UNCLASSIFIED

AD NUMBER

AD814687

LIMITATION CHANGES

TO:

Approved for public release; distribution is unlimited. Document partially illegible.

FROM:

Distribution authorized to U.S. Gov't. agencies and their contractors; Critical Technology; APR 1967. Other requests shall be referred to Air Force Technical Applications Center, Washington, DC. Document partially illegible. This document contains export-controlled technical data.

AUTHORITY

usaf ltr, 28 feb 1972

THIS PAGE IS UNCLASSIFIED

~ 8 GEIGENVALUES AND EIGENVECTORS OF SPECTRAL DENSITY MATRICES

7 April 1967

Prepared For

AIR FORCE TECHNICAL APPLICATIONS CENTER Washington, D. C.

By

N. R. Goodman MEASUREMENT ANALYSIS CORPORATION

Under

Project VELA UNIFORM

Sponsored By

ADVANCED RESEARCH PROJECTS AGENCY Nuclear Test Detection Office ARPA OTONE NO. 624

EIGENVALUES AND EIGENVECTORS OF SPECTRAL DENSITY MATRICES

SEISMIC DATA LABORATORY REPORT NO. 179

AFTAC Project No.: Project Title: ARPA Order No.: ARPA Program Code No.:

Name of Contractor:

Contract No.: Date of Contract: Amount of Contract: Contract Expiration Date: Project Manager: VELA T/6702 Seismic Data Laboratory 624 5810 TELEDYNE, INC. F 33657-67-C-1313 3 March 1967 \$ 1,735,617 2 March 1968 William C. Dean (703) 836-7644

P. O. Box 334, Alexandria, Virginia

AVAILABILITY

This document is subject to special export controls and each transmittal to foreign governments or foreign national may be made only with prior approval of Chief, AFTAC. This research was supported by the Advanced Research Projects Agency, Nuclear Test Detection Office, under Project VELA-UNIFORM and accomplished under the technical direction of the Air Force Technical Applications Center under Contract F 33657-67-C-1313.

Neither the Advanced Research Projects Agency nor the Air Force Technical Applications Center will be responsible for information contained herein which may have been supplied by other organizations or contractors, and this document is subject to later revision as may be necessary.

ABSTRACT

This report describes some interpretations and uses of eigenvalues and eigenvectors of spectral and sample spectral density matrices of multiple stationary time series.

The spectral density matrix of a zero-mean multiple stationary time series is defined. Eigenvalues and eigenvectors of the spectral density matrix are discussed and principal component theory is presented. Statistical distribution theory and related results are used to investigate the eigenvalues of a sample spectral density matrix. This investigation gives methods for obtaining simultaneous confidence bounds on the elements of the true spectral density matrix and its inverse, and also methods for obtaining confidence bounds on the eigenvalues of the true spectral density matrix.

CONTENTS

1.	Spectral Representation of a Zero-Mean	
	Multiple Stationary Time Series	1
2.	Eigenvalues and Eigenvectors of a	
	Spectral Density Matrix	2
3.	Principal Components	5
4.	Statistical Estimation of Eigenvalues and Eigenvectors of a Spectral Density Matrix	10
5.	Statistical Distribution Theory and Related Results	
	Pertaining to the Random Eigenvalues of a	
	Sample Spectral Density Matrix	11

EIGENVALUES AND EIGENVECTORS OF SPECTRAL DENSITY MATRICES

The present report describes some interpretations and uses of eigenvalues and eigenvectors of spectral and sample spectral density matrices of multiple stationary time series.

1. SPECTRAL REPRESENTATION OF A ZERO-MEAN MULTIPLE STATIONARY TIME SERIES

A zero-mean multiple stationary time series $\overline{X}(t)$ has the spectral representation

$$X(t) = \begin{bmatrix} X_{1}(t) \\ X_{2}(t) \\ \vdots \\ \vdots \\ \vdots \\ X_{p}(t) \end{bmatrix} = \int_{-\infty}^{\infty} e^{i\omega t} \begin{bmatrix} dZ_{1}(\omega) \\ dZ_{2}(\omega) \\ \vdots \\ \vdots \\ dZ_{p}(\omega) \end{bmatrix}$$

(1)

One has

$$\mathbf{E}\begin{bmatrix} dZ_{1}(\omega) \\ \vdots \\ dZ_{p}(\omega) \end{bmatrix} \begin{bmatrix} \overline{dZ_{1}}(\omega), \dots, \overline{dZ_{p}}(\omega) \end{bmatrix} = \Sigma(\omega) d\omega \qquad (2)$$

where the p x p Hermitian nonnegative definite matrix $\Sigma(\omega)$ is the spectral density matrix at frequency ω of $\overline{X}(t)$.

