Z)                        PEV00100
0003          REAL*4 RHOIN,RATIO,EPSI,THETAO                  PEV00110
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MM,NP1,NM1 PEV00120
0005          COMMON /EQAL/ H, M1, M2                          PEV00130
0006          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 PEV00140
0007          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO           PEV00150
0008          COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPSI,THETAO, PEV00160
          1RSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XRI(45),PRI, PEV00170
          2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45), PEV00180
          3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS PEV00190
0009          M=0.0                                           PEV00200
0010          RSIGMA=0.0                                       PEV00210
C NONVEGS IF INCLUDED ARE ADDED TO P--ARE POS. IN ALL PHASES PEV00220
0011          GO TO (10,30), NT2                                PEV00230
0012          10 DO 20 I=1,N                                    PEV00240
0013          20 RSIGMA=RSIGMA-RHO*DLOG(X(I))                  PEV00250
0014          30 GO TO (40,50,150), NPHASE                     PEV00260
C OBJECTIVE FUNCTION - SIGMA VIOL. CONSTS.                    PEV00270
0015          40 F=0.0                                         PEV00280
0016          50 IF (M) 100,100,60                             PEV00290
0017          60 DO 90 J=1,M                                    PEV00300
0018          IF (RJ(J)) 80,80,70                              PEV00310
0019          70 RSIGMA=RSIGMA-RHO*DLOG(RJ(J))                 PEV00320
0020          GO TO 90                                          PEV00330
0021          80 F=F-RJ(J)                                       PEV00340
0022          90 CONTINUE                                       PEV00350
C EQUALITIES NOT ADDED IN FEAS. PHASE                          PEV00360
0023          100 CONTINUE                                       PEV00370
0024          IF (M2) 140,140,110                               PEV00380
0025          110 GO TO (140,120,150), NPHASE                  PEV00390
0026          120 DO 130 I=1,M2                                  PEV00400
0027          K=M+I                                             PEV00410
0028          130 M=M+RJ(K)**2                                   PEV00420
0029          M=M/RHO                                           PEV00430
0030          140 MS=M+RSIGMA                                    PEV00440
0031          PO=F+MS                                            PEV00450
0032          HMS=2.*M-RHO*DFLOAT(M)                            PEV00460
0033          G=F+HMS                                            PEV00470
0034          IF (NT2.EQ.1) G=G-RHO*DFLOAT(M)                  PEV00480
0035          150 RETURN                                         PEV00490
0036          END                                               PEV00500

```

Figure 27.--Subroutine PEVALU.

6.20 PRESEN

This subroutine calculates the gradient of the optimal value function, using the gradient of the Lagrangian when taken with respect to the parameters (Step 9 of Algorithm 2.1). In addition, PRESEN (when invoked) performs a preliminary screening of the problem parameters to which the optimal value function is practically insensitive. The criterion used for this purpose is that if the change in optimal value function is less than 0.1 percent of its current value as a result of the introduced increment of a given parameter, then that parameter is

eliminated from further consideration. This parameter screening is invoked when the fifth variable in the second option card takes a value of zero, as shown in Table 4. Figure 28 lists PRESEN.

6.21 PUNCH

When a call to PUNCH is made, the current value of the array x is punched, along with some of the control cards that can be used to restart SENSUMT. Subroutine PUNCH is shown in Figure 29.

6.22 REJECT

Subroutine REJECT puts values of the point at x and the associated values of the objective function, constraint functions, and W function back into their normal locations. These values were temporarily stored in other locations by subroutine STORE. STORE appears as Figure 30.

6.23 RHOCOM

This subroutine is used to compute the initial value of r (RH0) for each NLP problem. As previously stated, the search for a feasible starting point can result in a sequence of NLP problems. The initial value of r is computed by the formula specified for the use of option 1 (NT1). RHOCOM is given as Figure 31.

6.24 SECONDD

SECONDD is used to query about the computer's clock. This is made possible by an internal clock which is called by this subroutine. The elapsed time obtained and monitored by this subroutine is used by the subroutines TIMEC, TCHECK, and SET. Figure 32 is subroutine SECONDD.

```

0001      SUBROUTINE PRESEN(DU,KTEST)
C
C          1 MARCH 1976
C
C SUBROUTINE PRESEN IS USED TO CALCULATE THE GRADIENT OF THE OPTIMAL
C VALUE FUNCTION USING THE GRADIENT OF THE LAGRANGIAN TAKEN WITH
C RESPECT TO THE PARAMETERS. PRESEN IS CALLED BY SENS WHEN VARIABLE
C NEXOPS = 0 OR =1. ADDITIONALLY, WHEN NEXOPS = 0, THE PARAMETERS WHICH
C AFFECT THE OPTIMAL VALUE FUNCTION BY LESS THAN 0.001 TIMES ITS
C CURRENT VALUE ARE ELIMINATED FROM FURTHER CONSIDERATION. THIS
C SUBROUTINE WAS CODED BY R. L. ARMACOST.
0002      IMPLICIT REAL*8(A-M,D-Z)
0003      COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,MN1
0004      COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO
0005      COMMON/EQUAL/ H, M1,MZ
0006      COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS
0007      COMMON/ABG/LY,LZ,PER(45)
0008      COMMON/ABG1/FE1,FE2
0009      COMMON/ABG2/DF1(45),DF2(45)
0010      DIMENSION GX(90),DU(90),KTEST(45),KLIST(45)
0011      MPHZ = M + MZ
0012      FTST = 0.001 * DABS(F)
0013      DO 100 J=1,NPAR
0014      KTEST(J) = 0
0015      PAR(J) = PAR(J) + DPAR(J)
0016      CALL RESTNT(0,DF)
0017      IF(MPHZ.EQ.0) GO TO 20
0018      DO 10 I=1,MPHZ
0019      CALL RESTNT(I,DU(I))
0020      CONTINUE
0021      10  DEM = 2. * DPAR(J)
0022      PAR(J) = PAR(J) - DEM
0023      CALL RESTNT(0,XF)
0024      IF(MPHZ.EQ.0) GO TO 40
0025      DO 30 I=1,MPHZ
0026      CALL RESTNT(I,GX(I))
0027      CONTINUE
0028      30  DFEPS = (DF - XF)/DEM
0029      IF(MPHZ.EQ.0) GO TO 60
0030      DO 50 I=1,MPHZ
0031      DU(I) = (DU(I) - GX(I))/DEM
0032      50  SUM = DFEPS
0033      IF(M.EQ.0) GO TO 80
0034      DO 70 I=1,M
0035      SUM = SUM - RHO/RJ(I)*DU(I)
0036      70  IF(MZ.EQ.0) GO TO 95
0037      TSUM = 0.
0038      DO 90 I=1,MZ
0039      IM = I+M
0040      TSUM = TSUM + RJ(IM)*DU(IM)
0041      90  SUM = SUM + TSUM * 2./RHO
0042      95  DEL(J) = SUM
0043      PAR(J) = PAR(J) + DPAR(J)
0044      DTEST = DABS(DEL(J))
0045      IF(DTEST.GE.FTST) KTEST(J) = 1
0046      CONTINUE
0047      100 WRITE(6,600)
0048      600 FORMAT(22X,34HOPTIMAL VALUE FUNCTION SENSITIVITY   //)
0049      DO 200 I=1,NPAR,5
0050      II=MINO(I+4,NPAR)
0051      WRITE(6,601) ((JJ,DEL(JJ)), JJ=I,II)
0052      601 FORMAT( 5(7H DF/DA(,I2,2H)=,G14.7))
0053      CONTINUE
0054      IF(LY .GT. 0.) GO TO 500
0055      DO 400 J=1,NPAR
0056      DF1(J)=DEL(J)
0057      400 CONTINUE
0058      GO TO 700
0059      500 DO 610 J=1,NPAR
0060      DF2(J)=DEL(J)
0061      610 CONTINUE
0062      IF(LY .EQ. 0.) GO TO 700
0063      PAR(LZ)=PAR(LZ)-PER(LZ)
0064      RETURN
0065      700 JJ = 0
0066      DO 250 J=1,NPAR
0067      IF(KTEST(J).EQ.0) GO TO 250
0068      JJ = JJ + 1
0069      KLIST(JJ) = J
0070      CONTINUE
0071      250 IF(JJ.EQ.0) GO TO 300
0072      WRITE(6,602)
0073      602 FORMAT( /31M DETAILED SENSITIVITY RESULTS FOLLOW FOR PARAMETERS )
0074      WRITE(6,603) (KLIST(I), I=1,JJ)
0075      603 FORMAT(1M 40(I2,2H ,))
0076      WRITE(6,604)
0077      604 FORMAT(/)
0078      RETURN
0079      300 WRITE(6,605)
0080      605 FORMAT( 42M THERE ARE NO DETAILED SENSITIVITY RESULTS   //)
0081      RETURN
0082      END

```

Figure 28.--Subroutine PRESEN.

