ADVANCED TECHNIQUES FOR BLACK BOX MODELING (EFFECT OF SIGNAL QU--ETC(U))
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ADVANCED TECHNIQUES FOR BLACK BOX MODELING (EFFECT OF SIGNAL QUANTIZATION; MULTIRATE SAMPLING OF WIDEBAND SYSTEMS)

University of South Florida

V. K. Jain

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**ABSTRACT**

The pencil-of-functions method is a black box modeling method. The research described here deals with the issues of (1) effect of quantized input output signals and (2) the modeling of wideband systems. The development shows that the pencil-of-functions method together with certain statistical corrections and/or the use of multirate sampling lead to enhanced transfer function models. Also, it is shown that LF, MF and HF test results, when employed in conjunction with the procedure...
Item 20 (Cont'd)

described for adjoining smallband transfer functions, yield successful
wideband identification.
PREFACE

This effort was conducted by University of South Florida under the sponsorship of the Rome Air Development Center Post-Doctoral Program for Rome Air Development Center. Mr. John F. Spina RADC/RBCT was the task project engineer and provided overall technical direction and guidance.

The RADC Post-Doctoral Program is a cooperative venture between RADC and some sixty-five universities eligible to participate in the program. Syracuse University (Department of Electrical Engineering), Purdue University (School of Electrical Engineering), Georgia Institute of Technology (School of Electrical Engineering), and State University of New York at Buffalo (Department of Electrical Engineering) act as prime contractor schools with other schools participating via sub-contracts with prime schools. The U. S. Air Force Academy (Department of Electrical Engineering), Air Force Institute of Technology (Department of Electrical Engineering), and the Naval Post Graduate School (Department of Electrical Engineering) also participate in the program.

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Further information about the RADC-Doctoral Program can be obtained from Mr. Jacob Scherer, RADC/RBC, Griffis AFB, NY, 13441, telephone Autovon 587-2543, Commercial (315) 330-2543.
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EVALUATION

The research described here deals with the issues of (1) the effect on black box identification accuracy of quantized (noisy) input/output data, and (2) the modeling of wideband systems by frequency partitioning and the use of multirate sampling within the sub bands. The development presented here shows that the pencil-of-function method together with selected statistical corrections on the contaminated data and/or the use of multirate sampling leads to enhanced transfer function identification. The enhancement is quantitatively described in terms of normalized mean square errors between the "true" transfer function, the identified transfer function without statistical corrections and the identified transfer function with the statistical corrections.

JOHN F. SPINA
Project Engineer
ADVANCED TECHNIQUES FOR BLACK-BOX MODELING

1. INTRODUCTION

The pencil-of-functions method is a black-box modeling method [1]-[2]. Given an input, output response pair of a system under test, the algorithm leads to a comprehensive description of the system in the form of a transfer function. Although the method was originally developed for use upon linear networks, its applicability has been extended by Weiner and Ewen [3]-[4] to nonlinear Volterra models. The method has been implemented in a FORTRAN program and is available from RADC together with necessary user instructions [5]. The research described here deals with the important issues of signal quantization during analog-to-digital conversion, and the black-box modeling of wideband systems.

1. Quantization

Practical analog-to-digital (A/D) converters employ small word lengths, typically 8 to 16 bits, and, as a rule, one can trade word length for higher conversion speed, cost remaining fixed. Unfortunately, small word lengths lead to degradation in the accuracy of the identified transfer function [6]. It is shown here that the statistical properties of the quantization error can be exploited to improve the accuracy and reliability of the identified parameters. The study thus demonstrates that higher speed implementations and/or additional cost benefits may be achieved for the pencil-of-functions method than have heretofore been realized.

2. Wideband Identification

Communication systems utilize many wideband circuits, for example, amplifiers for spread-spectrum signals. Black-box modeling, or identification, of these circuits poses both a theoretical and a practical challenge. A multirate sampling approach to identification of wideband transmittances is discussed. It permits determination of the transfer function of a four-to-five decade bandwidth system from simple transient tests. Clever selection of sampling rates and test inputs reduces the wideband problem into three, simpler smallband problems. The smallband transfer functions are identified via the pencil-of-functions method and then adjoined, systematically, to construct the
wideband transfer function estimate.

The report is structured as follows. Section 2 describes the pencil-of-functions method in brief. Theoretical details are omitted, for they can be found elsewhere [1], [2]. The description is included here for convenience of the reader, and also to emphasize the discrete-time version of the method. A computer program for conversion from s to z domain transfer functions is given in Appendix A. Section 3 presents the study on improvement of quantization-caused degradation, through a statistical approach. The key to this turns out to be the determinant of the Gram matrix of the integrated signals. A computer program, "GQUANT", developed for the particular case of impulse response modeling, is given in Appendix B. Section 4 discusses the results of the study on wideband systems. Included are equations and tables for ready selection by the test engineer of inputs and sampling rates for the LF, MF and HF band transient tests. These pulse inputs have been selected after careful study and are considered both effective and laboratory realizable. A computer program, "USPEC", which generates the amplitude spectra of the recommended pulses is given in Appendix C.
2. PENCIL-OF-FUNCTIONS METHOD

Recorded input, output responses of a network can be integrated to yield a family of signals, called measurement signals. Application of the pencil-of-functions theorem [1] to this family yields, in a closed form, the identified parameters of the network function. The procedure for this black-box modeling method is described below. Although proofs are omitted, the usefulness of the technique will be demonstrated with examples. Discrete-time signals are chosen for the presentation here, because of inherent computational advantages, although such signals must often be obtained by sampling a continuous-time system.

2.1 SIMULTANEOUS NUMERATOR AND DENOMINATOR DETERMINATION

Identification Problem

Given the input-output observations
\[ \{u(k), y(k)\}, \quad k=0,1,\ldots,K \]  
(1) arising from a physical system (see Fig. 1) believed to be linear, and of finite order, it is desired to find a system model

\[ H(z) = \frac{b_1 z^{-1} + \ldots + b_n z^{-n}}{1 + a_1 z^{-1} + \ldots + a_n z^{-n}} \]  
(2)

\[ = \frac{d_1 z^{-1}}{1 - c_1 z^{-1}} \]  
(3)

which best fits the observations, in some sense (see Fig. 2). A solution can be obtained by use of the pencil-of-functions theorem as discussed below.

For convenience denote sequences \( \{u(k)\} \) and \( \{y(k)\} \) simply as \( u \) and \( y \), respectively. Also, denote the inner-product of two sequences as

\[ \langle x, y \rangle = \sum_{k=0}^{K} x(k) y(k) \]  
(4)

Measurement Sequences

From the given sequences \( y \) and \( u \) we form the following set of sequences, called measurement sequences:
Fig. 1. Response-pair from system under test

Fig. 2. Identification problem: Find $H(z)$ so that $\hat{y}(k)$ is close to $y(k)$
\[ y_1(k) = y(k) \]
\[ y_2(k) = y_1(0)+y_1(1)+\ldots+y_1(k) \]
\[ \ldots \]
\[ y_{n+1}(k) = y_n(0)+y_n(1)+\ldots+y_n(k) \]  \hspace{1cm} (5)

\[ u_1(k) = u(k-1) \]
\[ u_2(k) = u_1(0)+u_1(1)+\ldots+u_1(k) \]
\[ \ldots \]
\[ u_{n+1}(k) = u_n(0)+u_n(1)+\ldots+u_n(k) \]  \hspace{1cm} (6)

where \( n \) is the order of the model desired. That is, \( n \) is the degree of the network function \( H \).

Note that these sequences represent repeated discrete integrations of the observed signals \( y(k) \) and \( u(k) \), respectively, i.e.,

\[ y_{j+1}(k) = \sum_{\ell=0}^{k} y_j(\ell) \quad j=1, \ldots, n \]  \hspace{1cm} (7)

\[ u_{j+1}(k) = \sum_{\ell=0}^{k} u_j(\ell) \quad j=1, \ldots, n \]  \hspace{1cm} (8)

Equivalently, \( y_{j+1}(k) \) is obtained by passing \( y_j(k) \) through the filter \( I(z) = z/(z-1) \) as shown in Fig. 3. Likewise, \( u_{j+1}(k) \) is obtained by passing \( u_j(k) \) through the discrete integrator \( I(z) \).

**Gram Matrix**

Next form the following inner-product matrix
Digital integrator $I(z) = \frac{z}{z - 1}$

Fig. 3. Generation of integrated signals and the Gram matrix
(u(k) included in the inner-product generator if model is required to have direct transmission)
where we have used the notation \( N = n+1 \) for convenience. This \((N+n) \times (N+n)\) dimensional matrix is the Gram matrix [10] of the \((N+n)\) dimensional vector sequence

\[
\{\hat{f}(k)\}, \quad k = 0,1,...,K
\]  

where

\[
\hat{f}(k) = \begin{bmatrix}
  y_1(k) \\
  y_2(k) \\
  \vdots \\
  y_N(k) \\
  u_2(k) \\
  \vdots \\
  u_N(k)
\end{bmatrix}
\]  

To state this observation formally, we have

\[
F = \sum_{k=0}^{K} \hat{f}(k)\hat{f}^T(k) \quad \text{(12)}
\]  

The entry \( u_1(k) \) is omitted in \( f(k) \), and therefore in the formation of the gram matrix \( F \), whenever direct transmission in the model is absent (that is when the coefficient \( b_0 \) in the function \( H(z) \) is constrained to be zero).
Diagonal Cofactors

Denote the diagonal cofactors of $F$ as $D_i$:

$$D_i = i,i \text{ cofactor of } F \quad (13)$$

Recall that the $i,i$ cofactor of a square matrix is the determinant of the matrix after deleting the $i$th row and the $i$th column.

Parameters of the Network Function

The parameters of the network function are given by the square-roots of $D_i$ up to a multiplicative constant. That is

$$\left[ \sum_{i=1}^{N} \sqrt{D_i} (1 - z^{-1}) \right] Y(z) = \left[ \sum_{i=1}^{N} \sqrt{D_{N+1}} (1 - z^{-1}) \right] U(z) \quad (14)$$

which can be normalized, by dividing by $D = \sqrt{D_1} + \ldots + \sqrt{D_N}$, so that the leading coefficient becomes unity. Clearly the computed transfer function becomes

$$H(z) = \frac{z^{-1} \left[ \sum_{i=1}^{N} \sqrt{D_{N+1}} (1 - z^{-1}) \right] / D}{\left[ \sum_{i=1}^{N} \sqrt{D_i} (1 - z^{-1}) \right] / D} \quad (15)$$

REMARKS

* Note that the first measurement signal is the network output itself, $y_1 = y$. Next follow its successive integrations. Each of these signals can be expressed directly in terms of $y(k)$. Indeed, if we let $I(z) = z/(z-1)$, $Y_{j+1}(z) = I_j(z) Y(z)$ so that $y_{j+1}(k) = i_j(k) \circ y(k)$ where $i_j(k)$ is the inverse transform of $I_j(z)$ and $\circ$ denotes discrete-time convolution.

* The dimensionality of the measurement vector $\mathbf{f}(k)$ is $2n+1 = N+n$ when the direct transmission term $b_o$ is constrained to be zero. If the network does have direct transmission, $u_1(k) = u(k)$ should be included in the vector $\mathbf{f}$ so that its dimensionality, as well as that of the corresponding Gram matrix $F$, becomes $2n+2 = 2N$. The right hand side of equation (14) modifies slightly as follows

$$\left[ \sum_{i=1}^{N} \sqrt{D_i} (1 - z^{-1}) \right] Y(z) = \left[ \sum_{i=1}^{N} \sqrt{D_{N+1}} (1 - z^{-1}) \right] U(z) \quad (16)$$

The counterpart of equation (15) follows from (16) and is therefore not given here.
To illustrate the steps of the method, a simple example is given next. (The reader, unfamiliar with the pencil-of-functions method, may wish to work the details with pencil and paper; others may skip this example.)

Example 1

Consider the setup of Fig. 4 where $u_1(k)$ denotes the input signal and $y_1(k)$ the output. The network is known to have direct transmission and of first order (i.e., the s-domain transfer function is of the type $(d_1s + d_0)/(s + c_0)$). The measurements are made every 1 ms for 5 samples, $k = 0, 1, \ldots, 4$.

Unit pulse input

Suppose the following signals are generated as a result of a unit pulse input (only $y_1$ and $u_1$ may have been recorded in real time):

$$
\begin{align*}
Y_1(k) & = 1.0 \quad 1.2 \quad 0.96 \quad 0.768 \quad 0.6144 \\
Y_2(k) & = 1.0 \quad 2.2 \quad 3.16 \quad 3.928 \quad 4.5424 \\
u_1(k) & = 1 \quad 0 \quad 0 \quad 0 \\
u_2(k) & = 1 \quad 1 \quad 1 \quad 1 \quad 1
\end{align*}
$$

The Gram matrix of the signals $y_1$, $y_2$, $u_1$ and $u_2$ is

$$
F = \begin{bmatrix}
4.3289 & 12.4811 & 1.0 & 4.5424 \\
12.4811 & 51.8881 & 1.0 & 14.8304 \\
1.0 & 1.0 & 1.0 & 1.0 \\
4.5424 & 14.8304 & 1.0 & 5.0
\end{bmatrix}
$$

which yields the following square-roots of the diagonal cofactors.

$$
\sqrt{D_1} = 3.5032 \quad \sqrt{D_2} = 0.87581 \quad \sqrt{D_3} = 1.7516 \quad \sqrt{D_4} = -6.1307
$$

Note that the signs of these square-roots are chosen in direct correspondence with the signs of the cofactors of the first row of $F$ [1]. Now, substitution into (16) and division by $(D_1 + D_2)$ leads to the equation

$$(1 - 0.8 \ z^{-1}) Y(z) = (1 + 0.4 \ z^{-1}) U(z)$$

Clearly, the true parameters have been recovered.
Unknown network function = \( \frac{1 + 0.4z^{-1}}{1 - 0.8z^{-1}} \)

**Fig. 4. A first order test system**
Results of computer simulation on a fourth order network function are presented next.

**Example 2.**

The network function considered is

\[
H(s) = \frac{[s^2 + 0.31(10^6)s + 0.003(10)^{12}]}{[s^4 + 0.804(10^6)s^3 + 1.4481(10^{12})s^2 + 0.009686(10^{18})s + 0.007056(10^{24})]}
\]

\[
= \frac{[s + 10^4][s + 0.3(10^6)]}{[s^2 + 0.004(10^6)s + 0.0049(10^{12})][s^2 + 0.8(10)^6s + 1.44(10^{12})]}
\]

s-poles: \((-0.002 \pm j 0.0699714)(10^6)\)

\((-0.400 \pm j 1.131371)(10^6)\)

It was converted to a digital equivalent form (using the programs STOZ in Appendix A and pole-zero \(z = e^{s\Delta} \) transform [5], [8]) for computer simulation. With a sampling interval \(\Delta = 0.5 \mu\text{s}\) the z-domain transfer function turns out to be

\[
H(z) = \frac{2.00z^{-2} - 3.7114409z^{-3} + 1.7128304z^{-4}}{1 - 3.379158z^{-1} + 4.428628z^{-2} - 2.718099z^{-3} + 0.6689807z^{-4}}
\]

The system was excited by a \(\pm\) square 5 \(\mu\text{s}\) pulse (see Fig. 5a). The model identified by the proposed method is

\[
\hat{H}(z) = \frac{2.00z^{-2} - 3.71150z^{-3} + 1.7128z^{-4}}{1 - 3.37922z^{-1} + 4.42868z^{-2} - 2.71812z^{-3} + 0.66898z^{-4}}
\]

s-poles: \((-0.002 \pm j 0.0699714)(10^6)\)

\((-0.399 \pm j 1.131373)(10^6)\)

Using the inverse of the pole-zero transform, the s-domain transfer function can be obtained. The poles turn out as shown above.

The response of the model and the actual network response are compared in Fig. 5b.
Fig. 5  IDENTIFICATION OF A FOURTH ORDER SYSTEM  
(data uncorrupted by noise)
REMARKS

• When the network under test is of order \( n \), i.e., when the model order is equal to the intrinsic order of the network, the rank of the matrix \( F \) equals its dimensionality minus one.

• The matrix \( F \) is positive semi-definite.

• In actual application the matrix \( F \) will be formed from quantized versions of signals \( y \) and \( u \). Call this corrupted matrix as \( G \). It will be shown that \( E(G) = F + \sigma^2 P \), where \( P \) denotes the correlation matrix of unit noise and \( E \) denotes the statistical expectation operator. It will be shown in Section III that \( E(G) \) has full rank (equal to the dimensionality of \( F \)).

As seen earlier, the pencil of functions method uses the square-roots of the diagonal cofactors of \( F \). A very important advantage of the method is the following.

"Since \( F \) is positive semi-definite (\( G \) positive definite with unit probability), its diagonal cofactors are non-negative (strictly positive). Hence, there is a built-in check and stopping point when, due to computational errors or wrong choice of model order, one or more of these cofactors turns out to be negative."

