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MEMORANDUM

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**TWO-PARAMETER EXPONENTIAL
AND RATIONAL FUNCTIONS FOR
LEAST-SQUARE APPROXIMATIONS**

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PREFACE

As a contribution to The RAND Corporation's studies in system analysis and synthesis, a scheme is delineated in this report for approximating prescribed system characteristics by sums of exponentials or by rational functions. The analytical results and computational aids which are derived are applicable to a broad class of system optimizations, such as those found in radar filter design; numerical examples are provided to illustrate their application to practical and important design problems. The contents of this Memorandum should be of interest to the Air Force Systems Command, as well as to others concerned with numerical methods of system design and signal analysis.

SUMMARY

In many system design problems, the representation of certain system attributes must be in terms of exponentially damped sinusoids or of rational functions in order to be physically meaningful. In this Memorandum, two sets of orthonormal elements are derived which should be useful for such approximation problems. One set constitutes a basis for exponential approximation and the other a basis for rational function approximation.

The closure properties of the two-parameter exponential and rational functions are examined first. Expressions are then presented for efficiently determining the orthonormal elements of each basis. Important characteristics of the sets are also discussed, and special relations among the coefficients generating the bases are deduced.

Once the general relations for the orthonormal elements are developed, the two sets are applied to typical approximation problems encountered in signal processing and system design. The computations involved in the solution of these problems are illustrated in the final portion of the study. The computer programs used for the sample problems are described in the appendices and should be helpful in similar design situations.

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I. INTRODUCTION

To ensure physical realizability in the synthesis of a system, it is frequently necessary to represent certain of the system's attributes by approximants other than algebraic polynomials. Such is the case, for instance, in the determination of optimum filters for smoothing and prediction where it is convenient to approximate empirical or analytical spectra of random processes by readily factorized rational functions.[†] The utility of exponential functions is well known in the design of linear networks for a prescribed transient response. Similarly, sums of exponentials often afford the most suitable approximations to cross-correlation measurements, to radioactive decay and gas absorption data, to mass spectrographs, and to ultracentrifuge analysis curves.

In the present treatment of rational and exponential function representations, two orthonormal bases for exponential and rational function approximations are derived. The bases consist of two-parameter elements which provide more efficient minimum mean-square-error approximation than two corresponding one-parameter sets investigated previously.⁽¹⁾ After the closure properties of the two orthonormal bases are examined in detail, new expressions are developed for efficiently generating the individual orthonormal elements. Several relations are then deduced which connect important properties of each basis. Useful identities and numerical techniques involving the basis coefficients are derived

[†] The term "rational function" denotes a ratio of algebraic polynomials in which the denominator polynomial is not generally a constant.

which obviate storage of either the form or selected values of the orthonormal approximants. Finally, several numerical examples are provided to illustrate both algorithms for exponential and rational function approximation.

II. SCOPE OF INVESTIGATION

In a previous investigation, the linearly independent functions

$$\hat{\varphi}_n(t) = \begin{cases} e^{-n\sigma t} & 0 < t < \infty \\ 0 & t \leq 0 \end{cases} \quad (1)$$

$$n = 1, 2, \dots; \operatorname{Im}(\sigma) = 0; \operatorname{Re}(\sigma) > 0$$

and

$$\hat{\psi}_n(\omega) = \frac{1}{n\sigma + j\omega} \quad -\infty < \omega < \infty \quad (2)$$

$$n = 1, 2, \dots; \operatorname{Im}(\sigma) = 0; \operatorname{Re}(\sigma) > 0$$

were examined in detail.¹ The more general sets comprised of

$$\varphi_n(t) = \begin{cases} e^{-n(\sigma - j\beta)t} & 0 < t < \infty \\ 0 & t \leq 0 \end{cases} \quad (3)$$

$$n = 1, 2, \dots; \operatorname{Im}(\sigma) = \operatorname{Im}(\beta) = 0; \operatorname{Re}(\sigma) > 0$$

and

$$\psi_n(\omega) = \frac{1}{n\sigma + j(\omega - n\beta)} \quad -\infty < \omega < \infty \quad (4)$$

$$n = 1, 2, \dots; \operatorname{Im}(\sigma) = \operatorname{Im}(\beta) = 0; \operatorname{Re}(\sigma) > 0$$

will be treated in this Memorandum. The primary motivation for such a study is that for a fixed number of orthonormal elements, an expansion of a prescribed function in terms of $\varphi_n(t)$ or $\psi_n(\omega)$ provides a minimum mean-square-error less than or equal to that for $\hat{\varphi}_n(t)$ or $\hat{\psi}_n(\omega)$, respectively.

¹See Tables 1-7 of Ref. 1.

In the remaining sections, the results obtained for the one-parameter sets, Eqs. (1) and (2), are extended to $\{\varphi_n(t)\}$ and $\{\psi_n(\omega)\}$, and numerical schemes are derived for simplifying least-square representations involving $\{\varphi_n(t)\}$ and $\{\psi_n(\omega)\}$. In so doing, the following topics are considered:

- o Closure of $\{\varphi_n(t)\}$ and $\{\psi_n(\omega)\}$
- o Determination of the orthonormalized sets[†]

$$X_m(t) = \sum_{n=1}^m \gamma_{mn} \varphi_n(t), \quad Y_m(t) = \frac{1}{\sqrt{2}} [X_m(t) + X_m^*(t)]$$

and

$$U_m(\omega) = \sum_{n=1}^m \lambda_{mn} \psi_n(\omega), \quad V_m(\omega) = \frac{1}{\sqrt{2}} [U_m(\omega) + U_m^*(\omega)]$$

- o Properties of X_m , Y_m , U_m , V_m , γ_{mn} , and λ_{mn}
- o Selection of σ and β
- o Computational aspects of approximants

$$g(t) \approx \sum_{m=1}^M a_m X_m(t) \quad \text{and} \quad h(\omega) \approx \sum_{m=1}^M b_m U_m(\omega)$$

[†]The symbol * denotes complex conjugate.

III. CLOSURE

In order to approximate prescribed functions arbitrarily closely by linear combinations of the elements $\varphi_n(t)$ and $\psi_n(\omega)$, it is necessary to verify the closure of the sets $\{\varphi_n(t)\}$ and $\{\psi_n(\omega)\}$. Since the functions $\varphi_n(t)$ and $\psi_n(\omega)$ are related through a linear operation, the Fourier transform, it will be possible to demonstrate closure of $\{\psi_n(\omega)\}$ in the space $L^2(-\infty, \infty)$ once closure of $\{\varphi_n(t)\}$ in $L^2(0, \infty)$ is shown.

An assemblage of functions $\{\varphi_n(t)\}$ of integrable square[†] is closed over (a, b) if the integral

$$\int_a^b y(t) \varphi_n^*(t) dt = 0 \quad n = 1, 2, \dots \quad (5)$$

implies that $y(t) \in L^2(a, b)$ vanishes everywhere in (a, b) except perhaps over a set of measure zero. The closure property derives its significance from a theorem that states that a set of functions is closed if and only if it is complete.⁽²⁾ Moreover, if a set $\{\varphi_n(t)\}$ is complete, then for any function $y(t) \in L^2(a, b)$ and any positive ϵ , there is a sum function

$$\hat{y}(t) = \sum_{n=1}^N a_n \varphi_n(t) \quad (6)$$

such that

$$\int_a^b |y(t) - \hat{y}(t)|^2 dt < \epsilon \quad (7)$$

[†]This property is symbolized by $\varphi_n(t) \in L^2(a, b)$.

Equations (5) and (6) represent a special case of the general situation in which a closed finite or infinite system of elements $\{y_n(x)\}$ in a Hilbert space Y permits arbitrarily close approximation (in the L^2 norm) to every element $y(x) \in Y$ by a finite linear combination of the $y_n(x)$. To establish this property for the elements $\varphi_n(t)$ in Eq. (3), Szasz's theorem[†] can be invoked:

A necessary and sufficient condition for closure $L^2(0,1)$ of the functions x^{λ_n} , $\text{Re}(\lambda_n) > -\frac{1}{2}$, is divergence of

$$\sum_{n=1}^{\infty} \frac{1 + 2\text{Re}[\lambda_n]}{1 + |\lambda_n|^2} \quad (8)$$

From the previous definition of closure, if $\{x^{\lambda_n}\}$ is not closed in $L^2(0,1)$, a function $y(x) \in L^2(0,1)$ exists which is orthogonal to every element x^{λ_n} ; that is

$$0 < \int_0^1 |y(x)|^2 dx < \infty \quad (9)$$

and

$$\int_0^1 y(x) x^{\lambda_n^*} dx = 0 \quad n = 1, 2, \dots \quad (10)$$

Conversely, if the set $\{x^{\lambda_n}\}$ is closed, a function $y(x) \in L^2(0,1)$, which is not identically zero, does not exist so that Eqs. (9) and (10) are satisfied. Accordingly, with the change of variable $x = e^t$ in the set $\{x^{\lambda_n}\}$, the conditions given in Eqs. (9) and (10) become

[†] See Ref. 2, pp. 32-36.

$$0 < \int_{-\infty}^0 |y(e^t)|^2 e^t dt < \infty \quad (9')$$

and

$$\int_{-\infty}^0 y(e^t) e^{t(1+\lambda_n^*)} dt = 0 \quad n = 1, 2, \dots \quad (10')$$

With the further transformation $y(e^t) e^{t/2} = \bar{y}(t)$, the last two relations give

$$0 < \int_{-\infty}^0 |\bar{y}(t)|^2 dt < \infty \quad (9'')$$

and

$$\int_{-\infty}^0 \bar{y}(t) e^{t(\frac{1}{2}+\lambda_n^*)} dt = 0 \quad n = 1, 2, \dots \quad (10'')$$

Thus, closure of the functions x^{λ_n} in the space $\bar{y}(x) \in L^2(0,1)$ is equivalent to existence of a function $\bar{y}(t)$ satisfying Eqs. (9'') and (10''). This in turn is equivalent to the closure of the functions $e^{t(\frac{1}{2}+\lambda_n)}$ in $L^2(-\infty,0)$, or $e^{-t(\frac{1}{2}+\lambda_n)}$ in $L^2(0,\infty)$.

In the application of interest, where $\varphi_n(t) = e^{-n(\sigma-j\beta)t}$, λ_n satisfies the equation

$$\lambda_n = (-\frac{1}{2} + n\sigma) - jn\beta \quad (11)$$

$$n = 1, 2, \dots; \quad \text{Im}(\sigma) = \text{Im}(\beta) = 0; \quad \text{Re}(\sigma) > 0$$

Consequently,

$$\text{Re}[\lambda_n] = -\frac{1}{2} + n\sigma > -\frac{1}{2} \quad n = 1, 2, \dots; \quad \text{Re}(\sigma) > 0, \quad \text{Im}(\sigma) = 0 \quad (12)$$

as required by Szasz's theorem. Substituting this value of λ_n in Eq. (8) yields

$$\sum_{n=1}^{\infty} \frac{1 + 2(-\frac{1}{2} + n\sigma)}{1 + (-\frac{1}{2} + n\sigma)^2 + (n\beta)^2} = \sum_{n=1}^{\infty} \frac{2n\sigma}{n^2(\sigma^2 + \beta^2) - n\sigma + 5/4} \quad (13)$$

According to Szasz's theorem, closure of $\{\varphi_n(t) = e^{-n(\sigma-j\beta)t}\}$ in the space $\bar{y}(t) \in L^2(0, \infty)$ rests on divergence of this sum.

Application of the integral test⁽³⁾ to the sum in Eq. (13) indicates that divergence of the infinite series depends on divergence of the integral

$$\int_1^{\infty} \frac{2\sigma\xi \, d\xi}{1 + (-\frac{1}{2} + \sigma\xi)^2 + (\beta\xi)^2} \quad (14)$$

The resultant integration is

$$\begin{aligned} & \frac{2\sigma^2}{(\sigma^2 + \beta^2)\sqrt{4\sigma^2 + 5\beta^2}} \tan^{-1} \left[\frac{2(\sigma^2 + \beta^2)\xi - \sigma}{\sqrt{4\sigma^2 + 5\beta^2}} \right] \Big|_1^{\infty} \\ & + \frac{\sigma}{(\sigma^2 + \beta^2)} \log [(\sigma^2 + \beta^2)\xi^2 - \sigma\xi + 5/4] \Big|_1^{\infty} \end{aligned} \quad (15a)$$

or

$$\begin{aligned} & \frac{2\sigma^2}{(\sigma^2 + \beta^2)(4\sigma^2 + 5\beta^2)^{\frac{1}{2}}} \left\{ \frac{\pi}{2} - \tan^{-1} \left[\frac{2(\sigma^2 + \beta^2) - \sigma}{(4\sigma^2 + 5\beta^2)^{\frac{1}{2}}} \right] \right\} - \frac{\sigma}{(\sigma^2 + \beta^2)} \log [\sigma^2 + \beta^2 - \sigma + 5/4] \\ & + \lim_{\xi \rightarrow \infty} \frac{\sigma}{2(\sigma^2 + \beta^2)} \log [(\sigma^2 + \beta^2)\xi^2 - \sigma\xi + 5/4] \rightarrow \infty \end{aligned} \quad (15b)$$

Since σ^2 and β^2 are positive and bounded, the last term of Eq. (15b) is unbounded. Consequently, the series of Eq. (13) is divergent, and the closure conditions of Szasz's theorem are satisfied by the set

$\{\varphi_n(t)\}$. It follows, therefore, that $\{\varphi_n(t) = e^{-n(\sigma-j\beta)t}\}$ is also complete in the space of functions $y(t) \in L^2(0, \infty)$.

The closure of $\{\psi_n(\omega)\}$ in the transform space $D(\omega) \in L^2(-\infty, \infty)$ is deducible directly from a lemma of Wiener's on invariance of closure.¹ His lemma states that the closure of a set of functions is preserved in any linear transformation which carries the whole of L^2 into itself and which conserves the integral of the squared modulus of each function. Quantitatively, the lemma states that:

Given a linear transformation such that to every function

$f(x) \in L^2(a, b)$, there corresponds a $g(y) \in L^2(c, d)$, if

$f_j(x) \rightarrow g_j(y)$ and

$$i. \quad c f_j(x) \rightarrow c g_j(y)$$

$$ii.^{11} \quad f_1(x) + f_2(x) \rightarrow g_1(y) + g_2(y)$$

$$iii. \quad \int_a^b |f_j(x)|^2 dx = \int_c^d |g_j(y)|^2 dy \quad j = 1, 2, \dots$$

then the closure properties of a sequence $\{f_j(x)\}$ are the same as those of $\{g_j(y)\}$.

The transformation which carries the set $\{\varphi_n(t)\}$ into the set $\{\psi_n(\omega)\}$ is the Fourier transform, since

$$\begin{aligned} \mathfrak{F} \{\varphi_n(t)\} &\equiv \int_{-\infty}^{\infty} \varphi_n(t) e^{-j\omega t} dt = \int_0^{\infty} e^{-(n\sigma+j\omega-jn\beta)t} dt \\ &= \frac{-e^{-(n\sigma+j\omega-jn\beta)t}}{n\sigma+j\omega-jn\beta} \bigg|_0^{\infty} = \frac{1}{n\sigma+j(\omega-n\beta)} \equiv \psi_n(\omega) \end{aligned} \quad (16)^{111}$$

¹ See Ref. 2, pp. 28-30.

¹¹ The symbol $\rightarrow$ denotes correspondence.

¹¹¹ $\mathfrak{F}$ denotes the Fourier transform.

Because the Fourier transform is a linear operator, properties (i) and (ii) of Wiener's lemma are satisfied. Moreover, Parseval's theorem⁽⁴⁾ indicates that

$$\int_0^{\infty} |\varphi_n(t)|^2 dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |\psi_n(\omega)|^2 d\omega = \int_{-\infty}^{\infty} |\psi_n(2\pi f)|^2 df \quad (17)$$

where $\omega = 2\pi f$. Hence, all the conditions of the closure invariance lemma are fulfilled, and the set $\{\psi_n(2\pi f)\}$ is also closed in $L^2(-\infty, \infty)$.

IV. ORTHONORMALIZATION OF $\{\varphi_n(t)\}$ and $\{\psi_n(\omega)\}$

The closure properties discussed in Section III ensure completeness of $\{\varphi_n(t)\}$ in the space of squared-integrable functions $d(t) \in L^2(0, \infty)$ and completeness of $\{\psi_n(\omega)\}$ in $D(\omega) \in L^2(-\infty, \infty)$. Consequently, the theory of Hilbert spaces can be applied to two further items of computational importance: orthonormalization of $\{\varphi_n(t)\}$ and $\{\psi_n(\omega)\}$, and representations of functions $d(t) \in L^2(0, \infty)$ by sums of elements $\varphi_n(t)$ and $\psi_n(\omega)$.

The task of orthonormalizing the set $\{\varphi_n(t)\}$ (or, analogously, the set $\{\psi_n(\omega)\}$) entails finding coefficients γ_{mn} in the functions $X_m(t)$,

$$X_m(t) \equiv \sum_{n=1}^m \gamma_{mn} \varphi_n(t) \quad m = 1, 2, \dots \quad (18)$$

such that

$$\int_0^\infty X_r(t) X_s^*(t) dt = \delta_{rs} \quad (19)'$$

The existence of these constants γ_{mn} is guaranteed by Theorem IV.1 of Ref. 5:

Let $\varphi_1, \varphi_2, \dots$ be a finite or infinite sequence of elements such that any finite number of elements $\varphi_1, \varphi_2, \dots, \varphi_k$ are linearly independent. Then constants

$$\begin{array}{ccc} \gamma_{11} & & \\ \gamma_{21} & \gamma_{22} & \\ \gamma_{31} & \gamma_{32} & \gamma_{33} \\ \vdots & & \end{array}$$

The symbol δ_{rs} is the Kronecker delta: $\delta_{rs} = 1$ for $r = s$; $\delta_{rs} = 0$, otherwise.

can be found such that the elements

$$\begin{aligned} X_1 &= \gamma_{11} \varphi_1 \\ X_2 &= \gamma_{21} \varphi_1 + \gamma_{22} \varphi_2 \\ X_3 &= \gamma_{31} \varphi_1 + \gamma_{32} \varphi_2 + \gamma_{33} \varphi_3 \\ &\vdots \end{aligned}$$

are orthonormal.

The proof of this theorem provides an iterative algorithm, the Gram-Schmidt orthogonalization procedure, for actually determining the functions $X_m(t)$. The recursion is given by the following:

$$\begin{aligned} x_1 &= \varphi_1 & \text{and } X_1 &= x_1 / \|x_1\| \\ x_2 &= \varphi_2 - (\varphi_2, X_1) X_1 & \text{and } X_2 &= x_2 / \|x_2\| \\ &\vdots & &\vdots \\ &\vdots & &\vdots \\ x_{m+1} &= \varphi_{m+1} - \sum_{k=1}^m (\varphi_{m+1}, X_k) X_k & \text{and } X_{m+1} &= x_{m+1} / \|x_{m+1}\| \end{aligned} \quad (20)'$$

Although the functions $X_m(t)$ can be determined by this iterative procedure, Eqs. (20) provide only an implicit expression for the coefficients γ_{mn} generating $\{X_m(t)\}$. An additional drawback to the above scheme is that the recursive evaluation of the basis elements is exceedingly laborious.¹¹

¹¹The notation $\|f\|$ denotes $\left[\int_0^\infty f(t) f^*(t) dt \right]^{1/2}$, and (f, g) symbolizes $\int_0^\infty f(t) g^*(t) dt$. Thus, $\|f\|^2 = (f, f)$.

¹¹See Ref. 1, pp. 12-14.

To circumvent these difficulties, it is necessary to abandon the Gram-Schmidt approach and to consider obtaining the γ_{mn} from methods based on the set $\{\psi_n(\omega)\}$ derived from the Fourier transform of $\varphi_n(t)$. With the aid of Parseval's theorem and other key results from the theory of functions of a complex variable, it will be possible to achieve an explicit relation for the coefficients γ_{mn} , as well as for the transform parameters λ_{mn} in the equation

$$U_m(\omega) = \sum_{n=1}^m \lambda_{mn} \psi_n(\omega) \quad (21)$$

where

$$\int_{-\infty}^{\infty} U_r(\omega) U_s^*(\omega) d\omega = \delta_{rs} \quad (22)$$

Application of the Fourier transform, Eq. (16), to the elements $\varphi_n(t)$ and $\psi_n(\omega)$ of Eqs. (3) and (4) reveals that

$$\begin{aligned} \frac{1}{2\pi} \int_{-\infty}^{\infty} U_m(\omega) e^{j\omega t} d\omega &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \sum_{n=1}^m \lambda_{mn} \psi_n(\omega) e^{j\omega t} d\omega \\ &= \sum_{n=1}^m \frac{\lambda_{mn}}{2\pi} \int_{-\infty}^{\infty} \psi_n(\omega) e^{j\omega t} d\omega = \sum_{n=1}^m \lambda_{mn} \varphi_n(t) \equiv \hat{X}_m(t) \end{aligned} \quad (23)$$

where m is finite. Hence, $\hat{X}_m(t)$ and $U_m(\omega)$ are Fourier transform pairs. Since $U_m(\omega)$ is assumed to be orthonormal, Parseval's theorem leads to

$$\int_{-\infty}^{\infty} U_m(\omega) U_n^*(\omega) d\omega = 2\pi \int_{-\infty}^{\infty} \hat{X}_m(t) \hat{X}_n^*(t) dt = \delta_{mn} \quad (24)$$

Consequently, if the coefficients γ_{mn} in $X_m(t)$ are chosen as

$$\gamma_{mn} = \sqrt{2\pi} \lambda_{mn} \quad (25)$$

then the set $\{X_m(t)\}$ will also be orthonormal, for

$$\begin{aligned} \delta_{mn} &= 2\pi \int_{-\infty}^{\infty} \left[\sum_{n=1}^m \lambda_{mn} \varphi_n(t) \right] \left[\sum_{k=1}^n \lambda_{nk}^* \varphi_k^*(t) \right] dt \\ &= \int_{-\infty}^{\infty} \left[\sum_{n=1}^m \sqrt{2\pi} \lambda_{mn} \varphi_n(t) \right] \left[\sum_{k=1}^n \sqrt{2\pi} \lambda_{nk}^* \varphi_k^*(t) \right] dt \\ &= \int_{-\infty}^{\infty} \left[\sum_{n=1}^m \gamma_{mn} \varphi_n(t) \right] \left[\sum_{k=1}^n \gamma_{nk}^* \varphi_k^*(t) \right] dt \\ &= \int_{-\infty}^{\infty} X_m(t) X_n^*(t) dt \end{aligned} \quad (26)$$

Once the orthonormal elements $\{U_m(\omega)\}$ are found, therefore, the orthonormal set $\{X_m(t)\}$ is also uniquely determined.

The condition of orthonormality and the structure of $\psi_n(x)$ suggest a scheme for finding the parameters λ_{mn} of $U_m(x)$. The procedure is based on two results of analytic function theory, and is an extension of the application of the theorems found in Refs. 1 and 5.

By observing the pole patterns of the rationalized functions $U_m(\omega)$, $m = 1, 2, \dots$, it is possible to compute the coefficients λ_{mn} which determine the m^{th} orthonormal function $U_m(\omega)$. In $U_1(\omega)$, for example,

$$U_1(\omega) = \frac{\lambda_{11}}{\sigma + j(\omega - \beta)} \quad (27)$$

a pole is located at $jx = -\sigma + j\beta$. In $U_2(w)$, where

$$U_2(w) = \frac{\lambda_{21}}{\sigma + j(w - \beta)} + \frac{\lambda_{22}}{2\sigma + j(w - 2\beta)} \quad (28)$$

$$= \frac{\sigma(2\lambda_{21} + \lambda_{22}) + j[x(\lambda_{21} + \lambda_{22}) - \beta(2\lambda_{21} + \lambda_{22})]}{[\sigma + j(x - \beta)][2\sigma + j(x - 2\beta)]}$$

poles are located at $jw = -\sigma + j\beta$ and $jw = -2\sigma + j2\beta$. In general, for

$$U_m(w) = \frac{\lambda_{m1}}{\sigma + j(w - \beta)} + \frac{\lambda_{m2}}{2\sigma + j(w - 2\beta)} + \dots + \frac{\lambda_{mn}}{n\sigma + j(w - n\beta)} \quad (29)$$

equally spaced left-half s-plane (LHP) poles are located at $jw = -\sigma + j\beta$, $-2\sigma + j2\beta$, ..., $-n\sigma + jn\beta$. Similarly, the conjugate function $U_m^*(w)$ has equally spaced right-half s-plane (RHP) poles at $jw = \sigma + j\beta$, $2\sigma + j2\beta$, ..., $n\sigma + jn\beta$. Therefore, if $U_1(x)$ is expressed as

$$U_1(x) = \frac{a_1}{\sigma + j(x - \beta)}, \quad a_1 = \text{normalizing constant} \quad (30)$$

then to orthogonalize $U_1(x)$ and $U_2(w)$, i.e., to ensure that

$$\int_{-\infty}^{\infty} U_2(w) U_1^*(w) dx = 0 \quad (31)$$

$U_2(w)$ must be chosen as

$$U_2(w) = \frac{a_2 [\sigma - j(w - \beta)]}{[\sigma + j(w - \beta)][2\sigma + j(w - 2\beta)]}, \quad a_2 = \text{normalizing constant} \quad (32)$$

In this way, the zero of $U_2(w)$ cancels the RHP pole of $U_1^*(w)$ in the product $U_2 U_1^*$, and the integrand of Eq. (31) becomes analytic in the RHP. Since $U_2 U_1^*$ also satisfies Jordan's lemma,⁽⁶⁾ the Cauchy integral theorem⁽⁷⁾ can be applied to the integral in Eq. (31) (with the contour

of integration taken in the clockwise sense as the $j\omega$ -axis and infinite RHP semicircle) to establish that

$$\int_{-\infty}^{\infty} U_2(\omega) U_1^*(\omega) d\omega = \frac{-1}{j} \oint_{-j\infty}^{j\infty} \frac{a_1^* a_2 ds}{[s+\sigma-j\beta] [s+2\sigma-j2\beta]} = 0 \quad (33)$$

Similarly, if $U_3(\omega)$ is selected as

$$U_3(\omega) = \frac{a_3 [\sigma-j(\omega-\beta)] [2\sigma-j(\omega-2\beta)]}{[\sigma+j(\omega-\beta)] [2\sigma+j(\omega-2\beta)] [3\sigma+j(\omega-3\beta)]}, \quad a_3 = \text{normalizing constant} \quad (34)$$

then both $U_3 U_2^*$ and $U_3 U_1^*$ are analytic in the RHP, and consequently both equalities

$$\int_{-\infty}^{\infty} U_3(\omega) U_2^*(\omega) d\omega = \frac{-1}{j} \oint_{-j\infty}^{j\infty} \frac{a_2^* a_3 ds}{[s+2\sigma-j2\beta] [s+3\sigma-j3\beta]} = 0 \quad (35)$$

and

$$\int_{-\infty}^{\infty} U_3(\omega) U_1^*(\omega) d\omega = \frac{-1}{j} \oint_{-j\infty}^{j\infty} \frac{a_1^* a_3 [-s+2\sigma+j2\beta] ds}{[s+\sigma-j\beta] [s+2\sigma-j2\beta] [s+3\sigma-j3\beta]} = 0 \quad (36)$$

are guaranteed by Jordan's lemma and the Cauchy integral theorem.

Extending the above sequence to the m^{th} element of the orthogonal basis, it is observed that the general form of $U_m(\omega)$ must be

$$U_m(\omega) = \begin{cases} \frac{a_1}{\sigma+j(\omega-\beta)}, & m = 1 \\ \frac{a_m [\sigma-j(\omega-\beta)] [2\sigma-j(\omega-2\beta)] \cdots [(m-1)\sigma-j(\omega-m\beta+\beta)]}{[\sigma+j(\omega-\beta)] [2\sigma+j(\omega-2\beta)] \cdots [m\sigma+j(\omega-m\beta)]}, & m = 2, 3, \dots \end{cases} \quad (37)$$

where a_m is a normalizing constant.

The constants a_m are readily determined by requiring that the set $\{U_m(\omega)\}$ be orthonormal; i.e.,

$$\int_{-\infty}^{\infty} U_m(\omega) U_m^*(\omega) d\omega = 1 \quad (38)$$

Substituting Eq. (37) for $U_m(\omega)$ in the normality condition Eq. (38)

gives

$$\begin{aligned} \int_{-\infty}^{\infty} d(j\omega) & \left\{ \frac{|a_m|^2 [\sigma - j(\omega - \beta)] [2\sigma - j(\omega - 2\beta)] \dots [(m-1)\sigma - j(\omega - m\beta + \beta)]}{j [\sigma + j(\omega - \beta)] [2\sigma + j(\omega - 2\beta)] \dots [m\sigma + j(\omega - m\beta)]} \right. \\ & \left. \cdot \frac{[\sigma + j(\omega - \beta)] [2\sigma + j(\omega - 2\beta)] \dots [(m-1)\sigma + j(\omega - m\beta + \beta)]}{[\sigma - j(\omega - \beta)] [2\sigma - j(\omega - 2\beta)] \dots [m\sigma - j(\omega - m\beta)]} \right\} \\ & = \int_{-\infty}^{\infty} \frac{|a_m|^2 d(j\omega)}{j [m\sigma + j(\omega - m\beta)] [m\sigma - j(\omega - m\beta)]} = 2\pi j \left\{ \frac{|a_m|^2}{j} \cdot \frac{1}{m\sigma + j m\beta - j\omega} \right\} \Big|_{j\omega = -m\sigma + j m\beta} \\ & = \frac{\pi |a_m|^2}{m\sigma} = 1, \quad m = 1, 2, \dots \end{aligned} \quad (39)$$

Since the $U_m(\omega)$ are already orthogonal, the normality condition Eq. (38) can be satisfied by choosing the a_m real in Eq. (39) and by assigning to a_m the value

$$a_m = \sqrt{\frac{m\sigma}{\pi}}, \quad m = 1, 2, \dots \quad (40)$$

In view of Eq. (37), the orthonormal basis functions can be finally written as

$$U_m(\omega) = \begin{cases} \sqrt{\frac{\sigma}{\pi}} \frac{1}{\sigma + j(\omega - \beta)} & , \quad m = 1 \\ \sqrt{\frac{m\sigma}{\pi}} \frac{\prod_{n=1}^{m-1} [n\sigma - j(\omega - n\beta)]}{\prod_{n=1}^m [n\sigma + j(\omega - n\beta)]} & , \quad m = 2, 3, \dots \end{cases} \quad (41)$$

The coefficients λ_{mn} in the series representation of $U_m(\omega)$, Eq. (21), can now be related to Eq. (41) by

$$\begin{aligned} & \sqrt{\frac{m\sigma}{\pi}} \frac{[\sigma-j(\omega-\beta)][2\sigma-j(\omega-2\beta)] \cdots [(m-1)\sigma-j(\omega-m\beta+\beta)]}{[\sigma+j(\omega-\beta)][2\sigma+j(\omega-2\beta)] \cdots [m\sigma+j(\omega-m\beta)]} \\ &= \sum_{k=1}^m \frac{\lambda_{mk}}{k\sigma+j(\omega-k\beta)}, \quad m = 2, 3, \dots \end{aligned} \quad (42)$$

Multiplying both sides of Eq. (42) by $n\sigma+j(\omega-n\beta)$, $1 \leq n \leq m$, gives

$$\begin{aligned} & \sqrt{\frac{m\sigma}{\pi}} \frac{[n\sigma+j(\omega-n\beta)][\sigma-j(\omega-\beta)][2\sigma-j(\omega-2\beta)] \cdots [(m-1)\sigma-j(\omega-m\beta+\beta)]}{[\sigma+j(\omega-\beta)][2\sigma+j(\omega-2\beta)] \cdots [n\sigma+j(\omega-n\beta)] \cdots [m\sigma+j(\omega-m\beta)]} \\ &= \sum_{k=1}^m \lambda_{mk} \left[\frac{n\sigma+j(\omega-n\beta)}{k\sigma+j(\omega-k\beta)} \right], \quad m = 2, 3, \dots; \quad 1 \leq n \leq m \end{aligned} \quad (43)$$

Since Eq. (43) must be valid for all ω , it must also be an identity for $\omega = n\beta + jn\sigma$, or $j\omega = -n\sigma + jn\beta$. Accordingly, with $j\omega = -n\sigma + jn\beta$, Eq. (43) reduces to

$$\begin{aligned} & \sqrt{\frac{m\sigma}{\pi}} \frac{\prod_{r=1}^{m-1} [\sigma(n+r) - j\beta(n-r)]}{\prod_{r=1}^{n-1} [-\sigma(n-r) + j\beta(n-r)] \prod_{r=n+1}^m [-\sigma(n-r) + j\beta(n-r)]} = \lambda_{mn} \end{aligned} \quad (44)$$

so that

$$\lambda_{mn} = \begin{cases} \sqrt{\frac{\sigma}{\pi}} & n = m = 1 \\ \sqrt{\frac{m\sigma}{\pi}} \frac{\prod_{r=1}^{m-1} \left[n + r \frac{(\sigma + j\beta)}{(\sigma - j\beta)} \right]}{\prod_{\substack{r=1 \\ r \neq n}}^m [r-n]} & m = 2, 3, \dots; 1 \leq n \leq m \\ 0 & m = 1, 2, \dots; n > m \end{cases} \quad (45)^{\dagger}$$

In view of Eq. (25), γ_{mn} is immediately determined for the orthonormal set $\{X_m(t)\}$ as $\gamma_{mn} = \sqrt{2\pi} \lambda_{mn}$.