2. EIGENVALUES AND EIGENVECTORS OF A SPECTRAL DENSITY MATRIX

There exists a $p \ge p$ unitary matrix $\beta(\omega)$, i.e.

$$\vec{\beta}(\omega) \ \beta(\omega) = I = \beta(\omega) \ \vec{\beta}'(\omega)$$
 (3)

such that

where the $\lambda_j(\omega)$, (j = 1, ..., p), are the eigenvalues (real and nonnegative) of $\Sigma(\omega)$ and

$$\lambda_{1}(\omega) \geq \lambda_{2}(\omega) \geq \ldots \geq \lambda_{p}(\omega) \geq 0$$
 (5)

Consider

$$\begin{bmatrix} dW_{1}(\omega) \\ \vdots \\ \vdots \\ \vdots \\ dW_{p}(\omega) \end{bmatrix} \equiv \overline{\beta}'(\omega) \begin{bmatrix} dZ_{1}(\omega) \\ \vdots \\ dZ_{1}(\omega) \\ \vdots \\ \vdots \\ dZ_{p}(\omega) \end{bmatrix}$$
(6)

one has

$$\mathbf{E}\begin{bmatrix} dW_{1}(\omega) \\ \vdots \\ \vdots \\ \vdots \\ dW_{p}(\omega) \end{bmatrix} \begin{bmatrix} d\overline{W}_{1}(\omega), \ldots, d\overline{W}_{p}(\omega) \\ \vdots \\ \vdots \\ dW_{p}(\omega) \end{bmatrix} = \mathbf{E} \overline{\beta}^{\dagger}(\omega) \begin{bmatrix} dZ_{1}(\omega) \\ \vdots \\ \vdots \\ dZ_{p}(\omega) \end{bmatrix} \begin{bmatrix} d\overline{Z}_{1}(\omega), \ldots, d\overline{Z}_{p}(\omega) \\ \vdots \\ dZ_{p}(\omega) \end{bmatrix}$$

(7)

$$= \overline{\beta}^{1}(\omega) \Sigma(\omega) \beta(\omega) d\omega = \begin{bmatrix} \lambda_{1}(\omega) & & \\ & \ddots & \\ & & \\ 0 & & \\ & & \lambda_{p}(\omega) \end{bmatrix} d\omega = \bigwedge_{\omega} (\omega) d\omega$$

From (7) one observes that the complex random variables $dW_1(\omega), \ldots, dW_p(\omega)$ are uncorrelated, and that

$$\operatorname{Var} \left(\mathrm{dW}_{j}(\omega) \right) = \mathbb{E} \left| \mathrm{dW}_{j}(\omega) \right|^{2} = \lambda_{j}(\omega) \, \mathrm{d}\omega, \quad (j = 1, \ldots, p) \tag{8}$$

Let $\beta_j(\omega)$, (j = 1, ..., p), denote the jth column vector of the unitary matrix $\beta(\omega)$. Since $\beta(\omega)$ is unitary

$$\overline{\beta}_{j}^{\prime}(\omega) \beta_{j}^{\prime}(\omega) = \begin{cases} 0 \text{ if } j \neq j^{\prime} \\ 1 \text{ if } j = j^{\prime} \\ \end{cases}, \quad (j = 1, \ldots, p)$$
(9)

Equation (9) expresses that the column vectors $\beta_j(\omega)$ of $\beta(\omega)$ are orthonormal, i.e., they are orthogonal and of unit norm. From (4) one has

$$\Sigma(\omega) \ \beta(\omega) = \beta(\omega) \ \Lambda(\omega) \tag{10}$$

i.e.

$$\Sigma(\omega) \beta_{i}(\omega) = \lambda_{i}(\omega) \beta_{i}(\omega) , \quad (j = 1, ..., p)$$
(11)

so that the column vectors $\beta_j(\omega)$ of $\beta(\omega)$ are the (normalized) eigenvectors of the spectral density matrix $\Sigma(\omega)$.

From (6),

$$dW_{j}(\omega) = \overline{\beta}_{j}(\omega) \left[\begin{array}{c} dZ_{1}(\omega) \\ \cdot \\ \cdot \\ dZ_{p}(\omega) \end{array} \right] \quad . \quad (j = 1, \dots, p)$$
(12)

so that $dW_j(\omega)$ is a linear combination of $dZ_1(\omega), \ldots, dZ_p(\omega)$ with coefficients equal to the complex conjugate of the components of the column vector $\beta_j(\omega)$, $(j = 1, \ldots, p)$.