```

FORTRAN IV G LEVEL 21          PUNCH          DATE = 80242          12/12/71

0001          SUBROUTINE PUNCH          PUN00040
          C          OCTOBER 1970          PUN00050
          C          PUN00060
          C          PUN00070
          C THIS SUBROUTINE PUNCHES THE STOPPING POINTS AND ASSOCIATED PARAMETERS PUN00080
          C SO THAT ANOTHER RUN MAY BE MADE STARTING WHERE THE CURRENT ONE PUN00090
          C STOPPED PUN00100
          C THIS ROUTINE IS CALLED IF NT6=2. PUN00110
0002          IMPLICIT REAL*(A-H,O-Z) PUN00120
0003          REAL*4 RHOIN,RATIO,EPSI,THETA0,XEP1,XEP2,T PUN00130
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 PUN00140
0005          COMMON /EQUAL/ H, HI, HZ PUN00150
0006          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 PUN00160
0007          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO PUN00170
0008          COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0, PUN00180
          IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PRI, PUN00190
          ZPR2,P1,F1,RJ1(90),DOTT,PGRADE(45),DIAG(45), PUN00200
          3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS PUN00210
0009          COMMON/EXPOPT/NEXP01,NEXP02,NEXP03,NEXP04,NEXP05,XEP1,XEP2 PUN00220
0010          T=60.0 PUN00230
0011          WRITE (7,10) EPSI,RHO,THETA0,RATIO,T,M,N,M, PUN00240
          C TMAX.IS SET TO 60. SECONDS PUN00250
0012          WRITE (7,20) (X(I),I=1,N) PUN00260
0013          NT1=3 PUN00270
          C SET RHO OPTION SO THIS VALUE OF RHO WILL BE USE FOR THE RESTART. PUN00280
0014          WRITE (7,30) NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 PUN00290
0015          WRITE (7,20) XEP1, XEP2 PUN00300
0016          WRITE (7,30) NEXP01,NEXP02,NEXP03 PUN00310
0017          RETURN PUN00320
          C PUN00330
0018          10 FORMAT (5E12.5,3I4) PUN00340
0019          20 FORMAT (6E12.5) PUN00350
0020          30 FORMAT (10I7) PUN00360
0021          END PUN00370

```

Figure 29.--Subroutine PUNCH.

```

FORTRAN IV G LEVEL 21          REJECT          DATE = 80242          12/13/73

J001          SUBROUTINE REJECT          REJ00040
          C          OCTOBER 1970          REJ00050
          C          REJ00060
          C          REJ00070
          C REJECT RETURNS THE STORED VALUES OF THE OBJECTIVE FUNCTION, THE REJ00080
          C CONSTRAINT FUNCTIONS AND THE PENALTY FUNCTION TO THEIR NORMAL LOCATION REJ00090
0002          IMPLICIT REAL*(A-H,O-Z) REJ00100
0003          REAL*4 RHOIN,RATIO,EPSI,THETA0 REJ00110
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 REJ00120
0005          COMMON /EQUAL/ H, HI, HZ REJ00130
0006          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO REJ00140
0007          COMMON/CRST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0, REJ00150
          IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PRI, REJ00160
          ZPR2,P1,F1,RJ1(90),DOTT,PGRADE(45),DIAG(45), REJ00170
          3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS REJ00180
0008          DO 10 I=1,N REJ00190
0009          X(I)=X1(I) REJ00200
0010          MMZ=M+NZ REJ00210
0011          DO 20 J=1,MMZ REJ00220
0012          RJ(J)=RJ1(J) REJ00230
0013          PO=P1 REJ00240
0014          RSIGMA=RSIG1 REJ00250
0015          G=G1 REJ00260
0016          F=F1 REJ00270
0017          H=H1 REJ00280
0018          RETURN REJ00290
0019          END REJ00300

```

Figure 30.--Subroutine REJECT.

```

FURTMAN IV G LEVEL 21          RHUCOM          DATE = 80262          12/39/21

0001          SUBROUTINE RHOCOM          RH000040
C          OCTOBER 1970          RH000050
C          RH000060
C          SUBROUTINE TO COMPUTE INITIAL RHO VALUE          RH000070
C          CONTROLLED BY CUL. 7 ON OPTION CARD          RH000080
0002          IMPLICIT REAL*8(A-H,O-Z)          RH000090
0003          REAL*4 RHOIN,RATIO,EPSI,THETA0          RH000100
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1          RH000110
0005          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10          RH000120
0006          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO          RH000130
0007          COMMON/CAST/DELX(45),DELX0(45),RHOIN,RATIO,EPSI,THETA0,          RH000140
          IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,          RH000150
          2PR2,P1,F1,RJ1(90),DOT1,PGRAD(45),DIAG(45),          RH000160
          3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS          RH000170
          GO TO (110,50,10,190), NT1          RH000180
0008          RHO=RHOIN          RH000190
0009          10          RHO=1.          RH000200
0010          20          IF (RHO) 30,30,40          RH000210
0011          30          RHO=1.          RH000220
0012          40          RETURN          RH000230
0013          50          NPAR1=1          RH000240
0014          60          RHO=1.          RH000250
C          2 MEANS RHO WHICH MINIMIZES GRADIENT MAG.          RH000260
0015          CALL GRAD (2)          RH000270
0016          DO TO I=1,N          RH000280
0017          70          PGRAD(I)=DELX0(I)          RH000290
0018          RHO=2.          RH000300
0019          CALL GRAD (2)          RH000310
0020          DO 80 I=1,N          RH000320
0021          DELX0(I)=DELX0(I)-PGRAD(I)          RH000330
0022          80          PGRAD(I)=PGRAD(I)-DELX0(I)          RH000340
0023          GO TO (90,130), NPAR1          RH000350
0024          90          DOT1=0.          RH000360
0025          DOT2=0.          RH000370
0026          DO 100 I=1,N          RH000380
0027          DOT1=DOT1+DELX0(I)*PGRAD(I)          RH000390
0028          100          DOT2=DOT2+DELX0(I)**2          RH000400
0029          RHO=DABS(DOT1/DOT2)          RH000410
0030          GO TO 20          RH000420
C          3 MEANS COMPUTE RHO SO AS TO MINIMIZE DEL P1/DDP/1.)DEL P          RH000430
0031          110          NPAR2=1          RH000440
0032          120          NPAR1=2          RH000450
C          USE DF AND DR SUBROUTINE          RH000460
0033          GO TO 60          RH000470
0034          130          RHO=1.          RH000480
C          ASSUME SIGMA TERM IS CONSID. GRTER THAN F TERM          RH000490
0035          CALL SECORD (2)          RH000500
0036          DO 140 I=1,N          RH000510
0037          140          DELX(I)=PGRAD(I)          RH000520
0038          CALL INVERS (1)          RH000530
0039          DO 150 I=1,N          RH000540
0040          X1(I)=DELX(I)          RH000550
0041          150          DELX(I)=DELX0(I)          RH000560
0042          CALL SECORD (2)          RH000570
0043          CALL INVERS (1)          RH000580
0044          DO 160 I=1,N          RH000590
0045          160          XR2(I)=DELX(I)          RH000600
0046          170          DOT1=0.          RH000610
0047          DOT2=0.          RH000620
0048          DO 180 I=1,N          RH000630
0049          DOT1=DOT1+PGRAD(I)*X1(I)          RH000640
0050          180          DOT2=DOT2+DELX0(I)*XR2(I)          RH000650
0051          RHO=DSORT(DABS(DOT1/DOT2))          RH000660
0052          GO TO 20          RH000670
0053          190          NPAR2=2          RH000680
0054          C          RHO MINIMIZES 2ND ORDER MOVE          RH000690
          GO TO 120          RH000700
0055          C          USES INTERNAL SUB. TO COM /DDP/-1 DF AND /DDP/- DR          RH000710
          DOT1=0.0          RH000720
          DOT2=0.0          RH000730
          DO 210 I=1,N          RH000740
          210          DOT1=X1(I)**2+DOT1          RH000750
          DOT2=X1(I)*XR2(I)+DOT2          RH000760
          RHO=DABS(DOT1/DOT2)          RH000770
          GO TO 20          RH000780
0062          END          RH000790
0063          END          RH000800

```

Figure 31.--Subroutine RHOCOM.

