PRINCIPAL COMPONENTS ANALYSIS (PCA)*

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Abstract—Principal Components Analysis (PCA) as a method of multivariate statistics was created before the Second World War. However, the wider application of this method only occurred in the 1960s, during the "Quantitative Revolution" in the Natural and Social Sciences.

The main reason for this time-lag was the huge difficulty posed by calculations involving this method. Only with the advent and development of computers did the almost unlimited application of multivariate statistical methods, including principal components, become possible.

At the same time, requirements arose for precise numerical methods concerning, among other things, the calculation of eigenvalues and eigenvectors, because the application of principal components to technical problems required absolute accuracy.

On the other hand, numerous applications in Social Sciences gave rise to a significant increase in the ability to interpret these nonobservable variables, which is just what the principal components are. In the application of principal components, the problem is not only to do with their formal properties but above all, their empirical origins.

The authors considered these two tendencies during the creation of the program for principal components. This program—entitled PCA—accompanies this paper. It analyzes consecutively, matrices of variance-covariance and correlations, and performs the following functions:

- —the determination of eigenvalues and eigenvectors of these matrices,
- -the testing of principal components,
- —the calculation of coefficients of determination between selected components and the initial variables, and the testing of these coefficients,
- —the determination of the share of variation of all the initial variables in the variation of particular components,
- -construction of a dendrite for the initial set of variables,
- -the construction of a dendrite for a selected pattern of the principal components,
- —the scatter of the objects studied in a selected coordinate system.

Thus, the PCA program performs many more functions especially in testing and graphics, than PCA programs in conventional statistical packages. Included in this paper are a theoretical description of principal components, the basic rules for their interpretation and also statistical testing.

Key Words: Principal Components Analysis, Variance-covariance matrix, Coefficients of determination, Eigenvalues, Eigenvectors, Correlation matrix, Bartlett's statistics, FORTRAN 77.

DESCRIPTION OF THE PRINCIPAL COMPONENTS METHOD

The basic aim of the analysis utilizing principal components is a reduction of the dimensions of the observation space in which given objects are studied (Kendall, 1983; Jackson and Hearne, 1973). The reduction is obtained by creating new linear combinations of variables characterizing the objects studied. These combinations, termed principal components, must satisfy certain mathematical and statistical conditions. They will be discussed in detail in subsequent sections.

The starting point in the principal components method is an observation matrix X in which column

vectors list observations characterizing an object with respect to random variables X_1, X_2, \ldots, X_n .

$$\mathbf{X} = \begin{bmatrix} x_{11} & x_{22} & \cdots & x_{1n} \\ x_{21} & x_{12} & \cdots & x_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ x_{p1} & x_{p2} & \cdots & x_{pn} \end{bmatrix}.$$

Each column vector represents a point in a p-dimensional space. Because the observation matrix \mathbf{X} is compiled for a sample of the entire population (numbers p and n are finite), the variance-covariance matrix \mathbf{S} derived from observations of random variables is an estimator of general variance-covariance matrix $\mathbf{\Sigma}$, whereas the vector of mean values $\mathbf{\bar{x}}$ is an estimator of the general vector \mathbf{U} . Thus as mentioned the task of the principal components method is to determine linear combinations with a maximum

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variance. Thus, the problem essentially is replacing the set of initial variables with their linear combinations, that is new variables with special properties. These new variables are termed principal components and are written in the form:

$$V = A'X \tag{1}$$

where

V is a matrix of the new variables,

A is a matrix of orthonormal eigenvectors of matrix S, and

X is the observation matrix.

Transformation (1) is possible after determinantal Equation (2) has been solved.*

$$|\mathbf{S} - l\mathbf{I}| = 0 \tag{2}$$

where

S is a variance—covariance matrix of order $(p \times p)$, l is the characteristic root of the determinantal equation, and

I is a unit matrix of order $(p \times p)$.

Equation (2) is a polynomial of degree p with respect to unknown l, hence it has p roots which can be ordered in such a way that

$$l_1 \geqslant l_2 \geqslant l_3 \geqslant \cdots \geqslant l_n \geqslant 0.$$

Because there is an orthonormal column eigenvector A_i corresponding to each root l_i , the variable V_1 derived from Equation (2) has the maximum value l_1 (maximum variance) and is termed the first principal component.

Because the sum $l_1 + l_2 + \cdots + l_p = \operatorname{tr} \mathbf{S}$ and is equal to the sum of the variances of matrix \mathbf{S} (i.e $\sigma_{11} + \sigma_{22} + \cdots + \sigma_{pp}$), l_1, l_2, \ldots, l_p defines the share of variability of particular principal components in the total variance of matrix \mathbf{S} . If we consider the quotients

$$\frac{l_1}{\text{tr }\mathbf{S}}100, \quad \frac{l_2}{\text{tr }\mathbf{S}}100, \quad \dots, \quad \frac{l_p}{\text{tr }\mathbf{S}}100, \quad (3)$$

we get the percent share of each component in the variance of matrix S. The algorithm for the calculation of principal components is such that this is a decreasing sequence, which indicates that $l_1/\text{tr} \, S \, 100$ is the largest quantity. Quantity l_2 corresponds to variable V_2 , which therefore is termed the second principal component. It is clear that there are as many principal components as initial variables.

Each root l_i has its corresponding column vector A_i such that

$$(\mathbf{S} - l\mathbf{I})\mathbf{A}_i = 0$$
 or $\mathbf{S}\mathbf{A}_i = l_i\mathbf{A}_i$. (4)

Because vectors A_1, A_2, \ldots, A_p are orthonormal, that is

$$\mathbf{A}_{i}'\mathbf{A}_{i} = 1, \quad \mathbf{A}_{i}\mathbf{A}_{i} = 0 \quad \text{for } i \neq j,$$
 (5)

and they satisfy (2) and (4), we have

$$\mathbf{A}_{i}'\mathbf{S}\mathbf{A}_{i} = l_{i}, \quad \mathbf{A}_{i}'\mathbf{S}\mathbf{A}_{i} = 0 \quad \text{for } i \neq j,$$
 (6)

$$\mathbf{I} = \mathbf{A}_1 \mathbf{A}_1' + \dots + \mathbf{A}_n \mathbf{A}_n' \tag{7}$$

and

$$S = l_1 A_1 A_1' + l_2 A_2 A_2' + \dots + l_n A_n A_n'.$$
 (8)

Expression (8) is termed a spectral decomposition of a matrix S.

The basic property of the new variables is the lack of correlation among them (in contrast to the initial variables). The variance of the *i*th component is l_i , or

$$Var(\mathbf{A}_i \mathbf{X}) = l_i, \tag{9}$$

whereas

$$Cov(\mathbf{A}_i\mathbf{X}, \mathbf{A}_i\mathbf{X}) = 0$$
 for $i \neq j$.

Because the primary aim of principal components analysis is the reduction of the dimensions of the observation space, it is necessary at some stage to decide on how many new variables should be taken into account further study. To help with the decision, the ratio of characteristic roots to the trace of the matrix is considered. For example, if the expression $l_1/\text{tr } S 100$ has a great value (e.g. 90%), the set of initial variables is replaced with the first component V_1 . When the ratio is not so high, the next components are taken into account. Naturally, the elimination of some components from further analysis cannot follow solely from the researcher's subjective evaluation of the $l_i/\text{tr } S$ 100 quotient, but must result from the testing of the components. This problem is discussed in detail in the next section.

An important issue in Principal Components Analysis is the interpretation of the components, to help determine, after the reduction of the observation space, which initial variables have the greatest shares in the variance of particular principal components. This information can be obtained using coefficients of determination established between the components and the initial variables. It should be added that interpretation of the components differs slightly depending on whether S or R is used.

Interpretation of principal components derived from variance-covariance matrix S

The coefficient of correlation between the *i*th component and the *j*th initial variable is defined by the equation:

$$r_{ij} = \frac{a_{ij}\sqrt{l_i}}{S_i}. (10)$$

Hence the coefficient of determination has the form:

$$r_{ij}^2 = \frac{a_{ij}^2 l_i}{S_i^2},\tag{11}$$

^{*}It also is possible to derive principal components by substituting the variance-covariance matrix S with correlation matrix R, that is by solving the equation $|\mathbf{R} - I\mathbf{I}| = 0$.

where a_{ij}^2 is the square of the element of the eigenvector A_i corresponding to the *i*th component and *j*th initial variable, l_i is the variance of the *i*th component, and S_i^2 is the variance of variable *j*.

On the basis of (8) and (9), making use of the variances and eigenvectors of all the principal components, it is possible to reconstruct the variance-covariance matrix S. Naturally, the product $l_1\mathbf{A}_1\mathbf{A}_1'$ has the greatest share in this reconstruction. Moreover, in for example matrix $l_1\mathbf{A}_1\mathbf{A}_1'$ the elements on the main diagonal are estimates (supplied with the help of the first component) of the variance of the jth initial variable, which can be computed from a general expression:

$$\hat{V}$$
ar $X_i = l_1 a_{1i}^2$, (12)

whereas the remaining elements are estimates of covariance, with the final element (matrix) of spectral decomposition $l_p \mathbf{A}_p \mathbf{A}_p'$ bringing the estimated variance and covariance up to real values. Given (12), (11) can be written in the form:

$$r_{ij}^2 = \frac{\hat{V}ar X_j}{Var X_i} = \frac{\hat{S}_j^2}{S_i^2}.$$
 (13)

It is clear that the coefficient of determination between component i and variable j is a ratio of the estimated variance of variable j to its real variance. If we consider any (ith) matrix from the spectral decomposition, then, by summing up the elements on the main diagonal

$$\sum_{j=1}^{p} \hat{S}_{j}^{2} = \sum_{j=1}^{p} l_{i} a_{ij}^{2}$$

$$\sum_{i=1}^{p} \hat{S}_{j}^{2} = l_{i} \sum_{i=1}^{p} a_{ij}^{2}$$

we get

$$\sum_{i=1}^{p} \hat{S}_{j}^{2} = l_{i}. \tag{14}$$

Thus, by adding together the estimated variances of particular variables we arrive at the variance of the *i*th component. This relationship can be the theoretical basis for the interpretation of the components. Using (14), we can write (3) alternatively as

$$\frac{\sum_{j=1}^{p} \hat{S}_{j}^{2}}{\text{tr } \mathbf{S}} 100. \tag{15}$$

Expression (15) assumes the greatest value for the first component; however, this measure should be used carefully. It follows from (14) that almost the entire variance of the *i*th component is made up of the estimated variance of a single variable, for example one that has a high absolute value in comparison with the remaining ones, that is one with high variance. Hence the necessity of a skillful formulation of the observation matrix, variables of which should be of a similar order of measure (15), the following dependence is proposed for use in the interpretation

of the components:

$$w = \frac{\sum_{i,j}^{p} r_{ij}^2}{n} 100 \tag{16}$$

where p is the number of variables of the observation matrix, and r_{ij}^2 is the coefficient of the determination between the *i*th component and *j*th initial variable.