The preceding derivation of the orthonormal basis $\{U_m(\omega)\}$ permits a simple determination of the allied real-valued functions of ω

$$V_m(\omega) = \frac{1}{\sqrt{2}} [U_m(\omega) + U_m^*(\omega)] = \frac{1}{\sqrt{2}} \sum_{n=1}^m [\lambda_{mn} \psi_n(\omega) + \lambda_{mn}^* \psi_n^*(\omega)] \quad (46)$$

Assurance of orthonormality for the set $\{V_m(\omega)\}$ follows from the relations

$$\int_{-\infty}^{\infty} V_m(\omega) V_n^*(\omega) d\omega = \frac{1}{2} \int_{-\infty}^{\infty} [U_m + U_m^*] [U_n^* + U_n] d\omega \quad (47)$$

or

$$\begin{aligned} \int_{-\infty}^{\infty} V_m(\omega) V_n^* d\omega &= \frac{1}{2} \int_{-\infty}^{\infty} U_m U_n^* d\omega + \frac{1}{2} \int_{-\infty}^{\infty} U_m^* U_n d\omega + \frac{1}{2} \int_{-\infty}^{\infty} U_m U_n d\omega \\ &\quad + \frac{1}{2} \int_{-\infty}^{\infty} U_m^* U_n^* d\omega \end{aligned} \quad (48)$$

[†] For the functions $\hat{\psi}_n(\omega)$ defined in Eq. (2), the result for λ_{mn} with $\beta = 0$ agrees with the quantity $\sqrt{\frac{m\sigma}{\pi}} \alpha_{mn}$ derived in Ref. 1.

Since the U_m are orthonormal (Eqs. (33-38)) the first two integrals on the right side of Eq. (48) yield the Kronecker delta

$$\int_{-\infty}^{\infty} V_m V_n^* d\omega = \delta_{mn} + \int_{-\infty}^{\infty} U_m U_n d\omega + \frac{1}{2} \int_{-\infty}^{\infty} U_m^* U_n^* d\omega \quad (49)$$

It is evident from Eq. (37) for $U_m(\omega)$ that the integrand $U_m U_n$ in the first integral of Eq. (49) contains only LHP poles. Since Jordan's lemma[†] is satisfied for $U_m U_n$, the first integral of Eq. (49) may be evaluated over the infinite RHP semicircle with diameter on the $j\omega$ -axis.

Since $U_m U_n$ is also holomorphic in this RHP region, Cauchy's integral theorem^{††} guarantees that the integral of $U_m U_n$ vanishes. Similarly, with $U_m^* U_n^*$ holomorphic in the entire LHP, it can be argued that the second integral of Eq. (49) is also zero. Consequently, Eq. (49) becomes

$$\int_{-\infty}^{\infty} V_m(\omega) V_n^*(\omega) d\omega = \delta_{mn} \quad (50)$$

The rational function expression of V_m may be written with the aid of Eqs. (46), (21), and (4) as

$$V_m(\omega) = \frac{1}{\sqrt{2}} \sum_{n=1}^m \frac{(\lambda_{mn} + \lambda_{mn}^*) n\sigma + j(-\lambda_{mn} + \lambda_{mn}^*)(\omega - n\beta)}{(n\sigma)^2 + (\omega - n\beta)^2} \quad (51)$$

or

$$V_m(\omega) = \sqrt{2} \sum_{n=1}^m a_{mn} \frac{n\sigma}{(n\sigma)^2 + (\omega - n\beta)^2} + \sqrt{2} \sum_{n=1}^m b_{mn} \frac{\omega - n\beta}{(n\sigma)^2 + (\omega - n\beta)^2} \quad (52)^{†††}$$

[†] See Ref. 6.

^{††} See Ref. 7.

^{†††} For $\beta = 0$, λ_{mn} is real and $b_{mn} = 0$. When $\sqrt{2} a_{mn}$ is also an integer, $m, n = 1, 2, \dots$, $V_m(\omega)$ takes the form of the envelope-delay components examined in Ref. 1.

where a_{mn} and b_{mn} are related to λ_{mn} , Eq. (45), as

$$\lambda_{mn} \equiv a_{mn} + j b_{mn} \quad (53)$$

Thus, Eq. (52) can be rationalized to give

$$V_m(\omega) = \begin{cases} \sqrt{\frac{2\sigma}{\pi}} \frac{\sigma}{\sigma^2 + (\omega - \beta)^2}, & m = 1 \\ \sqrt{\frac{2m\sigma}{\pi}} \operatorname{Re} \left\{ \frac{[m(\sigma + j\beta) - j\omega] \prod_{n=1}^{m-1} [n(\sigma + j\beta) - j\omega]^2}{\prod_{n=1}^m [(n\sigma)^2 + (\omega - n\beta)^2]} \right\}, & m = 2, 3, \dots \end{cases} \quad (54)$$

Similarly, in the transformed space, it can be shown that the functions

$$Y_m(t) = \frac{1}{\sqrt{2}} [X_m(t) + X_m^*(-t)] = \sqrt{\pi} \left[\sum_{n=1}^m \lambda_{mn} \varphi_n(t) + \sum_{n=1}^m \lambda_{mn}^* \varphi_n^*(-t) \right] \quad (55)$$

comprise an orthonormal basis. Equation (16) indicates that $\varphi_n(t)$ and $\psi_n(\omega)$, as defined in Eqs. (3) and (4), are Fourier transform pairs. Thus,

$$\varphi_n(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \psi_n(\omega) e^{j\omega t} d\omega \quad (56)$$

and

$$\varphi_n^*(-t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \psi_n^*(\omega) e^{j\omega t} d\omega \quad (57)$$

Substituting these relations into the expression for $V_m(\omega)$,

$$V_m(\omega) = \frac{1}{\sqrt{2}} [U_m(\omega) + U_m^*(\omega)] = \frac{1}{\sqrt{2}} \left[\sum_{n=1}^m \lambda_{mn} \psi_n(\omega) + \sum_{n=1}^m \lambda_{mn}^* \psi_{mn}^*(\omega) \right] \quad (46)$$

it follows that the Fourier transform of $V_m(\omega)$ is

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} V_m(\omega) e^{j\omega t} d\omega = \frac{1}{\sqrt{2}} \left[\sum_{n=1}^m \lambda_{mn} \varphi_n(t) + \sum_{n=1}^m \lambda_{mn}^* \varphi_n^*(-t) \right] = \sqrt{\frac{1}{2\pi}} Y_m(t) \quad (58)$$

Consequently, Parseval's theorem and the orthonormality relation,

Eq. (50), for $V_m(\omega)$ allow Eq. (50) to be written as

$$\int_0^{\infty} \frac{Y_m(t)}{\sqrt{2\pi}} \frac{Y_n^*(t)}{\sqrt{2\pi}} dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} V_m(\omega) V_n^*(\omega) d\omega = \frac{1}{2\pi} \delta_{mn} \quad (59)$$

so that

$$\int_0^{\infty} Y_m(t) Y_n^*(t) dt = \delta_{mn} \quad (60)$$

From Eqs. (55), (53), (45), and (3), the orthonormal basis functions

$Y_m(t)$ can be finally expressed as the complex-valued functions of t

$$Y_m(t) = \begin{cases} \sqrt{\pi} \sum_{n=1}^m \lambda_{mn} e^{-n(\sigma-j\beta)t} & t > 0 \\ \sqrt{\pi} \sum_{n=1}^m \lambda_{mn}^* e^{+n(\sigma+j\beta)t} & t < 0 \end{cases} \quad m = 1, 2, \dots \quad (61)$$

or, expressing λ_{mn} in polar form,

$$Y_m(t) = \sqrt{\pi} \sum_{n=1}^m |\lambda_{mn}| e^{-n\sigma|t|} \left\{ \cos \left[n\beta t + \frac{t}{|t|} \arg(\lambda_{mn}) \right] + \frac{t}{|t|} j \sin \left[n\beta|t| + \arg(\lambda_{mn}) \right] \right\} \quad m = 1, 2, \dots \quad \forall t \quad (62)$$

The key relations established for the orthonormal bases $X_m(t)$ and $U_m(\omega)$ and the equations determining the orthonormal basis coefficients λ_{mn} are summarized in Table 1.

Table 1^a

EQUIVALENT REPRESENTATIONS OF $X_m(t)$, $U_m(\omega)$, and λ_{mn}

$X_m(t)$	$U_m(\omega)$	λ_{mn}
Range		
$0 < t < \infty \quad [X_m(t) = 0, \quad t \leq 0]$	$-\infty < \omega < \infty$	$m = 2, 3, \dots$ $n = 1, 2, \dots, m \quad \lambda_{11} = 1$
$\sigma > 0, \operatorname{Im}(\sigma) = 0, \operatorname{Im}(\beta) = 0$	$\sigma > 0, \operatorname{Im}(\sigma) = 0, \operatorname{Im}(\beta) = 0$	$\lambda_{mn} = 0 \quad n > m$
Representation		
$\sum_{n=1}^m \gamma_{mn} \varphi_n(t)$	$\sum_{n=1}^m \lambda_{mn} \psi_n(\omega)$	$\sqrt{\frac{m\sigma}{\pi}} \prod_{r=1}^{m-1} \frac{[n+r(\sigma+j\beta)]}{[r-n]}$
$\sum_{n=1}^m \gamma_{mn} e^{-n(\sigma-j\beta)t}$	$\sum_{n=1}^m \lambda_{mn} \frac{1}{n\sigma+j(\omega-n\beta)}$	$\sqrt{\frac{m\sigma}{\pi}} \prod_{r=1}^{m-1} \frac{[n(\sigma-j\beta) + r(\sigma+j\beta)]}{(-1)^{n+1} (n-1)! (m-n)! (\sigma-j\beta)^{m-1}}$
$\sum_{n=1}^m \sqrt{2\pi} \lambda_{mn} e^{-n(\sigma-j\beta)t}$	$\sum_{n=1}^m \sqrt{\frac{m\sigma}{\pi}} \alpha_{mn} \frac{1}{n\sigma+j(\omega-n\beta)}$	$\sqrt{\frac{m\sigma}{\pi}} \frac{(-1)^{m+n+1}}{r!} \sum_{k=0}^m (-1)^k \binom{m}{n} S_m^{(k)} \binom{k}{n} \frac{[\sigma+j\beta]^{m-k}}{[\sigma-j\beta]}$
$\sqrt{2\pi} \mathcal{F}^{-1} \{U_m(\omega)\}$	$\sqrt{\frac{1}{2\pi}} \mathcal{F} \{X_m(t)\}$	$\sqrt{\frac{m}{m-1}} \frac{n(\sigma-j\beta) + (m-1)(\sigma+j\beta)}{(m-n)(\sigma-j\beta)} \lambda_{m-1,n} \quad n < m$

^aThe last two relations in the λ_{mn} column are derived in succeeding sections. $\mathcal{F}$ denotes the Fourier transform operator, and $\mathcal{F}^{-1}$ denotes the inverse Fourier transform operator (see Eq. (23)). The quantity α_{mn} is defined by Eq. (68).

V. SPECIAL PROPERTIES OF THE BASIS COEFFICIENTS λ_{mn} AND
THE ORTHONORMAL ELEMENTS $X_m(t)$, $U_m(\omega)$

The preceding analysis culminates in Eq. (45) for the basis coefficients λ_{mn} which generate the orthonormal elements $U_m(\omega)$, $V_m(\omega)$, $X_m(t)$, and $Y_m(t)$. In order to detect errors in the computation of these λ_{mn} , it is desirable to provide a check-sum relation for the λ_{mn} analogous to the expression for the coefficients α_{mn} of Ref. 1.

It is shown in Ref. 1 that for $\beta = 0$ an identity

$$\sum_{n=1}^m \hat{\alpha}_{mn} = (-1)^{m+1} \quad m = 1, 2, \dots \quad (63)^{\dagger}$$

exists in which the α_{mn} are given by

$$\hat{\alpha}_{mn} = \begin{cases} (-1)^{n-1} \binom{m-1}{n-1} \binom{m+n-1}{m-1} & 1 \leq n \leq m = 1, 2, \dots \\ 0 & n > m \end{cases} \quad (64)^{\dagger\dagger}$$

The $\hat{\alpha}_{mn}$ are related to the orthonormal elements $R_m(t)$ or $W_m(\omega)$ by

$$R_m(t) = \sum_{n=1}^m \sqrt{2m\sigma} \hat{\alpha}_{mn} e^{-n\sigma t} \quad 0 < t < \infty \quad m = 1, 2, \dots \quad (65)^{\dagger\dagger\dagger}$$

and

$$W_m(\omega) = \sum_{n=1}^m \sqrt{\frac{m\sigma}{\pi}} \hat{\alpha}_{mn} \frac{1}{n\sigma + j\omega} \quad |\omega| < \infty \quad m = 1, 2, \dots \quad (66)$$

[†] See Ref. 1, p. 22.

^{††} See Ref. 1, p. 21.

^{†††} See Ref. 1, Table 1.

For the λ_{mn} derived in this Memorandum, it is similarly demonstrable that

$$\sqrt{\frac{\pi}{m\sigma}} \sum_{n=1}^m \lambda_{mn} = \sum_{n=1}^m \alpha_{mn} = (-1)^{m+1} \quad m = 1, 2, \dots \quad (67)$$

where α_{mn} is defined as

$$\alpha_{mn} \equiv \sqrt{\frac{\pi}{m\sigma}} \lambda_{mn} = \begin{cases} 1 & m = n = 1 \\ \frac{\prod_{r=1}^{m-1} \left[n + r \frac{(\sigma + j\beta)}{(\sigma - j\beta)} \right]}{\prod_{r=1}^{m'} [r - n]} & m = 2, 3, \dots; 1 \leq n \leq m \\ 0 & m = 1, 2, \dots; n > m \end{cases} \quad (68)$$

For the special case in which $\beta = 0$, it is clear that $\alpha_{mn} = \hat{\alpha}_{mn}$, and the validity of Eq. (67) follows directly from the identity Eq. (63).¹ In general, for arbitrary real β , it must be established that

$$\sum_{n=1}^m \alpha_{mn} = \sum_{n=1}^m \frac{\prod_{r=1}^{m-1} \left[n + r \frac{(\sigma + j\beta)}{(\sigma - j\beta)} \right]}{\prod_{r=1}^{m'} [r - n]} = (-1)^{m+1} \quad (69)$$

or that

$$\sum_{n=1}^m \frac{\prod_{r=1}^{m-1} (n + rz)}{\prod_{r=1}^{m'} (r - n)} = (-1)^{m+1} \quad (70)$$

¹The prime used in the product notation $\prod_{r=1}^{m'} (r - n)$ signifies that r ranges only over the integers $1, 2, \dots, n-1, n+1, \dots, m$ so that the result is never zero.

¹See Ref. 1, p. 20, Eq. (64).

where, for convenience, the complex variable z is defined as

$$z = \frac{\sigma + j\beta}{\sigma - j\beta} \quad (71)$$

From the definition of the Stirling numbers of the first kind, (5) it is possible to write the factorial function relation

$$(x-1)(x-2) \cdots [x - (n-1)] = \sum_{k=0}^n S_n^{(k)} x^{k-1} \quad (72)$$

or

$$(x-1)(x-2) \cdots [x + (n-1)] = \sum_{k=0}^n S_n^{(k)} x^{k-1} (-1)^{n-k} \quad (73)$$

Hence, the numerators of Eq. (70) can be written as

$$\prod_{r=1}^{n-1} (n + rz) = \prod_{r=1}^{n-1} z \left(\frac{n}{z} + r \right) = z^{n-1} \sum_{k=0}^n S_n^{(k)} \left(\frac{n}{z} \right)^{k-1} (-1)^{k+n} \quad (74)$$

and the left side of Eq. (70) becomes

$$\sum_{n=1}^{\infty} \frac{\prod_{r=1}^{n-1} (n + rz)}{\prod_{r=1}^{\infty} (r-n)} = \sum_{n=1}^{\infty} \frac{z^{n-1}}{\prod_{r=1}^{\infty} (r-n)} \sum_{k=0}^n S_n^{(k)} \left(\frac{n}{z} \right)^{k-1} (-1)^{k+n} = \sum_{n=1}^{\infty} a_{nn} \quad (75)$$

In addition, the denominators of Eq. (70) can be expressed as

$$\prod_{r=1}^{\infty} (r-n) = (-1)^{n-1} \prod_{r=1}^{n-1} (n-r) = (-1)^{2n-n-1} \frac{n! (n-n)!}{n}, \quad 1 \leq n < \infty \quad (76)$$

or

$$\prod_{r=1}^{\infty} (r-n) = (-1)^{n+1} \frac{n! (n-n)!}{n} = (-1)^{n+1} \frac{n!}{n \binom{n}{n}} \quad (77)$$

so that Eqs. (69), (70), and (75) can be formulated as

$$\sum_{n=1}^m \alpha_{mn} = \frac{(-1)^{m+1}}{m!} \sum_{n=1}^m \sum_{k=0}^m (-1)^{k+n} \binom{m}{n} S_m^{(k)} n^k z^{m-k} \quad (78)$$

From this last relation, it is clear that Eq. (69) can be established as an identity if it can be verified that

$$\frac{(-1)^{m+1}}{m!} \sum_{k=0}^m \sum_{n=1}^m (-1)^{k+n} \binom{m}{n} S_m^{(k)} n^k z^{m-k} = (-1)^{m+1} \quad (79)$$

or, equivalently, that

$$\sum_{k=0}^m \frac{(-1)^k S_m^{(k)}}{m!} \left\{ \sum_{n=1}^m (-1)^n \binom{m}{n} n^k \right\} z^{m-k} - 1 = 0 \quad m = 1, 2, \dots \quad (80)$$

Since Eq. (80) is a polynomial of degree m in z , it follows from the fundamental theorem of algebra⁽⁹⁾ that Eq. (80) has only m values of z which make the left side equal to zero. In order for Eq. (80) to be generally valid, therefore, each coefficient of z^r , $r = 1, 2, \dots, m$, must be zero and the coefficient of z^0 must be unity (since $S_m^{(k)}/m!$ is also nonzero for $k = 1, 2, \dots, m$; $m = 1, 2, \dots$). Consequently, the proof of Eq. (69) rests on demonstrating that

$$\frac{(-1)^{m-k} S_m^{(m-k)}}{m!} \sum_{n=1}^m (-1)^n \binom{m}{n} n^{m-k} = \begin{cases} 0 & k = 1, 2, \dots, m \\ 1 & k = 0 \end{cases} \quad (81)$$

or, since $S_m^{(0)} = 0$, $S_m^{(m)} = 1$, and $S_m^{(k)} \neq 0$, $k = 1, 2, \dots, m$, $m = 1, 2, \dots$, that

$$\sum_{n=1}^m (-1)^n \binom{m}{n} n^{m-k} = \sum_{n=0}^m (-1)^n \binom{m}{n} n^{m-k} = \begin{cases} 0 & k = 1, 2, \dots, m-1 \\ (-1)^m m! & k = 0 \end{cases} \quad (82)$$

In order to verify Eq. (82), the following identity is utilized:

$$\Delta^m f(x) = \sum_{n=0}^m (-1)^{m-n} \binom{m}{n} f(x+n) \quad (83)'$$

where Δ denotes the forward difference operator. With the choice

$$f(x) = x^{m-k} \quad k = 0, 1, \dots, m; \quad m = 1, 2, \dots \quad (84)$$

Eq. (83) becomes

$$\Delta^m x^{m-k} = \sum_{n=0}^m (-1)^{m-n} \binom{m}{n} (x+n)^{m-k} \quad (85)$$

It is possible, now, to conclude the demonstration by applying to Eq. (85) the fundamental theorem of difference calculus:⁽¹⁰⁾

The n^{th} difference of a polynomial of degree n

$$y(x) = \sum_{j=0}^n a_j x^j \quad a_n \neq 0$$

is a constant, $a_n n! h$, and the $(n+1)^{\text{th}}$ difference is equal to zero. The first forward difference is defined as $\Delta y(x) \equiv y(x+h) - y(x)$; the second forward difference, as $\Delta^2 y(x) = \Delta y(x+h) - \Delta y(x)$; etc., and h is a constant.

Accordingly, for $k = 1, 2, \dots, m-1$

$$\Delta^m x^{m-k} = 0 \quad k = 1, 2, \dots, m-1 \quad (86)$$

[†] See Ref. 8, p. 823.

so that Eq. (85) becomes

$$\sum_{n=0}^m (-1)^{m-n} \binom{m}{n} (x+n)^{m-k} = 0 \quad k = 1, 2, \dots, m-1 \quad (87)$$

For the particular choice $x = 0$, Eq. (87) becomes

$$\sum_{n=0}^m (-1)^n \binom{m}{n} n^{m-k} = 0 \quad k = 1, 2, \dots, m-1 \quad (88)$$

which verifies Eq. (82) or Eq. (81) for all $k = 1, 2, \dots, m$ except $k = 0$. With $k = 0$, $h = 1$, Eq. (85) may be simplified through the finite difference theorem to

$$\Delta^m x^m = m! = \sum_{n=0}^m (-1)^{m-n} \binom{m}{n} (x+n)^m \quad (89)$$

or, with $x = 0$

$$\sum_{n=0}^m (-1)^n \binom{m}{n} n^m = (-1)^m m! \quad (90)$$

Since this last relation is recognized as Eq. (82) with $k = 0$, Eq. (82) and, therefore, Eq. (69) are established as identities for $m = 1, 2, \dots$; $k = 0, 1, \dots, m$. From Eqs. (68) and (69), it is clear that the coefficients λ_{mn} satisfy the check-sum expression

$$\sum_{n=1}^m \lambda_{mn} = (-1)^{m+1} \sqrt{\frac{mG}{n}} \quad m = 1, 2, \dots \quad (91)$$

Two additional derivations simplify the calculation of the initial values of the basis elements $X_m(t)$ and $U_m(\omega)$, as well as the evaluation of the integrals of these functions. In order to get these relations, it is necessary to verify the identity

$$\left[\frac{\sigma - j\beta}{\sigma + j\beta} \right]^{m-1} \sum_{n=1}^m \frac{\alpha_{mn}}{n} = 1 \quad (92)$$

or, using the definition of z in Eq. (71), to show that

$$z^{(1-m)} \sum_{n=1}^m \frac{\alpha_{mn}}{n} = 1 \quad (93)$$

By referring to Eq. (78), it is seen that Eq. (93) can be written as

$$\frac{m(-1)^{m+1}}{z^{m-1} m!} \sum_{n=1}^m \sum_{k=0}^m (-1)^{k+n} \binom{m}{n} S_m^{(k)} n^{k-1} z^{m-k} = 1 \quad (94)$$

or, reordering the summation,

$$\sum_{k=0}^m S_m^{(k)} \frac{(-1)^{m+1+k}}{(m-1)!} \sum_{n=1}^m (-1)^n \binom{m}{n} n^{k-1} z^{m-k} = z^{m-1} \quad (95)$$

Since $S_m^{(0)} = 0$, it is necessary to show that

$$\sum_{k=1}^m S_m^{(k)} \frac{(-1)^{m+1+k}}{(m-1)!} \left\{ \sum_{n=1}^m (-1)^n \binom{m}{n} n^{k-1} \right\} z^{m-k} - z^{m-1} = 0 \quad (96)$$

or, applying again the fundamental theorem of algebra, that

$$S_m^{(1)} \frac{(-1)^m}{(m-1)!} \sum_{n=1}^m (-1)^n \binom{m}{n} = 1 \quad (97)$$

and, since $S_m^{(k)} \neq 0$, $k = 1, 2, \dots, m$, that

$$\sum_{n=1}^m (-1)^n \binom{m}{n} n^{k-1} = 0 \quad k = 2, 3, \dots, m \quad (98)$$

Equation (98) follows immediately from the finite difference relation given in Eq. (83):

$$\Delta^m f(x) = \sum_{n=0}^m (-1)^{m-n} \binom{m}{n} f(x+n)$$

If $f(x)$ is chosen as

$$f(x) = x^{k-1} \quad (99)$$

then Eq. (83) states that

$$\Delta^m x^{k-1} = \sum_{n=0}^m (-1)^{m-n} \binom{m}{n} (x+n)^{k-1} \quad (100)$$

The previously cited theorem of difference calculus provides the relation

$$\Delta^m x^{k-1} = 0 \quad k = 2, 3, \dots, m \quad (101)$$

Consequently, Eq. (100) becomes

$$\sum_{n=0}^m (-1)^{m-n} \binom{m}{n} (x+n)^{k-1} = 0 \quad k = 2, 3, \dots, m \quad (102)$$

With the choice $x = 0$, this simplifies to

$$\sum_{n=0}^m (-1)^n \binom{m}{n} n^{k-1} = 0 \quad k = 2, 3, \dots, m \quad (103)$$

or, since the first term of the summand is zero

$$\sum_{n=1}^m (-1)^n \binom{m}{n} n^{k-1} = 0 \quad k = 2, 3, \dots, m \quad (104)$$

This last equation is recognized as Eq. (98).

It remains to demonstrate the validity of Eq. (97). Since

$S_m^{(1)}$ is given by

$$S_m^{(1)} = (-1)^{m-1} (m-1)! \quad m = 1, 2, \dots \quad (105)^\dagger$$

Eq. (97) becomes

$$- \sum_{n=1}^m (-1)^n \binom{m}{n} = 1 \quad (106)$$

or, adding and subtracting the term for $n = 0$,

$$\sum_{n=0}^m (-1)^n \binom{m}{n} = 0 \quad m = 1, 2, \dots \quad (107)$$

But Eq. (107) follows directly from the binomial expansion⁽¹¹⁾

$$(1+x)^m = \sum_{n=0}^m \binom{m}{n} x^n \quad m = 1, 2, \dots \quad (108)$$

Thus, with $x = -1$, Eq. (108) yields

$$0 = \sum_{n=0}^m \binom{m}{n} (-1)^n \quad m = 1, 2, \dots \quad (109)$$

Since this is precisely Eq. (107), Eq. (97) is verified, and thereby, the identity Eq. (92). Finally, utilizing Eqs. (92) and (68), it follows that

$$\sqrt{\frac{mn}{\sigma}} \left[\frac{\sigma - j\beta}{\sigma + j\beta} \right]^{m-1} \sum_{n=1}^m \frac{\lambda_{mn}}{n} = 1 \quad (110)$$

The identities given by Eqs. (91) and (110) can be used to evaluate $X_m(0)$, $U_m(0)$, $\int_0^\infty X_m(t) dt$, and $\int_{-\infty}^\infty U_m(\omega) d\omega$. From Eqs. (18) and (25), $X_m(t)$ becomes

[†] See Ref. 8, p. 824.

$$X_m(t) = \sum_{n=1}^m \sqrt{2\pi} \lambda_{mn} \varphi_n(t) \quad m = 1, 2, \dots \quad (111)$$

This can also be expressed, by Eq. (3) for $\varphi_n(t)$, as

$$X_m(t) = \sqrt{2\pi} \sum_{n=1}^m \lambda_{mn} e^{-n(\sigma-j\beta)t} \quad m = 1, 2, \dots; \quad t > 0 \quad (112)$$

Hence, $X_m(0)$ becomes

$$X_m(0) = \sqrt{2\pi} \sum_{n=1}^m \lambda_{mn} = (-1)^{m+1} \sqrt{2m\sigma} \quad m = 1, 2, \dots \quad (113)$$

Similarly, using Eqs. (4) and (21), $U_m(\omega)$ can be written as

$$U_m(\omega) = \sum_{n=1}^m \lambda_{mn} \left[\frac{1}{n\sigma + j(\omega - n\beta)} \right] \quad m = 1, 2, \dots; \quad |\omega| < \infty \quad (114)$$

so that

$$U_m(0) = \frac{1}{\sigma - j\beta} \sum_{n=1}^m \frac{\lambda_{mn}}{n} \quad m = 1, 2, \dots \quad (115)$$

or, in view of Eq. (110)

$$U_m(0) = \sqrt{\frac{\sigma}{m\pi}} \frac{(\sigma + j\beta)^{m-1}}{(\sigma - j\beta)^m} \quad m = 1, 2, \dots \quad (116)$$

The areas under $U_m(\omega)$ and $X_m(t)$ can be ascertained in a similar fashion. From Eq. (23)

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} U_m(\omega) e^{j\omega t} d\omega = \hat{X}_m(t) = \sum_{n=1}^m \lambda_{mn} \varphi_n(t)$$

If $\varphi_n(t)$ is substituted in Eq. (23), it follows that

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} U_m(\omega) d\omega = \sum_{n=1}^m \lambda_{mn} \varphi_n(0) = \sum_{n=1}^m \lambda_{mn} \quad (117)$$

Consequently, replacing the sum of the λ_{mn} with the right side of Eq. (91), the area under $U_m(\omega)$ becomes

$$\int_{-\infty}^{\infty} U_m(\omega) d\omega = (-1)^{m+1} \sqrt{4\pi m \sigma} \quad m = 1, 2, \dots \quad (118)$$

Since Eq. (23) relates $U_m(\omega)$ and $\hat{X}_m(t)$ as Fourier transform pairs, it is evident that

$$U_m(\omega) = \int_0^{\infty} \hat{X}_m(t) e^{-j\omega t} dt = \frac{1}{\sqrt{2\pi}} \int_0^{\infty} X_m(t) e^{-j\omega t} dt \quad (119)$$

Using Eqs. (21) and (4) for $U_m(\omega)$ and $\psi_n(\omega)$, Eq. (119) becomes

$$\frac{1}{\sqrt{2\pi}} \int_0^{\infty} X_m(t) e^{-j\omega t} dt = \sum_{n=1}^m \lambda_{mn} \left[\frac{1}{n\sigma + j(\omega - n\beta)} \right] \quad (120)$$

Setting $\omega = 0$ in this last equation yields

$$\int_0^{\infty} X_m(t) dt = \frac{\sqrt{2\pi}}{\sigma - j\beta} \sum_{n=1}^m \frac{\lambda_{mn}}{n} \quad (121)$$

and, utilizing the identity Eq. (110)

$$\int_0^{\infty} X_m(t) dt = \sqrt{\frac{2\sigma}{m}} \frac{(\sigma + j\beta)^{m-1}}{(\sigma - j\beta)^m} \quad m = 1, 2, \dots \quad (122)$$

The important identities derived in this section are summarized in Table 2.