3. PRINCIPAL COMPONENTS

The dW₁(ω), ..., dW_p(ω) are called principal components. With the eigenvalues $\lambda_1(\omega)$, ..., $\lambda_p(\omega)$ in the order indicated by (5), dW₁(ω) is called the first principal component, dW₂(ω) the second principal component, etc.

Consider the problem of determining $c_{11}(\omega)$, $c_{21}(\omega)$, ..., $c_{p1}(\omega)$ satisfying

$$\sum_{j=1}^{P} |c_{j1}(\omega)|^2 = 1$$
(13)

such that the linear combination $\sum_{j=1}^{p} \widetilde{c}_{j\lambda}(\omega) dZ_{j}(\omega) = d\widetilde{W}_{1}(\omega)$ has maximum variance, i.e., such that

$$\operatorname{Var} \left(d\widetilde{W}_{1}(\omega) \right) \stackrel{\text{\tiny E}}{=} \operatorname{Var} \left(\sum_{j=1}^{p} \overline{c}_{j1}(\omega) \ dZ_{j}(\omega) \right) \stackrel{\text{\tiny E}}{=} E \left| \sum_{j=1}^{p} \overline{c}_{j1}(\omega) \ dZ_{j}(\omega) \right|^{2} = \operatorname{maximum}$$
(14)

The general solution to that problem is

$$\begin{bmatrix} c_{11}(\omega), c_{21}(\omega), \dots, c_{p1}(\omega) \end{bmatrix} = u_1(\omega) \beta_1'(\omega)$$
(15)

where $u_1(\omega)$ denotes an arbitrary complex number of unit modulus and $\beta_1(\omega)$ is the eigenvector of $\Sigma(\omega)$ corresponding to the eigenvalue $\lambda_1(\omega)$ described previously. Thus, the variance of $dW_1(\omega)$ is a maximum if and only if

$$d\widetilde{W}_{1}(\omega) = u_{1}(\omega) \ dW_{1}(\omega)$$
(16)

Furthermore, that maximum variance is

$$\mathbf{E} \left| d\widetilde{W}_{1}(\omega) \right|^{2} = \mathbf{E} \left| dW_{1}(\omega) \right|^{2} = \mathbf{E} \overline{\beta}_{1}^{\prime}(\omega) \begin{bmatrix} dZ_{1}(\omega) \\ \vdots \\ \vdots \\ dZ_{p}(\omega) \end{bmatrix} \begin{bmatrix} d\overline{Z}_{1}(\omega), \ldots, d\overline{Z}_{p}(\omega) \end{bmatrix}$$

$$\beta_{1}(\omega)$$

$$(17)$$

 $= \overline{\beta}_{1}^{\prime}(\omega) \Sigma(\omega) \overline{\beta}_{1}(\omega) d\omega = \overline{\beta}_{1}^{\prime}(\omega) \overline{\lambda}_{1}(\omega) \beta_{1}(\omega) d\omega = \overline{\lambda}_{1}(\omega) d\omega$

Next, consider the problem of determining $c_{12}(\omega)$, $c_{22}(\omega)$, ..., $c_{p2}(\omega)$ satisfying

$$\sum_{j=1}^{p} |c_{j2}(\omega)|^{2} = 1$$
(18)

such that the linear combination $\sum_{j=1}^{p} \overline{c_{j2}}(\omega) dZ_{j}(\omega) = d\widetilde{W}_{2}(\omega)$ is uncorrelated with $dW_{1}(\omega)$ and has maximum variance, i.e., such that

$$\mathbf{E} \ \mathbf{d} \widetilde{\mathbf{W}}_{2}(\omega) \ \mathbf{d} \overline{\mathbf{W}}_{1}(\omega) = 0 \tag{19}$$

and

$$\operatorname{Var}\left(d\widetilde{W}_{2}(\omega)\right) = E \left|d\widetilde{W}_{2}(\omega)\right|^{2} = \operatorname{maximum}$$
(20)

The general solution to the problem is

$$\begin{bmatrix} c_{21}(\omega), c_{22}(\omega), \dots, c_{p2}(\omega) \end{bmatrix} = u_2(\omega) \beta_2(\omega)$$
 (21)

where $u_2(\omega)$ denotes an arbitrary complex number of unit modulus and $\beta_2(\omega)$ is the eigenvector of $\Sigma(\omega)$ corresponding to the eigenvector $\lambda_2(\omega)$ described previously. Thus, the variance of $dW_2(\omega)$ is a maximum if and only if

$$d\widetilde{W}_{2}(\omega) = u_{2}(\omega) dW_{2}(\omega)$$
 (22)