Equation (16) shows the percent of the variance of all the variables accounted for by the ith component. The results of measure (16) applied to components derived from the covariance matrix are as a rule lower than those of measure (15), because in reality one (e.g. the first) component seldom accounts for more than 50% of the variance of all the variables included in the observation matrix. When a variable in the observation matrix exceeds all others markedly in value, expression (15) gives a high value of the first component, and expression (16) a low one. It is so because the variance of the component depends in this situation on a single variable only, and may not be the most important one either (its significance depends solely on the adopted units of measurement). As can be seen, expression (16) shows the actual share of the ith component in the variance of all the variables. It is worth noting that the numerator of (16) is easy to calculate because the following equation holds:

$$\sum_{i,j}^{p} r_{ij}^{2} = I_{i} \sum_{j=1}^{p} \frac{a_{ij}^{2}}{S_{i}^{2}}.$$
 (17)

The variance of all the initial variables is accounted for by the component if:

$$\sum_{i,j}^{p} r_{ij}^2 = l_i \sum_{j=1}^{p} \frac{a_{ij}^2}{S_i^2} = p.$$

Interpretation of principal components derived from the correlation matrix R

The discrepancies between evaluations of principal components obtained on the basis of formulae (15) and (16) do not occur if we derive them from a correlation matrix (i.e. if use is made of normalized initial variables). This is so because the following dependencies hold by virtue of an appropriate transformation of the covariance matrix (Anderson, 1958):

$$r_{ii} = a_{ii} \sqrt{l_i}, \tag{18}$$

$$r_{ii}^2 = a_{ii}^2 l_i, \tag{19}$$

$$\sum_{i,i}^{p} r_{ij}^{2} = \sum_{i,i}^{p} a_{ij}^{2} l_{i},$$

$$\sum_{i}^{p} r_{ij}^2 = l_i. \tag{20}$$

Or,

$$\frac{l_i}{\text{tr S}} 100 = \frac{\sum_{i,j}^{p} r_{ij}^2}{p} 100.$$

Thus, when the correlation matrix is used, in order to determine what part of the variance of all the initial

variables is accounted for (in percent) by the *i*th component, it is possible to employ either measure (15) or (16). There is no such optionality in the situation of the variance—covariance matrix.

TESTING PRINCIPAL COMPONENTS

Because matrices S and R are only estimators of matrices Σ and P (covering the entire population), the results obtained must be subjected to a verification procedure. There are several approaches to the testing of principal components (cf. Mardia, Kent, and Bibby, 1979; Anderson, 1984). They are discussed next and then used in the program. It must be emphasized that the tests used in this program are based on the asymptotic distributions of the roots and vectors of variance—covariance and correlation matrices. Thus, the size of the sample used in the calculation process may have considerable influence on the results of the testing (see Anderson, 1984, p. 468).

Testing principal components derived from variance–covariance matrix S

Testing the hypothesis about the ratio of the sum of the least characteristic roots to the sum of all roots. If the sum of the characteristic roots of a few final components with relation to the trace of matrix S is relatively small, then there is a justifiable temptation to eliminate these components from further analysis. They can be rejected after the following hypothesis has been verified:

$$H_0$$
: $f(\lambda) = \frac{\lambda_{k+1} + \dots + \lambda_p}{\lambda_1 + \dots + \lambda_p} \geqslant \delta$ (21)

against an alternative hypothesis:

$$H_1$$
: $f(\lambda) < \delta$,

where δ is assumed known.

The asymptotic variance $f(\mathbf{l})$ is determined from the equation:

$$2\left(\frac{\delta}{\operatorname{tr}\boldsymbol{\Sigma}}\right)^{2}(\lambda_{1}^{2}+\cdots+\lambda_{k}^{2}) + 2\left(\frac{1-\delta}{\operatorname{tr}\boldsymbol{\Sigma}}\right)^{2}(\lambda_{k+1}^{2}+\cdots+\lambda_{p}^{2}). \quad (22)$$

Hypothesis H_0 is rejected if $\sqrt{n} [f(\mathbf{l}) - \delta]$ is smaller than the corresponding point of the standardized normal distribution multiplied by the root from (22) in which estimators are substituted for real values, that is:

$$2\left(\frac{\delta}{\operatorname{tr}\mathbf{S}}\right)^{2}(l_{1}^{2}+\cdots+l_{k}^{2}) + 2\left(\frac{1-\delta}{\operatorname{tr}\mathbf{S}}\right)^{2}(l_{k+1}^{2}+\cdots+l_{p}^{2}) \quad (23)$$

whereas

$$f(\mathbf{l}) = \frac{l_{k+1} + \dots + l_p}{l_1 + \dots + l_p}.$$

Determination of the confidence interval for the proportion of variance accounted for the successive principal components. The point value of the proportion of variance accounted for by successive principal components is determined from the equation:

$$\hat{\psi} = \frac{l_1 + \dots + l_k}{l_1 + \dots + l_n}.$$
 (24)

However, the determination of the confidence interval for quantity $\hat{\psi}$ may supply further significant information about the boundary values of the interval in which the true value of ψ may be contained.

Assuming that

$$\hat{\alpha} = \frac{l_1^2 + \dots + l_k^2}{l_1^2 + \dots + l_n^2},$$

the variance of estimator $\hat{\psi}$ is calculated from the equation:

$$\hat{\tau}^2 = \frac{2 \operatorname{tr} \mathbf{S}^2}{(n-1)(\operatorname{tr} \mathbf{S})^2} (\hat{\psi}^2 - 2\hat{\alpha}\hat{\psi} + \hat{\alpha}^2).$$
 (25)

In turn, the confidence interval of $\hat{\psi}$ (for $\alpha = 0.05$) has the form:

$$\hat{\psi} + 1.96(\hat{\tau}^2)^{1/2}.\tag{26}$$

Testing the hypothesis that (p-k) eigenvalues of matrix Σ are equal. After p principal components have been determined, it is advisable to verify the hypothesis that (p-k) eigenvalues are equal. Such a situation is termed isotropy and may suggest that variance is equal in all the directions of (p-k)-dimensional space spanned by the last (p-k) eigenvectors. The test helps determine the number of principal components used to describe objects under study.

In the testing procedure use is made of Bartlett's asymptotic approximation:

$$\left(n - \frac{2p+11}{6}\right)(p-k)\log\left(\frac{a_0}{g_0}\right) \sim \chi^2(p-k+2)(p-k-1)/2 \quad (27)$$

where

$$a_0 = \frac{l_{k+1} + \dots + l_p}{p-k}, \quad g_0 = (l_{k+1} \times \dots \times l_p)^{1/(p-k)},$$

n is the number of observations, and the number of degrees of freedom is calculated from the equation: $df = \frac{1}{2}(p - k + 2)(p - k + 1)$.

With the help of statistics (27) the zero hypothesis about the equality of eigenvalues of matrix Σ is tested:

$$H_0: \lambda_n = \lambda_{n-1} = \cdots = \lambda_{k+1}$$
.

Hypothesis H_0 usually is tested sequentially, taking k = 0, k = 1, etc. If $\chi^2_{\text{obl}} \ge \chi^2_{\text{x,df}}$, then H_0 is rejected.

Testing principal components derived from correlation matrix R

Testing the hypothesis that (p - k) least eigenvalues of matrix **P** are equal. When principal components are derived from correlation matrix **R** (which is the

estimator of the general correlation matrix P), to test the hypothesis that (p-k) of its eigenvalues are equal use is made of Bartlett's statistics having the form:

$$(n-1)(p-k)\log\left(\frac{a_0}{g_0}\right),\tag{28}$$

where

$$a_0 = \frac{l_{k+1} + \cdots + l_p}{p - k},$$

$$g_0 = (l_{k+1} \times \cdots \times l_p)^{1/(p-k)}, \ 0 < k < p-1.$$

This expression can be treated as chi-square with $\frac{1}{2}(p-k+2)(p-k-1)$ degrees only if the first k components account for a relatively large portion of variance. This statistic is used to test the hypothesis:

$$H_0$$
: $\lambda_p = \lambda_{p-1} = \cdots = \lambda_{k+1}$.

If $\chi^2_{\text{obl}} \geqslant \chi^2_{\alpha,df}$, then hypothesis H_0 is rejected.

INSTRUCTIONS FOR THE PCA SYSTEM

The Principal Components Analysis program (henceforth termed PCA) has been written in the FORTRAN-77 programming language. Being fully compatible with the FORTRAN-77 standard it can be used without further modifications on any computer equipped with a compiler of this language.

The high quality of the compilers used and the carefully selected (and tested) numerical procedures performing the main part of the calculations should guarantee a reliable and efficient performance of the program.

In this paper the PCA program is presented in the Appendix together with some comments relative to the Operating System DOS (for the IBM PC).

System requirements

The basis for a successful run of the PCA on IBM PC type machines is compliance with the system requirements and the constraints of the operating system DOS 3.30 (or later).

The PCA system is designed to be used in machines with at least 640 kbyte of random access memory (RAM). It does not use extended memory (provided that the user has not changed parameter settings in the beginning of the source code). If it is to be run in a given PC system, it is necessary to modify (once) the CONFIG.SYS file stored in this system by introducing into it the sentences

$$FILES = 20$$

 $BUFFERS = 10$.

This is an indispensable modification, because the operation of the PCA program utilizes many more files simultaneously than the DOS standard allows.

Each time the PCA system is initiated, all programs residing in the RAM (e.g. Side-Kick, Norton Commander, etc.) should be removed.

Data preparation

In order to simplify as much as possible the process of data input and the derivation of results in the PCA system, a "file-to-file" method of information transmission has been introduced. Thus data can be prepared earlier, independently of the system in question, using an appropriate wordprocessor and stored in a file with a specified name facilitating its identification. The results, in turn (also made into a file), also can be inspected, analyzed, or printed independently of the PCA system (with reservations as discussed). Further on, we present a sample data file 'd1-pca' which will serve to demonstrate how to operate the PCA program and how it runs. This file has been prepared according to a pattern which must be followed in any other data file.

6					
29					
1.08	7.43	0.60	1.27	8.00	0.36
1.00	9.01	0.71	1.08	9.01	0.36
1.13	7.19	0.49	1.24	8.14	0.40
1.03	6.24	0.55	1.82	6.63	0.47
1.04	7.07	0.57	1.50	7.35	0.38
1.17	7.63	0.59	0.88	8.90	0.42
0.89	7.16	0.67	1.38	6.37	0.33
1.04	9.05	0.65	1.60	9.40	0.38
1.04	10.23	0.81	1.80	10.60	0.37
1.15	6.49	0.51	0.95	7.49	0.41
1.13	6.24	0.43	1.11	7.06	0.40
1.14	6.38	0.53	1.23	7.26	0.41
1.05	7.21	0.63	1.20	7.54	0.38
1.15	6.97	0.54	0.95	7.99	0.40
1.06	7.07	0.53	1.13	7.49	0.38
1.03	7.31	0.67	1.17	7.49	0.36
1.05	8.63	0.62	1.38	9.09	0.39
1.06	6.05	0.46	1.29	6.42	0.38
1.05	7.86	0.45	2.09	8.28	0.37
1.19	5.34	0.42	0.93	6.33	0.43
0.95	6.76	0.49	1.63	6.44	0.35
1.12	6.37	0.53	1.0	7.13	0.39
0.95	5.63	0.53	1.60	5.34	0.33
0.94	7.46	0.55	1.82	6.92	0.33
1.04	6.76	0.53	1.08	7.26	0.39
1.04	8.27	0.67	1.27	8.62	0.38
1.07	9.12	0.51	1.07	9.75	0.38
1.05	9.01	0.74	1.38	9.49	0.39
1.04	6.95	0.51	1.50	7.22	0.35

There is a single figure in the first row of the 'd1-pca' file (further on interpreted as P—the number of variables) satisfying the condition $1 \le P \le 50$.

The second row contains one integer (N—the number of objects) which satisfies the condition $3 \le N \le 200$.

Next come elements of the observation matrix (in lexicographic order) in such a way that the figure placed in the ith row in the jth position represents the value of the jth variable of the ith object.