Table 2

SPECIAL PROPERTIES OF $X_m(t)$, $U_m(\omega)$ and λ_{mn}

$X_m(t)$	$U_m(\omega)$	λ_{mn}
Range		
$m = 1, 2, \dots$	$m = 1, 2, \dots$	$m = 1, 2, \dots$
$n = 1, 2, \dots, m$	$n = 1, 2, \dots, m$	$n = 1, 2, \dots, m$
Property		
$\int_0^m X_m(t) dt = \sqrt{\frac{2\sigma}{\omega}} \frac{(\sigma + j\beta)^{m-1}}{(\sigma - j\beta)^m}$	$\int_{-\infty}^{\infty} U_m(\omega) d\omega = (-1)^{m+1} \sqrt{4\pi m \sigma}$	$\sum_{n=1}^m \lambda_{mn} = (-1)^{m+1} \sqrt{\frac{m\sigma}{\pi}}$
$X_m(0) = (-1)^{m+1} \sqrt{2m\sigma}$	$U_m(0) = \sqrt{\frac{\sigma}{m\pi}} \frac{(\sigma + j\beta)^{m-1}}{(\sigma - j\beta)^m}$	$\sum_{n=1}^m \lambda_{mn} = \frac{(\sigma + j\beta)^{m-1}}{(\sigma - j\beta)^m} \sqrt{\frac{\sigma}{m\pi}}$
$\int_0^{\infty} X_m(t) X_n^*(t) dt = \delta_{mn}$	$\int_{-\infty}^{\infty} U_m(\omega) U_n^*(\omega) d\omega = \delta_{mn}$	$\lambda_{mn} = \sqrt{\frac{m\sigma}{\pi}} \frac{(-1)^{m+n+1}}{n!(m-n)!} \sum_{k=0}^m (-1)^k S_m^{(k)} \frac{(\sigma + j\beta)^{m-k}}{(\sigma - j\beta)^k}$
$\sqrt{2\pi} \mathcal{F}^{-1} \{U_m(\omega)\}$	$\sqrt{\frac{1}{2\pi}} \mathcal{F} \{X_m(t)\}$	$\lambda_{mn} = \sqrt{\frac{m\sigma}{\pi}} \frac{(-1)^{m+n+1}}{(\sigma - j\beta)^{m-1}} \sum_{l=1}^m \sum_{z=1}^{m-k+l} \sum_{p=1}^k Q_{mn}(k, r, p) \sigma^{(m-r-p+1)} \beta^{(r+p-2)}$

See Appendix E. $Q_{mn}(k, r, p) = \left[(-1)^{p+k+1} S_m^{(k)} \frac{k}{n} (m-k)!(k-1)! j^{(r+p)} \right] \div \left[(r-1)!(p-1)!(k-p)!(m-k-r+1)! \right]$.

VI. REPRESENTATIONS WITH $X_m(t)$ AND $U_m(\omega)$

The preceding sections have focused on generating the complete orthonormal exponential sums $X_m(t)$ and $Y_m(t)$ and the rational functions $U_m(\omega)$ and $V_m(\omega)$. Because the corresponding elements of these sets are related as Fourier transform pairs, it is a simple matter to determine simultaneously the least-mean-squared-error representations of prescribed functions $g(t) \in L^2(0, \infty)$ by $X_m(t)$ or $Y_m(t)$, and of functions $h(\omega) \in L^2(-\infty, \infty)$ by $U_m(\omega)$ or $V_m(\omega)$. The solution of the optimum expansion coefficients follows the usual treatment found in the literature on generalized Fourier analysis. Certain noteworthy simplifications evolve, however, because of the special nature of the orthonormal functions developed in this Memorandum.

When a specified function $g(x) \in L^2(0, \infty)$ is to be approximated in a least-integrated-squared-error sense by orthonormal elements $\{\theta_m(x)\}$ complete in $L^2(0, \infty)$, it is necessary to determine the coefficients a_m in the M^{th} partial sum

$$g(x) \approx \sum_{m=1}^M a_m \theta_m(x) \equiv \hat{g}(x) \quad x > 0 \quad (123)$$

so that the L^2 -norm

$$\|g(x) - \hat{g}(x)\|^2 \equiv \int_0^\infty |g(x) - \hat{g}(x)|^2 dx \quad (124)$$

is minimized. The necessary condition for achieving an extremum of Eq. (124) is that⁽¹²⁾

$$\frac{\partial \|g(x) - \hat{g}(x)\|^2}{\partial a_r} = 0 \quad r = 1, 2, \dots, M \quad (125)$$

or, substituting Eq. (123) for $g(x)$ in Eq. (125), that

$$\begin{aligned} \frac{\partial}{\partial a_r} \int_0^\infty \left[g(x) g^*(x) - g(x) \sum_{m=1}^M a_m^* \theta_m^*(x) - g^*(x) \sum_{m=1}^M a_m \theta_m(x) \right. \\ \left. + \sum_{m=1}^M \sum_{n=1}^M a_m a_n^* \theta_m(x) \theta_n^*(x) \right] dx = 0 \quad r = 1, 2, \dots, M \end{aligned} \quad (126)$$

Upon differentiating Eq. (126) and utilizing the orthonormality of the $\theta_m(x)$, Eq. (126) reduces to

$$- \int_0^\infty g(x) \theta_r^*(x) dx - \int_0^\infty g^*(x) \theta_r(x) dx + a_r^* + a_r = 0 \quad (127)$$

This last equation in a_r is obviously satisfied by

$$a_r = \int_0^\infty g(x) \theta_r^*(x) dx \quad r = 1, 2, \dots, M \quad (128)$$

The fact that this value of a_r leads to the desired minimization of the norm $\|g(x) - \hat{g}(x)\|^2$ can be seen from the sufficiency condition⁽¹³⁾

$$\frac{\partial \|g(x) - \hat{g}(x)\|^2}{\partial a_r} = 2 > 0 \quad r = 1, 2, \dots, M \quad (129)$$

For the particular set of functions $X_m(t)$ given by Eq. (18), it is clear from Eq. (128) that a least-squared-error representation $\hat{g}(t)$ of a function $g(t) \in L^2(0, \infty)$ of the form

$$g(t) \approx \sum_{m=1}^M a_m X_m(t) \equiv \hat{g}(t) \quad 0 < t < \infty \quad (130)$$

requires that the a_n be selected according to

$$a_n = \int_0^\infty g(t) X_n^*(t) dt = \int_0^\infty \sqrt{2\pi} \lambda_{nn}^* \int_0^\infty g(t) e^{-n(\sigma-j\beta)t} dt$$

$$n = 1, 2, \dots, M$$
(131)

The integrals determining these a_n are identifiable as the Laplace transform⁽¹⁴⁾ of $g(t)$ evaluated at the complex frequencies

$s_n = n(\sigma-j\beta)$, $n = 1, 2, \dots, M$, i.e.,

$$\int_0^\infty e^{-n(\sigma-j\beta)t} g(t) dt = \int_0^\infty e^{-s_n t} g(t) dt \Big|_{s=s_n} = \mathcal{L}[g(t)] \Big|_{s=s_n} = G(s_n)$$

$$n = 1, 2, \dots, M$$
(132)

Consequently, knowledge of the Laplace transform, $G(s_n)$, of $g(t)$ at the M complex frequencies $s_n = n(\sigma-j\beta)$, $n = 1, 2, \dots, M$, is sufficient to determine the expansion coefficients a_n of Eq. (131) and, thereby, the optimum integrated-squared-error approximation $\hat{g}(t)$ to the prescribed function $g(t)$ on $t > 0$. Alternatively, this algorithm for representing $g(t)$ in terms of $X_n(t)$ also provides a technique for obtaining an approximate numerical inverse $\hat{g}(t)$ to the prescribed Laplace transform, $G(s)$, of an unknown function $g(t)$.

Through Eqs. (130)-(132), the approximant $\hat{g}(t)$ of $g(t)$ can be expressed as

$$\hat{g}(t) = \sum_{m=1}^M \sum_{n=1}^m \sqrt{2\pi} \lambda_{mn}^* G(s_n) e^{-n(\sigma-j\beta)t} \quad t > 0$$
(133)

As a consequence of the Fourier transform relationship between corresponding elements $X_m(t)$ and $U_m(\omega)$, a least-integrated-squared-error representation $\hat{h}(\omega)$ to a specified function $h(\omega) \in L^2(-\infty, \infty)$ can also be readily computed. If $h(\omega)$ is the Fourier transform of $g(t)$ and if $\hat{h}(\omega)$ is defined as

$$\hat{h}(\omega) \equiv \sum_{m=1}^M b_m U_m(\omega) \approx h(\omega) \quad (134)$$

with the coefficients b_m selected to yield

$$\min_{\{b_m\}} \sum_{m=1}^M |h(\omega) - \hat{h}(\omega)|^2 \quad m = 1, 2, \dots, M \quad (135)$$

then, following the solution for the Fourier coefficients a_m , the b_m are determined as

$$b_m = \int_{-\infty}^{\infty} h(\omega) U_m^*(\omega) d\omega \quad m = 1, 2, \dots, M \quad (136)$$

But by Parseval's theorem and the previously established relation

$\mathcal{F}\{X_m(t)\} = \sqrt{2\pi} U_m(\omega)$, it follows that

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} h(\omega) U_m^*(\omega) d\omega = \frac{1}{\sqrt{2\pi}} \int_0^{\infty} g(t) X_m^*(t) dt \quad (137)$$

so that substitution of Eqs. (131) and (135) in Eq. (137) yields the numerical equivalence

$$b_m = \sqrt{2\pi} a_m \quad m = 1, 2, \dots, M \quad (138)$$

This result and another application of Parseval's theorem can provide a link between the approximation errors incurred in the

domains $\{X_n(t)\}$, $g(t) \in L^2(0, \infty)$, and $\{U_n(x)\}$, $h(x) \in L^2(-\infty, \infty)$. If ϵ denotes the error in the representation $\hat{g}(t) \sim g(t)$; i.e.,

$$\epsilon = \min_{\{a_n\}} \|g(t) - \hat{g}(t)\|^2 = \min_{\{a_n\}} \int_0^\infty |g(t) - \hat{g}(t)|^2 dt \quad (139)$$

then from Eq. (130) for $\hat{g}(t)$, ϵ can be written as

$$\begin{aligned} \epsilon = \int_0^\infty & \left[g(t) g^*(t) - g^*(t) \sum_{n=1}^M a_n X_n(t) - g(t) \sum_{n=1}^M a_n^* X_n^*(t) \right. \\ & \left. + \sum_{n=1}^M \sum_{m=1}^M a_n a_m^* X_n(t) X_m^*(t) \right] dt \end{aligned} \quad (140)$$

or, taking account of Eq. (131) for the optimum a_n and Eq. (26) for the orthonormality of the $X_n(t)$

$$\epsilon = \int_0^\infty |g(t)|^2 dt - \sum_{n=1}^M a_n a_n^* - \sum_{n=1}^M a_n^* a_n + \sum_{n=1}^M a_n a_n^* \quad (141)$$

so that

$$\epsilon = \int_0^\infty |g(t)|^2 dt - \sum_{n=1}^M |a_n|^2 \quad (142)$$

In a similar fashion, the error, $\mathcal{E}$, in the representation $\hat{h}(x)$ of $h(x) \in L^2(-\infty, \infty)$ can be derived from Eqs. (134)-(136) and (22) as

$$\mathcal{E} = \min_{\{b_m\}} \|h(x) - \hat{h}(x)\|^2 = \|h(x)\|^2 - \sum_{m=1}^M |b_m|^2 \quad (143)$$

From the Fourier transform relation of $g(t)$ to $h(\omega)$, Parseval's theorem guarantees that

$$g(t)^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} |h(\omega)|^2 d\omega \quad (144)$$

Consequently, ϵ and $\mathcal{E}$ are related as

$$\mathcal{E} = 2\pi \int_{-\infty}^{\infty} |g(t)|^2 dt = \sum_{n=1}^M 2\pi |a_n|^2 = 2\pi \epsilon \quad (145)$$

Since $|a_n|^2$ is nonnegative, Eq. (145) indicates that both ϵ and $\mathcal{E}$ are monotonically nonincreasing as M is increased. Therefore, for an arbitrarily prescribed approximation error ϵ or $\mathcal{E}$, M can be iteratively ascertained. Moreover, according to Eqs. (131) and (136) for a_n and b_n , M can be determined without recomputing any of the previously calculated coefficients a_n or b_n , $n = 1, 2, \dots, M-1$.

The key results of this section are recapitulated in Table 3.

In the table, the inner product notation $(x(\zeta), y(\zeta))$ denotes

$\int_{-r_1}^{r_2} x(\zeta) y^*(\zeta) d\zeta$, where the appropriate values of r_1 and r_2 are obvious from the context.

Table 3
REPRESENTATION RELATIONS

Property	Approximant		Relation
Form	$\hat{g}(t) = \sum_{m=1}^M a_m X_m(t)$	$\hat{h}(\omega) = \sum_{m=1}^M b_m U_m(\omega)$	$\hat{g}(t) = \sqrt{2\pi} \mathcal{F}^{-1} \{h(\omega)\}$
Admissible functions	$g(t) \in L^2(0, \infty)$	$h(\omega) \in L^2(-\infty, \infty)$	$g(t) = \mathcal{F}^{-1} \{h(\omega)\}$
Least-square coefficients	$a_m = (r(t), X_m(t))$	$b_m = (h(\omega), U_m(\omega))$	$a_m = b_m / \sqrt{2\pi}$
Least-square error	$\epsilon = \ g(t)\ ^2 = \sum_{m=1}^M a_m ^2$	$\epsilon = \ h(\omega)\ ^2 = \sum_{m=1}^M b_m ^2$	$\epsilon = \epsilon/2\pi$

VII. COMPUTATIONAL ASPECTS OF ORTHONORMAL EXPANSIONS IN $U_m(\omega)$ and $X_m(t)$

The computational labor in evaluating the coefficients λ_{mn} from Eq. (45) can be substantially reduced by employing a recursive procedure. In order to develop such a recursion scheme, the coefficients $\lambda_{m+1,n}$ must be expressed in terms of previously determined $\lambda_{m,n}$.[†] This can be accomplished by using Eq. (45).

With m replaced by $m+1$ in Eq. (45), the coefficient $\lambda_{m+1,n}$ becomes

$$\lambda_{m+1,n} = \begin{cases} \sqrt{\frac{(m+1)\sigma}{\pi}} \frac{\prod_{r=1}^m \left[n+r \left(\frac{\sigma+j\beta}{\sigma-j\beta} \right) \right]}{\prod_{r=1}^{m+1} [r-n]}, & \begin{matrix} m = 1, 2, \dots \\ 1 \leq n \leq m+1 \end{matrix} \\ 0 & \begin{matrix} m = 1, 2, \dots \\ n > m+1 \end{matrix} \end{cases} \quad (146)$$

Since λ_{mn} is also determined by Eq. (45), the ratio of $\lambda_{m+1,n}$ to λ_{mn} can be formed as

$$\frac{\lambda_{m+1,n}}{\lambda_{m,n}} = \begin{cases} \sqrt{2} (1+z) & \begin{matrix} m = 1 \\ n = 1 \end{matrix} \\ \sqrt{\frac{m+1}{m}} \frac{\prod_{r=1}^m (n+rz) \prod_{r=1}^m (r-n)}{\prod_{r=1}^{m+1} (r-n) \prod_{r=1}^{m-1} (n+rz)} & \begin{matrix} m = 2, 3, \dots \\ 1 \leq n \leq m \end{matrix} \end{cases} \quad (147)$$

[†]In order to avoid confusion, a comma is used to separate the subscripts m and n in the coefficients λ_{mn} . Accordingly, λ_{mn} is understood to signify $\lambda_{m,n}$, the n^{th} coefficient of the m^{th} basis function.

Consequently,

$$\lambda_{m+1,n} = \sqrt{\frac{m+1}{m}} \frac{n(\sigma-j\beta) + m(\sigma+j\beta)}{(m+1-n)(\sigma-j\beta)} \lambda_{m,n} \quad \begin{matrix} m = 1, 2, \dots \\ 1 \leq n \leq m \end{matrix} \quad (148)$$

Since λ_{mn} is zero for $n > m$, and since division by zero is invalid, the above expression for $\lambda_{m+1,n}$ in terms of the preceding $\lambda_{m,n}$ is limited to the range $1 \leq n \leq m$. In order to determine the remaining element $\lambda_{m+1,m+1}$, the identity given by Eq. (110) can be used in the following way

$$\sum_{n=1}^m \frac{\lambda_{mn}}{n} = \sqrt{\frac{\sigma}{m\pi}} \left[\frac{\sigma+j\beta}{\sigma-j\beta} \right]^{m-1} \quad m = 1, 2, \dots \quad (110)$$

With m replaced by $m+1$, Eq. (110) becomes

$$\sum_{n=1}^{m+1} \frac{\lambda_{m+1,n}}{n} = \sqrt{\frac{\sigma}{(m+1)\pi}} \left[\frac{\sigma+j\beta}{\sigma-j\beta} \right]^m \quad (149)$$

If the term $n = m+1$ is individually summed, Eq. (149) gives

$$\sum_{n=1}^m \frac{\lambda_{m+1,n}}{n} + \frac{1}{m+1} \lambda_{m+1,m+1} = \sqrt{\frac{\sigma}{(m+1)\pi}} \left[\frac{\sigma+j\beta}{\sigma-j\beta} \right]^m \quad (150)$$

so that

$$\lambda_{m+1,m+1} = \sqrt{\frac{(m+1)\sigma}{\pi}} \left[\frac{\sigma+j\beta}{\sigma-j\beta} \right]^m - \sum_{n=1}^m \frac{m+1}{n} \lambda_{m+1,n} \quad m = 1, 2, \dots \quad (151)$$

With this last relation and Eq. (148), it is evident that the coefficients $\lambda_{m,n}$ can be recursively evaluated by using the initial values given in Eq. (45), namely

$$\begin{aligned} \lambda_{11} &= \sqrt{\frac{\sigma}{\pi}} & \begin{matrix} m = 1 \\ n = 1 \end{matrix} \\ \lambda_{1n} &= 0 & n > 1 \end{aligned} \quad (152)$$

and by employing the expressions

$$\lambda_{m+1,n} = \begin{cases} \sqrt{\frac{m+1}{m}} \frac{n(\sigma-j\beta) + m(\sigma+j\beta)}{(m+1-n)(\sigma-j\beta)} \lambda_{m,n} & \begin{matrix} m = 1, 2, \dots \\ 1 \leq n \leq m \end{matrix} \\ \sqrt{\frac{(m+1)\sigma}{\pi}} \left[\frac{\sigma+j\beta}{\sigma-j\beta} \right]^m - (m+1) \sum_{k=1}^m \frac{\lambda_{m+1,k}}{k} & \begin{matrix} m = 1, 2, \dots \\ n = m+1 \end{matrix} \\ 0 & \begin{matrix} m = 1, 2, \dots \\ n > m+1 \end{matrix} \end{cases} \quad (153)$$

The check-sum relation Eq. (67) still applies and can be used to detect errors in the evaluation of each new row of basis coefficients $\lambda_{m+1,n}$ ($n = 1, 2, \dots, m+1$; $m = 1, 2, \dots$) generated from the previously computed $\lambda_{m,n}$.

It is also possible to provide a recursion formula for the efficient computation of the orthonormal elements $\{X_m(t)\}$ and $\{U_m(\omega)\}$. From Eqs. (3) and (18), $X_m(t)$ can be expressed as the sum of exponentials

$$X_m(t) = \sqrt{2\pi} \sum_{n=1}^m \lambda_{mn} e^{-n(\sigma-j\beta)t} \quad m = 1, 2, \dots \quad (154)$$

The fact that this sum involves integrally related decay factors can be exploited to evaluate $X_m(t)$ recursively for arbitrary values of $t > 0$. This is accomplished by first defining the associated quantities

$X_{m,r}$ as

$$X_{m,1} = \lambda_{mm} \quad r = 1 \quad (155)$$

with

$$\lambda_{m0} = 0 \quad (156)$$

and with λ_{mn} , $n \neq 0$, given by the iterative relation Eq. (153). With these definitions, it is clear that m iterations of the expression

$$X_{m,r+1} = e^{-(\sigma-j\beta)t} X_{m,r} + \lambda_{m,m-r} \quad \begin{matrix} r = 1, 2, \dots, m \\ m = 1, 2, \dots \end{matrix} \quad (157)$$

and final multiplication by $\sqrt{2\pi}$ yields the m^{th} basis function

$$X_m(t) = \sqrt{2\pi} X_{m,m+1}(t) \quad (158)$$

Thus, for arbitrary t , $X_m(t)$ can be calculated with only one evaluation of $e^{-(\sigma-j\beta)t}$, with $m+1$ multiplications, and with $m-1$ additions. This is an important economy in either manual or machine computation time as it results in only one exponential table look-up or one exponential subroutine entry.

By following the calculation of $X_m(t)$, $U_m(\omega)$ can also be iteratively obtained for any value of ω . From the rational function form, Eq. (41), of $U_m(\omega)$, it follows that

$$U_{m+1}(\omega) = \sqrt{\frac{(m+1)\sigma}{\pi}} \frac{\prod_{n=1}^m [n\sigma - j(\omega - n\beta)]}{\prod_{n=1}^{m+1} [n\sigma + j(\omega - n\beta)]} \quad m = 1, 2, \dots \quad (159)$$

Consequently,

$$U_{m+1}(\omega) = \sqrt{\frac{m+1}{m}} \frac{[m\sigma - j(\omega - m\beta)]}{[(m+1)\sigma + j(\omega - (m+1)\beta)]} U_m(\omega) \quad m = 1, 2, \dots \quad (160)$$

with

$$U_1(\omega) = \sqrt{\frac{\sigma}{\pi}} \frac{1}{\sigma + j(\omega - \beta)} \quad (161)$$

Once $U_1(\omega)$ is calculated for an arbitrary value of ω , then $U_2(\omega)$, $U_3(\omega)$, ..., follow by successive multiplications by $\sqrt{2(\sigma-j\omega+j\beta)/(2\sigma+j\omega-2j\beta)}$, $\sqrt{3/2} (2\sigma-j\omega+2j\beta)/(3\sigma+j\omega-3j\beta)$, and so forth.

It has already been noted that computation of the expansion coefficients for a least-integrated-squared-error representation of a prescribed $g(t)$ requires knowledge of the Laplace transform $G(s_n)$ of $g(t)$ at the M complex frequencies $s_n = n(\sigma+j\beta)$, $n = 1, 2, \dots, M$. Thus, evaluation of the coefficients a_m in Eq. (131) entails integrals of the type

$$L_n\{g(t)\} \equiv G(s_n) = \int_0^\infty g(t) e^{-n(\sigma+j\beta)t} dt \quad n = 1, 2, \dots, M \quad (162)$$

When the Laplace transform of $g(t)$ is not available, it can be approximately determined by a variety of quadrature schemes.⁽¹⁵⁾

If the linear functionals L_n are approximated by the quadrature formula

$$L_n \approx \sum_{k=1}^M w_k g(t_k) \quad n = 1, 2, \dots, M \quad (163)$$

then an appropriate criterion for obtaining the weights w_k , $k = 1, 2, \dots, M$ for assigned sample points $t_1, t_2, \dots, t_M$ is that L_n be exact for $g(t) = 1, e^{-(\sigma+j\beta)t}, e^{-2(\sigma+j\beta)t}, \dots, e^{-(M-1)(\sigma+j\beta)t}$. This criterion translates into the system of equations

$$\begin{aligned}
 \sum_{k=1}^M w_k &= \int_0^{\infty} e^{-n(\sigma+j\beta)t} dt = \frac{1}{n(\sigma+j\beta)} \\
 \sum_{k=1}^M w_k e^{-(\sigma+j\beta)t_k} &= \int_0^{\infty} e^{-(n+1)(\sigma+j\beta)t} dt = \frac{1}{(n+1)(\sigma+j\beta)} \\
 &\vdots \\
 \sum_{k=1}^M w_k e^{-(M-1)(\sigma+j\beta)t_k} &= \int_0^{\infty} e^{-(n+M-1)(\sigma+j\beta)t} dt = \frac{1}{(n+M-1)(\sigma+j\beta)}
 \end{aligned}
 \tag{164}$$

It is convenient to employ the following matrix notations in the above equations:

$$\underline{w} = \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_M \end{bmatrix} \tag{165}$$

$$\underline{v} = \begin{bmatrix} 1 & 1 & \dots & 1 \\ e^{-(\sigma+j\beta)t_1} & e^{-(\sigma+j\beta)t_2} & \dots & e^{-(\sigma+j\beta)t_M} \\ \vdots & \vdots & \vdots & \vdots \\ e^{-(M-1)(\sigma+j\beta)t_1} & e^{-(M-1)(\sigma+j\beta)t_2} & \dots & e^{-(M-1)(\sigma+j\beta)t_M} \end{bmatrix} \tag{166}$$

$$\underline{m} = \begin{bmatrix} \frac{1}{n(\sigma+j\beta)} \\ \frac{1}{(n+1)(\sigma+j\beta)} \\ \vdots \\ \frac{1}{(n+M-1)(\sigma+j\beta)} \end{bmatrix} \quad (167)$$

With these matrix definitions, the system of equations in Eq. (164) can be written as

$$\underline{V} \underline{w} = \underline{m} \quad (168)$$

and the symbolic solution for the weights becomes

$$\underline{w} = \underline{V}^{-1} \underline{m} \quad (169)$$

$\underline{V}$ is identifiable as the Vandermonde matrix and $\underline{V}^{-1}$ as its inverse.[†] Since σ must be nonzero and positive, and since the sample points t_i , $i = 1, 2, \dots, M$ are distinct, $\underline{V}$ is nonsingular^{††} and has an inverse, $\underline{V}^{-1}$. Once $\underline{V}^{-1}$ is evaluated, the solution for the quadrature weights can be obtained from Eq. (169) and the expansion coefficients a_m can be finally computed.

Numerous expressions are available for determining the elements of the inverse Vandermonde matrix.⁽¹⁵⁻¹⁸⁾ With a judicious choice

[†] See Ref. 10, p. 92.

^{††} See Ref. 10, p. 93.

of the sampling points, t_i , the inverse $\underline{V}^{-1}$ can be computed readily either from greatly simplified formulasⁱ or written directly from tables of Universal matrices.^{(19)††} An application of Universal matrices to a quadrature scheme similar to Eq. (163) can be found in Ref. 1.^{†††} Because of the direct analogy to the present situation (with σ replaced by $(\sigma + j\beta)$), the derivations will not be repeated here. By using the Universal matrices and the special sampling points t_i , it is possible to derive the weights $\underline{w}$ from a simple computation of the moment vector $\underline{m}$ and a premultiplication by the known matrix $\underline{V}^{-1}$. The expansion coefficients a_m then follow from Eqs. (163) and (131). Since a_m and b_m are related through Eq. (138), a similar development can be made for efficiently determining the coefficients b_m in the approximation $\hat{h}(\omega)$ to a prescribed function $h(\omega)$.

ⁱSee Ref. 16, pp. 96-98.

^{††} Universal matrices are the inverses of the Vandermonde matrices with sample points $t_i = i - n/2$, $i=0,1,\dots,n$ and with n equal to an integer.

^{†††} See Ref. 1, pp. 55-61.

VIII. SELECTION OF THE COMPLEX DECAY CONSTANT $(\sigma - j\beta)$

The orthonormal functions $X_m(t)$ and $U_m(\omega)$, the generating coefficients λ_{mn} and γ_{mn} , and the expansion coefficients a_m and b_m discussed in Section VII are dependent on σ and β . Except for the constraints that σ and β be real and that σ be greater than zero, the parameters σ and β have not yet been specified.

In attempting to formulate a criterion for selecting σ and β , several difficulties are immediately encountered. First, in data which arise from a sum of exponentially damped sinusoids of unknown decay and frequency constants, it may not be possible to obtain a unique complex decay value (or pole location) such that $-n(\sigma - j\beta)$, $n = 1, 2, \dots$, matches all the decay factors inherent in some prescribed data $g(t)$, or that matches all the poles comprised by a given $h(\omega)$. Second, there always exist functions $g(t)$ or $h(\omega)$ for which the optimum choice of σ and β in an M -term representation does not remain best as the approximation complexity is increased.

Another problem in solving for the complex decay constant $(\sigma - j\beta)$ occurs when the given data is impaired by noise or when the subsequent manipulation of the data is accompanied by round-off errors. The intractability of this classical problem is widely recognized and has been amply illustrated even in cases where four or fewer exponentials underlie the numerical data.⁽²⁰⁻³⁰⁾ Consequently, the present objective in solving for σ and β will not be to recover the original parameters imbedded in given data. Instead, an approximation to the

¹ See Ref. 15, pp. 272-288.

data will be sought which is best in an integral-square sense with respect to the Fourier coefficients a_m or b_m . Another more tractable criterion may be necessary to optimize the approximation with respect to σ and β .

Since σ and β are subject to only the two aforementioned constraints, their choice is essentially arbitrary and can be based on any criterion of optimality. One obvious criterion is the minimization over σ and β of the integral-square approximation error ϵ or $\hat{\epsilon}$ given by Eqs. (142) and (143). Although this objective would be consistent with the conditions leading to the expansion coefficients a_m and b_m , it unfortunately results in a nonlinear programming problem requiring iterative search procedures for its solution.[†]

A more tractable criterion than least-squares is one which requires matching the asymptotic approach to zero of the approximant and prescribed function for large t . More precisely, if $g(t)$ and its derivative $g'(t)$ exist for some $t \gg 1$ and are nonzero, both σ and β can be determined by requiring $g(t)$ and its approximant $\hat{g}(t)$ to have the same decay envelope for large t . This condition can be derived from Eqs. (130), (18), and (3), as follows.