The computations involve finding the cofactors of a \( 2n + 1 \) or \( 2n + 2 \) dimensional matrix. For the special case of impulse response modeling the calculation of denominator and numerator coefficients can be decoupled, so that computations involve only an \( n + 1 \) dimensional matrix. This will be discussed next.

2.2 DECOUPLED PROCEDURE FOR MODELING IMPULSE RESPONSES

Consider that \( y(k) \) is the impulse response of a network and that a suitable \( K \) has been selected such that \( y(k) = 0 \) for \( k > K \). We define the reverse-time integrated signals as follows [2], [11]

\[
\begin{align*}
y_1(k) & = y(k) \\
y_2(k) & = y_1(k) + \ldots + y_1(k), \\
& \quad \cdots \\
y_N(k) & = y_n(k) + \ldots + y_n(k),
\end{align*}
\]

(17)
(Recall, $N=n+1$). Let $F$ be defined as
\[
F = \begin{bmatrix}
\langle y_1, y_1 \rangle & \cdots & \langle y_1, y_N \rangle \\
\vdots & \ddots & \vdots \\
\langle y_N, y_1 \rangle & \cdots & \langle y_N, y_N \rangle
\end{bmatrix},
\langle y_i, y_j \rangle = \sum_{k=1}^{K} y_i(k)y_j(k) \tag{18}
\]

or, equivalently,
\[
F = \sum_{k=1}^{K} f(k)f_T(k) \tag{19}
\]

where $f_T(k) = [y_1(k) \ y_2(k) \ \ldots \ y_N(k)]$. Then, it can be shown that the denominator polynomial is given by
\[
A(z) = z^{-N} \left[ \sum_{i=1}^{N} \sqrt{D_i} (z-1)^{-1}/\sqrt{D_i} \right] \tag{20}
\]

where $D_i$ denotes the $i$th diagonal cofactor of the matrix $F$. Note the positive powers of $z$ on the right hand side. Further, the numerator coefficients are obtained as:
\[
\begin{bmatrix}
b_0 \\
\vdots \\
b_N
\end{bmatrix} = \begin{bmatrix}
p_{11} & \cdots & p_{1N} \\
\vdots & \ddots & \vdots \\
p_{N1} & \cdots & p_{NN}
\end{bmatrix}^{-1} \begin{bmatrix}
q_1 \\
\vdots \\
q_N
\end{bmatrix}, \tag{21a}
\]

\[
p_{ij} = \langle w(k+l-i), w(k+l-j) \rangle \tag{21b}
\]

\[
q_i = \langle y(k), w(k+l-i) \rangle \tag{21c}
\]

where $w(k)$ is the impulse response (i.e., inverse z-transform) of $1/A(z)$ and inner products are summed from $k=0$ to $K$. 

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If the network is known to have no direct transmission, i.e., \( b_0 \) is suspected to be zero, then \( N \) should be replaced by \( n \) on the right hand side, \( b_0 \) by \( b \) and in forming the inner products \( w(k+1-i) \) should be replaced by \( w(k-i) \) (likewise, \( w(k+1-j) \) should be replaced by \( w(k-j) \)).

Three examples will be presented next. The first is a simple, paper-pencil type example; it considers the same impulse response as did Example 1 (page 9) but with a long record length. The final example is interesting because it deals with an impulse response which, theoretically, requires an infinite order system (of type 1)) for exact reproduction; a fifth order model is computed by the pencil-of-functions method which yields a fractional energy error of 0.0359. In the first two examples the true transfer function is recovered by the modeling technique, i.e., the fractional energy error is zero.

Notation:

- \( \hat{y}(k) \) or \( \hat{y}(k) \): Model response
- \( \tilde{y}(k) \) or \( \tilde{y}(k) \): Model response error \( y(k)-\hat{y}(k) \)
- \( S = \sum_{k=0}^{K} y^2(k) \): Response energy
- \( \varepsilon = \sum_{k=0}^{K} y^2(k) \): Error energy
- \( \nu = \frac{\varepsilon}{S} \): Fractional energy error, or simply fractional error, or normalized mean square error
- \( \eta = 100(1-\nu) \): Per cent modeling efficiency

**Example 3**

Given the left hand side of \( y_1(k) = 1.5(0.8)^k - 0.5\delta_{k0} \), we find for \( K=40 \)
\[ F = \begin{bmatrix} 4 & 20 \\ 20 & 100 \end{bmatrix} \]  
(inner products are from \( k=1 \) to \( 40 \))

Then \( D_1 = 100 \) and \( D_2 = 4 \). Equation (20) yields

\[ A(z) = z^{-1}(10z-8)/10 = 1 - 0.8z^{-1} \]

Equation (21), in turn becomes

\[ \frac{1}{0.36} \begin{bmatrix} 1 & 0.8 \\ 0.8 & 1 \end{bmatrix} \begin{bmatrix} b_0 \\ b_1 \end{bmatrix} = \frac{1}{0.36} \begin{bmatrix} 1.5 - 0.18 \end{bmatrix} \]

which produces \( B(z) = (1 + 0.4z^{-1}) \). The model has been identified perfectly with zero fractional error.

**Example 4**

A fourth order network is known to have zero direct transmission \( (b_0=0) \). The numerical data of its impulse response,

\[ y(t) = 10 e^{-2t} \sin(2t) - 2 e^{-0.5t} \sin(4t) \]

is recorded at uniformly sampled intervals of \( \Delta = 0.2 \) sec. For \( K=150 \) (which signifies a long record; \( K = 30 \) sec), we find

\[ F = \begin{bmatrix} 4.51403 & 2.38856 & 0.779798 & 0.114119 & -0.05724740 \\ 2.09775 & 0.953032 & 0.253410 & 0.00074006 \\ 0.501610 & 0.157312 & 0.01499790 & 0.057232 \\ 0.057232 & 0.00960080 & 0.00352616 \end{bmatrix} \]

\[ \det F = 0.54E-14 \]

Note - All summations have employed a multiplication factor \( \Delta \), for scaling purposes, both in forming the integrated signals and in forming the inner-products. However, to undo the effect of this scaling, the ith diagonal cofactor has to be multiplied with \( \Delta^{2i} \) to yield \( D_i \). The entire process will be called \( \Delta \)-scaling.

The values of \( \frac{\sqrt{D_i}}{D_1} \) are

\[ 1 \quad 1.50410 \quad 1.33762 \quad 0.58517 \quad 0.11959 \]

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Fig. 6. Comparison of network and model impulse responses
Equations (20) and (21) yield the following denominator and numerator coefficients.

Denominator:  
1 \ -2.49588 \ 2.82521 \ -1.5760 \ 0.36786

Numerator:  
0 \ -1.31238 \ 1.68950 \ -1.55568 \ 0.00158

The fractional energy error turns out to be $\nu = 0.1E-6$. As seen from Fig. 6 the model response $\hat{y}$ is indistinguishable from the true response $y$.

**Example 5**

Here we consider a problem in approximation. This terminology, rather than identification, is appropriate since the square pulse

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

cannot be exactly reproduced as the impulse response of a finite order linear system. A fifth order model is desired whose impulse response approximates this signal. Using $\Delta = 0.05$ sec. and setting $y(0)=0$, $y(k)=1$ for $k=1,\ldots,20$, $y(k)=0$ for $k=21,\ldots,40$ the following Gram matrix is obtained.

$$F = \begin{bmatrix} 
1.0 & 0.525000 & 0.192500 & 0.0553438 & 0.01328250 & 0.002767190 \\
0.358750 & 0.146781 & 0.0448284 & 0.01117940 & 0.002391640 \\
0.0635731 & 0.0201175 & 0.00514064 & 0.001194800 & 0.0065189 \\
0.0065189 & 0.00169444 & 0.000373780 & 0.00044604 & 0.000022306 \\
0.00044604 & 0.000099363 & 0.000022306 & & \\
\end{bmatrix}$$

$$\det F = 0.344E-26 \quad (\text{note - } \Delta \text{ scaling is employed})$$

The values of $\sqrt{D_1/D_1}$ are

$$1 \quad 1.1458 \quad 0.67865 \quad 0.22900 \quad 0.042415 \quad 0.003383$$

Equations (20) and (21) yield the following z transfer function coefficients.

Denom.  
1 \ -3.854184 \ 6.095388 \ -4.93206 \ 2.037106 \ -0.0342867

Numerator  
0 \ -1.179760 \ 3.806755 \ -5.19377 \ 3.53997 \ -1.039771

The fractional energy error turns out to be $\nu = 0.0359$ with a corresponding modeling efficiency of 96.4%. The model response $\hat{y}$ is compared with
Fig. 7. Comparison of ideal and model impulse responses. Correction upon Gram matrix applied for (b).
the desired, ideal, response $y$ in Fig. 7a.

A brute force application of the correction procedure given in the next section (Section 3) results in the model response shown in Fig. 7b. Of course better approximations can only be obtained with higher order models.
3. QUANTIZATION ERROR: IMPROVEMENT OF ESTIMATES
(PENCIL-OF-FUNCTIONS METHOD)

Practical analog-to-digital (A/D) converters employ small word lengths, typically 8 to 16, and as a result incur quantization error in the representation of the signal. This, in turn, causes degradation in the accuracy of the identified transfer function [6]. It will be shown in this section that the statistical properties of the quantization error can be exploited to improve the accuracy of the parameter estimates. A computer program "GQUANT" incorporating the technique developed is given in Appendix B.

The principle of analog-to-digital conversion is explained well in references [6], [9]. For our purposes certain essential properties are most pertinent. If b bits are used (including the sign bit) and XMSB is the analog value of the most significant bit (next to the sign bit), then the following observations and properties hold.

(a) The step size equals

\[ \delta = \frac{XMSB}{2^{b-2}} \]  

(b) For an input \( y \) to the A/D convertor the analog value of the output is

\[ x = y + e \]  

where \(|e| < \frac{\delta}{2}\) for roundoff and \(|e| < \delta\) for truncation.

(c) If the signal excursions during each sampling time-interval \( \Delta \) are large compared to \( \delta \), then

\[ x(k) = y(k) + e(k) \quad y(k) \overset{\text{def}}{=} y(k\Delta) \]  

where \( e(k) \) is an independent sequence of random variables having a uniform distribution over one step size \( \delta \). In case of roundoff, this distribution is centered at zero, so that the random variable \( e(k) \) has a zero mean and a variance [7].

\[ \text{Var}(e(k)) = \frac{\delta^2}{12} \]  

In the ensuing discussion we will assume the A/D converter employs roundoff.

(d) Under the assumptions in (c) above, the error sequence \( e(k) \) is uncorrelated with the parent sequence \( y(k) \).
Simulation shows that neither of the properties (c) or (d) strictly hold in practice. However, we will use these properties exercising caution where necessary.

For definiteness we will discuss in detail the correction technique for impulse response modeling method of subsection 2.2. Parallel formulas are applicable to the simultaneous denominator and numerator modeling procedure of subsection 2.1, but will not be given here. Recall that the poles of the model are obtained from the Gram matrix of the signal y and its successive integrations. We therefore begin with the analysis and correction of the quantized Gram matrix.

3.1 GRAM MATRIX OF THE QUANTIZED SIGNAL

We will use the model of equation (24) for the quantized signal
\[ x(k) = y(k) + e(k) \]
where \[ E\{e(k)\} = 0, E\{e(k) e(\ell)\} = 0 \text{ and } E\{y(k) e(\ell)\} = 0 \text{ for all } k \text{ and } \ell. \] For the reversed time integrated signals we have
\[ x_i(k) = y_i(k) + e_i(k) \]
where \( e_i(k) \) are derived from \( e(k) \) through equations analogous to (17), i.e.,
\[ e_{i+1}(k) = e_i(k) + e_i(k+1) + \ldots + e_i(K) \]

Define also the vector sequences
\[ g(k) = \begin{bmatrix} x_1(k) \\ x_2(k) \\ \vdots \\ x_N(k) \end{bmatrix}, \quad p(k) = \begin{bmatrix} e_1(k) \\ e_2(k) \\ \vdots \\ e_N(k) \end{bmatrix}, \quad k=0,\ldots,K \] (28)

Then the Gram matrix of the quantized signal can be written as
\[ G = \sum_{k=0}^{K} g(k) g^T(k) \]
\[ = \sum_{k=0}^{K} \left[ f(k) f^T(k) + f(k) p^T(k) + p(k) f^T(k) + p(k) p^T(k) \right] \] (29)
Observation 1

$$E\{G\} = \sum_{k=0}^{K} F(k) F^T(k) + E\{\sum_{k=0}^{K} P(k) P^T(k)\}$$

$$= F + \sigma^2 P$$  \hspace{1cm} (30)

where \(P\) is the unit noise covariance matrix defined below. Further, if properties (c) and (d) strictly hold then

\[\sigma^2 = \frac{\delta^2}{12}\]

Observation 2

The unit noise covariance matrix is given by

$$P = E\{\sum_{k=0}^{K} P(k) P^T(k)\}$$ \hspace{1cm} (31)

where \(P(k) = [e_1(k) e_2(k) \ldots e_N(k)]^T\) as before, but \(e_1(k) = e(k)\) is taken to be a zero mean, unit variance, uncorrelated sequence.

Remark

If properties (c) and (d) do not strictly hold, then the value of \(\sigma^2\) (and possibly the definition of \(P\)) should be modified. We will estimate \(\sigma^2\) so as to satisfy the following criterion.

**Jain's Identification Criterion**

Consistent with the noise and bias models the estimated Gram matrix should achieve a minimum of the determinant.

Whatever method is used to choose the estimated Gram matrix, care should be taken to make sure that its determinant remains nonnegative, since the determinant of the true Gram matrix is nonnegative (see page 13). An approach to estimation of the Gram matrix is presented in subsection 3.4. First, however, we discuss the computation of the unit noise covariance matrix.

3.2 UNIT NOISE COVARIANCE MATRIX

Examination of the sequences

- \(e_1(k) = e(k)\)
- \(e_2(k) = e(k) + \ldots + e(K)\)
- \(e_3(k) = e(k) + \ldots + (K+1-k)e(K)\)
- \(e_4(k) = e(k) + \ldots + (K+1-k)^2e(K)\)
leads to the general formula

$$e_{i+1}(k) = \sum_{\ell=0}^{K-k} \ell^{i-1} e(k-\ell)$$

(33)

We then have (using the definitions in (28) and (31))

$$p(k) = \sum_{\ell=0}^{K-k} \ell^{i-1} e(k-\ell)$$

(34a)

$$r(\ell) = \sum_{\ell=0}^{K-k} \ell^{i-1} e(k-\ell)$$

(34b)

where \(r(\ell) = [\delta_{\ell,0}, 1, \ldots, \ell^{n-1}]\); \(\delta_{\ell,0}\) is the unit pulse sequence. Then

$$P = \sum_{k=0}^{K} \sum_{\ell=0}^{K-k} \sum_{m=0}^{K-k} r(\ell) r^T(m) e(k-\ell) e(k-m)$$

and, since \(e\) is a zero mean, unit variance, uncorrelated sequence,

$$P = \sum_{k=0}^{K} \sum_{\ell=0}^{K-k} r(\ell) r^T(\ell)$$

Note that \(P\) is determined entirely by the integers \(N\) and \(K\), the dimensionality of \(P\) (recall \(N=n+1\)) and the length of the observed sequences, respectively. Clearly, \(P\) can be precomputed and stored.

3.3 ESTIMATION OF QUANTIZATION ERROR VARIANCE

The discussion in subsection 3.1, specifically equation (30), leads us to estimate \(F\) as

$$\hat{F} = G - \sigma^2 P$$

(36)
where $\sigma^2$ will be chosen so as to minimize the determinant of $\hat{F}$.

One possible approach to this minimization is developed here. We use the fact that the rank of the true Gram matrix $F$ is $n$, i.e., its determinant is zero. Rewriting (36)

$$\hat{F} = G - \sigma^2 P$$

we set the determinant of both sides to zero. If the quantization error is small, we can approximate the determinant of the right hand side by the first two terms of the determinant expansion theorem. Thus

$$|\hat{F}| = 0 \approx G - \sigma^2 \sum \det(G,P)_i$$

where the notation $[G,P]_i$ means the matrix obtained by replacing the $i$th column of $G$ by the $i$th column of $P$.

Then

$$\hat{\sigma}^2 = \frac{|G|}{\sum \det(G,P)_i}$$

and, of course,

$$\hat{F} = G - \hat{\sigma}^2 P$$

Note that formula (39) can also be applied recursively, by replacing $G$ in (39) with the last estimate of $F$. An exit must be made when the determinant of the estimated matrix ceases to reduce further (or begins to increase).