That maximum variance is

$$E \left| d\widetilde{W}_{2}(\omega) \right|^{2} = E \left| dW_{2}(\omega) \right|^{2} = \lambda_{2}(\omega) d\omega$$
 (23)

Similarly, the $c_{13}(\omega)$, $c_{23}(\omega)$, ..., $c_{p3}(\omega)$ satisfying

$$\sum_{j=1}^{p} |c_{j3}(\omega)|^{2} = 1$$
 (24)

such that $d\widetilde{W}_{3}(\omega) \stackrel{\text{def}}{=} \sum_{j=1}^{p} \overline{c}_{j3}(\omega) dZ_{j}(\omega)$ is uncorrelated with $dW_{1}(\omega)$ and $dW_{2}(\omega)$ and has maximum variance is

$$\left[c_{13}(\omega), c_{23}(\omega), \ldots, c_{p3}(\omega)\right] = u_{3}(\omega) \beta_{3}(\omega) \qquad (25)$$

the $d\widetilde{W}_{3}(\omega)$ possessing maximum variance is

$$d\widetilde{W}_{3}(\omega) = u_{3}(\omega) \ dW_{3}(\omega) \tag{26}$$

and that maximum variance is

$$E \left| dW_{3}(\omega) \right|^{2} = \lambda_{3}(\omega) d\omega$$
 (27)

Proceeding inductively in the manner illustrated above it is seen that

$$d\widetilde{W}_{r+1}(\omega) = u_{r+1}(\omega) \ dW_{r+1}(\omega)$$
(28)

is the solution to the following problem:

Find the $d\widetilde{W}_{r+1}(\omega) \equiv \sum_{j=1}^{p} \overline{c}_{j, r+1}(\omega) dZ_{j}(\omega)$ possessing maximum variance subject to the constraint that $\sum_{j=1}^{p} |c_{j, r+1}(\omega)|^{2} = 1$ and $d\widetilde{W}_{r+1}(\omega)$ is uncorrelated with $dW_{1}(\omega)$, $dW_{2}(\omega)$, ..., $dW_{r}(\omega)$. Furthermore, that maximum variance is

$$\mathbf{E} \left| d\widetilde{W}_{r+1}(\omega) \right|^2 = \mathbf{E} \left| dW_{r+1}(\omega) \right|^2 = \lambda_{r+1}(\omega) dW$$
(29)

From the foregoing discussion it is seen that the transformation (6) of the $\left[dZ_1(\omega), \ldots, dZ_p(\omega)\right]$ to the $\left[dW_1(\omega), \ldots, dW_p(\omega)\right]$, i.e., the transformation to principal components, is a transformation to uncorrelated random variables possessing the special variance properties described above. In some studies the principal components with large

variance may be of special interest. When the principal components with large variance "account for most of the variability, " i.e., when the total variance of the other principal components is comparatively small, restricting attention (in exploratory investigations) to the principal components with large variance may constitute an effective way of reducing the "dimensionality" of a problem.

4. STATISTICAL ESTIMATION OF EIGENVALUES AND EIGENVECTORS OF A SPECTRAL DENSITY MATRIX

Let $\Sigma(\omega)$ denote a sample Hermitian nonnegative definite spectral density matrix constituting an estimator of the true spectral density matrix $\Sigma(\omega)$. The eigenvalues $\widehat{\lambda}_{j}(\omega)$, $(j = 1, \ldots, p)$, of $\widehat{\Sigma}(\omega)$ in descending order are, respectively, estimators for the eigenvalues $\lambda_{j}(\omega)$, $(j = 1, \ldots, p)$, of $\Sigma(\omega)$. Similarly, the corresponding (normalized) eigenve fors $\widehat{\beta}_{j}(\omega)$, $(j = 1, \ldots, p)$, of $\widehat{\Sigma}(\omega)$ are, respectively, estimators for the normalized eigenvectors $\beta_{j}(\omega)$, $(j = 1, \ldots, p)$, of $\Sigma(\omega)$. Under suitable hypotheses (i.e., hypotheses that make $\widehat{\Sigma}(\omega)$ a maximum likelihood estimator for $\Sigma(\omega)$), the $\widehat{\lambda}_{j}(\omega)$, $\widehat{\beta}_{j}(\omega)$, $(j = 1, \ldots, p)$, are, respectively, maximum likelihood estimators for $\lambda_{i}(\omega)$, $\beta_{i}(\omega)$, $(j = 1, \ldots, p)$.

