

AD-A050 206

UNIVERSITY OF SOUTH FLORIDA TAMPA  
THEORY AND COMPUTER PROGRAM FOR TRANSFER FUNCTION IDENTIFICATION--ETC(U)  
FEB 78 V K JAIN, G J DOBECK, L J LAWDERMILT N61331-75-C-0012  
NCSL-TM-204-78 NL

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1 OF 2

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N61331-75-C-0012---Main Contract  
N61339-75-C-0012---Subcontract No. 1-E-21A02  
From Georgia Institute of Technology per  
Ms. Marilyn Griffin, Naval Coastal Systems  
Laboratory

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DEPARTMENT OF THE NAVY  
NAVAL COASTAL SYSTEMS LABORATORY  
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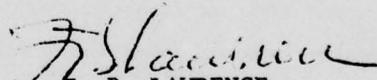
(1) Change Report Number (block 1) to: NCSL TM-204-78

(2) Change Contract or Grant Number(s) (block 8) to:

N61331-75-C-0012  
Subc. 1-E-21A02

(3) Change report date (block 12) to: February 1978

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REPORT DOCUMENTATION PAGE		READ INSTRUCTIONS BEFORE COMPLETING FORM
1. REPORT NUMBER NCSL TM-204-78 ✓	2. GOVT ACCESSION NO.	3. RECIPIENT'S CATALOG NUMBER
4. TITLE (and Subtitle) Theory and Computer Program for Transfer Function Identification by the GRAM Identifier with Application to Undersea Vehicles.		5. TYPE OF REPORT & PERIOD COVERED Technical Memo
7. AUTHOR(s) V. K./Jain, G. J./Dobeck L. J./Lawdermilt		6. PERFORMING ORG. REPORT NUMBER
9. PERFORMING ORGANIZATION NAME AND ADDRESS University of South Florida ✓ Tampa, Florida 33620		8. CONTRACT OR GRANT NUMBER(s) 15 N61331-75-C-0112 N61339-75-C-0122
11. CONTROLLING OFFICE NAME AND ADDRESS Naval Coastal Systems Laboratory Panama City, Florida 32407		10. PROGRAM ELEMENT, PROJECT, TASK AREA & WORK UNIT NUMBERS Task Area 16 ZF 61112001, Task Area No. SMW 2
14. MONITORING AGENCY NAME & ADDRESS (if different from Controlling Office) 18 NCSL 19 TM-204-78		12. REPORT DATE 11 FEBRUARY 1978
16. DISTRIBUTION STATEMENT (of this Report) Approved for Public Release - Distribution Unlimited 16 F61112, SMW 2		13. NUMBER OF PAGES 105 12 106 p.
17. DISTRIBUTION STATEMENT (of the abstract entered in Block 20, if different from Report)		15. SECURITY CLASS. (of this report) Unclassified
18. SUPPLEMENTARY NOTES		15a. DECLASSIFICATION/DOWNGRADING SCHEDULE N/A
19. KEY WORDS (Continue on reverse side if necessary and identify by block number) Computer Programs      Dynamics      Flight Test Data Underwater Vehicles      Identifiers:      Submerged Vehicles Vehicles      Parameters Estimates      Vehicle Dynamics Transfer Functions      Measurement Filters Models      GRAM Identifier		
20. ABSTRACT (Continue on reverse side if necessary and identify by block number) A method for the analysis of submerged vehicle dynamics is described. It enables the test engineer to identify the transfer function parameters from actual flight-test data (for post-flight analysis), or from simulated trajectories (for designing tow-test maneuvers), via the computer program GRAM. The method is noniterative and yields reliable estimates in the presence of disturbances and instrumentation noise. ←		

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TABLE OF CONTENTS

1.	INTRODUCTION	1
2.	GRAM IDENTIFIER	2
	Measurement Filter Theorem	6
	Measurement Vector	7
	Generalized Least-Squares Formulation of the Identification Problem	8
	Solution of the Identification Problem	9
3.	PROGRAM DESCRIPTION	13
	Input Data Cards	14
	Memory	18
	Output	19
	Program GRAM	21
4.	APPLICATION EXAMPLES	27
	Example 1	27
	Example 2	33
	APPENDIX A	A-1
	APPENDIX B	B-1
	APPENDIX C	C-1

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## LIST OF ILLUSTRATIONS

<u>Figure No.</u>		<u>Page No.</u>
1.	Single-input, Single-output Identification Problem	3
2.	Measurement Filter System	5
3.	Simulated Feedback Loop	13
4.	Experiment 3 - Third Order Pitch vs Stern-plane Identification	30
5.	Experiment 5 - Detection of Dominant Complex Poles	32
6.	Identification of Longitudinal Dynamics of SMS2619 Vehicle; Reconstruction of Pitch From Identified Model	34
7.	Identification of Lateral Dynamics of SMS2619 Vehicle; Reconstruction of Roll From Identified Model	34

## I. INTRODUCTION

A major impediment in the study of the dynamics of a particular submerged vehicle is a lack of accurate functional relationship between the motion variables and the control inputs. In the uncoupled linear case such relationships may consist of linear transfer functions. Although their general form including the degrees of the denominator and numerator polynomials is known [1], the coefficients of these transfer functions are often unknown. They must be determined either through analytical formulas involving hydrodynamic coefficients or through identification algorithms performed upon experimental flight test data. The purpose of this report is to present a new identification method 'GRAM Identifier' and a computer program for its application to flight test data of submerged vehicles.

The method discussed possesses the following advantages: a) it is noniterative and therefore computationally fast, and b) it is noise-worthy[2]-[4]. Only the single-input, single-output case is considered in the present report. It will therefore be assumed that the flight test data consist of single input maneuvers, each caused by the actuation of a single control surface while the remaining control surfaces are held at zero deflection. To aid the engineer, a computer program entitled GRAM has been written that performs the necessary computations. The program is suitable for analysis of actual flight test data as well as for a simulation mode. In the latter case the flight trajectories are first generated, incorporating synthetic disturbance and measurement noise, and identification is then performed on the simulated trajectories. The simulation mode is useful when, for example, the approximate transfer functions of the vehicle are known (from hydrodynamic computations) and it is desired to find efficient maneuvers so as to develop a flight test plan.

The structure of the report is as follows. Section II presents the theory of the GRAM Identifier. Section III gives a user oriented description of the computer program. The results of some case studies, including those performed on actual vehicles, are provided in Section IV. Appendix A includes the listing and flow charts of the subroutines used by GRAM. Appendix B provides a brief discussion on the equivalence of z-domain and s-domain transfer functions and Appendix C deals with the solution of a key equation.

## II. GRAM IDENTIFIER

The identification problem is formulated with reference to Fig. 1. The variable  $u$  represents a nonzero input variable -- the stern-plane angle, the rudder angle or other control surface deflection. The corresponding response  $y$  represents one of the motion variables -- pitch angle, yaw rate or some other. In part (a) of the figure is shown the vehicle, the instrumentation for the input-output variables, and the necessary samplers for digitization of these signals. Part (b) of the figure provides a discrete-time interpretation of the identification problem, which is stated as follows:

Given

- i) the input and output measurements  $v(k)$ ,  $x(k)$ ,  $k=1, \dots, K$ ,
- ii) the integers  $n$  and  $r$  in the model

$$y(k) + a_1 y(k-1) + \dots + a_n y(k-n) = b_0 u(k) + \dots + b_r u(k-r) \quad (1)$$

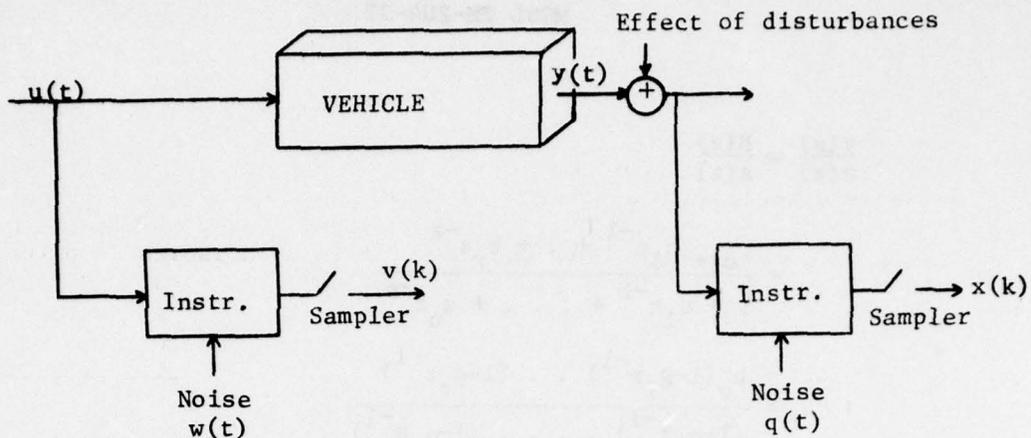
- iii) a statistical description of the noise processes  $w(k)$  and  $q(k)$ , find the unknown parameters  $a_i$  and  $b_i$  so that the model provides the best fit (in some sense) into the measured data.

Note that the quantities  $a_i$ ,  $b_i$ ,  $y(k)$  and  $u(k)$  are in fact to be estimated. Only  $x(k)$  and  $v(k)$  are directly available.

### Remarks

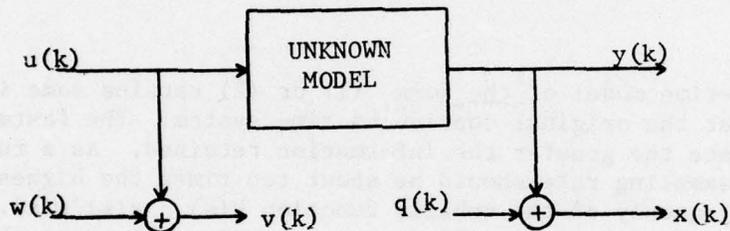
The reader familiar with the principles of signal sampling may wish to skip these remarks.

- Note that  $y(k) = y(k\Delta)$ ,  $u(k) = u(k\Delta)$ , etc., where  $\Delta$  is the sampling interval.
- In terms of the z-transform variable the relationship of (1) can be written as [5]



(a) Single-input, single-output maneuver

$$y(s)/u(s) = H(s)$$



(b) Discrete time identification problem

$$y(z)/u(z) = H(z)$$

Fig. 1. Single-input, single-output identification problem.

$$\frac{y(z)}{u(z)} = \frac{B(z)}{A(z)} \quad (2a)$$

$$= \frac{b_0 + b_1 z^{-1} + \dots + b_r z^{-r}}{1 + a_1 z^{-1} + \dots + a_n z^{-n}} \quad (2b)$$

$$= \frac{b_0 (1 - \beta_1 z^{-1}) \dots (1 - \beta_r z^{-1})}{(1 - \alpha_1 z^{-1}) \dots (1 - \alpha_n z^{-1})} \quad (2c)$$

The system described by (1), or (2) is stable if and only if each pole  $\alpha_i$  satisfies the condition  $|\alpha_i| < 1$

- A discrete-time model of the form (1) or (2) retains some information about the original continuous time system. The faster the sampling rate the greater the information retained. As a rule of thumb the sampling rate should be about ten times the highest critical frequency of the vehicle function  $H(s) = y(s)/u(s)$ . When this condition is satisfied a correspondence between the z-domain and s-domain functions may be achieved. Specifically, the equivalent continuous-time model becomes (see also Appendix B)

$$H(s) = \frac{\delta_0 (s+q_1) \dots (s+q_r)}{(s+p_1)(s+p_2) \dots (s+p_n)} \quad (3)$$

where

$$q_i = -\frac{1}{\Delta} \ln(\beta_i) \quad (\text{or } \beta_i = e^{-q_i \Delta})$$

$$p_i = -\frac{1}{\Delta} \ln(\alpha_i) \quad (\text{or } \alpha_i = e^{-p_i \Delta})$$

The relationship in (1), or equivalently (2b), may be written as

$$A^T \xi_n y(z) = B^T \xi_r u(z) \quad (4)$$

where

$$A^T = [1 \ a_1 \ \dots \ a_n]$$

$$B^T = [b_0 \ b_1 \ \dots \ b_r]$$

$$\xi_n^T = [1 \ z^{-1} \ \dots \ z^{-n}]$$

$$\xi_r^T = [1 \ z^{-1} \ \dots \ z^{-r}]$$

The super T denotes the transpose of a vector or matrix. Equation (4) should form the basis for modeling the vehicel dynamics. However, as discussed later, the use of the vector signals  $\xi_n^T y(z)$ ,  $\xi_r^T u(z)$  leads to poor results in identification. Instead GRAM<sup>n</sup> relies upon certain measurement signals shown in Fig. 2 generated by means of first order digital filters. Specifically, use is made of the vector signals

$$Y(k) = [y_0(k), y_1(k), \dots, y_n(k)]^T$$

$$U(k) = [u_{n-r}(k), \dots, u_n(k)]^T$$

consisting of the measurements at time instant  $k\Delta$ . They will be called output and input measurement vectors, respectively. Their z-transforms are denoted as  $Y(z)$  and  $U(z)$ . It can then be shown that

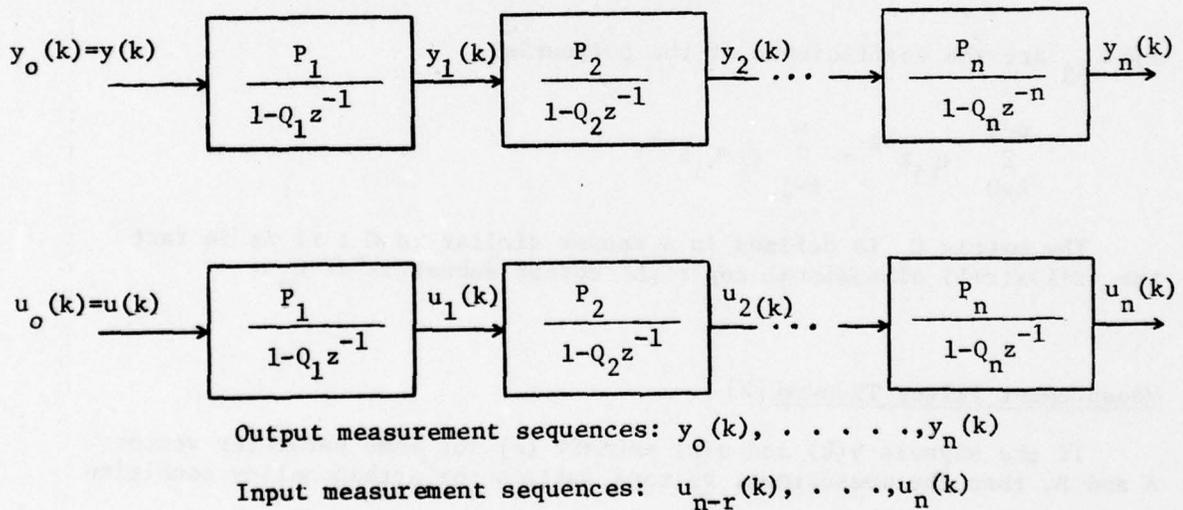


Fig. 2. Measurement filter system

$$Y(z) = \frac{1}{d_n(z)} C_n^T \xi_n y(z) \quad (5a)$$

$$U(z) = \frac{1}{d_n(z)} C_r^T \xi_r u(z) \quad (5b)$$

where

$$d_n(z) = \prod_{i=1}^n (1 - Q_i z^{-1}) / P_i$$

$$C_n = \begin{bmatrix} c_{00} & c_{01} & \dots & c_{0n} \\ c_{10} & c_{11} & & 0 \\ \cdot & \cdot & & 0 \\ \cdot & \cdot & & \cdot \\ c_{n0} & 0 & & 0 \end{bmatrix}$$

and  $c_{\ell j}$  are the coefficients of the polynomial

$$\sum_{\ell=0}^{n-j} c_{\ell j} z^{-\ell} = \prod_{i=j}^n (1 - Q_i z^{-1})$$

The matrix  $C_r$  is defined in a manner similar to  $C_n$ ; it is in fact the  $(r+1) \times (r+1)$  dimensional top right corner submatrix<sup>n</sup> of  $C_n$ .

Measurement Filter Theorem [2]

If the signals  $y(k)$  and  $u(k)$  satisfy (4) for some parameter vector  $A$  and  $B$ , then the measurement vectors satisfy the orthogonality condition

$$[\alpha^T - \beta^T] \begin{bmatrix} Y(k) \\ U(k) \end{bmatrix} = 0 \quad \text{for all } k$$

where

$$\alpha = C_n^{-1} A \tag{6a}$$

$$\beta = C_r^{-1} B \tag{6b}$$

Proof The matrices  $C_n$  and  $C_r$  are upper triangular about the cross-diagonal, the latter having nonzero entries. Hence these matrices are nonsingular. We can therefore rewrite (4) as

$$(C_n^{-1} A)^T C_n^T \xi_n y(z) = (C_r^{-1} B)^T C_r^T \xi_r u(z)$$

Substituting 6a and 6b and dividing through by  $d(z)$  one has

$$\alpha^T \frac{1}{d(z)} C_n^T \xi_n y(z) - \beta^T \frac{1}{d(z)} C_r^T \xi_r u(z) = 0$$

Upon substituting (5a) and (5b) this equation yields

$$[\alpha^T \quad -\beta^T] \begin{bmatrix} Y(z) \\ U(z) \end{bmatrix} = 0$$

The result sought by the theorem is obtained immediately upon taking the inverse transform. QED

Corollary: Let

$$\lambda = [\alpha^T \quad -\beta^T]^T \quad \text{Synthetic parameter vector}$$

and

$$f(k) = [Y^T(k) \quad U^T(k)]^T \quad \text{Model-measurement vector}$$

then

$$\lambda^T f(k) = 0 \quad \text{for all } k \tag{7}$$

Measurement vectors

As stated earlier, the sequences  $y(k)$  and  $u(k)$  are not actually available. Only  $x(k)$  and  $v(k)$  are, where

$$x(k) = y(k) + q(k)$$

$$v(k) = u(k) + w(k)$$

Because the system of measurement filters in Fig. 2. is linear, the following observation can now be made

Suppose that instead of processing  $y(k)$  the cascade of filters processes output noise  $q(k)$ . Similarly, let the lower cascade of filters process the noise sequence  $w(k)$ . And, let the resulting measurement sequences be denoted as  $q_i(k)$  and  $w_i(k)$ , respectively.

Then

$$x_i(k) = y_i(k) + q_i(k)$$

$$v_i(k) = u_i(k) + w_i(k)$$

where  $x_i(k)$  and  $v_i(k)$  are the data-measurement sequences obtained by processing  $x(k)$  and  $v(k)$  by the two cascades of measurement filters.

Let

$$Q(k) = [q_0(k), \dots, q_n(k)]$$

$$W(k) = [w_{n-r}(k), \dots, w_n(k)]$$

$$X(k) = [x_0(k), \dots, x_n(k)]$$

$$V(k) = [v_{n-r}(k), \dots, v_n(k)]$$

$$e(k) = [Q^T(k), W^T(k)]^T$$

noise measurement vector

$$g(k) = [X^T(k), V^T(k)]^T$$

data measurement vector

Then

$$g(k) = f(k) + e(k)$$

For convenience,  $f(k)$ ,  $e(k)$  and  $g(k)$  will be called model-measurement vector, noise-measurement vector and data-measurement vector, respectively. To emphasize, the model-measurement vector  $f(k)$  is obtained by passing the model sequences  $y(k)$  and  $u(k)$  through the measurement filters; the noise-measurement vector  $e(k)$  by passing the noise sequences  $q(k)$  and  $v(k)$  through the same filters; and the data-measurement vector by passing the data sequences  $x(k)$  and  $v(k)$  through the system of measurement filters.

Generalized Least-Squares Formulation of the Identification Problem

Given the data-measurement vectors  $g(k)$ ,  $k=1, \dots, K$  and the noise-measurement vector covariance

$$R = E \sum_{k=1}^K e(k) e^T(k) \quad (E: \text{ expected value operator})$$

find the synthetic parameter vector  $\lambda$  that minimizes

$$J = \sum_{k=1}^K [g(k) - f(k)]^T R^{-1} [g(k) - f(k)] \quad (8)$$

under the constraint

$$\lambda^T f(k) = 0 \quad (9)$$

Remark

(The reader may wish to skip this in the first reading)

If  $q(k)$  and  $w(k)$  are stationary white noise processes with variances  $\sigma_q^2$  and  $\sigma_w^2$  respectively and cross-correlation coefficient  $\rho$  (which could of course be zero) then

$$R = \sum_{k=1}^K (K-k+1) \begin{bmatrix} \tilde{Q}(k)\tilde{Q}^T(k)\sigma_q^2 & \rho\tilde{Q}(k)\tilde{W}^T(k)\sigma_q\sigma_w \\ \rho\tilde{W}(k)\tilde{Q}^T(k)\sigma_q\sigma_w & \tilde{W}(k)\tilde{W}^T(k)\sigma_w^2 \end{bmatrix} \quad (10)$$

Here,  $[\tilde{Q}(k), \tilde{W}(k)] = p(k)$  represents the measurement-vector sequences resulting from unit pulse ( $\delta_k = \{1, 0, 0, \dots\}$ ) stimuli at the measurement filter input terminals. We shall call  $p(k)$  the pulse-measurement vectors.

- Note that  $R$  has the form

$$R = \begin{bmatrix} R_{11}\sigma_q^2 & R_{12}\rho\sigma_q\sigma_w \\ R_{21}\rho\sigma_q\sigma_w & R_{22}\sigma_w^2 \end{bmatrix}$$

where the matrices  $R_{11}$ ,  $R_{12} = R_{21}^T$ , and  $R_{22}$  are known (without the knowledge of  $\sigma_q^2$ ,  $\sigma_w^2$ , and  $\rho$ ). They are determined entirely by the known measurement filters. When either  $\sigma_q$  or  $\sigma_w$  is zero, the matrix  $R$  becomes known up to a scalar multiple.

Solution of the Identification Problem

The solution  $\lambda$  and  $f(k)$  which minimize (8) under the constraint are obtained by the Lagrange multiplier method:

$$J^* = \sum_{k=1}^K \left\| g(k) - f(k) \right\|_{R^{-1}}^2 + \sum_{k=1}^K v_k (\lambda^T f(k)) \quad (11)$$

$$\frac{\partial J^*}{\partial f(k)} = -2R^{-1}(g(k) - f(k)) + v_k \lambda = 0 \quad (12a)$$

$$(g(k) - f(k)) = \frac{1}{2} v_k R \lambda \quad (12b)$$

$$\lambda^T (g(k) - f(k)) = \frac{1}{2} v_k \lambda^T R \lambda \quad (12c)$$

$$\frac{\partial J^*}{\partial v_k} = \lambda^T f(k) = 0 \quad (12d)$$

Equations (12c) and (12d) together yield

$$\begin{aligned} \lambda^T g(k) &= \frac{v_k}{2} \lambda^T R \lambda \\ \frac{v_k}{2} &= \frac{\lambda^T g(k)}{\lambda^T R \lambda} \end{aligned} \quad (12e)$$

Substitution of (12b), (12d) and (12e) gives the minima with respect to  $f(k)$  and  $v_k$ :

$$\begin{aligned} J^* &= \sum_{k=1}^K \frac{v_k}{2} (g(k) - f(k))^T \lambda \\ &= \sum_{k=1}^K \frac{v_k}{2} g^T(k) \lambda \\ &= \sum_{k=1}^K \frac{\lambda^T g(k) g^T(k) \lambda}{\lambda^T R \lambda} \end{aligned}$$

By defining the Gram matrix of the data-measurement vectors  $g(k)$

$$G = \sum_{k=1}^K g(k) g^T(k) \quad (13)$$

we may write

$$J^* = \frac{\lambda^T G \lambda}{\lambda^T R \lambda} \quad (14)$$

The problem now is to minimize (14) with respect to  $\lambda$ . This is quite readily shown to be the eigenvector solution of

$$(G - \mu R) \lambda = 0 \quad (15)$$

corresponding to the smallest eigenvalue  $\mu_1$ .

Furthermore, it turns out that  $J_{\text{minimum}} = \mu_1$  and  $f(k)$  are given by

$$f(k) = g(k) - \frac{\lambda^T g(k)}{\lambda^T R \lambda} R \lambda$$

### Remarks

- The actual solution of the eigenvector problem in (15) is contingent upon the form of the  $R$  matrix. Three different cases arise which are discussed in Appendix C.
- The standard deviations  $\sigma_q$  and  $\sigma_w$  of the output and input noise sequences are frequently unavailable. Under white noise assumption they can be estimated approximately as follows. Suppose that the useful signal frequencies are limited to the frequency band  $[0, f_1]$ , then by high-pass filtering one can estimate the power-density of the noise in the region  $f_1$  to  $f_1 + f_2$ ; call this density  $S(f_1)$ . Suppose the standard deviation of the high-pass filtered signal is  $\tilde{\sigma}$ , then the standard deviation of the original noise signal may be estimated as

$$\hat{\sigma}^2 = 2f_1 S(f_1) + \tilde{\sigma}^2 \quad \text{and} \quad S(f_1) = \tilde{\sigma}^2 / (2f_2)$$

- As mentioned earlier, a judicious choice of measurement filters can lead to rapid and successful identification of the vehicle transfer function. The primary consideration in this choice is that the resulting data-measurement sequences  $y_i(k)$  should be as linearly independent as possible (so that the cross-correlations between them are as small as possible). For, if the measurement sequences were highly correlated, the matrix  $G$  would become ill-conditioned. Correspondingly, the solution  $\hat{\lambda}$  would be unreliable. For example if the measurement filters were chosen to have  $Q_i \approx 0$  so as to have nearly all-pass characteristics (compared to the vehicle's critical frequencies) than the resulting measurement sequences have pair-wise correlations approximating unity.