```

FORTRAN IV G LEVEL 21          SECUND          DATE = 80242          12/14/02
0001          SUBROUTINE SECUND(SECS)          SEC00050
0002          CALL CLOCK(1)          SEC00060
0003          SECS = 1          SEC00070
0004          SECS = SECS / 100.0          SEC00080
0005          RETURN          SEC00090
0006          END          SEC00100

```

Figure 32.--Subroutine SECUND.

6.25 SECORD

Subroutine SECORD evaluates the matrix of second partials of the W function. It calls on the user-supplied subroutine MATRIX to evaluate the upper triangle of the matrix of the second partials of the objective and the constraint functions. Subroutine SECORD is listed in Figure 33.

6.26 SENS

This subroutine calculates the solution point and optimal value function sensitivities for each subproblem (if required), as well as the final problem, by implementing Steps 1-3 and 7-8 of Algorithm 2.1 of Section 2. The various choices for sensitivity calculations are given in options 3, 4, and 5 of the second option card (see Table 4). SENS is called by the program MAIN and the subroutine BODY. It calls subroutine LMULT to calculate the Lagrange multiplier sensitivities. SENS is shown as Figure 34.

6.27 SET

Subroutine SET is called once by MAIN at the start of the attempt to solve a problem. It obtains and stores the value of the computer's clock. It obtains the value of the clock by calling a system library or user coded subroutine that queries the computer's clock and returns its value as a floating point number in seconds. Figure 35 lists subroutine SET.

```

FORTRAN IV G LEVEL 21          SECOND          DATE = 80242          12/46/12

0001          SUBROUTINE SECORD (IS)          SEC00040
C          C          SEC00050
C          OCTOBER 1970          SEC00060
C          C          SEC00070
C SECORD EVALUATES THE MATRIX OF SECOND PARTIALS OF THE PENALTY          SEC00080
C FUNCTION.          SEC00090
C          C          SEC00100
C- (1) MEANS DONT COMPUTE GRAD. OUTER PRODUCT(IN SECORD)          SEC00110
0002          IMPLICIT REAL*8(A-M,D-Z)          SEC00120
0003          REAL*4 RHOIN,RATIO,EPSI,THETA0          SEC00130
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1          SEC00140
0005          COMMON /EQAL/ H, H1, MZ          SEC00150
0006          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NY5,NT6,NT7,NT8,NT9,NT10          SEC00160
0007          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO          SEC00170
0008          COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPSI,THETA0,          SEC00180
          IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PRI,          SEC00190
          2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),          SEC00200
          3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS          SEC00210
          DO 10 I=1,N          SFC00220
          DO 10 J=1,N          SEC00230
0011          10          A(I,J)=0.          SEC00240
0012          GO TO (230,20), IS          SEC00250
CGRAD. TERM NOT PREV. COMPUTED          SEC00260
0013          20          DO 30 I=1,N          SEC00270
0014          DO 30 J=1,I          SEC00280
0015          A(I,J)=0.0          SEC00290
0016          30          CONTINUE          SEC00300
0017          GO TO (40,60), NT2          SEC00310
0018          40          DO 50 I=1,N          SEC00320
0019          50          A(I,J)=RHO/X(I)**2          SEC00330
0020          60          CONTINUE          SEC00340
0021          IF (M.LE.0) GO TO 130          SEC00350
0022          DO 120 IN=1,M          SEC00360
0023          IF (RJ(IN)) 120,120,70          SEC00370
0024          70          CALL GRAD1 (IN)          SEC00380
0025          TT = RHO/RJ(IN)**2          SEC00390
0026          DO 110 I=1,N          SEC00400
0027          IF (DEL(I)) 80,110,80          SEC00410
0028          T=TT*DEL(I)          SEC00420
0029          80          DO 100 J=1,I          SEC00430
0030          IF (DEL(J)) 90,100,90          SEC00440
0031          A(I,J)=A(I,J)+T*DEL(J)          SEC00450
0032          100          CONTINUE          SEC00460
0033          110          CONTINUE          SEC00470
0034          120          CONTINUE          SEC00480
C EQUALITY CONSTRAINTS          SEC00490
0035          130          IF (MZ) 210,210,140          SEC00500
0036          140          GO TO (210,150,230), NPHASE          SEC00510
0037          150          RQ=2./RHO          SEC00520
0038          DO 200 JJ=1,MZ          SEC00530
0039          IN=M+JJ          SEC00540
0040          CALL GRAD1 (IN)          SEC00550
0041          DO 190 I=1,N          SEC00560
0042          IF (DEL(I)) 160,190,160          SEC00570
0043          160          T=RQ*DEL(I)          SEC00580
0044          DO 180 J=1,I          SEC00590
0045          IF (DEL(J)) 170,180,170          SEC00600
0046          A(I,J)=A(I,J)+T*DEL(J)          SEC00610
0047          180          CONTINUE          SEC00620
0048          190          CONTINUE          SEC00630
0049          200          CONTINUE          SEC00640
0050          210          DO 220 I=1,N          SEC00650
0051          DIAG(I)=A(I,I)          SEC00660
0052          220          A(I,I)=0.          SEC00670
C READY NOW FOR MATRIX OF 2ND PARTIALS OF RESTRAINTS          SEC00680
0053          230          GO TO (240,510,520), NT10          SEC00690
0054          240          IF (M.LE.0) GO TO 340          SEC00700
0055          DO 330 IN=1,M          SEC00710
0056          LORN=2          SEC00720
C CONSTRAINT ASSUMED NONLINEAR          SEC00730
0057          CALL MATRIX (IN,LORN)          SEC00740
0058          IF (LORN.LT.2) GO TO 330          SEC00750
0059          IF (RJ(IN) .GT. 0.0) GO TO 280          SEC00760
0060          DO 260 I=2,M          SEC00770
0061          IM1=I-1          SEC00780
0062          DO 260 J=1,IM1          SEC00790
0063          IF (A(J,I)) 250,260,250          SEC00800
0064          250          A(I,J)=A(I,J)-A(J,I)          SEC00810
0065          A(J,I)=0.          SEC00820
0066          260          CONTINUE          SEC00830
0067          DO 270 I=1,N          SFC00840
0068          DIAG(I)=DIAG(I)-A(I,I)          SEC00850
0069          270          A(I,I)=0.0          SEC00860
0070          GO TO 330          SEC00870
0071          280          T=-RHO/RJ(IN)          SEC00880
0072          DO 300 I=2,M          SEC00890
0073          IM1=I-1          SEC00900

```

Figure 33.--Subroutine SECORD.