5. STATISTICAL DISTRIBUTION THEORY AND RELATED RESULTS PERTAINING TO THE RANDOM EIGENVALUES OF A SAMPLE SPECTRAL DENSITY MATRIX

To simplify notation the dependence on the frequency ω will not be indicated here. For example, $\widehat{\Sigma}(\omega)$ will simply be denoted by $\widehat{\Sigma}$, the frequency dependence being understood. That is, it is understood that $\widehat{\Sigma}$ is a spectral density estimator pertaining to a small frequency band centered at frequency ω .

The distributional theory and related results described here are predicated on

$$\widehat{\Sigma} = \frac{1}{n} \mathbf{A}$$

where A is complex Wishart distributed with n degrees of freedom. The complex Wishart distribution with parameters n, p, Σ will be denoted by $W_{c}(\Sigma; n, p)$.

<u>Thm. 1.</u> If A has the distribution $W_c(I; n, p)$ with $n \ge p$, then the random eigenvalues $\widehat{K}_1 \ge \widehat{K}_2 \ge \ldots \ge \widehat{K}_p \ge 0$ of A have the joint probability density function

$$p(\widehat{K}_{1}, \ \widehat{K}_{2}, \ \dots, \ \widehat{K}_{p}) = C(p, n) \begin{pmatrix} p \\ \square \\ j=1 \end{pmatrix} \begin{pmatrix} \widehat{K}^{n-p} \\ j \end{pmatrix} e^{-\sum_{j=1}^{p} \widehat{K}_{j}} \prod_{\substack{j < k \\ j, k=1}} \left(\widehat{K}_{j} - \widehat{K}_{k} \right)^{2}$$

(31)

(30)

where the "constant" C(p, n) is given by

$$C(p, n) = \left[\begin{array}{c} p \\ \Pi & \Gamma & (n - p + j) \\ j = 1 \end{array} \right]^{-1}$$

The probability density function (31) is defined over the domain

 $\widehat{K}_{1} \geq \widehat{K}_{2} \geq \cdots \geq \widehat{K}_{p} \geq 0.$ Since the eigenvalues $\widehat{\lambda}_{j}$ of $\widehat{\Sigma}$ are related to the eigenvalues \widehat{K}_{j} of A by $\widehat{\lambda}_j = \frac{1}{n} \widehat{K}_j$, (j = 1, ..., p), one may regard (31) as giving the probability density function of the $\widehat{\lambda}_j$, (j = 1, ..., p).

<u>Thm. 2.</u> If $\widehat{K}_1, \widehat{K}_2, \ldots, \widehat{K}_p$ are random variables distributed with the probability density function (31), then for any constants a, b such that $0 \le a \le b$

$$\operatorname{Prob}\left[\mathbf{a} \leq \widehat{\mathbf{K}}_{p}, \ldots, \widehat{\mathbf{K}}_{2}, \widehat{\mathbf{K}}_{1} \leq \mathbf{b}\right] = C(p, n) \begin{vmatrix} Y_{0} & Y_{1} & \cdots & Y_{p-1} \\ Y_{1} & Y_{2} & \cdots & Y_{p} \\ \vdots & \vdots & \vdots & \vdots \\ Y_{p-1} & Y_{p} & \cdots & Y_{2p-2} \end{vmatrix}$$
(33)

where

$$Y_{i+j-2} \equiv \int_{a}^{b} v^{n-p+i+j-2} e^{-v} dv$$
, (i, j = 1, 2, ..., p)

(34)

(32)

Equation (33) gives a closed form expression for the probability that <u>all</u> the eigenvalues of a random Hermitian matrix A distributed $W_c(I; n, p)$ with $n \ge p$ will be between prescribed limits a and b. Equivalently, one may regard (33) as giving the probability that <u>all</u> the random eigenvalues of a sample spectral density matrix $\hat{\Sigma}$ will be between prescribed limits for the case when the true spectral density matrix $\Sigma = 1$. Confidence band results pertaining to general spectral density matrices are derivable from Thm. 2, so that the condition that $\Sigma = 1$ is not as restrictive as might appear at first glance.

Let A* denote a random Hermitian matrix distributed $W_c(I; n, p)$ where $n \ge p$. Let $ch_{min}(A*)$ denote the minimum eigenvalue of A* and $ch_{max}(A*)$ denote the maximum eigenvalue of A*. Let Σ denote a general nonsingular (p x p) spectral density matrix. From Thm. 2 for a chosen 0 < e < 1 one may obtain constants l* and u* such that

$$1 - e = \operatorname{Prob}\left[l^* \leq \operatorname{ch}_{\min}(A^*), \operatorname{ch}_{\max}(A^*) \leq u^* \right]$$
(35)

The probability statement (35) is equivalent to

$$1 - e = \operatorname{Prob}\left[l^{*} \leq \frac{\overline{a'A^{*}a}}{\overline{a'a}} \leq u^{*}, \text{ for all nonzero complex (p x 1) vectors a} \right]$$
(36)