The PCA system does not destroy the input data

The operation of the system and its control

Let us assume, for the sake of simplicity, that the PCA system to be operated appears, together with a suitable data file, on a selected path of a hard disc, and that the output file will be put on this path. too. To start the system, the user should write the name

PCA

and press the ENTER key. The course which the computation process will take depends crucially on the user's answers to questions asked by the system. The final part of this documentation contains an example of a dialog with the system and the results for which it provided a basic (for the d1-pca data file). Next we describe the principal elements of this dialog.

After the display of introductory information, the system asks the user for the name of the file. It should be noted that the name should be given in single quotation marks. With the assumptions mentioned previously and in the example now discussed, the correct answer to the question about the file name is

d1-pca.

The answer should be followed by pressing the ENTER key, and this holds for each next answer. Then follows the question about the name of the

output file. In our example the answer may be

wyn.1st.

It is assumed in the PCA system that the full name of the data file and an output file may be arbitrary, but its length should not exceed 60 characters.

When the names of the two files have been given, the processing and physical formulation of the output file takes place. At the same time the first few eigenvalues of the variance-covariance matrix are displayed.

Calculations are suspended temporarily when the following message appears.

EXAMINE YOUR OUTPUT FILE IN ORDER TO CHOOSE A NEW COORDINATE SYSTEM THEN PRESS (ENTER) IT IS USUALLY POSSIBLE TO EXECUTE ANY OS COMMAND NOW Execution suspended:

This is connected with the need for the user to evaluate the results obtained thus far and to make a decision about further calculations. If the information displayed on the screen so far is not sufficient to make such a decision, it is recommended to follow the DOS command:

type wyn.1st | more

(or its equivalent, depending on the name selected for the output file). This command will result in the part of the output file so far being displayed in page order on the screen.

Having completed the PCA analysis with the use of the variance-covariance matrix, the system proceeds automatically to the second stage, namely a PCA analysis making use of the correlation matrix. The questions that follow and the form of dialog are similar to those in the first stage of calculations.

The PCA program makes possible an analysis of the objects under study in new systems of coordinates (principal components) selected by the user. In the figure, the position of an object is denoted by a number from 1 to 9, depending on how many objects have identical coordinates in a new system. If there are more than nine objects sharing the same coordinates, a special graphic sign appears. On the righthand side of the figure identification numbers of the objects are printed.

Running the PCA system for the exemplary data series D1-PCA—Results (file WYN.LST)

* PRINCIPAL COMPONENTS ANALYSIS * ********************

FOR DATA FILE D1-PCA TUMBER OF VARIABLES = 6

NUMBER OF CASES= 29

VAR.-COVARIANCE MATRIX WAS TAKEN INTO ACCOUNT \$

DESCRIPTIVE PARAMETERS OF ORIGINAL VARIABLES

VARIABLE	MEAN	VARIANCE	STAND. DEV.	
X(1)	0.105793E+01	0.4864685E-02	0.6974729E-01	
I(2)	0.734103E+01	0.1300064E+01	0.1140204E+01	
I(3)	0.568621E+00	0.8680856E-02	0.9317111E-01	
I(4)	0.132241E+01	0.9271486E-01	0.3044912E+00	
X(5)	0.775897E+01	0.1416264E+01	0.1190069E+01	
I(6)	0.381724E+00	0.9039239E-03	0.3006533E-01	

```
##### VAR.-COVARIANCE NATRIX #####
            COL( 1)
                                      COL(2)
                                                                 COL(3)
                                                                                         COL( 4)
                                                                                                                      CDL(5)
        0.486468D-02 -0.158013D-01 -0.229941D-02 -0.128019D-01 0.160151D-01
 1
 2 -0.158013D-01 0.130006D+01 0.777635D-01 0.779941D-01 0.124907D+01
 3 \quad -0.229941 \\ D-02 \quad 0.777635 \\ D-01 \quad 0.868086 \\ D-02 \quad 0.365850 \\ D-02 \quad 0.652537 \\ D-01 \\ D-02 \quad 0.652537 \\ D-01 \\ D-02 \quad 0.868086 
      -0.128019D-01 0.779941D-01 0.365850D-02 0.927149D-01 -0.205957D-02 0.160151D-01 0.124907D+01 0.652537D-01 -0.205957D-02 0.141626D+01
 4
 5
       0.151046D-02 -0.728109D-02 -0.670036D-03 -0.289727D-02 0.335006D-02
 6
            COL( 6)
 1
       0.151046D-02
       -0.728109D-02
 3
      -0.670036D-03
 4 -0.289727D-02
 5
       0.335006D-02
 6
       0.903924D-03
TESTING THE PRINCIPAL COMPONENTS
                                  (FOR VARIANCE-COVARIANCE MATRIX)
___ MO.__CRITICAL DELTA__ LEFTCOWFID.___ RIGHTCOWFID.
                   100
                                                89.20
                                                                                       96.00
          1
          2
                       4
                                                97.60
                                                                                      99.10
          3
                                                99.80
                                                                                     99.90
                      1
           4
                                               100.00
                                                                                    100.00
          5
                       0
                                                  0.00
                                                                                       0.00
          6
                       ٥
                                                  0.00
                                                                                        0.00
 BARTLET'S TEST REJECTS EIGENVECTORS NO.
          5
          6
                  ****** #ON-ZERO EIGENVALUES *******
         BR.
                                               LAMBDA
                                                                     PERCENTAGE OF TRACE
          1
                             0.2613593905742508E+01 92.57
                              0.1631374931478640E+00
                                                                                    5.78
           3
                              0.4304858866144624E-01
                                                                                     1.52
           4
                              0.3247796870052141E-02
                                                                                       0.12
           5
                              0.3270209024478446E-03
                                                                                       0.01
                              0.1384170300235664E-03
                                                                                       0.00
         SELECTED EIGENVECTORS (FROM 1 TO 2 ) OF THE VAR.-COVARIANCE MATRIX
                A( 1)
                                          A( 2)
 1 -0.0001155169 0.1594767997
 2 -0.6902897618 -0.5309246613
 3 -0.0387275414 -0.0614179494
       -0.0208237984 -0.6376345788
 5 -0.7221949295 0.5291913710
       0.0010306523 0.0478813818
******** ***** VALUES OF THE PRINCIPAL COMPONENTS ***********
                     V(1)
                                                                                       V( 2)
                       -0.1095584883804461D+02
                                                                                        -0.3684137088632902D+00
                      -0.1277621780691318D+02
                                                                                2 -0.5711549374131692D+00
                       -0.1088636639158975D+02
                                                                               3 -0.1311308911594027D+00
                      -0.9154394532615441D+01
                                                                               4
                                                                                       -0.8119405490473295D+00
                4
                Б
                        -0.1024152023331932D+02
                                                                                         -0.6714900809364176D+00
                6
                       -0.1173532222750920D+02
                                                                               6
                                                                                        0.26819105287359840+00
                       -0.9597303384400675D+01
                                                                                       -0.1193822078561360D+01
                                                                               8
                8
                      -0.1309397415020718D+02
                                                                                       -0.7065554935458379D+00
               9
                       -0.1478552145730417D+02
                                                                                 9
                                                                                         -0.8358495501355532D+00
                                                                              10
              10
                      -0.9928464507290487D+01
                                                                                        0.83895999249758550-01
                       -0.9445589847792558D+01
                                                                             11
                                                                                       -0.1117015714194618D+00
              11
                                                                             12
              12
                      -0.9693031858935081D+01
                                                                                       -0.1607771123805059D+00
              13
                       -0.1047145550469394D+02
                                                                              13
                                                                                        -0.4560731083644219D+00
              14
                       -0.1062207319044447D+02
                                                                              14
                                                                                         0.9132649503541419D-01
                       -0.1033337592670940D+02
                                                                              15
                                                                                       -0.3558322408807868D+00
              15
                      -0.1050531742480956D+02
                                                                             16
                                                                                       -0.5230999872863084D+00
              16
                       -0.1257441980887520D+02
                                                                              17
                                                                                         -0.5034207331926865D+00
              17
                      -0.8857152675013044D+01
                                                                              18
                                                                                       -0.4782461294909233D+00
              18
                                                                             19
                       -0.1146614062732514D+02
                                                                                       -0.9664908816771722D+00
              19
              20
                      -0.8292967215995477D+01
                                                                             20
                                                                                       0.1062143760071039D+00
                       -0.9369962434875720D+01
                                                                              21
                                                                                        -0.1082235996258010D+01
              21
              22
                      -0.9587472449555461D+01
                                                                              22
                                                                                       -0.8175395451033177D-01
                      -0.7796465582402760D+01
                                                                             23 -0.1048686945368204D+01
              23
              24
                      -0.1020611846632740D+02
                                                                              24
                                                                                       -0.1327259443636310D+01
              25
                       -0.9952227459998097D+01
                                                                              25
                                                                                        -0.2837886044528224D+00
                                                                              26 -0.4960124751256308D+00
                      -0.1198613878838233D+02
              26
```

```
27 -0.1337860765538985D+02
                                            27 -0.2071740963255488D+00
         28
             -0.1313025519510916D+02
                                          28 -0.5008657100114776D+00
             -0.1006250738816002D+02
         29
                                           29 -0.6743253643014311D+00
 COEFFICIENTS OF DETERMINATION AND CORRELATION BETWEEN COMPONENTS V(I)
                       AND ORIGINAL VARIABLES X(J)
  BETWEEN V( 1) AND ORIGINAL VARIABLES FROM X(1) TO X( 6)
X( 1) R^2= 0.00000717 R=
                               -0.00267755 SIGNIFICANT FOR ALPHA
                                                                   > 0.1
X( 2) R^2= 0.95793511 R=
                             -0.97874160 SIGNIFICANT FOR ALPHA
                                                                   < 0.001
                             -0.67198210 SIGNIFICANT FOR ALPHA < 0.001
X(3) R^2 = 0.45155994 R =
                               -0.11056160 SIGNIFICANT FOR ALPHA > 0.1
I( 4)
        R^2= 0.01222387 R=
I( 5)
        R^2= 0.96250418 R=
                               -0.98107298 SIGNIFICANT FOR ALPHA
                                                                    < 0.001
I( 6) R^2= 0.00307136 R=
                               0.05541985 SIGNIFICANT FOR ALPHA > 0.1
 TOTAL PERCENTAGE OF VARIABILITY OF ALL ORIGINAL VARIABLES
     ACCOUNTED FOR BY V( 1) MEASURE W( 1) = 39.79%
  BETWEEN V( 2) AND ORIGINAL VARIABLES FROM X(1) TO X( 6)
I( 1) R^2= 0.85289210 R=
                              0.92352158 SIGNIFICANT FOR ALPHA
                                                                   < 0.001
I( 2) R^2= 0.03537160 R=
                               -0.18807339 SIGNIFICANT FOR ALPRA > 0.1
I( 3) R^2= 0.07088949 R=
                             -0.26625080 SIGNIFICANT FOR ALPHA > 0.1
                               -0.84581245 SIGNIFICANT FOR ALPHA
                                                                   < 0.001
        R^2= 0.71539870 R=
I( 4)
                                                                    > 0.1
        R^2= 0.03225781 R=
                                0.17960461 SIGNIFICANT FOR ALPHA
X( 5)
X( 6) R^2= 0.41376644 R=
                               0.64324680 SIGNIFICANT FOR ALPHA < 0.001
TOTAL PERCENTAGE OF VARIABILITY OF ALL ORIGINAL VARIABLES
     ACCOUNTED FOR BY V( 2) MEASURE W( 2) = 35.34%
     CONSTRUCTING A DENDRITE FOR THE EUCLIDEAN DISTANCE MATRIX
             ORIGINAL VARIABLES SPACE IS CONSIDERED
.D( 1, 3) = 0.307083
.D( 3, 6) = 0.955405
..D( 6, 26) = 0.815476
....D( 26, 17) = 0.604401
....D( 17, 2) = 0.502295
....D( 17, 28) = 0.564624
.....D( 28, 8) = 0.257682
.....D( 28, 27) = 0.478748
\dotsD( 27, 9) = 1.60577
..D( 3, 14) = 0.397367
.D( 1, 13) = 0.516817
.D( 13, 5) = 0.386523
...D( 5, 24) = 0.672532
...D( 5, 29) = 0.189209

..D( 13, 15) = 0.192614

...D( 15, 25) = 0.389871
....D( 25, 10) = 0.394462
....D( 10, 12) = 0.379341
.....D( 12, 22) = 0.265895
.....D( 22, 11) = 0.210000
.....D( 11, 18) = 0.695916
.....D( 18, 4) = 0.614980
.....D( 4, 21) = 0.605806
.....D( 21, 7) = 0.513614
.....D( 18, 20) = 0.814125
.....D( 18, 23) = 1.20764
..D( 13, 16) = 0.125698
.D( 1, 19) = 0.979388
                                0.55868870
  MEAH
  STD. DEV=
                                0.32723257
                       =
  MRAE + 2*(STD.DEV)
                                1.21315384
     CONSTRUCTING A DEMORITE FOR THE EUCLIDEAN DISTANCE MATRIX
     DISTANCE MATRIX IN A COORDINATE SYSTEM WITH AXES V( 1)-V( 2) IS CONSIDERED
.D(1,3) = 0.247247
..p( 3, 14) = 0.345454
.D( 1, 16) = 0.476347
..D(16, 5) = 0.302669
...D( 5, 24) = 0.656724
....D( 24, 7) = 0.623267
.....D( 7, 21) = 0.253249
.....D( 21, 4) = 0.345730
......D( 4, 18) = 0.446883
 .....D( 18, 20) = 0.812342
.....D( 18, 23) = 1.20435
...D( 5, 29) = 0.179036
```