```

0074          DO 300 J=1,IM1
0075          IF (A(J,I)) 290,300,290
0076          290 A(I,J)=A(I,J)+T*A(J,I)
0077          A(J,I)=0.
0078          300 CONTINUE
0079          DO 320 I=1,N
0080          IF (A(I,I)) 310,320,310
0081          310 DIAG(I)=DIAG(I)+T*A(I,I)
0082          A(I,I)=0.
0083          320 CONTINUE
0084          330 CONTINUE
0085          340 CONTINUE
0086          GO TO (520,350,520), NPHASE
0087          350 IF (MZ.EQ.0) GO TO 420
C--          EQUALITY SECOND PARTIALS HERE
          IF (INTIO.GE.2) GO TO 420
0088          DO 410 I=1,MZ
0089          IN=M+I
0090          LORN=2
0091          CALL MATRIX (IN,LORN)
0092          IF (LORN.LT.2) GO TO 410
0093          T=2.*RJ(IN)/RMD
0094          DO 380 I=2,M
0095          IM1=I-1
0096          DO 370 J=1,IM1
0097          IF (A(J,I)) 360,370,360
0098          360 A(I,J)=A(I,J)+T*A(J,I)
0099          A(J,I)=0.0
0100          370 CONTINUE
0101          380 CONTINUE
0102          DO 400 I=1,N
0103          IF (A(I,I)) 390,400,390
0104          390 DIAG(I)=DIAG(I)+T*A(I,I)
0105          A(I,I)=0.0
0106          400 CONTINUE
0107          410 CONTINUE
0108          C GET MATRIX OF 2ND PARTIALS OF OBJECTIVE FUNCTION
0109          420 LLL=2
0110          CALL MATRIX (0,LLL)
0111          IF (LLL.LT.2) GO TO 490
0112          DO 440 I=2,N
0113          IM1=I-1
0114          DO 440 J=1,IM1
0115          IF (A(J,I)) 430,440,430
0116          430 A(I,J)=A(I,J)+A(J,I)
0117          440 A(J,I)=A(I,J)
0118          DO 470 I=1,N
0119          IF (A(I,I)) 450,460,450
0120          450 A(I,I)=DIAG(I)+A(I,I)
0121          GO TO 470
0122          460 A(I,I)=DIAG(I)
0123          470 CONTINUE
0124          480 RETURN
0125          490 DO 500 I=1,N
0126          A(I,I)=DIAG(I)
0127          DO 500 J=1,N
0128          A(I,J)=A(J,I)
0129          GO TO 480
0130          510 GO TO (520,350,350), NPHASE
0131          520 DO 530 I=2,N
0132          IM1=I-1
0133          DO 530 J=1,IM1
0134          A(J,I)=A(I,J)
0135          DO 540 I=1,N
0136          540 A(I,I)=DIAG(I)
0137          GO TO 480
0138          END
SEC00910
SEC00920
SEC00930
SEC00940
SEC00950
SEC00960
SEC00970
SEC00980
SEC00990
SEC01000
SEC01010
SEC01020
SEC01030
SEC01040
SEC01050
SEC01060
SEC01070
SEC01080
SEC01090
SEC01100
SEC01110
SEC01120
SEC01130
SEC01140
SEC01150
SEC01160
SEC01170
SEC01180
SEC01190
SEC01200
SEC01210
SEC01220
SEC01230
SEC01240
SEC01250
SEC01260
SEC01270
SEC01280
SEC01290
SEC01300
SEC01310
SEC01320
SEC01330
SEC01340
SEC01350
SEC01360
SEC01370
SEC01380
SEC01390
SEC01400
SEC01410
SEC01420
SEC01430
SEC01440
SEC01450
SEC01460
SEC01470
SEC01480
SEC01490
SEC01500
SEC01510
SEC01520
SEC01530
SEC01540
SEC01550
SEC01560
SEC01570

```

Figure 33.--continued

FURTRAN IV G LEVEL 21

SENS

DATE = 80242

12/27/55

```

0001          SUBROUTINE SENS                                SEN00060
C                                                     SEN00070
C               1 MARCH 1976                                SEN00080
C                                                     SEN00090
C THIS VERSION OF THE SENSITIVITY ANALYSIS SUBROUTINE IS USED TO SEN00100
C COMPUTE THE DIRECTIONAL DERIVATIVES OF X AND F WITH RESPECT TO SEN00110
C CERTAIN PARAMETERS CODED IN THE ARRAY PAR(20). THE DIRECTIONAL SEN00120
C DERIVATIVES ARE ESTIMATED FOR ONE PARAMETER AT A TIME WITH NPAR SEN00130
C THE NUMBER OF PARAMETERS INVOLVED IN THE SENSITIVITY ANALYSIS. THE SEN00140
C USE OF THE PARAMETERS PAR(20) MUST BE CONSISTENT THROUGHOUT THE SEN00150
C USER'S SUBROUTINES.                                     SEN00160
C THE SUBROUTINE IS USED FOR A SENSITIVITY ANALYSIS AT THE FINAL SUB- SEN00170
C PROBLEM OR FOR A SENSITIVITY ANALYSIS AT EACH SUBPROBLEM ALONG THE SEN00180
C MINIMIZING TRAJECTORY. DPAR(20) IS THE ARRAY OF DIFFERENCING SEN00190
C INTERVALS CORRESPONDING TO THE PARAMETERS PAR(20). DPAR(20) IS SEN00200
C ASSIGNED VALUES IN SUBROUTINE PARDIF.                 SEN00210
C THIS APPROACH TO SENSITIVITY ANALYSIS IS DUE TO A. V. FIACCO. THE SEN00220
C FIRST VERSION WAS CODED BY B. CAUSEY. THE SECOND VERSION WAS CODED SEN00230
C BY M. C. MYLANDER. THE THIRD VERSION WAS AN EXTENSION OF THE SEN00240
C SECOND VERSION TO PERMIT SENSITIVITY ANALYSIS ALONG THE MINIMIZING SEN00250
C TRAJECTORY, AND WAS CODED BY R. L. ARMACOST. THIS IS THE FOURTH SEN00260
C VERSION WHICH INCORPORATES THE OPTION TO CALCULATE THE PARTIAL SEN00270
C DERIVATIVES OF THE LAGRANGE MULTIPLIERS (SUBROUTINE LMULT) AND TO SEN00280
C CONDUCT A PRELIMINARY SCREENING OF THE PARAMETERS BY ESTIMATING THE SEN00290
C FIRST ORDER SENSITIVITY OF THE OPTIMAL VALUE FUNCTION (SUBROUTINE SEN00300
C PRESEN). THIS VERSION WAS CODED BY R. L. ARMACOST.      SEN00310
C THIS ROUTINE TOGETHER WITH OTHER SENSITIVITY ROUTINE WERE SEN00320
C REDIMENSIONED TO 45 I.E. X(45) AND PAR(45). FURTHERMORE SEN00330
C IT WAS SLIGHTLY MODIFIED TO INTERFACE BOUND CALCULATION SEN00340
C ROUTINES. GHAEMI(1979).                                SEN00350