Another interesting case worth mentioning is that when each measurement filter, instead of being chosen as a recursive first order digital filter, is replaced by a unit delay. In this event the formulation coincides with that considered by Levin [6]. If the sampling rate for discretizing the motion variables is adequately high, which is usually true, the corresponding sequences  $y_i(k) \equiv y(k-i)$  are highly correlated, rendering this choice undesirable [7]. Poor identification results therefore accrue.

- Since the measurement filters  $(1-Q_i)/(1-Q_i z^{-1})$  have low-pass frequency characteristics with unit d.c. gain, a convenient way to control the correlation between the resulting signals is to choose  $Q_i$  so that the mean power of measurement sequences diminish in a sensible manner. Call the mean power of the output of the  $i$ th filter as  $\bar{p}_i$ ; the  $Q_i$  could be selected so that

$$\bar{p}_i \approx \frac{n-i+1}{n+1} \bar{p}_0 \quad i = 1, 2, \dots, n$$

This choice of measurement filters has been implemented in the computer program GRAM as is available to the engineer on a select option basis

- As discussed earlier the minimum value achieved by the criterion function  $J$  is given by  $\hat{\mu}$ , the smallest eigenvalue of the equation posed in (15). This value will be called the algorithm error. However, since the engineer is interested really in the fidelity of output reconstruction a simple measure of fidelity may be used. Let  $\hat{y}(k)$  be the reconstructed signal obtained by processing the measured input  $v(k)$  by the estimated transfer function (the true input  $u(k)$  is used when simulation mode is used in the computer program GRAM). Then the percent reconstruction error is defined as

$$ESR = 100 \sqrt{\frac{\sum_{k=1}^K (y(k) - \hat{y}(k))^2}{\sum_{k=1}^K y^2(k)}}$$

## III. PROGRAM DESCRIPTION

The purpose of this FORTRAN program is to determine a linear model from flight-test data of a submerged vehicle. The method used is the 'GRAM Identifier' discussed in Section II. The computer program is designed to work under three different modes. When ISIM = 0 analysis of actual flight test data is performed; with ISIM = 1 and ISIM = 2 flight trajectories are first simulated and then identification is performed.

When ISIM is either one or two the input data is generated in subroutine FILLV where the type of control input is specified by the parameter INPT. The flight trajectory is simulated using the z-domain vehicle transfer function when ISIM = 1 (subroutine RESPON) or using the vehicle impulse response when ISIM = 2 (subroutine CONVOL). Once the flight trajectories are generated the program uses the facility of adding synthetic white Gaussian noise (subroutine CORUPT) to the simulated flight trajectories. Next the model identification is performed based upon the actual or simulated flight trajectory through the 'GRAM Identifier' method. The identified model is used to reconstruct a flight trajectory which is compared to the actual or simulated flight trajectory for analysis.

When ISIM = 1 it is possible also to simulate a feedback system as shown in Fig. 3 wherein the vehicle transfer function, the input function (via option parameter INPT), the compensator constants and the gain constant are specified to the program.

The last of these is not read via data cards; it is entered directly in the subroutine RESPON. Identification is then performed upon the vehicle input-output. For reconstruction one may use open-loop reconstruction or closed-loop reconstruction.

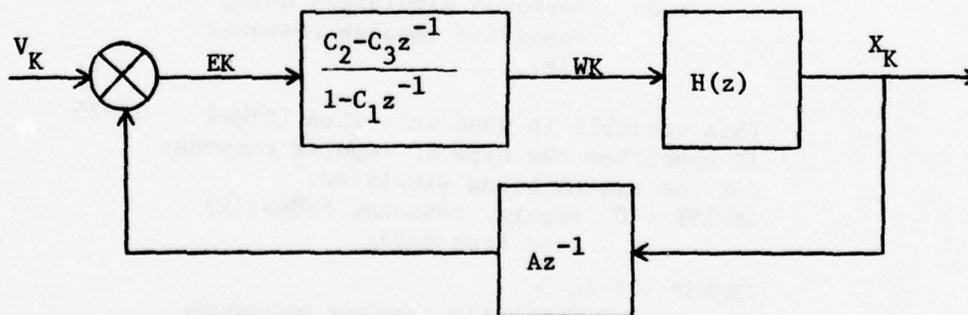


Fig. 3. Simulated feedback loop;  $C_1 \equiv \text{COMPS}(I)$ ,  $A = \text{GAIN}$

The input data cards on the subsequent pages give a description of all input variables, and in so doing provide an understanding of the program use.

INPUT DATA CARDS

CARD # 1                    The first card is a title card. Columns 1 through 80 are available for an alphanumeric title.

CARD # 2                    First option card which contains eight variables.

Variable Name (Format)	Description	Columns	Preferred Value (if any)
N (I5)	Order of system	1-5	-
MP1 (I5)	Number of data points	6-10	-
IPLT (I5)	Plotter option; IPLT = 0    No plots = 1    Plots only on line printer = 2    Plots on printer and CALCOMP plotter	11-15	1
ISIM (I5)	Simulation option ISIM = 0    Performs identification upon flight test data = 1    Performs simulation using z-domain transfer function coefficients = 2    Performs simulation using specified impulse response h(k).	16-20	-
IMRESP (I5)	This variable is used only when ISIM=2 it specifies the type of impulse response for the system being simulated. IMRESP = 0    Impulse response HPULSE(k) read from cards  IMRESP = 1 to 5 Synthetic impulse responses generated (see subroutine CONVOL)	21-25	-
NPULSE (I5)	Number of impulse response points	26-30	-



INPUT = 5 Square wave followed by exponential  
 = 6 Exponential  
 = 7 Periodic impulse  
 = 8 Triangular wave  
 = 9 Exponential + decaying sinusoid  
 = 10 Random noise  
 = 11-20 shifted functions 1-10

IREM (I2) Option used to describe the order of the numerator compared to the denominator. This parameter controls the numerator of the z-domain model transfer function. Specifically, it limits the numerator degree in  $z^{-1}$  to N-IREM. For example if IREM = 1, then the model seeks a numerator.

$$\text{NUM}(z) = b_1 + b_2 z^{-1} + \dots + b_n z^{-(n-1)}$$

IZTS (I2) This option determines the type of z domain to s domain transformation that is performed.

IZTS = 0 Z domain to S domain conversion is not performed. That is an equivalent continuous time system is not found

= 1 An equivalent continuous time system is found (from the discrete time transfer function  $\hat{H}(z)$  based on a logarithmic z to s transformation).

= 2 An equivalent continuous time system is found (from the discrete time transfer function  $H(z)$  based on a pulse delayed z to s transformation).

QOPT (I2) Option used to determine measurement filter pole(s). If QOPT = 0 each of the measurement filter poles is set equal to the data value read as QSAV. If QOPT=1 the measurement filter poles are calculated in subroutine FINDQ.

FDBACK (I2)	This option allows a negative feedback path to be added to simulate a feedback system for which the vehicle is the plant.  FDBACK = 0 No feedback = 1 Feedback is simulated	9-10	-
FDREC	FDREC is the variable to determine the type of reconstruction desired.  FDREC = 0 Open loop reconstruction = 1 Closed loop reconstruction  (Note FDREC must equal zero if FDBACK equals zero)	11-12	-
ILEVIN	This option is used when the LEVIN identification technique is desired.  ILEVIN = 0 Gram identification technique performed.  = 1 Levin identification technique performed.	13-14	0
IDLX	Delay introduced on input numerator  $\text{NUM}(z) = z^{-\text{IDLX}}(b_1 + \dots + b_{(n+1-\text{IREM})} * z^{-(N-\text{IREM})})$	15-16	0
N (I5)	Order of model	21-25	-
MP1 (I5)	Number of data points	26-30	-
NPRD (I5)	Time scale parameter for input signal (useful only when ISIM = 1 or 2)	31-35	-
ISKIP (I5)	This variable determines the sequence of points plotted on the printer. If ISKIP = 1 every data point is plotted and if ISKIP = 5 every fifth point is plotted, etc.	36-40	-
DELTA (F5.0)	Sampling interval	41-45	-

QSAV (F5.0)	QSAV is the measurement filter pole(s) used only if QOPT = 0 and disregarded if QOPT = 1.	46-50	0.8 to 0.95
NSPQ (F5.0)	Noise to signal <u>power</u> ratio of the output sequence (used only if ISIM = 1 or 2).	51-55	0.01
NSPW (F5.0)	Noise to signal <u>power</u> ratio of the input sequence (used only if ISIM = 1 or 2)	56-60	0.01

CARD # IX + 2

This card contains the coefficients of a first order compensator in the forward path preceding the vehicle. The general form of the compensator is;

$$C(z) = \frac{\text{COMPS}(2) - \text{COMPS}(3)z^{-1}}{1 - \text{COMPS}(1)z^{-1}} \quad \left( \begin{array}{l} \approx \frac{cs+b}{s+a} \\ - \end{array} \right)$$

The coefficients are read in the form of 3F10.1 fields. If no compensation is desired a blank card should be inserted in this position.

END OF FILE CARD

// Card on IBM 360 system.

## MEMORY

The total storage required for the program is 152K bytes, or approximately 40K words, on an IBM 360/75 computer system. This will, of course, change if the array dimensions are changed to meet the users test requirements. Presently the program can accept up to one thousand data points each for the input and output signals. The model sought (or the simulated model entered when ISIM = 1) can be as high as ninth order. Under the third simulation mode (ISIM = 2) the impulse response can be of a length up to sixty four data points. It is important to note that the square matrices G and Z and the vectors GAMMA, XLAMDA, and COEFF should have a dimension at least as large as (N+N+2) where N is the order of the model.

## OUTPUT

The first line of output is the title which is followed by a column of program variables as follows:

## STARTING SIMULATION

```
SYSTEM ORDER =      4
M + 1 =      500
```

```
INPT =      9
IREM =      1
IZTS =      2
```

```
NSPQ =      0.010000
NSPW =      0.010000
```

```
SAMPLING INTERVAL =      0.500000
Q PARAMETER =      0.900000
```

```
QOPT =      1
IPLT =      1
FDBACK=      0
FDREC=      0
ILLVIN=      0
```

```
IDLY =      1
INORM=      1
COMPS(I) =      .0
```

The next portion of the output is a plot of the vehicle input and output. On the left hand side three columns of printout list the serial number of the data point, the instantaneous value of the output and the instantaneous value of the input respectively. In case a feedback loop is employed (i.e. FDBACK = 1) then the command input (i.e. the input to the feedback system) is plotted preceding the vehicle input-output plot. Next the following title is printed

## GRAM IDENTIFIER

The next output line lists the values of measurement filter poles (Q(I)). Note that N measurement filters are used where N is the system order. All Q(I) are equal if QOPT = 0.

Following the Q parameters is the listing of the gram matrix G. This is an NPNP2 x NPNP2 matrix where NPNP2 = N+N+2. The item printed next is the noise correction matrix Z which is generated in the subroutine BUILDZ. This matrix is also NPNP2 x NPNP2 dimensional. The next line of printout is the Synthetic Coefficient Vector XLAMDA which is generated using the subroutines SOLVE1, 2, and 3. Following the Synthetic Coefficient Vector is the transformation matrix A (again an NPNP2 x NPNP2 matrix) which is premultiplied with XLAMDA to obtain the desired parameter vector GAMMA. It is generated in the

subroutine BUILDA. The next value printed is the estimate of bias in data followed by the variable NN. At this point in the output the following are printed

- a) the z-domain denominator, numerator and poles,
- b) the s-domain poles, numerator constants of a partial fraction expansion (available only when IZTS = 2 or 3), the denominator and numerator.

The values of the denominator and numerator coefficients for both the Z and S-domain are printed in ascending order of the degree of the term it multiplies, starting with the constant term and ascending to the appropriate highest order term. At this point the reconstruction from the model is obtained via RESPON. The subroutine ERROR calculates and prints the reconstruction error

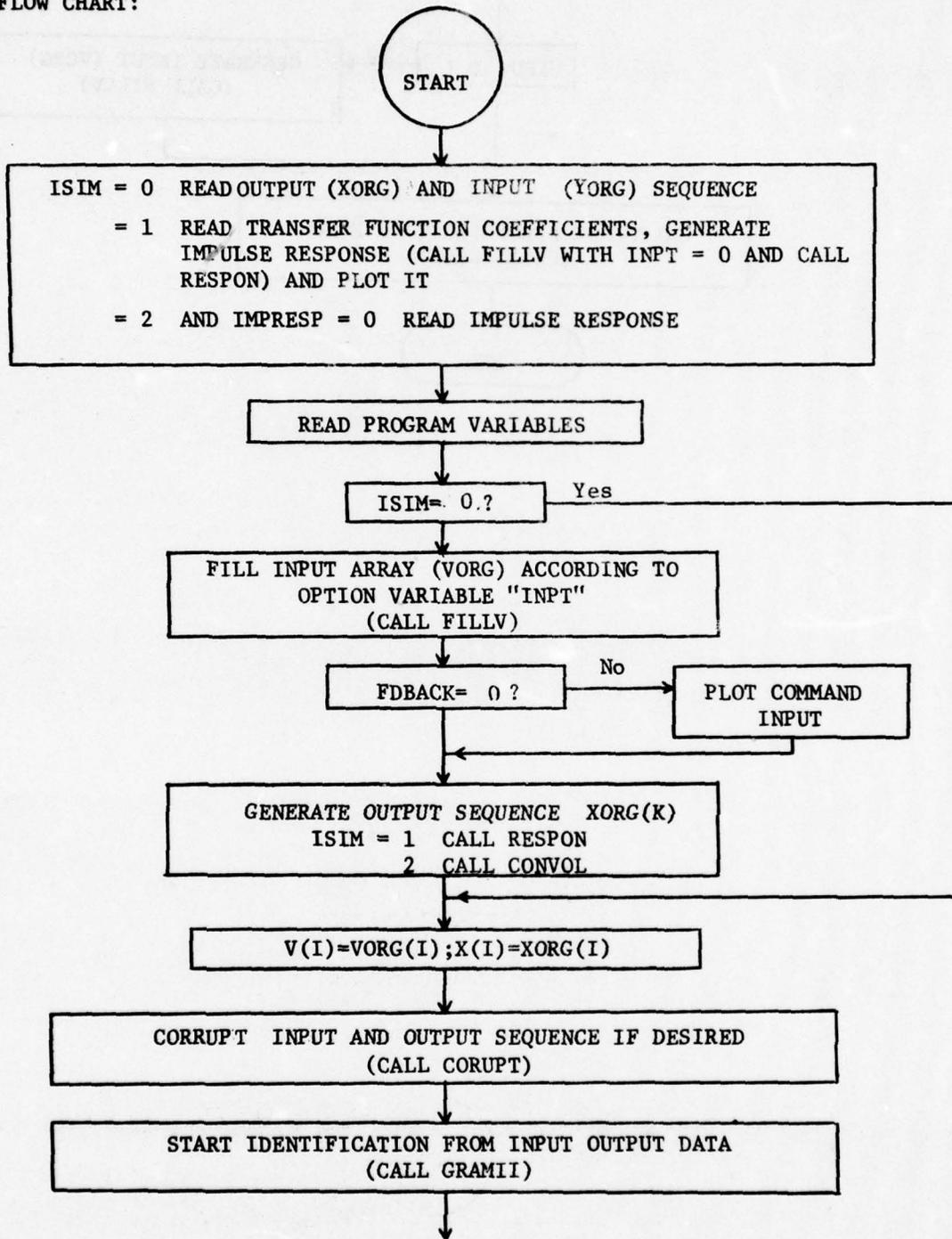
PER CENT MEAN POWER ERROR OF RECONSTRUCTION 0.000

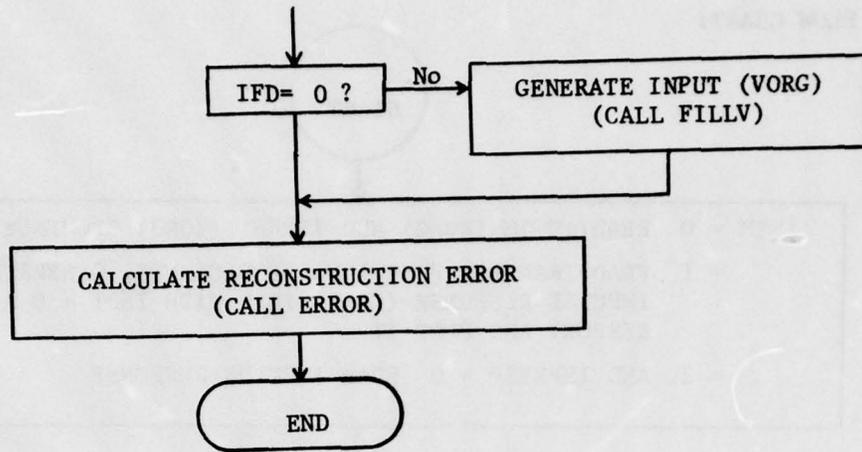
PER CENT OF SQUARE ROOT OF POWER ERROR IN RECONSTRUCTION 0.002

The last output is the plot of the true response (when ISIM = 1 or 2) or actual flight test data when ISIM = 0 and the reconstructed response. The same format is used as the previous one for the vehicle input output plot.

PROGRAM GRAM

FLOW CHART:





```

C   PROGRAM GRAM
      DIMENSION X(1000),V(1000),XORG(1000),VORG(1000),XREC(1000)
      DIMENSION DATA(1000,2),DATA2(1000,2),BUFF(3000)
      DIMENSION G(20,20),Z(20,20),GAMMA(20),XLAMDA(20),COEFF(20)
      DIMENSION HPULSE(64)
      DIMENSION TITLE(80)
      COMMON NN
      REAL*8 G,Z,GAMMA,XLAMDA,COEFF,COMPS
      REAL*8 DELTA,Q,QSAV,DELSAV,AVGQ,AVGW,SUMV2,XSAV
      REAL NSPQ,NSPW
      INTEGER QOPT,FDBACK,FDREC
      EQUIVALENCE (Z(1,1),BUFF(1)),(G(1,1),BUFF(1501))
      EQUIVALENCE (DATA(1,1),X(1)),(DATA(1,2),V(1))
      EQUIVALENCE (XORG(1),DATA2(1,1)),(XREC(1),DATA2(1,2))
      EQUIVALENCE (NSPQ,SIGQ),(NSPW,SIGW)
      COMMON /COMPEN/COMPS(10)
      COMMON /GKRD/IGKR

C
      WRITE(6,1022)
      READ(5,1021)TITLE
      WRITE(6,1021)TITLE
      WRITE(6,1023)
      MAXPL=1000
      MAX=20
4320  READ(5,1001)  N,MP1,IPLT,ISIM,IMRESP,NPULSE,INORM
      NPNP2=N+N+2
      RDEL=0.01
      IF(ISIM.EQ.0)READ(5,6995)(XORG(K),K=1,MP1),(VORG(K),K=1,MP1)
      IF(ISIM.EQ.2.AND.IMRESP.EQ.0)READ(5,6995)(HPULSE(K),K=1,NPULSE)
      IF(ISIM.NE.1)GO TO 6622
C     READ DIFFERENCE EQUATION PARAMETERS
      READ(5,701,END=1234)(COEFF(I),I=1,NPNP2)
      CALL FILLV(VORG,MP1,0,NPRD)
      CALL RESPON(X,VORG,N,COEFF,XLAMDA,MP1,0)
      IF(IPLT.NE.2)GO TO 6622
      CALL PLOP8(MP1,1,X,MAXPL,C.0,RDEL,3,
117HIMPULSE RESPONSE_,
216HTIME IN SECONDS_,BUFF)
6622  CONTINUE
      KKKK=0

C
C
4321  READ(5,1,END=1234)INPT,IREM,IZTS,QOPT,FDBACK,FDREC,ILEVIN,IDLY,
      INDEK,MP1,NPRD,ISKIP,DELTA,QSAV,NSPQ,NSPW
      READ(5,6995)(COMPS(I),I=1,3)
      IF(N.EQ.10000)GO TO 4320
      Q=QSAV
      WRITE(6,1000)N,MP1,INPT,IREM,IZTS,NSPQ,NSPW,DELTA,Q,QOPT,
1IPLT,FDBACK,FDREC,ILEVIN,IGKR,IDLY,INORM,  LCOMPS(I),I=1,3)
      NM1=N-1
      NP1=N+1
      NP2=N+2
      NPNP1=N+N+1
      NPNP2=N+N+2
      RHO=0.0
      NN=N-IREM

```

```

IF(ISIM.EQ.0)GO TO 23
C
C      FILLING THE INPUT ARRAY ACCORDING TO OPTION PARAMETER, INPT
CALL FILLV(VORG,MP1,INPT,NPRD)
IF(FDBACK.EQ.0)GO TO 6626
D022I=1,MP1
V(I)=VORG(I)
22  X(I)=0.0
    IF(KKKK.NE.0) GO TO 6616
CALL PLOTIT(DATA ,2,MP1,1,MP1,ISKIP,MAXPL,1,1.0)
6616 CONTINUE
    IF(IPLT.NE.2 .OR. KKKK.GT.0) GO TO 6626
CALL PLOP8(MP1,2,DATA ,MAXPL,0.0,RDEL,3,
125HINPUT TO FEEDBACK SYSTEM_,
216HTIME IN SECONDS_,BUFF)
6626 CONTINUE
C
C      GENERATING SEQUENCE X(K)
IF(ISIM.EQ.1)CALL RESPON(XORG,VORG,N,COEFF,XLAMDA,MP1,FDBACK)
IF(ISIM.EQ.2)CALL CONVCL(HPULSE,VORG,XORG,NPULSE,MP1,IMRESP)
23  D024I=1,MP1
    V(I)=VORG(I)
24  X(I)=XORG(I)
    IF((NSPQ+NSPW)*ISIM .NE.0)CALL CORUPT(X,V,SIGQ,SIGW,MP1)
    IF(KKKK.NE.0) GO TO 6611
WRITE(6,1003)
CALL PLOTIT(DATA ,2,MP1,1,MP1,ISKIP,MAXPL,1,1.0)
6611 CONTINUE
    IF(IPLT.NE.2 .OR. KKKK.GT.0) GO TO 6633
CALL PLOP8(MP1,2,DATA ,MAXPL,0.0,RDEL,3,
127HCORRUPTED INPUT AND OUTPUT_,
216HTIME IN SECONDS_,BUFF)
6633 CONTINUE
    N=NDEN
C      START IDENTIFICATION FROM INPUT OUTPUT DATA
C
CALL GRAMII(X,V,MP1,SIGQ,SIGW,RHC,N,DELTA,Q,QOPT,IREF,I2TS,GAMMA,
IXLAMDA,G,Z,MAX,ILEVIN,IDLY,INORM)
IFD=0
IF(FDBACK.GE.1 .AND. FDREC.EQ.1) IFD=FDBACK
IF(IFD.NE.0) CALL FILLV(VORG,MP1,INPT,NPRD)
CALL ERROR(XREC,VORG,GAMMA,MP1,N,XLAMDA,XORG,IFD,IDLY)
C
C      PLOT RECONSTRUCTION
WRITE(6,66)
WRITE(6,1004)
IF(IPLT.EQ.0) GO TO 6544
CALL PLOTIT(DATA2,2,MP1,1,MP1,ISKIP,MAXPL,1,1.0)
6544 CONTINUE
99  RDEL=DELTA
    IF(IPLT.NE. 2) GO TO 6644
CALL PLOP8(MP1,2,DATA2,MAXPL,0.0,RDEL,3,
140HTRUE RESPONSE VS RECONSTRUCTED RESPONSE_,
216HTIME IN SECONDS_,BUFF)
6644 CONTINUE
C
    KKKK=K+1
100 GO TO 4321
1234 CALL PICSIZ(0,0,0.0)
1022 FORMAT(////////////////////)
1021 FORMAT(80A1)

```





## IV. APPLICATION EXAMPLES

## Example 1

For a six-man submersible the transfer functions describing its dynamics were obtained from the hydrodynamic coefficients via a computer program RGEORGE. The pitch vs. stern-plane dynamics will be used here to demonstrate the application of GRAM Identifier (ILEVIN = 0 in the program). Specifically the transfer function relating these variables is

$$\begin{aligned} H_4(s) &= \frac{-0.34320s^2 - 0.17384s - 0.008631}{s^4 + 1.47989s^3 + 0.11833s^2 + 0.02048s + 0.00102} \\ &= \frac{-0.08348 + j0.48366}{s + 0.00919 + j0.11378} + \frac{-0.08348 - j0.48366}{s + 0.00919 - j0.11378} \\ &\quad + \frac{0.16696}{s + 1.4057} + \frac{0.00002}{s + 0.05579} \end{aligned}$$

For all practical purposes this is seen to be equivalent to

$$\begin{aligned} H_4(s) = H_3(s) &= \frac{-0.34320s - 0.15469}{s^3 + 1.58950s^2 + 0.26054s + 0.00306} \\ &= \frac{-0.08349 + j0.48367}{s + 0.00919 + j0.11378} + \frac{-0.08349 - j0.48367}{s + 0.00919 - j0.11378} + \frac{0.16696}{s + 1.4057} \end{aligned}$$

In this third order function the energy of the complex pole pair is 26.9 while the energy associated with the real pole is 0.01; the latter thus represents only 0.037% of the energy at the dominant complex pole pair. Therefore, unless the input is such that its spectral content is rich in radian frequencies around 1.4, this mode will be extremely feeble. We will in fact call this mode a micromode [3]. When this micromode is not appropriately excited the vehicle transfer function may be approximated as

$$\begin{aligned} H_4(s) \approx H_2(s) &= \frac{-0.16698s + 0.10853}{s^2 + 0.1838s + 0.01303} \\ &= \frac{-0.08349 + j0.48367}{s + 0.00919 + j0.11378} + \frac{-0.08349 - j0.48367}{s + 0.00919 - j0.11378} \end{aligned}$$

Before describing the various experiments conducted, the z-domain description of the pitch transfer function is first provided. The transfer function  $H_4(s)$  was transformed by the Leading-Edge-Pulse Equivalence method (Appendix B). A sampling interval  $\Delta = 0.5$  second was used yielding

$$H_4(z) = \frac{(10^{-4})(0.20744z^{-1} - 0.19065z^{-2} - 0.15179z^{-3} + 0.13714z^{-4})}{1 - 3.45527z^{-1} + 4.38952z^{-2} - 2.41135z^{-3} + 0.47714z^{-4}}$$

In all of the five experiments performed 250 seconds of simulated data (MP1 = 500) were used. This represents approximately 2 1/2 time constants of the dominant mode.