..D(16, 13) = 0.750947E-1

```
...D( 13, 15) = 0.170629
....D( 15, 25) = 0.387897
....D( 25, 12) = 0.286905
.....D( 12, 10) = 0.339548
.....D( 12, 22) = 0.131862
......D( 22, 11) = 0.145009
.D( 1, 19) = 0.786190
.D( 19, 26) = 0.701248
...D( 26, 6) = 0.804311

...D( 26, 17) = 0.588328

...D( 17, 2) = 0.212862

....D( 2, 8) = 0.345401
.....D( 8, 28) = 0.208865
.....D( 28, 27) = 0.384622
.....D( 27, 9) = 1.54099
                            0.46439481
                     =
                              0.33063257
 STD. DEV=
 MEAN + 2+(STD.DEV)
                              1.12565994
          SCATTERED PLOT IN V( 1)-V( 2) COORDINATE SYSTEM
     0.600+
 0.400+
 0.200+
                                                                 20
 22
                                                                 3
                                                                 12
-0.200+
                                                                 27
                                                                 25
                                                                 1
                                                                     15
-0.400+
                                                                 13
                                                                 26
                                                                     18
                                                                 28
                  1
                                                                     17
                                                                          16
                                                                  2
-0.600+
                                  11
                                                                     29
                                                                  8
-0.800+
-1.000+
                                                                 21
-1.200+
                                                                  7
                                  1
                                                                 24
```

-14.250 -12.750 -11.250 -9.750 -8.250 -6.750

PRITTO	UFDF	DI OTTEN	TE	THE	FOLLOWING	UDUED	
LOTE 12	MULL	PLUITED	1.5	1111	LOTTORING	UKDEK	:

NO.	*	X (I)	*	Y(I)	* OBIECT NO.
1	-0	. 1174E+02		0.2682E+00	6
2	-0	. 8293E+01		0.1062E+00	20
3	-0	. 1062E+02		0.9133E-01	14
4	-0	. 9928E+01		0.8390E-01	10
				-0.8175E-01	
6	-0	. 1089E+02		-0.1311E+00	3
7	-0	. 9446E+01		-0.1117E+00	11
8	-0	. 9693E+01		-0.1608E+00	12
9	-0	. 1338E+02		-0.2072E+00	27
10	-0	. 9952E+01		-0.2838E+00	25
11	-0	. 1096E+02		-0.3684E+00	1
12	-0	. 1033E+02		-0.3558E+00	15
13	-0	. 1047E+02		-0.4561E+00	13
14	-0	. 11 99E+ 02		-0.4960E+00	26
15	-0	. 8857E+01		-0.4782E+00	18
16	-0	. 1313E+02		-0.5009E+00	28
17	-0	. 1257E+02		-0.5034E+00	17
18	-0	. 1051E+02		-0.5231E+00	16
19	-0	. 1278E+02		-0.5712E+00	2
20	-0	.1024E+02		-0.6715E+00	5
21	-0	.1006E+02		-0.6743E+00	29
22	-0	. 1309E+02		-0.7066E+00	8
23	-0	.9154E+01		-0.8119E+00	4
24	-0	. 1479E+02		-0.8358E+00	9
25	-0	. 1147E+02		-0.9665E+00	19
26	-0	. 7796E+01		-0.1049E+01	23
27	-0	.9370E+01		-0.1082E+01	21
28	-0	.9597E+01		-0.1194E+01	7
29	-0	. 1021E+02		-0.1327E+01	24

#UMBER OF VARIABLES = 6 #UMBER OF CASES= 29

CORRELATION HATRIX WAS TAKEN INTO ACCOUNT

<

Now the same analysis as for the variance-covariance matrix will be done.

REFERENCES

Anderson, T. W., 1958, An introduction to multivariate statistical analysis: John Wiley & Sons, New York, 675 p.

Anderson, T. W., 1984, An introduction to multivariate

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Jackson, J. E., and Hearne, F. T., 1973, Relationships among coefficients of vectors used in principal components: Technometrics v. 15, no. 3, p. 601-610.

Kendall, M. G., 1938, The geographical distribution of crop productivity in England: Roy. Statist. Soc., v. 102, p. 21-48.

Mardia, K. V. Kent, J. T., and Bibby, J. M., 1979, Multivariate analysis: Academic Press, London, 521 p.