0002          IMPLICIT REAL*8(A-H,O-Z)                    SEN00360
0003          REAL*4 RHOIN,RATIO,EPSI,THETAO,XEP1,XEP2     SEN00370
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 SEN00380
0005          COMMON/EQUAL/H,M1,MZ                          SEN00390
0006          COMMON/OPTNS/NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 SEN00400
0007          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RHO       SEN00410
0008          COMMON/CRST/DELX(45),DELXO(45),RHOIN,RATIO,EPSI,THETAO, SEN00420
          INTCTR,NUMINI,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1, SEN00430
          ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),     SEN00440
          3PREV3,ADELX,RSIG1,G1,NPHASE,NSATIS            SEN00450
          COMMON/SEN/PAR(45),DPAR(45),NPAR,ISENS        SEN00460
          COMMON/EXPOPT/NEXOP1,NEXOP2,NEXOP3,NEXOP4,NEXOP5,XEP1,XEP2 SEN00470
          COMMON/ABG/LY,LZ,PER(45)                       SEN00480
          COMMON/ABG1/FE1,FE2                             SEN00490
          COMMON/ABG2/DF1(45),DF2(45)                    SEN00500
          COMMON/ABG5/TX(45),TDELX(45)                   SEN00510
          COMMON/OP6/NEXOP6                               SEN00520
          DIMENSION DELT(45),DELTX(45),X2M(45),X3M(45)  SEN00530
          DIMENSION DELMU(90),DUM(90),KTEST(45)         SEN00540
          DO 5 I=1,N                                      SEN00550
          DELTX(I) = DELX(I)                              SEN00560
          5 DELT(I) = DELXO(I)                            SEN00570
          CALL STORE                                       SEN00580
          AVAL = 1.0E-10                                   SEN00590
          MPHZ = M + MZ                                    SEN00600
          ISENS = ISENS + 1                               SEN00610
          CALL PARDIF                                     SEN00620
C WRITE OUT X, RHO AND PAR.                               SEN00630
0026          WRITE(6,10) RHO                             SEN00640
0027          10 FORMAT(1H //30X,20HSENSITIVITY ANALYSIS // SEN00650
          15X,22HTHE VALUE OF R(RHO) IS ,E12.5)          SEN00660
          WRITE(6,20)                                     SEN00670
0028          20 FORMAT(/6X, 64HTHE POINT AT WHICH THE ESTIMATE OF SENSITIVITY WILL SEN00680
          1BE MADE IS )                                   SEN00690
          DO 40 I=1,N,6                                    SEN00700
          II=MINO(I+5,N)                                  SEN00710
          WRITE(6,30) ((J,X(J)), J=I,II)                 SEN00720
          30 FORMAT( 6(2X,2MX(,12,2M)=,G14.7) )         SEN00730
          40 CONTINUE                                     SEN00740
C CALCULATION OF GP VARIABLE VALUE                        SEN00750
          IF(NEXOP6.NE.1) GO TO 660                      SEN00760
          INDEX=0.                                        SEN00770
          CALL TRANS(INDEX)                               SEN00780
          WRITE(6,600)                                    SEN00790
          600 FORMAT(16X,'CORRESPONDING POINT FOR GP PROBLEM IS') SEN00800
          DO 640 I=1,N,6                                  SEN00810
          II=MINO(I+5,N)                                  SEN00820
          640 WRITE(6,30)((J,X(J)),J=I,II)               SEN00830
          DO 650 I=1,N                                    SEN00840
          650 X(I)=TX(I)                                  SEN00850
          660 CONTINUE                                    SEN00860
          WRITE(6,50) ((I,PAR(I),DPAR(I)), I=1,NPAR)     SEN00870
          50 FORMAT(45H0 PARAMETER VALUE DIFFERENCING INTERVAL / SEN00880
          1 (14,2X,G14.6,6X,G14.5) )                    SEN00890
          WRITE(6,55)                                     SEN00900
          55 FORMAT(//)                                   SEN00910
C EVALUATE F, G AND H.                                    SEN00920

```

Figure 34.--Subroutine SENS.

```

0050          CALL RESTNT(I,F)
0051          IF(MPMZ.EQ.0) GO TO 85
0052          DO 80 J=1,NPMZ
0053          CALL RESTNT(J,RJ(J))
0054          80 CONTINUE
0055          IF(NEXOP5.EQ.2) GO TO 85
0056          CALL PRESENIDU,KTEST)
0057          IF(LY.GT.0) GO TO 400
C COMPUTE DEL F.
0058          85 CALL GRAD(10)
0059          DO 90 I=1,N
0060          X3H(I) = DEL(I)
0061          90 CONTINUE
C COMPUTE (DEL)**2 P - STORED IN A.
0062          CALL SECORD(2)
C PERFORM THE L-U DECOMPOSITION OF A.
0063          DO 100 I=1,N
0064          DELX(I)=0.0
0065          100 CONTINUE
0066          NP=NT3
0067          NT3=1
0068          CALL INVERS(1)
C CHECK TO MAKE SURE AN ORTHOGONAL MOVE IS NOT ATTEMPTED.
0069          DO 110 I=1,N
0070          IF(DELX(I).EQ.0.0) GO TO 110
0071          WRITE(6,105)
0072          105 FORMAT(94H0 THE MATRIX OF SECOND PARTIALS IS NOT POSITIVE DEFINITE
1, SENSITIVITY ANALYSIS IS TERMINATED )
0073          GO TO 202
0074          110 CONTINUE
0075          DO 120 I=1,NPAR
0076          X2H(I) = PAR(I)
0077          120 CONTINUE
0078          DO 200 J=1,NPAR
0079          IF(NEXOP5.NE.0) GO TO 115
0080          IF(KTEST(J).EQ.0) GO TO 200
0081          115 IF(NEXOP4.EQ.0) GO TO 121
0082          CALL LMULT(0,DELMU,DEM,DU)
C COMPUTE D(DEL P)/DA(J) AND D(F)/DA(J) USING CENTRAL DIFFERENCING.
0083          121 PAR(J) = PAR(J) + DPAR(J)
0084          CALL RESTNT(0,DF)
0085          IF(M.EQ.0) GO TO 126
0086          DO 125 I=1,M
0087          CALL RESTNT(I,RJ(I))
0088          IF(RJ(I).GT.AVAL) GO TO 125
0089          123 DPAR(J)=0.1*DPAR(J)
0090          PAR(J) = X2H(I)
0091          WRITE(6,124) J,DPAR(J)
0092          124 FORMAT(16H RESETTING DPAR(I,2,3H) = ,G14.5)
0093          IF(DPAR(J).EQ.0.) GO TO 201
0094          GO TO 121
0095          125 CONTINUE
0096          126 IF(MZ.EQ.0) GO TO 128
0097          DO 127 I=1,MZ
0098          MZPI=M+I
0099          CALL RESTNT(MZPI,RJ(MZPI))
0100          127 CONTINUE
0101          128 IF(NEXOP4.EQ.0) GO TO 129
0102          CALL LMULT(1,DELMU,DEM,DU)
0103          129 CALL GRAD(2)
0104          DO 130 I=1,N
0105          DELX(I)=DELX0(I)
0106          130 CONTINUE
0107          DEM=2.0*DPAR(J)
0108          PAR(J)=PAR(J) - DEM
0109          CALL RESTNT(0,VAL)
0110          IF(M.EQ.0) GO TO 136
0111          DO 135 I=1,M
0112          CALL RESTNT(I,RJ(I))
0113          IF(RJ(I).GT.AVAL) GO TO 135
0114          GO TO 123
0115          135 CONTINUE
0116          136 IF(MZ.EQ.0) GO TO 138
0117          DO 137 I=1,MZ
0118          MZPI=M+I
0119          CALL RESTNT(MZPI,RJ(MZPI))
0120          137 CONTINUE
0121          138 IF(NEXOP4.EQ.0) GO TO 139
0122          CALL LMULT(2,DELMU,DEM,DU)
0123          139 CALL GRAD(2)
0124          DF=(DF-VAL)/DEM
0125          DO 140 I=1,N
0126          DELX(I)=(DELX(I) - DELX0(I))/DEM
0127          140 CONTINUE
0128          PAR(J) = X2H(I)
C HAVING ALREADY FACTORED A, SOLVE A*X=B FOR X,
C WHERE B=DELX AND X=DX/DA(I).
SEN00930
SEN00940
SEN00950
SEN00960
SEN00970
SEN00980
SEN00990
SEN01000
SEN01010
SEN01020
SFN01030
SEN01040
SEN01050
SEN01060
SEN01070
SEN01080
SFN01090
SEN01100
SEN01110
SEN01120
SFN01130
SEN01140
SEN01150
SEN01160
SEN01170
SEN01180
SFN01190
SEN01200
SEN01210
SEN01220
SEN01230
SEN01240
SEN01250
SEN01260
SEN01270
SEN01280
SFN01290
SEN01300
SEN01310
SEN01320
SEN01330
SEN01340
SEN01350
SEN01360
SEN01370
SEN01380
SEN01390
SEN01400
SEN01410
SEN01420
SEN01430
SEN01440
SEN01450
SEN01460
SEN01470
SEN01480
SEN01490
SEN01500
SEN01510
SEN01520
SEN01530
SEN01540
SEN01550
SEN01560
SEN01570
SEN01580
SEN01590
SEN01600
SEN01610
SEN01620
SEN01630
SEN01640
SEN01650
SEN01660
SEN01670
SEN01680
SEN01690
SEN01700
SEN01710
SEN01720
SEN01730
SEN01740
SEN01750
SEN01760
SEN01770
SEN01780
SEN01790

```

Figure 34.--continued