#### Experiment 1

The function  $H_4(z)$  is employed to simulate the discrete-time trajectory  $\theta(k)$  to a given stern-plane input  $\delta_s(k)$  (INPT = 9, NPUL = 100). As stated earlier  $\theta(s)/\delta_s(s)$  is effectively a third order function, however, a fourth order identification was first performed without masking the data with noise (NSPW = NSPQ = 0.0). The identification yielded the following  $\hat{H}_4(s)$  with the option variables chosen as ISIM = 1, IREM = 1, IDLY = 1, QOPT = 1 and IZTS = 2:

$$\hat{H}_4(s) = \frac{-0.34323s^2 - 0.17394s - 0.00867}{s^4 + 1.48029s^2 + 0.11867s^2 + 0.02049s + 0.00103}$$

ESR = 0.000

(see page 12)

Clearly this identification is good since the model found is almost identical to the given  $H_4(s)$ . A comparison of the poles of the given  $H_4(s)$  and the identified poles is shown below.

$H_4(s)$ Poles	Identified Poles
-0.00919 + j0.11378	-0.00919 + j0.11378
-0.00919 - j0.11378	-0.00919 - j0.11378
-1.4057	-1.4058
-0.05579	-0.05603

Although the vehicle transfer function was identified perfectly, any quick conclusions as to the effectiveness of the identification method are misleading. Because even the slightest amount of noise on the data will mask the micro-micro-mode  $\left(\frac{0.00002}{s+0.05579}\right)$  making its identification impossible. Therefore the remaining experiments will pertain to third order identification except the last one. The latter is a second order run demonstrating the detection of the dominant pole pair.

#### Experiment 2

This run is identical to the previous one except that a third order model was sought (N = 3). The transfer function found was

$$\hat{H}_3(s) = \frac{-0.00435s^2 - 0.33603s - 0.14986}{s^3 + 1.38026s^2 + 0.03806s + 0.01774}$$

ESR = 0.447

The identified poles are compared to the true poles below

$H_3(s)$ Poles	Identified Poles
-0.00919 + j0.11378	-0.00919 + j0.11378
-0.00919 - j0.11378	-0.00919 - j0.11378
-1.4057	-1.36187

### Experiment 3

Ten percent rms noise was added to both the input and output data ( $NSPW = NSPQ = (0.1)^2$ ). A third order identification was performed using an input generated via option  $INPT = 9$ ,  $NPUL = 100$ . The test failed due to the extremely poor spectral content of the input. Specifically the input signal did not have sufficient energy at radian frequencies around 1.4, consequently it did not properly excite the pole at that location. A different input was therefore used ( $INPT = 5$ ,  $NPUL = 10$ ,  $QOPT = 1$ ). The input and output signals are shown in Fig. 4a. The corresponding results yielded by GRAM are as follows:

$$\hat{H}_3(s) = \frac{-0.50521s^2 - 2.75286s - 2.5146}{s^3 + 1.76398s^2 + 0.04714s + 0.02249}$$

$$ESR = 5.275$$

A comparison of the identified poles to the true poles is given below.

$H_3(s)$ Poles	Identified Poles
-0.00919 + j0.11378	-0.00982 + j0.11312
-0.00919 - j0.11378	-0.00982 - j0.11312
-1.4057	-1.74435

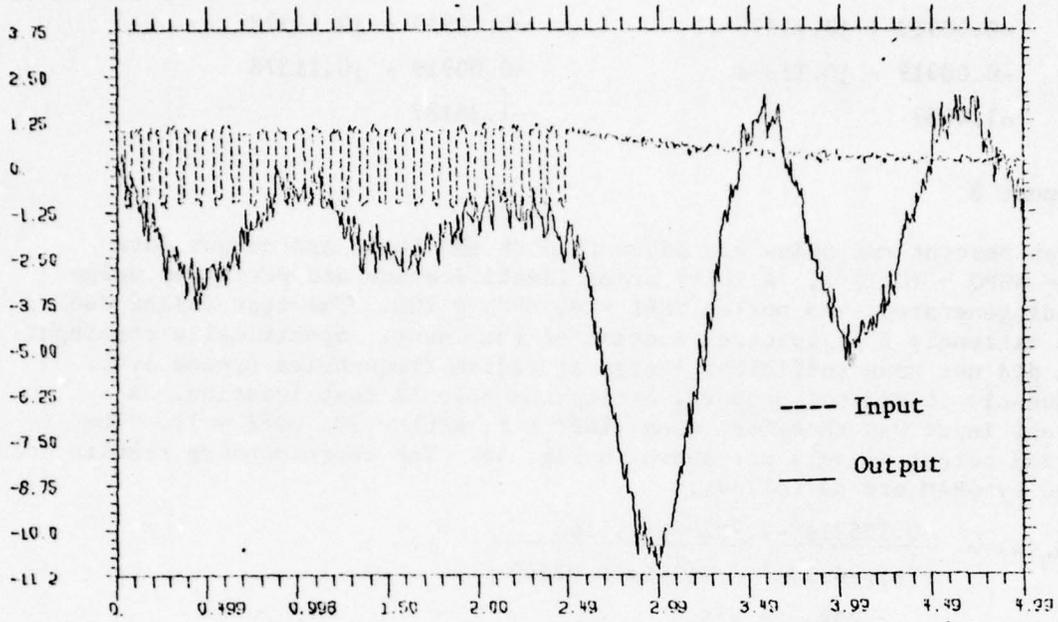
This input signal contained sufficient energy around the radian frequency 1.4 to excite the micro-mode just enough for identification purposes. Figure 4b shows the reconstructed output comparing it to the true output.

### Experiment 4

This experiment demonstrates the importance of the choice of the measurement filter pole(s). The measurement filter pole  $QSAV$  was varied ( $QOPT = 0$  to disable automatic filter pole selection) and its effect studied on the identification algorithm. Each of the runs performed uses the options  $INPT = 5$ ,  $NPUL = 10$ ,  $IREM = 1$  and  $IDLY = 0$  seeking a third order model. Ten percent noise was added of the same manner as in Experiment 3.

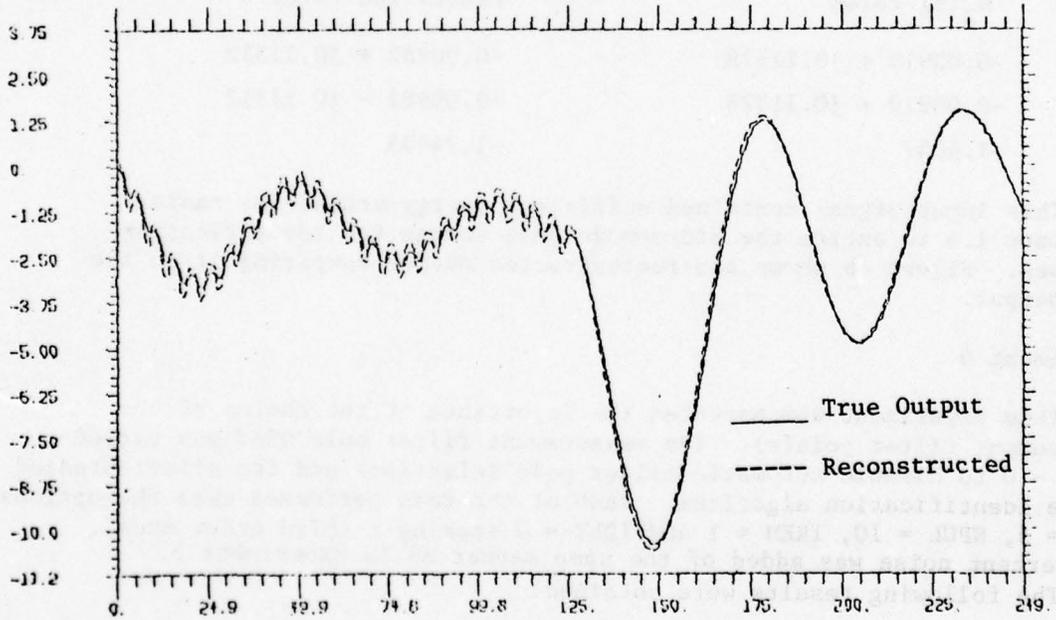
The following results were obtained:

CORRUPTED INPUT AND OUTPUT



(a)

TRUE RESPONSE VS RECONSTRUCTED RESPONSE



(b)

Figure 4. Experiment 3 - Third Order pitch vs Stern-Plane Identification

QSAV	ESR (Percent Error to Signal ratio RMS)
0.70	$\infty$
0.80	33.986
0.85	14.374
0.90	3.947
0.95	4.556
0.98	5.696

This example should make the user aware of the flexibility provided to the test engineer by the measurement filter pole(s). The value of the measurement filter pole should be such that each successive measurement filter attenuates the input signal by a reasonable fraction; in particular the output of the last measurement filter should not be an order of magnitude lower in power than the input signal to the first measurement filter. More information on the choice of measurement filters is available in reference [2].

#### Experiment 5

This final experiment demonstrates the detection of the dominant pole pair. The input (INPT = 6, NPUL = 200) and output signals were masked with 10% rms noise (NSPW = NSPQ =  $(0.1)^2$ ) and the following option parameters were used: QSAV = 0.95 IREM = 1, IDLY = 0. The input used (INPT = 6) is an exponentially decaying function whose time constant, and therefore the cutoff of the power spectrum curve, is controlled by the value of NPUL (Time constant = NPUL \* DELTA). Therefore to identify the slow poles ( $-0.00919 \pm j0.11378$ ) an input rich in radian frequencies around 0.114 is desired. A value of NPUL = 200 produces an adequate power spectrum. The input and output signals are shown in Fig. 5a. The computer program identified the second order mode as follows

$$\hat{H}_2(s) = \frac{0.02308s + 0.09245}{s^2 + 0.01904s + 0.01318}$$

ESR = 2.641

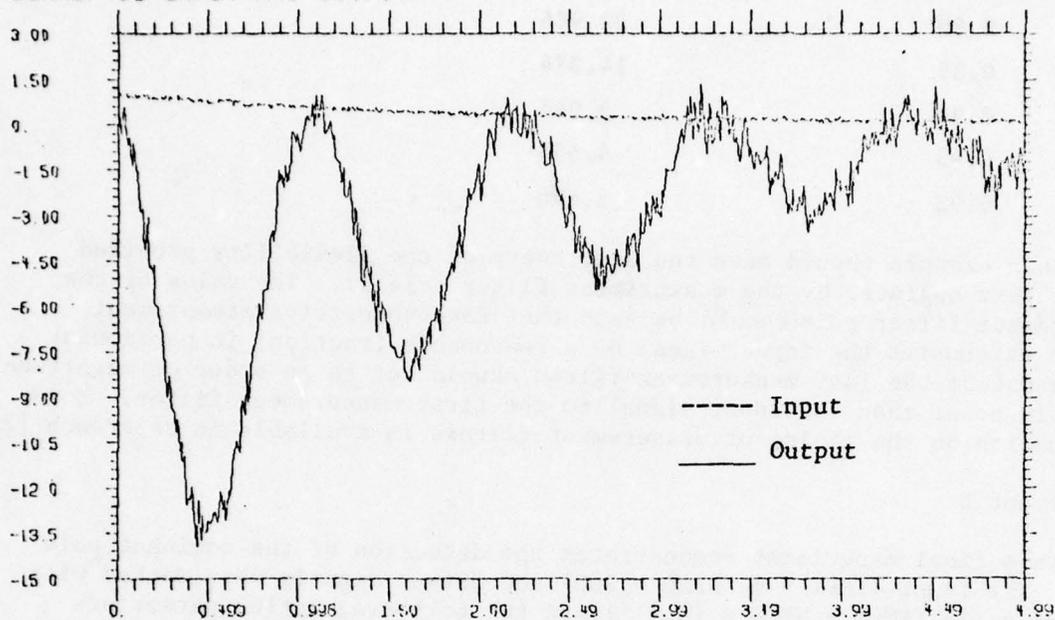
A comparison of the identified poles and the true poles is given below.

$H_2(s)$ Poles	Identified Poles
$-0.00919 + j0.11378$	$-0.00952 + j0.11439$
$-0.00919 - j0.11378$	$-0.00952 - j0.11439$

Figure 5b shows the reconstructed output comparing it to the true output.

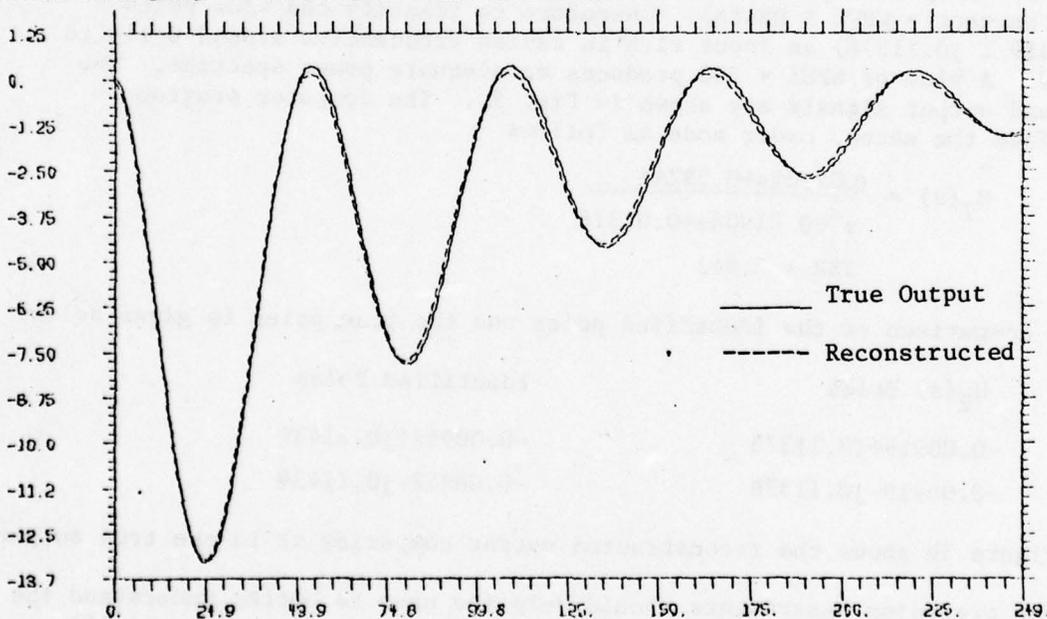
The preceding experiments should help the user to better understand the significance of the option variables (originally defined on pages 14-18).

CORRUPTED INPUT AND OUTPUT



(a)

TRUE RESPONSE VS RECONSTRUCTED RESPONSE



(b)

Figure 5. Experiment 5 - Detection of Dominant Complex Poles.

## Example 2

Flight tests on a towed-sonar vehicle (SMS2619--- one-third scale model), designed by the Naval Coastal Systems Laboratory, were conducted at the Naval Ship Research and Development Center, Carderock, Md. The test data were recorded on magnetic tapes at a sampling rate of 30Hz. In this example the results of two experiments are presented, one pertaining to pitch vs. stern-plane deflection and the other pertaining to roll vs. rudder deflection. In both cases the data were preprocessed by a 2Hz digital filter to substantially remove an undesirable 3.3Hz oscillation, suspected to be caused by an artifact in instrumentation. The data were then sampled at 5Hz (Nyquist frequency = 2.5Hz  $\approx$  15.71 radian/sec) for use by the identification program.

For the pitch vs. stern-plane data a second order (N=2) identification was performed which yielded the model

$$\hat{H}_2 = \theta(s)/\delta_s(s) = \frac{0.784(s + 0.028)}{(s + 1.21)(s + 0.03)}$$

The reconstructed output is shown in Fig. 6 together with the output data used for identification. It is worth mentioning that higher order models were also attempted, however, the additional poles found had insignificant energies associated with them.

For the roll vs. rudder data the results of a sixth order (N = 6) identification are presented. The model found is

$$\begin{aligned} H_6(s) &= \frac{-0.09682s^5 + 1.01434s^4 + 0.19137s^3 + 0.04870s^2 + 0.00132s + 0.00013}{s^6 + 0.67744s^5 + 0.64971s^4 + 0.12743s^3 + 0.02647s^2 + 0.00078s + 0.00007} \\ &= \frac{-0.35521 \pm j0.79885}{s + 0.23747 \pm j0.66789} + \frac{-0.00106 \pm j0.00045}{s + 0.01006 \pm j0.05316} + \frac{-0.01183 \pm 0.01648}{s + 0.09119 \pm j0.19015} \end{aligned}$$

The reconstructed output is shown in Fig. 7 together with the output data. However, the energies associated with the last two pairs of poles are quite small so that a second order identification would therefore seem desirable. In our analysis of the SMS vehicle we also conducted multiinput-multioutput identification, and such analysis showed that the lateral dynamics is essentially governed by two pole-pairs, one pole pair reasonably observable in the roll data and the other in yaw-rate data. The multiinput, multioutput version of GRAM identifier is discussed in [9].

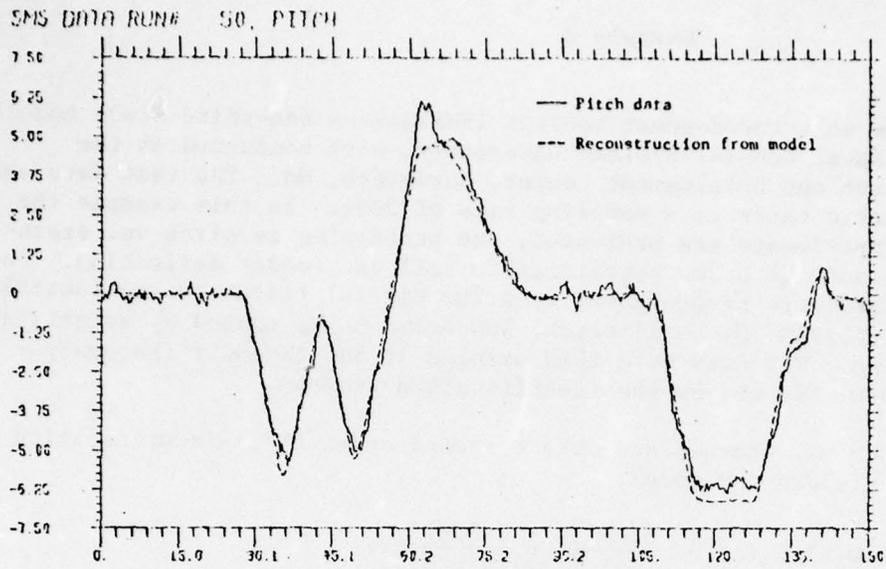


Fig. 6 Identification of longitudinal dynamics of SMS2619 vehicle; reconstruction of pitch from identified model.

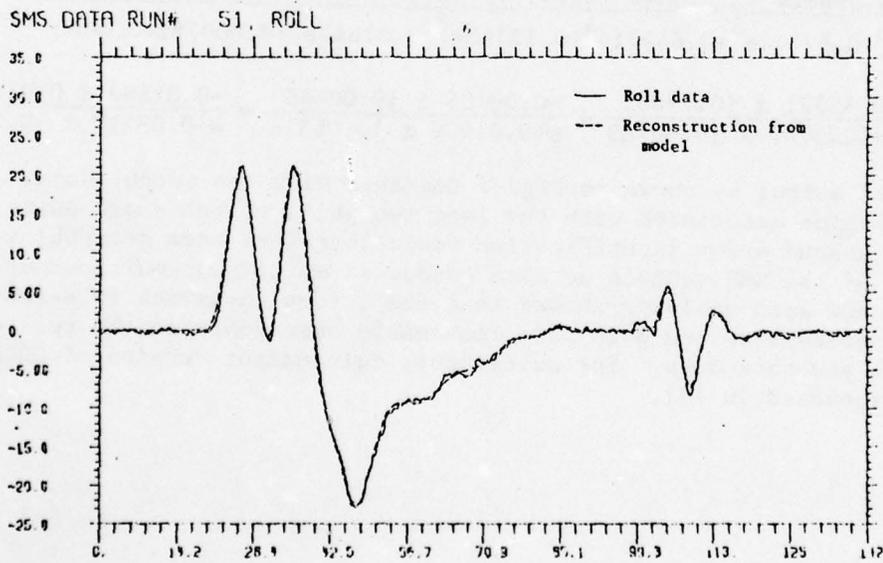


Fig. 7. Identification of lateral dynamics of SMS2619 vehicle; reconstruction of roll from identified model.

REFERENCES

- [1] D. E. Humphreys, "Development of the Equations of Motion and Transfer Functions for Underwater Vehicles", Naval Coastal Systems Laboratory Report TR-287-76, July 1976.
- [2] G. J. Dobeck, System Identification and Application to Undersea Vehicles, Ph.D. Dissertation, College of Engineering, University of South Florida, 1976.
- [3] V. K. Jain, "Extraction of Vehicle Transfer Functions from Noisy Flight Test Data via a Discrete Decoupled Technique (Phase I)", Engineering Research Report, University of South Florida, (submitted to NCSL in November 1974), 1974.
- [4] V. K. Jain, "Extraction of Vehicle Transfer Functions from Noisy Flight Test Data via a Discrete Decoupled Technique (Phase II)", Engineering Research Report, University of South Florida, (submitted to NCSL in August 1975), 1975.
- [5] H. Freeman, Discrete Time Systems, New York; John Wiley, 1965.
- [6] M. J. Levin, "Estimation of a System Pulse Transfer Function in the Presence of Noise", IEEE Trans. A.C., vol. AC-9, 229-235, July, 1964.
- [7] G. J. Dobeck, System and Signal Identification by Digital Methods, M.S. Thesis, University of South Florida, Tampa, Florida, 1973.
- [8] M. Nichols, "On the z-domain Description of Navy Vehicles", correspondence to Naval Coastal System Laboratory, Panama City, FL, July 1976.
- [9] V. K. Jain and G. J. Dobeck, "Multiinput-Multioutput Identification of Vehicle Dynamics via Gram Identifies", Research report in Preparation.

APPENDIX A

SUBROUTINES USED IN PROGRAM

The following subroutines used by the GRAM Identifier program are described in the appendix.

BUILDA  
 BUILDZ  
 CONVOL  
 CORUPT  
 ERROR  
 FILLV  
 FINDQ  
 GRAMII  
 IZTOS  
 POLCON  
 PRCVEC  
 PRMAT  
 PRVEC  
 RESPON  
 SOLVE1  
 SOLVE2  
 SOLVE3  
 ZTOS

The subroutines DOBINV, PLOP8 and POLRT are not detailed. Their function is indicated below and they can be substituted by standard routines from a scientific package.

DOBINV	Inversion of a square matrix
PLOP8	X-Y plotter (CALCOMP) routine The subroutines called by PLOP8 are also not discussed here
POLRT	Computes the real and complex roots of a real polynomial

SUBROUTINE: BUILD A

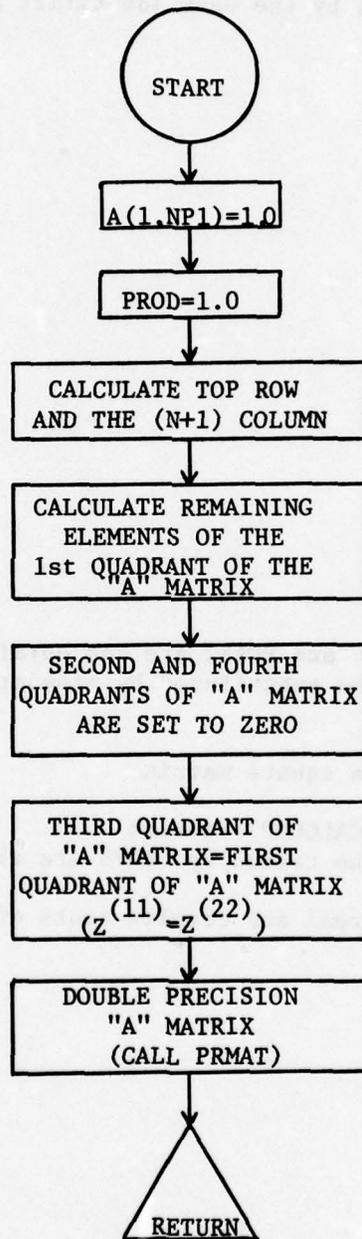
PURPOSE: CALCULATES THE TRANSFORMATION MATRIX TO CONVERT SYNTHETIC  
PARAMETER VECTOR TO THE DESIRED PARAMETER VECTOR

EQUATIONS:  $A_{ij} = A_{i,j+1} - \frac{Q_j A_{i-1,j+1}}{P_j}$  after 1st row and (n+1)<sup>th</sup> column

"A"'s are generated by other means with  $i = 2,3,\dots, n+1$ ;

$j=1,n-1,\dots,1$

FLOW CHART:



SUBROUTINE: BUILD A

DESCRIPTION: The "A" Matrix is formed in the following manner

$$A = \begin{bmatrix} \Gamma & 0 \\ 0 & \Gamma \end{bmatrix}$$

where each quadrant is an  $(n+1) \times (n+1)$  square matrix.