APPENDIX

FORTRAN Program Listing

PROGRAM PCA

***	*********	**********	*****		
*	PRINCIPAL COMPONE	ENTS ANALYSIS PROGRAM	*		
*	LANGUAGE: F	ORTRAN77	*		
*	ВУ		*		
*	ANDRZEJ MACKIEWICZ	WALDEMAR RATAJCZAK	*		
*	TECHNICAL UNIVERSITY	ADAM MICKIEWICZ UNIVERSITY	*		
*	POZNAN, POLAND				
*			*		
*	FLOPPY DISK DEL	IVERED BY REQUEST	*		
***	*********	*********	*****		
*	MAXIMUM NUMBER OF VARIABLES	IS SET AT 50;	*		
*	MAXIMUM NUMBER OF CASES IS	SET AT 200:	*		

```
FOR CHANGING SIZES REPLACE 50 AND 200 IN THE PARAMETER
  STATEMENT THREE LINES BELOW.
  INTEGER I, J, J1, K, L, LA, LB, N, M, NM, MM, LL, IERR, PRMODE
  PARAMETER (NM=50, MM=200)
   ----- CHANGED ?.
  PARAMETER (PRMODE=1)
    ----- CONVERT THIS PARAMETER TO O IF YOUR
                 PRINTER CAN PLOT ONLY 133 CHARACTERS PER LINE
                 (STANDARD PRMODE VALUE IS 1)
  REAL
                   RINFM, RXXXM, RINF
  DOUBLE PRECISION XINF, XMIN, XXXF, XXXN
  PARAMETER (RXXXM= 3.35E+38, XXXF= 1.7D+308, XXXN= 2.3D-308)
   ----- ^ ----- ^ ----- ^ ----- ^ -CHANGED ?.
  VAX CONSTANTS: 1.7E+38
                                  1.7D+38
                                                 5.9D-39
  THESE PARAMETERS ARE MACHINE-DEPENDENT (HERE THEY HAVE BEEN
  SET FOR THE IBM PC OR ANY OTHER MACHINE EQUIPPED WITH THE
  INTEL 8087, 80287, 80387 OR COMPATIBLE. SUGGESTIONS FOR THE VAX
   ARE PRESENTED TWO LINES BELOW THE LAST PARAMETER STATEMENT).
   RXXXM - IS THE LARGEST VALUE POSITIVE (NORMALIZED) REAL DATA.
  XXXF - IS THE LARGEST VALUE POSITIVE (NORMALIZED) DOUBLE
         PRECISION DATA.
  XXXN - IS THE SMALLEST VALUE POSITIVE (NORMALIZED) DOUBLE
          PRECISION DATA.
  TO ALTER THESE PARAMETERS FOR A PARTICULAR ENVIRONMENT
   CHECK THE MANUAL OF IT OR SEE THE FOLLOWING PAPER:
   FOX P.A., HALL A.D., SCHRYER N.L., *FRAMEWORK FOR A
  PORTABLE LIBRARY*, ACM TRANSACTIONS ON MATHEMATICAL
   SOFTWARE, VOL. 4, NO. 2, JUNE 1978, PP. 177-188.
  INTEGER
                  BB(MM), PROC(NM), H1(MM), R1(MM), ISORT(MM), NSTEPY
   DOUBLE PRECISION A(NM,NM),D(NM),E(NM),X(NM),S(NM),AC(NM)
  DOUBLE PRECISION AS1, TRACE, BX, DX, R, SX1, LCO(NM)
  DOUBLE PRECISION U(MM), V(MM), RCO(NM)
  REAL
                 TQUANT, Y(MM, MM), WDIS(MM*(MM-1)/2), DD(MM), P(4)
  LOGICAL
  CHARACTER
                 INSET*60,OUTSET*60,STR*9,YON*1
  EQUIVALENCE
                 (Y,WDIS),(DD,X),(E,U),(S,V)
                  RINFM, XINF, XMIN
  COMMON
  RINFM= -RXXXM
  XINF= XXXF
  XMIN= XXXN
  WRITE(*,'(12X,''PRINCIPAL COMPONENTS ANALYSIS-INTERACTION PROGRAM
  & >>//\)>)
  & '',///)')
               ****** GIVE THE PATH NAME OF THE INPUT FILE - AT
  WRITE(*,'(''
 &MOST 60 CHARACTERS *******'/)')
  READ(*,5) INSET
               ****** GIVE THE NAME OF THE OUTPUT FILE - AT MOST
  WRITE(*,'(''
 &60 CHARACTERS *******')')
  READ(*,5) OUTSET
5 FORMAT (A)
  OPEN(2,STATUS='UNKNOWN',FILE=OUTSET)
```

```
WRITE(2,'(22X, ''* PRINCIPAL COMPONENTS AWALYSIS *'')')
    WRITE(2,'(/, ''* '', '' FOR DATA FILE '', A60,''* '')') INSET
    ST=.FALSE.
10
   IF (ST) THEN
       OPEN(1,STATUS='OLD',FILE=INSET)
       OPEN(1,STATUS='UNKNOWN',FILE=INSET)
    ENDIF
    READ(1,*) N
    IF (N.GT.NM) THEN
       WRITE(2,'(/9X,'' MAXIMUM NUMBER OF VARIABLES IS EQUAL TO NM''
   &/'' ARRAY SIZE IN THE CALLING PROGRAM MUST BE CHANGED FOR YOUR''/
        ,21X,''DATA BEFORE USE ''/)')
       STOP
    ENDIF
    READ(1,*) M
    IF (M.GT.MM) THEN
       WRITE(2,'(/9x,'' MAXIMUM NUMBER OF CASES IS EQUAL TO MM''
   &/'' ARRAYS SIZE IN THE CALLING PROGRAM MUST BE CHANGED FOR YOUR''/
        ,21X,''DATA BEFORE USE ''/)')
       STOP
    ENDIF
    IF(M.LT.3) THEN
       WRITE(2,'(/15X,''TOO FEW CASES. MINIMUM NUMBER OF CASES IS EQUA
   &L TO 3''/)')
       STOP
    ENDIF
    WRITE (2,'(//)')
    WRITE(2,'('' NUMBER OF VARIABLES = '', 13,33%,''NUMBER OF CASES=''.
   &13)') N,M
    IF (M.LE.300) THEN
       NSTEPY≈50
       ELSE IF (M.LE.600) THEN
       NSTEPY=100
       ELSE
       NSTEPY=200
    ENDIF
    WRITE(2,*)
    WRITE(2,*) ('<>',I=1,39)
    IF(ST) THEN
    WRITE(2,'(/19X,''CORRELATION MATRIX WAS TAKEN INTO ACCOUNT''/)')
    WRITE(2,'(/17X,''VAR.-COVARIANCE MATRIX WAS TAKEN INTO ACCOUNT''/
   £)')
    ENDIF
    WRITE(2,*) ('<>',I=1,39)
    DETERMINING THE VARIANCE-COVARIANCE MATRICES
    CALL COVM (NM,M,N,S,A,D,E)
    WRITE(2,*)
    WRITE(2,'(17X,''DESCRIPTIVE PARAMETERS OF ORIGINAL ''.
   &''VARIABLES ''/)')
    WRITE(2,'(4X,''VARIABLE'',8X,''MEAN'',13X,''VARIANCE'',7X,
```

```
&''STAND. DEV.''/)')
    DO 20 I=1,N
    WRITE(2,'(4X,''X('',I3,'')'',3X,E16.6,3X,2E16.7)') I,S(I),A(I,I),
   & DSQRT(A(I,I))
20
   AC(I)=A(I,I)
     OPTIONAL DETERMINATION OF THE CORRELATION MATRIX
     IF (ST) THEN
        DO 40 I=2,N
           DO 30 J=1,I-1
              A(I,J)=A(I,J)/DSQRT(A(I,I)*A(J,J))
 30
           CONTINUE
 40
      CONTINUE
        DO 50 I=1,N
           A(I,I)=1
50
        CONTINUE
         WRITE(2,'(//16X,', ##### CORRELATION MATRIX #####'')')
         WRITE(2,'(//16X,'' ##### VAR.-COVARIANCE MATRIX #####'')')
     ENDIF
      PRINTING THE CORRELATION MATRIX AND THE VAR.-COVAR MATRIX
     DO 70 I=1,N
       DO 60 J=1,I
          A(J,I)=A(I,J)
60
       CONTINUE
70
    CONTINUE
     K=N
     DO 110 L=0,K-1,5
        WRITE(2,*)
        J1=MINO(L+5,K)
        WRITE(2,120)(I,I=L+1,J1)
        WRITE(2,*)
        DO 100 J=1,N
           WRITE(2,130) J,(A(J,I),I=L+1,J1)
100
        CONTINUE
110 CONTINUE
120 FORMAT (5X,5(3X:,'COL(',12,')',5X))
130 FORMAT (I3,5(1X:,D14.6))
     DETERMINING EIGENVECTORS AND EIGENVALUES
CALL TRED2(NM,N,A,D,E,A)
     CALL TQL2(NM, N, D, E, A, IERR)
     IF (IERR.GT.O) THEN
        WRITE(2,*) 'QL ALGORITHM FAILS - FUTURE PROGRESS IMPOSSIBLE'
        STOP
     ENDIF
     TRACE OF THE MATRIX CONSIDERED IS NOW DETERMINED
     TRACE=0
     DO 140 I=1,N
```

```
TRACE=TRACE+DMAX1(ODO,D(I))
 140 CONTINUE
    BARTLET'S TEST
    IF (.NOT.ST) THEN
      CALL BARTLE (N.M.D.TRACE, PROC.LCO, RCO)
      ELSE
      CALL BART2(N,M,D)
    ENDIF
   PRINTING OF EIGENVALUES
    WRITE(2,'(5X,'' NR. '',15X,'' LAMBDA'',8X,
   & ''PERCENTAGE OF TRACE''/)')
    DO 150 I=1.N
       L=N-I+1
       IF (D(L).LE.ODO) THEN
           GOTO 170
           ELSE
           WRITE(2,160) I,D(L),SNGL(100D0*D(L)/TRACE)
        ENDIF
 150 CONTINUE
 160 FORMAT(5X,13,10X,E22.16,7X,F6.2)
    PRINTING EIGENVALUES ON THE SCREEN
 170 IF(ST) THEN
    WRITE(*,'(/10X,''+++ CORRELATION MATRIX IS CONSIDERED +++ ''/)')
    WRITE(*,'(/10X,''- VAR.-COVARIANCE MATRIX IS CONSIDERED -''/)')
    ENDIF
    WRITE(*,'(/10X,'' ******* FIRST NO-ZERO EIGENVALUES *******')')
    WRITE(*,'(5X,'' NR. '',15X,'' LAMBDA'',8X,
    & ''PERCENTAGE OF TRACE''/)')
    DO 180 I=1,AMINO(10,N)
       L=N-I+1
       IF (D(L).LE.ODO) THEN
           GOTO 190
           WRITE(*,160) I,D(L),SNGL(100D0*D(L)/TRACE)
        ENDIF
180 CONTINUE
190 WRITE(*,'(//1X,'' GIVE THE NUMBER OF THE PRINCIPAL COMPONENTS YOU
   &WANT TO CONSIDER ''/)')
    READ(*,*) K
   END OF MONITORING
    IF(K.GT.O) THEN
       IF(ST) THEN
          WRITE(2,*) ('.',I=1,78)
          WRITE(2,'(/5X, '' SELECTED EIGENVECTORS (FROM '', I1,
          '' TO'', I3, '' ) OF THE CORRELATION MATRIX'')') 1, K
```

```
ELSE
          WRITE(2,'(/5x, '' SELECTED EIGENVECTORS (FROM ''.11.
          '' TO '', I3, '' ) OF THE VAR.-COVARIANCE MATRIX'')') 1,K
       ENDIF
       PRINTING THE EIGENVECTORS
       DO 230 L=0.K-1.5
          WRITE(2,*)
          J1=MINO(L+5,K)
          WRITE(2,240) (I,I=L+1,J1)
          WRITE(2,*)
          DO 220 J=1,N
             WRITE(2,250) J, (A(J,N-I+1),I=L+1,J1)
220
          CONTINUE
230
       CONTINUE
    ENDIF
240 FORMAT (5(10X:,'A(',I2,')'))
250 FORMAT (I3,5(1X:,F14.10))
    WRITE(2,'(//)')
    WRITE(2,'('' ************** VALUES OF THE PRINCIPAL COMPONENTS
    DO 280 I=1,K
       CLOSE(1,STATUS='KEEP')
       OPEN(1,STATUS='OLD',FILE=INSET)
       READ(1,*) N
       READ(1.*) M
       WRITE(2,*)
       WRITE(2,'(12X,'' V('',I2,'')''/)') I
       DO 270 J=1,M
          READ(1,*) (X(LL),LL=1,N)
          BX=ODO
          DO 260 LA=1,N
             BX=BX+X(LA)*A(LA,N-I+1)
260
          CONTINUE
          WRITE(2, '(8X, 13, 3X, D23.16)') J, BX
270
      CONTINUE
280 CONTINUE
    DETERMINATION AND CORRELATION COEFFICIENTS ARE NOW CALCULATED
    WRITE(2,*)
    WRITE(2,*) ('.', I=1,72)
    WRITE(2,*) ' COEFFICIENTS OF DETERMINATION AND CORRELATION BETWE
   &EN COMPONENTS V(I) '