Now, there exists a nonsingular (p x p) matrix M such that

$$M' M = \Sigma$$
(37)

a = Mb

and

Let

$$\Lambda = \mathbf{M}' \mathbf{A} * \mathbf{M} \tag{39}$$

The Hermitian positive definite matrix A is distributed $W_{c}(\Sigma; n, p)$. Upon substituting (38) in (36) and using (37) and (39) one obtains

$$1 - e = \operatorname{Prob}\left[l * \leq \frac{\overline{b'} A b}{\overline{b'} \Sigma b} \leq u^* , \text{ for all nonzero complex (p x 1) vectors } b \right]$$

(38)

Using (30) one may state (40) in the form

 $1 - a = \operatorname{Prob}\left[\frac{1*}{n} \leq \frac{\overline{b}' \widehat{\Sigma} b}{\overline{b}' \Sigma b} \leq \frac{u*}{n}, \text{ for all nonzero complex (p x 1) vectors b}\right]$ (41)

From (41) one obtains the simultaneous confidence band result:

 $1 - e = \operatorname{Prob}\left[\frac{n}{u^*} \, \overline{b}' \, \widehat{\Sigma} b \leq \overline{b}' \, \Sigma b \leq \frac{n}{\ell^*} \, \overline{b}' \, \widehat{\Sigma} b \right]$ simultaneously for all complex (p x 1) vectors b (42)

The vectors b may be freely chosen in (42) with <u>all</u> bounds on the resulting linear combinations of the elements of Σ holding simultaneously with probability 1 - e. From simultaneous confidence bounds on suitably chosen linear combinations of the elements of Σ one may obtain, for example, simultaneous confidence bounds on <u>all</u> the spectra, co-spectra, and quadrature-spectra of Σ . One may also view (42) as giving simultaneous confidence bands for values of an Hermitian quadratic form where Σ is the matrix of the quadratic form. Equation (42) algo yields

$$1 - \mathbf{e} = \operatorname{Prob}\left[\frac{n}{u^{*}} \operatorname{ch}_{\min}(\widehat{\Sigma}) \leq \lambda_{p}, \ldots, \lambda_{2}, \lambda_{1} \leq \frac{n}{\ell^{*}} \operatorname{ch}_{\max}(\widehat{\Sigma})\right]$$
(43)

where $ch_{\min}(\widehat{\Sigma})$ denotes the minimum eigenvalue of the sample spectral density matrix $\widehat{\Sigma}$ and $ch_{\max}(\widehat{\Sigma})$ denotes the maximum eigenvalue. Equation (43) is a confidence bound for all the eigenvalues of the true spectral density matrix Σ .

From (35) one also obtains

$$1 - c = \operatorname{Prob}\left[\frac{l^{*}}{n} \overline{c'} \widehat{\Sigma}^{-1} c \leq \overline{c'} \Sigma^{-1} c \leq \frac{u^{*}}{n} \overline{c'} \widehat{\Sigma}^{-1} c, \operatorname{simultaneously for all complex (p x 1) vectors c}\right]$$
(44)

The simultaneous confidence band statement (44) yields confidence bound results pertaining to Σ^{-1} directly analogous to those for Σ described above. Viewing (44) as giving simultaneous confidence bands for values of an Hermitian quadratic form where Σ^{-1} is the matrix of the quadratic form may be especially important. Such quadratic forms occur in quadratic signal detection methods. In conclusion it is noted that both (42) and (44) are derived from (35) and hold simultaneously, i.e.

$$1 - e = \operatorname{Prob} \begin{bmatrix} \frac{n}{u^*} \ \widehat{b}^{!} \ \widehat{\Sigma}b \le \overline{b}^{!} \ \Sigma b \le \overline{b}^{!} \ \widehat{\Sigma}b \le \overline{b}^{!} \ \widehat{\Sigma}b \le \overline{b}^{!} \ \widehat{\Sigma}b \end{vmatrix}, \quad \begin{array}{c} \operatorname{simultaneously for all} \\ \operatorname{complex} (p \ x \ 1) \ \text{vectors} \ b \end{aligned}$$

$$\frac{I^*}{n} \ \overline{c}^{!} \ \widehat{\Sigma}^{-1} \ c \le \overline{c}^{!} \ \Sigma^{-1} \ c \le \frac{u^*}{n} \ \overline{c}^{!} \ \widehat{\Sigma}^{-1} \ c \end{aligned}, \quad \begin{array}{c} \operatorname{simultaneously for all} \\ \operatorname{simultaneously for all} \\ \operatorname{complex} (p \ x \ 1) \ \text{vectors} \ c \end{aligned}$$