```

0129          CALL INVERS(2)
C PRINT OUT DX/DA(J)
0130          WRITE(6,150) J
0131          150 FORMAT(47H X-DERIVATIVES ARE WITH RESPECT TO PARAMETER ,I2)
0132          DO 170 I=1,N,6
0133             II=MINO(I+5,N)
0134             WRITE(6,160) (JJ,DELX(JJ), JJ=I,II)
0135          160 FORMAT( 6(4H DX(,I2,2H)=,G14.7) )
0136          170 CONTINUE
C CALCULATION OF GP VARIABLE SENSITIVITY
0137          IF(NEXOP6.NE.1) GO TO 670
0138          INDEX=1
0139          CALL TRANS(INDEX)
0140          WRITE(6,610) J
0141          610 FORMAT(2X,'CORRESPONDING DERIVATIVES OF GP VARIABLES WRT'
I,' PARAMETER',I2,'ARE')
0142          DO 680 I=1,N,6
0143             II=MINO(I+5,N)
0144             WRITE(6,160)(JJ,DELX(JJ),JJ=I,II)
0145          680 CONTINUE
0146          DO 690 I=1,N
0147             690 DELX(I)=TDELX(I)
0148          670 CONTINUE
0149          IF(NEXOP6.EQ.0) GO TO 375
0150          IF(N.EQ.0) GO TO 301
0151          CALL LMULT(3,DELMU,DEM,DU)
0152          WRITE(6,350) J
0153          350 FORMAT(/ 41H U-DERIVATIVES WITH RESPECT TO PARAMETER ,I2)
0154          DO 351 I=1,M,6
0155             II = MINO(I+5,M)
0156             WRITE(6,352) ((JJ,DU(JJ)),JJ=I,II)
0157          352 FORMAT(6(4H DU(,I2,2H)=,G14.7))
0158          351 CONTINUE
0159          301 IF(MZ.EQ.0) GO TO 375
0160          CALL LMULT(4,DELMU,DEM,DU)
0161          WRITE(6,360) J
0162          360 FORMAT(/ 41H W-DERIVATIVES WITH RESPECT TO PARAMETER ,I2)
0163          DO 361 I=1,MZ,6
0164             II = MINO(I+5,MZ)
0165             WRITE(6,362) ((JJ,DU(JJ+M)), JJ=I,II)
0166          362 FORMAT(6(4H DW(,I2,2H)=,G14.7))
0167          361 CONTINUE
0168          375 CONTINUE
C COMPUTE DF/DA(J).
0169          DO 180 I=1,N
0170             DF = DF + X3H(I)*DELX(I)
0171          180 CONTINUE
C PRINT DF/DA(J).
0172          WRITE(6,190) DF
0173          190 FORMAT(/ 10X,13HDF(X(H))/DA= ,G14.6/10H *****)
0174          200 CONTINUE
0175          GO TO 202
0176          201 WRITE(6,106) J
0177          106 FORMAT(1H ,2INTERMINATING PARAMETER,13,16H DUE TO DPAR = 0 /)
0178          GO TO 200
0179          202 NT3=NP
0180          CALL REJECT
0181          DO 205 I=1,N
0182             DELX(I) = DELT(I)
0183          205 DELX(I) = DELT(I)
0184          400 CONTINUE
0185          500 RETURN
0186          END
SEN01800
SEN01810
SEN01820
SEN01830
SEN01840
SEN01850
SEN01860
SEN01870
SEN01880
SEN01890
SEN01900
SEN01910
SEN01920
SEN01930
SEN01940
SEN01950
SEN01960
SEN01970
SEN01980
SEN01990
SEN02000
SEN02010
SEN02020
SEN02030
SEN02040
SEN02050
SEN02060
SEN02070
SEN02080
SEN02090
SEN02100
SEN02110
SEN02120
SEN02130
SEN02140
SEN02150
SEN02160
SEN02170
SEN02180
SEN02190
SEN02200
SEN02210
SEN02220
SEN02230
SEN02240
SEN02250
SEN02260
SEN02270
SEN02280
SEN02290
SEN02300
SEN02310
SEN02320
SEN02330
SEN02340
SEN02350
SEN02360
SEN02370
SEN02380
SEN02390
SEN02400
SEN02410
SEN02420

```

Figure 34.--continued

```

FORTRAN IV G LEVEL 21          SET          DATE = 80242          12/14/52

0001          SUBROUTINE SET (TMAX)          SET00040
          C          SET00050
          C          FEBRUARY 1971          SET00060
          C          SET00070
          C SET STORES THE TIME AT WHICH THE PROBLEM IS BEGUN          SET00080
          COMMON /TSM/ NSMW          SET00090
          COMMON /TMX/ TMO,EXT,EXT90          SET00100
          C          SET00110
          C          SET00120
          C          SECOND GIVES JOB CPU EXECUTION TIME IN 1/1000 OF A SECOND          SET00130
          C          SET00140
          C          SET00150
          C          SET00160
          C          SET00170
          C          SET00180
          C          SET00190

0004          CALL SECOND (TMO)          SET00140
0005          EXT=TMAX+TMO          SET00150
0006          EXT90= TMO + 0.90*TMAX          SET00160
0007          NSMW=1          SET00170
0008          RETURN          SET00180
0009          END          SET00190

```

Figure 35.--Subroutine SET.

6.28 STORE

Subroutine STORE stores the values of the current point x and the associated values of F , W , and G in a temporary storage area. These values are restored by a call to REJECT. The listing for subroutine STORE appears as Figure 36.