3.4 SIMULATION EXAMPLES

As stated earlier, a FORTRAN IV computer program "GQUANT" has been developed for simulation and modeling of quantized impulse responses. A rational model of the type given in equation (1) is produced, except that $b_0$ is constrained to zero; i.e., the network is assumed to have no direct transmission. (Slight modification in the computation of numerator coefficients enables this constraint to be removed.) Equivalent s-domain description can be obtained through appropriate z to s transformation. Salient features of the program are the following.

It can be used in either a simulation mode (IRESP=1 or 2) or in network-response-data entry mode (IRESP=0).

Model can be obtained for unquantized signal (ISPN=-1, IFIX=-1, NFIX immaterial) when in the simulation mode, or actual response-data when in data entry mode.
Model can be obtained for the quantized signal (ISPN=1) without any statistical correction (IFIX=−1). Intended for use in simulation mode. Model can be obtained for the quantized signal (ISIM=1 or 2 and ISPN=1, or ISIM=0 and ISPN=0) with statistical correction (IFIX=1); the use of IBIAS=1 performs a bias correction in addition to statistical noise correction.

Two examples are given below, one in which a second order network response is simulated and another in which a fourth order response is simulated. Thus both examples pertain to simulated impulse responses.

- $H_{\text{ideal}}(z)$: True transfer function of the network.
- $H(z)$: Transfer function obtained by application of pencil-of-functions method upon unquantized signal. Note that $H(z)$ need not be equal to $H_{\text{ideal}}(z)$; among the reasons for this are computation errors, and the use of $K \neq \infty$.
- $H_{\text{quant}}(z)$: Transfer function obtained by from the quantized signal. (no correction is applied)
- $\hat{H}(z)$: Transfer function obtained from the quantized signal; one or more iterations of statistical correction for quantization errors are used.
- $\hat{H}(z)$: Transfer function obtained from the quantized signal; in addition to statistical correction for quantization errors, correction is also applied for possible bias in the data. It should be mentioned that the usefulness of bias correction arises both because the quantization errors in particular record of data may not be zero-mean, and also because $K \neq \infty$ may produce an apparent bias in data.
- NDIG: Length of binary word $b_{\text{NDIG}} \ldots b_2 b_1$ (note $b_{\text{NDIG}}$ is the sign bit, $b_{\text{NDIG}-1}$ the most significant bit, ..., and $b_1$ the least significant bit; also, we have employed a mid-tread type of quantizer in simulation)
- XMSB: The analog weight (or value) of the most significant bit.
Example 6

A second order network with zero direct transmission is simulated. Its impulse response

\[ y(t) = 2e^{-2t} - 2e^{-0.5t} \]

is sampled uniformly at intervals \( \Delta = 0.2 \) sec. apart. The coefficients of the transfer function \( H_{\text{ideal}}(z) \) are

Denominator

1  -1.575157  0.606530

Numerator

0  -0.469035  0

Without quantization the modeling program yields (using ISPN = -1, IFIX = -1, NFIX immaterial) the following results:

\( H(z) \) (using ISPN = -1, IFIX = -1, NFIX immaterial) \( \nu = 0.6E-8 \)

Denominator

1  -1.575180  0.606551

Numerator

0  0.469050  -0.000041

Experiment 1

For XMSB = 5.0 Volts and NDIG = 10, the program yields the following results:

\( H_{\text{quant}}(z) \) (using ISPN = 1, IFIX = -1, NFIX immaterial) \( \nu = 0.84E-3 \)

Denominator

1  -1.635921  0.662757

Numerator

0  0.513516  -0.114095

\( H(z) \) (using ISPN = 1, IFIX = 1, NFIX = 3; includes bias correction) \( \nu = 0.62E-3 \)

Denominator

1  -1.628129  0.655569

Numerator

0  0.506869  -0.098299

The impulse responses of \( H_{\text{quant}}(z) \) and \( H(z) \) are compared with that of \( H_{\text{ideal}}(z) \) in Fig. 8a and 8b. (The quantized signal used in determining these transfer functions is shown in Fig. 8c.) Although the improvement through statistical correction is hard to discern from these figures, the fractional energy error clearly points to a slight improvement.
Fig. 8a Comparison of network and model impulse responses.  
Model is obtained from quantized data.  
(No statistical correction is applied)
Fig. 8b Comparison of network and model impulse responses. Model is obtained from quantized data; statistical correction is applied.
Fig. 8c. Quantized signal used for the determination of model transfer function (see Fig. 8a and 8b for model response)
A more impressive improvement is achieved in the next experiment.

Experiment 2

For \( \text{XMSB} = 5.0 \) Volts and \( \text{NDIG} = 7 \), the program yields the following results:

\[ H_{\text{quant}}(z) \text{ (using ISPN} = 1, \text{ IFIX} = -1, \text{ NFIX immaterial)} \quad \nu = 0.0061 \]

Denominator

\[
\begin{array}{ccc}
1 & -1.721641 & 0.743261 \\
0 & 0.571452 & -0.263324 \\
0 & 0.550597 & -0.219990 \\
0 & 0.543307 & -0.204371 \\
\end{array}
\]

Numerator

\[
\begin{array}{c}
0 \end{array}
\]

\[ H(z) \text{ (using ISPN} = 1, \text{ IFIX} = 1, \text{ NFIX} = 1; \text{ IBIAS} = 0) \quad \nu = 0.0047 \]

Denominator

\[
\begin{array}{ccc}
1 & -1.703728 & 0.726890 \\
0 & 0.550597 & -0.219990 \\
0 & 0.543307 & -0.204371 \\
\end{array}
\]

Numerator

\[
\begin{array}{c}
0 \end{array}
\]

\[ H(z) \text{ (using ISPN} = 1, \text{ IFIX} = 1, \text{ NFIX} = 1; \text{ IBIAS} = 1) \quad \nu = 0.0042 \]

Denominator

\[
\begin{array}{ccc}
1 & -1.696989 & 0.720718 \\
0 & 0.543307 & -0.204371 \\
\end{array}
\]

Numerator

\[
\begin{array}{c}
0 \end{array}
\]

Example 7

A fourth order network with zero direct transmission is simulated. The impulse response

\[ y(t) = 10e^{-2t} \sin(2t) - 2e^{-0.5t} \sin(4t) \]

is sampled uniformly at intervals \( \Delta = 0.2 \) sec. apart. The coefficients of the transfer function \( H_{\text{ideal}}(z) \) are

Denominator

\[
\begin{array}{cccc}
1 & -2.495629 & 2.824925 & -1.577498 & 0.367879 \\
0 & 1.312168 & -1.688152 & 1.553863 & 0 \\
\end{array}
\]

Numerator

\[
\begin{array}{c}
0 \end{array}
\]

Without quantization the modeling program yields (using ISPN = -1, IFIX = -1, NFIX immaterial) the following results:

\[ H(z) \text{ (using ISPN} = -1, \text{ IFIX} = -1, \text{ NFIX immaterial)} \quad \nu = 0.1 \text{E-6} \]
Denominator
1 -2.495883 2.825209 -1.577598 0.367863
Numerator
0 -1.312376 1.689499 -1.555676 0.001578
For XMSB = 5.0 Volts and NDIG = 10, the program yields the following
results
\( H_{\text{quant}}(z) \) (using ISPN = 1, IFIX = -1, NFIX immaterial) \( \nu = 0.076 \)
Denominator
1 -3.003323 3.619004 -2.057970 0.455114
Numerator
0 -1.290046 2.315347 -1.991812 0.834767
\( H(z) \) (using ISPN = 1, IFIX = 1, NFIX = 1; IBIAS = 0) \( \nu = 0.045 \)
Denominator
1 -2.930222 3.483747 -1.979050 0.440495
Numerator
0 -1.368196 2.599335 -2.567285 1.185892
\( H(z) \) (using ISPN = 1, IFIX = 1, NFIX = 1; IBIAS = 1) \( \nu = 0.040 \)
Denominator
1 -2.910943 3.447190 -1.956858 0.436115
Numerator
0 -1.393888 2.690009 -2.731991 1.280967
 Clearly, a reduction in energy error has been achieved via statistical
correction.

Remarks
The application of the statistical correction was predicated upon several
assumptions. Experiments show that these assumptions are not satisfactorily
met. The following comments therefore arise.
  • The quantization error process \( e(k) \) is not white. It might be useful
    in future work to model this error process as a first order process and estimate
    the corner frequency of this process together with its intensity.
  • The correlation between the quantization error \( e(k) \) and the input
    signal \( y(k) \) is not zero. This may be ameliorated by the use of a well known
    technique [13] namely the addition and, after quantization, the subtraction
    of a dither signal\(^2\). This is shown in Fig. 9. The application of this

\(^2\)Pseudo-random binary signals are often used as dither signals.
technique to our problem, and the extent of improvement achieved [14], remain subjects of future investigation.

\[ y(t) \rightarrow y'(t) \rightarrow \text{A/D converter} \rightarrow x'(k) \rightarrow \text{Computer} \]

\[ x(k) = x'(k) - r(k) = y(k) + e(k) \]

Fig. 9 Use of dither signal to decorrelate \( y(k) \) and \( e(k) \)

- In estimating the intensity (variance) of noise via equation (36) only the first two terms of the determinant expansion were retained. Perhaps three terms, i.e., constant linear and quadratic, should be retained in order to get a more accurate estimate of \( \sigma^2 \). However, we feel that the benefit of this step would be realized only after the steps 1 and 2 stated above have been taken.
4. WIDEBAND IDENTIFICATION

Determining the transfer function of a network from its observed input-output responses represents the inverse of the analysis problem. Interest in this problem arises from the frequent need for a relatively simple mathematical description of the system so that behavior for other anticipated inputs may be predicted up to acceptable accuracies. However, the identification of wideband networks presents some unique difficulties. Consider, for example, a network whose frequencies of interest range from $f_0$ Hz to $(10^5)f_0$ Hz. To identify the corner frequencies at the low end, one would require an observation record of length $T = 1/f_0$ sec. On the other hand, in order to avoid aliasing effects the sampling rate must be chosen in excess of $2(10^5)f_0$, say $f_S = (10^6)f_0$. A million samples of data for both input and output are thus produced. Apart from the difficulties of storing this staggering amount of data and the impracticality of processing them, serious numerical difficulties also arise from this simple minded approach to identification; for instance, the low frequency poles cannot be represented in z-domain accurately even with a 64-bit computer word. A possible remedy is to break the problem into two or three smallband problems. The network dynamics can be identified for each of these, and this information can be used to estimate the wideband transfer function.

A multirate sampling approach to identification of wideband transmittances is presented in this section. It permits efficient

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3A frequency band of less than two decades will be termed as small-band.
determination of the transfer function of a four-to-five decade bandwidth system from transient tests. Clever selection of sampling rates and exciting inputs reduces the wideband problem into three, simpler smallband problems. Each smallband problem encompasses only one-to-two decades of bandwidth. The three transfer functions $H_L(s)$, $H_M(s)$, and $H_H(s)$ are easily identified via the pencil-of-functions method, and then adjoined to build the wideband transfer function estimate $H(s)$. The technique is demonstrated by simple illustrative examples and a realistic RF amplifier example.

**Frequency regions** (sub-bands)

The concept of small band descriptions begins by splitting the wideband region into three regions. As shown in Fig. 10, these regions may be termed as low-frequency band, medium frequency band, and the high-frequency band; in short, LF, MF, and HF. These regions may be chosen covering approximately equal ranges on the logarithmic scale. Denote the band edges as $f_L$, $f_M$, and $f_H$, and the respective mid-region frequencies as $f_{L\prime}$, $f_{M\prime}$, and $f_{H\prime}$. The latter may be -- although not necessarily, chosen as the geometric means of the band edge frequencies.

By design, the following inequalities hold

\[ 0.1f_L < f_0 < f_L \cdot f_M < 10f_L \] \hspace{1cm} (40a)
\[ 0.1f_M < f_1 < f_M \cdot f_H < 10f_M \] \hspace{1cm} (40b)
\[ 0.1f_H < f_2 < f_H \cdot f_M < 10f_H \] \hspace{1cm} (40c)

In some cases where prior knowledge of the approximate frequency characteristic of the network is available, it may be more appropriate to choose the regions as LF, MF and HF, or as LF, MIF and HF.
Fig. 10 Definition of LF, MF and HF frequency bands
Sampling Rates

For the three small band problems the sampling intervals are chosen as

\[ \Delta_L = \frac{1}{100f_L} \quad (41a) \]
\[ \Delta_M = \frac{1}{100f_M} \quad (41b) \]
\[ \Delta_H = \frac{1}{100f_H} \quad (41c) \]

The sampling rates are of course the reciprocals of these numbers.

By this choice -- and in view of (40), the sampling rates become at least ten times the highest frequency of interest in the respective bands. If the system input are selected so as to excite frequencies only in one of the bands, then by using the prescribed sampling rate aliasing effects would be avoided.

Now, if \( K \), the number of samples used for the identification procedure, is taken as 1000, the length of the record would be 2 times the longest time constant of the band under consideration (for example, with this choice of \( K \) for the LF case, \( T_L = 1000 \Delta_L = \frac{10}{f_L} = \frac{1}{2\pi f_L} \)). Such record lengths are considered adequate for practical identification of low edge corner frequencies, and storing and processing 1000 samples of data is well within today's minicomputer capability.

Inputs

The key to the conversion of the wideband problem to three small band problems is the careful selection of inputs which excite...
frequencies essentially limited to one of the bands. At first this would seem to pose no real difficulty, for we can choose a narrowband signal for the test. However, a little thought would reveal that testing with very narrowband signals would be in direct conflict with the basic philosophy of system identification, which is broadband modeling with transient tests. Therefore a judicious compromise must be made between these conflicting requirements.

The following inputs are suggested as a rough guide. Experimentation and experience leads to a much richer variety of signals which meet the above compromise strategy. Two different considerations have been kept in mind in the selection of these inputs: the spectral requirement stated above and, equally important, easy realizability in the laboratory.

a) LF Input

For the low frequency band the input selected is a triangular pulse, either a full cycle $\text{TR}_{+,-}(t)$ or a half cycle $\text{TR}_+(t)$ (see Fig. 11). In either case, the total duration of the pulse is taken to be $T_L/2$ and the pulse is followed by zero input for the remainder of the time, i.e., from $T_L/2$ to $T_L$. The magnitude spectra of these inputs can be shown to be

$$|\text{TR}_{+,-}(f)| = \frac{T_L}{4\Delta_L} \left| \frac{\sin \pi f T_L}{\pi f T_L} \right|^2 |\sin 2\pi f T_L/8|$$  \hspace{2cm} (42)

$$|\text{TR}_+(f)| = \frac{T_L}{4\Delta_L} \left| \frac{\sin \pi f T_L/4}{\pi f T_L/4} \right|^2$$  \hspace{2cm} (43)

where, keeping (15a) in mind, $T_L = K/100f_L$.

![Input waveforms for LF tests.](image_url)

Figure 11. Input waveforms for LF tests.

---

6 For networks which pass d.c., $\text{TR}_{+,-}(t)$, i.e., a full cycle triangular pulse, is recommended; this reduces the predominance of a d.c component in the network response.
Unit peak values for the pulses have been assumed. The amplitude spectra are tabulated in Tables 1 and 2.

### TABLE 1
Magnitude Spectra of +, - triangular pulse ($\Lambda_{-}$)

| $f/f_L$ | ITR$_{+,-(f)}$ | $|TR_{+,-(f)}|^2$ |
|---------|----------------|------------------|
|         | $K=20$ | $K=200$ | $K=1000$ | $K=2000$ |
| 0.01    | -56.1 dB | -36.1 dB | -22.1 dB | -16.1 dB |
| 0.1     | -36.1   | -16.1   | -3.5     | -1.8     |
| 0.5     | -22.1   | -3.5    | -16.1    | -29.8    |
| 1.0     | -16.1   | -1.8    | -29.8    | -35.8    |
| 2.0     | -10.3   | -∞      | -∞      | -∞      |
| 10.0    | -1.8    | -∞      | -∞      | -∞      |
| max in band | -1.8 | -1.8 | -1.8 | -1.8 |

Note: zero dB corresponds to a magnitude of $2b/\Delta_L=K/4$

### TABLE 2
Magnitude spectra of a + triangular pulse ($\Lambda_{+}$)

| $f/f_L$ | $|TR_{+,+(f)}|^2$ |
|---------|------------------|
|         | $K=20$ | $K=200$ | $K=1000$ | $K=2000$ |
| 0.01    | -0.0 dB | -0.0 dB | -0.0 dB | -0.1 dB |
| 0.1     | -0.0    | -0.1    | -1.8    | -7.8    |
| 0.5     | -0.0    | -1.8    | -29.8   | -35.8   |
| 1.0     | -0.1    | -7.8    | -35.8   | -∞      |
| 2.0     | -0.3    | -∞      | -∞      | -∞      |
| 10.0    | -7.8    | -∞      | -∞      | -∞      |
| max in band | 0.0 | -0.1 | -1.8 | -7.8 |

Note: zero dB corresponds to a magnitude of $2b/\Delta_L=K/4$

---

7 Minus infinity is used whenever the spectral amplitude is less than -200 dB below reference level.
It is clear from Tables 1 and 2 that the spectra of these pulse inputs diminishes to -30 dB or more (below in-band maxima) at the LF-MF boundary, provided \( N \) is chosen greater than or equal to 200. This insures that the frequencies in the MF region are not excited by application of these inputs. A possible exception is the case where there is a sharp resonant peak in the MF band, particularly at the LF-MF boundary. However, the presence of such a peak is generally known beforehand; such a resonant component in the output can be filtered before performing identification on the LF test data.