This matrix embodies the relationship between the synthetic coefficient vector XLAMDA derived from the GRAM matrix and the true coefficient vector of the systems transfer function, GAMMA. This relationship is dependent upon the values used for the measurement filters. Multiplying XLAMDA by a matrix of order  $2(n+1)$  with  $\Gamma$  in the upper left and lower right quadrants, yields GAMMA.

PROGRAM VARIABLES:	A	"A" MATRIX
	DEL	MEASUREMENT FILTER NUMERATOR
	MAX	MAXIMUM ROWS PERMISSIBLE
	N	ORDER OF SYSTEM
	Q	MEASUREMENT FILTER POLE(S)

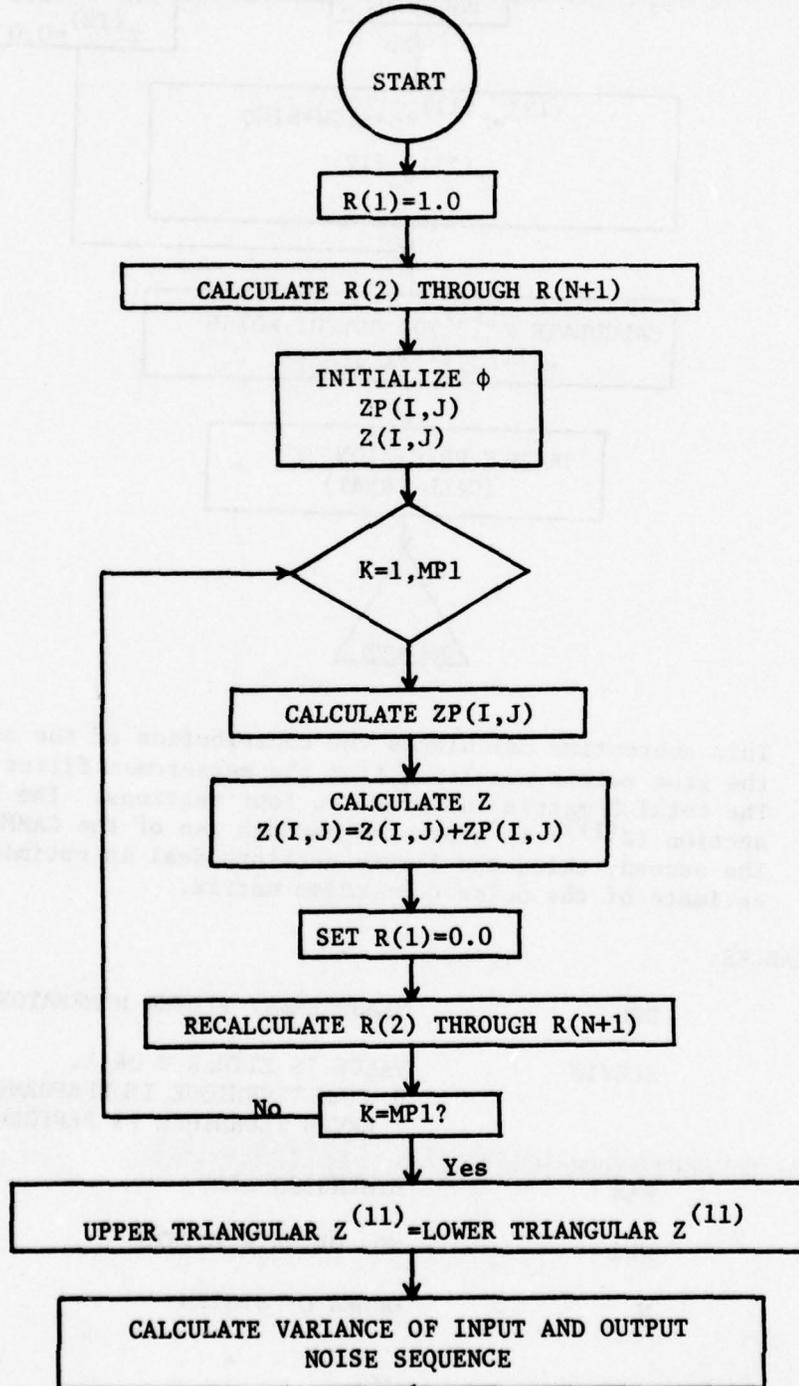
```
SUBROUTINE BUILDA(A,Q,DEL,N,MAX)
REAL*8 A(MAX,1),Q(1),DEL(1),PROD
NP1=N+1
NPNP2=N+N+2
A(1,NP1)=1.0000
PROD=1.0000
DO312K=1,N
I=NP1-K
PROD=PROD/DEL(I)
312 A(K+1,NP1)=0.0000
DO313I=2,NP1
DO313K=1,N
J=NP1-K
313 A(I,J)=(A(I,J+1)-Q(J)*A(I-1,J+1))/DEL(J)
DO314I=1,NP1
DO314J=1,NP1
A(I,J+NP1)=0.0000
A(I+NP1,J)=0.0000
314 A(I+NP1,J+NP1)=A(I,J)
WRITE(6,1005)
1005 FFORMAT(1X,'A-MATRIX')
CALL PRMAT(A,NPNP2,NPNP2,MAX)
RETURN
END
```

SUBROUTINE: BUILD Z

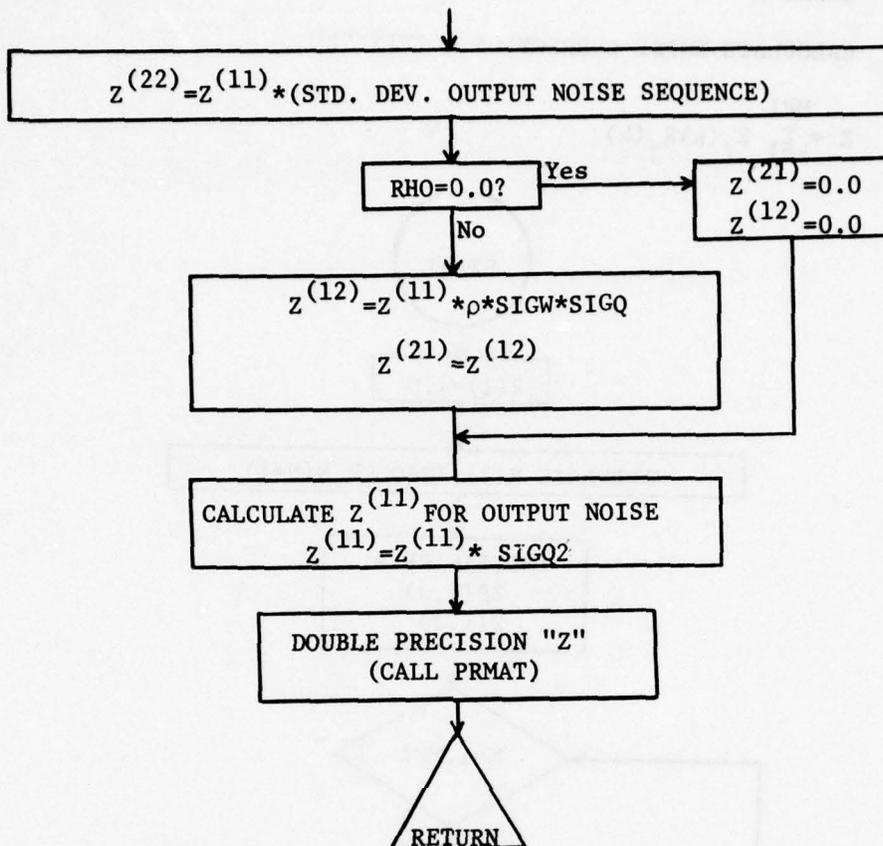
PURPOSE: CALCULATE NOISE CORRECTION MATRIX "Z"

EQUATION: 
$$Z = \sum_{k=1}^{MP1} R_1(k)R_j(k)$$

FLOW CHART:



SUBROUTINE: BUILD Z



DESCRIPTION: This subroutine calculates the contribution of the noise to the gram matrix resulting from the measurement filter output. The total Z matrix is formed in four sections. The First section ( $Z^{(11)}$ ) is generated through use of the GAMMA matrix. The second, third and fourth sections deal in optimizing the estimate of the noise correction matrix.

PROGRAM VARIABLES:

DEL	MEASUREMENT FILTER NUMERATOR
ILEVIN	VALUE IS EITHER 0 OR 1. 0 GRAM TECHNIQUE IS PERFORMED 1 LEVEN TECHNIQUE IS PERFORMED
MAX	DIMENSION SIZE
MP1	NO. OF DATA POINTS
N	ORDER OF SYSTEM

SUBROUTINE: BUILD Z

Q	MEASUREMENT FILTER POLE
R	COEFFICIENT VECTOR
RHO	EXPECTATION OF $(W(K)*Q(K))$
SIGQ	STANDARD DEVIATION OF OUTPUT NOISE SEQUENCE
SIGW	STANDARD DEVIATION OF INPUT NOISE SEQUENCE
Z	NOISE CORRECTION MATRIX
ZP	WORKING ARRAY

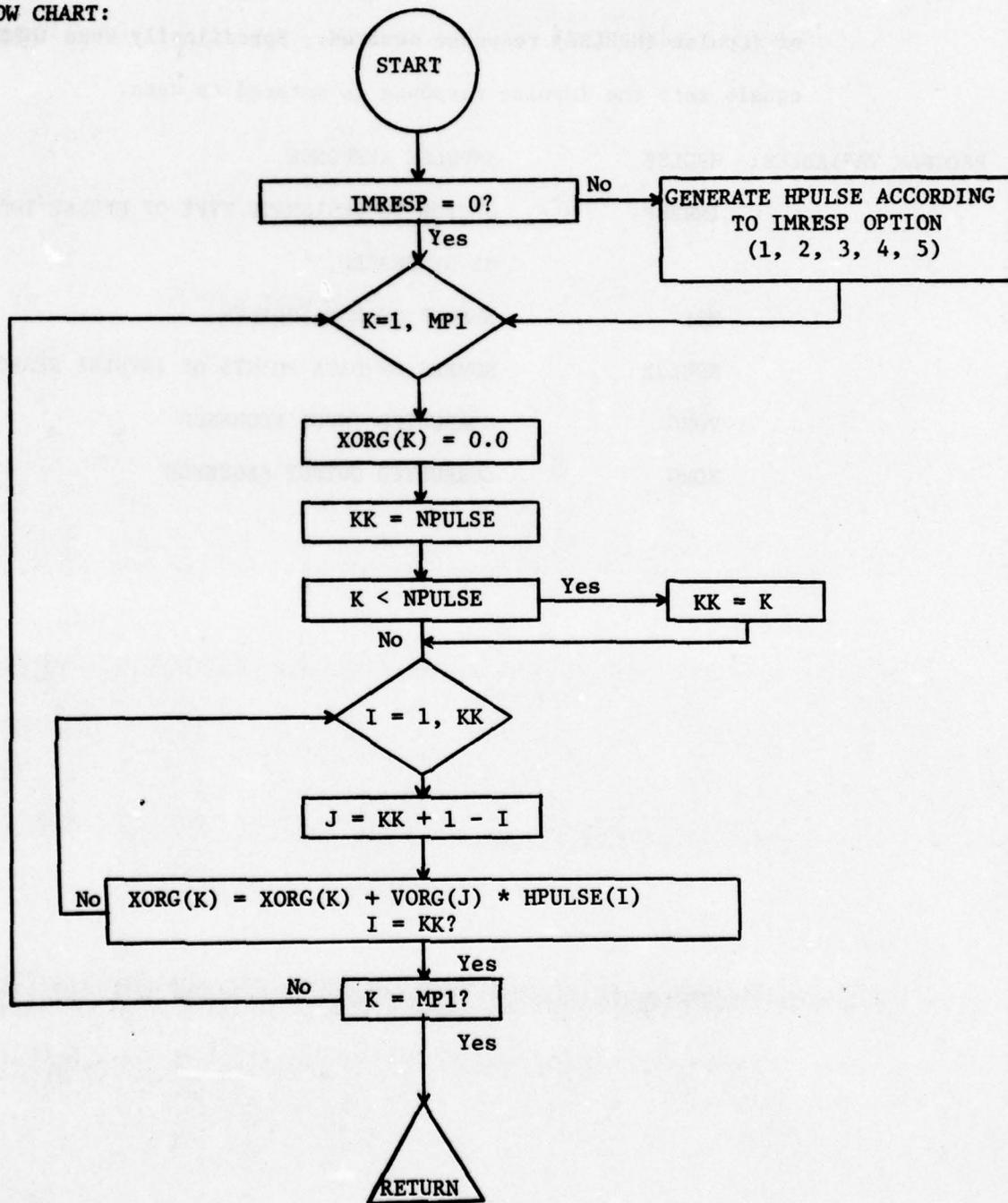


SUBROUTINE: CONVOL

PURPOSE: PERFORM CONVOLUTION OF HPULSE AND VORG

EQUATION: 
$$XORG = \sum_{m=1}^{NPULSE} VORG_{k+1-m} HPULSE_m$$

FLOW CHART:



SUBROUTINE: CONVOL

DESCRIPTION: This subroutine determines the convolution (XORG(K)) of VORG(I) and HPULSE(J). NPULSE is the length of the impulse response and the variable IMRESP is the option used to specify the type of impulse (HPULSE) response desired. Specifically when IMRESP equals zero the impulse response is entered as data.

PROGRAM VARIABLES:	HPULSE	IMPULSE RESPONSE
	IMRESP	OPTION TO DESIGNATE TYPE OF HPULSE TO BE GENERATED
	MPI	NUMBER OF DATA POINTS
	NPULSE	NUMBER OF DATA POINTS OF IMPULSE RESPONSE
	VORG	CORRUPTED INPUT SEQUENCE
	XORG	CORRUPTED OUTPUT SEQUENCE

```

C
C
C
SUBROUTINE CONVOL(HPULSE,VORG,XORG,NPULSE,MP1,IMRESP)
PERFORMS CONVOLUTION OF HPULSE AND VORG
XORG(K)= HPULSE(1)*VORG(K)+ ..... +HPULSE(KK)*VORG(K-KK+1),
WHERE KK=NPULSE OR K WHICHEVER SMALLER

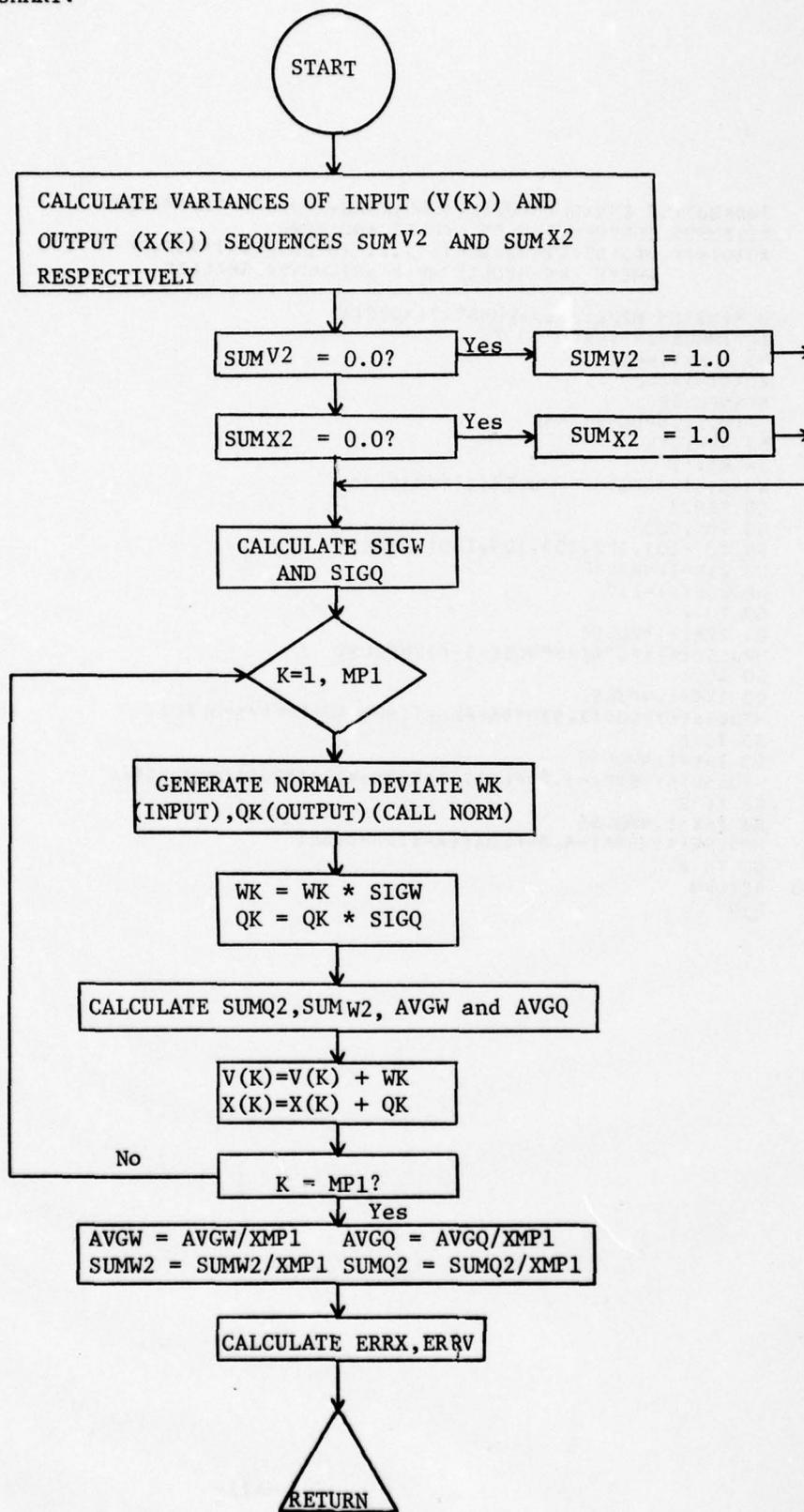
DIMENSION HPULSE(1),VORG(1),XORG(1)
IF(IMRESP.NE.0)GO TO 2:
2  DO 5 K=1,MP1
   XORG(K)=0.0
   KK=NPULSE
   IF(K.LT.NPULSE)KK=K
   DO 4 I=1,KK
    J= K+1-I
4  XORG(K)=XORG(K)+HPULSE(I)*VORG(J)
5  CONTINUE
   GO TO 1000
20  GO TO (101,102,103,104,105),IMRESP
101 DO 21K=1,NPULSE
21  HPULSE(K)=1.0
   GO TO 2
102 DO 22K=1,NPULSE
22  HPULSE(K)=FLOAT(NPULSE+1-K)/NPULSE
   GO TO 2
103 DO 23K=1,NPULSE
23  HPULSE(K)=COS(1.570796*FLOAT(NPULSE+1-K)/NPULSE)
   GO TO 2
104 DO 24K=1,NPULSE
24  HPULSE(K)=EXP(-8.0*FLOAT(K*K-2*K+1)/(NPULSE*NPULSE))
   GO TO 2
105 DO 25K=1,NPULSE
25  HPULSE(K)=EXP(-4.0*FLOAT(K-1)/NPULSE)
   GO TO 2
1000 RETURN
END

```

SUBROUTINE: CORUPT

EQUATION: See attached sheet

FLOW CHART:



SUBROUTINE: CORUPT

EQUATION:	$AVGW = \frac{1}{XMP1} \sum_{k=1}^{MP1} WK$		Sample mean of input noise seq.
	$AVGQ = \frac{1}{XMP1} \sum_{k=1}^{MP1} QK$		Sample mean of output noise seq.
	$SUMQ2 = \frac{1}{XMP1} \sum_{k=1}^{MP1} QK^2$		Sample mean of power of output noise
	$SUMW2 = \frac{1}{XMP1} \sum_{k=1}^{MP1} WK^2$		Sample mean power of input noise
	$SUMX2 = \frac{1}{XMP1} \sum_{k=1}^{MP1} X(k)^2$		Sample mean power of output seq.
	$SUMV2 = \frac{1}{XMP1} \sum_{k=1}^{MP1} V(k)^2$		Sample mean power of input seq.

SUBROUTINE: CORUPT

DESCRIPTION: This subroutine calculates the standard deviation of Input (SIGW) noise sequence and Output(SIGQ) noise sequence. The values of the mean power of input and output noise along with the mean power of input and output sequence are also calculated in this subroutine. The final calculation is of the noise to signal power ratio of both input and output.

PROGRAM VARIABLES:	MP1	NUMBER OF DATA POINTS
	SIGQ	STANDARD DEVIATION OF THE OUTPUT NOISE SEQUENCE Q(k)
	SIGW	STANDARD DEVIATION OF THE INPUT NOISE SEQUENCE W(k)
	V	CORRUPTED INPUT SEQUENCE
	X	CORRUPTED OUTPUT SEQUENCE

```

SUBROUTINE CORUPT(X,V,SIGQ,SIGW,MPI)
DIMENSION X(1),V(1)
REAL*8 SUMV2,SUMX2,SUMQ2,SUMW2,AVGQ,AVGW

      IX INITIALIZES UNIFORM RANDOM NUMBER GENERATOR (IBM SUBROUTINE RANDU)
      RANDU IS CALLED BY SUBROUTINE NORM WHICH GENERATES NORMAL DEVIATES

      INITIATE THE RANDOM SEQUENCE GENERATOR
      IX=65549
      XMP1=MPI
      SUMX2=3.7000
      SUMV2=0.0000
      DJ39K=1,MPI
      SUMX2=SUMX2+X(K)*X(K)
      SUMV2=SUMV2+V(K)*V(K)
39  CONTINUE
      SUMV2=SUMV2/XMP1
      SUMX2=SUMX2/XMP1
      IF(SUMV2.EQ.0.000)SUMV2=1.000
      IF(SUMX2.EQ.0.000)SUMX2=1.000

      FOR INPUT=0, SIGW AND SIGQ BECOME STD. DEV. OF NOISE

      SIGW=DSQRT(SUMV2*SIGW)
      SIGQ=DSQRT(SUMX2*SIGQ)
      WRITE(6,897)SIGQ,SIGW
897  FORMAT(//30X,'SIGQ=',G17.10,5X,'SIGW=',G17.10,/)
      SUMQ2=0.0000
      SUMW2=0.0000
      AVGQ=0.0000
      AVGW=0.0000
      DJ40K=1,MPI
      CALL NDRM(WK,IX)
      CALL NDRM(QK,IX)
      WK=WK*SIGW
      QK=QK*SIGQ
      SUMQ2=SUMQ2+QK*QK
      SUMW2=SUMW2+WK*WK
      AVGW=AVGW+WK
      AVGQ=AVGQ+QK
      V(K)=V(K)+WK
40  X(K)=X(K)+QK
      AVGW=AVGW/XMP1
      AVGQ=AVGQ/XMP1
      SUMW2=SUMW2/XMP1
      SUMQ2=SUMQ2/XMP1
      ERRX=DSQRT(SUMQ2/SUMX2)*100.0
      ERRV=DSQRT(SUMW2/SUMV2)*100.0
      WRITE(6,1001)AVGQ,AVGW,SUMQ2,SUMW2,SUMX2,SUMV2,ERRX,ERRV
1001 FORMAT(///,20X,'SAMPLE MEAN OF OUTPUT NOISE SEQUENCE = ',E11.4,/,
1,20X,'SAMPLE MEAN OF INPUT NOISE SEQUENCE = ',E11.4,/,20X,
2'SAMPLE MEAN POWER OF OUTPUT NOISE = ',E11.4,/,20X,
3'SAMPLE MEAN POWER OF INPUT NOISE = ',E11.4,/,20X,
4'SAMPLE MEAN POWER OF OUTPUT SEQUENCE = ',E11.4,/,20X,

5'SAMPLE MEAN POWER OF INPUT SEQUENCE = ',E11.4,/,20X,
6'100.0 TIMES THE SQUARE ROOT OF THE NOISE TO SIGNAL POWER RATIO OF
7 THE OUTPUT = ',F7.3,/,20X,
8'100.0 TIMES THE SQUARE ROOT OF THE NOISE TO SIGNAL POWER RATIO OF
9 THE INPUT = ',F7.3)
      RETURN
      END

```

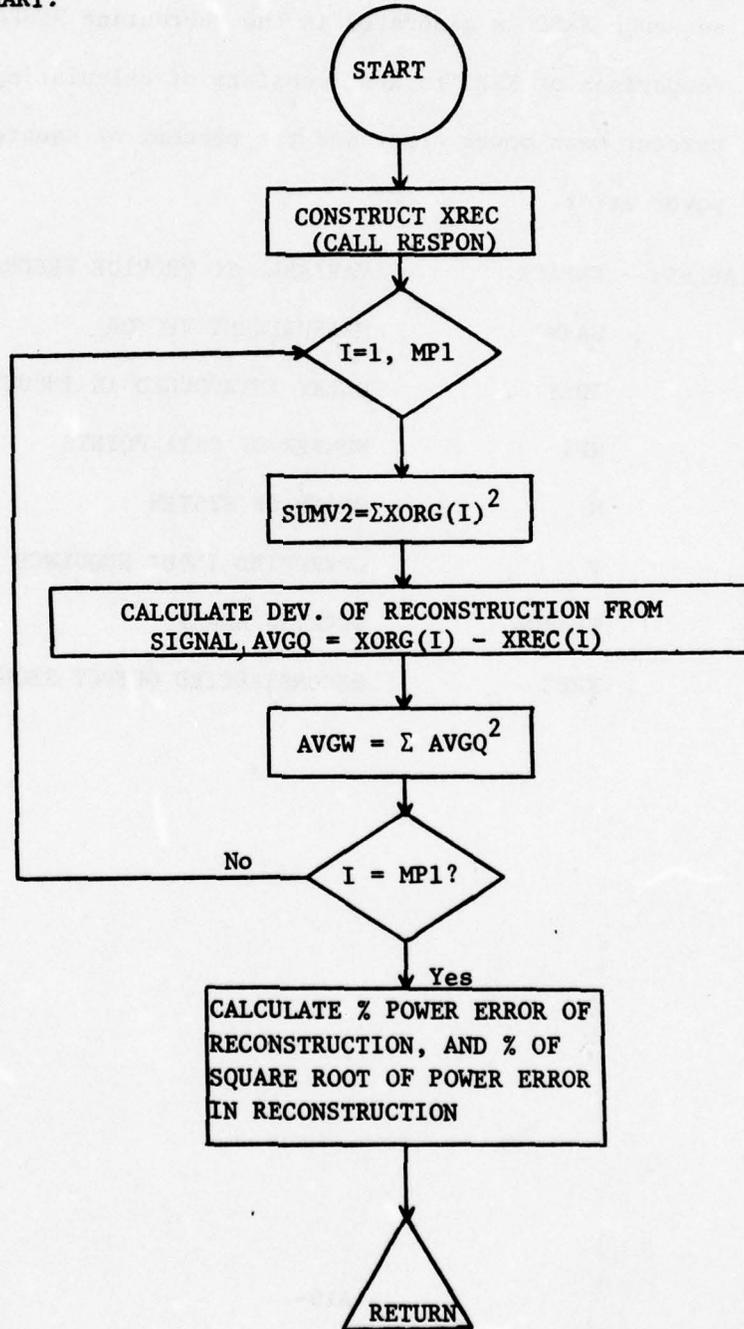
SUBROUTINE: ERROR

PURPOSE: CALCULATE PERCENT MEAN POWER ERROR IN RECONSTRUCTION AND PERCENT OF SQUARE ROOT OF POWER ERROR IN RECONSTRUCTION.