Since

$$g(t) \approx \hat{g}(t) = \sum_{m=1}^M a_m X_m(t) = \sum_{m=1}^M a_m \sum_{n=1}^m \sqrt{2\pi} \lambda_{mn} e^{-n(\sigma-j\beta)t} \quad t > 0 \quad (170)$$

[†] See Ref. 1, p. 40.

the first derivative of $g(t)$ can be written as

$$g'(t) \approx \hat{g}'(t) = -\sqrt{2\pi} (\sigma - j\beta) \sum_{m=1}^M a_m \sum_{n=1}^m n \lambda_{mn} e^{-n(\sigma - j\beta)t} \quad t > 0 \quad (171)$$

For sufficiently large t and with $\sigma > 0$, it is clear that only the terms with the smallest decay factor, $-\sigma$, predominate in Eqs. (170) and (171). Consequently, Eqs. (170) and (171) can be simplified for $t \gg 1$ to

$$g(t) \approx \hat{g}(t) \approx \sqrt{2\pi} e^{-(\sigma - j\beta)t} \sum_{m=1}^M a_m \lambda_{m1} \quad t \gg 1 \quad (172)$$

and

$$g'(t) \approx \hat{g}'(t) \approx -\sqrt{2\pi} (\sigma - j\beta) e^{-(\sigma - j\beta)t} \sum_{m=1}^M a_m \lambda_{m1} \quad t \gg 1 \quad (173)$$

If the ratio of these last two equations is formed, then for $\sigma > 0$

$$-\frac{g'(t)}{g(t)} \bigg|_{t \gg 1} \approx -\frac{\hat{g}'(t)}{\hat{g}(t)} \bigg|_{t \gg 1} \approx \sigma - j\beta \quad g(t) \big|_{t \gg 1} \neq 0 \quad (174)$$

Hence,

$$\sigma \approx -\operatorname{Re} \left[\frac{g'(t)}{g(t)} \bigg|_{t \gg 1} \right] \quad g(t) \neq 0, \quad g'(t) \neq 0 \quad (175a)$$

and

$$\beta \approx \operatorname{Im} \left[\frac{g'(t)}{g(t)} \bigg|_{t \gg 1} \right] \quad (175b)$$

When the prescribed function $g(t)$ is real, the solution for β given by Eq. (175b) indicates that β can be set equal to zero. However,

this may be an unsatisfactory choice for β if $g(t)$ exhibits an oscillatory nature for some values of t . In such cases, β and σ can be resolved by representing $g(t)$ not as in Eq. (170), but as

$$g(t) \approx \hat{g}(t) = \sum_{m=1}^M c_m Y_m(t) = \frac{1}{\sqrt{2}} \sum_{m=1}^M c_m [X_m(t) + X_m^*(-t)] \quad (176)$$

where the orthonormal functions $Y_m(t)$ are defined by Eqs. (55), (61) and (62). In view of Eq. (62), $\hat{g}(t)$ becomes

$$g(t) \approx \sqrt{\pi} \sum_{m=1}^M c_m \sum_{n=1}^m |\lambda_{mn}| e^{-n\sigma|t|} \left\{ \cos \left[n\beta t + \frac{t}{|t|} \arg(\lambda_{mn}) \right] + \frac{t}{|t|} j \sin[n\beta|t| + \arg(\lambda_{mn})] \right\} \quad |t| < \infty \quad (177)$$

For $t \gg 1$, the terms for which $n = 1$ predominate, as before, so that

$$g(t) \approx \sqrt{\pi} e^{-\sigma t} \sum_{m=1}^M c_m |\lambda_{m1}| \left\{ \cos[\beta t + \arg(\lambda_{m1})] + j \sin[\beta t + \arg(\lambda_{m1})] \right\} \quad t \gg 1 \quad (178)$$

with

$$g'(t) \approx -\sigma g(t) + \sqrt{\pi} e^{-\sigma t} \sum_{m=1}^M c_m |\lambda_{m1}| \left\{ -\beta \sin[\beta t + \arg(\lambda_{m1})] + j\beta \cos[\beta t + \arg(\lambda_{m1})] \right\} \quad t \gg 1 \quad (179)$$

and

$$\begin{aligned}
 g''(t) \approx & -\sigma g'(t) + \sqrt{\pi} e^{-\sigma t} \sigma \beta \sum_{m=1}^M c_m |\lambda_{m1}| \left\{ \sin[\beta t + \arg(\lambda_{m1})] \right. \\
 & \left. - j \cos[\beta t + \arg(\lambda_{m1})] \right\} - \sqrt{\pi} e^{-\sigma t} \beta^2 \sum_{m=1}^M c_m |\lambda_{m1}| \left\{ \cos[\beta t + \arg(\lambda_{m1})] \right. \\
 & \left. + j \sin[\beta t + \arg(\lambda_{m1})] \right\} \quad t \gg 1 \quad (180)
 \end{aligned}$$

or

$$g''(t) \approx -2\sigma g'(t) - (\sigma^2 + \beta^2) g(t) \quad t \gg 1 \quad (181)$$

Similarly,

$$g'''(t) \approx -2\sigma g''(t) - (\sigma^2 + \beta^2) g'(t) \quad t \gg 1 \quad (182)$$

The preceding two equations relate the unknowns σ and β to the given function $g(t)$ and its derivatives at some large value of t . Equation (182) can be solved for β^2 to yield

$$\beta^2 \approx - \frac{g'''(t) + 2\sigma g''(t) + \sigma^2 g'(t)}{g'(t)} \bigg|_{t \gg 1} \quad g'(t) \big|_{t \gg 1} \neq 0 \quad (183)$$

When this expression for β^2 is substituted into Eq. (181), the following relation for σ is obtained

$$\sigma \approx \frac{g'''(t)g(t) - g'(t)g''(t)}{2[g'^2(t) - g(t)g''(t)]} \bigg|_{t \gg 1} \quad g'(t) \neq \pm \sqrt{g(t)g''(t)}, \quad t \gg 1 \quad (184)$$

Thus, β is also explicitly related to $g(t)$ and its derivatives for some large value of t as

$$\beta \approx \left\{ -\frac{g'''(t)}{g''(t)} - \frac{g''(t)}{g'(t)} \cdot \frac{g'''(t)g(t) - g'(t)g''(t)}{g'^2(t) - g(t)g''(t)} - \frac{g''^2(t)g^2(t) + g''^2(t)g'^2(t) - 2g(t)g'(t)g''(t)g'''(t)}{4[g'^4(t) + g^2(t)g''^2(t) - 2g'^2(t)g(t)g''(t)]} \right\}^{\frac{1}{2}} \quad (185)$$

$$g'(t) \neq 0, \quad g'(t) \neq \pm R(c)g''(t), \quad t \gg 1$$

Several qualifications are noteworthy regarding these equations for σ and β . It is clear that Eqs. (175), or Eqs. (184) and (185), provide suitable values of $(\sigma - j\beta)$ when the prescribed $g(t)$ is given as a long-time record. When $g(t)$ is a transient or pulse-type function, or when the complete time history of $g(t)$ is unknown, the asymptotic properties of $g(t)$ are not available for estimating σ and β by the aforementioned equations. For these pulsatile functions, it is necessary to resort to more elaborate techniques for obtaining $(\sigma - j\beta)$.

One procedure for obtaining σ and β relies on the discrete version of Prony's exponential approximation method.[†] In essence, an application of the Prony algorithm⁽³¹⁾ enables a selection of $(\sigma - j\beta)$ based on many samples of $g(t)$, rather than on a match of only the asymptotic values of $g(t)$ and its derivatives for large t . Other versions of Prony's scheme⁽³²⁻³³⁾ which are available can also be used to select σ and β . Depending on the procedure chosen, the approximant $\hat{g}(t)$ satisfies in a least-squares sense a finite difference or differential equation involving the sampled or continuous ordinates of the specified $g(t)$. Since all these approaches are conceptually similar, and since

[†]See Ref. 1, pp. 42-51.

the discrete version of Prony's method has been detailed in Ref. 1 for the case in which $\beta = 0$, the derivations will not be repeated here with $\beta \neq 0$. It will suffice to state that the derivations in Ref. 1 apply to the condition $\beta \neq 0$ considered in this Memorandum by merely substituting $(\sigma - j\beta)$ for σ in Eqs. (127)-(154) of Ref. 1.

Though no mention has yet been made of the associated problem of selecting σ and β for approximations of functions $h(\omega) \in L^2(-\infty, \infty)$ by sums of the orthonormal elements $U_m(\omega)$, it turns out that all the techniques discussed so far are applicable, albeit indirectly and with additional computational labor. Since the approximants $\hat{h}(\omega)$ and $\hat{g}(t)$ are Fourier transform pairs, the Fourier transform can be taken of a graphically or analytically specified $h(\omega)$ to produce a $g(t)$ or samples of $g(t)$ at S points $t = 0, 1, \dots, S-1$.[†] In turn, these ordinates of $g(t)$ can be utilized in all of the Prony schemes cited earlier, and a choice of $(\sigma - j\beta)$ can again be made based on $g(t)$'s satisfying a difference or differential equation.

[†] See Ref. 30, pp. 67-75.

IX. NUMERICAL EXAMPLES AND APPLICATIONS TO FILTER DESIGN

In the introduction, several important practical applications are enumerated for the orthonormal sets $X_m(t)$, $Y_m(t)$, $U_m(\omega)$, and $V_m(\omega)$. In order to clarify the use of the algorithms developed in this Memorandum, two filter design problems will be illustrated in this section.

The first example pertains to an approximation problem in network synthesis.⁽³⁴⁾ It requires finding a physically realizable transfer function for a finite, lumped-element, passive, linear network such that the network's impulse response is a replica of the waveform depicted in Fig. 1. The approximating response must be close enough to the prescribed response that the mean-square error between them is less than $\frac{1}{8} \times 10^{-2}$ over the time interval (0,10). In quantitative terms, the prescribed transient response is given as

$$g(t) = \begin{cases} 0 & t < 0 \\ 2t & 0 \leq t \leq 0.5 \\ -2(t-1) & 0.5 \leq t \leq 0.95 \\ e^{-2.42377t} & t > 0.95 \end{cases} \quad (186)$$

and it is necessary to find a_k and N for the approximant $\hat{g}(t)$ such that

$$\hat{g}(t) = \sum_{k=1}^N a_k X_k(t) = \sum_{k=1}^N a_k \sqrt{2\pi} \sum_{m=1}^k \lambda_{km} e^{-m(\sigma-j\beta)t} \approx g(t) \quad (187)$$

and

$$\epsilon_N = \frac{1}{10} \int_0^{10} |g(t) - \hat{g}(t)|^2 dt \leq \delta = \frac{1}{8} \times 10^{-2} \quad (188)$$

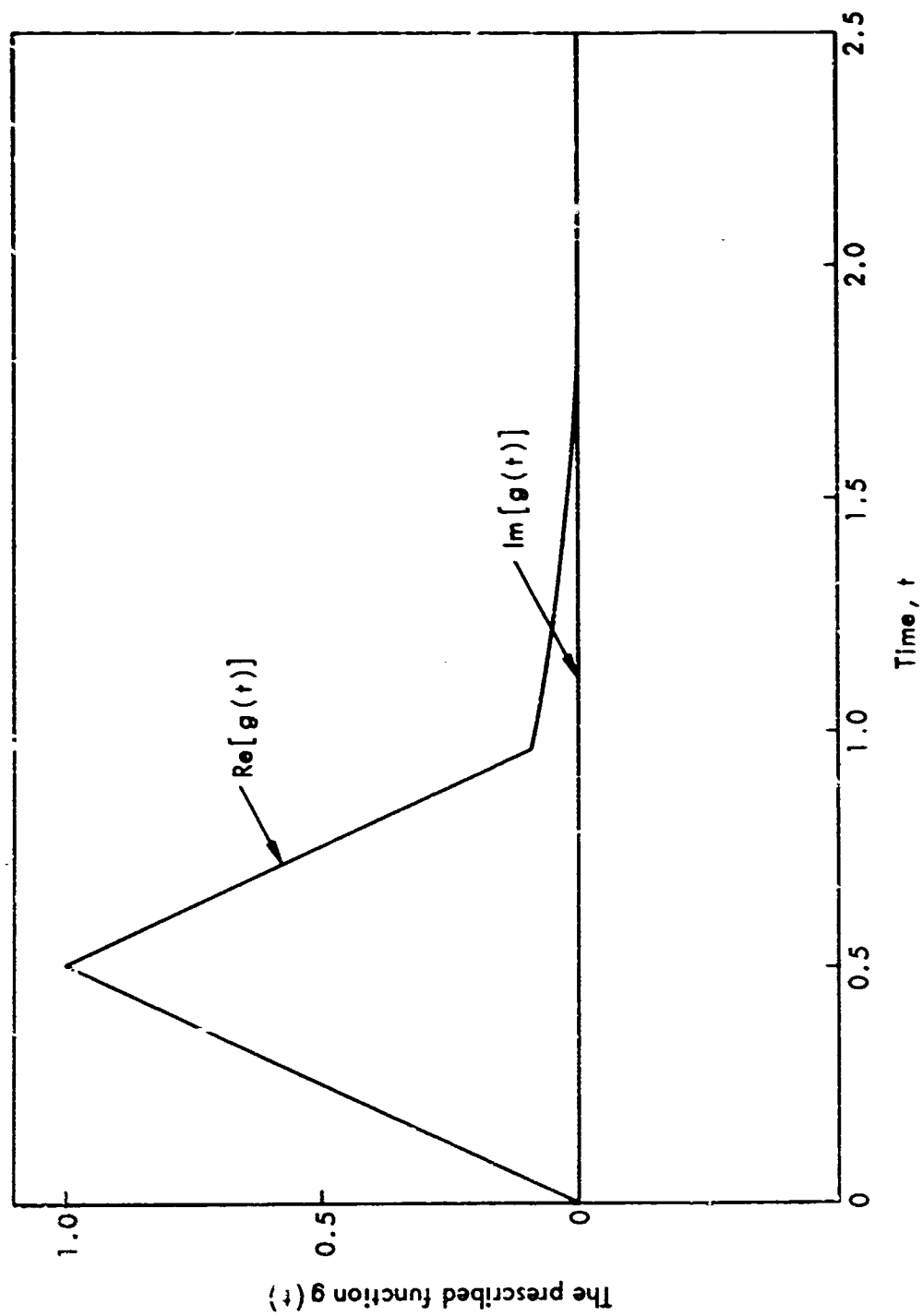


Fig. 1 — Graph of prescribed $g(t)$

In order to meet this design specification, the coefficients a_k , $k = 1, 2, \dots, N$ must be computed for some choice of σ , β , and N . In so doing it is convenient to recall that the expansion coefficients a_k are independent of N . Thus, N can be selected arbitrarily, ϵ_N can be evaluated and compared with δ , and N can be iteratively adjusted to be smaller or larger depending on whether $\epsilon_N < \delta$ or $\epsilon_N > \delta$. In either case, the a_k are determined from Eq. (131) as

$$a_k = \int_0^{\infty} g(t) X_m^*(t) dt = \sqrt{2\pi} \sum_{m=1}^k \lambda_{km}^* \int_0^{\infty} g(t) e^{-m(\sigma+j\beta)t} dt \quad (189)$$

$$k = 1, 2, \dots, N$$

Thus, evaluation of all the a_k entails integrating the N quantities

$$G_m = \int_0^{\infty} g(t) e^{-m(\sigma+j\beta)t} dt \quad m = 1, 2, \dots, N \quad (190)$$

Let

$$G_m = 2 \int_0^{0.5} t e^{-m(\sigma+j\beta)t} dt + 2 \int_{0.5}^{0.95} (1-t) e^{-m(\sigma+j\beta)t} dt \quad (191)$$

$$+ \int_{0.95}^{\infty} e^{-(m\sigma+mj\beta+2.42377)t} dt$$

$$= \frac{1}{m^2(\sigma+j\beta)^2} \left[-4e^{-0.5m(\sigma+j\beta)} + [2 - .1(\sigma+j\beta)m] e^{-0.95m(\sigma+j\beta)} + 2 \right] \quad (192)$$

$$m = 1, 2, \dots, N$$

In order to determine the moments G_m (the sampled Laplace transform of $g(t)$) by this last relation, a selection of σ and β is necessary. Since $g(t)$ is a pulse-like function with an exponential

tail and no oscillatory character, Eq. (175) is appropriate for determining σ and β . Accordingly

$$\sigma \approx -\operatorname{Re} \left[\frac{g'(t)}{g(t)} \mid t \gg 1 \right] = -\operatorname{Re} \left[\frac{-2.42377 e^{-2.42377 t}}{e^{-2.42377 t}} \mid t \gg 1 \right] \quad (193)$$

or

$$\sigma \approx 2.42377 > 0 \quad (194)$$

and

$$\beta \approx \operatorname{Im} \left[\frac{g'(t)}{g(t)} \mid t \gg 1 \right] = 0 \quad (195)$$

With these choices for σ and β , the first nine coefficients[†] a_k are found from Eqs. (192), (189), and the tabulated λ_{mn} of Appendix A as

$$\begin{aligned} a_1 &= 0.3738 + j 0. & a_5 &= -0.1227 + j 0. \\ a_2 &= 0.3957 + j 0. & a_6 &= -0.0412 + j 0. \\ a_3 &= 0.0694 + j 0. & a_7 &= 0.0284 + j 0. \\ a_4 &= -0.1164 + j 0. & a_8 &= 0.0441 + j 0. \\ a_9 &= 0.0227 + j 0. \end{aligned} \quad (196)$$

These values for a_k can be inserted in Eq. (187) along with the tabulated λ_{mn} to find $\hat{g}(t)$ for $t > 0$. When this is done, the rms error over the interval (0,10) can be computed from Eq. (188) to give

$$\epsilon_9 = 0.2 \times 10^{-2} = \delta \quad N = 9 \quad (197)$$

[†]The values of a_k listed in Eq. (196) have been rounded to four significant digits.

Since $\epsilon_8 = 0.3 \times 10^{-2} > \delta$, it is clear that the solution for $N = 9$, $\sigma = 2.42377$, $\beta = 0$, satisfies the design specification. For these values of N , σ , and β , graphs of $g(t)$ and $\hat{g}(t)$ can be compared (see Figs. 2 and 3).

Finally, since the Laplace transform of $X_m(t)$ is

$$\begin{aligned} \int_0^\infty X_m(t) e^{-st} dt &= \sum_{n=1}^m \sqrt{2\pi} \lambda_{mn} \int_0^\infty e^{(-n\sigma - s + jn\beta)t} dt \\ &= \sqrt{2\pi} \sum_{n=1}^m \lambda_{mn} \frac{1}{s + n(\sigma - j\beta)} \end{aligned} \quad \begin{cases} s = \alpha + j\beta \\ \alpha + n\sigma > 0 \\ n = 1, 2, \dots \end{cases} \quad (198a)$$

the approximating realizable network transfer function can be written as

$$\hat{H}(s) = \sqrt{2\pi} \sum_{m=1}^N \sum_{n=1}^m \lambda_{mn} \frac{a_m}{s + n(\sigma - j\beta)} \approx H(s) \equiv \int_0^\infty g(t) e^{-st} dt \quad (198b)$$

Thus, with σ equal to a real number and β equal to zero (as in the present example) $\hat{H}(s)$, the network approximant, can be implemented with either RC or RL elements to give the impulse response $\hat{g}(t) \approx g(t)$.

The second numerical example deals with a problem of optimal filtering. In processing signals impaired by additive noise, it is possible to accomplish smoothing and prediction by constructing an appropriate filter.⁽³⁵⁾ Usually this requires approximating the spectral densities of the signal and random noise process by rational functions. This procedure in turn permits the analytically optimum filter to be approximated in the form of a linear, lumped-constant network. Since the power spectral density is a rational function of frequency for a signal whose correlation function is a sum of

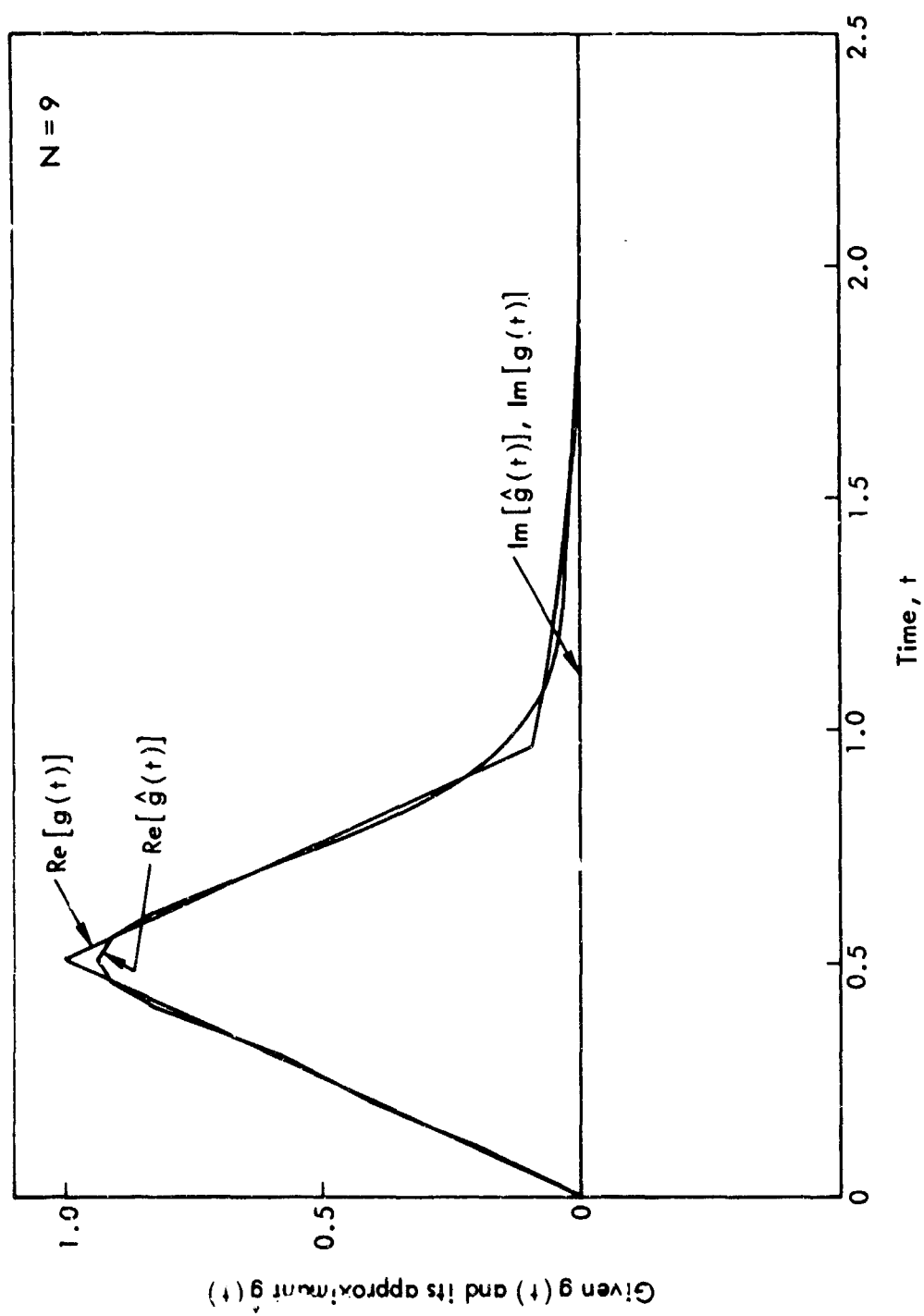


Fig.2 — Graphs of $g(t)$ and $\hat{g}(t)$ for $N = 9$

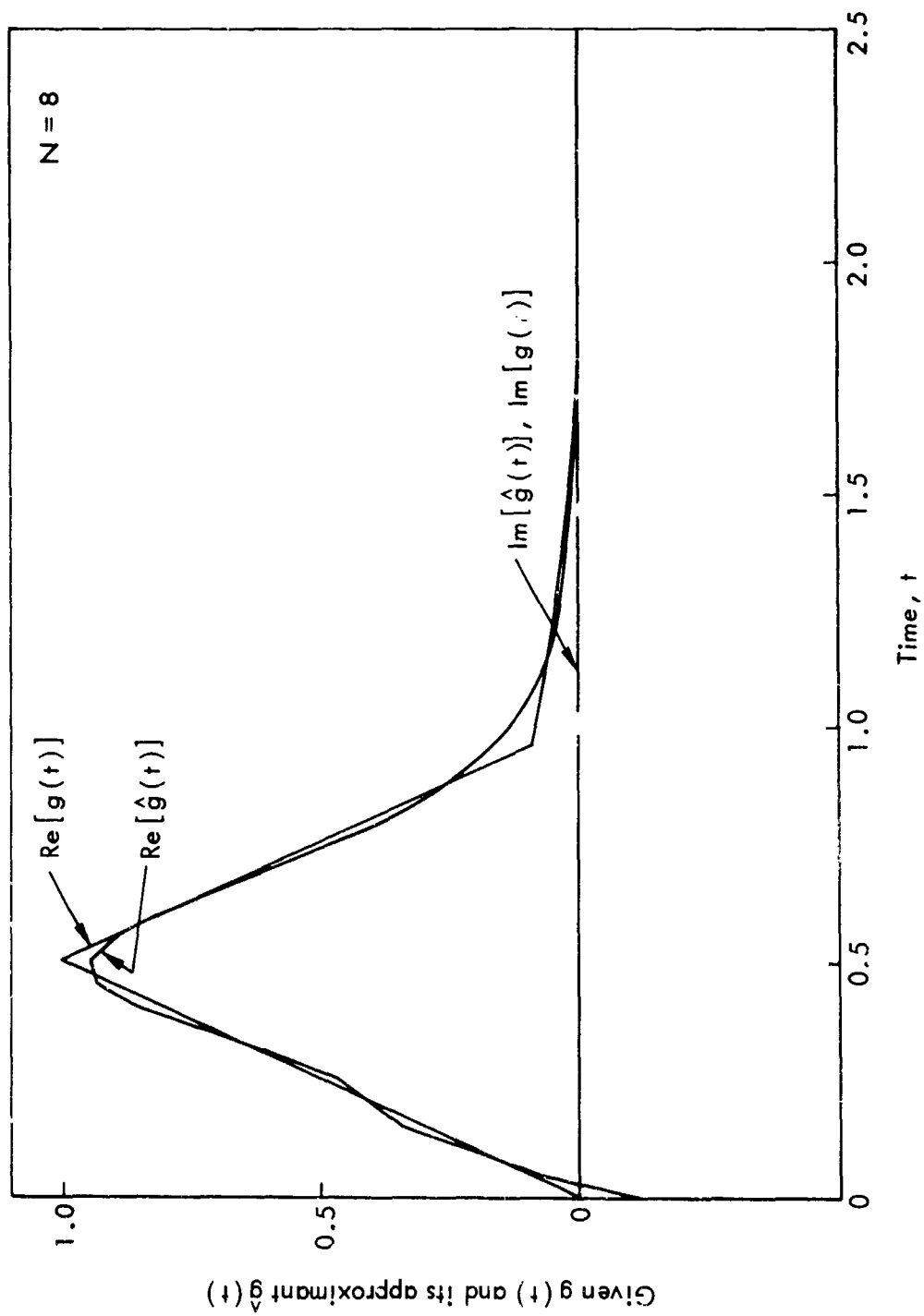


Fig.3 — Graphs of $g(t)$ and $\hat{g}(t)$ for $N = 8$

exponentials, it is possible to use advantageously the sets $\{X_m(t)\}$ and $\{Y_m(t)\}$ in the approximation of empirical correlation data.

A commonly observed autocorrelation function for a stationary random process is the exponentially damped cosine⁽³⁶⁾

$$\Psi(\tau) = e^{-2|\tau|} \cos \pi \tau \quad |\tau| > 0 \quad (199)$$

Though the best approximant of $\Psi(\tau)$ of exponential form is clearly $\Psi(\tau)$ itself, it is instructive to see how efficiently this exponentially damped cosinusoid can be approximated in an integral-square sense by the orthonormal elements $X_m(\tau)$. Since $\Psi(\tau)$ is defined over the doubly infinite interval, $|\tau| < \infty$, and since the $X_m(\tau)$ are non-zero only over $\tau > 0$, it is necessary to represent $\Psi(\tau)$ over $(-\infty, \infty)$ as

$$\Psi(\tau) \approx \hat{\Psi}_+(\tau) = \sum_{k=1}^N a_k X_k(\tau) \quad 0 \leq \tau < \infty \quad (200)$$

and

$$\Psi(\tau) \approx \hat{\Psi}_-(\tau) = \sum_{k=1}^N b_k X_k(\tau) \quad -\infty < \tau \leq 0 \quad (201)$$

It is evident from Eq. (199) that $\Psi(\tau)$ is an even function of τ ; i.e.,

$$\Psi(\tau) = \Psi(-\tau) \quad (202)$$

Hence

$$\hat{\Psi}_+(\tau) = \hat{\Psi}_-(-\tau) \quad (203)$$

and

$$a_k = b_k \quad k = 1, 2, \dots, N \quad (204)$$

Consequently, the moments G_m for $\Psi(\tau)$ can be related simultaneously to a_k and b_k through Eqs. (189), (190), and (204) as

$$G_m = \int_0^\infty e^{-(m\sigma + jm\beta + 2)\tau} \cos \pi\tau \, d\tau \quad (205)$$

$$= \frac{m(\sigma + j\beta) + 2}{[m(\sigma + j\beta) + 2]^2 + \pi^2} \quad m = 1, 2, \dots, N$$

In terms of these G_m , the expansion coefficients for $\hat{\Psi}_+(\tau)$ and $\hat{\Psi}_-(\tau)$ become

$$a_k = \sqrt{2\pi} \sum_{m=1}^k \lambda_{km}^* G_m = b_k \quad k = 1, 2, \dots, N \quad (206)$$

where the λ_{km}^* are tabulated in Appendix A as rational functions of σ and β .