	ONTROL DATA - BAD					
(Security cleant/icetion of title, body of abstract and inde	axing annotation must be onto	red when	the overall superi is electlied)			
. ORIGINATING ACTIVITY (Corporate author)	1	20. REPORT SECURITY CLASSIFICATION				
TELEDYNE, INC.						
ALEXANDRIA, VIRGINIA 22314						
REPORT TITLE						
EIGENVALUES AND EIGENVECTORS	OF SPECTRAL DEN	SITY	MATRICES			
DESCRIPTIVE NOTES (Type of report and inclusive dates) Scientific						
AUTHOR(3) (Lost name, Stat name, Initial)						
Goodman, N. R.						
7 April 1967	70. TOTAL NO. OF PAR	8 19	75. NO. OF REFS			
. CONTRACT OR GRANT NO.	S. ORIGINATOR'S REP		BER(8)			
F 33657-67-C-1313	179					
VELA T/6702	1.5					
	S. OTHER REPORT NO	(8) (Any	other numbers that may be seeigned			
ARPA Urder No.: 524						
AVAILABILITY/LINITATION NOTICES						
This document is subject to sp	pecial export of	contro	ls and each			
transmittal to foreign governm	ments or foreig	n nat	ional may be			
I SUPPLEMENTARY NOTES	12. SPONSORING MILITA	RY ACTI	VITY			
	ADVANCED RESEARCH PROJECTS AGENCY					
	NUCLEAR TEST WASHINGTON	CTION OFFICE				
ABSTRACT	1					
This report describes some int	terpretations a	and us	ses of eigenvalues			
and eigenvectors of spectral a	and sample spec	tral	density matrices			
of multiple stationary time se	eries.					
The spectral densities antenin of			1. abatismana			
time series is defined Figer	t a zero-mean m	ultip onvoc	tors of the			
anastrol density matrix are di	ivaraco and cre	incir	al component			
Spectral density matrix are up	iscussed and pr					
theory is presented. Statisti	iscussed and pr ical distributi	on th	eory and related			
theory is presented. Statisti results are used to investigat	iscussed and pr ical distributi te the eigenval	on th	eory and related of a sample spectra			
theory is presented. Statistic results are used to investigat density matrix. This investig	iscussed and pr ical distributi te the eigenval gation gives me	on th ues c thods	eory and related of a sample spectra for obtaining			
theory is presented. Statisti results are used to investigat density matrix. This investig simultaneous confidence bounds	iscussed and prical distributi te the eigenval gation gives me s on the elemen	on th ues of thods	eory and related of a sample spectra for obtaining the true spectral			
theory is presented. Statistic results are used to investigat density matrix. This investig simultaneous confidence bounds density matrix and its inverse confidence bounds on the eiger	iscussed and prical distributi te the eigenval gation gives me s on the elements, and also met	on th ues of thods thods	neory and related of a sample spectra for obtaining the true spectral for obtaining spectral density			
theory is presented. Statistic results are used to investigat density matrix. This investig simultaneous confidence bounds density matrix and its inverse confidence bounds on the eigen matrix.	iscussed and prical distributi te the eigenval gation gives me s on the elemen e, and also met nvalues of the	on th ues of thods thods true	neory and related of a sample spectra for obtaining the true spectral for obtaining spectral density			
theory is presented. Statistic results are used to investigate density matrix. This investigate density matrix and its inverse confidence bounds on the eigen matrix.	iscussed and prical distributi te the eigenval gation gives me s on the elemen e, and also met nvalues of the	on th ues of thods thods true	eory and related of a sample spectra for obtaining the true spectral for obtaining spectral density			
theory is presented. Statisti results are used to investigat density matrix. This investig simultaneous confidence bounds density matrix and its inverse confidence bounds on the eiger matrix.	iscussed and prical distributi te the eigenval gation gives me s on the elemen e, and also met nvalues of the	on th ues of thods thods true	eory and related of a sample spectra for obtaining the true spectral for obtaining spectral density			
theory is presented. Statistic results are used to investigat density matrix. This investig simultaneous confidence bounds density matrix and its inverse confidence bounds on the eigen matrix.	iscussed and prical distributi te the eigenval gation gives me s on the elemen e, and also met nvalues of the	on th ues of thods thods true	eory and related of a sample spectra for obtaining the true spectral for obtaining spectral density			
theory is presented. Statisti results are used to investigat density matrix. This investig simultaneous confidence bounds density matrix and its inverse confidence bounds on the eigen matrix.	iscussed and pr ical distributi te the eigenval gation gives me s on the elemen e, and also met nvalues of the	on th ues of thods tts of thods true	eory and related of a sample spectra for obtaining the true spectral for obtaining spectral density			
theory is presented. Statisti results are used to investigat density matrix. This investig simultaneous confidence bounds density matrix and its inverse confidence bounds on the eiger matrix.	iscussed and pr ical distributi te the eigenval gation gives me s on the elemen e, and also met nvalues of the	on th ues of thods tts of thods true	eory and related of a sample spectra for obtaining the true spectral for obtaining spectral density			