    WRITE(2,'(23X, '' AND ORIGINAL VARIABLES X(J)'')')
    WRITE(2,*)
    J1=M-2
    P(1)≈0.1
    P(2)=0.05
    P(3)=0.01
    P(4)=0.001
    DO 320 I=1,K
    WRITE(2,*)
    WRITE(2,'('' BETWEEN V('', I3,'') AND ORIGINAL VARIABLES FROM X(1)
```

```
& TO X('', I3, '')'') I, N
 WRITE(2,*)
 AS1=0
    DO 310 J=1,N
       R=A(J,N-I+1)*A(J,N-I+1)*D(N-I+1)
       IF (.NOT.ST) R=R/AC(J)
       AS1=AS1+R
       DX=DSIGN(DSQRT(R),A(J,N-I+1))
       DO 290 LA=1,4
          CALL TQ(J1,P(LA),TQUANT)
          BX=TQUANT*TQUANT
          TQUANT=SQRT(BX/(BX+J1))
          IF (DABS(DX).LE.TQUANT) THEN
             L=LA-1
             GOTO 300
          ENDIF
290
          CONTINUE
          L=5
300
          IF (L.EQ.O) THEN
              STR=' > 0.1 '
              ELSEIF(L.EQ.1) THEN
              STR=' = .1 '
              ELSEIF(L.EQ.2) THEN
              STR=' = 0.05 '
              ELSEIF(L.EQ.3) THEN
              STR=' = 0.01 '
              ELSEIF(L.EQ.4) THEN
              STR=' = 0.001'
              ELSE
              STR=' < 0.001'
          WRITE(2,'(''X('',I3,'') R^2='',F11.8,'' R= '',F14.8, '' S
    &IGNIFICANT FOR ALPHA '',1X,A9)') J,R,DX,STR
310
        CONTINUE
     WRITE(2,*)
     WRITE(2,'('' TOTAL PERCENTAGE OF VARIABILITY OF ALL ORIGINAL VARIA
    &BLES'')')
                    ACCOUNTED FOR BY V('', 13,'') MEASURE W('', 13,''
     WRITE(2,'(''
    &) ='', F6.2, ''%'')') I,I,AS1*100.0/DBLE(N)
320 CONTINUE

    EUCLIDEAN DISTANCES (IN ORIGINAL VARIABLE SPACE) ARE NOW DETERMINED

*------
     WRITE(2,*)
     WRITE(2,*) ('=',I=1,78)
     IF(.NOT.ST) THEN
        CLOSE(1,STATUS='KEEP')
        OPEN(1,STATUS='OLD',FILE=INSET)
        READ(1,*) N
        READ(1,*) M
        DO 330 J=1,M
          READ(1,*) (Y(LL,J),LL=1,N)
330
        CONTINUE
        LL=1
        DO 370 I=1,M-1
```

```
DO 340 K=1,N
              X(K)=Y(K,I)
340
           CONTINUE
           DO 360 J=I+1,M
              SX1=0.0
              DO 350 K=1,N
                SX1=SX1+(X(K)-Y(K,J))**2
350
              CONTINUE
              Y(J,I)=SQRT(SX1)
360
           CONTINUE
           Y(I,I)=0.0
370
        CONTINUE
        DO 390 I=2,M
           DO 380 J=1,I-1
             Y(J,I)=Y(I,J)
380
           CONTINUE
390
        CONTINUE
        DO 410 J=2.M
          DO 400 I=1,J-1
             WDIS(LL)=Y(I,J)
             LL=LL+1
400
          CONTINUE
410
        CONTINUE
        CLOSE(1,STATUS='KEEP')
        WRITE(2,'(/5X,''CONSTRUCTING A DENDRITE FOR THE EUCLIDEAN DISTA
    ANCE MATRIX'')')
        WRITE(2,'(/15X,''ORIGINAL VARIABLES SPACE IS CONSIDERED'',/)')
        RINF=RXXXM+1E-1
        CALL DENDRI (M, WDIS, RINF, 1, 1, BB, DD, H1, R1, LL, PRMODE)
        CLOSE(1,STATUS='KEEP')
    ENDIF
    DISTANCE MATRIX IN COORDINATE SYSTEM V(1)-V(2) IS NOW COMPUTED
    CLOSE(1,STATUS='KEEP')
420 CLOSE(2,STATUS='KEEP')
    WRITE(+.+)'EXAMINE YOUR OUTPUT FILE IN ORDER TO CHOOSE A NEW COORD
    AINATE SYSTEM'
    WRITE (*,*)'THEN PRESS <ENTER>'
    WRITE(*,*) 'IT IS USUALLY POSSIBLE TO EXECUTE ANY OS COMMAND NOW'
    PAUSE
415 WRITE(*,*)'TYPE CAPITAL Y IF YOU WANT TO CONSIDER A NEW PRINCIPAL
    & COMPONENT '
     WRITE(*,*) 'COORDINATE SYSTEM OR TYPE CAPITAL N
    READ(*,425) YON
    IF(YON.NE.'Y' .AND. YON.NE.'N') GOTO 415
425 FORMAT(A1)
    OPEN(2,STATUS='OLD',ACCESS='APPEND',FILE=OUTSET)
    IF(YON.EQ.'N') GOTO 490
430 WRITE(*,*) 'GIVE THE NUMBER OF THE FIRST COORDINATE AXIS '
    READ(*,*) LA
    WRITE(*,*) 'GIVE THE NUMBER OF THE SECOND COORDINATE AXIS '
    READ(*,*) LB
    IF(LA.GT.N.OR.LB.GT.N) THEN
        WRITE (*,*) 'DATA ERROR - TRY AGAIN'
```

```
GO TO 430
     ENDIF
     OPEN(1,STATUS='OLD',FILE=INSET)
     READ(1,*) N
     READ(1,*) M
     DO 440 J=1,M
        READ(1,*) (Y(J,LL),LL=1,N)
440 CONTINUE
     CLOSE(1,STATUS='KEEP')
     DATA FOR PLOTTING
     LA=N+1-LA
     LB=N+1-LB
     DO 460 J=1,M
        BX=ODO
        DX=ODO
        DO 450 K=1.N
           BX=BX+DBLE(Y(J,K))*A(K,LA)
           DX=DX+DBLE(Y(J,K))*A(K,LB)
 450
        CONTINUE
        V(J)=DX
        U(J)=BX
 460 CONTINUE
     LL=1
     DO 480 I=2,M
       DO 470 J=1,I-1
          WDIS(LL)=SNGL(DSQRT((U(J)-U(I))**2+(V(J)-V(I))**2))
          LL=LL+1
470
       CONTINUE
480 CONTINUE
     WRITE(2,'(/5x,''CONSTRUCTING A DENDRITE FOR THE EUCLIDEAN DISTANCE
    & MATRIX''))
     WRITE(2,'(/5X,''DISTANCE MATRIX IN A COORDINATE SYSTEM WITH AXES
    &V('', I2,'')-V('', I2,'') IS CONSIDERED'',/)') N+1-LA, N+1-LB
     RINF=RXXXM+1E-1
     CALL DENDRI(M, WDIS, RINF, 1, 1, BB, DD, H1, R1, LL, PRMODE)
     WRITE(2,*)
     WRITE(2,'(/17X,'' SCATTERED PLOT IN V('', I2,'')-V('', I2,'') COORDI
    AWATE SYSTEM''/)') N+1-LA,N+1-LB
     CALL PLOTDR (U, V, M, ISORT, 60, NSTEPY, PRMODE)
     WRITE(2,'(//)')
     WRITE(2,'(14X,''POINTS WERE PLOTTED IN THE FOLLOWING ORDER :'')')
     WRITE(2,'(12X,60A1)') ('-',I=1,47)
     WRITE(2,'(12X,'' NO. * X(I)
                                         Y(I) * OBJECT NO.'')')
     WRITE(2,'(12X,60A1)') ('-',I=1,47)
     DO 485 I=1,M
        WRITE(2,'(12X,I3,2X,E11.4,3X,E11.4,6X,I3)')
                        I,U(ISORT(I)),V(I),ISORT(I)
485 CONTINUE
     WRITE(2,*)
    GOTO 420
    PRINCIPAL COMPONENTS ANALYSIS FOR THE CORRELATION MATRIX OR STOP
```

```
490 WRITE(2,*) ('=',I=1,78)
     IF(.NOT.ST)THEN
        ST=.TRUE.
        CLOSE(1,STATUS='KEEP')
        GOTO 10
     ENDIF
     END
SUBROUTINE TQ(N,P,TQUANT)
***********
     STUDENT'S T-QUANTILES
************
     INTEGER N
     REAL
            P, TQUANT, HP, A, B, C, D, X, Y
     IF (N.LT.1 .OR. P.LT..O .OR. P.GT.1.0) THEN
        TQUANT=0
        RETURN
     ENDIF
     HP=1.57077963268
     IF (N.EQ.1) THEN
        X=P*HP
        TQUANT=COS(X)/SIN(X)
     ELSEIF (N.EQ.2) THEN
        TQUANT=SQRT(2.0/(P*(2.0-P))-2.0)
     ELSE
        A=1.0/(N-.5)
        B=48.0/(A*A)
        C=((20700.0*A/B-98.0)*A-16.0)*A+96.36
        D=((94.5/(B+C)-3.0)/B+1.0)*SQRT(A*HP)*H
        X=D*P
        Y = X * * (2.0/N)
        IF(Y.GT..O5+A) THEN
           IF(ABS(P-.1).LT.1D-7) THEN
             X=-1.644854
          ELSEIF(ABS(P-.05).LT.1D-7) THEN
             X=-1.959964
          ELSEIF(ABS(P-.01).LT.1D-7) THEN
             X=-2.575829
          ELSEIF(ABS(P-.001).LT.1D-7) THEN
             X=-3.290527
          ELSE
             X=1D30
          ENDIF
          Y=X*X
          IF(N.LT.5) C=C+.3*(N-4.5)*(X+.6)
          C=(((.05*D*X-5.0)*X-7.0)*X-2.0)*X+B+C
          Y = (((((.04*Y+6.3)*Y+36.0)*Y+94.5)/C-Y-3.0)/B+1.0)*X
          Y=A+Y+Y
          IF (Y.GT..1) THEN
             Y=EXP(Y)-1.0
          ELSE
             Y=((Y+4.0)*Y+12.0)*Y*Y/24.0+Y
          ENDIF
       ELSE
          Y=((1.0/((N+6.0)/(N+Y)-.089+D-.822)+(N+2.0)+3.0)
```

```
Ł
           +.5/(N+4.0))*Y-1.0)*(N+1.0)/(N+2.0)+1.0/Y
       ENDIF
       TQUANT=SQRT(N*Y)
    ENDIF
     END
SUBROUTINE DENDRI (N,D,INF,S,PR,B,C,H,R,ST,PRMODE)
******************
    PLOTTING THE MINIMUM SPANNING TREE
*******************
     INTEGER N, PR, ST, PRMODE
     INTEGER B(N), H(N), S1, S2, I, J, K, K1, L, M1, R(N), S, II, JJ
     REAL
             C(N),D((N+(N-1))/2),DI,INF,MIN,SEL,MEAN,STDEV
     CHARACTER LINOUT+200, SHORT+133
     MEAN=0.0
     STDEV=0.0
     SEL=ODO
     IF (N.LT.1) THEN
       ST=2
       RETURN
     ENDIF
     IF (PRMODE.EQ.1) THEM
       JJ=200
       ELSE
       JJ=133
     ENDIF
     ST=0
     INF=INF+1.0E0
     DO 10 I=1.N
       H(I)=I
       C(I)=INF
       B(I)=0
  10 CONTINUE
     H(S)≈N
     S1=S
     DO 30 I=N-1,1,-1
       MIN=INF+1EO
       S2=((S1-1)*(S1-2))/2
       DO 20 K=1,I
          J=H(K)
          IF (S1.GT.J) THEN
            L=S2+J
            ELSE
            L=((J-1)*(J-2))/2+S1
          ENDIF
          DI=D(L)
          IF (DI.LT.C(J)) THEN
            C(J)=DI
            B(J)=S1
            ELSE
            DI=C(J)
          ENDIF
          IF (DI.LT.MIN) THEN
            MIN=DI
            K1≈K
```

```
ENDIF
20
       CONTINUE
       S1=H(K1)
       H(K1)=H(I)
30 CONTINUE
   DO 40 I=1,N
      IF (B(I).EQ.0) GOTO 40
      IF (C(I).LT.INF-1EO) THEN
          SEL=SEL+C(I)
          ELSE
         SEL=INF-1E0
          ST=1
      ENDIF
40 CONTINUE
   IF (PR.EQ.O.OR.S.NE.1) RETURN
   DO 50 I=1,N
      H(I)=0
50 CONTINUE
   DO 60 I=2, M
      J=B(I)
      H(J)=H(J)+1
60 CONTINUE
   R(1)=1
   J=1
   K=1
   DO 100 I=2,N
      IF (H(K).EQ.O) THEM
         J=J-1
         K=R(J)
         GOTO 70