```

FORTRAN IV G LEVEL 21          STORE          DATE = 80242          12/15/22

0001          SUBROUTINE STORE          ST000040
          C          ST000050
          C          OCTOBER 1970          ST000060
          C          ST000070
          C STORE STORES THE VALUES OF THE CURRENT POINT AND THE ASSOCIATED          ST000080
          C VALUES OF THE FUNCTIONS IN A TEMPORARY AREA.          ST000090
          IMPLICIT REAL*8(A-H,O-Z)          ST000100
          REAL*4 RMOIN,RATIO,EPSI,THETA0          ST000110
          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1          ST000120
          COMMON /EQUAL/ M, M1, M2          ST000130
          COMMON/VALUE/F,G,PO,RSIGMA,RJ(90),RMD          ST000140
          COMMON/CRST/DELX(45),DELX0(45),RMOIN,RATIO,EPSI,THETA0,          ST000150
          IRSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XRI(45),PRI,          ST000160
          ZPRZ,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),          ST000170
          3 PREV3,ADELX, NTCTR, NUMINI, NPHASE, NSATIS          ST000180
          DO 10 I=1,N          ST000190
          X1(I)=X(I)          ST000200
          MNZ=M+MZ          ST000210
          DO 20 J=1,MNZ          ST000220
          RJ(IJ)=RJ(IJ)          ST000230
          P1=PO          ST000240
          F1=F          ST000250
          G1=G          ST000260
          RSIG1=RSIGMA          ST000270
          M1=M          ST000280
          RETURN          ST000290
          END          ST000300

0008          DO 10 I=1,N          ST000190
0009          X1(I)=X(I)          ST000200
0010          MNZ=M+MZ          ST000210
0011          DO 20 J=1,MNZ          ST000220
0012          RJ(IJ)=RJ(IJ)          ST000230
0013          P1=PO          ST000240
0014          F1=F          ST000250
0015          G1=G          ST000260
0016          RSIG1=RSIGMA          ST000270
0017          M1=M          ST000280
0018          RETURN          ST000290
0019          END          ST000300

```

Figure 36.--Subroutine STORE.

6.29 TCHECK

If the print option has been set so that printouts occur only after a subproblem has been called, then TCHECK is called after each iteration of the algorithm used to minimize the W function. The elapsed time is computed as is done in TIMEC but it is not printed out. If the elapsed time exceeds 90 percent of the estimated time limit, then the print option is changed so a printout occurs after each iteration of the procedure being used to minimize the W function. If the elapsed time exceeds the user specified time limit, then the routine will cause termination of the attempt to solve the problem by setting NSWW equal to 2. Subroutine TCHECK appears as Figure 37.

```

FORTRAN IV G LEVEL 21          TCHECK          DATE = 80243          10/40/07

0001          SUBROUTINE TCHECK                      TCM00050
          C          FEBRUARY 1971                    TCM00060
          C          TCM00070
          C TCHECK CHECKS THE NUMBER OF SECONDS THAT HAVE ELAPSED SINCE THE START TCM00080
          C OF THE PROBLEM. IF THE SOLUTION IS TAKING LONGER THAN 90 PER-CENT TCM00090
          C OF THE ESTIMATED MAXIMUM TIME, A SWITCH IS SET TO GIVE MORE OUTPUT. TCM00100
0002          COMMON /TSM/ NSWW                      TCM00110
0003          COMMON /OPTNS/ NT1,NT2,NT3,NT4,NT5,NT6,NT7,NT8,NT9,NT10 TCM00120
0004          COMMON /TMX/ TMO,EXT,EXT90              TCM00130
0005          CALL SECUND (SECS)                      TCM00140
0006          IF (SECS.LT.EXT90) RETURN                TCM00150
          C GETTING CLOSE TO EXCEEDING THE TIME LIMIT SET OUTPUT OPTION TO GIVE TCM00160
          C MORE OUTPUT.                              TCM00170
0007          NT3=2                                    TCM00180
0008          X=SECS - TMO                             TCM00190
0009          WRITE(6,10) X                            TCM00200
0010          10  FORMAT (6X,5HTIME=,F9.3,8H SECONDS ) TCM00210
0011          IF (SECS .GT. EXT) NSWW=2               TCM00220
0012          CALL OUTPUT (1)                          TCM00230
0013          RETURN                                    TCM00240
0014          END                                      TCM00250

```

Figure 37.--Subroutine TCHECK.

6.30 TIMEC

Subroutine TIMEC calls the same routine as called by SET to obtain the current status of the computer's time clock. It then prints out the elapsed time since the call to SET. If the elapsed time is greater than the user specified maximum time limit, TIMEC will cause the termination of the attempt to solve the problem by setting NSWW

equal to 2. The listing for subroutine TIMEC is shown below in Figure 38.

```

FORTRAN IV G LEVEL 21          TIMEC          DATE = 80242          12/15/53
0001          SUBROUTINE TIMEC          TIM00030
          C          TIM00040
          C          FEBRUARY 1971          TIM00050
          C          TIM00060
          C TIMEC CHECKS THE NUMBER OF SECONDS THAT HAVE ELAPSED SINCE THE START          TIM00070
          C OF THE PROBLEM. IT PRINTS THIS NUMBER. IF THE SOLUTION IS TAKING          TIM00080
          C LONGER THAN THE ESTIMATED MAXIMUM TIME, A SWITCH IS SET TO TERMINATE          TIM00090
          C THE RUN.          TIM00100
0002          COMMON /TSW/ NSWM          TIM00110
0003          COMMON /TMX/ TMO,EXT,EXT90          TIM00120
          C          TIM00130
          C SECOND GIVES JOB CPU EXECUTION TIME IN 1/1000 OF A SECOND          TIM00140
          C          TIM00150
0004          CALL SECOND (SECS)          TIM00160
0005          X=SECS-TMO          TIM00170
0006          WRITE (6,20) X          TIM00180
0007          IF (SECS.LT.EXT) GO TO 10          TIM00190
0008          NSWM=2          TIM00200
0009          10          RETURN          TIM00210
          C          TIM00220
0010          20          FORMAT (6X,5HTIME=F9.3,0M SECONDS)          TIM00230
0011          END          TIM00240

```

Figure 38.--Subroutine TIMEC.

6.31 TRANS

TRANS is a special purpose subroutine that performs an exponential transformation of the solution vector and sensitivity results. This subroutine is useful when solving problem $c(x)$, the convex equivalent of the geometric programming problem $G(t)$. The former problem is obtained from the latter via the transformation $t = e^{-x_i}$, $i=1, \dots, n$. TRANS is used to transform the results to the original space of the variable t . The motivation for coding this subroutine was the computational difficulty experienced by some users while solving geometric programming problems with SUMT. TRANS is called by the subroutines OUTPUT and SENS. Figure 39 is a listing of subroutine TRANS.

6.32 XMØVE

Subroutine XMØVE contains most of the logic of the algorithm used to minimize the W function for a given value of r (RHØ). First, a