EQUATIONS:  $AVGW = \sum \left[ \frac{(XORG(I) - XREC(I))}{XORG(I)} \right]^2 * 100$

$AVGQ = \sqrt{AVGW} * 100$

FLOW CHART:



SUBROUTINE: ERROR

DESCRIPTION: The subroutine ERROR calculates the Output sequence from the Input sequence and the GAMMA Matrix. The sequence called XREC is compared to XORG, the original output sequence. The sequence XREC is generated in the subroutine RESPON. The comparison of XREC to XORG consists of calculating the percent mean power error and the percent of square root of power error.

PROGRAM VARIABLES:	FDBACK	VARIABLE TO PROVIDE FEEDBACK IF DESIRED
	GAMMA	MEASUREMENT VECTOR
	IDLY	DELAY INTRODUCED IN INPUT NUMERATOR
	MP1	NUMBER OF DATA POINTS
	N	ORDER OF SYSTEM
	V	CORRUPTED INPUT SEQUENCE
	XLAMDA	WORKING ARRAY
	XREC	RECONSTRUCTED OUTPUT SEQUENCE

```

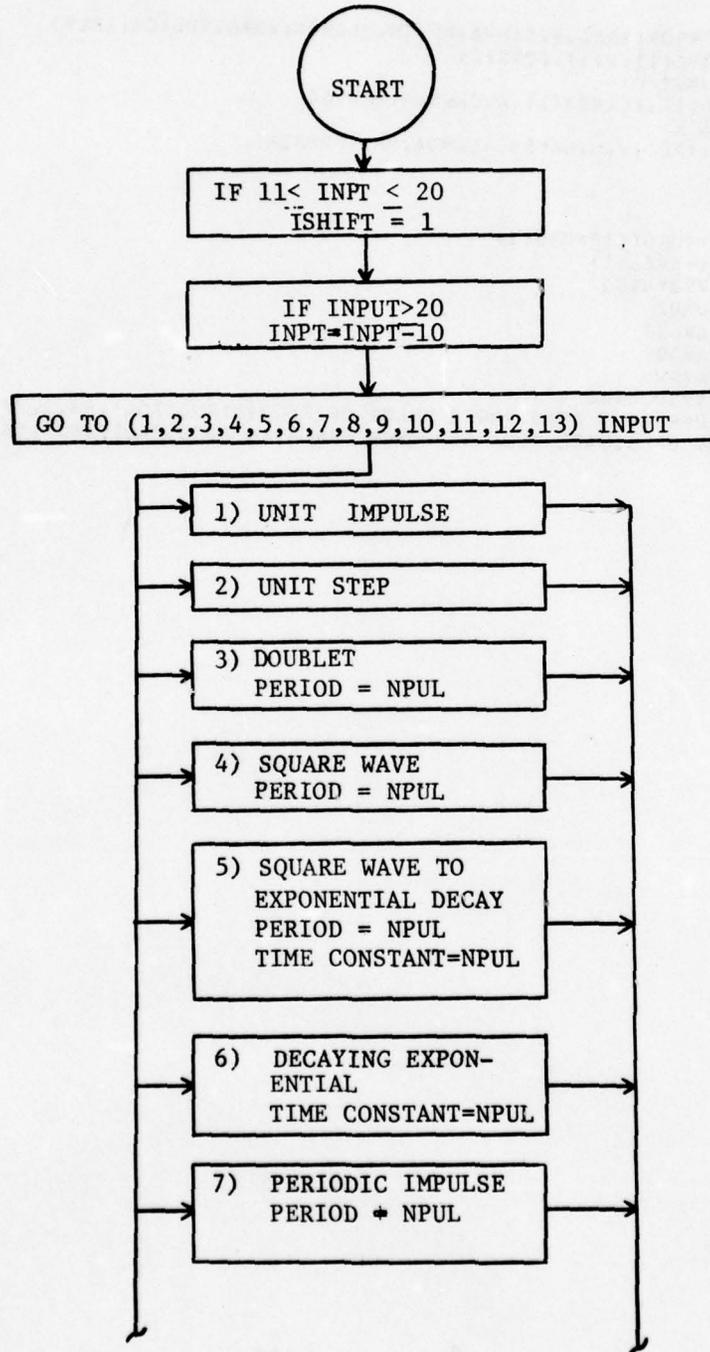
SUBROUTINE ERROR(XREC,V,GAMMA,MP1,N,XLAMDA,XORG,FDBACK,IDL)
DIMENSION XREC(1),V(1),XORG(1)
DIMENSION VVV(20)
REAL*8 GAMMA(1),XLAMDA(1),AVGW,SUMV2,AVGQ
INTEGER FDBACK
CALL RESPON(XREC,V,N,GAMMA,XLAMDA,MP1,FDBACK)
AVGW=0.0000
SUMV2=0.000
DO26I=1,MP1
SUMV2=SUMV2+XORG(I)*XORG(I)
AVGQ=XORG(I)-XREC(I)
26  AVGW=AVGW+AVGQ*AVGQ
    AVGW=AVGW/SUMV2
    AVGQ=DSQRT(AVGW)
    AVGQ=100.0*AVGQ
    AVGW=100.0*AVGW
27  WRITE(6,27)AVGW,AVGQ
    FORMAT(1X,'PER CENT MEAN POWER ERROR OF RECONSTRUCTION',F8.3,///,
    11X,'PER CENT OF SQUARE ROOT OF POWER ERROR IN RECOSTRUCTION',F8.3)
    RETURN
    END

```

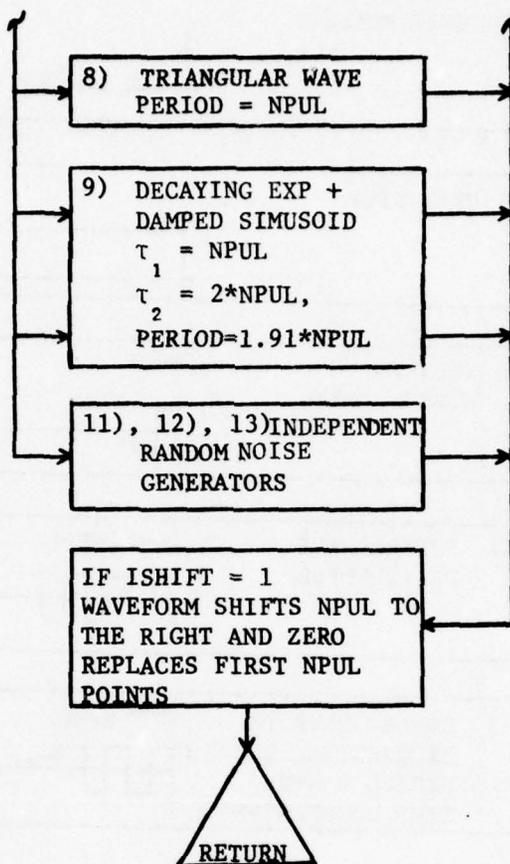
SUBROUTINE: FILLV

PURPOSE: GENERATION OF DISCRETE POINTS FOR A VARIETY OF WAVEFORMS  
 (For the correspondance of the waveshapes and input parameter  
 INPT see page )

FLOW CHART:



SUBROUTINE: FILLV

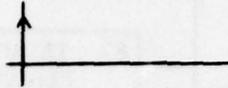


DESCRIPTION: This subroutine builds an array of NPT points defined by the chosen wave form and parameter (NPUL) of that waveform. It is useful in approximating input signals for excitation of control system.

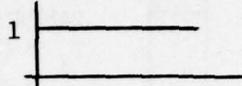
PROGRAM VARIABLES:	INPUT	DESIRED WAVEFORM OPTION
	NPT	NUMBER OF DATA POINTS
	NPUL	WAVEFORM PARAMETER
	V	GENERATED OUTPUT SEQUENCE

SUBROUTINE: FILLV

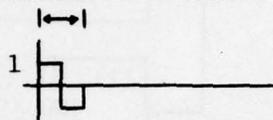
1) UNIT PULSE



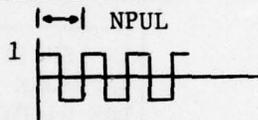
2) UNIT STEP



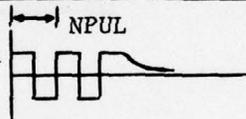
3) DOUBLET  
WIDTH = NPUL



4) SQUARE WAVE  
PERIOD=NPUL



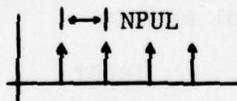
5) SQUARE WAVE TO  
EXPONENTIAL DECAY 1  
PERIOD = NPUL  
TIME CONSTANT=NPUL



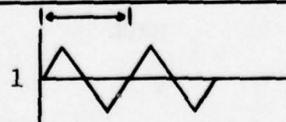
6) EXPONENTIAL  
TIME CONSTANT=NPUL



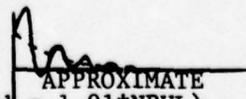
7) PERIODIC IMPULSE  
PERIOD=NPUL



8) TRIANGULAR WAVE  
PERIOD=NPUL



9) DECAYING EXP +  
DAMPED SINUSOID  
 $\tau_1 = NPUL$   
( $\tau_2 = 2*NPUL$ , Period =  $1.91*NPUL$ )



```

SUBROUTINE FILLV(V,NPT,INPUT,NPUL)
C   FILLS THE ARRAY FOR INPUT ACCORDING TO INPUT OPTION DESIGNATED
    DIMENSION V(1)
    GO TO (1,2,3,4,5,6,7,8,9,10),INPUT
1   V(1)=1.0                                IMPULSE
    DO 101 I=2,NPT
101  V(I)=0.0
    GO TO 999
2   DO 102 I=1,NPT                          STEP
102  V(I)=1.0
    GO TO 999
3   DO 103 I=1,NPT                          DOUBLET
    V(I)=0.0
    IF(I.LT.NPUL/2)V(I)=1.
    IF(I.GE.NPUL.AND.I.LT. NPUL) V(I)=-1.0
103  CONTINUE
    GO TO 999
4   DO 104 I=1,NPT                          SQ WAVE
    V(I)=1.0
    TPUL=I/NPUL
    IF(I-TPUL*NPUL.GE.NPUL/2) V(I)=-1.0
104  CONTINUE
    GO TO 999
5   DO 105 I=1,NPT                          SQ-EXP
    V(I)=1.0
    IF(I.GE.NPUL/2.AND.I.LT.NPUL) V(I)=-1.0
    IF(I.GE.(1.5)*NPUL.AND.I.LT.2*NPUL) V(I)=-1.0
    ARG1=2.5-FLOAT(I)/FLOAT(NPUL)
    IF(I.GE.(2.5)*NPUL) V(I)=EXP(ARG1)
105  CONTINUE
    GO TO 999
6   DO 106 I=1,NPT                          EXP
    ARG2=-FLOAT(I)/FLOAT(NPUL)
106  V(I)=EXP(ARG2)
    GO TO 999
7   DO 107 I=1,NPT                          PRD IMPL
107  V(I)=0.0
    N=NPT/NPUL
    V(1)=1.0
    DO 1071 J=1,N
    I=J*NPUL
1071 V(I)=1.0
    GO TO 999
8   DO 108 I=1,NPT                          TRI WAVE
    TPUL=I/NPUL
    ITPUL=TPUL
    V(I)=(2.*FLOAT(I)/FLOAT(NPUL)-2.*TPUL)*(-1.0)**ITPUL
    IF(I-TPUL*NPUL.GE.NPUL/2) V(I)=2*(1+TPUL-FLOAT(I)/FLOAT(NPUL))*(-1
    1.0)**ITPUL
108  CONTINUE
    GO TO 999
9   DO 109 I=1,NPT                          EXP+OSC
    ARG3=-FLOAT(I)/FLOAT(NPUL)
    ARG4=-.5*FLOAT(I)/FLOAT(NPUL)
    ARG5=3.296*FLOAT(I)/FLOAT(NPUL)

```

```
109 V(I)=EXP(ARG2)+EXP(ARG4)*SIN(ARG5)
    GO TO 999
10 IX=619327213
    DO 110 I=1,NPT
    A=0.0
    DO 1101 K=1,12
    IY=IX*65539
    IF(IY)1102,1103,1103
1102 IY=IY+2147483647+1
1103 YFL=IY
    YFL=YFL*.4656613E-9
    IX=IY
    A=A+YFL
1101 CONTINUE
    V(I)=A-5.0
110 CONTINUE
    GO TO 999
999 CONTINUE
    RETURN
    END
```

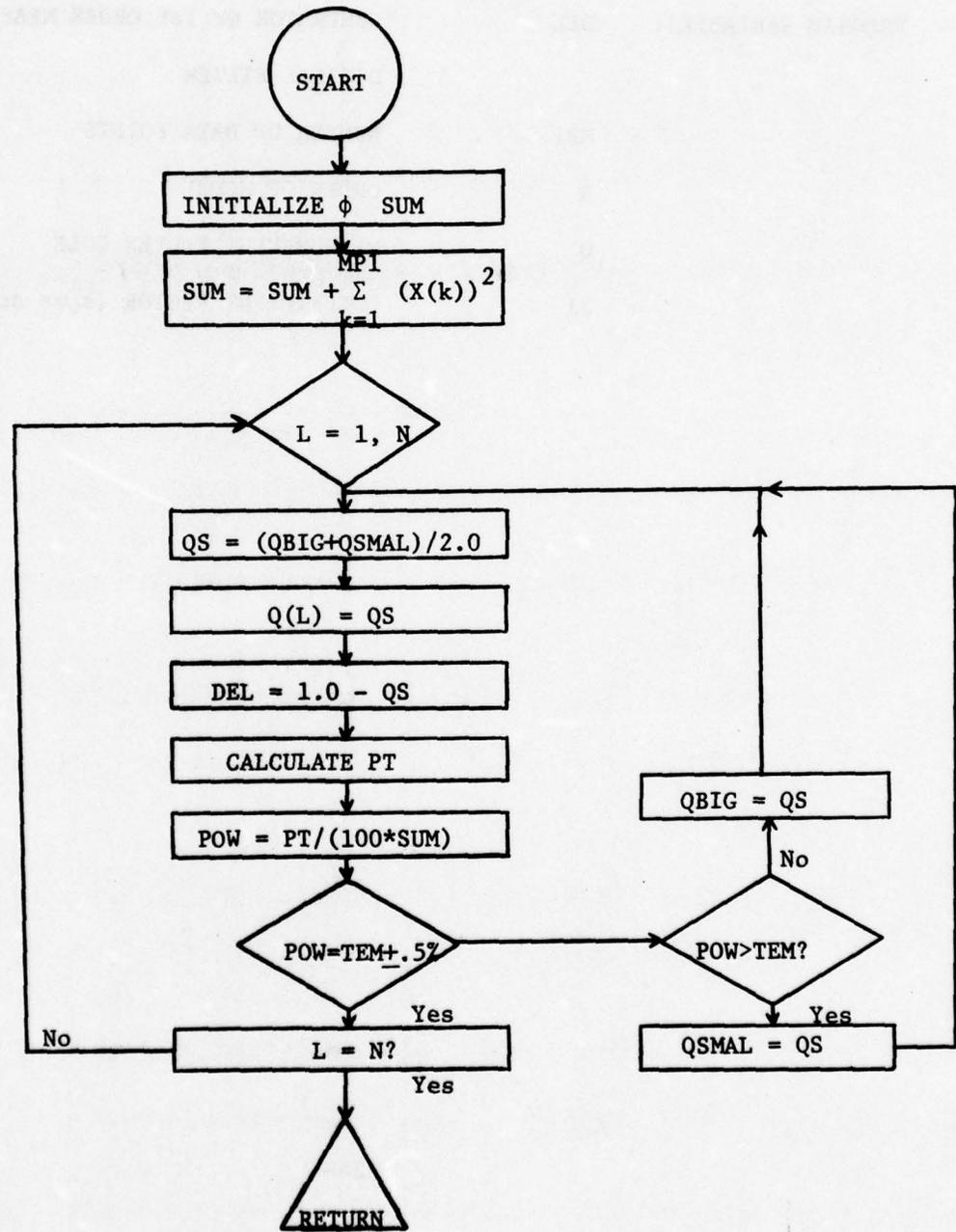
RANDOM

SUBROUTINE: FINDQ

PURPOSE: CALCULATES MEASUREMENT FILTER POLE(Q) AND NUMERATOR (DEL).

EQUATION:  $H_m(z) = \frac{DEL}{1-QZ^{-1}}$

FLOW CHART:



SUBROUTINE: FINDQ

DESCRIPTION: FINDQ determines the measurement filter pole and numerator.  
 The subroutine uses an iterative process of calculating DEL and Q. The iteration is satisfied when the variable POW is within  $\pm .5\%$  TEM.

PROGRAM VARIABLES:	DEL	NUMERATOR OF 1st ORDER MEASUREMENT DIGITAL FILTER
	MP1	NUMBER OF DATA POINTS
	N	ORDER OF MODEL
	Q	MEASUREMENT FILTER POLE
	X1	COEFFICIENT VECTOR (same as Gamma)

```

SUBROUTINE FINDQ(Q,DEL,X1,X,MP1,N)
DIMENSION X(1)
REAL*8 Q(1),DEL(1),TEM,POW,PT,QS,QBIG,QSMAL,X1(1),SUM
SUM=0.000
DO667K=1,MP1
667 SUM=SUM+X(K)**2
NP1=N+1
DO11L=1,N
LP1=L+1
TEM=100.000/DFLOAT(NP1)*DFLOAT(NP1-L)
QBIG=1.000
QSMAL=0.000
100 QS=(QBIG+QSMAL)/2.000
Q(L)=QS
DEL(L)=1.000-QS
PT=0.000
DO4I=1,LP1
4 X1(I)=0.000
DO3K=1,MP1
X1(I)=X(K)
DO5I=1,L
5 X1(I+1)=X1(I+1)*Q(I)+X1(I)*DEL(I)
3 PT=PT+X1(LP1)*X1(LP1)
POW=PT/SUM*100.000
IF(POW.LE.1.00500*TEM.AND.POW.GE.TEM*0.99500)GO TO 1
IF(POW.GT.TEM)GO TO 6
QBIG=QS
GO TO 100
6 QSMAL=QS
GO TO 100
1 CONTINUE
RETURN
END

```

SUBROUTINE: GRAMII

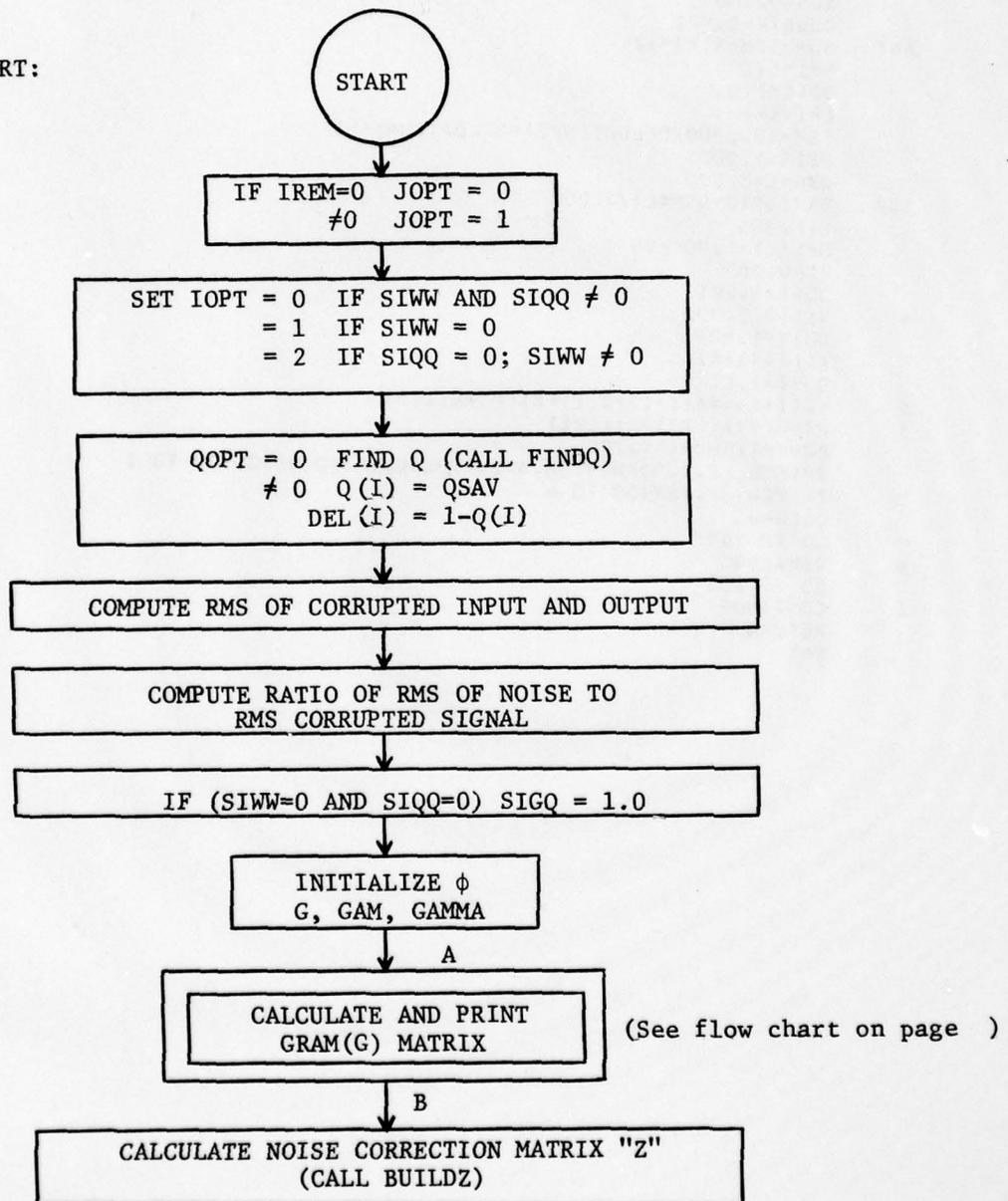
PURPOSE: PERFORMS GRAM II TECHNIQUE WHICH YIELDS THE GRAM MATRIX (G), NOISE CORRECTION MATRIX (Z) AND THE COEFFICIENT MATRIX (GAMMA).