**MF Input** -

For the medium frequency band the input selected is an oscillatory pulse, modulated either by a decaying exponential OEX\( (t) \) or by a diminishing one-quarter-cycle triangular wave OTR\( (t) \). In either case, the frequency of oscillation is taken to be \( f_M \), the center frequency of the band. The duration of the oscillation is taken to be \( T_M/2 \) (see Fig. 12), followed by zero input for the remainder of the time, i.e., from \( T_M/2 \) to \( T_M \). In presenting the spectral analysis below it is assumed that the on-set of the pulse begins with the maxima of the oscillation, i.e., the pulse is triggered at its maximum value. Thus \( u(t) = m(t) \cos 2\pi f_M t \) where \( m(t) \) denotes the modulating envelope. The spectra of these inputs can be shown to be as follows:

\[
|OEX(f)| = \frac{1}{2} \left| M(f + f_M) + M(f - f_M) \right|,
\]

\[
M(f) = \frac{1}{(a+j\omega)} \left[ 1 - e^{-\frac{(a+j\omega)T_M}{2}} \right]
\]

\[
|OTR(f)| = \frac{1}{2} \left| M(f + f_M) + M(f - f_M) \right|,
\]

\[
M(f) = \frac{1}{j\omega} \left[ 1 - \frac{\sin T_M f/2}{\pi T_M f/2} \left( \cos T_M f/2 - j\sin T_M f/2 \right) \right]
\]

If the sampling interval were chosen five times the value suggested in (41a) the magnitude spectrum diminishes to -30 dB at the LF-MF boundary even for \( N=20 \).
where ω=2πf, 'a' is the inverse time-constant associated with the exponential decay and, keeping (41b) in mind, $T_M = k/100f_M$. Unit peak values have been assumed.

![Input waveforms for MF test.](image)

In order to delineate the spectral characteristics of the input $OEX(t)$, three different values of 'a' will be considered: $a=0$, $a=2/T_M$, and $a=4/T_M$. The corresponding waveshapes will be denoted as $OEX_0(t)$, $OEX_1(t)$ and $OEX_2(t)$, respectively. The amplitude spectra of $OEX_0(t)$, $OEX_1(t)$, $OEX_2(t)$ and $OTR(t)$ are tabulated in Tables 3 to 6 respectively.

**TABLE 3**

<table>
<thead>
<tr>
<th>$f/f_M$</th>
<th>$K = 20$</th>
<th>$K = 200$</th>
<th>$K = 1000$</th>
<th>$K = 2000$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.010</td>
<td>0.1 dB</td>
<td>-74.0 dB</td>
<td>-74.0 dB</td>
<td>-74.1 dB</td>
</tr>
<tr>
<td>0.100</td>
<td>0.1</td>
<td>-34.0</td>
<td>-37.8</td>
<td>-∞</td>
</tr>
<tr>
<td>0.500</td>
<td>0.1</td>
<td>-7.4</td>
<td>-21.4</td>
<td>-∞</td>
</tr>
<tr>
<td>0.909</td>
<td>0.0</td>
<td>-0.5</td>
<td>-3.6</td>
<td>-20.5</td>
</tr>
<tr>
<td>0.990</td>
<td>0.0</td>
<td>-0.0</td>
<td>-0.1</td>
<td>-0.2</td>
</tr>
<tr>
<td>1.000</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
</tr>
<tr>
<td>1.010</td>
<td>-0.0</td>
<td>0.0</td>
<td>-0.0</td>
<td>-0.1</td>
</tr>
<tr>
<td>1.100</td>
<td>-0.0</td>
<td>0.3</td>
<td>-3.5</td>
<td>-∞</td>
</tr>
<tr>
<td>2.000</td>
<td>-0.4</td>
<td>-∞</td>
<td>-∞</td>
<td>-∞</td>
</tr>
<tr>
<td>10.000</td>
<td>-29.1</td>
<td>-∞</td>
<td>-∞</td>
<td>-∞</td>
</tr>
</tbody>
</table>

Note: zero dB corresponds to the resonant peak at $f_M$. Minus infinity is used whenever the spectral amplitude is less than -200dB below reference level.
**TABLE 4**

Magnitude spectra of an Oscillatory exp. pulse

<table>
<thead>
<tr>
<th>$f/f_M$</th>
<th>$K = 20$</th>
<th>$K = 200$</th>
<th>$K = 1000$</th>
<th>$K = 2000$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.010</td>
<td>-1.7 dB</td>
<td>-26.2 dB</td>
<td>-53.1 dB</td>
<td>-63.1 dB</td>
</tr>
<tr>
<td>0.100</td>
<td>-1.6</td>
<td>-23.4</td>
<td>-36.7</td>
<td>-49.7</td>
</tr>
<tr>
<td>0.500</td>
<td>-1.1</td>
<td>-6.9</td>
<td>-20.7</td>
<td>-33.5</td>
</tr>
<tr>
<td>0.909</td>
<td>-0.2</td>
<td>-0.5</td>
<td>-3.4</td>
<td>-14.6</td>
</tr>
<tr>
<td>0.990</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
</tr>
<tr>
<td>1.000</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
</tr>
<tr>
<td>1.010</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
</tr>
<tr>
<td>1.100</td>
<td>0.2</td>
<td>0.3</td>
<td>-3.2</td>
<td>-15.7</td>
</tr>
<tr>
<td>2.000</td>
<td>1.2</td>
<td>-13.6</td>
<td>-27.4</td>
<td>-33.5</td>
</tr>
<tr>
<td>10.000</td>
<td>-12.9</td>
<td>-29.9</td>
<td>-43.8</td>
<td>-49.9</td>
</tr>
</tbody>
</table>

Note: zero dB corresponds to the resonant peak at $f_M$

**TABLE 5**

Magnitude spectra of an Oscillatory exp. pulse

<table>
<thead>
<tr>
<th>$f/f_M$</th>
<th>$K = 20$</th>
<th>$K = 200$</th>
<th>$K = 1000$</th>
<th>$K = 2000$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.010</td>
<td>-0.0 dB</td>
<td>-15.0 dB</td>
<td>-41.7 dB</td>
<td>-53.2 dB</td>
</tr>
<tr>
<td>0.100</td>
<td>-0.0</td>
<td>-14.2</td>
<td>-34.0</td>
<td>-43.4</td>
</tr>
<tr>
<td>0.500</td>
<td>-0.0</td>
<td>-5.6</td>
<td>-19.1</td>
<td>-27.4</td>
</tr>
<tr>
<td>0.909</td>
<td>-0.0</td>
<td>-0.4</td>
<td>-2.9</td>
<td>-9.8</td>
</tr>
<tr>
<td>0.990</td>
<td>0.0</td>
<td>0.0</td>
<td>-0.1</td>
<td>0.2</td>
</tr>
<tr>
<td>1.000</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
</tr>
<tr>
<td>1.010</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
</tr>
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</tr>
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<td>-0.1</td>
<td>-8.1</td>
<td>-21.4</td>
<td>-27.4</td>
</tr>
<tr>
<td>10.000</td>
<td>-9.3</td>
<td>-24.1</td>
<td>-37.8</td>
<td>-43.8</td>
</tr>
</tbody>
</table>

Note: zero dB corresponds to the resonant peak at $f_M$.
## TABLE 6

Magnitude Spectra of an Oscillatory triangular pulse

<table>
<thead>
<tr>
<th>( f / f_M )</th>
<th>( K = 20 )</th>
<th>( K = 200 )</th>
<th>( K = 1000 )</th>
<th>( K = 2000 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.010</td>
<td>0.1 dB</td>
<td>-38.0 dB</td>
<td>-51.9 dB</td>
<td>-58.1 dB</td>
</tr>
<tr>
<td>0.100</td>
<td>-18.0</td>
<td>-36.3</td>
<td>-43.8</td>
<td></td>
</tr>
<tr>
<td>0.500</td>
<td>-4.3</td>
<td>-21.2</td>
<td>-27.4</td>
<td></td>
</tr>
<tr>
<td>0.909</td>
<td>0.0</td>
<td>-0.3</td>
<td>-2.3</td>
<td>-8.7</td>
</tr>
<tr>
<td>0.990</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
<td>-0.1</td>
</tr>
<tr>
<td>1.000</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
</tr>
<tr>
<td>1.010</td>
<td>-0.0</td>
<td>0.0</td>
<td>0.0</td>
<td>-0.1</td>
</tr>
<tr>
<td>1.100</td>
<td>0.0</td>
<td>0.2</td>
<td>-2.1</td>
<td>-9.5</td>
</tr>
<tr>
<td>2.000</td>
<td>-0.3</td>
<td>-7.6</td>
<td>-21.4</td>
<td>-27.4</td>
</tr>
<tr>
<td>10.000</td>
<td>-9.3</td>
<td>-23.9</td>
<td>-37.8</td>
<td>-43.8</td>
</tr>
</tbody>
</table>

Note: zero dB corresponds to the resonant peak at \( f_M \)

It is clear from Tables 3 and 4 that the spectra of pulses \( OEX_0(t) \) and \( OEX_1(t) \) diminish to \(-25\) dB or more at the MF-LF and MF-HF boundaries provided \( N \) is chosen greater than or equal to 200\(^9\). This insures that the frequencies in the LF and HF regions are not excited significantly by application of these inputs. Tables 3 and 6 show that \( OEX_2 \) and OTR spectra diminish only to \(-15\) dB at these boundaries; these pulses are useful in the initial stages of testing, or when the network's corner frequencies are spread over the band.

### HF Input -

The inputs used for the MF band are equally useful for the HF band with \( T_M \) replaced by \( T_H \). Tables 3 to 6 also hold with \( f_M \) replaced by \( f_H \).

\(^9\)If the sampling interval is chosen five times the value suggested in (41b), the magnitude spectrum diminishes to \(-30\) dB at the boundaries even for \( N=20\).
Small band Identification

Input-output data obtained from smallband tests can be analyzed by use of the Fortran program IGRAM [5] and the s-domain smallband transfer functions obtained therefrom. The program, however, requires that the transfer function order (degree of the denominator polynomial) be specified. If the order of a smallband transfer function is known from circuit considerations, then the identification is performed for this order and for at least one order higher and lower. For example, if the LF band behavior is expected to be of order 4, then identification should be performed for n=3, 4 and 5. The lowest order model yielding satisfactory fractional error (see page 13) should be accepted as the model for that smallband. If, on the other hand, the smallband order is not known, then an upward modeling strategy must be adopted. Starting from an initial order, a low guess, increasingly higher orders are attempted until the fractional error in identification turns out to be acceptably small.

Thus, the smallband transfer functions \( H_L(s) \), \( H_M(s) \) and \( H_H(s) \) become available. From these the overlapping critical frequencies, or ideally speaking common critical frequencies, are carefully isolated. This isolation of common critical frequencies is useful in the next, and final, step in wideband identification.

Adjoined Wideband Transfer Function

The transfer functions obtained from the smallband tests must be adjoined to form the overall wideband transfer function. For convenience we will drop the hat (carat) on the identified TFs, smallband or wideband. The reader must, however, bear this in mind.

In the notation to follow we will use \( C \) to denote gain constant; and \( H \) with suitable subscripts to denote transfer functions, which are assumed to be in the Bode canonical form

\[
C \frac{s^k (1 + s/z_1) \ldots (1 + s/z_m)}{(1 + s/p_1) \ldots (1 + s/p_n)} \quad (k = 0, \text{positive or negative integer})
\]

(46)

The first subscript (on \( H \)) refers to the test from which the transfer
function is obtained; the second subscript, if any, denotes the band with which the critical frequencies (poles and/or zeroes) are shared. For example, $H_{MH}(s)$ denotes the part of $H_N(s)$ whose critical frequencies are shared with $H(s)$. Ideally speaking, of course, $H_{MH}(s) = H_{HM}(s)$.

Thus we have

$$H_L(s) = C_L H_{LL}(s) H_{LM}(s) \quad (47a)$$

$$H_M(s) = C_M H_{ML}(s) H_{MM}(s) H_{MH}(s) \quad (47b)$$

$$H_H(s) = C_H H_{HM}(s) H_{HH}(s) \quad (47c)$$

Critical Frequency Adjustment

In practice the overlapping critical frequencies in two small-band tests will not turn out to be identical through the corresponding identifications. For example, a critical frequency $s'_i$ (pole, or zero) common to $H_N(s)$ and $H_H(s)$, may be identified as $s_{i-}$ in the MF identification and as $s_{i+}$ in the HF identification. We will adjust both of them to a common value given by their geometric mean

$$s_i = \sqrt{s_{i-} s_{i+}} \quad (48)$$

Assume that this process has been performed on $H_{LM}$ and $H_{ML}$, and likewise on $H_{MH}$ and $H_{HM}$. In order to avoid unduly complicated notation we will let the original symbols denote these adjusted transfer functions, so that now

$$H_{LM}(s) = H_{ML}(s) \quad (49a)$$

$$H_{MH}(s) = H_{HM}(s) \quad (49b)$$

Other adjustments include setting $s_i$ to 0 when it turns out to be well below $2\pi f_-$, where $f_-$ denotes the left boundary of the frequency band, but is known to be zero from circuit considerations; care should be taken in this case to let $s_i(1 + s/s_i)$ to $s$. Thus, in the canonical form of (46) the gain should be divided by $s_i$ when the term $(1 + s/s_i)$ is replaced by $s$. Another case of adjustment occurs when $s_i$ turns out to be much larger than $2\pi f_+$, where $f_+$ denotes the right boundary of the frequency band; it
is then useful to set the term \((1 + s/s_i)\) to just 1. These two types of adjustments occur in Example 8 on page 47.

Gain Adjustment

To obtain equalization at the boundaries of the frequency bands the gains are adjusted as follows.

\[
C_L = C_L' \left[ \frac{H_L(\lambda_1)H_H(\lambda_2)}{H_L(\lambda_1)H_H(\lambda_2)} \right]^{1/2}
\]

(50a)

\[
C_M = C_M' \left[ \frac{H_L(\lambda_1)H_H(\lambda_2)}{H_M(\lambda_1)H_H(\lambda_2)} \right]^{1/2}
\]

(50b)

\[
C_H = C_H' \left[ \frac{H_L(\lambda_1)H_H(\lambda_2)}{H_M(\lambda_1)H_H(\lambda_2)} \right]^{1/2}
\]

(50c)

where

\[
\lambda_1 = j2\pi f_1
\]

\[
\lambda_2 = j2\pi f_2
\]

Recall that \(f_1\) is the boundary between the LF and MF regions, and \(f_2\) is the boundary between the MF and HF regions (see Fig. 10, page 36).

As stated earlier, the purpose of this gain adjustment is to minimize gain discontinuity at the boundaries. However, phase mismatch may still exist at these boundaries for the redefined smallband transfer functions. These transfer functions are

\[
H_L(s) = C_L H_{LL}(s)H_{LM}(s)
\]

(51a)

\[
H_M(s) = C_M H_{ML}(s)H_{MN}(s)H_{HM}(s)
\]

(51b)

\[
H_H(s) = C_H H_{HM}(s)H_{HH}(s)
\]

(51c)

Wideband Transfer Function -

The wideband transfer function is taken to be

\[
H(s) = C H_{LL}(s)H_{LM}(s)H_{MM}(s)H_{HM}(s)H_{HH}(s)
\]

(52)

10 For certain wideband networks only two smallbands, LF and HF (with boundary frequencies \(f_1\) and \(f_2\)), might be necessary. In such cases only (50a) and (50c) are needed with \(H_M\) deleted and \(\lambda_1=\lambda_2=j2\pi f_1\); likewise, only (51a) and (51c) are needed wherein the subscript \(M\) is replaced by the subscript \(H\).
where the constant C is selected to match the gain of one of the smallband transfer functions at a chosen frequency, perhaps $H_M(s)$ at the midband frequency $f_M$.