      ELSE
         H(K)=H(K)-1
         DO 80 M1=2,N
            IF (K.EQ.B(M1)) GOTO 90
80
         CONTINUE
90
         IF (PRMODE.EQ.1) THEM
             WRITE(LINOUT(1:),'(200A)') ('.', II=1,200)
             IF (J.LE.JJ-25) THEN
                WRITE(LINOUT(J+1:),'(''D('',I4,'','',I4,'') = '',
                G12.6E1)') K,M1,C(M1)
             ELSE
                WRITE(LINOUT((JJ-25):),'(A)')' SORRY, LINE TOO LONG'
             ENDIF
             WRITE(2,'(200A)') LINOUT
         ELSE
             WRITE(SHORT(1:),'(133A)') ('.',II=1,133)
             IF (J.LE.JJ-25 ) THEN
                WRITE(SHORT(J+1:),'(''D('',I4,'','',I4,'') = '',
                G12.6E1)') K,M1,C(M1)
             ELSE
                WRITE(SHORT((JJ-25):),'(A)') ' SORRY, LINE TOO LONG '
             WRITE(2,'(133A)') SHORT
         ENDIF
         MEAN=MEAN+C(M1)
```

```
J=J+1
          K=M1
          R(J)=K
          B(M1)=-B(M1)
       ENDIF
 100 CONTINUE
     DO 110 I=2.N
        B(I)=IABS(B(I))
 110 CONTINUE
     MEAN=MEAN/FLOAT(N-1)
     DO 120 I=2,N
       SEL=(C(I)-MEAN)
       STDEV=STDEV+SEL*SEL
 120 CONTINUE
     STDEV=SQRT(STDEV/FLOAT(N-1))
     WRITE(2,*)
     WRITE(2,*)' MEAN
                                  = ', MEAN
     WRITE(2,*)' STD. DEV=
                                  = ', STDEV
     WRITE(2,*)' MEAN + 2*(STD.DEV)
                                  = ', MEAN+2*STDEV
     END
SUBROUTINE COVM (MN,M,N,S,A,D,E)
********************
     CALCULATION OF MEANS AND THE VARIANCE-COVARIANCE MATRIX
************************
     INTEGER
                       MN, M, N, I, J, N1, I1, J1
     DOUBLE PRECISION
                      DR,DIJ,DIV
     DOUBLE PRECISION
                       D(N), E(N), S(N), A(MN, N)
     DO 20 I=1,N
       D(I)=0D0
       S(I)=ODO
       DO 10 J=1,N
          A(I,J)=ODO
        CONTINUE
 10
     CONTINUE
20
 30
     N1=0
     DO 60 I=1,M
        READ(1,*) (E(J),J=1,N)
        N1=N1+1
       DIV=1DO/DBLE(N1)
       DO 50 I1=1,N
          DR=E(I1)-S(I1)
          D(I1)=DR
          S(I1)=S(I1)+DR*DIV
          DO 40 J1=1.I1
             DIJ=DR*D(J1)
             A(I1,J1)=A(I1,J1)+DIJ-DIJ*DIV
          CONTINUE
40
 50
        CONTINUE
     CONTINUE
 60
     DO 80 I=1,N
        DO 70 J=1,I
          A(I,J)=A(I,J)/DBLE(M)
 70
        CONTINUE
     CONTINUE
 80
     END
```

```
SUBROUTINE BART2(N,M,D)
***************
    BARTLET'S TEST FOR THE COVARIANCE MATRIX
****************
     INTEGER
                  N,M,I,IER,J,K1,P
     REAL
                  CHIPRO
    DOUBLE PRECISION AO, AOO, D(N)
     LOGICAL
                  LOGI
    LOGI=.FALSE.
     P≃0
     DO 10 I=N,1,-1
       IF (D(I).GT.1D-13) P=P+1
     CONTINUE
 10
     J≈MINO(N-2,P)
     DO 30 K1=0,J
       AO=ODO
       A00≈1D0
       DO 20 I=K1+1,N
          AO=AO+D(N-I+1)
          A00=A00*DABS(D(N-I+1))
 20
       CONTINUE
       AO=AO/(N-K1)
       A00=A00**(1.D0/(N-K1))
       IF(AO/AOO.GE.1D30) THEN
         LOGI=.TRUE.
         GOTO 40
       ENDIF
       CALL CDTR(SNGL((M-1)*(P-K1)*DLOG(AO/AOO)),
               REAL((P-K1+2)*(P-K1-1)*0.5), CHIPRO, IER)
       IF(1.0-CHIPRO.GT.O.O5.OR.IER.NE.O) GOTO 40
30
    CONTINUE
    WRITE(2,*)
40
    WRITE(2,*) ('*', I=1,80)
    WRITE(2.*)
    WRITE(2,'(20X,''TESTING THE PRINCIPAL COMPONENTS '')')
    WRITE(2,'(25X,''(FOR CORRELATION MATRIX)'')')
    WRITE(2.*)
    IF(IER.NE.O.OR.LOGI) THEN
       WRITE (2,*) 'BARTLET''S TEST CAN NOT BE USED HERE '
    ENDIF
    WRITE(2,*)
    IF (K1.LT.N) THEN
       WRITE(2,*) ' BARTLET''S TEST REJECTS EIGENVECTORS NO. '
       DO 50 I=K1+1.N
         WRITE(2,'(5X,I3)') I
       CONTINUE
50
    ENDIF
    WRITE(2,*)
    END
SUBROUTINE BARTLE(N, M, D, TRACE, PROC, LCO, RCO)
**************
             BARTLET'S TEST
*****************
```

```
INTEGER
                       N,M,K,I,IER,J,L,K1,PROC(N)
     REAL
                       CHIPRO
     DOUBLE PRECISION AO, AOO, DELTA, TR1, TR2, TR3, TRACE, D(N), LCO(N), RCO(N)
     LOGICAL LOGI
     LOGI=.FALSE.
     DO 20 K1=0,N-2
         A0=ODO
         A00=1D0
        DO 10 I=K1+1,N
           AO=AO+DABS(D(N-I+1))
           A00=A00*DABS(D(N-I+1))
10
        CONTINUE
         AO=AO/(N-K1)
         A00=A00**(1.D0/(N-K1))
        IF(AO/AOO.GE.1D30) THEN
           LOGI=.TRUE.
            GOTO 30
        ENDIF
        CALL CDTR(SNGL((M-(2*N+11)/6D0)*(N-K1)*DLOG(AO/AOO)).
              REAL((N-K1+2)*(N-K1-1)*0.5), CHIPRO, IER)
        IF (IER.NE.O) GOTO 30
        IF(1.0-CHIPRO.GT.0.05) GOTO 30
     CONTINUE
20
30
     IF(LOGI.OR.IER.NE.O) THEN
        K1=N
     ENDIF
   TESTING PRINCIPAL COMPONENTS
     DO 40 I=K1+1,N
        PROC(N-I+1)=0
        LCO(N-I+1)=ODO
        RCO(N-I+1)=ODO
40
     CONTINUE
     DO 80 K=0,K1-1
        TR1=0
        TR2=0
        TR3=0
        DO 50 I=1,K
          TR1=TR1+D(N-I+1)**2
50
        CONTINUE
        DO 60 I=K+1,N
           TR2=TR2+D(N-I+1)**2
           TR3=TR3+D(N-I+1)
60
        CONTINUE
        DO 70 L=0,100
          DELTA=0.01*DBLE(L)
       IF(DSQRT(DBLE(M))*(TR3/TRACE-DELTA)-1.96*DSQRT(DABS(2*(DELTA
    Ł
             /TRACE)) ** 2 * TR1 + 2 * ((1 - DELTA) / TRACE) ** 2 * TR2) . LT.0) THEN
          PROC(N-K)=L
          GOTO 80
       ENDIF
70
        CONTINUE
        PROC(N-K)=100
     CONTINUE
80
```

```
CONFIDENCE INTERVAL IS NOW DETERMINED
     TR1=ODO
     DO 90 I=1,N
       TR1=D(I)*D(I)+TR1
90
    CONTINUE
     DO 110 K=1,K1
       DELTA=ODO
       TR2=ODO
       DO 100 J=1,K
          TR2=TR2+D(N-J+1)**2
          DELTA=DELTA+D(N-J+1)
100
       CONTINUE
       TR3=DELTA/TRACE
       A00=TR2/TR1
       A0=(2D0*TR1*(TR3**2-2D0*TR3*A00+A00**2))/((M-1)*TRACE**2)
       RCD(N-K+1)=1.96*DSQRT(DABS(AO))
       LCO(N-K+1)=TR3-RCO(N-K+1)
       RCO(N-K+1)=TR3+RCO(N-K+1)
110 CONTINUE
     WRITE(2,*)
     WRITE(2,*) ('*', I=1,80)
     WRITE(2.*)
     WRITE(2,'(20X,''TESTING THE PRINCIPAL COMPONENTS '')')
     WRITE(2,'(20X,''(FOR VARIANCE-COVARIANCE MATRIX)'')')
     WRITE(2.*)
     IF(LOGI.OR.IER.NE.O) THEN
       WRITE (2,*) ' BARTLET''S TEST CAN NOT BE USED HERE '
     WRITE(2,*)'___ NO.__CRITICAL DELTA__ LEFTCONFID.___ RIGHTCONFID.'
     WRITE(2,*)
     DO 120 I=N,1,-1
        WRITE(2,'(18,2X,15,6X,F12.2,4X,F17.2)')
    &n-I+1, PROC(I), IDNINT(LCO(I)*1000)/10D0, IDNINT(RCO(I)*1000)/10D0
120 CONTINUE
     ENDIF
     WRITE(2,*)
     IF (K1.LT.N) THEN
       WRITE(2,*) ' BARTLET''S TEST REJECTS EIGENVECTORS NO. '
       DO 130 I=K1+1,N
          WRITE(2,'(5X,13)') I
130
       CONTINUE
     ENDIF
     WRITE(2,*)
     END
SUBROUTINE TQL2(NM,N,D,E,Z,IERR)
*******************
     THIS SUBROUTINE FINDS THE EIGENVALUES AND EIGENVECTORS *
     OF A SYMMETRIC TRIDIAGONAL MATRIX BY THE QL METHOD.
          It was taken from the LINPACK library
******************
     INTEGER I, J, K, L, M, N, II, L1, L2, NM, MML, IERR
     DOUBLE PRECISION D(N), E(N), Z(NM, N)
```

```
DOUBLE PRECISION C, C2, C3, DL1, EL1, F, G, H, P, R, S, S2, TST1, TST2, PYTHAG
   IERR = 0
   IF (W .EQ. 1) GO TO 1001
   DO 100 I = 2, N
100 E(I-1) = E(I)
   F = 0.0D0
   TST1 = 0.0D0
   E(H) = 0.000
   DO 240 L = 1, N
       J = 0
       H = DABS(D(L)) + DABS(E(L))
       IF (TST1 .LT. H) TST1 = H
       DO 110 M = L, N
          TST2 = TST1 + DABS(E(M))
          IF (TST2 .EQ. TST1) GO TO 120
110
       CONTINUE
       IF (M .EQ. L) GO TO 220
120
130
       IF (J .EQ. 30) GO TO 1000
       J = J + 1
       L1 = L + 1
       L2 = L1 + 1
       G = D(L)
       P = (D(L1) - G) / (2.0D0 * E(L))
       R = PYTHAG(P.1.0D0)
       D(L) = E(L) / (P + DSIGH(R,P))
       D(L1) = E(L) * (P + DSIGN(R,P))
       DL1 = D(L1)
       H = G - D(L)
       IF (L2 .GT. W) GO TO 145
       DO 140 I = L2, N
          D(I) = D(I) - H
140
       F = F + H
145
       P = D(M)
       C = 1.0D0
       C2 = C
       EL1 = E(L1)
       S = 0.0D0
       MML = M - L
       DO 200 II = 1, MML
         C3 = C2
         C2 = C
         S2 = S
         I = M - II
         G = C * E(I)
         H = C * P
         R = PYTHAG(P,E(I))
         E(I+1) = S * R
         S = E(I) / R
         C = P / R
         P = C * D(I) - S * G
         D(I+1) = H + S * (C * G + S * D(I))
         DO 180 K = 1, N
            H = Z(K,I+1)
            Z(K,I+1) = S * Z(K,I) + C * H
            Z(K,I) = C * Z(K,I) - S * H
```

```
180
        CONTINUE
 200
       CONTINUE
       P = -S * S2 * C3 * EL1 * E(L) / DL1
       E(L) = S * P
       D(L) = C * P