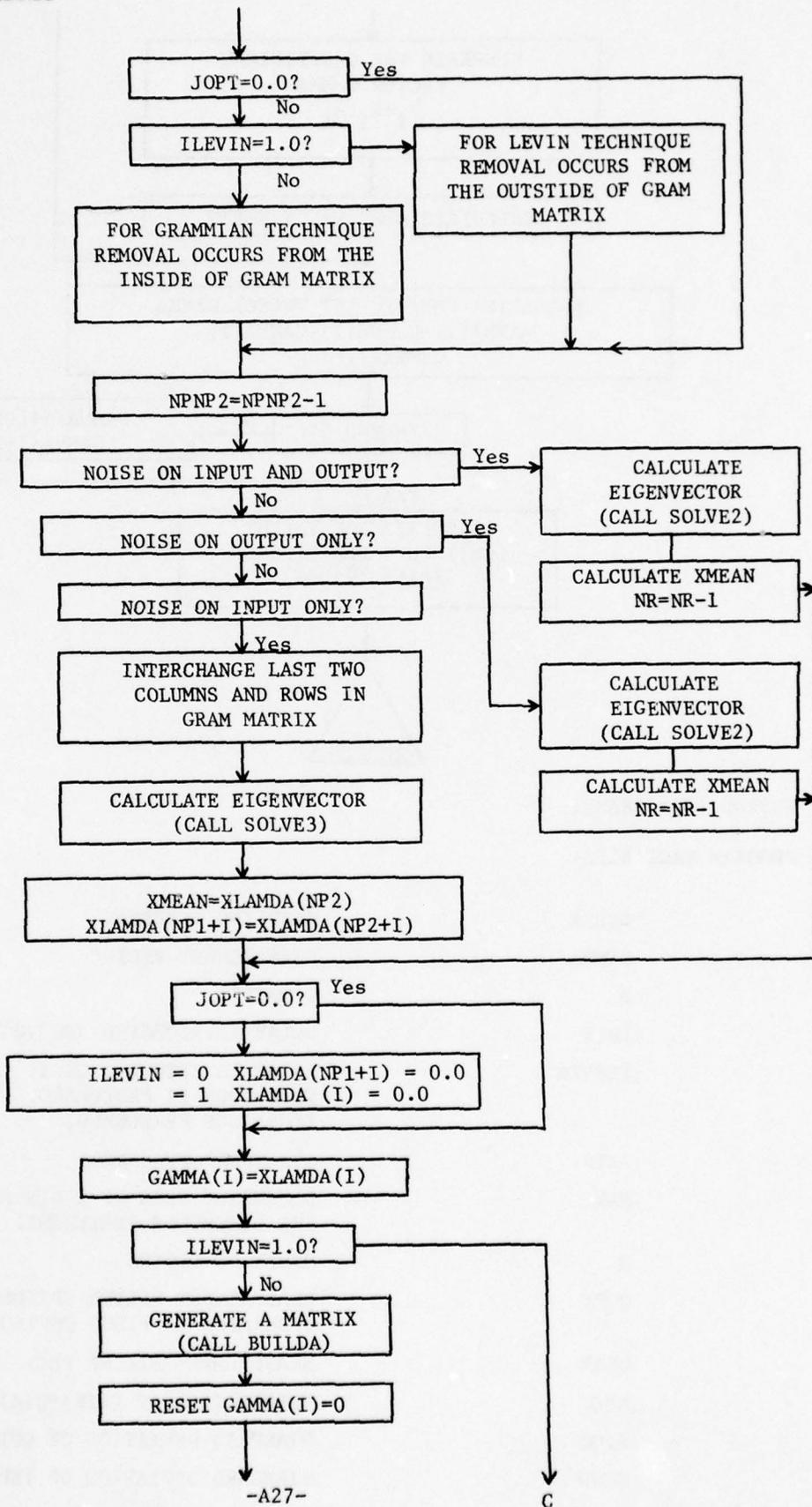
EQUATION:

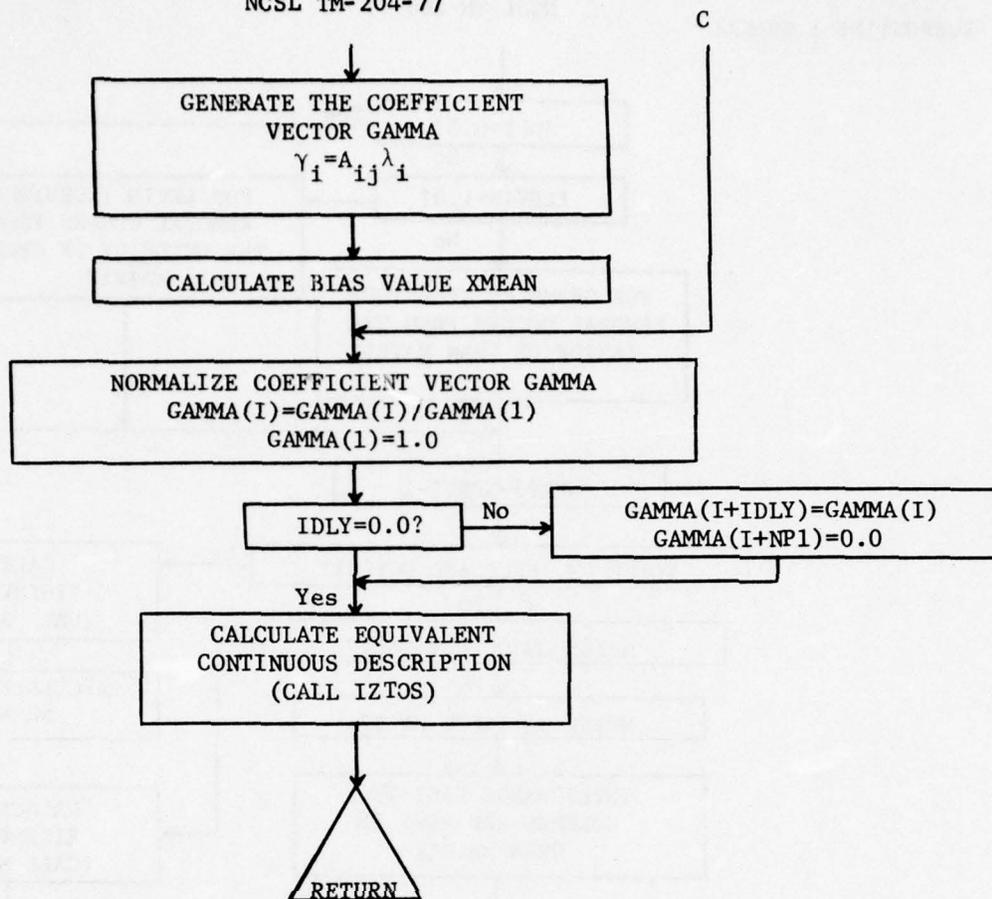
$$G_{ij} = \sum_{k=1}^{MP1} \text{GAMMA}_i(k) * \text{GAMMA}_j(k)$$

$$\gamma_i = A_{ij} \lambda_j$$

FLOW CHART:







SUBROUTINE: GRAMII

PROGRAM VARIABLES:

DELTA	SAMPLING INTERVAL
GAMMA	MEASUREMENT VECTOR
G	G MATRIX
IDLY	DELAY INTRODUCED ON INPUT NUMERATOR
I Levin	VALUE IS EITHER 0 OR 1. 0 GRAM TECHNIQUE IS PERFORMED. 1 LEVIN TECHNIQUE PERFORMED.
IZTS	SEE SUBROUTINE ZTOS
MAX	DIMENSION SIZE OF 2 DIM ARRAYS IN THE DIMENSION STATEMENT.
N	ORDER OF SYSTEM
QOPT	MEASUREMENT FILTER OPTION. QOPT=1 Q(I) GENERATED IN FINDQ QOPT=0 Q(I)=QSAV
QSAV	MEASUREMENT FILTER POLE
RHO	EXPECTATION OF (W(K)*Q(K))
SIQQ	STANDARD DEVIATION OF OUTPUT NOISE SEQUENCE
SIWW	STANDARD DEVIATION OF INPUT NOISE SEQUENCE

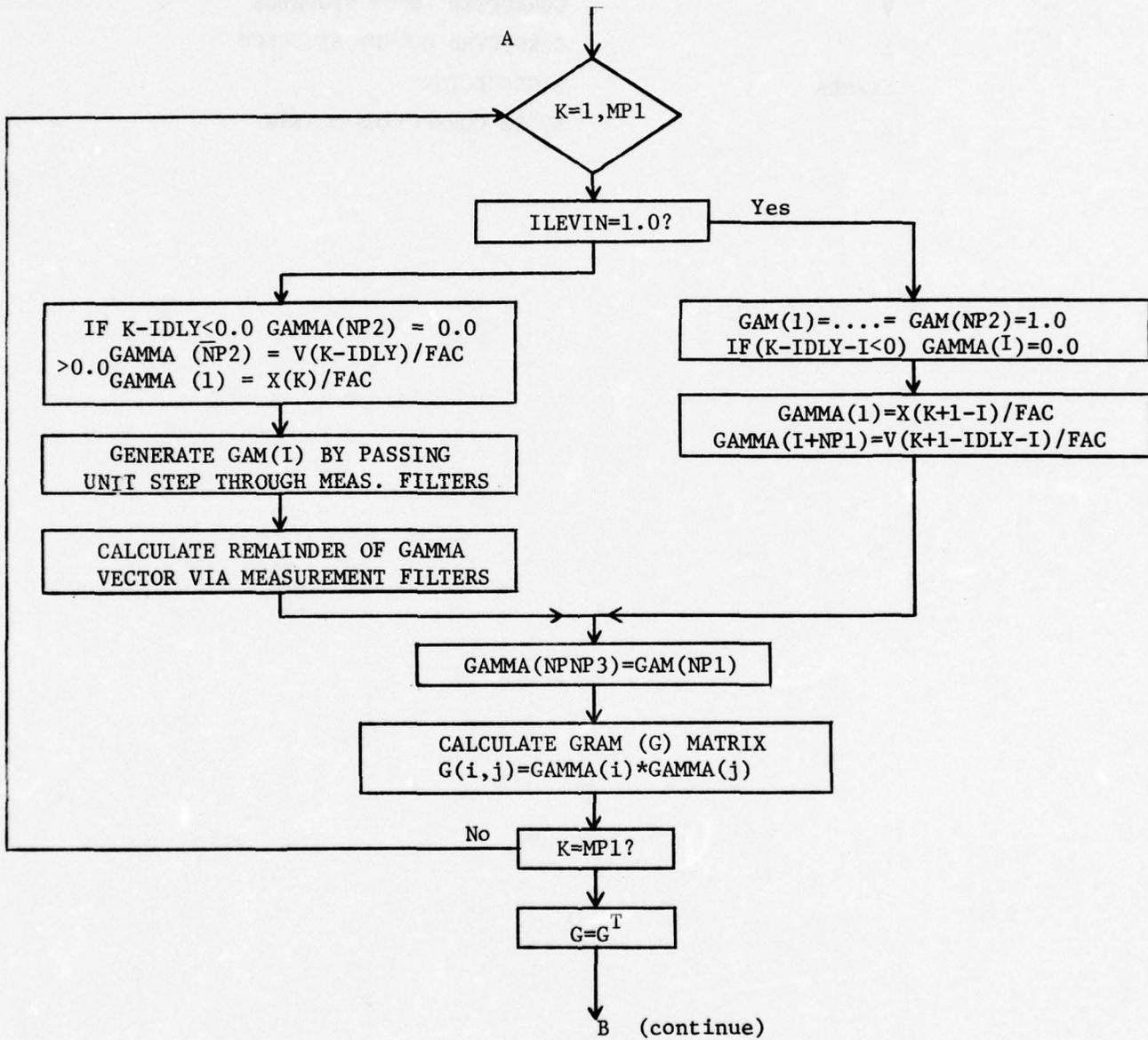
SUBROUTINE: GRAMII

PROGRAM VARIABLES CONTINUED:

V	CORRUPTED INPUT SEQUENCE
X	CORRUPTED OUTPUT SEQUENCE
XLAMDA	EIGENVECTOR
Z	NOISE CORRECTION MATRIX

SUBROUTINE: GRAMII

CALCULATION OF GRAM MATRIX "G"



```

SUBROUTINE GRAMII(X,V,MP1,SIQQ,SIWW,RHO,N,DELTA,QSAV,QOPT,IREM,
11ZTS,GAMMA,XLAMDA,G,Z,MAX,ILEVIN,IDLY,INORM)
C
C THIS SUBROUTINE PERFORMS THE GRAM II TECHNIQUE
C
DIMENSION X(1),V(1),G(MAX,1),Z(MAX,1),GAMMA(1),XLAMDA(1),Q(20),
IDEL(20)
DOUBLE PRECISION G,Z,GAMMA,XLAMDA,DELTA,DEL,PROD,Q,QSAV
INTEGER QOPT
REAL*8 XMEAN,GAM(25),FAC
REAL*8 S,F,S1,S2,VARQ,VARW,GI
COMMON /MATRIX/S(20,20),F(20,20),GI(20,20),S1(10,10),S2(10,10)
MAX2=MAX/2
WRITE(6,1000)
1000 FORMAT(1H1,20X,'THE GRAM II TECHNIQUE')
C JOPT = 0 IF DIRECT TRANSMISSION IS ASSUMED
C JOPT=0
C JOPT = 1 IF NO DIRECT TRANSMISSION IS ASSUMED
IF(IREM.NE.0)JOPT=1
C
C IOPT = 0 NOISE ON BOTH INPUT AND OUTPUT IS ASSUMED
C IOPT = 1 NOISE ON OUTPUT ONLY IS ASSUMED
C IOPT = 2 NOISE ON INPUT ONLY IS ASSUMED
C
C IOPT=0
IF(SIWW.EQ.0.0)IOPT=1
IF(SIQQ.EQ.0.0.AND.SIWW.NE.0.0)IOPT=2
C
C DEL IS THE NUMERATOR OF THE KNOWN FIRST ORDER DIGITAL FILTERS
IF(QOPT.NE.0) GO TO 21
DO19I=1,N
DEL(I)=1.0D00-QSAV
19 Q(I)=QSAV
GO TO 22
21 CALL FINDQ(Q,DEL,GAMMA,X,MP1,N)
22 CONTINUE
WRITE(6,2020)
2020 FORMAT(30X,'Q PARAMETERS')
CALL PRVEC(Q,N)
NP1=N+1
NP2=N+2
NPNP2=N+N+2
NR=NP1-IREM
NP1PIR=NP1+IREM
VARW=0.0
VARQ=0.0
DO300I=1,MP1
VARW=VARW+V(I)*V(I)
300 VARQ=VARQ+X(I)*X(I)
VARQ=DSQRT(VARQ/MP1)
VARW=DSQRT(VARW/MP1)
SIQQ=SIQQ/VARQ
SIWW=SIWW/VARW
IF(SIWW.EQ.0.0.AND.SIQQ.EQ.0.0)SIQQ=1.0
NPNP2=NPNP2+1

```

```

NR=NR+1
DO10I=1,NPNP2
GAM(I)=0.000
GAMMA(I)=0.0000
DO10J=I,NPNP2
10 G(I,J)=0.0000
GAM(I)=1.000
C
C      CALCULATING THE G MATRIX
DO50K=1,MP1
IF(ILEVIN.EQ.1)GO TO 31
IF(K-IDLY)25,25,24
25 GAMMA(NP2)=0.0000
GO TO 26
24 FAC=1.0
IF(INORM.EQ.1)FAC=VARW
GAMMA(NP2)=V(K-IDLY)/FAC
FAC=1.0
IF(INORM.EQ.1)FAC=VARQ
GAMMA(1)=X(K)/FAC
26 CONTINUE
DO30I=1,N
GAM(I+1)=GAM(I+1)*Q(I)+GAM(I)*DEL(I)
GAMMA(I+1)=GAMMA(I)*DEL(I)+GAMMA(I+1)*Q(I)
30 GAMMA(I+NP2)=GAMMA(I+NP1)*DEL(I)+GAMMA(I+NP2)*Q(I)
GO TO 35
31 CONTINUE
DO 32 I=1,NP1
GAM(I+1)=GAM(I)
IF(K-IDLY-I.LT.0) GAMMA(I)=0.0000
IF(K-IDLY-I.LT.0) GAMMA(I+NP1)=0.0000
IF(K-IDLY-I.LT.0) GO TO 32
FAC=1.0
IF(INORM.EQ.1)FAC=VARQ
GAMMA(1)=X(K+1-I)/FAC
FAC=1.0
IF(INORM.EQ.1)FAC=VARW
GAMMA(I+NP1)=V(K+1-IDLY-I)/FAC
32 CONTINUE
35 CONTINUE
GAMMA(NPNP2)=GAM(NP1)
DO40I=1,NPNP2
DO40J=I,NPNP2
40 G(I,J)=G(I,J)+GAMMA(I)*GAMMA(J)
50 CONTINUE
DO60I=2,NPNP2
K=I-1
DO60J=1,K
60 G(I,J)=G(J,I)
WRITE(6,1002)
1002 FORMAT(20X,'THE G MATRIX')
CALL PRMAT(G,NPNP2,NPNP2,MAX)
C
C      CALCULATING THE NOISE CORRECTION MATRIX Z BY SUBROUTINE BUILDZ
CALL BUILDZ(Z,S,GAMMA,N,MP1,SIGW,SIGQ,RHO,DEL,Q,MAX,ILEVIN)
IF(JOPT)70,90,70
70 CONTINUE
IF(ILEVIN.EQ.1)GO TO 81
DO80J=1,NPNP2
DO80I=1,NR
Z(NP1+I,J)=Z(NP1PIR+I,J)

```

```

80  G(NP1+I,J)=G(NP1PIR+I,J)
      NPNP2=NPNP2-IREM
      DO85J=1,NPNP2
      DO85I=1,NR
      Z(J,NP1+I)=Z(J,NP1PIR+I)
85  G(J,NP1+I)=G(J,NP1PIR+I)
      GO TO 90
81  DO 82 J=1,NPNP2
      Z(NP1+NR+1,J)=G(NPNP2,J)
82  G(NP1+NR+1,J)=G(NPNP2,J)
      NPNP2=NPNP2-IREM
      DO83I=1,NPNP2
      Z(I,NP1+NR+1)=Z(I,NP1+NP1+1)
83  G(I,NP1+NR+1)=G(I,NP1+NP1+1)
      CALL PRMAT(G,NPNP2,NPNP2,MAX)
90  NPNP2=NPNP2-1
      IF(IOPT-1)617,605,618
C
C      NOISE ON BOTH INPUT AND OUTPUT
617 CALL SOLVE2(Z,G,GAMMA,XLAMDA,NPNP2,1,MAX)
      XMEAN=XLAMDA(NPNP2+1)
      NR=NR-1
      GO TO 606
C
C      NOISE ON OUTPUT ONLY
605 CALL SOLVE2(Z,G,GAMMA,XLAMDA,NP1,NR,MAX)
      XMEAN=XLAMDA(NPNP2+1)
      NR=NR-1
      GO TO 606
C
C      NOISE ON INPUT ONLY
618 NPP=NPNP2+1
      NR=NR-1
      DO550I=1,NPP
850 GAMMA(I)=G(I,NPP)
      DO551J=1,NR
      JJ=NR-J+1
      DO551I=1,NPP
851 G(I,NP2+JJ)=G(I,NP1+JJ)
      DO552I=1,NPP
852 G(I,NP2)=GAMMA(I)
      DO553I=1,NPP
853 GAMMA(I)=G(NPP,I)
      DO554I=1,NR
      II=NR-I+1
      DO554J=1,NPP
854 G(NP2+II,J)=G(NP1+II,J)
      DO555I=1,NPP
855 G(NP2,I)=GAMMA(I)
      CALL SOLVE3(Z,G,GAMMA,XLAMDA,NP2,NR,MAX)
      XMEAN=XLAMDA(NP2)
      DO556I=1,NR
856 XLAMDA(NP1+I)=XLAMDA(NP2+I)
606 IF(JOPT)120,130,120
120 NPNP2=NPNP2+IREM
      IF(ILEVIN.EQ.1)GO TO 124
      DO122I=1,NR
122 XLAMDA(NPNP2-I+1)=XLAMDA(NP2+NR-I)
      DO123I=1,IREM
123 XLAMDA(NP1+I)=0.0000
      GO TO 130

```

```

124 NN3MIR=NP2+1-IREM
    DOI25I=NN3MIR,NP2
125 XLAMDA(I)=0.0000
130 CONTINUE
    FAC=1.0
    IF(INORM.EQ.1)FAC=VARQ/VARW
    DO30II=NP2,NP2
301 XLAMDA(I)=XLAMDA(I)*FAC
    WRITE(6,1001)
1001 FORMAT(10X,'THE SYNTHETIC COEFFICIENT VECTOR, XLAMDA, IS')
    CALL PRVEC(XLAMDA,NP2)
    DO 150 I=1,NP2
150 GAMMA(I)=XLAMDA(I)
    IF(ILEVIN.EQ.1)GO TO 165
C
C     GENERATING GAMMA FROM XLAMDA
    CALL BUILDA(S,Q,DEL,N,PAX)
    DO160I=1,NP2
    GAMMA(I)=0.0000
    FAC=1.0
    IF(INORM.EQ.1)FAC=VARQ
    DO160J=1,NP2
160 GAMMA(I)=GAMMA(I)+S(I,J)*XLAMDA(J)
    XMEAN=XMEAN*S(1,NP1)*FAC/GAMMA(I)
165 CONTINUE
    WRITE(6,655)XMEAN
655 FORMAT(1X,'MEAN COEFFICIENT IS ',D13.6,/)
    DO200I=2,NP2
200 GAMMA(I)=GAMMA(I)/GAMMA(1)
    GAMMA(1)=1.0000
    IF(IDLY.EQ.0)GO TO 172
    IDLY1=IDLY+1
    DO 170 II=IDLY1,NP1
    I=NP2+1-II
170 GAMMA(I+IDLY)=GAMMA(II)
    DO 172 I=1,IDLY
    GAMMA(I+NP1)=0.0000
172 CONTINUE
C
C     CALCULATING THE EQUIVALENT CONTINUOUS DESCRIPTION
    CALL IZTOS(GAMMA,N,DELTA,IZTS)
    WRITE(6,1003)
1003 FORMAT(///,1X,100(1H-),/,1X,100(1H-))
    RETURN
    END

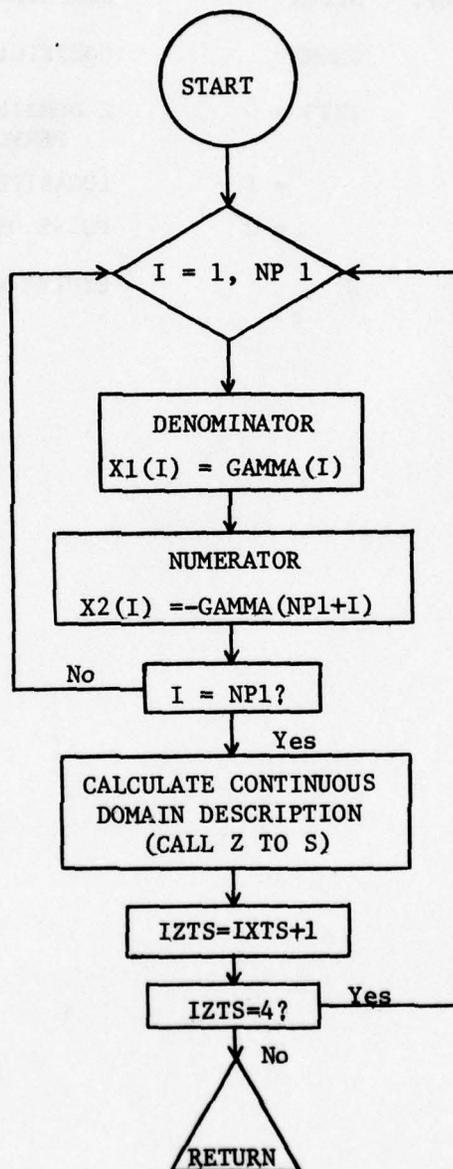
```

SUBROUTINE: IZTOS

PURPOSE: SEPARATES THE NUMERATOR FROM THE DENOMINATOR PARAMETERS IN GAMMA.

EQUATION: DENOMINATOR;  $X1(I) = \text{GAMMA}(I)$   
 NUMERATOR;  $X2(I) = \text{GAMMA}(NP1 + I)$

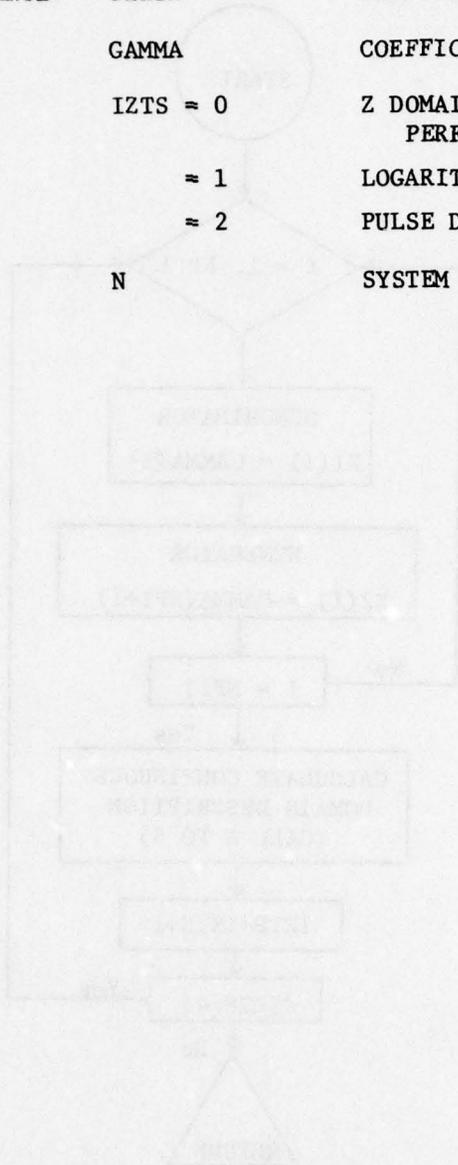
FLOW CHART:



SUBROUTINE: IZTOS

DESCRIPTION: This subroutine takes the coefficient vector GAMMA and separates the vector into the numerator (X2(I) array) and denominator (X1(I) array).