**Example 8**

Consider an a.c. coupled network believed to have frequencies of interest from 0.02 MHz to 50 MHz, thus encompassing 3.3 decades. The parameters of the system are given in Table 7. In this case it is adequate to break up the frequency region into two smallbands as follows:

- **LF** $f_o = 0.02$ MHz to $f_1 = 1.0$ MHz $f_L = 0.1$ MHz (53a)
- **HF** $f_1 = 1.0$ MHz to $f_2 = 50.0$ MHz $f_H = 10.0$ MHz (53b)

The inequalities in (40) are clearly satisfied. From (41) and Tables 1 and 3 it appears reasonable to choose

$$
\begin{align*}
\Delta_L &= 0.1 \mu s & K_L &= 200 & T_L &= 20 \mu s \\
\Delta_H &= 0.001 \mu s & K_H &= 200 & T_H &= 0.2 \mu s
\end{align*}
$$

and the inputs $u_L(t) = TR_{+,o}(t)$, $u_H(t) = OEX_{o}(t)$. We, however, select

$$
\begin{align*}
\Delta_L &= 0.05 \mu s & K_L &= 200 & T_L &= 10 \mu s \\
\Delta_H &= 0.002 \mu s & K_H &= 100 & T_H &= 0.2 \mu s
\end{align*}
$$

and define the test pulses explicitly as follows

$$
\begin{align*}
u_L(t) &= \begin{cases} 
\text{One complete cycle triangular wave over 0 to 5 sec.} \\
\text{0 level over 5 to 10 sec.}
\end{cases} \\
u_H(t) &= \cos\left(\frac{2\pi}{0.05} t\right), \quad 0 \leq t \leq 0.1 \\
&= 0, \quad 0.1 < t \leq 0.2
\end{align*}
$$

To simulate the LF test response, the system function $H_{true}(z)$ corresponding to $\Delta_L' = 0.005 \mu sec.$ is excited by $u_L(k)$ of (55a); the response is then resampled at 1/10th rate (i.e., every 10th output sample is picked up). In a laboratory test this artifice of using a high sampling rate $H(z)$ to preserve the integrity of the network response, and then resampling the output, is of course not necessary. The network output can be sampled directly at the desired rate $1/\Delta_L$.

\text{\textsuperscript{11} Recall, } \Delta \text{ denotes sampling interval, } K \text{ the total number of samples and } T \text{ the total duration of the test. Of course } T = KA.\text{ \textsuperscript{47}}
The results of identification from IGRAM using the method of Subsection 2.1 are given below.

**LF Test**

Both first and second order identifications were performed. Since the first order model (predicted response) has a fit error of $\nu = 0.8E-4$, it is decided the LF behavior is first order. For $n=1$ the Gram matrix, the square-roots of the diagonal cofactors, the z-domain model and the corresponding s-domain model (using pulse-invariant conversion [5], pages 80-82) are given below:

$$
F = \begin{bmatrix}
0.00074285 & 0.00037153 & -0.075250 & 5.1001 \\
1.03620000 & -5.19880000 & -354.5100 & 1266.7000 \\
33.336000 & 346090.00 & 1.0297(10^{-8})s & -1.8167327z^{-1} + 0.81709493z^{-2} \\

det F = 0.128 \end{bmatrix}
$$

The values of $\sqrt{D_1/D_1}$ are

\[ L(z) = 0.0101(1-z^{-1}) \\
 1-0.95215z^{-1} \]

\[ L(s) = \frac{0.0101(s-0.14(10^6))}{(s+0.981(10^6))} = \frac{1.0297(10^{-8})s}{(s/0.981(10^6)+1)} \]  

(Frequency adjusted)
HF test –

As in the LF case, here also a first order model is found adequate producing a fit error (fractional energy error) $\nu = 0.40E-4$. For $n=1$ the Gram matrix, the square-roots of the diagonal cofactors, the z-domain model and the corresponding z-domain model (using pulse-invariant conversion) are given below.

$$F = \begin{bmatrix}
0.00090304 & 0.00045161 & -0.073612 & -0.52301 \\
0.01842400 & 0.449400 & 1.61760 & 1.25000 \\
25.000000 & 12.50000 & 410.38000 \\
\end{bmatrix}$$

$$\det F = 0.22E-4$$

The values of $\sqrt{D_1/D_1}$ are

$$1 \quad 0.22817 \quad -0.0023153 \quad 0.0022408$$

$$\hat{H}_H(z) = \frac{-0.00006 + 0.00188517z^{-1}}{1 - 0.814217z^{-1}}$$

$$\hat{H}_H(s) = \frac{0.0023s + 1.009(10^6)}{s + 102.76(10^6)} = \frac{0.00982}{s/1.028(10^8) + 1}$$

(Frequency adjusted)

Gain Adjustment –

At $s = j2\pi f_1$, where $f_1=1$MHz, the gains of the LF and HF transfer functions turn out to be 0.00998 and 0.00981 respectively. The adjusted gain constants (using (50) and (51)) are $C_L=1.021(10^{-8})$ and $C_H=0.00990$.

Wideband Transfer Function –

The wideband transfer function is

$$H(s) = C \frac{s}{(s/0.981(10^6) + 1)(s/1.028(10^8) + 1)}$$

where $C=1.023(10^{-8})$ is obtained from gain matching at $s=j2\pi f_1; f_1=1$MHz.

Comparison –

The Bode plots of $H(s)$ and $H_{true}(s)$ (of Table 7) are compared in Fig. 13. It appears that satisfactory wideband identification has been achieved.

Remark

The procedure of adjoining the smallband transfer functions can of course be programmed.
Fig. 13 Comparison of the magnitude (Bode) plots of the identified transfer function and the network function of a wideband system.
Example 9

As a second example of wideband identification consider the RF amplifier (Fig. 14) of reference [5]. The frequency regions are

\[
\begin{align*}
\text{LF} & : f_0 = 0.002 \text{ MHz to } f_1 = 0.1 \text{ MHz} \\
\text{MF} & : f_1 = 0.1 \text{ MHz to } f_2 = 10 \text{ MHz} \\
\text{HF} & : f_2 = 10 \text{ MHz to } f_3 = 1000 \text{ MHz}
\end{align*}
\]

The smallband transfer functions, identified from LF MF HF tests through IGRAM [5] are

\[
\begin{align*}
H_L(s) &= -20.125 \frac{(s-0.0015(10^6))(s+0.0012(10^6))}{(s+0.034(10^6))(s+0.075(10^6))} -0.7892(10^{-8}) \frac{s^2}{[s/0.034(10^6)+1][s/0.075(10^6)+1]} \\
H_M(s) &= -20(10^6) \frac{1}{(s+24.92(10^6))} -20.55 \frac{1}{(s/25.31(10^6)+1)} \\
H_H(s) &= 2.79(10^7) \frac{(s-19060(10^6))}{(s+25.7(10^6))(s+1140(10^6))} -18.432 \frac{(-s/19606(10^6)+1)}{[s/25.31(10^6)+1][s/1140(10^6)+1]}
\end{align*}
\]

The second step of each of the above is obtained after frequency adjustment as outlined on page 45.

Gain Adjustment -

At \( f_1 = 0.1 \text{ MHz} \) the gains of the LF and MF transfer functions turn out to be 19.9536 and 20.5437 respectively. At \( f_2 = 10 \text{ MHz} \) the gains of the MF and HF transfer functions are computed to be 7.6776 and 6.8759 respectively. The adjusted gain constants (using (50) and (51)) are \( C_L = (0.9602)C_L = 0.7578, C_M = (0.9327)C_M = 19.17, \) and \( C_H = (1.0414)C_H = 19.19. \)

Wideband Transfer Function -

The transfer function of the network is estimated as

\[
\hat{H}(s) = -C \frac{s^2}{[s/0.034(10^6)+1][s/0.075(10^6)+1][s/25.31(10^6)+1][s/1140(10^6)+1]}
\]

\[
\begin{align*}
\hat{H}(s) &= -18.432 \frac{(-s/19606(10^6)+1)}{[s/25.31(10^6)+1][s/1140(10^6)+1]} \frac{s}{[s/0.034(10^6)+1][s/0.075(10^6)+1][s/25.31(10^6)+1][s/1140(10^6)+1]} \frac{1}{51}
\end{align*}
\]
Fig. 14 A common-emitter wideband amplifier
where $C=8.0596 \times 10^{-9}$ is obtained from gain matching with $\hat{H}_M(s)$ at $s=j2\pi f_M$, $f_M=1$ MHz.

Comparison -

The Bode plots of $\hat{H}(s)$ and $H_{\text{true}}(s)$ (of Fig. 14) are compared in Fig. 15. It appears that satisfactory wideband identification has been achieved from smallband time domain tests.
Fig. 15 Comparison of the magnitude (Bode) plots of the identified transfer function and the network function of an RF amplifier.
REFERENCES


APPENDIX A

LISTING OF

PROGRAM

STOZ
STOZ

GIVEN THE CONTINUOUS DESCRIPTION, PROGRAM COMPUTED THE
EQUIVALENT DISCRETE DOMIAN DESCRIPTION OF A LINEAR
DYNAMIC SYSTEM.

STOZ GENERATES H(z) AND THE CORRESPONDING DIFFERENCE
EQUATION FROM THE TRANSFER FUNCTION H(s).

THE INPUT ARRAYS A AND B ARE FILLED ACCORDING TO THE
DIFFERENTIAL EQUATION

\[ a(1)u(t) + a(2)u(t-1) + \ldots + a(n+1)u(t-n) = b(1)y(t) + b(2)y(t-1) + \ldots + b(n+1)y(t-n) \]

WHERE \( u(t) \) = THE MTH TIME DERATIVE OF FUNCTION, \( u \)
\( a(n+1) \) MUST BE EQUAL 1.

RETURNS ARRAYS A AND B CONTAINING THE EQUIVALENT DISCRETE
DESCRIPTION STORED ACCORDING TO THE DIFFERENCE EQUATION

\[ b(n+1)u(t) + b(n+2)u(t-1) + \ldots + b(2)u(t-n) = a(1)y(t) + a(2)y(t-1) + \ldots + a(n+1)y(t-n) \]

B(1) ALWAYS EQUALS 1.

THE POLES OF THE CONTINUOUS DOMAIN MUST BE DISTINCT AND
NON-ZERO FOR THE TRANSFORMATION TO BE VALID.

---------------------------

DATA CARD SET PREPARATION

N = ORDER OF SYSTEM
A (MAXIMUM) = ONE LESS THAN THE DIMENSION SUBSCRIPT

I(TH) = 0 FOR THE IMPULSE IN Variant DESCRIPTION,
= 1 FOR THE PULSE INVARIANT DESCRIPTION,
= 2 FOR THE TRAFFIC INVARIANT DESCRIPTION,
= 3 FOR THE LOGISTIC TRANSFORM DESCRIPTION.

IPOLZ = 1 IF POLES AND ZEROS ARE REAL
MUST BE COMPLEX (REAL, IMAGINARY) WITH ONE REAL ASSOCIATED
NEGATIVES OF POLES AND ZEROS READ AT 1.
FOR A POLZ OF (5+2), INPUT +2... +5... +1... +2.
(2FIF 4 PER POLZ, 4 PER CARD)

= 0 IF DENOMINATOR AND NUMERATOR ARE REAL
IN POLYNOMIAL FORM.
COEFFICIENTS ORDERED FROM LOW TO HIGH
DENOMINATOR INFORMATION always REAL FIRST.
HIGHEST ORDER DENOMINATOR COEFF MUST BE 1.

NOTE: WHEN POLE-ZERO DATA IS ENTERED A GAIN CARD
MUST FOLLOW THE LAST POLE CARD.
IF THERE ARE NO ZEROS, USE BLANK CARDS IN
THE NORMAL ZEROS POSITIONS.

cELTA = SAMPLING INTERVAL
START EACH DATA CARD SET WITH A DESCRIPTION CARD CONTAINING UP TO 51 CHARACTERS COLS 2-52.
FIRST DATA CARD CONTAINS:
N, IMTHD, IPOLZ, IN DFS FORMAT, PLUS DELTA IN DFS FORMAT.
SECOND GROUP OF DATA CARDS IS A POLES OR N+1 DENOMINATOR COEFFICIENTS, AFTER LAST POLE CARD, USE A GATH CARD.
LAST GROUP OF DATA CARDS IS N ZEROS (OR BLANKS).
OR N+1 NUMERATOR COEFFICIENTS (BLANKS FOR ZERO COEFFICIENTS).
THE DATA FORMAT FOR EACH OF THE SYSTEM PARAMETER CARDS IS 8F15.5.
AS MANY SETS OF DATA CARDS MAY BE RUN AS DESIRED.

STOZ MAIN PROGRAM

REAL B(2:1), A(20), RR(2C), RI(2C), DELTA, TEMP(2C)
COMPLEX C(2C), CA(2C), CB(2C), CAI(2C), CBI(2C)
1 TEM(2D), CONI, CON2, CONT
DIMENSION TITLE(70)

READ TITLE AND FIRST DATA CARD
130 READ(5,92) TITLE
IF(EOF(5).NE.0) GO TO 5995
WRITE(6,910)
91C FORMAT(6(/))
WRITE(6,920) TITLE
92C FORMAT(7(51))
WRITE(6,912)
916 FORMAT(6/1X,7(1('')))
READ(5,942) N, IMTHD, IPOLZ, DELTA
960 FORMAT(3(5,1F10.6))
WRITE(6,930) IMTHD, IPOLZ, DELTA
95C FORMAT(//3X,4N1,145X,1M10 C**I4.5X,**IPOLZ = **I4.5X
1 *DELTA = **G17.1,1.11)
NP1 = N+1
NP2 = N+2
NP3 = N+3
NP4 = N+4
IF(IPOLZ.EQ.1) GO TO 303

READ POLES AND ZEROS
110 READ(5,963)(CA(I), I=1,N)
960 FORMAT(NF10.6)
CALL POLCON(0R, TEM, 0) ,N)
DO 109 I=1,NP1
109 R(I) = TEM(I)

READ GAIN CARD
READ(5,961) RK

READ ZEROS
READ(5,965)(CA(I), I=1,N)
CALL POLCON(CAI, TEM, 0) ,N)
DO 209 I=1,NP1
209 A(I) = TEM(I)*RK
GO TO 310

READ DENOMINATOR AND NUMERATOR COEFFICIENTS
C 30C READ(S,96C) (B(I),I=1,NP1)
   READ(S,96C) (A(I),I=1,NP1)
   CONTINUE
C PRINT DENOMINATOR AND NUMERATOR COEFFICIENTS
C WRITE(S,97C)
97C FORMAT(* S-DOMAIN DENOMINATOR*)
   CALL PRVEC(S,NP1)
   WRITE(S,97C)
98C FORMAT(* S-DOMAIN NUMERATOR*)
   CALL PRVEC(A,NP1)
C DETERMINE ORDER OF NUMERATOR
C NN=N
   DO 309 I=1,NP1
      IF (*.NE.0) GO TO 40C
      NN=NN-I
   CONTINUE
WRITE(S,990) NN
309 CONTINUE
C FACTOR DENOMINATOR TO FIND POLES
C IF(IPOLZ.NE.0) GO TO 50C
   CALL PCLRT(A,TEMP,N,RR,R1,IER)
   DO 409 I=1,N
      CR(I)=CMPLX(PR(I),RI(I))
   CONTINUE
WRITE(S,991)
991 FORMAT(* POLES OF S-DOMAIN*)
   CALL PRVEC(CR,N)
C IF(IPOLZ.NE.1) GO TO 150C
C LOGARITHMIC TRANSFORM
C CONTINUE
WRITE(S,990)
990 FORMAT(* LOGARITHMIC TRANSFORM*)
C WORK ON NUMERATOR
C IF(NN.EQ.0) GO TO 130C
   CALL PCLRT(A,TEMP,N,RR,R1,IER)
   DO 1009 I=1,NN
      CA(I)=CMPLX(RF(I),RI(I))
   CONTINUE
WRITE(S,990)
100C CONTINUE
990 FORMAT(* ZEROS IN S-DOMAIN*)
   CALL PRVEC(CR,N)
   DO 1029 I=1,NN
      CA(I)=CMPLX(CR(I),DELTA)
   CONTINUE
1029 IF(NN.EQ.N) GO TO 110C
   DO 1030 I=NP1,NP1
      CA(I)=0.D0
   CONTINUE
1030 CONTINUE
C NOW THE FIRST NN ENTRIES OF CA CONTAIN THE
Z-Domain zeros of the transfer function, while the remaining entries are zero-cut.

Work on denominator

DO 1129 I=1,N
1129 CR(I)=EXP(CR(I)*DELTA)

Now CR contains the N Z-domain poles

Form numerator and denominator

Z-domain polynomials

IF(INN.EQ.1) CAA(I)=1.0
IF(INN.NE.1) CALL PCSTZ(CA,CAA,C,NN)
CALL PCSTZ(CR,CR,0,N)

Now CB contains the N+1 Z-domain denominator coefficients, and CAA contains the N+1 numerator coefficients.

Adjust DC gain constant

A1=A(1)/B(1)
A2=1.0
DO 1209 I=1,NN
1209 A2=A2+CAA(I)*1.0
1219 A2=A2+CR(I)*1.0
FAC=A1*A2/A2
DO 1229 I=1,NPI
1229 CAA(I)=CAA(I)*FAC

Now CAA contains the adjusted Z-domain numerator coefficients and FAC contains the gain factor used for the adjustment.