In order to complete the approximation $\hat{\Psi}(\tau) \approx \Psi(\tau)$, a selection of the parameters σ and β is necessary. This will enable the evaluation of G_m , a_k , and b_k of Eqs. (205) and (206) and, thereby, of $\hat{\Psi}_+(\tau)$ and $\hat{\Psi}_-(\tau)$ in Eqs. (200) and (201). Since $\Psi(\tau)$ is given as a real, oscillatory, exponentially decaying function, Eqs. (183)-(185) can be used to obtain σ and β , with β not necessarily zero (as would be the case if Eq. (175) were used). Accordingly, the first three derivatives of $\Psi(\tau)$ for $\tau > 0$ become

$$\Psi'(\tau) = -e^{-2\tau} [2 \cos \pi\tau + \pi \sin \pi\tau] \quad (207)$$

$$\Psi''(\tau) = e^{-2\tau} [(4 - \pi^2) \cos \pi\tau + 4\pi \sin \pi\tau] \quad (208)$$

$$\Psi'''(\tau) = e^{-2\tau} [(-8 + 6\pi^2) \cos \pi\tau + (\pi^2 - 12) \pi \sin \pi\tau] \quad (209)$$

Putting these values in Eq. (184) for σ results in

$$\sigma \approx \frac{16 \pi^2 e^{-4\tau}}{8 \pi^2 e^{-4\tau}} \bigg|_{\tau \gg 1} = 2 \quad (210)$$

Similarly, after simplifying the trigonometric terms generated from substituting Eqs. (207)-(210) in Eq. (184) for β^2 , β^2 becomes

$$\begin{aligned} \beta^2 \approx & \frac{-2(4-3\pi^2) e^{-2\tau} \cos \pi\tau + (\pi^2-12) \pi e^{-2\tau} \sin \pi\tau + \dots}{-2 e^{-2\tau} \cos \pi\tau - \pi e^{-2\tau} \sin \pi\tau} \\ & \frac{4(4-\pi^2) e^{-2\tau} \cos \pi\tau + 16 \pi e^{-2\tau} \sin \pi\tau + \dots}{3 e^{-2\tau} \cos \pi\tau - 4\pi e^{-2\tau} \sin \pi\tau} \bigg|_{\tau \gg 1} \end{aligned} \quad (211)$$

or

$$\beta^2 \approx \frac{2\pi^2 e^{-2\tau} (\cos \pi\tau + \pi \sin \pi\tau)}{2 e^{-2\tau} (\cos \pi\tau + \pi \sin \pi\tau)} \bigg|_{\tau \gg 1} = \pi^2 \quad (212)$$

Consequently, the principal root of Eq. (212) gives

$$\beta \approx \pi \quad (213)$$

For $\Psi(\tau)$ prescribed in Eq. (199), the values $\sigma \approx 2$ and $\beta \approx \pi$ just derived have special intuitive appeal. When these values are used, and when a mean-square-error tolerance, Eq. (188), of $\delta = 0.9 \times 10^{-2}$ is selected, it is necessary to compute twelve expansion coefficients. Thus, with $N = 12$, $\sigma = 2$, and $\beta = \pi$, the a_k of Eq. (189) become

$$a_k = \sqrt{2\pi} \sum_{m=1}^k \lambda_{km}^* G_m = b_k \quad k = 1, 2, \dots, N \quad (214)$$

where now

$$G_m = \int_0^{\infty} \Psi(\tau) e^{-m(\sigma+j\beta)\tau} d\tau = \int_0^{\infty} e^{-[m(\sigma+j\beta)+2]\tau} \cos \pi\tau d\tau \quad (215)$$

or

$$G_m = \frac{m(\sigma+j\beta) + 2}{[m(\sigma+j\beta)+2]^2 + \pi^2} \quad m = 1, 2, \dots, N \quad (216)$$

Thus, the expansion coefficients, rounded to four significant digits,[†] are

$$\begin{aligned} a_1 &= 0.3221 - j 0.1132 & a_5 &= -0.0316 + j 0.0488 & a_9 &= 0.0347 - j 0.0048 \\ a_2 &= -0.0967 + j 0.0452 & a_6 &= 0.0500 - j 0.0014 & a_{10} &= -0.0197 - j 0.0249 \\ a_3 &= 0.0798 + j 0.0281 & a_7 &= -0.0249 - j 0.0361 & a_{11} &= -0.0107 + j 0.0270 \\ a_4 &= -0.0242 - j 0.0648 & a_8 &= -0.0154 + j 0.0358 & a_{12} &= 0.0265 - j 0.0037 \end{aligned} \quad (217)$$

with

$$\epsilon_{12} = 0.895 \times 10^{-2} < \delta \quad (218)$$

and

$$\epsilon_{11} = 0.934 \times 10^{-2} > \delta \quad (219)$$

The corresponding approximation $\hat{\Psi}(\tau)$ and the given autocorrelation function $\Psi(\tau)$ are compared in Figs. 4-7.

Finally, since the Wiener-Khintchine Theorem⁽³⁷⁾ relates the autocorrelation function and spectral density as Fourier transform pairs, the power density spectrum associated with the approximation

$$\hat{\Psi}_+(\tau) = \sum_{k=1}^{12} a_k X_k(\tau) \approx \Psi(\tau) \quad \tau \geq 0 \quad (220)$$

[†] Sixteen significant digits were computed in all the calculations of a_k since the λ_{mn} may become large for certain values of σ , β , and m .

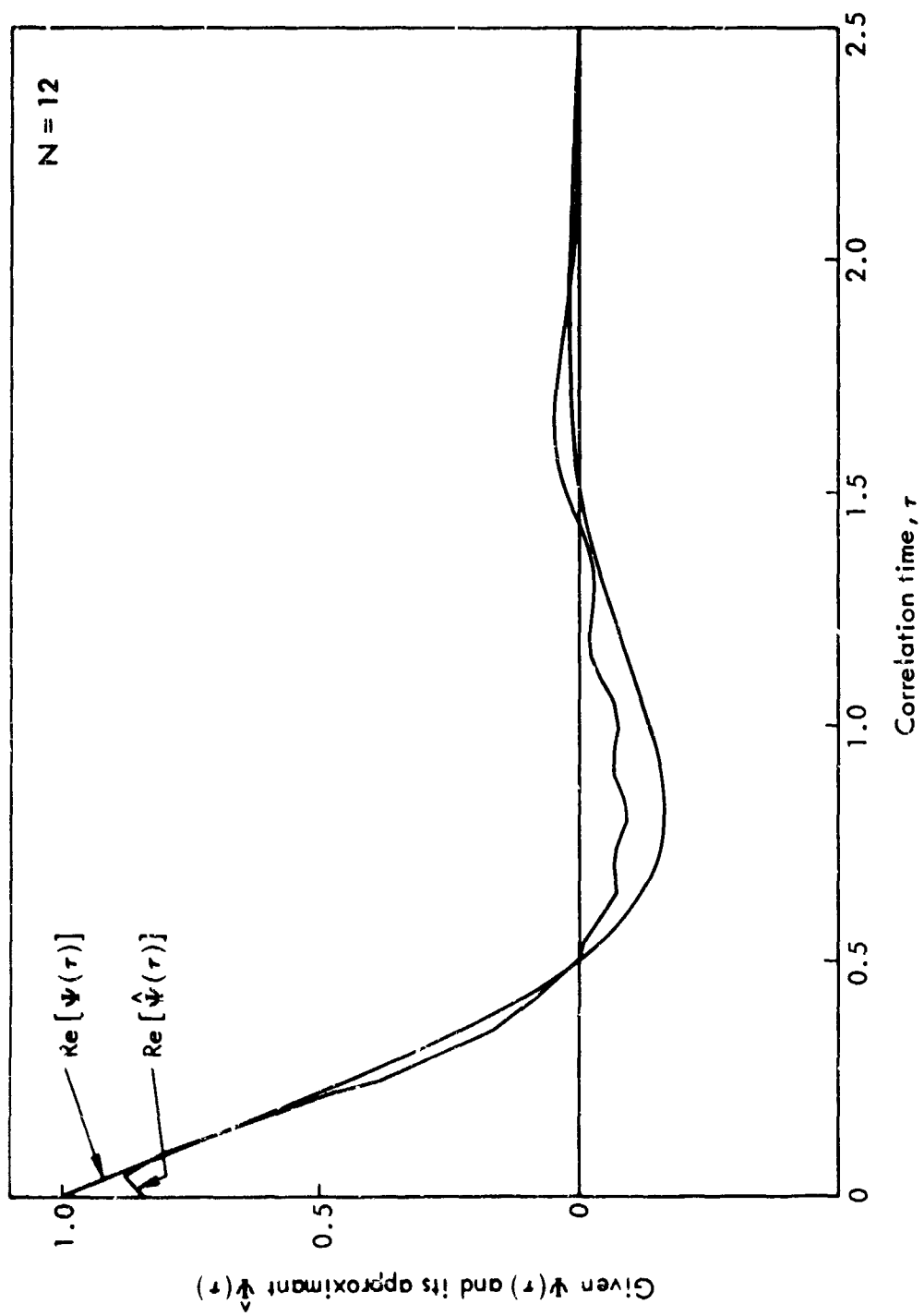


Fig. 4 — Graphs of the real parts of $\Psi(\tau)$ and $\hat{\Psi}(\tau)$ for $N = 12$

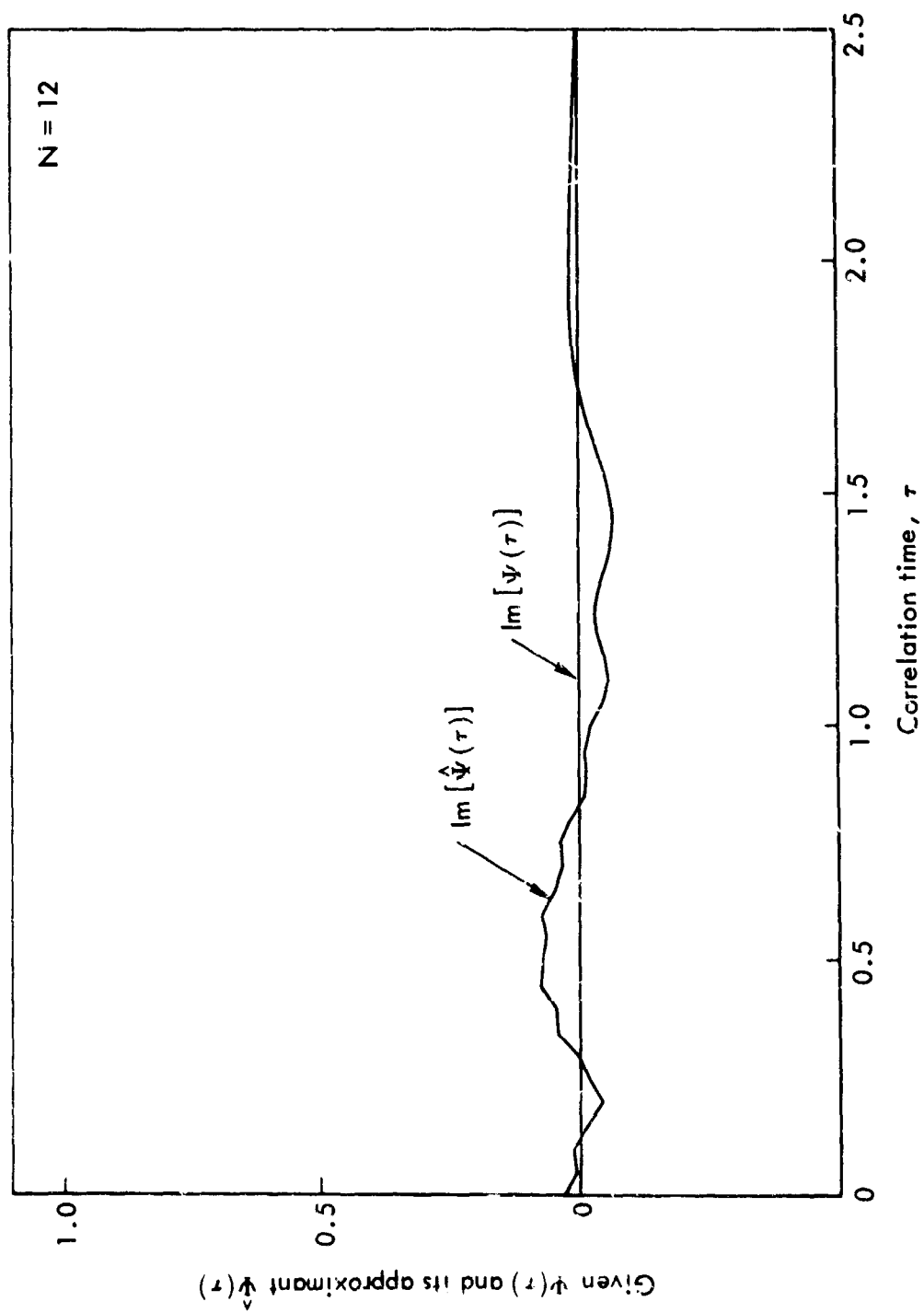


Fig. 5--- Graphs of the imaginary parts of $\Psi(\tau)$ and $\hat{\Psi}(\tau)$ for $N = 12$

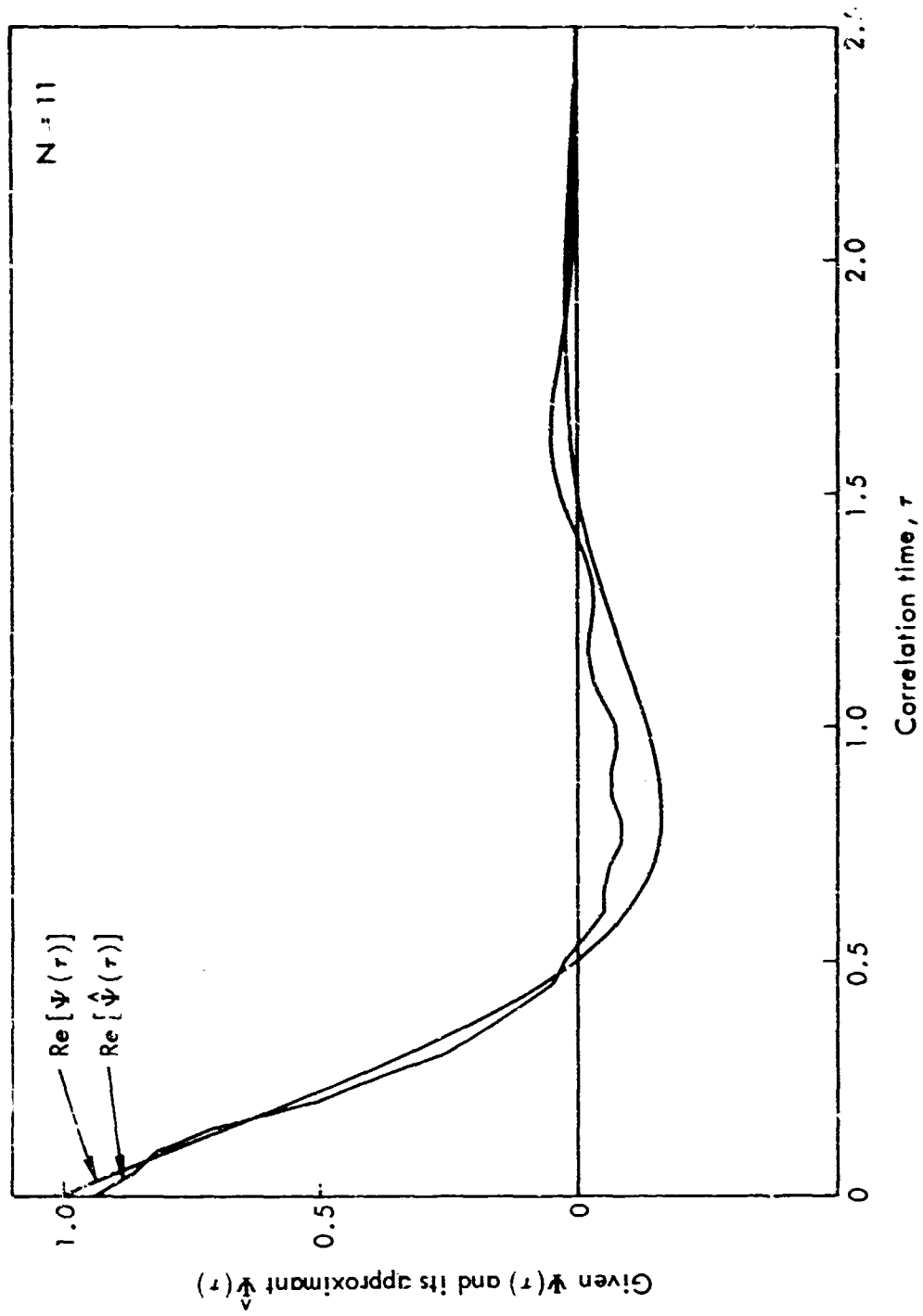


Fig. 6— Graphs of the real parts of $\Psi(\tau)$ and $\hat{\Psi}(\tau)$ for $N = 11$

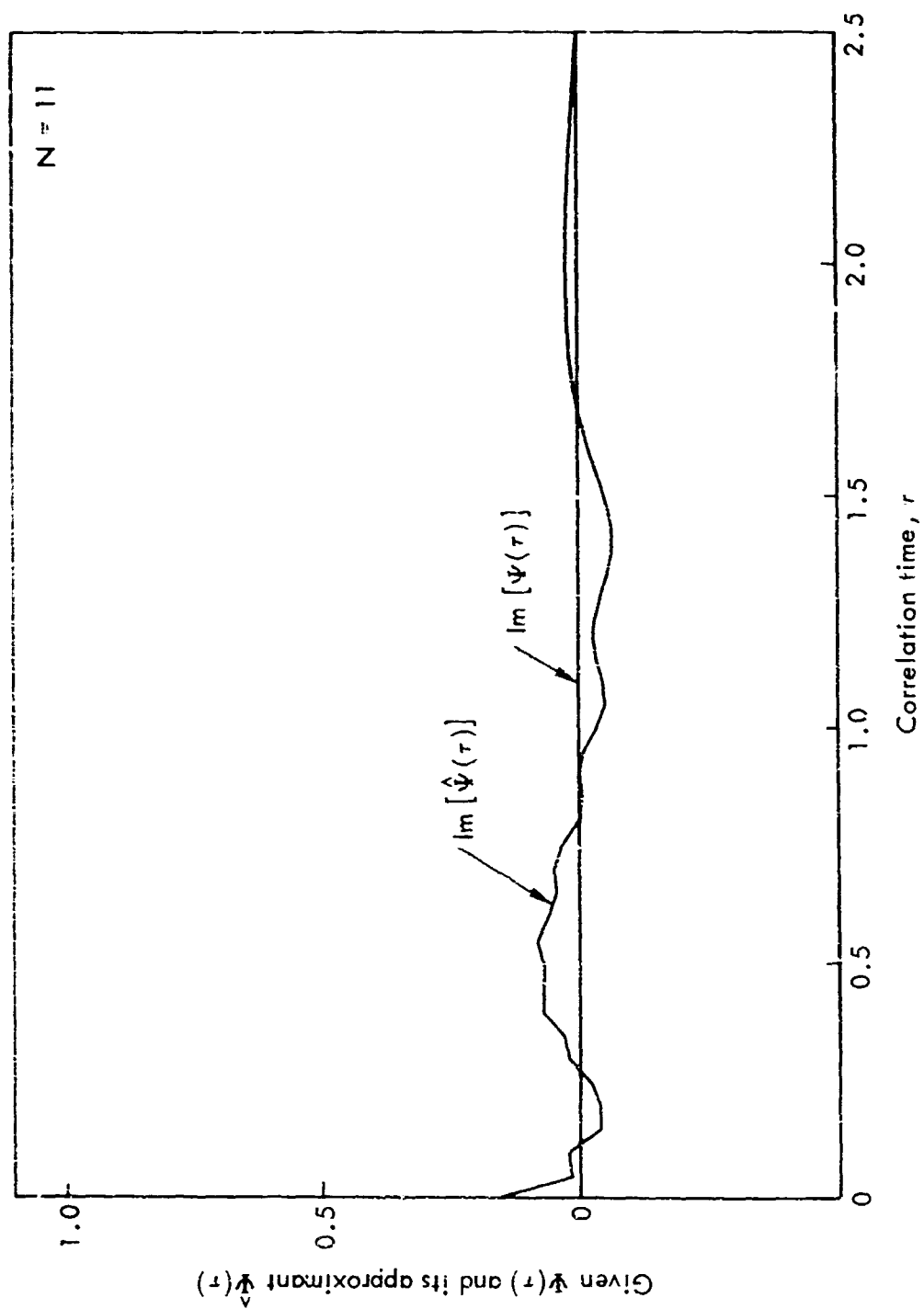


Fig. 7 — Graphs of the imaginary parts of $\Psi(\tau)$ and $\hat{\Psi}(\tau)$ for $N = 11$

$$\hat{Y}_-(\tau) = \sum_{k=1}^{12} a_k X_k(-\tau) \approx \Psi(\tau) \quad \tau \leq 0 \quad (221)$$

is

$$\hat{\Phi}(\omega) = \int_0^{\infty} \hat{Y}_+(\tau) e^{-j\omega\tau} d\tau + \int_{-\infty}^0 \hat{Y}_-(\tau) e^{-j\omega\tau} d\tau \quad (222)$$

$$\approx \int_{-\infty}^{\infty} \Psi(\tau) e^{-j\omega\tau} d\tau = \Phi(\omega)$$

Thus, by changing the variable in the second integral of Eq. (222) and by noting Eq. (221), Eq. (222) becomes

$$\hat{\Phi}(\omega) \approx \sum_{k=1}^{12} a_k \left\{ \int_0^{\infty} X_k(\tau) e^{-j\omega\tau} d\tau + \int_0^{\infty} X_k(\tau) e^{j\omega\tau} d\tau \right\} \quad (223)$$

Since $X_k(\tau)$ and $U_k(\omega)$ are related as Fourier transforms (to within a scale factor; see Eq. (23)), Eq. (223) can be written as

$$\hat{\Phi}(\omega) \approx \sum_{k=1}^{12} \sqrt{2\pi} a_k [U_m(\omega) + U_m^*(\omega)] \quad (224)$$

or, in view of Eqs. (21) and (4), as

$$\hat{\Phi}(\omega) \approx 2\sqrt{2\pi} \sum_{k=1}^{12} \sum_{n=1}^k a_k \left\{ \frac{n\sigma \operatorname{Re}[\lambda_{kn}] + (\omega - n\beta) \operatorname{Im}[\lambda_{kn}]}{(n\sigma)^2 + (\omega - n\beta)^2} \right\} = \hat{\hat{\Phi}}(\omega) \quad (225)$$

Thus, for any value of N , once the a_k are determined for the auto-correlation function $\Psi(\tau)$, the approximating spectral density can be written by inspection. In view of Eq. (145), moreover, the integral-square error between $\hat{\Phi}(\omega)$ and $\hat{\hat{\Phi}}(\omega)$ in Eq. (225) is simply $2\pi\epsilon_N$.

Appendix A

THE ORTHONORMAL BASIS COEFFICIENTS λ_{mn}

The rational function of σ and β given in Eq. (45) can be used to generate the coefficients λ_{mn} . As indicated in Section VII, however, there is considerable computational advantage in recursively evaluating λ_{mn} by Eq. (153). A program for doing this is presented in Appendix B.

In order to tabulate the algebraic λ_{mn} obtained by computer, the following definitions are convenient. From Eq. (45), λ_{mn} is expressed as the proper rational function

$$\lambda_{mn} = \begin{cases} \sqrt{\frac{\sigma}{\pi}} & n = m = 1 \\ \sqrt{\frac{m\sigma}{\pi}} \frac{\prod_{r=1}^{m-1} [n(\sigma - j\beta) + r(\sigma + j\beta)]}{(\sigma - j\beta)^{m-1} \prod_{r=1}^m (r-n)} & \begin{matrix} m = 2, 3, \dots \\ 1 \leq n \leq m \end{matrix} \\ 0 & \begin{matrix} m = 1, 2, \dots \\ n > m \end{matrix} \end{cases} \quad (226)$$

If η and Λ_{mn} are defined as

$$\eta \equiv j\beta \quad (227)$$

and

$$\Lambda_{mn} \equiv \frac{\prod_{r=1}^{m-1} [n(\sigma - j\beta) + r(\sigma + j\beta)]}{\prod_{r=1}^m (r-n)} = \frac{\prod_{r=1}^{m-1} [n(\sigma - \eta) + r(\sigma + \eta)]}{\prod_{r=1}^m (r-n)} = (\sigma - \eta)^{m-1} \alpha_{mn} \quad (228)$$

where α_{mn} is given by Eq. (68), then λ_{mn} can be reexpressed in terms of the polynomial Λ_{mn} in σ and η as

$$\lambda_{mn} = \begin{cases} \sqrt{\frac{\sigma}{\pi}} & n = m = 1 \\ \frac{\sqrt{\frac{m\sigma}{\pi}}}{(\sigma-\eta)^{m-1}} \Lambda_{mn} & \begin{matrix} m = 2, 3, \dots \\ 1 \leq n \leq m \end{matrix} \\ 0 & \begin{matrix} m = 1, 2, \dots \\ n > m \end{matrix} \end{cases} \quad (229)$$

The quantities Λ_{mn} defined by Eq. (228) are shown in Table 4 for the range $m = 1, 2, \dots, 10$ and $n = 1, 2, \dots, m$. For each value of m , the check-sum relation Eq. (67) holds, thereby verifying the exactness of the rational λ_{mn} derived from the polynomials Λ_{mn} .

The Λ_{mn} of Table 4 have been generated by encoding the recursive relation, Eq. (153), in ALTRAN, a symbolic manipulation language, and executing the program on a digital computer.⁽³⁸⁾ The ALTRAN compiler, which is required for execution of the program listed in Appendix B, produces MAP output consisting of transfers to ALPAK, a group of sub-routines for computing and simplifying polynomials and certain rational functions.

Since ALTRAN is not widely available and since it requires large computer storage for evaluating Λ_{mn} for m as high as ten, another program is given in Appendix C for computing the Λ_{mn} . The Fortran IV routine listed in Appendix C computes the integer coefficients of the variables σ and β in the polynomials Λ_{mn} . The computational basis

for the program is discussed in Appendix D where a series representation for Λ_{mn} is derived to replace the product form given by Eq. (228). Finally, in Appendix E a Fortran IV program is presented which enables arbitrary functions $g(t) \in L^2(0, \infty)$ to be expanded in terms of the orthonormal elements $X_m(t)$.

Table 4

THE ASSOCIATED ORTHONORMAL BASIS COEFFICIENTS Λ_{mn}

$$\Lambda_{11} = 1$$

$$\Lambda_{21} = 2\sigma$$

$$\Lambda_{22} = -3\sigma + \eta$$

$$\Lambda_{31} = 3\sigma^2 + \sigma\eta$$

$$\Lambda_{32} = -12\sigma^2 + 4\sigma\eta$$

$$\Lambda_{33} = 10\sigma^2 - 7\sigma\eta + \eta^2$$

$$\Lambda_{41} = (12\sigma^3 + 10\sigma^2\eta + 2\sigma\eta^2)/3$$

$$\Lambda_{42} = -3\sigma^3 + 4\sigma^2\eta + 2\sigma\eta^2$$

$$\Lambda_{43} = 60\sigma^3 - 42\sigma^2\eta + 6\sigma\eta^2$$

$$\Lambda_{44} = (-105\sigma^3 + 113\sigma^2\eta - 35\sigma\eta^2 + 3\eta^3)/3$$

$$\Lambda_{51} = (30\sigma^4 + 43\sigma^3\eta + 20\sigma^2\eta^2 + 3\sigma\eta^3)/6$$

$$\Lambda_{52} = (-180\sigma^4 - 36\sigma^3\eta + 20\sigma^2\eta^2 + 4\sigma\eta^3)/3$$

$$\Lambda_{53} = 210\sigma^4 - 117\sigma^3\eta + 3\sigma\eta^3$$

$$\Lambda_{54} = (-840\sigma^4 + 904\sigma^3\eta - 280\sigma^2\eta^2 + 24\sigma\eta^3)/3$$

$$\Lambda_{55} = (756\sigma^4 - 1101\sigma^3\eta + 536\sigma^2\eta^2 - 101\sigma\eta^3 + 6\eta^4)/6$$

$$\Lambda_{61} = (90\sigma^5 + 189\sigma^4\eta + 146\sigma^3\eta^2 + 49\sigma^2\eta^3 + 6\sigma\eta^4)/15$$

$$\Lambda_{62} = (-315\sigma^5 - 198\sigma^4\eta + 8\sigma^3\eta^2 + 22\sigma^2\eta^3 + 3\sigma\eta^4)/3$$

$$\Lambda_{63} = 560\sigma^5 - 172\sigma^4\eta - 78\sigma^3\eta^2 + 8\sigma^2\eta^3 + 2\sigma\eta^4$$

Table 4--continued

$$\Lambda_{64} = (-3780\sigma^5 + 3648\sigma^4\eta - 808\sigma^3\eta^2 - 32\sigma^2\eta^3 + 12\sigma\eta^4)/3$$

$$\Lambda_{65} = (3780\sigma^5 - 5505\sigma^4\eta + 2680\sigma^3\eta^2 - 505\sigma^2\eta^3 + 30\sigma\eta^4)/3$$

$$\Lambda_{66} = (-2310\sigma^5 + 4247\sigma^4\eta - 2842\sigma^3\eta^2 + 852\sigma^2\eta^3 - 112\sigma\eta^4 + 5\eta^5)/5$$

$$\Lambda_{71} = (630\sigma^6 + 1773\sigma^5\eta + 1967\sigma^4\eta^2 + 1073\sigma^3\eta^3 + 287\sigma^2\eta^4 + 30\sigma\eta^5)/90$$

$$\Lambda_{72} = (-2520\sigma^6 - 2844\sigma^5\eta - 728\sigma^4\eta^2 + 208\sigma^3\eta^3 + 112\sigma^2\eta^4 + 12\sigma\eta^5)/15$$

$$\Lambda_{73} = (2520\sigma^6 + 66\sigma^5\eta - 609\sigma^4\eta^2 - 81\sigma^3\eta^3 + 21\sigma^2\eta^4 + 3\sigma\eta^5)/2$$

$$\Lambda_{74} = (-37800\sigma^6 + 28920\sigma^5\eta - 784\sigma^4\eta^2 - 1936\sigma^3\eta^3 + 56\sigma^2\eta^4 + 24\sigma\eta^5)/9$$

$$\Lambda_{75} = (41580\sigma^6 - 56775\sigma^5\eta + 23975\sigma^4\eta^2 - 2875\sigma^3\eta^3 - 175\sigma^2\eta^4 + 30\sigma\eta^5)/6$$

$$\Lambda_{76} = (-27720\sigma^6 + 50964\sigma^5\eta - 34104\sigma^4\eta^2 + 10224\sigma^3\eta^3 - 1344\sigma^2\eta^4 + 60\sigma\eta^5)/5$$

$$\Lambda_{77} = (154440\sigma^6 - 343146\sigma^5\eta + 293243\sigma^4\eta^2 - 122023\sigma^3\eta^3 + 25703\sigma^2\eta^4 - 2547\sigma\eta^5 + 90\eta^6)/90$$

$$\Lambda_{81} = (2520\sigma^7 + 8982\sigma^6\eta + 13187\sigma^5\eta^2 + 10193\sigma^4\eta^3 + 4367\sigma^3\eta^4 + 981\sigma^2\eta^5 + 90\sigma\eta^6)/315$$

$$\Lambda_{82} = (-11340\sigma^7 - 19098\sigma^6\eta - 10586\sigma^5\eta^2 - 884\sigma^4\eta^3 + 1024\sigma^3\eta^4 + 334\sigma^2\eta^5 + 30\sigma\eta^6)/45$$

$$\Lambda_{83} = (12600\sigma^7 + 5370\sigma^6\eta - 2913\sigma^5\eta^2 - 1623\sigma^4\eta^3 - 57\sigma^3\eta^4 + 57\sigma^2\eta^5 + 6\sigma\eta^6)/5$$

$$\Lambda_{84} = (-103950\sigma^7 + 51180\sigma^6\eta + 19534\sigma^5\eta^2 - 5912\sigma^4\eta^3 - 1298\sigma^3\eta^4 + 108\sigma^2\eta^5 + 18\sigma\eta^6)/9$$