PROGRAM VARIANCE: DELTA                    SAMPLING INTERVAL  
                  GAMMA                    COEFFICIENT VECTOR  
                  IZTS = 0                Z DOMAIN TO S DOMAIN CONVERSION NOT PERFORMED  
  = 1                LOGARITHMIC TRANSFORMATION IS PERFORMED  
  = 2                PULSE DELAYED TRANSFORMATION IS PERFORMED  
                  N                        SYSTEM ORDER



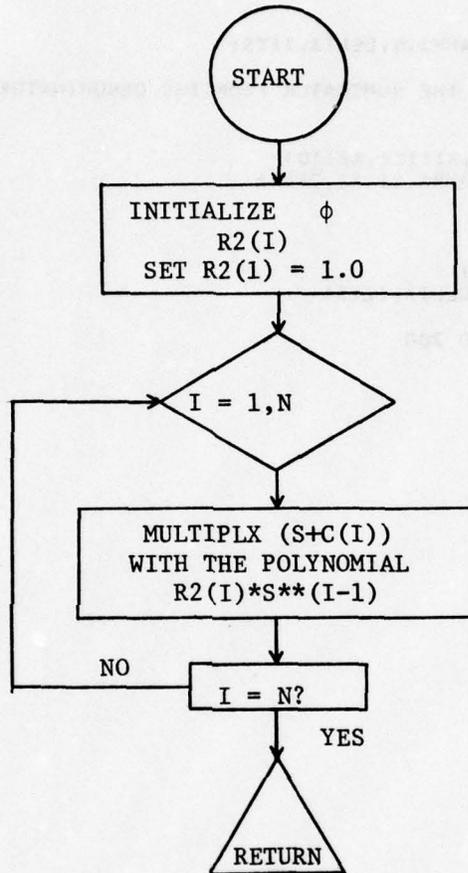
SUBROUTINE IZTOS(GAMMA,N,DELTA,IZTS)

```
C
C      IZTOS SEPARATES THE NUMERATOR FROM THE DENOMINATOR PARAMETERS
C      IN GAMMA
C
DIMENSION GAMMA(1),X1(10),X2(10)
DOUBLE PRECISION GAMMA,X1,X2,DELTA
NP1=N+1
200 DO3 I=1,NP1
    X1(I)=GAMMA(I)
3   X2(I)=-GAMMA(NP1+I)
    CALL ZTOS(X1,X2,N,DELTA,IZTS)
    IZTS=IZTS+1
    IF(IZTS.EQ.4) GO TO 200
RETURN
END
```

SUBROUTINE: POLCON

PURPOSE: CONSTRUCTS POLYNOMIAL FROM ITS ROOTS

FLOW CHART:



DESCRIPTION: This subroutine constructs a polynomial from its roots and the polynomial coefficients are stored in an array R2(I) in ascending order.

PROGRAM VARIABLES

C	ROOTS USED TO FORM POLYNOMIAL
K	OPTION WHICH SUPPRESSES POLYNOMIAL CONSTRUCTION WITH SPECIFIED ROOT(S)
N	ORDER OF SYSTEM
R2	COEFFICIENTS OF POLYNOMIAL

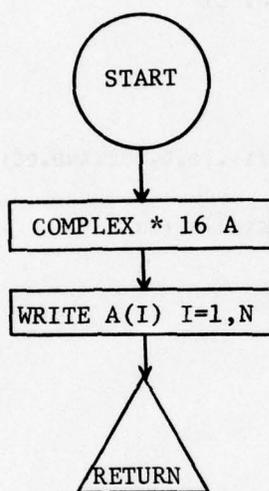


SUBROUTINE: PRCVEC

PURPOSE: This subroutine prints out a complex single dimensioned array.

EQUATION: Complex number  $A + BJ$  is printed (A, BJ)

FLOW CHART:



DESCRIPTION: This subroutine is called in the ZTOS subroutine when the poles and zeroes in the S domain are needed.

PROGRAM VARIABLES:    A                    ARRAY TO BE OUTPUT  
                           N                    NUMBER OF ELEMENT IN ARRAY

```

SUBROUTINE PROCVEC(A,N)
C
C THIS SUBROUTINE PRINTS OUT A COMPLEX SINGLE DIMENSIONED ARRAY
C A COMPLEX NUMBER OF THE FORM A + B J IS OUTPUTTED IN THE FORM
C ( A, B J) WHERE J = SQUARE ROOT OF -1
DIMENSION A(1)
COMPLEX*16 A
WRITE(6,2)
WRITE(6,1)(A(I),I=1,N)
1 FORMAT(1X,1H(,D17.10,1H,,D17.10,3H J))
WRITE(6,2)
WRITE(6,2)
2 FORMAT(/)
RETURN
END

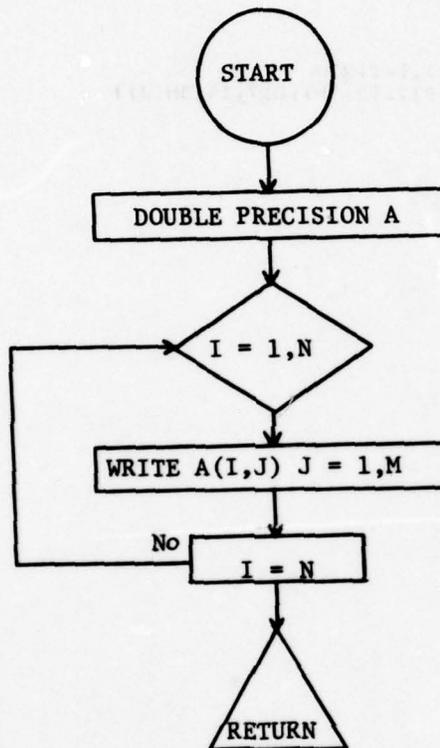
```



SUBROUTINE: PRMAT

PURPOSE: SUBROUTINE OUTPUTS DOUBLE PRECISION DOUBLE DIMENSION ARRAY

FLOW CHART:



DESCRIPTION: This subroutine is called in GRAM II and takes an array of two dimensions and gives an output of the same two dimensional array in double precision.

PROGRAM VARIABLES:	A	OUTPUT DOUBLE PRECISION ARRAY
	M	MATRIX "A" COLUMN DIMENSION
	N	MATRIX "A" ROW DIMENSION
	NMAX	DIMENSION SIZE OF TWO DIMENSIONAL ARRAY

```
      SUBROUTINE PRMAT(A,N,M,NMAX)
      DOUBLE PRECISION A
      C
      THIS SUBROUTINE OUTPUTS DOUBLE PRECISION DOUBLE DIMENSIONED ARRAY
      DIMENSION A(NMAX,1)
      WRITE(6,1)
      D32I=1,N
      2  WRITE(6,3)(A(I,J),J=1,M)
      3  FORMAT(1X,10013.5)
      WRITE(6,1)
      WRITE(6,1)
      1  FORMAT(/)
      RETURN
      END
```



```

SUBROUTINE PRVEC(A,N)
C
C THIS SUBROUTINE OUTPUTS DOUBLE PRECISION SINGLE DIMENSIONED ARRAY
DIMENSION A(1)
DOUBLE PRECISION A
WRITE(6,31)
WRITE(6,1)(A(I),I=1,N)
1  FORMAT(1X,10D13.5)
WRITE(6,31)
WRITE(6,31)
31  FORMAT(/)
RETURN
END

```

SUBROUTINE: RESPON

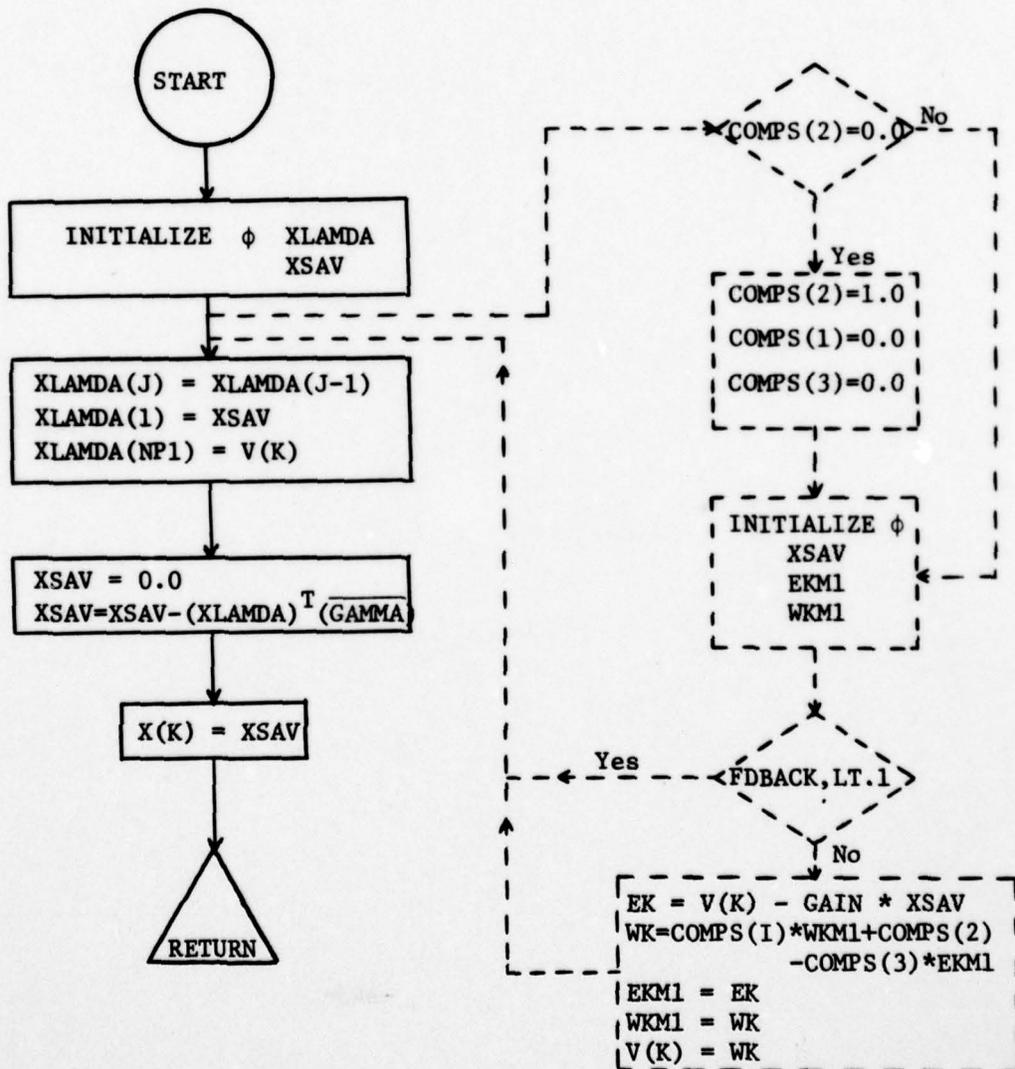
PURPOSE: CALCULATES RESPONSE ( $X_K$ ) FROM COEFFICIENT VECTOR (GAMMA)  
TIMES THE ARRAY XLAMDA.

EQUATIONS:  $[X_K, (XLAMDA)] [GAMMA] = 0$

$$\text{or } X_K = - [X_{k-1} \dots X_{k-n} \quad v_k \dots v_{k-n}] \begin{bmatrix} \gamma(2) \\ \vdots \\ \gamma(N+N+2) \end{bmatrix}$$

$$\text{also } X_K = - (XLAMDA)^T (\overline{GAMMA})$$

FLOW CHART:



SUBROUTINE: RESPON

DESCRIPTION: Subroutine RESPON determines the response of

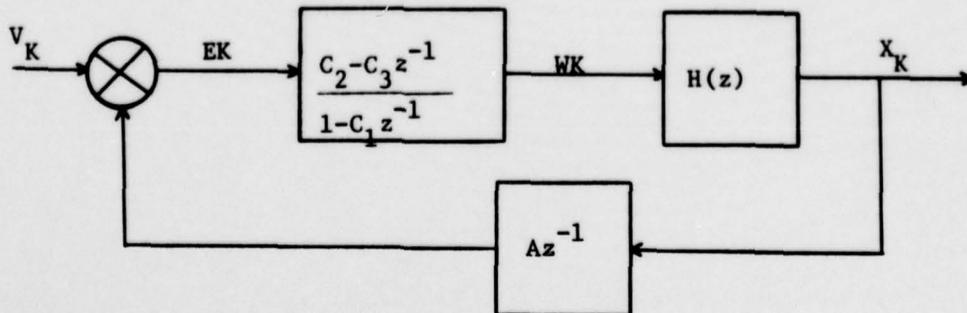
$$H(z) = \frac{b_0 + b_1 z^{-1} + b_2 z^{-2} + \dots + b_n z^{-n}}{1 + a_1 z^{-1} + a_2 z^{-2} + \dots + a_n z^{-n}} \text{ to the}$$

input sequence  $V(K)$ . The coefficients are entered as an NPNP2 =

$N + N + 2$  vector  $\text{GAMMA} = (1, a_1, \dots, a_n, -b_0, \dots, -b_n)$ .

PROGRAM VARIABLES:	FDBACK	NO FEEDBACK FDBACK = 0
		NEGATIVE FEEDBACK FDBACK = 1
	GAMMA	COEFFICIENT VECTOR
	MP1	NUMBER OF DATA POINTS
	N	MODEL ORDER
	V	INPUT SEQUENCE
	X	OUTPUT SEQUENCE
	XLAMDA	WORKING ARRAY

FEEDBACK AND COMPENSATION:



SUBROUTINE: RESPON

PURPOSE: CALCULATES RESPONSE ( $x_k$ ) FOR THE SYSTEM SHOWN IN ABOVE FIGURE.  
 THIS ADDITION TO SUBROUTINE RESPON INCORPORATES THE FLEXABILITY  
 OF ADDING NEGATIVE FEEDBACK AND CASCADE COMPENSATOR IN THE  
 FORWARD LOOP FOR OPTIMIZATION.

EQUATION:  $E_k = V_k - Ax_{k-1}$

$W_k = WKMI * COMPS (1) + E_k * COMPS (2) - COMPS (3) * EKMI$

FLOW CHART: (See flow chart for RESPON the dotted section is for Feedback  
 and Compensation network.)



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```

SUBROUTINE RESPONSE(X,V,GAMMA,KLAMDA,MP1,FDBACK)
DIMENSION X(1),V(1),GAMMA(1),KLAMDA(1)
REAL*8 XSAV,GAMMA,KLAMDA,GAIN
INTEGER FDBACK
    
```

THESE CARDS FOR DETERMINING A STABLE FEEDBACK SYSTEM  
 FEEDBACK GAIN CALLED 'GAIN' IS DETERMINED

```

REAL*8 CN,GMN,GM,TEM,      KCOF(2),ROOTR(2),ROOTI(2),COF(2)
REAL*8 COMPS
COMMON /COMPRN/COMPS(1)
GAIN=-.18
NP1=N+1
NP2=N+2
ISTABL=0
    
```

1000000

1000000

19

899

211

21

22

23

20

```

NM1=N-1
NP1=N+1
NPNP1=N+N+1
NPNP2=N+N+2
DO19I=1,NPNP1
KLAMDA(1)=0.000
IF(COMPS(2).NE.0.)GO TO 899
COMPS(2)=1.0
COMPS(1)=2.0
COMPS(3)=0.0
CONTINUE
KSAV=0.0000
TKM1=0.
WKM1=0.
DO20K=1,MP1
IF(FDBACK.LT.1)GO TO 211
EK=V(K)-GAIN*KSAV
WK=COMPS(1)*WKM1+COMPS(2)*EK-COMPS(3)*EKM1
WKM1=WK
EKM1=EK
V(K)=WK
DO21I=1,NM1
J=NPNP1-I
KLAMDA(J)=KLAMDA(J-1)
DO22I=1,N
J=NPNP2-I
KLAMDA(I)=KLAMDA(I-1)
KLAMDA(1)=KSAV
KLAMDA(NP1)=V(K)
KSAV=0.0000
DO23I=1,NPNP1
KSAV=KSAV-GAMMA(I+1)*KLAMDA(I)
IF(DABS(KSAV).GT.1.001)GO TO 27
X(K)=KSAV
RETURN
    
```

27

28

```

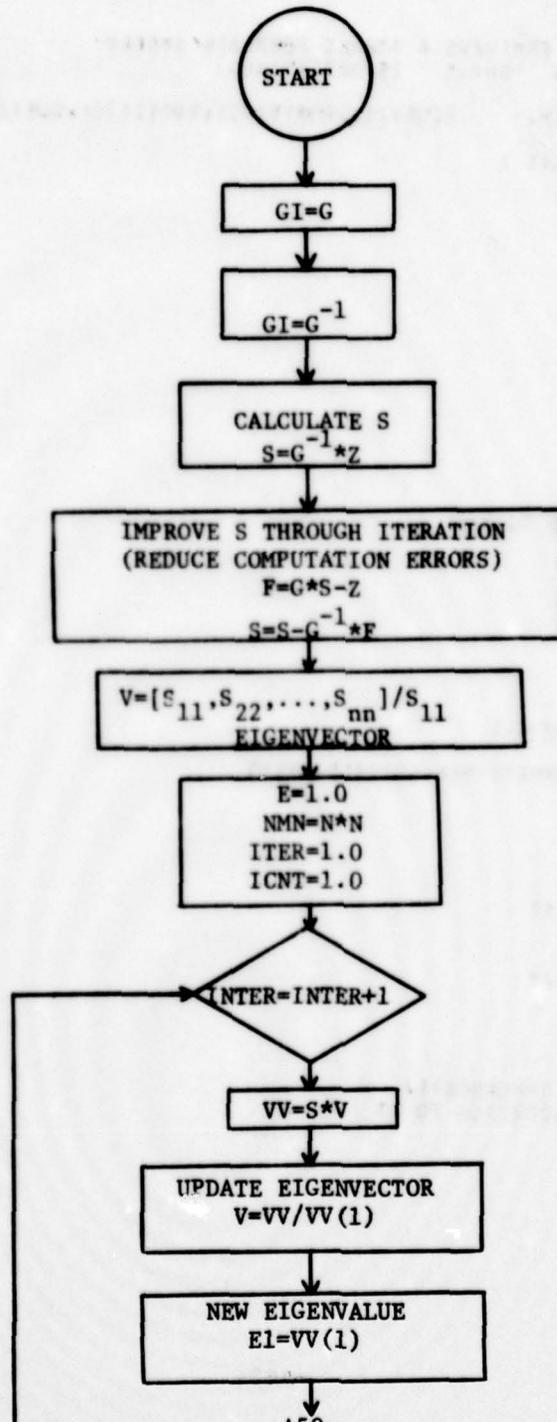
DO28I=K,MP1
X(I)=0.0
RETURN
END
    
```

SUBROUTINE: SOLVE1

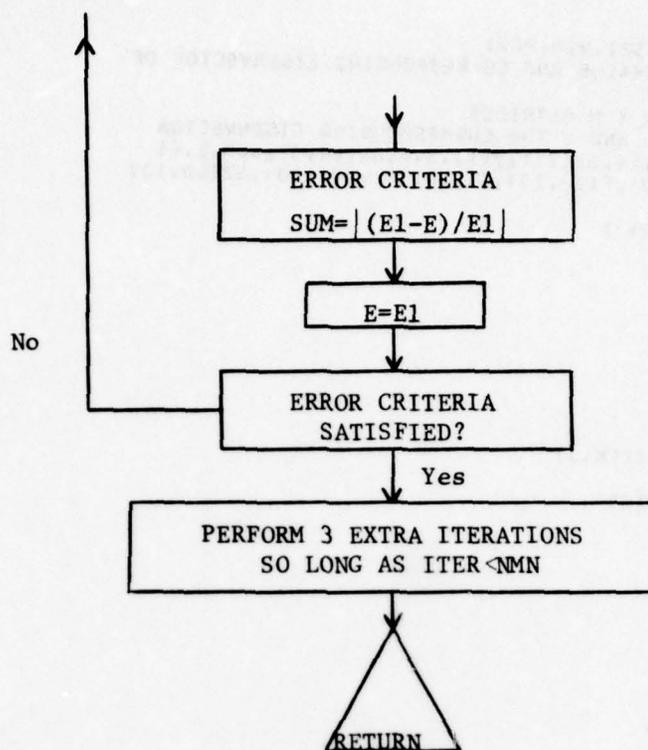
PURPOSE: FINDS MAXIMUM EIGENVALUE AND CORRESPONDING EIGENVECTOR OF

$$(Z-E*G)*V=0$$

FLOW CHART:



SUBROUTINE: SOLVE1



DESCRIPTION: Subroutine SOLVE1 uses the first quadrant of the GRAM matrix and noise correction matrix " " to calculate the output portion of the EIGENVECTOR (N+1 elements) plus the maximum EIGENVALUE .

PROGRAM VARIABLES:

G	FIRST QUADRANT OF GRAM MATRIX
MAX	MAXIMUM ROWS PERMISSIBLE
N	N+1 (ORDER OF SYSTEM + 1)
S21	COEFFICIENT MATRIX
V	EIGENVECTOR
Z	NOISE CORRECTION MATRIX

```

SUBROUTINE SOLVE1(Z,G,S21,V,N,MAX)
  FINDS MAXIMUM EIGENVALUE AND CORRESPONDING EIGENVECTOR OF
  ( Z - E*G ) * V = 0
  WHERE Z AND G ARE N X N MATRICES
  E IS THE EIGENVALUE AND V THE CORRESPONDING EIGENVECTOR
  REAL*8 Z(MAX,1),G(MAX,1),S21(1),V(1),S,F,GI,VV,S2,SUM,E,E1
  COMMON /MATRIX/S(2),20),F(20,20),GI(20,20),VV(100),S2(10,10)

  CALCULATE S=(G INVERSE)*Z

  DO1I=1,N
  DO1J=1,N
  1  GI(I,J)=G(I,J)
  CALL DOBINV(GI,N,MAX)
  DO2I=1,N
  DO2J=1,N
  S(I,J)=0.000
  DO2K=1,N
  2  S(I,J)=S(I,J)+GI(I,K)*Z(K,J)

  IMPROVE S THRU ITERATION
  F=G*S-Z
  S=S-(G INVERSE)*F

  ND2=N
  DO100ITER=1,ND2
  DO3I=1,N
  DO3J=1,N
  SUM=0.000
  DO4K=1,N
  4  SUM=SUM+G(I,K)*S(K,J)
  3  F(I,J)=SUM-Z(I,J)
  DO5I=1,N
  DO5J=1,N
  SUM=0.000
  DO6K=1,N
  5  SUM=SUM+GI(I,K)*F(K,J)
  5  S(I,J)=S(I,J)-SUM
  100 CONTINUE

  INITIALLY V=(S(1,1), . . . , S(N,N))/S(1,1), E=1
  ITERATE: EVFC VV=S*V, EVAL E1=VV(1)

  DO7I=1,N
  7  V(I)=S(I,I)/S(1,1)
  E=1.000
  NMN=N*N
  ITER=1
  ICNT=1
  8  ITER=ITER+1
  DO9I=1,N
  VV(I)=0.000
  DO9J=1,N
  9  VV(I)=VV(I)+S(I,J)*V(J)
  DO10I=1,N