KEY WARA		LIN	LINK B		LINKC				
		HOLE	WT	ROLE	WT	ROLE	WT		
eigenvectors eigenvalues multichannel time series spectral density matrix inverse matrix confidence bounds principal components									
4. MARY 19.	licence								
ORIGINATING ACTIVITY: Enter the name and address	Impose	he constant	of each #	atlan			-		
REPORT SECURITY CLASSIFICATION: Enter the over- security classification of the report. Indicate whether estricted Date" is included. Marking is to be in accord- ce with appropriate security regulations. GROUP: Automatic downgrading is specified in DoD Di- tive 5200, 10 and Armed Forces Inducted Marked. Enter	 (i) "Qualified requesters may obtain copies of this report from DDC." (2) "Foreign announcement and dissemination of this report by DDC is not authorized." (3) "U. E. Government agencies may obtain copies of this report directly from DDC. Other qualified DDC users shall request through " (4) "U. E. military sgencies may obtain copies of this report directly from DDC. Other qualified users shall request through " (4) "U. E. military sgencies may obtain copies of this report directly from DDC. Other qualified users shall request through " (5) "All distribution of this report is controlled. Qualified DDC users shall request through " (5) "All distribution of this report is controlled. Qualified DDC users shall request through " (6) "All distribution of this report is controlled. Qualified DDC users shall request through " (7) If the report has been furnished to the Offics of Technical Services, Department of Commerce, for sais to the public, indicate this fact and enter the price, if known. (1) SUPPLEPZENTARY NOTES: Use for additional explanatory notes. (2) SPONSORING MILITARY ACT VITY: Enter the name of the departmental project office or isboratory sponsoring (papering for) the research and development. Include address (3) ABSTRACT: Enter an sbatract giving 5 brief and factuat summery of the document indicative of the report, even though it may also appear elsawhere is the body of the technical report. If additional space is required, a continuation sheet shall 								
ne group number. Also, when applicable, show that option/i larkings have been used for Group 3 and Group 4 as suthor- sed. REPORT TITLE: Enter the complete report title in all apital letters. Titles is all cases should be unclassified. a meaningful title cannot he selected without classifica- on, show title classification in all capitals in parenthesis amediately following the title. DESCRIPTIVE NOTES: If appropriate, suter the type of port, e.g., laterim, progress, summary, annual, or final. ive the inclusive dates when a specific reporting period is									
AUTHOR(S): Enter the name(s) of author(s) as shown on in the report. Enter last name, first name, middle initial, military, show rank and branch of service. The name of a principal author is an absolute sainimum requirement. REPORT DATE: Enter the date of the report as day, nath, year; or month, yeas. If rest; then one date appears the report, use date of publication. . TOTAL NUMBER OF PAGES: The total page count outd follow normal paginalies procedures, Les, enter the mber of pages containing information.									
NUMBER OF REFERENCER Enter the total number of erences cited in the report. CONTRACT OR GRANT NUMBER: if appropriate, enter applicable number of the contract or grant under which	it is highly desirable that the abstract of classified reports be uncleasified. Each paragraph of the abstract shall end with an indication of the military security classification of the in- formation in the paragraph, represented as (TS), (S), (C), or (7).								
s report was written. 8c, & 8d. PROJECT NUMBER: Enter the appropriate litary department identification, such as project another.	There is no limitation on the length of the abstract. How- aver, the suggested length is from 150 to 225 words.								
bproject member, system numbers, task number, etc. O"IGINATOR'S REPORT NUMBER(S): Enter the offi- il report number by which the document will be identified d controlled by the originating ectivity. This number must enique to this report. OTHER REPORT NUMBER(S): if the report has been algoed any other report numbers (either by the originator by the originator.	14. KEY WORDS: Ney words are technically meaningful terms or short phrases that characterize a report and may be used as index entries for estaloging the report. Key words must be astected so that no security classification to required. Menti- flars, such as equipment model designation, trade nome, military project code name, geographic location, may be used as key words but will be followed by an indication of technical con- text. The assignment of links. rules, and weights is optional.								
AVAILABILITY/LIMITATION NOTICES: Salar any lim-									