Table 4--continued

$$\Lambda_{85} = (249480\sigma^7 - 299070\sigma^6\eta + 87075\sigma^5\eta^2 + 6725\sigma^4\eta^3 - 3925\sigma^3\eta^4 + 5\sigma^2\eta^5 + 30\sigma\eta^6)/9$$

$$\Lambda_{86} = (-180180\sigma^7 + 317406\sigma^6\eta - 196194\sigma^5\eta^2 + 49404\sigma^4\eta^3 - 3624\sigma^3\eta^4 - 282\sigma^2\eta^5 + 30\sigma\eta^6)/5$$

$$\Lambda_{87} = (1081080\sigma^7 - 2402022\sigma^6\eta + 2052701\sigma^5\eta^2 - 854161\sigma^4\eta^3 + 179921\sigma^3\eta^4 - 17829\sigma^2\eta^5 + 630\sigma\eta^6)/45$$

$$\Lambda_{88} = (-2027025\sigma^7 + 5282325\sigma^6\eta - 5503581\sigma^5\eta^2 + 2947489\sigma^4\eta^3 - 867299\sigma^3\eta^4 + 138319\sigma^2\eta^5 - 10863\sigma\eta^6 + 315\eta^7)/315$$

$$\Lambda_{91} = (22680\sigma^8 + 98478\sigma^7\eta + 181557\sigma^6\eta^2 + 184046\sigma^5\eta^3 + 110654\sigma^4\eta^4 + 39398\sigma^3\eta^5 + 7677\sigma^2\eta^6 + 630\sigma\eta^7)/2520$$

$$\Lambda_{92} = (-113400\sigma^8 - 259020\sigma^7\eta - 218448\sigma^6\eta^2 - 71156\sigma^5\eta^3 + 4936\sigma^4\eta^4 + 9484\sigma^3\eta^5 + 2304\sigma^2\eta^6 + 180\sigma\eta^7)/315$$

$$\Lambda_{93} = (46200\sigma^8 + 40690\sigma^7\eta - 1731\sigma^6\eta^2 - 10806\sigma^5\eta^3 - 2914\sigma^4\eta^4 + 114\sigma^3\eta^5 + 117\sigma^2\eta^6 + 10\sigma\eta^7)/10$$

$$\Lambda_{94} = (-1247400\sigma^8 + 198360\sigma^7\eta + 439128\sigma^6\eta^2 + 7192\sigma^5\eta^3 - 39224\sigma^4\eta^4 - 3896\sigma^3\eta^5 + 648\sigma^2\eta^6 + 72\sigma\eta^7)/45$$

$$\Lambda_{95} = (2243240\sigma^8 - 3139470\sigma^7\eta + 234765\sigma^6\eta^2 + 348650\sigma^5\eta^3 - 30850\sigma^4\eta^4 - 11710\sigma^3\eta^5 + 405\sigma^2\eta^6 + 90\sigma\eta^7)/36$$

$$\Lambda_{96} = (-840840\sigma^8 + 1361108\sigma^7\eta - 703968\sigma^6\eta^2 + 99756\sigma^5\eta^3 + 16024\sigma^4\eta^4 - 3732\sigma^3\eta^5 - 48\sigma^2\eta^6 + 20\sigma\eta^7)/5$$

$$\Lambda_{97} = (16216200\sigma^8 - 34949250\sigma^7\eta + 2388493\sigma^6\eta^2 - 10759714\sigma^5\eta^3 + 1844654\sigma^4\eta^4 - 87514\sigma^3\eta^5 - 8379\sigma^2\eta^6 + 630\sigma\eta^7)/90$$

Table 4--continued

$$\begin{aligned}
 \Lambda_{98} &= (-32432400\sigma^8 + 84517200\sigma^7\eta - 88057296\sigma^6\eta^2 + 47159824\sigma^5\eta^3 \\
 &\quad - 13876784\sigma^4\eta^4 + 2213104\sigma^3\eta^5 - 173808\sigma^2\eta^6 + 5040\sigma\eta^7)/315 \\
 \Lambda_{99} &= (6806800\sigma^8 - 20355850\sigma^7\eta + 25047613\sigma^6\eta^2 - 16458298\sigma^5\eta^3 \\
 &\quad + 6267166\sigma^4\eta^4 - 1402306\sigma^3\eta^5 + 177733\sigma^2\eta^6 - 11458\sigma\eta^7 + 280\eta^8)/280 \\
 \Lambda_{10\ 1} &= (11340\sigma^9 + 583110\sigma^8\eta + 1301697\sigma^7\eta^2 + 1646458\sigma^6\eta^3 + 1289454\sigma^5\eta^4 \\
 &\quad + 639606\sigma^4\eta^5 + 195977\sigma^3\eta^6 + 33858\sigma^2\eta^7 + 2520\sigma\eta^8)/11340 \\
 \Lambda_{10\ 2} &= (-311850\sigma^9 - 910755\sigma^8\eta - 1054017\sigma^7\eta^2 - 577963\sigma^6\eta^3 - 110949\sigma^5\eta^4 \\
 &\quad + 34719\sigma^4\eta^5 + 22933\sigma^3\eta^6 + 4527\sigma^2\eta^7 + 315\sigma\eta^8)/630 \\
 \Lambda_{10\ 3} &= (277200\sigma^9 + 382740\sigma^8\eta + 111684\sigma^7\eta^2 - 70029\sigma^6\eta^3 - 49902\sigma^5\eta^4 \\
 &\quad - 8058\sigma^4\eta^5 + 1044\sigma^3\eta^6 + 411\sigma^2\eta^7 + 30\sigma\eta^8)/35 \\
 \Lambda_{10\ 4} &= (-8108100\sigma^9 - 1829160\sigma^8\eta + 3350232\sigma^7\eta^2 + 1144568\sigma^6\eta^3 \\
 &\quad - 236976\sigma^5\eta^4 - 123384\sigma^4\eta^5 - 5528\sigma^3\eta^6 + 2088\sigma^2\eta^7 + 180\sigma\eta^8)/135 \\
 \Lambda_{10\ 5} &= (4540536\sigma^9 - 3097962\sigma^8\eta - 927117\sigma^7\eta^2 + 582016\sigma^6\eta^3 + 96270\sigma^5\eta^4 \\
 &\quad - 28734\sigma^4\eta^5 - 4117\sigma^3\eta^6 + 288\sigma^2\eta^7 + 36\sigma\eta^8)/18 \\
 \Lambda_{10\ 6} &= (-3153150\sigma^9 + 4473525\sigma^8\eta - 1619049\sigma^7\eta^2 - 153891\sigma^6\eta^3 \\
 &\quad + 134907\sigma^5\eta^4 - 1977\sigma^4\eta^5 - 2979\sigma^3\eta^6 + 39\sigma^2\eta^7 + 15\sigma\eta^8)/5 \\
 \Lambda_{10\ 7} &= (129729600\sigma^9 - 263377800\sigma^8\eta + 192158694\sigma^7\eta^2 - 57689219\sigma^6\eta^3 \\
 &\quad + 3997518\sigma^5\eta^4 + 1144542\sigma^4\eta^5 - 154546\sigma^3\eta^6 - 3339\sigma^2\eta^7 + 630\sigma\eta^8)/135 \\
 \Lambda_{10\ 8} &= (275675400\sigma^9 + 702180000\sigma^8\eta - 706228416\sigma^7\eta^2 + 356829856\sigma^6\eta^3 \\
 &\quad - 94372752\sigma^5\eta^4 + 11872992\sigma^4\eta^5 - 370816\sigma^3\eta^6 - 44064\sigma^2\eta^7 \\
 &\quad + 2520\sigma\eta^8)/315
 \end{aligned}$$

Table 4--continued

$$\begin{aligned} \Lambda_{10 \ 9} = & (61261200\sigma^9 - 183202650\sigma^8\eta + 225428517\sigma^7\eta^2 - 148124682\sigma^6\eta^3 \\ & + 56404494\sigma^5\eta^4 - 12620754\sigma^4\eta^5 + 1599597\sigma^3\eta^6 - 103122\sigma^2\eta^7 \\ & + 2520\sigma\eta^8)/140 \end{aligned}$$

$$\begin{aligned} \Lambda_{10 \ 10} = & (-104756652\sigma^9 + 353598543\sigma^8\eta - 502107309\sigma^7\eta^2 + 391672865\sigma^6\eta^3 \\ & - 183840261\sigma^5\eta^4 + 53443743\sigma^4\eta^5 - 9532127\sigma^3\eta^6 + 993411\sigma^2\eta^7 \\ & - 53955\sigma\eta^8 + 1134\eta^9)/1134 \end{aligned}$$

Appendix B

AN ALTRAN PROGRAM FOR GENERATING Λ_{mn}

The ALTRAN program listed in this appendix is used for computing the polynomial's Λ_{mn} of Eq. (228). A recurrence relation similar to Eq. (153) is the basis for the routine.

In order to execute this program, an ALTRAN compiler is required to produce MAP and, thereby, the final object code.[†] The accompanying routine and its associated control statements are appropriate for execution on The RAND Corporation's IBM 7044 computer. The prologue to the listings describes the program's parameters, usage, and limitations.

[†] The ALTRAN compiler consists of transfers to ALPAK, a group of subroutines for operating on certain polynomials and rational functions. The details of ALTRAN and ALPAK, including the format for representing polynomials, are discussed in Ref. 38.

C
C
C THE FOLLOWING LISTING IS AN ALTRAN PROGRAM FOR COMPUTING
C THE POLYNOMIALS $\Lambda(M,N)$ IN THE VARIABLES SIGMA AND BETA.
C THE POLYNOMIALS ARE EXPRESSED AS CONSTANTS TIMES $(\text{SIGMA}^{**}(M-1)) \times$
C $(\text{BETA}^{**}1)$ FOR $I=0,1,\dots,M$ AND FOR $M=1,2,\dots$ AND FOR $N=1,2,\dots,M$
C THE $\Lambda(M,N)$ ARE ZERO FOR N GREATER THAN M.
C THE FOLLOWING PROGRAM CALCULATES $\Lambda(M,N)$ FOR $M=1,2,\dots,10$.
C THE CONTROL CARDS LISTED BELOW INDICATE THE APPROPRIATE DECK
C SET UP FOR EXECUTION ON THE IBM 7044. BINARY DECKS ARE NOT
C FULLY LISTED. THE $\Lambda(M,N)$ ARE MADE AVAILABLE AS BOTH
C PRINTED AND PUNCHED CARD OUTPUT.
C

C
C
\$CLOSE S.SU07,REWIND
\$IBJOB MAP,FILES
\$FILE 'S.FBIA',NONF,*,BLOCK=10
\$FILE 'S.FBOA',NONE,*,BLOCK=10
\$IEDIT U07,SRCH
\$IBLDR TMG
\$IBLDR TMG10
\$IBLDR TMGDFN
\$IBLDR ALTRAN
\$ENTRY TMG
STORAGE 13000
LAYOUT (L) SIGMA 18, GAMMA 18
ALGEBRAIC (L) $\Lambda(10,10)$
INTEGER M,N,R,I
 $\Lambda(1,1)=1$
 $\Lambda(2,1)=2*\text{SIGMA}$
 $\Lambda(2,2)=-3*\text{SIGMA}+\text{GAMMA}$
PRINT $\Lambda(1,1),\Lambda(2,1),\Lambda(2,2)$
PUNCH $\Lambda(1,1),\Lambda(2,1),\Lambda(2,2)$
DO 20 M=3,10
I=M-1
DO 15 N=1,I
 $\Lambda(M,N)=\Lambda(M-1,N)*(N*(\text{SIGMA}-\text{GAMMA})+I*(\text{SIGMA}+\text{GAMMA}))/(M-N)$
PRINT M,N, $\Lambda(M,N)$
PUNCH $\Lambda(M,N)$
15 CONTINUE
 $\Lambda(M,M)=1$
DO 16 R=1,I
 $\Lambda(M,M)=\Lambda(M,M)*(M*(\text{SIGMA}-\text{GAMMA})+R*(\text{SIGMA}+\text{GAMMA}))/ (R-M)$
16 CONTINUE
PRINT $\Lambda(M,M)$
PUNCH $\Lambda(M,M)$
20 CONTINUE
STOP
END
FINISH

\$IBSYS
\$CLOSE S.SU06,REWIND
\$IBJOB MAP,100P1
\$IEDIT U06,SRCH

\$IBMAP ALTRAN
\$EDIT U07,SRCH
\$IBLDR ALFSRT
\$IBLDR READF
\$IBLDR READD
\$IBLDR READI
\$IBLDR OUT
\$IBLDR PUNCHP
\$IBLDR ALF
\$IBLDR ALP
\$EDIT IN
\$IBLDR IOREO 10/28/66
\$CDICT IOREO
C BINARY CARDS DELETED
\$TEXT IOREO
C BINARY CARDS DELETED
\$DKEND IOREO
\$IBLDR POSTXX 10/28/66
\$CDICT POSTXX
C BINARY CARDS DELETED
\$TEXT POSTXX
C BINARY CARDS DELETED
\$DKEND POSTXX
\$ENTRY ALT
\$IBSYS
\$CLOSE S.SU07,REMOVE

Appendix C

A FORTRAN IV PROGRAM FOR GENERATING THE COEFFICIENTS IN Λ_{mn}

The FORTRAN IV program listed in this appendix is used to compute the constants in the polynomials Λ_{mn} of Eq. (228). The routine is based on Eqs. (237), (240), and (241) of Appendix D.

The following program is compatible with the IBM 7044, 7094, and 360 series FORTRAN IV compilers. The program obviates the large storage needed in the ALTRAN routine discussed in Appendix B. The routine also allows cross-checking with the results of ALTRAN.

The prologue to the listing describes the parameters, usage, and limitations of the program, as well as the format of the printed results. All of the computation is performed in double precision.

\$IBFTC COEFF

C
C
C THIS ROUTINE PRODUCES THE MATRIX OF LAMBDA(M,N) COEFFICIENTS.
C M RANGES FROM 1 TO 15 AND N RANGES FROM 1 TO M.
C THE COEFFICIENTS ARE ZERO FOR ALL N GREATER THAN M.
C THE COEFFICIENTS OF THE VARIABLES SIGMA**I X BETA**J ARE GIVEN
C FOR EACH POLYNOMIAL LAMBDA(M,N). THE OUTPUT FORMAT IS AS FOLLOWS
C M= , N= , POWER OF SIGMA= , POWER OF BETA= , COEFFICIENT= .
C ALL COMPUTATION IS IN DOUBLE PRECISION. FOR M GREATER THAN 15
C HIGHER PRECISION IS REQUIRED, THE INDEXING MUST BE INCREASED
C AND THE DIMENSION STATEMENTS MUST BE ADJUSTED.
C

DOUBLE PRECISION S(16,16),F(16),AM,AN,A,ANS,B,C
INTEGER R,P
COMPLEX Q
LOGICAL TEST
WRITE(6,6)
6 FORMAT(1H1////10X,1HM,3X, 1HN,3X,14HPOWER OF SIGMA,3X,
.13HPOWER OF BETA,3X,30HCOEFFICIENT OF SIGMA-BETA TERM ////)
DO 1 M=1,15
S(M,M)=1.00
AM=DBLE(FLOAT(M))
S(M+1,1)=-AM*S(M,1)
M1=M+1
DO 1 K=M1,15
S(M,K)=0.00
1 CONTINUE
DO 4 M=2,15
AM=DBLE(FLOAT(M))
DO 4 K=2,15
4 S(M+1,K)=S(M,K-1)-AM*S(M,K)
F(1)=1.00
F(2)=F(1)
DO 2 I=2,14
2 F(I+1)=F(I)*DBLE(FLOAT(I))
DO 100 M=1,10
AM=DBLE(FLOAT(M))
M2=M-1
DO 100 N=1,M
AN=DBLE(FLOAT(N))
MN=M-N+1
A=((-1.00)**(M+N))/(F(N+1)*F(MN))
DO 100 II=1,M
ISIGMA=IABS(M-II)
IF(II.EQ.M.AND.N.NE.M) GO TO 100
IBETA=IABS(M2-ISIGMA)
ANS=0.00
DO 3 K=1,M
MK=M-K +1
U=(-AN)**K*S(M,K)*F(MK)*F(K)
DO 3 R=1,MK
MKR=M-K-R+2
DO 3 P=1,K

```
C=B*(-1.00)**(P-1)
KP=K-P+1
IF((M-R-P+1).EQ.ISIGMA) ANS=ANS+C/(F(R)*F(P)*F(MKR)*F(KP))
3 CONTINUE
ANS=-ANS*A
Q=CMPLX(0.,1.)**IBETA
IQ1=REAL(Q)
IQ2=AIMAG(Q)
TEST=IQ2.EQ.0.AND.IQ1.LT.0
IF(TEST) ANS=-ANS
IF(TEST) WRITE(6,7) M,N,ISIGMA,IBETA,ANS
7 FORMAT(9X,I2,2X,I2,9X,I2,15X,I2,11X,D24.16)
IF(TEST) GO TO 100
TEST=IQ1.EQ.0.AND.IQ2.LT.0
IF(TEST) ANS=-ANS
IF(TEST) WRITE(6,5) M,N,ISIGMA,IBETA,ANS
5 FORMAT(9X,I2,2X,I2,9X,I2,15X,I2,11X,D24.16,2X,1H1)
IF(TEST) GO TO 100
IF(IQ1.EQ.0.AND.IQ2.GE.0) WRITE(6,5) M,N,ISIGMA,IBETA,ANS
IF(IQ2.EQ.0.AND.IQ1.GE.0) WRITE(6,7) M,N,ISIGMA,IBETA,ANS
100 CONTINUE
CALL EXIT
END
```

Appendix D

A SERIES REPRESENTATION FOR Λ_{mn}

For reasons discussed in Appendix A, a series representation for Λ_{mn} is a useful supplement to the product form given in Eq. (228).

Starting with Eqs. (68) and (228) for α_{mn} and Λ_{mn} , one can write

$$\alpha_{mn} = \sqrt{\frac{\pi}{m\sigma}} \lambda_{mn} \quad (230)$$

and

$$\Lambda_{mn} = (\sigma - j\beta)^{m-1} \alpha_{mn} \quad (231)$$

From Eq. (78) it follows that

$$\alpha_{mn} = \frac{(-1)^{m+1}}{m!} \sum_{k=0}^m (-1)^{k+n} \binom{m}{n} S_m^{(k)} n^k z^{m-k} \quad (232)$$

where z is defined by Eq. (71), and $S_m^{(k)}$ are the Stirling numbers of the first kind given by Eq. (73).

Thus

$$\Lambda_{mn} = \frac{(-1)^{m+n+1}}{(m-n)! n!} \sum_{k=0}^m (-1)^k S_m^{(k)} n^k (\sigma + j\beta)^{m-k} (\sigma - j\beta)^{k-1} \quad (233)$$

The binominal theorem⁽³⁷⁾ indicates that

$$(\sigma + j\beta)^{m-k} = \sum_{r=0}^{m-k} \sigma^{m-k-r} (j\beta)^r \binom{m-k}{r} \quad (234)$$

and

$$(\sigma - j\beta)^{k-1} = \sum_{p=0}^{k-1} \sigma^{k-1-p} (-j\beta)^p \binom{k-1}{p} \quad (235)$$

Substitution of Eqs. (234) and (235) in Eq. (233) therefore results in

$$\Lambda_{mn} = \frac{(-1)^{m+n+1}}{n!(m-n)!} \sum_{k=0}^m \sum_{r=0}^{m-k} \sum_{p=0}^{k-1} \left\{ \frac{(-1)^{k+p} S_m^{(k)} n^k (m-k)!(k-1)!}{r!p!(m-k-r)!(k-p-1)!} \right\} \sigma^{m-r-p-1} (j\beta)^{r+p} \quad (236)$$

Since $S_m^{(0)} = 0$, the series representation for Λ_{mn} simplifies to

$$\Lambda_{mn} = \frac{(-1)^{m+n+1}}{n!(m-n)!} \sum_{k=1}^m (-1)^k S_m^{(k)} n^k (m-k)!(k-1)! \sum_{r=1}^{m-k+1} \sum_{p=1}^k \left\{ \frac{(-1)^{p(j)} (j)^{r+p}}{(r-1)!(p-1)!(k-p)!(m-k-r+1)!} \right\} \sigma^{m-r-p+1} \beta^{r+p-2} \quad (237)$$

In order to obtain the coefficients c_i in the representation

$$\Lambda_{mn} = \sum_{i=0}^{m-1} c_i \sigma^{m-1-i} \beta^i \quad (238)$$

which follows from Eq. (237), one notes that r and p must be selected according to

$$\begin{cases} \sigma^{m-1-i} = \sigma^{m-r-p+1} & r = 1, 2, \dots, m-k+1 \\ \beta^i = \beta^{r+p-2} & p = 1, 2, \dots, k \\ & k = 1, 2, \dots, m \end{cases} \quad (239)$$

Thus, for any value of i and r

$$m-1-i = m-r-p+1 \quad i = 0, 1, \dots, m-1; \quad k = 1, 2, \dots, m; \quad r = 1, 2, \dots, m-k+1 \quad (240)$$

so that p is constrained to the value(s)

$$p = i-r+2 \quad \text{with } p \in \{1, 2, \dots, k\} \quad (241)$$

Consequently, c_i is the aggregate of all terms of Eq. (237) obtained by allowing k to range over the integers from 1 to m ; r , from 1 to $m-k+1$; and $p = i-r+2$, with p contained in the set $k = 1, 2, \dots, m$. Such a summation is easily programmed for digital computation; a FORTRAN IV routine for finding the c_i is provided in Appendix C.

Appendix E

A FORTRAN IV PROGRAM FOR COMPUTING EXPONENTIAL APPROXIMATIONS $\hat{g}(t)$

The accompanying program is designed to calculate the exponential approximation $\hat{g}(t)$ to a prescribed function $g(t) \in L^2(0, \infty)$. The algorithm is based on the recurrence relations Eqs. (157) and (158) for the m^{th} orthonormal basis function $X_m(t)$, on recurrence relation Eq. (153) for λ_{mn} , and on Eqs. (132) or (162) for the moments of $g(t)$. These moments are approximated by a 64-point Gaussian quadrature scheme and are used to form the Fourier expansion coefficients a_m according to Eq. (131). The approximant $\hat{g}(t)$ is finally obtained from Eq. (133).

The numerical examples discussed in Section IX have been solved with the use of program APRX. The values of σ and β obtained from Eqs. (175), or from Eqs. (184) and (185), are input parameters to the program as described in the prologue of APRX. All other program options, definitions, and limitations are also clarified in the listings.

The real and imaginary parts of $g(t)$ must be supplied as the double-precision function subroutines named GR(T) and GI(T), respectively. Once APRX is entered, GR, GI, and all the supporting routines are automatically invoked to produce a 51-point tabular and graphical display of $\hat{g}(t)$. Card output of $\hat{g}(t)$ for subsequent processing is also available.

CC

```

DOUBLE PRECISION CFR(20),CFI(20),GHATR,GHATI,T,SIG,BET
.,ERRORR,RMSE,AA,BB,S2P,ESIGT,PI,
.LSR(20,20),LSI(20,20),SI(20),S2(20)
COMMON /COEFF/ CFR,CFI
COMMON /MOMENT/S1,S2
COMMON /LSCOM/LSR,LSI
COMMON SIG,BET,MC,NNN,AA,BB,S2P,ESIGT,T,GHATR,GHATI
EXTERNAL ERRORR
REAL GRAPH(1000),ORDR(51),ORDI(51),ABSC(51),H(51),HH(51)
DATA PI/3.141592653589793D0/
S2P=DSQRT(2.D0*PI)
1 READ(5,8) MC,SIG,BET,AA,BB,GRMAX,GRMIN,TMAX,TMIN,IPUNCH
8 FORMAT(13,2D16.1, 2D5.1,4F5.1,15)
WRITE(6,23) SIG,BET,MC,AA,BB
23 FORMAT(1H1,6HSIGMA=,D24.16//6H BETA=,D24.16//8H FOR M= ,12,6H TERM
.S //9H FROM T= ,D24.16//7H TO T= ,D24.16//)
CALL LAMBDA
WRITE(6,18)
18 FORMAT(//27H THE MATRIX OF LAMBDA(M,N) //)

```

```

DO 19 M=1,MC
19 WRITE(6,20) (M,N,LSR(M,N),LSI(M,N),N=1,M)
20 FORMAT(3H L(,I2,1H,I2,2H)=,D24.16,4H +I ,D24.16)
WRITE(6,17)
17 FORMAT(//10H THE MOMENT VECTOR      //)
CALL COEF
WRITE(6,16) (I,S1(I),S2(I),I=1,MC)
16 FORMAT(3H G(,I2,2H)=,D24.16,4H +I ,D24.16 )
WRITE(6,900)
900 FORMAT(///70H THE REAL AND IMAGINARY PARTS OF THE EXPANSION COEFF
.ICIENTS , A(M)                      ///)
WRITE(6,901) (I,CFR(I),CFI(I),I=1,MC)
901 FORMAT(4H A(,I2,2H)=,D24.16,5H +J ,D24.16)
DIV=(TMAX-TMIN)/50.
DO 10 I=1,51
ABSC(I)=AA+FLOAT(I-1)*DIV
T=DBLE(ABSC(I))
H(I)=GR(T)
HH(I)=GI(T)
CALL GHAT
ORDR(I)=GHATR
ORDI(I)=GMATI
IF(IPUNCH) 11,10,11
11 WRITE(7,7) ABSC(I),H(I),HH(I),ORDR(I),ORDI(I)
7 FORMAT(5F10.4)
10 CONTINUE
CALL INTGRL(AA,BB,ERRORR,RMSE)
RMSE=DSQRT(RMSE)/(BB-AA)
CALL PLOT2(GRAPH,TMAX,TMIN,GRMAX,GRMIN)
CALL PLOT3(1H+,ABSC(1),H(1),51)
CALL PLOT3(1H*,ABSC(1),ORDR(1),51)
WRITE(6,808)
808 FORMAT(///)
WRITE(6,800) (ABSC(I),ORDR(I),ORDI(I),H(I),HH(I),I=1,51 )
800 FORMAT(4H T=,F6.2,3X,9HRE(GHAT)=,E17.8,3X,9HIM(GHAT)=,E17.8,3X,
.6HRE(G)=,E17.8,3X,6HIM(G)=,E17.8 )
WRITE(6,812) RMSE
812 FORMAT(/////15H THE RMS ERROR= , E10.3/)
WRITE(6,809 )
809 FORMAT(1H1)
CALL PLOT4(1,1H )
WRITE(6,801)
801 FORMAT(35H0+=RE(G(T))      *=RE(GHAT(T))      )
CALL PLOT2(GRAPH,TMAX,TMIN,GRMAX,GRMIN)
CALL PLOT3(1H+,ABSC(1),HH(1),51)
CALL PLOT3(1H*,ABSC(1),ORDI(1),51)
WRITE(6,809)
CALL PLOT4(1,1H )
WRITE(6,802)
802 FORMAT(35H0+=IM(G(T))      *=IM(GHAT(T))      )
GO TO 1
CALL EXIT
END

```

\$IBFTC LAMBDA

C
C
C
C
C
C
C

THIS SUBROUTINE COMPUTES THE COEFFICIENTS LAMBDA(M,N) IN DOUBLE PRECISION. THE REAL AND IMAGINARY PARTS OF LAMBDA(M,N) ARE STORED IN ARRAYS LSR(M,N) AND LSI(M,N) RESPECTIVELY.

```

SUBROUTINE LAMBDA
  DOUBLE PRECISION LSR( 20,20),LSI(20,20),SIG,BET,PI,A,B,C,AM,AN,
  Q,E,F,G,H,AK,D
  DATA PI/3.141592653589793/
  COMMON SIG,BET,MC
  COMMON /LSCOM/LSR,LSI
  A=SIG*SIG+BET*BET
  B=SIG*SIG-BET*BET
  C=2.DO*SIG*BET
  LSR(1,1)=DSQRT(SIG/PI)
  LSI(1,1)=0.DO
  DO 1 M=1,MC
    AM=M
    E=(-1)**M
    F=C*AM
    G=B*AM
    U=DSQRT((AM+1.DO)/AM)
    DO 2 N=1,M
      AN=N
      Q=D/(A*(AM+1.DO-AN))
      LSR(M+1,N)=(LSR(M,N)*(AN*A+ G)-LSI(M,N)*F)*Q
2    LSI(M+1,N)=(LSI(M,N)*(AN*A+ G)+LSR(M,N)*F)*Q
      H=DSQRT(AM+1.DO)
      F=E*H*LSR(1,1)
      G=0.DO
    DO 3 K=1,M
      F=F-LSR(M+1,K)
3    G=G-LSI(M+1,K)
      LSR(M+1,M+1)=F
1    LSI(M+1,M+1)=G
  RETURN
END

```

\$IBFTC COEF

C
C
C
C
C
C
C

THIS SUBROUTINE COMPUTES THE FOURIER COEFFICIENTS A(M) IN DOUBLE PRECISION. THE REAL AND IMAGINARY PARTS OF THE RESULTS ARE STORED IN THE ARRAYS CFR(M) AND CFI(M) RESPECTIVELY.

```

SUBROUTINE COEF
  EXTERNAL RINTG,IINTG
  DOUBLE PRECISION IINTG,RINTG,SIG,BET,S2P,LSR(20,20),AA,BB

```

```

.,LSI(20,20),CFR(20),CFI(20),SI(20),S2(20),PI,C1(20,20),C2(20,20)
.,C3(20,20),C4(20,20),ESIGT,T
COMMON /COEFF/CFR,CFI
COMMON /LSCOM/LSR,LSI
COMMON /ARRAY/C1,C2,C3,C4
COMMON /MOMENT/S1,S2
COMMON SIG,BET,MC,NNN,AA,BB,S2P,ESIGT,T,GHATR,GHATI
DO 10 N=1,MC
NNN=N
CALL INTGRL(AA,BB,INTG,S1(N))
CALL INTGRL(AA,BB,IINTG,S2(N))
10 CONTINUE
DO 1 M=1,MC
CFR(M)=0.00
CFI(M)=0.00
DO 5 N=1,M
CFR(M)=CFR(M)+LSR(M,N)*S1(N)+LSI(M,N)*S2(N)
CFI(M)=CFI(M)+LSR(M,N)*S2(N)-LSI(M,N)*S1(N)
5 CONTINUE
CFR(M)=S2P*CFR(M)
CFI(M)=S2P*CFI(M)
1 CONTINUE
DO 2 M=1,MC
DO 2 N=1,M
C1(M,N)=CFR(M)*LSR(M,N)-CFI(M)*LSI(M,N)
C2(M,N)=-CFR(M)*LSI(M,N)-CFI(M)*LSR(M,N)
C3(M,N)=CFI(M)*LSR(M,N)+CFR(M)*LSI(M,N)
2 C4(M,N)=-CFI(M)*LSI(M,N)+CFR(M)*LSR(M,N)
RETURN
END

```

\$IBFTC GHAT

C
C
C THIS SUBROUTINE FORMS THE REAL AND IMAGINARY PARTS OF THE
C APPROXIMATION GHAT(T) TO THE PRESCRIBED FUNCTION G(T). ALL THE
C COMPUTATION IS IN DOUBLE PRECISION. THE REAL AND IMAGINARY PARTS
C OF THE APPROXIMANT ARE STORED IN GHATR AND GHATI RESPECTIVELY
C FOR EACH VALUE OF THE INDEPENDENT VARIABLE T.
C
C

```

SUBROUTINE GHAT
DOUBLE PRECISION C1(20,20),C2(20,20),C3(20,20),C4(20,20),EE,DE
.,GHATR,GHATI,T,SIG,BET,AA,BB,S2P,ESIGT,BT,CC,SS
COMMON /ARRAY/C1,C2,C3,C4
COMMON SIG,BET,MC,NNN,AA,BB,S2P,ESIGT,T,GHATR,GHATI
BT=BET*T
ESIGT=DEXP(-SIG*T)
GHATR=0.00
GHATI=0.00
DO 1 M=1,MC
DE=1.00
DO 1 N=1,M

```

```

DE=DE*ESIGT
EE=BT*DBLE(FLOAT(N))
CC=DCOS(EE)
SS=DSIN(EE)
GHATR=GHATR+DE*(CC*C1(M,N)+SS*C2(M,N))
1 GHATI=GHATI+DE*(CC*C3(M,N)+SS*C4(M,N))
GHATR=GHATR*S2P
GHATI=GHATI*S2P
RETURN
END

```

\$IBFTC RINTG

```

C
C
C THIS SUBROUTINE COMPUTES THE REAL PART OF THE MOMENTS OF THE
C PRESCRIBED FUNCTION G(T). THESE ARE USED IN COMPUTING THE
C FOURIER EXPANSION COEFFICIENTS. COMPUTATION IS IN DOUBLE PRECISION.
C
C
C DOUBLE PRECISION FUNCTION RINTG(T)
C DOUBLE PRECISION GR,GI,SIG,BET,AN,T
C COMMON SIG,BET,MC,N
C AN=N
C RINTG=DEXP(-AN*SIG*T)*(GR(T)*DCOS(AN*BET*T)+GI(T)*DSIN(AN*BET*T))
C RETURN
C END

```

\$IBFTC IINTG

```

C
C
C THIS ROUTINE COMPUTES THE IMAGINARY PART OF THE MOMENTS OF THE
C PRESCRIBED FUNCTION G(T). THESE ARE USED IN COMPUTING THE
C FOURIER EXPANSION COEFFICIENTS. COMPUTATION IS IN DOUBLE PRECISION.
C
C
C DOUBLE PRECISION FUNCTION IINTG(T)
C DOUBLE PRECISION GR,GI,SIG,BET,AN,T
C COMMON SIG,BET,MC,N
C AN=N
C IINTG=DEXP(-AN*SIG*T)*(-GR(T)*DSIN(AN*BET*T)+GI(T)*DCOS(AN*BET*T))
C RETURN
C END

```

\$IBFTC INTGRL

```

C
C
C THIS PROGRAM PERFORMS INTEGRATION IN DOUBLE PRECISION AND
C IS BASED ON THE 64 POINT GAUSSIAN QUADRATURE FORMULA.
C A=LOWER LIMIT OF INTEGRATION IN DOUBLE PRECISION.