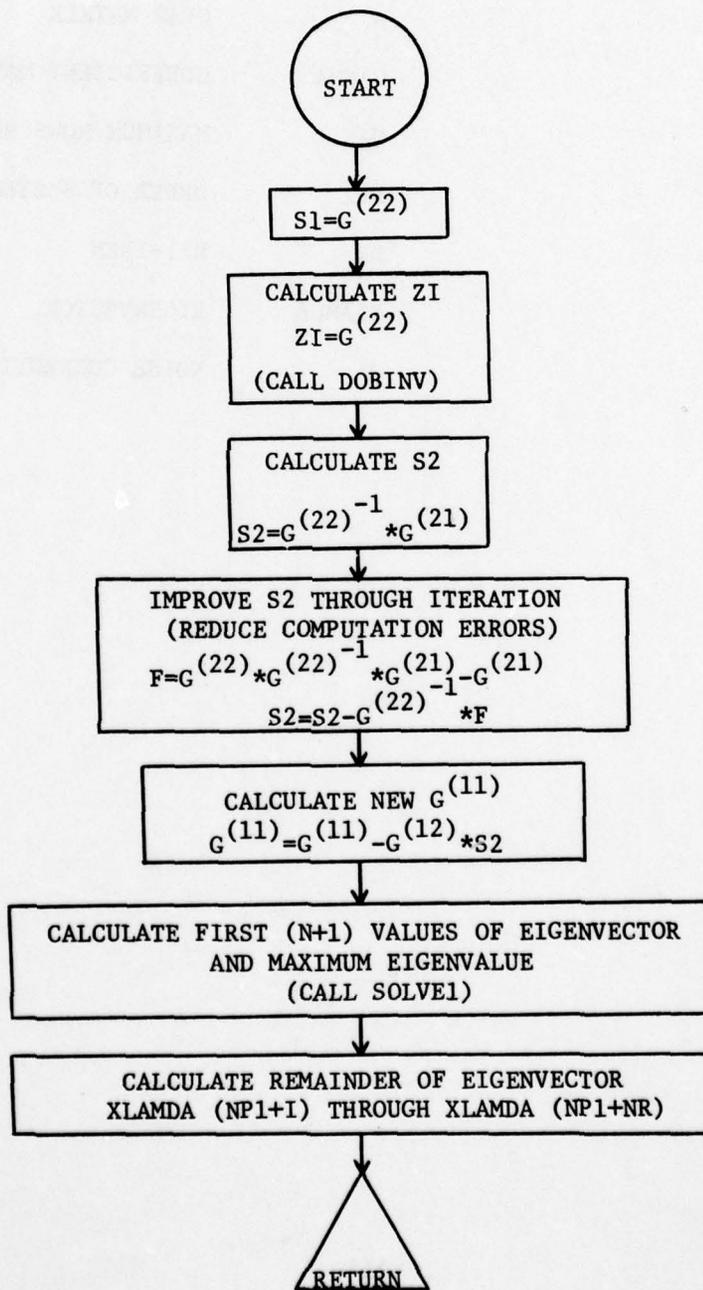
  10  V(I)=VV(I)/VV(1)
  E1=VV(1)
  SUM=DABS((E1-E)/E1)
  E=E1
  IF(SUM.GT.1.0D-8.AND.ITER.LT.NMN)GO TO 8
  ICNT=ICNT+1
  IF(ICNT.LT.5.AND.ITER.LT.NMN)GO TO 8
  WRITE(6,11)ITER,E,SUM
  11  FORMAT(10X,'ITER=',I3,' MAX.EIGENVALUE=',D13.6,' ERROR=',D13.6)
  RETURN
  END

```

SUBROUTINE: SOLVE2

PURPOSE: CALCULATE MAXIMUM EIGNEVALUE AND EIGENVECTOR

FLOW CHART:



SUBROUTINE: SOLVE2

DESCRIPTION: Subroutine SOLVE2 calculates the maximum EIGENVALUE and EIGENVECTOR when noise is present on both input and output or the noise is present only on the output. SOLVE2 uses subroutine SOLVE1 to calculate the first (N+1) EIGENVALUES and the maximum EIGENVALUE. SOLVE2 uses an iteration process (similar to SOLVE1 and SOLVE3) to reduce computation errors.

PROGRAM VARIABLES:

G	GRAM MATRIX
GAMMA	COEFFICIENT MATRIX
MAX	MAXIMUM ROWS PERMISSIBLE
NP1	ORDER OF SYSTEM+1 (N+1)
NR	NP1-IREM
XLAMDA	EIGENVECTOR
Z	NOISE CORRECTION MATRIX

AD-A050 206

UNIVERSITY OF SOUTH FLORIDA TAMPA  
THEORY AND COMPUTER PROGRAM FOR TRANSFER FUNCTION IDENTIFICATION--ETC(U)  
FEB 78 V K JAIN, G J DOBECK, L J LAUDERMILT N61331-75-C-0012  
NCSL-TM-204-78 NL

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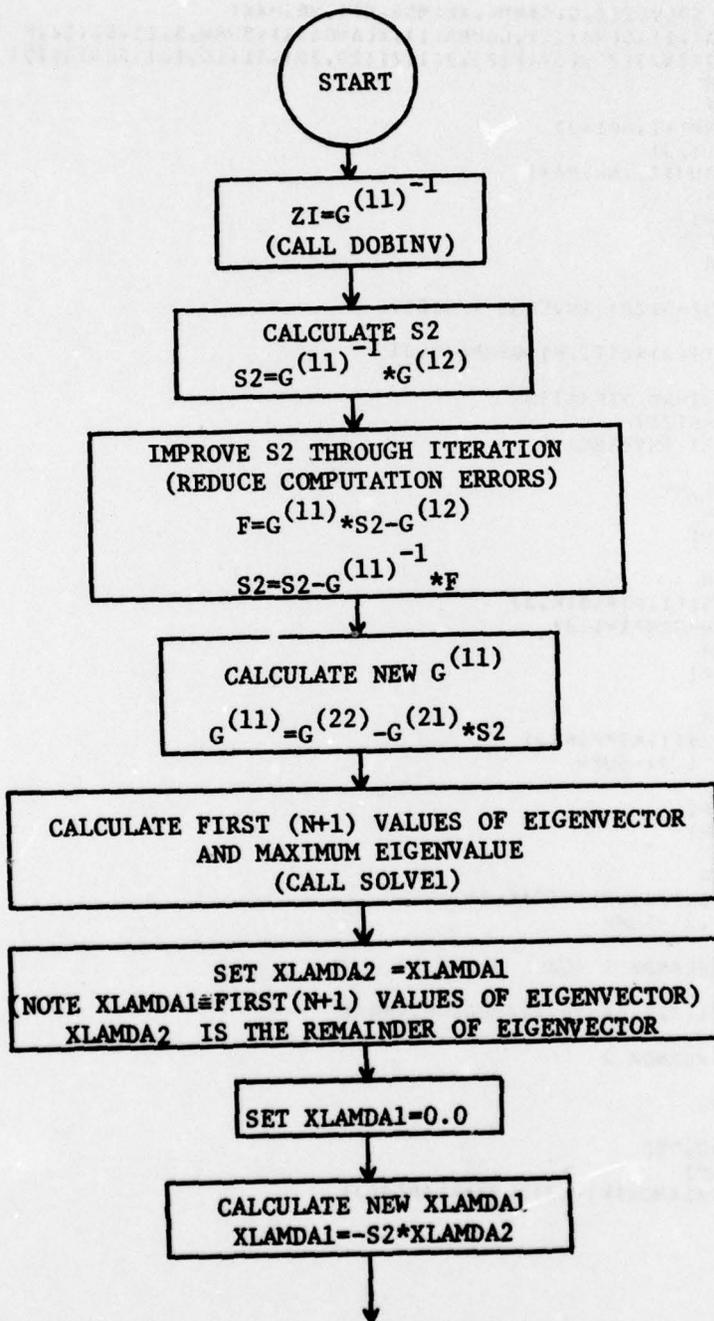
SUBROUTINE SOLVE2(Z,G,GAMMA,XLAMDA,NP1,NR,MAX)
REAL*8 Z(MAX,1),G(MAX,1),GAMMA(1),XLAMDA(1),SUMW,S,ZI,S1,S2,F
COMMON /MATRIX/S(20,20),F(20,20),ZI(20,20),S1(10,10),S2(10,10)
DO620I=1,NR
DO620J=1,NR
620 ZI(I,J)=G(NP1+I,NP1+J)
S1(I,J)=ZI(I,J)
CALL DORINV(ZI,NR,MAX)
DO621I=1,NR
DO621J=1,NP1
S2(I,J)=0.0D0
DO621K=1,NR
C
C CALCULATE S2=G(22) INVERSE * G(21)
C
621 S2(I,J)=S2(I,J)+ZI(I,K)*G(NP1+K,J)
C
C IMPROVE S2 THRU ITERATION
C
C F=G(22)*S2-G(21)
C
C S2=S2-(G(22) INVERSE)*F
C
DO624ITER=1,NR
DO622I=1,NR
DO622J=1,NP1
SUMW=0.0D0
DO623K=1,NR
623 SUMW=SUMW+S1(I,K)*S2(K,J)
622 F(I,J)=SUMW-G(NP1+I,J)
DO625I=1,NR
DO625J=1,NP1
SUMW=0.0D0
DO626K=1,NR
626 SUMW=SUMW+ZI(I,K)*F(K,J)
625 S2(I,J)=S2(I,J)-SUMW
624 CONTINUE
DO627I=1,NP1
DO627J=1,NP1
SUMW=0.0D0
DO628K=1,NR
628 SUMW=SUMW+G(I,K+NP1)*S2(K,J)
627 G(I,J)=G(I,J)-SUMW
C
C CALCULATE XLAMDA 1 (CALL SOLVE 1)
C
CALL SOLVE1(Z,G,GAMMA,XLAMDA,NP1,MAX)
C
C CALCULATE XLAMDA 2
C
DO629I=1,NR
K=NP1+I
XLAMDA(K)=0.0D0
DO629J=1,NP1
629 XLAMDA(K)=XLAMDA(K)-S2(I,J)*XLAMDA(J)
RETURN
END

```

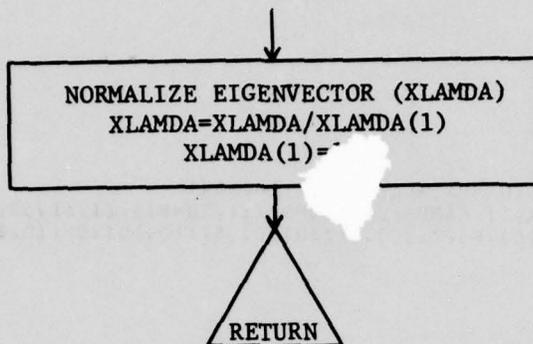
SUBROUTINE: SOLVE 3

PURPOSE: CALCULATE MAXIMUM EIGENVALUE AND CORRESPONDING EIGENVECTOR  
(NOISE ON INPUT ONLY)

FLOW CHART:



SUBROUTINE: SOLVE3



DESCRIPTION: SOLVE3 is used to calculate the EIGENVECTOR when the system has noise only on input. SOLVE3 uses SOLVE1 to calculate the maximum EIGENVALUE and the (N+1) EIGENVALUES. The remaining EIGENVECTOR elements are calculated in subroutine SOLVE3. SOLVE3 uses an iteration process (similar to SOLVE1 and SOLVE2) to reduce computation errors.

## PROGRAM VARIABLES:

G	GRAM MATRIX
GAMMA	COEFFICIENT MATRIX
MAX	MAXIMUM ROWS PERMISSIBLE
NP1	SYSTEM ORDER +1, (N+1)
NR	NP1-IREM
XLAMDA	EIGENVECTOR
Z	NOISE CORRECTION MATRIX

```

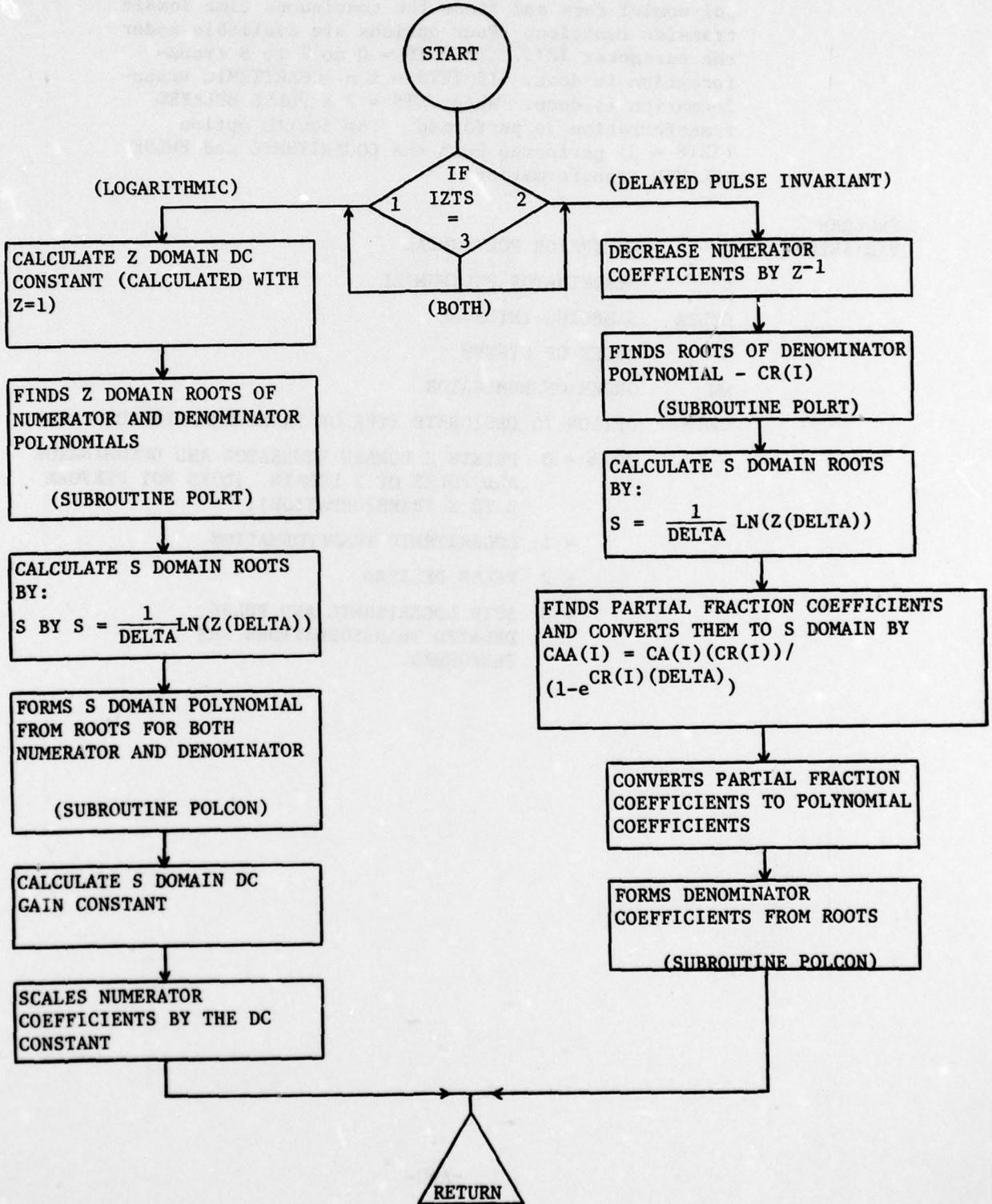
SUBROUTINE SOLVE3(Z,G,GAMMA,XLAMDA,NP1,NR,MAX)
REAL*8 Z(MAX,1),G(MAX,1),GAMMA(1),XLAMDA(1),SUMW,S,ZI,S1,S2,F
COMMON /MATRIX/S(20,20),F(20,20),ZI(20,20),S1(10,10),S2(10,10)
NPNP2=NP1+NR
DO1I=1,NP1
DO1J=1,NP1
1  ZI(I,J)=G(I,J)
   CALL DOBINV(ZI,NP1,MAX)
   DO2I=1,NP1
   DO2J=1,NR
   S2(I,J)=0.0D0
   DO2K=1,NP1
2  S2(I,J)=S2(I,J)+ZI(I,K)*G(K,J+NP1)
   DO3ITER=1,NP1
   DO4I=1,NP1
   DO4J=1,NR
   SUMW=0.0D0
   DO5K=1,NP1
5  SUMW=SUMW+G(I,K)*S2(K,J)
4  F(I,J)=SUMW-G(I,J+NP1)
   DO6I=1,NP1
   DO6J=1,NR
   SUMW=0.0D0
   DO7K=1,NP1
7  SUMW=SUMW+ZI(I,K)*F(K,J)
6  S2(I,J)=S2(I,J)-SUMW
3  CONTINUE
   DO8I=1,NR
   DO8J=1,NR
   SUMW=0.0D0
   DO9K=1,NP1
9  SUMW=SUMW+G(I+NP1,K)*S2(K,J)
8  G(I,J)=G(I+NP1,J+NP1)-SUMW
   CALL SOLVE1(7,G,GAMMA,XLAMDA,NR,MAX)
   DO10I=1,NR
10  XLAMDA(NP1+I)=XLAMDA(I)
   DO11I=1,NP1
   XLAMDA(I)=0.0D0
   DO11J=1,NR
11  XLAMDA(I)=XLAMDA(I)-S2(I,J)*XLAMDA(J+NP1)
   DO12I=2,NPNP2
12  XLAMDA(I)=XLAMDA(I)/XLAMDA(1)
   XLAMDA(1)=1.0D0
   RETURN
   END

```

SUBROUTINE: ZTOS

PURPOSE: CONVERTS A DISCRETE TIME TRANSFER FUNCTION H(Z) TO A CONTINUOUS TIME TRANSFER FUNCTION H(S).

FLOW CHART:



## SUBROUTINE ZTOS

DESCRIPTION: This subroutine uses the  $H(Z)$  transfer function in polynomial form and finds the continuous time domain transfer function. Four options are available under the parameter IZTS. If IZTS = 0 no Z to S transformation is done. If IZTS = 1 a LOGARITHMIC transformation is done. When IZTS = 2 a PULSE DELAYED transformation is performed. The fourth option (IZTS = 3) performs both the LOGARITHMIC and PULSE DELAYED transformations.

## PROGRAM

## VARIABLES:

A	NUMERATOR POLYNOMIAL
B	DENOMINATOR POLYNOMIAL
DELTA	SAMPLING INTERVAL
N	ORDER OF SYSTEM
NN	ORDER OF NUMERATOR
IZTS	OPTION TO DESIGNATE TYPE OF TRANSFORMATION DESIRED
	IZTS = 0 PRINTS Z DOMAIN NUMERATOR AND DENOMINATOR AND POLES OF Z DOMAIN. (DOES NOT PERFORM Z TO S TRANSFORMATION).
	= 1 LOGARITHMIC TRANSFORMATION
	= 2 PULSE DELAYED
	= 3 BOTH LOGARITHMIC AND PULSE DELAYED TRANSFORMATIONS ARE PERFORMED.



```

IF(NN.EQ.0)GO TO 469
CALL POLRT(A,TEMP,NN,RR,RI,IER)
DO15 I=1,NN
15 CA(I)=DCMLX(RR(I),RI(I))
DO 7 I=1,NN
7 CA(I)=(+1.0/DELTA)*CLOG(CA(I))
IF(NN.EQ.N) GOTO471
469 CONTINUE
DO 470 I=NNP1,NP1
CAA(I)=0.000
470 CA(I)=0.000
471 CONTINUE
IF(NN.EQ.0)CAA(1)=1.000

```

C  
C  
C  
C  
C

NOW THE FIRST NN ENTRIES OF CA CONTAIN THE S-DOMAIN ZEROS OF NUMERATOR AND THE REMAINING ENTRIES ARE ZEROED OUT.

```
IF(NN.NE.0)CALL POLCON(CA,CAA,0,N)
```

C  
C  
C  
C  
C

WORK ON DENOMINATOR

```

CALL POLRT(B,TEMP,N,RR,RI,IER)
DO16 I=1,N
CR(I)=DCMLX(RR(I),RI(I))
16 CF(I)=1.0000/CR(I)
909 WRITE(6,1002)
CALL PRCVEC(CF,N)
IF(IZTS.EQ.0) GO TO 900
235 DO6 I=1,N
6 CR(I)=(-1.0/DELTA)*CLOG(CR(I))
WRITE(6,240)
240 FORMAT(' LOGARITHMIC TRANSFORMATION')
WRITE(6,999)
WRITE(6,2000)
2000 FORMAT(' POLES IN S DOMAIN')
CALL PRCVEC(CR,N)
DO3000I=1,N
3000 CR(I)=-CR(I)
CALL POLCON(CR,CB,0,N)

```

C  
C  
C  
C  
C

ADJUST DC GAIN CONSTANT

```

A2=CAA(1)
B2=CB(1)
FAC=(A1/B1)*(B2/A2)
DO 603 I=1,NNP1
603 CAA(I)=CAA(I)*FAC
GO TO 2010

```

C  
C  
C  
C  
C  
C  
C

DELAYED PULSE INVARIANT TRANSFORMATION

SHIFTS NUMERATOR COEFFICIENTS FOR DELAY

```

250 CONT = A(1)
    DO 300 I=1,N
300  A(I)=A(I+1)-CONT*B(I+1)
    A(NP1)=0.0
400  CALL POLRT(B,TEMP,N,RR,RI,IER)
    DO61I=1,N
    CR(I)=DCMLX(RR(I),RI(I))
61   CF(I)=1.0000/CR(I)
    WRITE(6,1002)
1002 FORMAT(1X,'THE POLES OF THE Z-DOMAIN')
    CALL PRCVEC(CF,N)
C
C
C   PARTIAL FRACTION EXPANSION
C
C
    DO31=1,N
    CON1=1.0000
    CON2=0.0000
    DO4J=1,N
    CON2=CON2*CR(I)+A(N-J+1)
    IF(I-J)5,4,5
5    CON1=CON1*(1.0000-CR(I)*CF(J))
4    CONTINUE
3    CA(I)=CON2/CON1
C
C
C   TRANSFORMATION OF DENOMINATOR AND NUMERATOR
C
C
224  DO21=1,N
    CON1=CDLOG(CR(I))/DELTA
    CA(I)=CA(I)*CR(I)*CON1/(CR(I)-1.0000)
2    CR(I)=CON1
    WRITE(6,241)
241  FORMAT(' DELAYED PULSE TRANSFORMATION')
    WRITE(6,999)
226  WRITE(6,1004)
1004 FORMAT(' NEGATIVE OF THE POLES IN THE S-DOMAIN')
    CALL PRCVEC(CR,N)
    WRITE(6,1003)
1003 FORMAT(1X,'NUMERATOR CONSTANTS OF FACTORIZED H(S)')
    CALL PRCVEC(CA,N)
    CALL POLCON(CR,CB,O,N)
    DO71I=1,NP1
71   CAA(I)=0.0000
    DO9K=1,N
    CALL POLCON(CR,CF1,K,N)
    DO9J=1,N
9    CAA(J)=CAA(J)+CF1(J)*CA(K)
    CAA(NP1)=0.0000
2010 CONTINUE
    DO450I = 1,NP1
    CAA(I)= CAA(I)+CONT*CB(I)
C
403  WRITE(6,1005)
1005 FORMAT(' S-DOMAIN DENOMINATOR')
    CALL PRCVEC(CB,NP1)
    WRITE(6,1006)
1006 FORMAT(' S-DOMAIN NUMERATOR')
    CALL PRCVEC(CAA,NP1)
    DO20I=1,NP1

    B(I)=CB(I)
20   A(I)=CAA(I)
900  RETURN
    END

```

(Reverse Page A64 Blank)

APPENDIX Bz TO s EQUIVALENCE TRANSFORMSLogarithmic Equivalence

Equation (3) on page 4 describes a simple way of finding an s-domain transfer function  $H(s)$  corresponding to a z-domain transfer function  $H(z)$ . It is implemented in the computer program whenever  $IZTS = 1$ . The basis for this correspondence lies in the logarithmic mapping  $s = -\frac{1}{T} \ln(z)$  of the poles and zeros from the z-plane to the s-plane. Appropriately, it is called the Logarithmic Equivalence Transform. The inverse mapping is similarly defined and we note that the left-hand-side of the s-plane maps into the interior of the unit circle in the z-plane.

The primary advantage of the logarithmic equivalence transform is that it preserves the degree of the numerator from the z-domain to the s-domain. However, it does not yield a good degree of invariance of the output between the continuous-time and discrete-time equivalent systems [8]. As a consequence, this method requires a finer sampling interval compared to the 'pulse interpolation' method (described below) in order to achieve a satisfactory invariance of the output.

Leading-Edge-Pulse Equivalence

This method aims for invariance of the output at the sampling instants. Strictly speaking this objective cannot be achieved for every arbitrary input because the sampled input signal loses some of the information of the original signal. Suitable restrictions must therefore be placed on the class of inputs for which the output invariance is sought. For example it is assumed that the bandwidth of the input signal and the highest frequency of the passband of the system are small compared to the sampling frequency (say one-tenth or smaller). Under such an assumption the input may be approximated by a train of rectangular pulses:

$$u(t) \cong \bar{u}(t) \triangleq \sum_{k=-\infty}^{\infty} u(k\Delta) p(t-k\Delta)$$

where

$$p(t) = \begin{cases} 1 & \text{for } 0 \leq t < \Delta \\ 0 & \text{otherwise} \end{cases}$$

Invariance of the outputs of a)  $H(z)$  excited by  $u(k\Delta)$  and that of b)  $H(s)$  excited by  $\bar{u}(t)$  can then be achieved by equivalencing  $H(z)$  and  $H(s)$  in the following manner:

$$H(z) = \sum_{i=1}^n \frac{\gamma_i}{1-\alpha_i z^{-1}} \quad \Leftrightarrow \quad \sum_{i=1}^n \frac{r_i}{s+p_i} = H(s),$$

$$p_i = -\frac{1}{T} \ln(\alpha_i)$$

$$r_i = \frac{\gamma_i p_i}{(1-\alpha_i)}$$

This method yields a high degree of invariance between the outputs of a)  $H(z)$  excited by  $u(k\Delta)$  and b)  $H(s)$  excited by the actual  $u(t)$ . In this respect its superiority over the logarithmic equivalence method has been demonstrated by case studies on several Navy vehicles. However, it suffers from the disadvantage that the degrees of the numerators in the z-domain and the s-domain do not, in general, equal each other.

APPENDIX C  
SOLUTION OF A KEY EQUATION

As stated in Section II three different cases for equation (15) arise depending upon whether  $\sigma_q$ ,  $\sigma_w$ , or both are nonzero. The solution for these different cases is discussed below.

Case 1: Noise on Both Input and Output (NSPQ  $\neq$  0, NSPW  $\neq$  0)

Equation (15) may be written as

$$(\mu I - G^{-1}Z)\lambda = 0 \quad (C1)$$

so that the desired solution  $\lambda$  is the eigenvector of  $G^{-1}Z$  corresponding to its largest eigenvalue.

Case 2: Noise on Output Only (NSPQ  $\neq$  0, NSPW = 0)

By partitioning the matrix  $G$  into four  $(n+1) \times (n+1)$  blocks and correspondingly partitioning  $\lambda$  one obtains

$$\begin{bmatrix} G_{11} & G_{12} \\ G_{21} & G_{22} \end{bmatrix} - \beta \begin{bmatrix} I & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \lambda^{(1)} \\ \lambda^{(2)} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad (C2)$$

which is equivalent to solving the pair

$$[(G_{11} - G_{12}G_{22}^{-1}G_{21}) - \beta I] \lambda^{(1)} = 0 \quad (C3)$$

$$\lambda^{(2)} = -G_{22}^{-1}G_{21} \lambda^{(1)} \quad (C4)$$

The first part is solved as a usual eigenvalue problem. The eigenvector  $\lambda^{(1)}$  corresponding to the minimum eigenvalue is selected, and, then, from the second equation  $\lambda^{(2)}$  is obtained. The desired parameter vector is finally obtained as

$$\hat{\lambda} = \frac{1}{\lambda^{(1)}(1)} \begin{bmatrix} \lambda^{(1)} \\ \lambda^{(2)} \end{bmatrix} \quad (C5)$$

Case 3: Noise on Input Only (NSPQ = 0, NSPW  $\neq$  0)

This case is quite similar in nature to case 2 above and is treated accordingly.