Go to 500:

1500 CONTINUE

Non-logarithmic transformations.

Adjust for direct transmission

This routine requires that B(NP1) = 1.0

CONT=(0.0,0.0)
IF(INN.LT.1) GO TO 1510
CONT=A(NP1)
DO1569 I=1,N
1569 A(I)=A(I)-CONT*B(I)

Find numerator constants for partial fraction expansion

1510 DO1529 I=1,N
1519 CONT=0.0
DO1519 J=1,N
1519 CONT=CONT+CR(J)*A(N-J+1)
1512 CONT=CONT*(CR(J)-CR(J-1))
1519 CONTINUE
1529 C(I)=CONT/CRA
WRITE(6,930)
930 FORMAT(* Numerator constants of the factorized H(s) = *)

CALL PRCCVFCCCN)
CONVERT THE FIRST ORDER PARTIAL FRACTIONS TO Z EOMAT.

IMTHO=IMTHO+1
GO TO 120(8,300+400(I), IMTHO)

IMPULSE INVARIANT

DO 2009 I=1,N
CA(I)=CA(I)*DELTA
CR(I)=CEXP(CR(I)*DELTA)
GO TO 450

PULSE INVARIANT

DO 3009 I=1,N
CON=CEXP(CR(I)*DELTA)
CA(I)=CA(I)*CON/CR(I)
CR(I)=CON
GO TO 450

TRAPEZOIDAL INVARIANT

ICHECK=2
DO 4009 I=1,N
CON=CEXP(CR(I)*DELTA)
CON2=CA(I)/CR(I)*CR(I)*DELTA*CON
CONT=CONT+CON*((1.3-CR(I))*DELTA*CON-1.0)
CA(I)=CON*((1.3-CONI)*CR(I)-CON)
CR(I)=CON
GO TO 450

CONSTRUCT THE Z DOMAIN DENOMINATOR AND NUMERATOR POLYNOMIALS

CONTINUE
CALL PCST7(CP,CB9,ON)
DO 4509 I=1,N
4509 CA(I)=J-I+1

CALL PCST7(CR,CA,K,N)
DO 4519 J=1,N
4519 CA(J)=CA(J-J)+CA(J)*CA(K)
CA(NPl)=0.0

******
IF(IMTHO.NE.1)GO TO 4521
DC 4523 I=1,N
II=NPl=1
4522 CA(I+1)=CA(I)
CA(1)=0.0
CONTINUE

ADJUST FOR DIRECT TRANSMISSION

DTXC=(0.0,0.0)
IF(NN,NE,1) CONT=DTXC
CA(NPl+1)=CONT*CR(I)*NPl)
DO 4529 I=1,N
4529 CA(I)=CA(I)+CONT*CB(I)

SHIFT NUMERATOR TO COMPLETE PULSE INVARIAT TRANSFORM
WHEN NUMERATOR HAS LOWER ORDER THAN DENOMINATOR

61
CONTINUE

PRINT THE TRANSFORMED COEFFICIENTS

WRITE (6, 9510)
CALL PROVECGR(N)
WRITE (6, 9520) FAC
WRITE (6, 9530)
CALL PROVECGR(NN1)
WRITE (6, 9540)
CALL PROVECGR(NP1)
WRITE (6, 9550)
CALL PROVECGR(NP11)
WRITE (6, 9560)
WRITE (6, 9570)
STOP

SUBROUTINE PROSTZ(C, R, N)
C
C      PROSTZ CONSTRUCTS A Z-DOMAIN POLYNOMIAL COEFFICIENT ARRAY
C      FROM AN ARRAY OF ITS ROOTS.
C
DIMENSION C(N), R(N)
COMPLEX C, R
NPI = N + 1
DO I = 2, NPI
R(I) = 0.0
DO J = 1, N
IF (I - K16, 3, 6
DO W = 1, N
IF (W - K16, 3, 6
CONTINUE
RETURN
END

SUBROUTINE PROVEC(A, N)
C
C      PROVEC PRINTS A COMPLEX VECTOR
C
DIMENSION A(N)
COMPLEX A
IF (N .EQ. 0) GO TO 100
WRITE (6, 920)
WRITE (6, 910)(A(I), I = 1, N)
FORMAT (3, 1X, 1E15.14, 1E15.14, 1X, 1E15.14, 1X, 1E15.14, 1X, 1E15.14)
100 WRITE (6, 920)
920 FORMAT (2I1, 1X)
SUBROUTINE PRVEC(A,N)
C
THIS SUBROUTINE OUTPUTS A SINGLE DIMENSIONED ARRAY
DIMENSION A(I)
C
WRITE(6,1)(A(I),I=1,N)
I
RETURN
END
SUBROUTINE POLCON(C,Q2,K,K)
C
A POLYNOMIAL CONSTRUCTION PROGRAM NEEDED FOR Z10S
C
DIMENSION C(I),Q2(I)
COMPLEX C,R2
REAL DQ
EQUIVALENCE (COMP,DQ)
NP1=NP+1
I01=2*NP+1
I02=I1+1
COMP=0(I)
IF(I.EQ.K.OR.Q(I).EQ.0.0.AND.Q(2).EQ.0.0)GO TO 4
DQ2=I1+1
J=I-J1+1
2
Q2(J)=R*(J+1)*C(I)*C(J)
R2(1)=R(I)*C(I)
CONTINUE
RETURN
END
SUBROUTINE POLRT(COF,GOF,ROOTI,ROOT,IER)
C
COMPUTES THE REAL AND COMPLEX ROOTS OF A REAL POLYNOMIAL
C
DESCRIPTION OF PARAMETERS
C
COF -VECTOR OF M+1 COEFFICIENTS OF THE POLYNOMIAL
C
ORDERED FROM SMALLEST TO LARGEST POWER
C
GOF -WORKING VECTOR OF LENGTH M+1
C
M -ORDER OF POLYNOMIAL
C
ROOTI-RESULTANT VECTOR OF LENGTH M CONTAINING REAL ROOTS
C
OF THE POLYNOMIAL
C
ROOT-RESULTANT VECTOR OF LENGTH M CONTAINING THE
C
CORRESPONDING IMAGINARY ROOTS OF THE POLYNOMIAL
C
IER -ERROR CODE WHERE
C
IER=0 NO ERROR
C
IER=1 M LESS THAN ONE
C
IER=2 M GREATER THAN 36
C
IER=3 UNABLE TO DETERMINE ROOT WITH 50C INTERATION?
C
ON 5 STARTING VALUES
C
IER=4 HIGH ORDER COEFFICIENT IS ZERO
C
DIMENSION XCOF(1),COF(1),ROOT(1),ROOTI(1)
C
LIMITED TO 36TH DEGREE POLYNOMIAL OR LESS.
FLOATING POINT OVERFLOW MAY OCCUR FOR HIGH ORDER
POLYNOMIALS BUT WILL NOT AFFECT THE ACCURACY OF THE RESULT
C
METHOD

63
NEwTON-RAPHSON ITERATIVE TECHNIQUE. THE FINAL ITERATIONS ON EACH ROOT ARE PERFORMED USING THE ORIGINAL POLYNOMIAL RATHER THAN THE REDUCED POLYNOMIAL TO AVOID ACCUMULATION OF ERRORS IN THE REDUCED POLYNOMIAL.

```
C IFIT=0
  N=M
  IER=0
  IF(XCOF(N+11)10,25,15
10 IF(N) 15,15,32
C SET ERROR CODE TO 1
C 15 IEP=1
20 IF(N=23) 20,20G
20 IF(N=1.2)33IIE
203 FORMAT(1X,*ERROR CALLED FROM POLYT, IEP= *,I3)
201 RETURN
C SET ERROR CODE TO 4
C 25 IER=4
GO TO 20
C SET ERROR CODE TO 2
C 30 IEP=2
GO TO 20
32 IF(N-3)=5,5,5
35 NX=N
NX=N+1
N2=1
KJ1 = N+1
DO 40 L=1,KJ1
MT=KJ1-L+1
43 COF(MT)=X**OF(L)
C SET INITIAL VALUES
C 45 XO=0.05311111
YO=0.15611111
C ZERO INITIAL VALUE COUNTER
C IN=0
50 X=X
C INCREMENT INITIAL VALUES AND COUNTER
C XO=10.3*YO
YO=10.8*X
C SET X AND Y TO CURRENT VALUE
C X=X
Y=Y
IN=IN+1
GO TO 50
55 IFIT=1
XPR=X
YPR=Y
C EVALUATE POLYNOMIAL AND DERIVATIVES
```
140 COF(2)=COF(2)+ALPHA*COF(1)
145 DO 150 L=2,N
150 COF(L+1)=COF(L+1)+ALPHA*COF(L)-SUMSQ*COF(L-1)
155 ROOTI(N2)=Y
       ROOTR(N2)=X
       N2=N2+1
IF(SUMSQ) 160,165,166
160 Y=-Y
       SUMSQ=0.0
       GO TO 155
165 IF(N) 23,20,45
       END
**PROGRAM "GQUANT"**
**IMPULSE-RESPONSE MODELING**
**BY PENCIL-CF-FUNCTIONS METHOD**
**DEC 1979 (FOR RAC E)**

**PROGRAM "GQUANT" USES CHARACTERISTICS OF QUANTIZATION ERROR IN PENCIL-CF-FUNCTION METHOD TO PRODUCE IMPROVED TRANSFER FUNCTION. THIS MODEL IMPULSE-RESPONSE OF CHANNEL NETWORK CAN BE USED IN SIMULATION MODE OR ON EXPERIMENTALLY RECORDED RESPONSES.**

********************************************************************************
DIMENSION E(EO),U(EO),U(ESC),X(5C),G(8),A(8),AP(A,8)
DIMENSION ON(8),G(8),GCUM(8),U(8),C(8),C(8)
DIMENSION V(16),P(16),SHIFT(8),SPH(8),SPH(8)
DIMENSION TITLE(7),IBUF(512)
DOUBLE PRECISION DT,AC,3D,ERROR
COMMON /IO/ISP,F,DELTA,SIG2,CT,DI,BIAS,IBIAS
COMMON /IO/ISIM,BIAS,AMP,FESUM,FESUM
COMMON /IO/IO,ILT,IPR,ICUMD,IPLT
RECORD
MAXPL=512
MAX=8
MAX2=2*MAX
IR=5
ILT=6
ISKIP=0
MSTART=2
CALL VEQUAT(MAXPL,U,F,0,10)
CALL VEQUAT(MAXPL,U,F,1,10)
CALL VEQUAT(MAX2,V,VV,1,10)
WRITE(ILT,2)
READ(IR,9)(TITLE(I),I=1,7C)
WRITE(ILT,10)(TITLE(I),I=1,7C)
READ(IR,4)(TITLE(I),I=1,7C)
READ(IR,8)(TITLE(I),I=1,7C)
READ(IR,NPNT,IPRP,NGIG,N,ISIM,NCOMP,IPLT,NPNT)
X=ISP,CT,BIAS
NP1=N+1
NP2=N+1
NP3=N+3
NP4P2=N+42
NPNP1=N+1
IF(NPNT.EQ.0)NPNT=NPT
IF(C(CT,EO).LT.0)DT=1.0
IF(ISIM.EQ.0)READ(IR,14C)(F(K),K,1,NPNT)
IF(ISIM.EQ.0)GO TO 61
IF(ISIM.EQ.1)READ(IR,14C)(F(V(I),I,1,NPNP2)
IF(ISIM.EQ.1)CALL RESPON(F,U,NV,VV,NPNT)
IF(ISIM.EQ.0)GO TO 61
DO 60 I=1,NCOMP
READ(IR,5)AMP(I),SR(I),SI(I),SPM(I)
60 WRITE(ILT,11)(AMP(I),SR(I),SI(I),SPM(I)
CALL SIGNAL(F,NPNT,AMP,SR,SI,SPH,CT,NCOMP)
CONTINUE
IF(IPLT.GE.2)CALL PLOTS(IBUF,512,9)
1111 READ(IN,9)(TITLE(I),I=1,7C)
IF(EOF(IP),N,1)GO TO 998

67
WRITE(ILT,3)
WRITE(ILT,18) (TITLE(I),I=1,76)
READ(IR,4) IR,IREP,ISPN,IFIX,IBIAS,IYY,IZZ,CI
IF(ROUND=0)
C ROUND OFF OPTION
C 410 CONTINUE
IF(IROUND.NE.0) CALL QUANT2(F,NDIG,IFLAG,NPT,MAXFL)
CONTINUE
IF(IROUND.NE.0) GO TO 99
DO 30 K=1,NPT
30 X(K)=X(K)+BIAS
CONTINUE
IF(NP1.GT.1) CALL INGRAT(F,NPT,NP1,MAXFL,-1)
C COMPUTE GAP MATIX
C NP=NPT
IF(IBIAS.NE.0) NP=NP2
DO 44 I=1,NPT
44 J=1,NPT
AD=0.2
IF(ISPN.EQ.0 .AND. IRCUND.EQ.1) GO TO 43
DO 42 K=NPT,-1
42 GN(I,J)=AD
43 CONTINUE
IF(IROUND.EQ.1) GN(I,J)=GN(I,J)
CONTINUE
IF(ISPN.EQ.0 .OR. IRCUND.EQ.1)
CALL GROCT1(G,E,DET,V,OFF,NPT,NP2,NOT,1)
IF(IROUND.EQ.1) WRITE(ILT,171) DET
IF(IRUND.EQ.1) WRITE(ILT,172) DET
IF(IROUND.EQ.1) WRITE (ILT,171) DET
IF(IFIX.EQ.0 .AND. ISPN.NE.1) GO TO 41
C WRITE(ILT,1)
IF(IROUND.EQ.1) IRCUND=IRCUND+1
IF(IFIX.EQ.0 .AND. ISPN.NE.1) GO TO 41
C ESTIMATE CF ** G
C 150 CALL RUL02D(M,V,NP1,NPNT,MAX,IFIX)
--- NP1 REPLACED BY KPP NEXT 3 CAPS--
CALL FIXGK(DUM,AP,GEST,E,V,NPP,MAX,IFIX)
IF(IFIX.EQ.0) WRITE(ILT,482) SIG2
CALL GROCT1(G,E,DET,V,NPP,NPP,NP2)
WRITE(ILT,162) DET
IF(IFIX.EQ.0) CALL PRMAT(GEST,NP1,NP1,NP1,MAX,0)
DO 154 I=1,NP1
154 J=1,NP1
DUM(I,J)=GEST(I,J)
IF(IFIX.EQ.0) GO TO 156
ISKIP=1
156 CONTINUE
C CALCULATE ERR CR MATIX
C 150 IF(IFIX.LE.2 .OR. ISKIF.EQ.0) GO TO 151
DO 32 I=1,NP1
32 J=1,NP1
G

WRITE(ILT,161)
CALL PQT4AT(E,NFl9NPt,..q4Xv-l)
WRITE(ILT,163)
CALL PRTMAT (EN,NNP1,NP1,MAX,-1)
C CONTINUE
C DETERMINE NUMERATOR
C
CONTINUE
IF(ISPN.EQ.0) GO TO 998
CALL VEQUAT(NP1,VP(NP2),VV,0,10)
CALL RESPON(X(1,11),U..,X,NNP1)
CALL INSTR4(NP1,NNP1-IREM,MAXPL,2)
C CHANGES MADE HERE FOR E(i)=C
L=NN-IREM
IF(IBIAS.NE.0) L=NN+1
LPl=L+1
LP2=L+2
IF(IBIAS.NE.0) CALL VEQUAT(NP1,X(1,LPl),U,F,G)
CALL VEQUAT(NP1,X(1,LP2),PIAS,G,3)
DO 210 I=1, L
DO 216 J=1,LP1
G(I,J)=0.
DO 215 K=NNP1
215 G(I,J)=G(I,J)*DET
CALL VEQUAT(NP1,NPTX(LP),X(NP1,LP2),F,G,1)
CALL VEQUAT(NP1,NPTX(LP),FC,1)
CALL VEQUAT(NP1,NPTX(LP),Pias,1)
DO 216 J=1,L
DO 219 K=NN-1
219 VV(I)=VV(I)+E(I,J)*G(J,LPl)
DET = DET * G(I,J)
211 WRITE(ILT,327)
WRITE(ILT,21C) (VV(I),I=1,NP1)
WRITE(ILT,21C) (VV(I),I=NP1+1,NNP1)
CALL RESPON(X(1,2),U,N,N,NNP1)
ERROR=8.C
FFSUM=8.C
FIF(K,E).GO TO 990
IF(K.EQ.1) WRITE(ILT,140)(K(I),I=1,NP1)
ERROR=ERROR+K(I)*X(K,I)
X(K,1)=ENER
213 ERROR=ERROR+K(I,3)*X(K,3)
FFSUM=FFSUM+DT
ERROR=ERROR*RT
RATIO=ERROR/FFSUM
WRITE(ILT,384) ERROR,RATIO,FFSUM,RATIO
IF(IPR.GE.2) WRITE(ILT,11C)(F(K),K=1,NNP1)
WRITE(ILT,11C)
IF(IPR.GE.2) WRITE(ILT,11C) (X(K,2),K=1,NNP1)
WRITE(ILT,11C)
IF(IPR.GE.2) WRITE(ILT,11C) (X(K,3),K=1,NNP1)
C TO=6.0
IF(IPLT.GE.2) CALL PLOP(NNPT,2,X.MAXP,L,TO,1MY,HINT,IMUF)
FORMAT STATEMENTS