```

C
C
C
C
C

B=UPPER LIMIT OF INTEGRATION IN DOUBLE PRECISION.
F=THE NAME OF THE DOUBLE PRECISION FUNCTION TO BE INTEGRATED.
ANS=THE RESULTANT INTEGRATION IN DOUBLE PRECISION.

```
SUBROUTINE INTGRL(A,B,F,ANS)
EXTERNAL F
DOUBLE PRECISION A,B,F,ANS,X(32),W(32),S1,S2,U(28),V(4),Y(28),Z(4)
DATA U/.024350292663424432509,.072993121787799039450,.121462819296
1120554470,.169644420423992818037,.217423643740007084150,.264687162
2208767416374,.31132287190210956158,.357220158337668115950,.402270
3157963991603696,.446366017253464087985,.489403145707052957479,.531
4279464019894545658,.571895646202634034284,.611155355172393250249,.
5648965471254657339858,.685236313054233242564,.71988185017161082684
69,.752819907260531896612,.783972358943341407610,.81326531512279755
79742,.840629296252580362752,.865999398154092819761,.88931544599511
84105853,.910522137078502805756,.929569172131939575821,.94641137485
98402816062,.961008799652053718919,.973326827789910963742/
DATA V/.98333625
103884625956931,.991013371476744320739,.996340116771955279347,.9993
1105041735772139457/
DATA Y/.048690957009139720383,.048575467441503426935,.048344762234
1802957170,.047999368596458307728,.047540165714830308662,.046968182
2816210017325,.046284796581314417296,.045491627927418144480,.044590
3558163756563060,.043583724529323453377,.042473515123653589007,.041
4262563242623528610,.039953741132720341387,.038550153178615629129,.
5037055128540240046040,.035472213256882383811,.03380516183714160939
62,.032057928354851553585,.030234657072402478868,.02833967261425948
73228,.026377469715054658672,.024352702568710873338,.02227017380838
83254159,.020134823153530209372,.017951715775697343085,.01572603047
96024719322,.013463047896718642598,.011168139460131128819/
DATA Z/.00884675
109826363947723,.006504457968978362856,.004147033260562467635,.0017
1183280721696432947/
DATA ISTART/+1/
S1=(B-A)/2.00
S2=(B+A)/2.00
IF(ISTART) 4,4,5
5 DO 2 I=1,28
X(I)=U(I)
2 W(I)=Y(I)
DO 3 I=1,4
X(I+28)=V(I)
3 W(I+28)=Z(I)
4 ANS=0.00
DO 1 I=1,32
1 ANS=ANS+W(I)*(F(S1*X(I)+S2)+F(-S1*X(I)+S2))
ANS=ANS*S1
ISTART=-0
RETURN
END
```

\$IBFTC ERRORR

C
C
C
C
C
C
C

THIS ROUTINE COMPUTES THE INTEGRAL SQUARE ERROR BETWEEN THE
PRESCRIBED FUNCTION AND THE APPROXIMANT GENERATED BY THE
ACCOMPANYING PROGRAMS. COMPUTATION IS IN DOUBLE PRECISION.

```
DOUBLE PRECISION FUNCTION ERRORR(X)
DOUBLE PRECISION SIG,BET,AA,BB,ESIGT,S2P,T,X,GHATR,GHATI,GR,GI
COMMON SIG,BET,MC,NNN,AA,BB,S2P,ESIGT,T,GHATR,GHATI
T=X
CALL GHAT
ERRORR=(GHATR-GR(X))**2+(GHATI-GI(X))**2
RETURN
END
```

DOUBLE PRECISION FUNCTION XR(M,T)
\$IBFTC XR

C
C
C
C
C
C
C
C
C

THIS ROUTINE COMPUTES THE REAL PART OF THE M TH ORTHONORMAL
BASIS FUNCTION X(M,T). COMPUTATION IS IN DOUBLE PRECISION.
THIS PROGRAM IS USED ONLY FOR CHECKING ORTHONORMALITY AND IS
NOT OTHERWISE USED BY THE ACCOMPANYING PROGRAMS.

```
DOUBLE PRECISION LSR(20,20),LSI(20,20),S2P,AN,SIG,BET,T,PI
,ESIGT,DE,EE,AA,BB
COMMON /LSCOM/LSR,LSI
COMMON SIG,BET,MC,NNN,AA,BB,S2P,ESIGT
DATA PI/3.141592653589793/
ESIGT=DEXP(-SIG*T)
DE=1.00
XR=0.00
DO 1 N=1,M
AN=N
DE=DE*ESIGT
EE=AN*BET*T
1 XR=XR+DE*(LSR(M,N)*DCOS(EE)-LSI(M,N)*DSIN(EE))
XR=S2P*XR
RETURN
END
```

\$IBFTC XI

C
C
C
C
C
C
C
C

THIS ROUTINE COMPUTES THE IMAGINARY PART OF THE M TH ORTHONORMAL
BASIS FUNCTION X(M,T). COMPUTATION IS IN DOUBLE PRECISION.
THIS PROGRAM IS USED ONLY FOR CHECKING ORTHONORMALITY AND IS
NOT OTHERWISE USED BY THE ACCOMPANYING PROGRAMS.

C

```

DOUBLE PRECISION FUNCTION XI(M,T)
DOUBLE PRECISION LSR(20,20),LSI(20 20),S2P,AN,SIG,BET,T,PI
.,ESIGT,DE,EE,AA,BB
COMMON /LSCOM/LSR,LSI
COMMON SIG,BET,MC,NNN,AA,BB,S2P,ESIGT
ESIGT=DEXP(-SIG*T)
DE=1.D0
XI=0.D0
DO 1 N=1,M
AN=N
EE=AN*BET*T
DE=DE*ESIGT
1 XI=XI+DE*(LSR(M,N)*DSIN(EE)+LSI(M,N)*DCOS(EE))
XI=S2P*XI
RETURN
END

```

\$IBFTC GR

C
C
C
C
C
C
C

THIS ROUTINE SUPPLIES THE REAL PART OF A SAMPLE FUNCTION G(T).
T=THE INDEPENDENT VARIABLE IN DOUBLE PRECISION.
GR=THE REAL PART OF G(T) IN DOUBLE PRECISION.

```

DOUBLE PRECISION FUNCTION GR(T)
DOUBLE PRECISION T
IF(T.LE.0.D0) GO TO 1
IF(T.LE..5D0) GO TO 2
IF(T.LE..95D0) GO TO 3
GR=DEXP(-2.42377D0*T)
GO TO 10
1 GR=0.D0
GO TO 10
2 GR=2.D0*T
GO TO 10
3 GR=-2.D0*(T-1.D0)
10 RETURN
END

```

\$IBFTC GI

C
C
C
C
C
C
C

THIS ROUTINE SUPPLIES THE IMAGINARY PART OF A SAMPLE FUNCTION G(T).
T=THE INDEPENDENT VARIABLE IN DOUBLE PRECISION.
GI=THE IMAGINARY PART OF G(T) IN DOUBLE PRECISION.

```

DOUBLE PRECISION FUNCTION GI(T)
DOUBLE PRECISION T

```

GI=0.00
RETURN
END

\$IBMAP PLOT

C THIS PLOTTING ROUTINE IS AN ADAPTATION OF SHARE UMPLT FOR THE
C IBM7044 . THE ROUTINE UTILIZES THE FORTRAN-MAP INPUT-OUTPUT
C PROGRAM FACILITY WHICH IS SUPPLIED AS A SEPARATE SUBROUTINE
* WRITES GRAPH IMAGES ON OUTPUT UNIT G

* CALLING SEQUENCES ARE

* CALL PLOT1(NSCALE,NH,SBH,NV,SBV)
* CALL PLOT2(IMAGE,XMAX,XMIN,YMAX,YMIN)
* CALL PLOT3(BCD,X,Y,NDATA)
* CALL PLOT4(NCHAR,NHABCD...)
* CALL OMIT(ARG) HJS
* CALL PLTAPE(ITAPE) HJS

* ENTRY PLOT1 *
* ENTRY PLOT2 *
* ENTRY PLOT3 *
* ENTRY PLOT4 *
* ENTRY FPLT4 G
* ENTRY OMIT *
* ENTRY PLTAPE G

* EXTERN WDATA G

*
* COL EQU 132 COLUMNS IN OUTPUT LINE (1401)
* SPACS EQU 6 SET UNUSED SPACES AT RIGHT EDGE OF PAGE
* REM OR CARD. SPACS MUST BE AT LEAST 6
G EQU COL-11-SPACS COLUMNS IN OUTPUT LINE AVAILABLE FOR IMAGE
* SPACE 5 G

* PLOT1
* REM MAIN JOB OF PLOT1 IS TO EXAMINE ARGUMENTS AND PREPARE
* REM SAMPLE GRIDLINE (DASH TO DASH-WORDS+1) AND SAMPLE
* REM NON-GRID LINE (BLANK TO BLANK-WORDS+1) FOR PLOT2

*
* PLOT1 SAVE 1,2 ENTRY TO PLOT1 G
* G
* G

CLA 3,4 SCALE FACTORS AND DECIMAL POINT POSITIONS

STA DELTA

ADD FIVE

STA DELT

STZ WRON1

WRON1 = 0 CLEAR ERROR FLAG, PLOT1

CLA ONE

STO WRON3

WRON3 = 1 SET ERROR FLAG, MISSING PLOT2

CLA* 4,4

NH, NUMBER OF HORIZONTAL GRID LINES

TSX FIX,2

TZE ERK1

ZERO ARGUMENT ILLEGAL, ERROR RETURN

STO NH

CLA* 5,4

SBH, NO. OF SPACES BETWEEN HORIZ. GRID LINES

TSX FIX,2

TZE ERK1

ZERO ARGUMENT ILLEGAL, ERROR RETURN

```

STO SBH
LDQ SBH
MPY NH
STQ LINES          LINES = NH*SBH      MAXIMUM LINE INDEX
CLA*    5,4        NV, NUMBER OF VERTICAL GRID LINES
TSX FIX,2
TZE ERR1          ZERO ARGUMENT ILLEGAL, ERROR RETURN
STO NV
CLA*    7,4        SBV, NO. OF SPACES BETWEEN VERT. GRID LINES
TSX FIX,2
TZE ERR1          ZERO ARGUMENT ILLEGAL, ERROR RETURN
STO SBV
LDQ SBV
MPY NV
STQ TOT           TOT = SBV*NV         MAXIMUM COLUMN INDEX
CLA     TOT
ADD ONE
STO TOTAL         TOTAL = TOT + 1      TOTAL COLUMNS PER LINE
SUB GWID          WHENEVER TOTAL .G. GRAPH WIDTH, ERROR RETURN
TMI PASS

*
*
*
ERR1 CAL OTAPE
CALL WDATA
PZE FORM
PZE ER1,0,1
PZE WRONG,0,3
PZE 0
CLA FPONE
STO WRON1
RETURN PLOT1

*
PASS CLA TOTAL
TSX FLOAT,2
FDP SIXF
STQ TEMP
CLA TEMP
FAD N999
TSX FIX1,2
STO WORDS         WORDS = TOTAL/6.   ROUNDED UP TO NEAREST INTEGER
LDQ WORDS         WORDS, NUMBER OF MACHINE LOCATIONS PER LINE
MPY SIX
STQ TOTLS         TOTLS = WORDS*6    BCD CHARACTERS PER LINE
LXA WORDS,2
CLA TOTLS
SUB TOTAL
PAX 0,1
CLA DSH,1         LAST WORD OF A HORIZONTAL GRID LINE
STO DASH+1,2      SET UP LAST WORD IN HORIZONTAL GRID LINE IMAGE
LDQ BLNKK         LAST WORD OF NON GRID LINE
STQ BLANK+1,2     SET UP LAST WORD IN NON GRID LINE IMAGE
TIX GA1,2,1
TRA GA2           ONE WORD PER LINE CASE

```

GA1	CLA	DSH		G
GA3	STO	DASH+1,2	SET REMAINDER OF HORIZ. GRID	G
	STQ	BLANK+1,2	SET UP REMAINDER OF NON GRID LINE IMAGE	
	TIX	GA3,2,1		G
*				G
GA2	STZ	I	COL. INDEX FOR VERTICAL GRID	G
*				G
GAMMA	TSX	PLACB,4	PUT VERTICAL GRID I IN IMAGE	G
	PZE	IEYE		
	PZE	I		
	PZE	BLANK		
*				G
	TSX	PLACB,4	PUT + AT INTERSECTIONS	G
	PZE	IPLUS		
	PZE	I		
	PZE	DASH		
*				G
	CLA	I		
	ADD	SBV		
	STO	I	I=I+SBV, INCREMENT COLUMN INDEX FOR VERT GRID	
	SUB	TOTAL		
	TZE	GAMMA	IF ZERO OR MINUS LINE IS UNFINISHED, RETURN	
	TMI	GAMMA		
*				G
DELTA	CLA **	NSCALE	NSCALE, DETERMINES SCALE FACTOR MODIFICATION	
	TZE	ETA	STANDARD SCALE FACTORS AND DEC POINT POSITIONS	
*				G
	AXT	4,4	G7= DEC POINT POSITION FOR X	
DELT	CLA	**4 NSCALE+5	G5 = SCALE FACTOR FOR X	G
	TSX	TFIX,2	G4= DEC POINT POSITION FOR Y	
	STO	G3+4,4	G3 = SCALE FACTOR FOR Y	G
	TIX	DELT,4,1		
*				G
ETA	CLA	G4		
	TZE	GA7		G
	TPL	GA7		G
	ZAC		NEG.DEC.PT. POSITION = 0	G
	TRA	GA8		G
GA7	CAS	EIGHT	MAX. DEC. PLACES, ORDINATE, =8	G
	CLA	EIGHT	IF G4 GTR THAN 8, SET G4=8	
	NOP			G
GA8	STO	G4		G
	CLA	G7		G
GA9	SUB	TEN	ABSCISSA DEC. POINT IS MOD 10	G
	TPL	GA9		G
	ADD	TEN		
	TZE	GA10		G
	TPL	GA10		G
	ZAC		NEG.DEC.PT. POSITION = 0	G
GA10	STO	G7		G
	CLA	SBV	SBV, COLUMNS AVAILABLE FOR EACH ABSCISSA VALUE	
	STO	G9	G9 = SBV FIELD WIDTH FOR ABSCISSA VALUES	G
	CAS	G7	IF G7 GTR THAN OR EQU TO G9, SET G7=G9-1	
	TRA	PASS4	ENSURES DEC POINT INSIDE FIELD	
	NOP			

	SUB ONE			
	STO G7	G9-1		G
PASS4	CLA TWELV	TWELVE SPACES ON LEFT FOR ORDINATE AND LABEL		
	ADD G7			
	STO G6	$G6 = G7 + 12$	FIELD WIDTH, LEFT ABSCISSA VALUE	
	CLA RECLT			
	SUB TOTAL			
	SUB G6			
	TPL PASS5	RECLT-TOTAL-G6.LT.0, REDUCE G6		G
	ADD G6			
	STO G6	G6 REDUCED TO NUMBER OF COLUMNS AVAILABLE		
PASS5	CLA G6	WHENEVER $G7 \geq G6$, $G7 = G6 - 1$		
	CAS G7	ENSURES DEC POINT INSIDE LEFTMOST FIELD		
	TRA EXIT			
	NOP			
	SUB ONE			
	STO G7	$G6 - 1$		G
*				G
*				G
*		SET THE FORMATS		G
*				G
EXIT	LDQ G3			HJS
	TSL BCDCON			HJS
	SLW FM1A			HJS
	CLA G3	IF G3 IS NEGATIVE SET SCALE		HJS
	TZE HJS1	FACTOR IN FM1 TO NEGATIVE		HJS
	TPL HJS1			HJS
	MSM FM1A			HJS
HJS1	LDQ G4			HJS
	TSL BCDCON			G
	SLW FM1B			G
	LDQ G5			HJS
	TSL BCDCON			HJS
	SLW FM3A			HJS
	CLA G5	IF G5 IS NEGATIVE SET SCALE		HJS
	TZE HJS2	FACTOR IN FM3 TO NEGATIVE		HJS
	TPL HJS2			HJS
	MSM FM3A			HJS
HJS2	LDQ G6			HJS
	TSL BCDCON			G
	SLW FM3B			G
	LDQ G7			G
	TSL BCDCON			G
	SLW FM3C			G
	SLW FM3F			G
	LDQ NV			G
	TSL BCDCON			G
	SLW FM3D			G
	LDQ G9			G
	TSL BCDCON			G
	SLW FM3E			G
*				G
	ZAC			G
	RETURN PLOT1	EXIT, SUCCESSFUL PLOT1		
	SPACE 5			G

* G *	PLOT2		HJS
	REM	MAIN JOB OF PLOT2 IS TO REPEATEDLY LAY DOWN SAMPLE	
	REM	GRIDLINE (DASH TO DASH-WORDS+1) FOLLOWED BY (SBH-1)	
	REM	NON-GRID LINES (BLANK TO BLANK-WORDS+1) TO FORM THE	
	REM	GRID IN THE IMAGE REGION	
			G
			G
	PLOT2	SAVE 1,2	G
*		ENTRY TO PLOT2	G
	STZ WRON3	WRON3 = 0 CLEAR ERROR FLAG, MISSING PLOT2	G
	STZ WRON2	CLEAR ERROR FLAG, PLOT2	G
	CLA WRON1		
	TNZ GA12		G
	CLA 3,4	IMAGE ADDRESS	G
	STA PLY22		G
	STA PLT23		G
	STA PLT37		
	CLA* 4,4	XMAX, MAX. ABSCISSA VALUE	G
	TSX TSTFP,2	TEST FOR FLOATING POINT ARGUMENT	
	TRA BAD		G
	STO XMAX		
	CLA* 5,4	XMIN, MIN. ABSCISSA VALUE	G
	TSX TSTFP,2	TEST FOR FLOATING POINT ARGUMENT	
	TRA BAD		G
	STO XMIN		
	CLA* 6,4	YMAX, MAX. ORDINATE VALUE	G
	TSX TSTFP,2	TEST FOR FLOATING POINT ARGUMENT	
	TRA BAD		G
	STO YMAX		
	CLA* 7,4	YMIN, MIN. ORDINATE VALUE	G
	TSX TSTFP,2	TEST FOR FLOATING POINT ARGUMENT	
	TRA BAD		G
	STO YMIN		G
*			
	CLS XMIN		
	FAD XMAX		
	TZE BAD	ERROR IF XMIN .EQ. XMAX	G
	STO SPANX	SPANX = XMAX - XMIN ABSCISSA SPAN	G
*			
	CLS YMIN		
	FAD YMAX		
	TZE BAD	ERROR IF YMIN .EQ. YMAX	G
	STO SPANY	SPANY = YMAX - YMIN ORDINATE SPAN	G
*			
	CLA LINES		
	TSX FLOAT,2		
	FDP SPANY		
	STQ U	U = LINES/SPANY NUMBER OF LINES PER UNIT Y	G
*			
	CLA TOT		
	TSX FLOAT,2		
	FDP SPANX		
	STQ V	V = TOT/SPANX NUMBER OF COLUMNS PER UNIT X	G
*			
	CLA NV		

```

TSX FLOAT,2
STO TEMP
CLA SPANX
FDP TEMP
STQ DELTX          DETLX = SPANX/NV      X INCR BETWN VERT GRID LINES
*                                                         G
CLA NH
TSX FLOAT,2
STO TEMP
CLA SPANY
FDP TEMP
STQ DELTY          DELTY = SPANY/NH      Y INCR BETWN HORZ GRID LINES
*                                                         G
STZ I              I=0  INITIALIZE WORD COUNTER FOR IMAGE REGION
STZ J              J=0  INITIALIZE LINE COUNTER FOR IMAGE REGION
*                                                         G
*                                                         G
TNEXT STZ K        K=0  INITIALIZE WORD COUNTER FOR HORZ GRID LINE
*                                                         G
LNEXT CLA K        LOOP TO PLACE ONE HORIZONTAL
PAX 0,2            GRID LINE IMAGE (DASH REGION) INTO IMAGE REGION
ADD I
PAC 0,4            G
CLA DASH,2
FLT22 STO **,*4  IMAGE
CLA K
ADD ONE
STO K              K = K+1          INCREMENT WORD COUNTER FOR LINE
SUB WORDS
TNZ LNEXT          IF NON-ZERO, LINE NOT FINISHED
*                                                         G
*                                                         G
*                                                         G
CLA I
ADD WORDS
STO I              I = I+WORDS      INCR WORD COUNTER FOR NEW LINE
*                                                         G
CLA NH
SUB J              SEE IF FINISHED
TZE GA13           WHEN J.EQ.NH, IMAGE GRID COMPLETE
*                                                         G
CLA ONE
STO TEMP          TEMP = 1  INITIALIZE BETWEEN GRID LINE COUNTER
*                                                         G
TFIN STZ K        K=0  INITIALIZE WORD COUNTER FOR EACH LINE
*                                                         G
LFIN CLA K        LOOP TO PLACE SBH NON-GRID LINE
PAX 0,2            IMAGES (BLANK REGION) INTO THE IMAGE REGION
ADD I
PAC 0,4            G
CLA BLANK,2
PLT23 STO **,*4  IMAGE
CLA K
ADD ONE
STO K              K = K+1          INCREMENT WORD COUNTER FOR LINE
SUB WORDS

```

	TNZ	LFIN	IF NON-ZERO, LINE NOT FINISHED	G
	CLA	:		G
	ADD	WORDS		
	STO	I	I = I+WORDS INCR WORD COUNTER FOR NEW LINE	
	CLA	TEMP		
	ADD	ONE		
	STO	TEMP	TEMP = TEMP+1 INCREMENT BETWN GRID LINE COUNTER	
	SUB	SBH		
	TNZ	TFIN	IF NON-ZERO, MORE LINES REQUIRED	G
	CLA	J		G
	ADD	ONE		
	STO	J	J = J+1 INCREMENT LINE COUNT FOR IMAGE REGION	
	TRA	TNEXT	RETURN FOR ANOTHER HORIZ. GRID LINE	G
				G
				G
			RETURN 2, UNSUCCESSFUL PLOT2	G
			SET ERROR FLAG, PLOT2	G
BAD	CAL	CTAPE		G
	CALL	WDATA		G
	PZE	FORM		G
	PZE	ER2,0,1		G
	PZE	WRONG,0,3		G
	PZE	0		G
				G
GA12	CLA	ONE	SET 'NO PLOT2' FLAG	G
	STO	WRON3		G
	CLA	FPTWO		G
	STO	WRON2		G
				G
				G
GA13	RETURN	PLOT2		G
	SPACE	5		HJS
	PLOT3			
	REM		PLOT3 EXAMINES THE DATA POINT TO MAKE SURE IT IS	
	REM		FLOATING POINT AND THEN PLACES I, IN THE PROPER SPOT	
	REM		IN THE IMAGE REGION	
				G
PLOT3	SAVE	1,2	PLOT3 ENTRY POINT	G
				G
	STZ	FLAG1	FLAG1 = 0 PLOT3 RETURN PRESET TO ZERO	
	CLA	3,4	ADDRESS, PLOTTING CHARACTER	G
	STA	PLT36		
	CLA	4,4	BASE ADDRESS, X COORDINATES	G
	STA	PLT35		
	CLA	5,4	BASE ADDRESS, Y COORDINATES	G
	STA	PLT34		G
				G
	CLA	WRON1		
	GRA	WRON2		
	TNZ	GA16	OUT IF BAD PLOT1 OR PLOT2	G
				G
				G
	ORA	WRON3		G

*	TZE	GA14	OUT IF PREVIOUS PLOT2	G
				G
	CALL	OTAPE	PLOT3 W/O PLOT2	G
	CALL	WDATA		G
	PZE	FORM		G
	PZE	ER3,0,3		G
	PZE	0		G
GA16	CLA	FTHRE		G
*				G
GA18	RETURN	PLOT3		G
	SPACE	5		G
*				G
GA14	CLA*	6,4	NDATA, NO. OF POINTS	G
	TSX	FIX.2		
	TNZ	GA17		G
*				G
	CLS	FTHRE	IF NDATA = 0, NO DATA POINTS. RETURN MINUS THREE	G
	TRA	GA18		G
*				G
GA17	STO	NDATA	NO. OF POINTS	G
	STZ	K	K=0 INITIALIZE DATA POINT COUNTER	G
*				G
LTEND	LAC	K,1		G
*				G
PLT34	CLS	** ,1	Y(K) Y COORDINATE OF (K+1)TH DATA POINT	
	TSX	TSTFP,2	TEST FOR FLOATING POINT ARGUMENT	
	TRA	GA21		G
	FAD	YMAX		
	LRS	35		
	FMP	U		
	TPL	GA19		G
	FSB	N05		
	TRA	GA20		G
GA19	FAD	N05		G
GA20	TSX	FIX1,2		G
	STO	I	I=(YMAX-Y(K))*U +OR- 0.5, LINE INDEX FOR DATA PT	
	TZE	PLT35	Y LIES ON TOP GRID LINE	G
	TNI	GA21	REJECT Y IF ABOVE TOP GRID LINE	G
	SUB	LINES		
	TZE	PLT35	Y LIES ON BOTTOM GRID LINE	G
	TPL	GA21	REJECT Y IF BELOW BOTTOM GRID LINE	G
*				G
PLT35	CLA	** ,1	X(K) X COORDINATE OF (K+1)TH DATA POINT	
	TSX	TSTFP,2	TEST FOR FLOATING POINT ARGUMENT	
	TRA	GA21		G
	FSB	XMIN		
	LRS	35		
	FMP	V		
	TPL	GA22		G
	FSB	N05		
	TRA	GA23		G
GA22	FAD	N05		G
GA23	TSX	FIX1,2		G
	STO	J	J=(X(K)-XMIN)*V +OR- 0.5, COL INDEX FOR DATA PT	
	TZE	GA24	X LIES ON LEFTMOST GRID LINE	G

	TNI	GA21	REJECT X IF LEFT OF GRID	G
	SUB	TOT		
	TZE	GA24	X LIES ON RIGHT GRID LINE	G
	TPL	GA21	REJECT X IF RIGHT OF GRID	G
GA24	LDQ	TOTLS		G
	MPY	I		
	LLS	35		
	ADD	J		
	STO	L	L=TOTLS*I+J, CHARACTER POSITION IN IMAGE REGION	
	TSX	PLACF,4	PLACE BCD IN L-TH	G
PLT36	PZE	** BCD		G
	PZE	L		G
PLT37	PZE	** IMAGE		G
*				G
THEND	CLA	K		
	ADD	ONE		
	STO	K	K=K+1 INCREMENT DATA POINT COUNTER	
	SUB	NDATA		
	TNZ	LTEND	IF NONZERO, MORE DATA POINTS TO BE PLOTTED	
*				G
	CLA	FLAG1	PLOT3 RETURN	
	RETURN	PLOT3		G
*				G
GA21	CLS	FTHRE	PLOT3 REJECTED POINT	G
	STO	FLAG1		G
	TRA	THEND		G
	SPACE	5		G
*PLOT4	(FPLOT4)			HJS
	REM		PLOT4 DECOMPOSES THE STRING OF CHARACTERS IN LABEL	
	REM		AND WRITES THE CURRENT GRAPH ON TAPE OTAPE	
*				G
*				G
PLOT4	SAVE	1,2	ENTRY TO PLOT4	G
*				G
*				G
			NO34 ASSUMES ALL ARRAYS ARE STORED FORWARD	HJS
FPLOT4	SYN	PLOT4	THEREFORE BOTH ENTRIES ARE THE SAME	G
*				G
	CLA	WRON1		
	ORA	WRON2		
	TNZ	GA26	OUT IF BAD PLOT1 OR PLOT2	G
*				G
*				G
	ORA	WRON3		G
	TZE	GA27	OUT IF PREVIOUS PLOT2	G
*				G
	CAL	OTAPE	UNSUCCESSFUL PLOT4	G
	CALL	WDATA		G
	PZE	FORM		G
	PZE	ER3,0,3	NO PREVIOUS PLOT2	G
	PZE	0		G
*				G
GA26	CLA	FPFOR	RETURN 4, UNSUCCESSFUL PLOT4	G
	RETURN	PLOT4		G
*				G
GA27	CLA	4,4	LABEL BASE ADDRESS	G

STMT	OPERATION	OPERANDS	DESCRIPTION	UNIT
STA	PLT41			G
STA	PLT42			G
CLA	PLT37		IMAGE ADDRESS	G
STA	ADDB			G
STA	GA49			G
LXA	WORDS,2		WORDS PER LINE	G
SXD	ADDB,2			G
SXD	GA49,2			G
CLA*	3,4		NCHAR, NO. OF CHARACTERS IN LABEL	G
TSX	FIX,2			
ADD	ONE			
PAX	0,4		NCHAR+1	G
AXT	6,1		SET CHAR. COUNT FOR LABEL WORD	HJS
CLA	YMAX		SET TOP LINE	G
STO	YAXIS		ORDINATE VALUE	G
* PLT41	LDQ	** LABEL	GET FIRST LABEL WORD	G
	STQ	LABEL		
	CLA	LINES		
	STO	FIXV	FIXV = LINES MAXIMUM HORIZONTAL LINE INDEX	
	CAL	IFONT		
	ANA	IFOUR	IF IFONT = 4,5,6, OR 7, DELETE BOTTOM GRID LINE	G
	TZE	GA29		
	CLA	LINES		
	SUB	ONE		
	STG	FIXV	FIXV = LINES-1, MAX LINE INDEX WITH NO BOTT LINE	G
* CA29	STZ	I	INITIALIZE LINE COUNT FOR IMAGE	G
	AXT	0,2	INITIALIZE WORD COUNTER FOR IMAGE REGION	HJS
* * CHECK	CLA	FIXV		
	SUB	I		
	TMI	GA30	FIXV-I NEGATIVE, IMAGE PRINT COMPLETE	G
* HJS3	TXH	HJS3,4,1	IF LABEL HAS BEEN COMPLETELY	HJS
	CAL	BLNKK	PRINTED, OR IF NO LABEL IS WANTED,	HJS
	SLW	LABEL	SET LABEL TO BLANK	HJS
	ZAC			HJS
	LDQ	I		
	DYP	SBH		
	TNZ	SKIP	IF NON-INTEGRAL, BYPASS ORDINATE PREPARATION	
	CAL	IFONT		
	ANA	TWO		
	TNZ	SKIP	IF IFONT=2,3,OR 6, DELETE ORDINATE VALUE	
* ADDB	CAL	OTAPE	GRID-LINE IMAGE	G
	CALL	WDATA		G
	PZE	FM1		G
	PZE	LABEL,0,1	LABEL CHARACTER	G
	PZE	YAXIS,0,1		G
	PZE	** ,0,**	IMAGE,0,WORDS	G
	PZE	0		G
* ADDB	CLA	YAXIS	ADJUST ORDINATE VALUE	G