```

FORTRAN IV G LEVEL 21          TRANS          DATE = 80243          12/19/30
0001          SUBROUTINE TRANS(INDEX)          TRA00050
C SUBROUTINE TRANS CODED BY GHAEMI(1979) TRANSFORMS THE OPTIMAL          TRA00060
CAP AND SENSITIVITY RESULTSTO T SPACE WHERE          TRA00070
C Y=EXP(-X). THIS ROUTINE IS USEFULL WHEN USER SOLVES THE CONVEX          TRA00080
C EQUIVALENT OF GP PROBLEMS OBTAINED BY THE ABOVE EXPONENTIAL          TRA00090
C TRANSFORMATION.          TRA00100
0002          IMPLICIT REAL*8(A-H,D-Z)          TRA00110
0003          COMMON/SHAPE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1          TRA00120
0004          COMMON/CRST/DELX(45),DELX0(45),RHOIN,KATIO,EPSI,THETA0,          TRA00130
          IRS101,G1,X1(45),X2(45),X3(45),XR2(45),XR1(45),PR1,          TRA00140
          ZPR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),          TRA00150
          3 PHEV3,ADELX, NCTR, NUMINI, NPHASE, NSATIS          TRA00160
0005          COMMON/ABGS/TX(45),TDELX(45)          TRA00170
0006          COMMON/ABGS/XX(45)          TRA00180
0007          IF(INDEX.EQ.1) GO TO 20          TRA00190
0008          DO 10 I=1,N          TRA00200
0009          TX(I)=X(I)          TRA00210
0010          X(I)=DEXP(-X(I))          TRA00220
0011          10 XX(I)=X(I)          TRA00230
0012          RETURN          TRA00240
0013          20 DO 30 I=1,N          TRA00250
0014          TDELX(I)=DELX(I)          TRA00260
0015          30 DELX(I)=-DEXP(-X(I))*DELX(I)          TRA00270
0016          RETURN          TRA00280
0017          END          TRA00290

```

Figure 39.--Subroutine TRANS.

direction is chosen in which it is believed the W function will initially decrease. Then \emptyset PT is used to find a minimum of the W function in that direction. Having generated a new point giving a lower value of the W function, $X\emptyset$ VE returns control to $B\emptyset$ DY.

Subroutine $X\emptyset$ VE contains three procedures for generating a direction vector. The choice of the procedure used is controlled by experimental option 2 ($NEX\emptyset$ P2). If $NEX\emptyset$ P2 = 1, then Newton's method is used to choose the direction vector S , that $S^i = -[V^2W(x^i, r_k)]^{-1} \nabla W(x^i)$. If $V^2W(x^i, r^k)$ is not a positive definite matrix, S^i is generated by rules outlined on page 167 in Fiacco and McCormick [10].

If $NEX\emptyset$ P2 = 2, then the negative of the gradient is used as the direction vector, $S^i = -\nabla W(x^i, r^k)$.

When $NEX\emptyset$ P2 = 3 a variable matrix method is used to generate S^i . This method is McCormick's modification of the Fletcher-Power-Davidon, as described on pages 170-175 in Fiacco and McCormick [10]. Subroutine $X\emptyset$ VE is shown as Figure 40.

```

FORTRAN IV G LEVEL 21          XMOVE          DATE = 80242          12/16/26

0001          SUBROUTINE XMOVE                                XMO00040
C                                                    XMO00050
C          MARCH 1971                                       XMO00060
C                                                    XMO00070
C XMOVE DETERMINES THE VECTOR ALONG WHICH THE SEARCH FOR A MINIMUM IS
C USING OPT.                                               XMO00080
0002          IMPLICIT REAL*8(A-H,O-Z)                     XMO00100
0003          REAL*4 RMDIN,RATIO,EPSI,THETA0,XEP1,XEP2      XMO00110
0004          COMMON/SHARE/X(45),DEL(45),A(45,45),N,M,MN,NP1,NM1 XMO00120
0005          COMMON/CRST/DELX(45),DELXO(45),RMDIN,RATIO,EPSI,THETA0,
          1RSIG1,G1,X1(45),X2(45),X3(45),XR2(45),XRI(45),PR1,
          2PR2,P1,F1,RJ1(90),DOTT,PGRAD(45),DIAG(45),
          3PREV3,ADELX,MTCTR,NUMINI,NPHASE,NSATIS
0006          COMMON/XEPOPT/NEXP1,NEXP2,NEXP3,NEXP4,NEXP5,XFP1,XEP2 XMO00170
0007          COMMON/XVE/SIG(45),YI(45),XX(45),DELL(45)
C--NEXP2 DETERMINES HOW MOVE IS TO BE MADE
C NEXP2 = 1 USE MODIFIED NEWTON RAPHSON METHOD.
C          = 2 USE MODIFIED NEWTON RAPHSON METHOD, BUT ADD DELXO TO
C          ORTHOGONAL MOVE VECTOR IF HESSIAN IS INDEFINITE.
C          = 3 USE STEEPEST DESCENT METHOD.
C          = 4 USE MCCORMICK,S MODIFICATION OF THE FLETCHER-POWELL
C          METHOD.
0008          GO TO (10,10,180,30), NEXP2
C--NEWTON -RAPH WITH WHATEVER METHOD IS IN INVERSE
0009          10 CALL GRAD (1)
C--ONE (1) MEANS ACCUMULATE MATRIX OF SECOND PARTIAL DERIVATIVES
0010          CALL SECND (1)
0011          DO 20 I=1,N
0012          20 DELX(I)=DELXO(I)
0013          CALL INVERS (1)
C IF A NONPOSITIVE PIVOT IS ENCOUNTERED IN INVERSE AN ATTEMPT IS MADE TO
C COMPUTE A VECTOR HAVING A POSITIVE DOT PRODUCT WITH A NEGATIVE
C EIGENVECTOR AND THE NEGATIVE OF DEL P.
0014          CALL STORE
0015          CALL OPT
0016          RETURN
C--F-P-D-MCC MOVE
0017          30 CALL GRAD (2)
C--MN IS NO. OF MOVES FOR THIS VALUE OF RMD
0018          IF (MN.NE.0) GO TO 70
0019          40 IREP=0
0020          IT=0
C--SET INITIAL GUESS INVERS MATRIX OF SECOND PARTIAL DERIVATIVES
C-- USE PARTIAL INVERSE IF KNOWN
0021          DO 50 I=1,N
0022          DO 50 J=1,N
0023          50 A(I,J)=0.0
0024          DO 60 I=1,N
0025          60 A(I,I)=1.0
0026          DO 80 I=1,N
0027          80 DELX(I)=DELXO(I)
0028          IF (IREP.GT.N) GO TO 40
0029          IF (IT.EQ.0) GO TO 130
0030          DO 90 I=1,N
0031          SIG(I)=X(I)-XXX(I)
0032          90 YI(I)=DELL(I)-DELXO(I)
C--NEGATIVE GRADIENT STORED AND COMPUTED
C--COMPUTE MY
0033          DO 100 I=1,N
0034          DELX(I)=0.0
0035          DO 100 J=1,N
0036          100 DELX(I)=DELX(I)+A(I,J)*YI(J)
C--COMPUTE Y(SIG -MY)-1
0037          ZCON=0.0
0038          DO 110 I=1,N
0039          110 ZCON=ZCON+YI(I)*(SIG(I)-DELX(I))
0040          IF (ZCON.EQ.0.0) GO TO 130
0041          IREP=IREP+1
0042          ZC=1./ZCON
C-- UPDATE H MATRIX USING MCC FORMULA WHEN SCALAR ME 0
0043          DO 120 I=1,N
0044          T1=ZC*(SIG(I)-DELX(I))
0045          DO 120 J=1,N
0046          A(I,J)=A(I,J)+T1*(DELX(J)+SIG(J))
0047          120 A(J,I)=A(I,J)
C-- STORE CURRENT POINT AND CURRENT GRADIENT (NEG)
0048          130 DO 140 I=1,N
0049          XXX(I)=X(I)
0050          140 DELL(I)=DELXO(I)
0051          DO 150 I=1,N
0052          DELX(I)=0.0
0053          DO 150 J=1,N
0054          150 DELX(I)=DELX(I)+A(I,J)*DELXO(J)
0055          ZC1=0.0
0056          DO 160 I=1,N
0057          160 ZC1=DELX(I)**2+ZC1
0058          70 ZC1=DSORT(ZC1)

```

Figure 40.--Subroutine XMOVE.

```
0059          DO 170 I=1,N          XMO00910
0060      170  DELX(I)=DELX(I)/ZC1    XMO00920
0061          CALL STORE            XMO00930
0062          CALL OPT              XMO00940
0063          IT=IT+1              XMO00950
0064          RETURN               XMO00960
0065      180  CONTINUE            XMO00970
          C STEEPEST DESCENT      XMO00980
0066          CALL GRAD (2)        XMO00990
0067          DO 190 I=1,N          XMO01000
0068      190  DELX(I)=DELX(I)      XMO01010
0069          CALL STORE            XMO01020
0070          CALL OPT              XMO01030
0071          RETURN               XMO01040
0072          END                  XMO01050
```

Figure 40.--*continued*

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