5 FORMAT(5F10.3)
6 FORMAT(14.6I2,14.4F10.6)
8 FORMAT(7DA1)
10 FORMAT(2X,7CA1)
92 FORMAT(12(5X,F5.1))
94 FORMAT(12(5X,F5.15))
11 FORMAT(2Y,12,*,AMP=*,F8.2,*,S=*,F10.4,*,J*,F11.4)
146 FORMAT(1C,6B4M MATRIX)
150 FORMAT(15X,6B4M MATRIX)
161 FORMAT(10,12HW TRUE = GEST)
163 FORMAT(12X,14HM TRUE = GCISY)
162 FORMAT(12X,14HM TRUE = MATRIX,*) (DET=*,G13.6,*!*)
210 FORMAT(7X,5D8.6,G13.6))
110 FORMAT(254X,F5.2))
410 FORMAT(2X,10I2X,F15.5))
178 FORMAT(10X,14HNOI5Y X MATRIX)
179 FORMAT(12X,16H TRUE GRAM MATRIX,*) (DET=*,G13.6,*!*)
172 FORMAT(10X,17HNOI5Y GRAM MATRIX,*) (DET=*,G13.6,*!*)
180 FORMAT(5F10.6)
323 FORMAT(2X, *EST NUM NENOP VECTOR*)
334 FORMAT(2X, X*ERROR=*,G13.6,**)FSUM=*,G13.6,**)LC=*,G13.6,**)1
395 FORMAT(2X,ESTIMATED MEAN=*,G13.5)
432 FORMAT(* ESTIMATED NOISE VAR=*,G17.5)
1 FORMAT(/)
2 FORMAT(10)
3 FORMAT(/)
4 ISIM=0 FOR MODELING ACTUAL RESPONSE DATA
5 1 OR 2 FOR SIMULATION (SIM(1)) (2SUMS OF EXP AND GSC)
11 NOSIM=1 FOR SIMULATING SH-BIT, OR SIMULATION MATHEMATICALLY NOUNCEFF IN MAT
21 IPAD=2 FOR BINARY, 11 FOR DECIMAL
21 NP1=INTEGRATED FUNCTIONS, THE FIRST IS DATA
1 IPR=0 FOR MINIMAL PRINTING, OTHERWISE 1 OR 2
2 ISP=IF ANALYSIS OF WKF, ERRORS SIGNAL ONLY; 1 FOR FITTED SIGN
3 -1 IF INTEGRATION OF TRUE (APPROXIMATE) SIGNAL ONLY
4 NCOMP= COMPONENTS (#EXP(EP T) * SINISI T) TYPE
5 INT=1 (OR 2) FOR FORWARD INTEGRATION, -1 FOR REVERSE
6 DT=Sampling INTERVAL XWF=WEIGHT OF MSA (4n-1)
7 NPT=DATA POINTS, N=DEGREE OF MCF
8 IFIX=1 FOR NO CORRECTION, 1
9 IFIX=0 IF GEST=GH-AK, 1 IF NOISE VAR TO BE ESTIMATED
10 -1 IF NO CORRECTION IS TO BE APPLIED (ISIM PLST BE 1)

GO TO 111

998 CONTINUE

CALL PLOTCV...

STOP
END

SUBROUTINE SIGNAL(F,NPT,AMP,SR,SI,SPH,DT,NCP)

DIMENSION F(L),AMP(1),SR(1),SI(1),SPH(1)
COMMON /ADIF/K,ILI,IPK,ICUNO

DOUBLE PRECISION A,B,C,V

00 12 K=1,NPT
12 XK=1+1

DO 20 I=1,NCP

A=SR(I)*DT

DO 20 I=1,NCP

A=SI(I)*DT

70
C=SP4(I)
DO 15 KK=1,NPT
K=KK-1
X=AMP(I)
IF(A,NE.,0.D0)X=X*EXP(A*K)
IF(A,NE.,0.D0)X=X*DSIN(B*K+C)
15 CONTINUE
C
IF(IPR.LT.2)GO TO 3C
WRITE(ILT,9)
WRITE(ILT,6)(F(K),K=1,NPT)
WRITE(ILT,1)
3C CONTINUE
1 FCPRAT() 6 FCPRAT(20,1,X,FS.2)
9 FCPRAT(10*X,F SIGNAL*)
RETURN
END
SUBROUTINE QUNTZ(F,X,NDIG,IPAD,NPT,NOI)
C ------------------------------------------
DIMENSION F(I),X(NDIG)
DOUBLE PRECISION CT,AC,SC
COMMON /046/I SPs,FMSS,DELTASIG2,DT .01, BIAS,NDI
COMMON /DA1/FbArF,E6AR,FEUm,EESUM
COMMON /IO/IR,ILT,IPP,IFR,FRING
C C
FF=F+EBAR=6
EESUM=0.
C C
BINARY QUANTIZATION
C WASC=SN
E=XX=F(KP*BGTAS
IF(KP.GE.1.0)XX=XX
OIF=ABS(XX-XLEV)
IF(OIF.LE.OE-3)XX=XX
XX=XTM
C WRITE(ILT,21C )XX,0C.OIFXLEV9X0
62 CONTINUE
IF(XX.LT. DEL) XX=0.
1
X(K,11=SN*XQ
a1 CONTINUE
GO DO TO 711
71
C
C BINARY QUANTIZATION
C WORD=SN BIT=59,....,LSB MAX NFG=-2*X+SB
IF(IU,NE.,2)GO TO 551
IF(XMSB.EQ.6.0)X=SB=5.6
NOI=NI-1
DELTA=(2.0*XMSB)/(2.**NOI)
DEL=DELTA/2.0
SIG2=DELTA*DELTA/12.0
WRITE(ILT,6)DELTA,SIG2
DO 61 K=1,NPT
XLEV=0.0
SN=1.0
OXY=F(K)+BIAS
IF(OXY.GE.0.0)SN=1.0
OXY=SN*OXY
DD=2.0*XMSB
XX=2.0*XMSB-DELTA
DO 2 K=1,NOI
O0=DD/2.0
OIF=ABS(OXY-XLEV)
IF(OIF.LE.DEL )XQ=XLEV
XTM=XLEV+DD
IF(OXY,GE.XLEV)XTM=XLEV+DD
XLEV=XTM
C WRITE(ILT,21C )XX,0C.OIFXLEV9X0
62 CONTINUE
IF(OXY.LT. DEL )XQ=0.0
XX(K)=SN*XQ
81 CONTINUE
GO TO 711
C
C4.
C43.
C5.
C6.

DECRALD QUINTIZATION

CONTINUE
IF (IRAD.N.E.10) GO TO 711
AAA=10.0*NCIG
DELTA=1.0/AAA
SIG2=DELTA*DELTA/12.0
WRITE(ILT,49)DELTA,SIG2
DO 31 K=1,NPT
X(K,1)=0.
FB=F(K)+BIAS
XTEM=AES(FB)
SN=1.0
IF(FBLT:.5)SN=-1.0
XTEM=XTEM+AAA
XTEM=XTEM+.5
IX=XTEM
XTEM=IX
XTEM=XTEM/AAA
X(K,1)=SN*XTEM
CONTINUE
C
C SSQ VALUE:
C
711 CONTINUE
DO 211 K=1,NPT
FB=F(K)+BIAS
X(K,2)=X(K,1)-FP
FBAR=FBAR+FB
FBAR=FBAR*X(K,2)
EESUM=EESUM+F(K,2)*X(K,2)
FESUM=FESUM+F(K,2)
IF(ISPN.EQ.0)X(K,1)=X(K,2)
CONTINUE
EESUM=EESUM*0.5
FESUM=2.0*FESUM*0.5
FBAR=FBAR/NPT
FBAR=FBAR/NPT
WRITE(ILT,40)FBAR,EBAR,FESUM,EESUM
IF(IPR.LE.2) GO TO 411
WRITE(ILT,8)
WRITE(ILT,110)X(K,1),K=1,NPT
IF(ISPN.EQ.1) GO TO 411
WRITE(ILT,19)
WRITE(ILT,115)X(K,2),K=1,NPT
WRITE(ILT,11)
411 CONTINUE
999 CONTINUE
C
C FORMAT STATEMENTS
C
8 FORMAT(1G*y,*ROUNDED F+BIAS SIGNAL*)
18 FORMAT(1X,*ROUNDED ERROR F*)
210 FORMAT(2X,5(F5.1,4X))
110 FORMAT(20(1X,F5.2))
115 FORMAT(1X,20(I1,X,F5.3))
178 FORMAT(10X,*HNCISY X MATRIX)
179 FORMAT(10X,*H MATRIX)
449 FORMAT(2X,6HBAR=,E11.4+6H EBAR=,E11.4+5H FB2=,E11.8+4H \ E2=,E11.8+)
449 FORMAT(2X,6HDELTA=,F12.3+5H SIG=,E12.4)
1 FORMAT(/)
RETURN
END
SUBROUTINE COMUT(F,X,NOIG,IPAD,NPT,NDIM)
C ADDS NOI:.
DIMENSION F(1),X(NGIN,1)
DOUBLE PRECISION DT,AD,PC
COMMON /DA0/ISPNN,*SR4,DELTA,SIG2,DT,Q,BIAS,IBIAS
COMMON /DQ1/BAF,EBAR,FESUM,EESUM
COMMON /I7/IR,ILT,IPR,IRECLS
C
C FBA=
FBA=
FESUM=
C NOIG1=NOIG=1
DELTA=(2.1xSIG8)/(2.((**NCIG1)
SIG2=DELTA*DELTA/12.0
WRITE(ILT,489)DELTA,SIG2
C
C IS=2458165
IS2=397665
SIGMA=SQU**(SIG2)
CALL NR4L(NPT,1.1,0.,SIGMA,IS,IS2,X(1,2),0)
DO 26 K=1,NPT
26 X(K,1)=F(K)+BIAS*X(K,2)
C
C DO 211 K=1,NPT
FBA=F(K)+BIAS
FBAR=FBA+FB
EBAR=EBA*X(K,2)
EESUM=EESUM*X(K,2)*X(K,2)
FESUM=FESUM+FB*X(K,2)
IF (ISP2,0.0) X(K,1)=X(K,2)
CONTINUE
C EESUM=EESUM*DT
FESUM=FESUM*FESUM*DT
FBA=FBA/NPT
FBAR=FBAR/NPT
WRITE(ILT,484)FBA,EBAR,FESUM,EESUM
IF (IPR.LE.2) GO TO 411
WRITE(ILT,8)
WRITE(ILT,115)(X(K,1),K=1,NPT)
IF (ISP2,M.0) GO TO 411
WRITE(ILT,18)
WRITE(ILT,115)(X(K,2),K=1,NPT)
WRITE(ILT,11)
C
CONTINUE
999 CONTINUE
C FORMAT STATEMENTS
C
8 FORMAT(18X,16B3) "GROUNDED F SIGNAL)"
18 FORMAT(18X,16B3) "GROUNDED ERROR E)"
210 FORMAT(2X,6(E11.4)
115 FORMAT(2X,1(E11.4,F5.5))
115 FORMAT(2X,1(E11.4,F5.5))
482 FORMAT(2X,6HBAF=,E11.4,6H EBAR=,E11.4,5H FE2=,E11.4,4H EF=,E11.4)
1 FORMAT(/)
RETURN
END
SUBROUTINE INGRAT(K,NPT,NP1,NGIM,INT)
C -----------------------------------------------
DIMENSION X(NGIM,1)
DOUBLE PRECISION IT,SC,9C
COMMON /D/ISPX,MSR,DELTA,SIG2,CT,DI,IBIAS,10IAS
COMMON /IO/ILT,IPR,IPRND
C GENERATE INTEGRATED SIGNALS FROM DATA IN X(K,I)
C INT=1 OR 2 FOR FORWARD INT., -1 FOR REVERSE
C INT=2 FOR UNIT DELAYS X(K,I+1)=X(K,I,II)
N=NFI-1
NFI=NFI+1
GO TO 51,11,11,91,10PR
C FORWARD INTEGRATION
11 CONTINUE
DO 40 J=1,N
JJ=J+1
X(J,JJ)=X(J,JJ)
DO 40 K=2,NPT
K1=K-1
X(K,JJ)=X(K,JJ)+X(K,J)
40 CONTINUE
GO TO 73
C REVERSE INTEGRATION
51 CONTINUE
DO 60 J=1,N
JJ=J-1
X(NPT,1)=X(NPT,1)
DO 60 K=2,NPT
K1=K+1
K=NPT-1-K
DO 60 xxxxx
K=NPT-1-K
Y(K,JJ)=Y(K,JJ)
60 CONTINUE
IF(IPR.GT.0.4)GO TO 62
IPWR=1B1A -1
DO 61 K=2,NPT
TIME=DT*KK
K=NPT-1-K
61 X(K,NPT)=TIME**IPWR
62 CONTINUE
GO TO 73
C GENERATE UNIT DELAYS
91 CONTINUE
DO 93 I=2,NFI
II=I-1
X(1,II)=X(1,II)
DO 93 K=2,NPT
K1=K-1
93 X(K,II)=X(K,II)
GO TO 81
70 CONTINUE
SC=1.6
DO 80 J=2,NFI
SC=SC*0.7
DO 80 K=2,NPT
80 X(K,1)=SC*X(K,1)
81 CONTINUE
IF(IPR.LT.41)GO TO 99
IF(IPRND.EQ.1)WRITE(ILT,171)
IF(IPRND.EQ.0)WRITE(ILT,179)
DO 100 K=1,NPT
100 WRITE(ILT,111)X(K,1),K=1,NPT
C WRITE(ILT,1)
99 CONTINUE
C

C110 FORMAT(4(1X,F12.6))
110 FORMAT(25(1X,F5.2))
178 FORMAT(10,1X,NOISY X MATRIX)
179 FORMAT(10X,8X MATRIX)
1 FORMAT(/)
RETURN
END

SUBROUTINE FIX(G,F,C,D,X,Y,NC,SIG,NDIM,IFIX)
C
C
C ESTIMATE NOISE INTENSITY SIG (ASSUME WHITE NOISE)
C CORRECT NOISY MATRIX = C
C (P) DENOTES NOISE MATRIX FOR UNIT NOISE
C NC IS THE NONZERO SUBMATRIX OF P = C*G*OF NOISE
C
DIMENSION G(NDIM,1),P(NDIM,1),D(NDIM,1),X(1),Y(1)
IF(IFIX.EQ.0) GO TO 51
JCT=0
SIG=0.0
3 SUMDET=0.0
CALL GKROCT(G,D,GOET,X,G,NDIM,0)
JCT=JCT+1
IF(JCT.EQ.1) DET=GOET
DO 5 J=1,NC
DO 7 II=1,N
DO 7 JJ=1,N
C(II,JJ)=G(II,JJ)
IF(JJ.EQ.I) DET=DET*P(II,II)
7 CONTINUE
CALL GKROCT(G,P,OGET,X,P,NDIM,3)
SUMDET=SUMDET*DET
5 CONTINUE
SIG=GOET/SUMDET
51 CONTINUE
DO 9 I=1,N
DO 9 J=1,N
C(I,J)=G(I,J)-SIG*F(I,J)
IF(IFIX.EQ.0) GO TO 11
CALL GKROCT(G,D,GOET,X,G,NDIM,0)
IF(GOET.LT.3.0) GO TO 51
IF(JCT.GT.5) GO TO 11
IF(GOET/DETG.GT.0.1) CALL FEOUTA(N,N,G,NC,IFIX)
SIG=SIG-ST
11 RETURN
END