	FSB	DELTY		G
	STO	YAXIS		G
GA50	CAL	ADDR	MOVE TO NEXT IMAGE LINE	G
	ADD	WORDS		G
	STA	ADDR		G
	STA	GA49		G
	CLA	I	COUNT LINES	G
	ADD	ONE		G
	STO	I		G
	TNX	CHECK,4,1	DECREMENT LABEL CHAR. COUNT	HJS
	CAL	LABEL	SET NEXT LABEL CHARACTER	G
	ALS	6		G
	TIX	GA32,1,1	DECREMENT CHARACTER COUNTER IN LABEL WORD	HJS
	AXT	6,1	REINITIALIZE CHAR. COUNT FOR LABEL WORD	HJS
	TXI	++1,2,-1	MOVE TO AND	G
PLT42	CAL	++,2 LABEL	GET NEXT LABEL WORD	G
GA32	SLW	LABEL	SAVE LABEL WORD	G
*				G
	TRA	CHECK	AROUND FOR NEXT LINE	G
*				G
*				G
SKIP	CAL	OTAPE	WRITE IMAGE, NON-GRID-LINE	G
	CALL	WDATA		G
	PZE	FM2		G
	PZE	LABEL,0,1		G
GA49	PZE	**,0,** IMAGE,0,WORDS		G
	PZE	0		G
*				G
	TRA	GA50		G
*				G
*				G
GA30	CAL	IFONT		G
	ANA	ONE		
	TNZ	EXIT2	IF IFONT=1,3,5, OR 7, DELETE ABSCISSA PRINTOUT	
*				G
	CLA	XMIN	FIRST ABSCISSA VALUE	G
	STO	ABS		G
	LXA	NV,4	FORM ABSCISSA VALUES	G
	TXI	++1,4,1	NV+1	G
	SXD	GA51,4		G
	TXI	++1,4,-1	NV	G
	AXT	0,1		G
	CLA	ABS,1		G
LS2	FAD	DELTX		G
	STO	ABS+1,1		G
	TXI	++1,1,-1		G
	TIX	LS2,4,1		G
*			PUT OUT THE ABSCISSA LINE	G
	CAL	OTAPE		
	CALL	WDATA		G
	PZE	FM3		G
GA51	PZE	ABS,0,** NV+1		G
	PZE	0		G

*	EXIT2	ZAC			G
		RETURN	PLOT4		HJS
		SPACE	5		HJS
*		OMIT			G
*					
*	OMIT	SAVE	2	DELETE PRINTOUT SEGMENTS	G
		CLA*	3,4		G
*				ARGUMENT TAKEN MOD 8, AS FOLLOWS	G
*				ARG=1, DELETE ABSCISSA VALUE PRINT	G
		TSX	FIX,2	ARG = 2, DELETE ORDINATE VALUE PRINTOUT	
		TMI	GA37	ARG=3, 1 AND 2	G
		ORA	IFONT	ARG=4, DELETE BOTTOM GRID LINE	G
		SLW	IFONT		G
		RETURN	OMIT	ARG=5, 1 AND 4	G
*					G
	GA37	COM		ARG=6, 2 AND 4	G
		ANA	IFONT	ARG=7, 1, 2, AND 4	G
		SLW	IFONT		G
		RETURN	OMIT	TO RESTORE, EXECUTE OMIT	G
		REM		WITH THE NEGATIVES OF THE ABOVE ARGUMENTS	
		SPACE	5		G
*		PLTAPE			
*					
*	PLTAPE	SAVE	2	CHANGE OUTPUT TAPE NO.	G
		CLA*	3,4		G
		TSX	FIX,2		G
		STO	OTAPE		G
		RETURN	PLTAPE		G
		SPACE	5		G
*		BCDCON			
*					
*				CONVERT INTEGER IN MQ	G
*				TO BCD (999999 MAX.)	G
*				IN LOGICAL AC.	G
*					G
*					G
*	BCDCON	AXT	**0		G
		AXT	6,4		G
		ZAC			G
	BCD1	SLW	BCD2		G
		ZAC			G
		DVP	TEN		G
		XEC	BCD2,4		G
		GRA	BCD2		G
		TIX	BCD1,4,1		G
		TRA*	BCDCON		G
		NOP			G
		ALS	6		G
		ALS	12		G
		ALS	18		G
		ALS	24		G
		ALS	30		G
	BCD2	DEC	0		G
*					

*	SPACE	5		G
*	TSTFP			G
*				
TSTFP	TZE	2,2	CHECK FOR FLOATING POINT	G
	STO TEMP		ITS CALLING SEQUENCE IS	
	SSP		TSX TSTFP,2	
	SUB MS004		THE ARGUMENT IS IN THE ACCUMULATOR	
	TMI	GA11		G
	CLA TEMP			
	TRA	2,2	FLOATING-POINT RETURN	G
GA11	CLA	TEMP		G
	TRA	1,2	NON-FLOATING-POINT RETURN	G
	SPACE	5		G
*	FLOAT			
*				
FLOAT	ORA CONST		FLOAT FLOATS A NUMBER KNOWN TO BE AN INTEGER	
	FAD CONST		THE CALLING SEQUENCE IS	
	TRA 1,2		TSX FLOAT,2 WITH ARGUMENT IN ACCUMULATOR	
	SPACE	5		G
*	FIX			
*				
*			CONVERT ARGUMENT TO FULL-WORD INTEGER	
FIX	SSP		POSITIVE RESULT	G
TFIX	LRS	26	SIGNED RESULT	G
	TNZ	FIX2	IF NON-ZERO, CONSIDERED FLOATING	G
	LLS	26	ALREADY FIXED	G
	TRA	1,2		
FIX2	LLS	26	RESTORE FLOATING NUMBER	G
FIX1	UFA CONST		FIXES A NUMBER KNOWN TO BE IN FLOATING POINT	
	LRS 27			
	ZAC			G
	LLS 27			
	TRA 1,2			
	SPACE	5		G
*	PLACF, PLACB			
*				
*			PLACE BCD CHARACTER IN	G
*			I-TH CHARACTER POSITION OF	G
*			A SPECIFIED REGION	G
*			TSX PLACE,4	G
*			PZE BCD	G
*			PZE I	G
*			PZE REGION	G
PLACF	MSM	GA36	-GA36, FORWARD ARRAYS	G
	TRA	GA35		G
*				G
PLACB	MSP	GA36	+GA36, BACKWARD ARRAYS	G
*				G
GA35	LDQ*	2,4	I	G
	STQ	TEM		G
	ZAC			G
	DVP SIX			
	STQ TEM1		TEM1=1/6, INDEX OF WORD CONTAINING CHAR POSITION	
	LAC	TEM1,2	SET FOR FORWARD ARRAYS	G

MIT	GA36	TEST IF FORWARD ARRAY	G
LXA	TEM1,2	NO, SET FOR BACKWARD ARRAYS	G
MPY	SIX		G
STQ	TEM1	TEM1=TEM1*6, CONTAINS TOTAL CHARACTER POSITIONS	
CLA	TEM	TEM, CONTAINS THE CHARACTER POSITION	
SUB	TEM1		
PAX	0,1	TEM-TEM1 = CHARACTER POSITION IN WORD	
CLA	3,4		
STA	GA4		G
STA	GA5		G
CAL	PLCMK,1		
GA4	ANA **2	ZERO THE CHARACTER POSITION	G
	SLW* GA4		G
	CAL* 1,4	GET AND	G
	ANA MASK	ISOLATE THE CHARACTER	G
	TNX GA5,1,0		G
GA6	ARS 6	SHIFT IT INTO POSITION	G
	TIX GA6,1,1		G
GA5	ORA **2	PUT CHARACTER INTO WORD	G
	SLW* GA5		G
	TRA 4,4	RETURN	
*			G
GA36	PZE 0	--=FORWARD ARRAY, +=BACKWARD ARRAY	G
	SPACE 5		G
ABS	BSS COL/2	ABSCISSA VALUES	G
	BCI 3,	I	
	BCI 1,1	SAMPLE LINE IMAGE	G
	BCI 1,	FOR NON-GRID LINES.	
	BCI 1, I	THIS IS LINE IMAGE USED IN STANDARD GRID.	
	BCI 1,	EXECUTION OF PLOT1 SETS UP NEW VALUES.	
	BCI 1, I		
	BCI 1,1		
	BCI 1,		
	BCI 1, I		
	BCI 1,		
	BCI 1, I		
	BCI 1,		
	BCI 1, I		
	BCI 1,		
	BCI 1, I		
	BCI 1,		
	BCI 1, I		
BLANK	BCI 1,1	FIRST WORD OF LINE IMAGE FOR NON-GRID LINES	
BLNKK	BCI 1,		
CONST	OCT 233000000000		
	BCI 3,	-----+	
	BCI 1,+-----	SAMPLE LINE IMAGE	G
	BCI 1,-----	HORIZONTAL GRID LINES.	
	BCI 1,---+---	THIS IS THE LINE IMAGE USED IN STANDARD GRID.	
	BCI 1,-----	EXECUTION OF PLOT1 PRODUCES NEW VALUES.	
	BCI 1,---+---		
	BCI 1,+-----		
	BCI 1,-----		
	BCI 1,---+---		
	BCI 1,-----		
	BCI 1,---+---		

	BCI	1,+-----			
	BCI	1,-----			
	BCI	1,--+-			
	BCI	1,-----			
	BCI	1,-----+			
DASH	BCI	1,+-----	1ST WORD OF LINE IMAGE FOR HORIZ GRID LINES		
DELTX	DEC	0.	INCR. BETWEEN VERTICAL LINES		G
DELY	DEC	0.	INCR. BETWEEN HORIZ. LINES		G
	BCI	1,-	DSH TO DSH-5 USED BY PLOT1 TO FILL OUT LAST		
	BCI	1,--	WORD OF HORIZ LINE IMAGE		
	BCI	1,---			
	BCI	1,----			
	BCI	1,-----			
DSH	BCI	1,-----			
EIGHT	DEC	8			
ER1	BCI	1,0PLOT1			G
ER2	BCI	1,0PLOT2			
ER3	BCI	3,0NO PREVIOUS PLOT2			
FIVE	DEC	5			
FIXV			MAXIMUM HORIZONTAL LINE INDEX FOR PRINTING		
FLAG1					
*					
*			FORMATS FOR PRINTING THE IMAGE		G
*					G
FM1	BCI	1,(1XA1,	GRID LINES		G
FM1A	BCI	1, 0	G3		G
	BCI	1,PF9.			G
FM1B	BCI	1, 3	G4		G
	BCI	2,,1X20A6)			G
*					G
*			NON-GRID LINES		G
FM2	BCI	3,(1XA1,10X20A6)			G
*					G
*			ABSCISSA LINE		G
FM3	BCI	1,(1HO			G
FM3A	BCI	1, 0	G5		G
	BCI	1,PF			G
FM3B	BCI	1, 15	G6		G
	BCI	1,,			G
FM3C	BCI	1, 3	G7		G
	BCI	1,,			G
FM3D	BCI	1, 10	NV		G
	BCI	1,F			G
FM3E	BCI	1, 10	G9		G
	BCI	1,,			G
FM3F	BCI	1, 3	G7		G
	BCI	1,)			G
FORM	BCI	1,(22A6)			G
FPONE	DEC	1.			
FPTWO	DEC	2.			
FTHRE	DEC	3.			
FPPOR	DEC	4.			
	SPACE	5			
*			FORMAT PARAMETERS		G
*			MODIFIABLE BY PLOT1		G

[illegible]

```

V      DEC      0.          LINES PER UNIT X
WORDS  DEC 17      *** NUMBER OF MACHINE LOCATIONS PER LINE
WRONG  BCI      3, IMPROPER ARGUMENT
WRON1  DEC 0      *** EQUALS 1 FOR UNSUCCESSFUL PLOT1
WRON2          EQUALS 1 FOR UNSUCCESSFUL PLOT2
WRON3  DEC 1      *** EQUALS 1 UNTIL SUCCESSFUL PLOT2
YAXIS
XMAX
XMIN
YMAX
YMIN

```

END THIS IS THE LAST CARD

\$IBMAP IO

C THIS PROGRAM IS USED IN CONJUNCTION WITH THE PLOTTING ROUTINE
C TO ALLOW FORTRAN INPUT-OUTPUT FACILITY IN A MACHINE LANGUAGE
C PROGRAM

-----RAND W038 I/O ROUTINE FOR MAP USERS
WHO WISH TO REFER TO THE STD FN4 I/O PACKAGE.

THIS ROUTINE IS IDENTICAL IN FUNCTION
TO THE 7090 ROUTINE X022. THIS 7044 VERSION IS
OFFERED COURTESY OF J D BABCOCK WITH THE BLESSINGS
G W ARMERDING WHO WAS RESPONSIBLE FOR THIS MESS
ON THE 7090.

CALLS ARE---

```

CAL      L (LOGICAL TPAE NO. IN DECREMENT)
CALL      (ROUTINE ENTRY)
PZE      FMT          (BCD ONLY)
CP       A1,T1,N1
OP       A2,T2,N2
.        . . .
.        . . .
PZE      0          - LAST MUST BE ZERO.
          (RETURN)

```

FMT IS LOCATION OF A STANDARD FN4 FORMAT STATEMENT
A1,T1 IS THE ADDR. OF FIRST DATA WORD (T=0,1,2)
N1 IS NUMBER OF WORDS (A(1)A-N+1)
OP IS PZE FOR DIRECT ADDRESSING
OP IS MZE FOR INDIRECT ADDRESSING

ROUTINE ENTRIES ARE---

- (1) RDATA --- BCD INPUT
- (2) IN ---- BCD INPUT FOR STD INPUT UNIT
(CAL L NOT REQUIRED)
- (3) WDATA BCD OUTPUT
- (4) OUT BCD OUTPUT FOR STD OUTPUT UNIT
(CAL L NOT REQUIRED)
- (5) PUNCH BCD OUTPUT FOR STD PUNCH UNIT
(CAL L NOT REQUIRED)

(6) WBIN BINARY OUTPUT
(7) RBIN BINARY INPUT

*			
*			
*			
*			
	ENTRY	WBIN	
	ENTRY	RBIN	
	ENTRY	IN	
	ENTRY	RDATA	
	ENTRY	OUT	
	ENTRY	WDATA	
	ENTRY	PUNCH	
*			
WBIN	AXT	0,0	
	SXA	IR4,4	
	AXT	3,4	
	TRA	D1	
*			
RBIN	AXT	0,0	
	SXA	IR4,4	
	AXT	2,4	
	TRA	D1	
*			
PUNCH	AXT	0,0	ALLON CALLS OF
	CAL	=7	TSL
	TRA	WDATA	OR
*			TSX VARIETY
IN	AXT	0,0	
	CAL	=5	STD INPUT UNIT
RDATA	AXT	0,0	
	SXA	IR4,4	
	AXT	0,4	
	TRA	D1	
*			
OUT	AXT	0,0	
	CAL	=6	STD OUTPUT UNIT
WDATA	AXT	0,0	
	SXA	IR4,4	
	AXT	1,4	
D1	SLW	DT	SAVE LOGICAL TAPE NO.
	SXA	DR4,4	
	AXT	7,4	
	CAL*	ENTRY+7,4	TEST FOR TSX OR TSL ENTRY
	ANA	ADT	
	TNZ	D7	
	TIX	*-3,4,1	
	MSP	ENTRY	
DR4	AXT	** ,4	WAS A TRUE TSX ENTRY
	TRA	D6	
D7	PAC	0,4	FIX
	TXI	*+1,4,-1	UP TRY TO SIMULATE
	CLA	IR4	TSX ENTRY
	STA	TSLR4	SAVE PROG. IR4
	SXA	IR4,4	
	LXA	DR4,4	
	MSM	ENTRY	

D6	CLA	SEL,4	TAPE
	STO	DSEL	SET UP TSH, STH
	CLA	END,4	AND FIL, RTN
	STO	DFND	
	CLA	CV1,4	
	STO	DT1	
	CLA	DV2,4	
	STO	DT2	
	YXH	DBIN,4,1	TEST IF BCD OR BIN CALL
	CLA	HNL	WAS BCD
	STO	CNVT	SET CONVERT CALL
	LXA	IR4,4	
	CAL	1,4	SET UP FORMAT
	TXI	++1,4,-1	BUMP FOR FIRST DATA CALL
	SXA	IR4,4	
	ANA	ADT	
	ORA	BCDFM	
	TRA	GO	
DBIN	CAL	NOP	FOR BINARY NOP 2,4
	LDQ	BNL	
	STQ	CNVT	
GO	SLW	DFMT	
	CAL	DT	CALL TO SET UP
	TSX	UTVAR.,4	FILE NAME
*		GIVE INITIAL CALL	
DSEL	***	**	TSX TSHIO,4 OR TSX STHIO,4
*			OR TSBIO.,STBIO.
DFILE	PZE	**	FIL XX.
DFMT	PZE	**,0,**	FORMAT FROM CALL (BCD=FORMAT,,BIN=NO
*			
*		NOW SET UP BASIC LOOP--	
	LXA	IR4,4	PREPARE TO PULL OUT ARGUMENTS
D2	SXA	IR4,4	
	CLA	1,4	
	TZE	DEND	CHECK IF DONE
	PDC	0,4	COUNT
	TXL	IR4,4,0	OUT IF NONE
	SXD	DTST,4	SET TEST FOR NO. OF WORDS
	TPL	D3	CHECK IF INDIRECT
	ANA	ADT	YES,, KILL DECREMENT + PRFX.
	ORA	DCLA	SET UP CLA
	SLW	++1	
	***	**	GETS CLA A,T IF INDIRECT
D3	STA	D5	SAVE A
	ANA	TAG	PICK UP
	ORA	SXD4	PROGRAMMER INDEX REGISTER
	SLW	++1	AND PUT THE COMP. OF IT
	***	**	
	LDC	D4,4	INTO IR4
	SXD	D4,4	
D5	AXT	**,4	COMPUTE ADDR OF A-T
D4	TXI	++1,4,**	
	SXA	DT1,4	SET ADDR.S OF PUTS
	SXA	DT2,4	AND GETS
	AXT	0,4	BEGIN LOOP

DT1	***		NOP OR CLA DATA, T
CNVT	***	**	T=0(1)-N (TSL HNLIO. OR BNLIO.
DT2	***		NOP OR STO DATA, T
*			NOP= AXT 0,0
	TXI	*+1,4,-1	
DTST	TXH	DT1,4,**	-N IN DECR
IR4	AXT	** ,4	
	TXI	02,4,-1	
*			
DEND	***	**	FINAL EXIT, TSX FILIO., 4
*			OR TSX RTNIO.,4
*			RESTORE AXT 0,0 ENTRIES
	AXT	7,4	
	PXA	0,0	
	STA*	ENTRY+7,4	
	TIX	*-1,4,1	
	LXA	IR4,4	
*			
	MIT	ENTRY	WAS IT TSX OR TSL
	TRA	DEND1	WAS TSX OK-EXIT VIA 2,4
	TXI	*+1,4,-2	WAS TSL, CALCULATE RETURN TRA
	PXA	0,4	
	PAC	0,4	ADDR
	SXA	DEND1-1,4	
	LXA	TSLR4,4	RESTORE PROG. IR4
	TRA	**	RETURN TO MAIN PROGRAM
DEND1	TRA	2,4	
*			
BCDFM	MZE	** , , FMTSC.	
BNL	TSL	BNLIO.	
HNL	TSL	HNLIO.	
DT	PZE		LOGICAL TAPE
ADT	OCT	777777	SAVE ADDR AND TAG
DCLA	CLA	** ,0	CLA ORDER
*			TABLE OF INITIAL CALLS
*			
	TSX	STBIO.,4	WBIN=IR4 OF 3
	TSX	TSBIO.,4	RBIN=IR4 OF 2
	TSX	STHIO.,4	WDATA = IR4 OF 1
SEL	TSX	TSHIO.,4	RDATA = IR4 OF 0
*			
	TSX	WLRIO.,4	WBUN
	TSX	RLRIO.,4	RBUN
	TSX	FILIO.,4	WDATA
END	TSX	RTNIO.,4	RDATA
*			SET UP HNLIO. ,BNLIO. LOOPS
	CLA	** ,4	3
	AXT	0,0	2
	CLA	** ,4	1
DV1	AXT	0,0	0
*			
	AXT	0,0	3
	STO	** ,4	2
	AXT	0,0	1
DV2	STO	** ,4	0

```

*
+
ENTRY  PZE  PUNCH
        PZE  WDATA
        PZE  OUT
        PZE  RDATA
        PZE  IN
        PZE  WBIN
        PZE  RBIN
*
TSLR4  PZE  **
TAG    OCT  000000700000
SXD4   SXD  D4,0
NOP    EQU  OV1

```

```

*
        EXTERN  UTVAR.
        EXTERN  TSBIO.
        EXTERN  STBIO.
        EXTERN  BNLIO.
        EXTERN  TSHIO.
        EXTERN  STHIO.
        EXTERN  HNLIO.
        EXTERN  FMTSC.
        EXTERN  RTNIO.
        EXTERN  FILIO.
        EXTERN  WLRIO.
        EXTERN  RLRIO.
        END

```

```

$IBMAP  UTV
        ENTRY  UTVAR.
        EXTERN ERLOC.
UTVAR.  SXA    UTVX,4      SAVE RETURN INDEX
        LAC    UTVX,4
        SXA    ERLOC.,4
        LXA    UTVX,4
        LAS    NFILES
        TRA    USTOP
        NOP
        PAC    ,4
        CLA    IOU,4      PICKUP ADDRESS OF FCB POINTER
        PAX    ,4
        TXL    USTOP-2,4,0
UTVX    AXT    **,4      STOP IF UNIT IS UNDEFINED
        STO    2,4      RESTORE RETURN INDEX
        TRA    1,4      SET LOCATION OF FCB
        LXA    UTVX,4    RETURN TO MAIN PROGRAM
        CLA*   -1,4
USTOP   TSL    FESEM.     RESTORE UNIT DESIGNATION
        PZE    EXIT,,32   ERROR, ILLEGAL UNIT REQUESTED.
                                NO OPTIGNAL RETURN
*INPUT-OUTPUT LOGICAL UNIT TABLE
*ADDITIONS OR DELETIONS SHOULD BE MADE BETWEEN IOU AND NFILES
IOU     PZE    FILOO.

```

```

PZE          FIL01.
PZE          FIL02.
PZE          FIL03.
PZE          FIL04.
PZE          FIL05.
PZE          FIL06.
PZE          FIL07.
PZE          FIL08.
PZE          FIL09.
PZE          FIL10.
PZE          FIL11.
PZE          FIL12.
PZE          FIL13.
PZE          FIL14.
PZE          FIL15.
NFILS PZE    *-IOU-1
          EXTERN FIL00.
          EXTERN FIL01.
          EXTERN FIL02.
          EXTERN FIL03.
          EXTERN FIL04.
          EXTERN FIL05.
          EXTERN FIL06.
          EXTERN FIL07.
          EXTERN FIL08.
          EXTERN FIL09.
          EXTERN FIL10.
          EXTERN FIL11.
          EXTERN FIL12.
          EXTERN FIL13.
          EXTERN FIL14.
          EXTERN FIL15.
          EXTERN FEXEM.,EXIT
          END

```

REFERENCES

1. Liechenstein, M. I., *Orthogonal Bases for Exponential and Rational Function Approximations of Network and Signal Characteristics*, The RAND Corporation, RM-5177-PR, January 1967.
2. Wiener, N., and R. E. A. C. Paley, "Fourier Transforms in the Complex Domain," *Am. Math. Soc. Colloq. Pubs.*, Vol. 19, New York, 1934, p. 26.
3. Sokolnikoff, I. S., and R. M. Redheffer, *Mathematics of Physics and Modern Engineering*, McGraw-Hill Book Co., Inc., New York, 1958, pp. 116-120.
4. Wiener, N., *The Fourier Integral and Certain of Its Applications*, Cambridge University Press, Cambridge, England, 1933, pp. 60-70.
5. Davis, P. J., *Interpolation and Approximation*, Blaisdell Publishing Co., New York, 1963, pp. 165-166.
6. Whittaker, E. T., and G. N. Watson, *A Course of Modern Analysis*, Cambridge University Press, Cambridge, England, 1962, pp. 115-116.
7. Kaplan, W., *Operational Methods for Linear Systems*, Addison-Wesley Publishing Co., Inc., Palo Alto, California, 1962, pp. 125-131.
8. Abramowitz, M., and I. A. Stegun (eds.), *Handbook of Mathematical Functions*, National Bureau of Standards Applied Mathematics Series, AMS 55, Washington, D. C., 1964, pp. 824-825.
9. Burnside, W. S., and A. W. Panton, *The Theory of Equations*, Vol. 1, Dover Publications, Inc., New York, 1960, pp. 22-23, 259-261.
10. Hamming, R. W., *Numerical Methods for Scientists and Engineers*, McGraw-Hill Book Co., Inc., New York, 1962, p. 8.
11. Stewart, B. M., *Theory of Numbers*, The Macmillan Company, New York, 1952, p. 19.
12. Hadley, G., *Nonlinear and Dynamic Programming*, Addison-Wesley Publishing Co., Inc., Palo Alto, California, 1964, pp. 53-58.
13. Leitmann, G. (ed.), *Optimization Techniques*, Academic Press, New York, 1962, p. 6.
14. Bellman, R., R. E. Kalaba, and J. Lockett, *Numerical Inversion of the Laplace Transform*, American Elsevier Publishing Co., Inc., New York, 1966, Chap. 1.
15. Lanczos, C., *Applied Analysis*, Prentice Hall, Inc., Englewood Cliffs, New Jersey, 1961, pp. 280-303.

16. Macon, N., and A. Spitzbart, "Inverses of Vandermonde Matrices," *Am. Math. Monthly*, Vol. 65, February 1958, pp. 95-100.
17. Laning, J. H., and R. H. Battin, *Random Processes in Automatic Control*, McGraw-Hill Book Co., Inc., New York, 1956, pp. 422-423.
18. Klinger, A., "The Vandermonde Matrix," *Am. Math. Monthly*, Vol. 74, No. 5, May 1967, pp. 571-574.
19. Liechenstein, M. I., *Evaluation of the Determinantal Polynomial of a Matrix of Polynomial Elements and an Application to AMTI Radar*, The RAND Corporation, RM-483 PR, January 1966, pp. 11-15, 43-44, 83-120.
20. Hildebrand, F. B., *Introduction to Numerical Analysis*, McGraw-Hill Book Co., Inc., New York, 1956, pp. 378-382.
21. Bellman, R., *On the Separation of Exponentials*, The RAND Corporation, P-1846, December 1959.
22. Gardner, D. G., J. C. Gardner, G. Laush, and W. W. Meinke, "Method for the Analysis of Multicomponent Exponential Decay Curves," *J. Chem. Phys.*, Vol. 31, No. 4, October 1959, pp. 978-986.
23. Wynn, P., "A Note on the Fitting of Certain Types of Experimental Data," *Statistica Neerlandica*, Vol. 16, 1962, pp. 143-150.
24. Cornell, R. G., *A New Estimation Procedure for Linear Combinations of Exponentials*, Oak Ridge National Laboratory, Report ORNL-2120, September 1956.
25. Slepian, D., "Estimation of Signal Parameters in the Presence of Noise," *IRE Trans. Info. Theory*, Vol. IT-3, March 1954, pp. 68-89.
26. Bellman, R. E., and R. E. Kalaba, *Quasilinearization and Nonlinear Boundary-Value Problems*, American Elsevier Publishing Co., Inc., New York, 1965, pp. 139-150.
27. Hudson, G. E., "Trend Analysis of Physical Data," *Am. J. Phys.*, Vol. 21, No. 5, May 1953, pp. 362-367.
28. Litman, S., and W. H. Huggins, "Growing Exponentials as a Probing Signal for System Identification," *Proc. IEEE*, Vol. 51, No. 6, June 1963, pp. 917-923.
29. Worsley, B. H., "Analysis of Decay-Type Data," *Comm. ACM*, Vol. 7, No. 1, January 1964, pp. 39-44.
30. Liechenstein, M. I., "Rational Function Approximations for Arbitrary Envelope Delay," Ph.D. dissertation, Yale University, June 1965, Chap. 2.

31. Prony, R., "Essai expérimental et analytique," *Journal de l'école polytechnique*, Paris, Vol. 1, No. 2, 1795, pp. 24-35.
32. Householder, A. S., *On Prony's Method of Fitting Exponential Decay Curves and Multiple-Hit Survival Curves*, Oak Ridge National Laboratory, Report ORNL-455, September 1949.
33. Kautz, W. H., "Transient Synthesis in the Time Domain," *Trans. IRE*, Vol. CT-1, No. 3, September 1964, pp. 29-39.
34. Mathers, G. W. C., *The Synthesis of Lumped-Element Circuits for Optimum Transient Response*, Electronics Research Laboratory, Stanford University, Tech. Rept. 38, November 1951.
35. Wiener, N., *Extrapolation, Interpolation, and Smoothing of Stationary Time Series*, John Wiley and Sons, Inc., New York, 1960, Chaps. 2,3.
36. Bendat, J. S., *Principles and Applications of Random Noise Theory*, John Wiley and Sons, Inc., New York, 1958, Section 7.3.
37. Harman, W. W., *Principles of the Statistical Theory of Communication*, McGraw-Hill Book Co., Inc., New York, 1963, pp. 43-48.
38. Brown, W. S., "A Language and System for Symbolic Algebra," in *System Analysis by Digital Computer*, F. F. Kuo and J. F. Kaiser (eds.), John Wiley and Sons, Inc., New York, 1966, pp. 349-373.

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10. ABSTRACT A computationally efficient scheme for approximating system characteristics by sums of exponentials or by rational functions. Results are applicable to a broad class of system optimizations, such as those in network synthesis and radar filter design, and to the understanding of cross-correlation measurements, radioactive decay, gas absorption, and mass spectrograph and ultracentrifuge analysis curves. Two sets of two-parameter orthonormal elements are derived: one set constitutes a basis for exponential approximation and the other a basis for rational function approximation. The closure properties of the two orthonormal bases are examined and new expressions are developed for efficiently determining the orthonormal elements of each basis. Several relations are then deduced that connect important properties of each basis. Useful identities and numerical techniques involving the basis coefficients are derived that obviate storage of either the form or selected values of the orthonormal approximants. Finally, several numerical examples illustrate algorithms for both exponential and rational function approximation. Computer programs in ALTRAN and in the more efficient FORTRAN IV are appended.		11. KEY WORDS Curve fitting Numerical methods and processes Radar Radio Communication systems Signal processing	