SUBROUTINE BUILOR(A,X,N,PAX)
C
C CONVERSION MATRIX FROM REVERSE INTEGRATION -- I. R. MODELING
C DIMENSION A(MAX,13),X(11),Y(11)
C DOUBLE PRECISION CT,Y
C COMMON /DA0/ISP,WMSB,DELTA,SIG2,CT,0I,IAS,18IAS
C COMMON /IO/IR,ILT,IPR,IRCUND
NM=N-1
DO 11 I=1,N
Y(I)=0.0
DO 11 J=1,N
A(I,J)=0.0
A(N,N)=1.0
11 CONTINUE
C
NO 20 J=1,N,CM1
J=4-JJ
75
C
DIMENSION XLAPDA(1), Y(MAX,1)
DOUBLE PRECISION A*E,C,D,E
INTEGER NL,H2,Z2
DOUBLE PRECISION CT, SC, HD, RC
COMMON /IO/ ISDN, XPSI, DELTA, SIG2, CT, SIG1, RIAS, INTAS
COMMON /IL/ IL, ILT, IPR, IPRC, ICOND
IGKR=1 IF(N,N.E.1) GO TO 3
Y(1,1)=1.
DET=DET*1.
GO TO 61 3
CONTINUE
DO 6 I=1,N
Y(I,1)=X(I,1)
A=1.0
GO TO 43
6 CONTINUE
Y(J,1)=K(J,1)
GO TO 61
43 CONTINUE
I=I+1
GO TO 61
5 CONTINUE
IF (ABS(Y(K,J)).LE.DABS(9.)) GO TO 18
B=ABS(Y(K,J))
G=B
18 CONTINUE
CONTINUE
M=K
L=J
19 CONTINUE
IF(L.EQ.1) GO TO 24
DO 23 J=1,N
C=Y(L,J)
Y(L,J)=Y(I,J)
23 Y(I,J)=C
C
INTERCANGE COLUMNS

24 IF(N.EQ.I)GO TO 29
DO 28 J=1,N
C=Y(J,I)
Y(J.I)=Y(J.I)
29 NUM(1.I)=I
NUM(2.I)=1
A=Y(I.I)
Y(I.I)=A
DO 42 J=1,N
IF(J.EQ.I)GO TO 42
C=-Y(I,J)
Y(I,J)=B
DO =I K=1,N
D=Y(K.I)*C
E=Y(K,J)*B+D
41 IF(CABS(E)-LT.1.0)GO TO 42
42 CONTINUE
A=3

RESTORE COLUMNS

DO 58 I=2,N
J=N+1-I
K=NUM(2,J)
IF(K.EQ.I)GO TO 52
DO 51 L=1,N
C=Y(K,l)
Y(K.L)=Y(J,L)
51 Y(J.L)=C
CONTINUE

A=0

RESTORE ROWS

IF(K.EQ.J)GO TO 5A
DO 57 L=1,N
C=Y(L,K)
Y(L.J)=Y(L,J)
57 Y(L.J)=C
CONTINUE

IF(COPT.NE.1)GO TO 1000
IF(Y(I,1).LT.0.0)GO TO 1000
A=X(1)
B=Y(I,1)
C=X(1)/Y(I,1)
X(1)=B/C
X(1)=ABS(X(1))

1000 CONTINUE

DO 200 I=2,N
SC=SC+D
IF(IGKR.EQ.0)XLAMDA(I)=Y(I,1)/Y(I,1)
IF(IGKR.EQ.0)GO TO 200
A=Y(I,1)
B=Y(I,1)
C=Y(I,1)/B
D=Y(I,1)/C
E=Y(I,1)/C

IF(Y(I,1).LT.0.0)GO TO 1000

1000 CONTINUE

X(1)=1.000
IF(IPR.GE.1)WRITE(6,106)(XLAMDA(I),I=1,N)
NPP=N
IF(IBM4.NE.0)NPP=N-1
CALL BUILDITY(KLAMCA,NPP,MAX)
100 FORMAT(5X,*SYNTHETIC PARAMETER VECTOR*,/,10G12.5)
100 CONTINUE
RETURN
END

FUNCTION COMB(N,M)
C
CALCULATE COMBINATION M CUT OF N
IF(N.LE.G)GO TO 99
L=1
LD=1
IF(M.EQ.0)GO TO 8
MN=1
5 L=L+1
DO 7 I=1,M
7 LD=LD*I
8 CONTINUE
99 RETURN
END

SUBROUTINE BUILDIT(Z,NP1,NPT,NCIP,NFIX)
C
DIMENSION Z(NP1,1),R(1)
COMMON /OIPSNS,KMSP,DELTA,GSQ,AS,IAS,IAS
COMMON /OIPSNS,ILT,IPR,IPCNC
TIME=DT*NOT
IFP=NFIX
GO TO 201.1,101.2(1)
101 SC=DT
DO 24 I=1,NP1
24 Z(I,1)=SC*NPT
IF(I.GE.2)Z(I,2)=DT*SC*COPRNPT-1,I
24 SC=SC*DT
DO 166 J=3,NP1
166 Z(I,J)=Z(J,J-1)*AC
166 AC=1.0/(I+J-2)
166 CONTINUE
DO 210 J=1,NP1
210 Z(I,J)=Z(J,J-1)*AC
210 CONTINUE
WRITE(ILT,160)
160 FORFIT(10EF QUANT.
170 CONTINUE
RETURN
END

SUBROUTINE PRINTMAT(A,M,N,NCIM,LOC)
C
C PRINTER A MATRIX, AND AN INTEGER (perhaps A LOCATION) IF LOC.GE.1
C
DIMENSION A(MDIM,1)
IF(LG.GE.1)WRITE(6,1)LOC
DO 31 I=1,N
31 WRITE(6,19)EA(I,J),J=1,N
5 FORMAT(* LOCATION/INTEGER=*,I5)
19 FORMAT(2X,10G13.6)
RETURN
END
SUBROUTINE RESPON(K,Y,N,GAMA,XLAMCA,NPI)

C******************************************************************************
DIMENSION X(I),Y(J),GAMA(J),XLAMCA(J)
DOUBLE PRECISION XSAV,A0,80
NPI=N-1
NPF=N+1
NPNP1=N*N+1
NPNP2=N*N+2
DO 19 I=1,NPI
19 XLAMDA(I)=0.0
XSAV=0.0
DO 26 K=1,NPI
IF(N.EQ.1)GO TO 25
DO 21 I=1,NPI
J=NPI-I
21 XLAMDA(I)=XLAMDA(I-1)
CONTINUE
DO 22 I=1,NPI
J=NPNI-I
22 XLAMCA(J)=XLAMCA(J-1)
XLAMDA(I)=XSAV
XLAMDA(NPI)=Y(K)
XSAV=0.0
DO 23 I=1,NPNP1
23 XSAV=XSAV-GAMA(I)*XLAMCA(I)
IF(DABS(XSAV).GE.1.0E10)XSAV=C.3
20 Y(K)=XSAV
RETURN
END
SUBROUTINE VEQUAT(NPT,Y,X,NPUL,IOFT)
C******************************************************************************
C IOPT=0 SET Y TO ZERO
C 1 OR 2 SET Y=X (PRINT IF 2)
C 3 SET Y=Y+GOMET X
C 9 SET Y TO ZERO
C 10 SET Y=STEP
C 11 SET Y=STEP
DIMENSION X(I),Y(J)
IF(IOPT.EQ.3)IOPT=9
DO 33 K=1,NPT
IF(IOPT.EQ.1,Y(K)=X(K))
IF(IOPT.EQ.3,Y(K)=Y(K)+X(K))
IF(IOPT.EQ.2,Y(K)=0.0)
IF(IOPT.EQ.11,Y(K)=1.0)
33 CONTINUE
IF(IOPT.EQ.2)WRITE(6,6)Y(K),K=1,NPT
6 FORMAT(2X,10G12.5)
IF(IOPT.EQ.10)Y(1)=1.0
RETURN
END
SUBROUTINE MEQUAT(M,N,B,A,NDIP,IOPT)
C******************************************************************************
C IOPT=0 SET b TO ZERO
C 1 B EQUAL TO A
C 10 B TO IDENT
DIMENSION A(NDIP+1),B(NDIP+1)
DO 33 I=1,M
DO 33 J=1,N
IF (I/OPT.NE.1) B(I,J)=0.0
IF (I/OPT.EQ.1.AND.I.EQ.J) B(I,J)=1.0
IF (I/OPT.EQ.1) B(I,J)=A(I,J)
33 CONTINUE
RETURN
END

SUBROUTINE PLOT(NPT,NF,Y,NDIM,TI,CT,LABEL,LNEP,LF,NF)
C
C...NUMBER OF TIME FITS WARNING: NNE SHOULD BE LGE NPT+2
C
C...NUMBER OF FUNS
C
C Y(K,1) DATA ARRAY OF DIMENSION NDIM,NF
C T0=INITIAL TIME, DT=TIME INCREMENT
C LABEL, INDEP = TITLES FOR Y AND X AXES
C DIMENSION Y(NDIM,NF),Y(2),LABEL(1),INDEP(1)
C DIMENSION X(512),IBUF(512)
COMMON /IO/IR,1LTU,TPLT
C
M=NPT*NDF
M1=M+1
M2=M+2
NPT1=NPT+1
NPT2=NPT+2
X(1)=T0
DO 9 K=2,NPT
X(K)=X(K-1)+DT
9 DO 8 I=1,NF
DO 6 K=NPT1,NDF
C
C INITIALIZATION (L0,INK,12IN,FAPF) MAY LENGTH=6 CI1
CALL PLOTMX(60.0)
C
C SET OPTIONS
C
CALL PLOT(6,-5.31)
CALL FACT R(6.0/6.5)
C
BEGIN PLOTTING
CALL SCALE(X(6.5,NPT+1)
CALL SCALE(Y(1,1),12.5,NPT+1)
CALL AXIS(C,...,113,TIME (SEC.),...
*16.6,5.2,...X(NPT1),Y(NPT2))
CALL AXIS(0.0,0.0,RESPONSES Y Y ,...
*16.11,92.0,Y(PT1),Y(PT2)
WRITE(6,6)X(NPT1),X(NPT2)
WRITE(6,7)Y(PT1),Y(PT2)
6 FORMAT(1X,T6.5,DIV),4(I1,F7.3))
7 FORMAT(IX,YL,DIV (17,01),4(I1,F7.3))
DO 10 I=1,NF
Y(NPT1,I)=Y(I)
Y(NPT2,I)=Y(I)
10 IF (I.EQ.1.OR.IPLT.EQ.3) CALL LINE(X,Y(1,1),NPT1,1,1-1,1)
IF (I.EQ.2.AND.IPLT.EQ.2) CALL DASHLN(X,Y(1,2),NPT,1)
CONTINUE
CALL PLOT(15,33,-3)
RETURN
END
APPENDIX C

LISTING OF

PROGRAM

USPEC
USPEC

"USPEC" FOR R40C
EVALUATES THE MAGNITUDE SPECTRUM OF THE FOLLOWING PULSE INPUTS (SEE SECTION 4 OF THE REPORT): 
1. TR+-(T) ONE CYCLE TRI WAVE, FOLLOWED BY ZERO LEVEL 
2. TP+-(T) ONE HALF CYCLE TRI WAVE, FOLLOWED BY ZERO LEVEL 
3. CEV+(T) OSCILLATORY PULSE (NO DECAY), FOLLOWED BY ZERO LEVEL 
4. CEY+(T) SAME AS 3, WITH EXP. DECREASING ENV. - ONE TIME CONST 
5. OEV+(T) OSCILLATORY PULSE WITH TRI ENV., FOLLOWED BY ZERO LEVEL 
FOR PULSES 3 TO 6 THE OSCILLATION IS A COSINE WAVE

DIMENSION N(1:1,F(1:2),F(1:2),D(1:2),I(2:5)
DATA N(1/-25/4(21/110/4/31/20/0/48/100/8/11/22/1/2/3/4/5/22/11
DATA F(1/-01/1/40/2/11/-4/5/5/5/0/-5/1/31/4/-/-3/3/4/-1/42/51/42/3
DATA F(3/-09/-2/7/1/2/1/31/1/-1/2/1/-1/3/-1/3/-3/3/4/-1/-3/31/31
DATA F(1/-3/-3/0/1/-3/-3/-3/-3/4/-3/-1/3/-3/-3/-3/3/-3/3/-3/3/-3/3
P1=+.3*AT/N(1,2)
P2=2,3*PI
NN=3
NF=13
IS1=1
IS2=6
NF=13
DO 31 J=1,NF
IF (J(J),F(J)=1,1)NF=NF+1
30 CONTINUE
I=TEST=0
IF (TEST,NE.1)GO TO 29
N(I)=10
F(I)=0:1
IS1=1
IS2=IS1
NN=1
NF=1
IR=J
31 CONTINUE
I=TEST=0
IF (TEST,NE.1)IPF=0
DO 100 ISG=0,401
NFF=NF
IF (ISIG=1F(2)=2) NFF=NF-NF
DO 95 J=1,NN
DO 94 J=1,NFF
F(J)=F(J)-1,1
IF (ISIG.EQ.6)AND,F(J).LT,.5)FR(J)=-1,0
IF (ISIG.EQ.3)AND,JTEST.EQ.1100 TO 31
F(J)=F(J+NFF)-1,0
31 CONTINUE
GO TO (4,1,4,5,1,5,1,5,1,7)ISIG
41 CONTINUE
NF=10,NSG=21*N(1)
IF (IPX .GE. 3) WRITE (6, 13) BF
BF = BF * FR(J)
IF (IPX .GE. 3) WRITE (6, 13) THETA = PI*FR
IF (ISIG .EQ. 2) WRITE (6, 13) THETA = 2.0 * THETA
IF (ISIG .EQ. 2) WRITE (6, 13) THETA = 1.0
IF (THETA .EQ. 0.0) XY = SIN(THETA) * THETA
IF (IPX .GE. 2) WRITE (6, 13) XY = XY
IF (IPX .GE. 3) WRITE (6, 13) YY = YY
IF (ISIG .EQ. 1) IF (THETA .EQ. 2.0 * THETA)
IF (ISIG .EQ. 1) XX = XX
51 CONTINUE
A = 5.0
IF (ISIG .EQ. 1) B = 6.0 * N(I)
IF (ISIG .EQ. 4) B = 2.0 * N(I)
IF (ISIG .EQ. 3) B = 4.0 * N(I)
EA = 2.0 * A * B
IF (IPX .GE. 2) WRITE (6, 13) 
XX = 0.0
XI = 2.0
DO 57 KF = 2, 2
FF = FR(J1 + 2, 2 * (KF - 1))
C = GSC(WB) * S(4 * WB)
EF = SIN(WB)
IF (IPX .GE. 2) WRITE (6, 13) EF
VR = 1.0 - EA * C
YT = EA * Y
PT = A * W + W
PT = A * W
YT = 2.0 / Y
XI = XI + ZI*0
57 CONTINUE
XY = 2.0 * XY + XY * YT
IF (IPX .GE. 2) WRITE (6, 13) 
YY = XY
X = 2.0
DO 67 XX = 2, 2
FF = FR(J1 + 2, 2 * (KF - 1))
XY = BF * FF
THETA = PI*FR
60 IF (THETA .EQ. 0.0) CF = CF
IF (THETA .EQ. 0.0) CF = CF
C = COS(THETA)
S = SIN(THETA)
YY = 2.0 * W * CF
YY = YY
XY = XY
X = XX
67 CONTINUE
XY = SQRT(XP*X + XP + XP)*X
IF (IPX .GE. 2) WRITE (6, 13) 
X = XX
IF (J .EQ. N(J1) ) GO TO 93
GO TO 93
83
CONTINUE
X(J,I)=XY
CONTINUE
DO 96 K=J1,NFF
XX=X(J,I)
IF(ISIG.GT.3.AND.NNF.EQ.NF1)*XX=XX(J,JX)/X(I,J)
X(J,I)=2.0
IF(xx.GE.1.2E-15)X(J,I)=2L.*4LOG10(xx)
CONTINUE
95 CONTINUE
WRITE(6,11)ISIG,(N(I),I=1,NN)
WRITE(6,11)
CONTINUE
97 CONTINUE
WRITE(6,11)FFOR,(X(I,J),J=1,N)
WRITE(6,11)
CONTINUE
C
C
1 FORMAT(/)
6 FORMAT(2X,F12.3,F12.1)
7 FORMAT(2X,THETA=*,F13.6)
8 FORMAT(2X,Y=*,F17.6)
9 FORMAT(2X,Y=*,F13.6)
11 FORMAT(2X,*FFOR. IN TRY FOR DIFFERENT freq.,NFT=*, ISIG=, 10(4x,4x,F15.4))
12 FORMAT(2X,*FFOR. IN TRY FOR DIFFERENT freq.,NFT=*, ISIG=, 10(4x,4x,F15.4))
13 FORMAT(2X,*FFOR. IN TRY FOR DIFFERENT freq.,NFT=*, ISIG=, 10(4x,4x,F15.4))
14 FORMAT(2X,*FFOR. IN TRY FOR DIFFERENT freq.,NFT=*, ISIG=, 10(4x,4x,F15.4))
15 FORMAT(2X,*FFOR. IN TRY FOR DIFFERENT freq.,NFT=*, ISIG=, 10(4x,4x,F15.4))
21 FORMAT(2X,*FFOR. IN TRY FOR DIFFERENT freq.,NFT=*, ISIG=, 10(4x,4x,F15.4))
22 FORMAT(2X,*FFOR. IN TRY FOR DIFFERENT freq.,NFT=*, ISIG=, 10(4x,4x,F15.4))
STOP
END
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