       TST2 = TST1 + DABS(E(L))
       IF (TST2 .GT. TST1) GO TO 130
 220
       D(L) = D(L) + F
 240 CONTINUE
    DO 300 II = 2, N
       I = II - 1
       K = I
       P = D(I)
       DO 260 J = II, N
         IF (D(J) .GE. P) GO TO 260
         K = J
         P = D(J)
 260
       CONTINUE
       IF (K .EQ. I) GO TO 300
       D(K) = D(I)
       D(I) = P
       DO 280 J = 1, N
          P = Z(J,I)
          Z(J,I) = Z(J,K)
          Z(J,K) = P
 280
       CONTINUE
 300 CONTINUE
     GO TO 1001
1000 IERR = L
1001 RETURN
     RND
SUBROUTINE TRED2(NM, N, A, D, E, Z)
************************
     THIS SUBROUTINE REDUCES A REAL SYMMETRIC MATRIX TO A
     SYMMETRIC TRIDIAGONAL MATRIX USING AND ACCUMULATING
     ORTHOGONAL SIMILARITY TRANSFORMATIONS.
          It was taken from the LIMPACK library
***********************
     INTEGER I, J, K, L, N, II, NM, JP1
     DOUBLE PRECISION A(NM, N), D(N), E(N), Z(NM, N)
     DOUBLE PRECISION F,G,H,HH,SCALE
    DO 100 I = 1, N
       DO 80 J = I, N
         Z(J,I) = A(J,I)
  80
       D(I) = A(N,I)
 100 CONTINUE
     IF (N .EQ. 1) GO TO 510
     DO 300 II = 2, N
       I = N + 2 - II
       L = I - 1
       H = 0.000
       SCALE = 0.0D0
       IF (L .LT. 2) GO TO 130
       DO 120 K = 1, L
```

```
120
           SCALE = SCALE + DABS(D(K))
        IF (SCALE .NE. 0.0D0) GO TO 140
130
       E(I) = D(L)
       DO 135 J = 1, L
           D(J) = Z(L,J)
           Z(I,J) = 0.0D0
           Z(J,I) = 0.000
135
       CONTINUE
       GO TO 290
       DO 150 K = 1, L
140
           D(K) = D(K) / SCALE
           H = H + D(K) * D(K)
150
       CONTINUE
       F = D(L)
       G = -DSIGN(DSQRT(H), F)
       E(I) = SCALE * G
       H = H - F * G
       D(L) = F - G
       DO 170 J = 1, L
170
          E(J) = 0.0D0
       DO 240 J = 1, L
          F = D(J)
          Z(J,I) = F
          G = E(J) + Z(J,J) * F
          JP1 = J + 1
          IF (L .LT. JP1) GO TO 220
          DO 200 K = JP1, L
             G = G + Z(K,J) * D(K)
             E(K) = E(K) + Z(K,J) * F
200
          CONTINUE
220
          E(J) = G
240
       CONTINUE
       F = 0.0D0
       DO 245 J = 1, L
          E(J) = E(J) / H
          F = F + E(J) * D(J)
245
       CONTINUE
       HH = F / (H + H)
       DO 250 J = 1, L
250
          E(J) = E(J) - HH + D(J)
       DO 280 J = 1, L
          F = D(J)
          G = E(J)
          DO 260 K = J, L
260
             Z(K,J) = Z(K,J) - F * E(K) - G * D(K)
          D(J) = Z(L,J)
          Z(I,J) = 0.0D0
280
       CONTINUE
290
       D(I) = H
300 CONTINUE
    DO 500 I = 2, N
       L = I - 1
       Z(N,L) = Z(L,L)
       Z(L,L) = 1.0D0
       H = D(I)
```

```
IF (H .EQ. 0.0D0) GO TO 380
       DO 330 K = 1, L
 330
         D(K) = Z(K,I) / H
       DO 360 J = 1, L
       G = 0.0D0
       DO 340 K = 1, L
 340
        G = G + Z(K,I) * Z(K,J)
       DO 360 K = 1, L
        Z(K,J) = Z(K,J) - G * D(K)
 360
       CONTINUE
 380
       DO 400 K = 1, L
 400
         Z(K,I) = 0.0D0
 500 CONTINUE
 510 DO 520 I = 1, N
       D(I) = Z(N,I)
       Z(N,I) = 0.0D0
 520 CONTINUE
    Z(N,N) = 1.0D0
    E(1) = 0.0D0
    RETURN
    END
DOUBLE PRECISION FUNCTION PYTHAG(A,B)
*************
   FINDS DSQRT(A**2+B**2) WITHOUT OVERFLOW OR DESTRUCTIVE UNDERFLOW *
**********************
    DOUBLE PRECISION A, B
    DOUBLE PRECISION P,R,S,T,U
    P = DMAX1(DABS(A), DABS(B))
    IF (P .EQ. 0.0D0) GO TO 20
    R = (DMIN1(DABS(A), DABS(B))/P)**2
 10 CONTINUE
    T = 4.0D0 + R
    IF (T .EQ. 4.0DO) GO TO 20
    S = R/T
    U = 1.0D0 + 2.0D0*S
    P = U*P
    R = (S/U)**2 * R
    GO TO 10
  20 PYTHAG = P
    RETURN
    END
SUBROUTINE PLOTDR(X,Y,NOBS,KEY,NSX,NSY,PRMODE)
*******************
           PLOTTING SCATTEGRAM - DRIVER
******************
    INTEGER PRMODE, NOBS, NSX, NSY
    DOUBLE PRECISION X(NOBS), Y(NOBS), XAXIS(28), T
    INTEGER KEY(NOBS), IA(133), IXLINE(133)
    DOUBLE PRECISION XSTEP, YSTEP, XMN, XMX, XMIN, YMN, YMX, YMIN
    INTEGER J, MAXA, MAXYA, MAXB, MAXYB, NSX1, NSY1
    INTEGER POM, LOGNB2, I, K, L, M, NN
    LOGNB2=INT(ALOG(FLOAT(NOBS)) *1.4426950+1E-5)
    DO 10 I=1, NOBS
```

```
KEY(I)=I
  10 CONTINUE
      M=NOBS
      DO 40 NN=1,LOGNB2
        M=M/2
        K=NOBS-M
        DO 30 J=1.K
          I=J
  20
          CONTINUE
          L=I+M
          IF(Y(L).GT.Y(I)) THEN
            T=Y(I)
            POM=KEY(I)
           Y(I)=Y(L)
           KEY(I)=KEY(L)
           Y(L)=T
           KEY(L)=POM
           I≈I-M
           IF(I.GE.1)GO TO 20
         ENDIF
   30 CONTINUE
   40 CONTINUE
     YMX = Y(1)
      YMN = Y(NOBS)
     XMX = X(1)
     XMN = XMX
     DO 50 J=2,NOBS
        XMN = DMIN1(X(J), XMN)
        XMX = DMAX1(X(J),XMX)
   50 CONTINUE
      IF (YMN.EQ.YMX) THEN
        YMN = YMN - 1.0D0
        YMX = YMX + 1.0D0
     ENDIF
     IF (XMN.EQ.XMX) THEN
        XMN = XMN - 1.0DO
        XMX = XMX + 1.0D0
     ENDIF
     NSX1 = NSX
     MSY1 = NSY
     CALL PLOTAX (XMN, XMX, NSX1, XMIN, XSTEP, MAXA, MAXB)
     IF (XMIN.EQ.XMN) THEN
        XMIN = XMIN - XSTEP
        MSX1 = MSX1 + 1
     ENDIF
     CALL PLOTAX (YMN, YMX, NSY1, YMIN, YSTEP, MAXYA, MAXYB)
     IF (YMIN.EQ.YMN) THEN
        YMIN = YMIN - YSTEP
        MSY1 = MSY1 + 1
     ENDIF
     CALL PLOTMN(X,Y,XMIN,XSTEP,NSX1,YMIN,YSTEP,NSY1,NOBS,MAXA,
    &MAXB, KEY, MAXYA, MAXYB, IA, IXLINE, XAXIS, PRMODE)
     RETURN
     END
```

```
SUBROUTINE PLOTAX (ZMN, ZMX, NSTEP, ZNMIN, STEP, MAXA, MAXB)
********************
         PLOTTING SCATTEGRAM - AUXILIARY SUBROUTINE
********************
     DOUBLE PRECISION STEP, ZMN, ZMX, ZNMIN
     INTEGER MAXA, MAXB, NSTEP
     DOUBLE PRECISION AR, RINT, RNSTPZ, TENN, ZNM, ZNMAX
     INTEGER I.II.IT.J.MINT
     DOUBLE PRECISION R(9)
     DATA R /0.1D0,0.15D0,0.2D0,0.25D0,0.4D0,0.5D0,0.6D0,0.75D0,0.8D0/
     RNSTPZ = NSTEP
     RINT = (ZMX-ZMN)/(RNSTPZ+0.1D0)
  10 MINT = DLOG10(RINT) - 2.0D0
     TENN = 10.0D0**MINT
     DO 30 J=1,11
        DO 20 I=1,9
           AR = R(I)
           IF (AR*TENN.GE.RINT) GO TO 40
  20
        CONTINUE
        TENN = TENN*10.0D0
  30 CONTINUE
  40 STEP = TENN*AR
     IT = ZMN/STEP
     AR = STEP*DBLE(IT)
  50 IF ((AR-STEP*0.05D0).GT.ZMN) THEN
         AR = AR - STEP
         GO TO 50
      ENDIF
      ZNMIN = AR
      ZNMAX = ZNMIN + STEP*(RNSTPZ+0.04D0)
      IF (ZNMAX.LT.ZMX) THEN
         RINT = RINT*1.05D0
         GO TO 10
      ENDIF
      MAXA = DLOG10(DABS(ZNMAX)) + 1.0D0
      IF (DABS(ZNMIN).GT.DABS(ZNMAX)) MAXA = DLOG10(DABS(ZNMIN)) +
     & 1.0DO
      MAXA = MAXO(O, MAXA)
      MAXB = DLOG10(STEP) - 2.5D0
      MAXB = -MINO(MAXB, 0)
      IF (MAXA+MAXB.GE.10) GO TO 70
      ZNM = ZNMIN
      DO 60 II=1,10
        I = II - 1
        IF (ZNM+STEP*(RNSTPZ+0.04D0).LT.ZMX) GO TO 70
        ZNMIN = ZNM
        ZNM = (ZNMIN*10.0D0**(MAXB~I))
        IF (ZNM.LT.O.ODO) ZNM = ZNM - 1.ODO
        ZNM = DINT(ZNM)/10.0D0**(MAXB-I)
  60 CONTINUE
  70 RETURN
      END
SUBROUTINE PLOTMN (X, Y, XMIN, XSTEP, NSTEPX, YMIN, YSTEP, NSTEPY,
     & NOBS, MAXA, MAXB, KEY, MAXYA, MAXYB, IA, IXLINE, XAXIS, PRMODE)
```

```
*********************
          PLOTTING SCATTEGRAM - MAIN SUBROUTINE
*************
     DOUBLE PRECISION XMIN, XSTEP, YMIN, YSTEP, Z,Q
     INTEGER MAXA, MAXB, MAXYA, MAXYB, NOBS, NSTEPX, NSTEPY, P. L1, K1, NN
     DOUBLE PRECISION X(NOBS), XAXIS(28), Y(NOBS)
     INTEGER IA(133), IXLINE(133), KEY(NOBS)
     DOUBLE PRECISION ZDIFF, YAXIS, YDIFF
     INTEGER I, IAY, IFOR, II, ISP1, ISP2, ISPACE, IXAX, IXZERO,
    & IYZERO, J, J1, K, MAXF, MAXYF, NCOUNT, NEND, NSTPX2, PRMODE
     LOGICAL LXE.LYE
     CHARACTER MA(27), MB(22), MO(17), ME(39), NB(10)
     INTEGER ICODE(39), IFMA(27), IFMB(22), IFMO(17), NUMB(10)
     DATA MO /'(','','','X',',','2','8','(','F','',',',',',',',',
    &' ','X',')',')'/
    &' ',' ','A','1',',','1','X',',','3','0','(','I','5',':',')',')'/
     &'X',',','3','0','(','I','5',':',')',')'/
     DATA ME /'.','+',' ','1','2','3','4','5',
    &'6','7','8','9','A','B','C','D','E','F','G','H','I','J','K','L',
    &'M','N','O','P','Q','R','S','T','U','V','W','X','Y','Z','*'/
    DATA NB /'0','1','2','3','4','5','6','7','8','9'/
    DO 1 I=1.17
      IFMO(I)=ICHAR(MO(I))
 1
    DO 2 I=1,27
 2
      IFMA(I)=ICHAR(MA(I))
    DO 3 I=1,22
 3
      IFMB(I)=ICHAR(MB(I))
    DO 4 I=1,39
 4
      ICODE(I)=ICHAR(ME(I))
    DO 5 I=1,10
 5
      NUMB(I)=ICHAR(NB(I))
    YDIFF = 0.0D0
    ZDIFF = 0.0D0
    IXZERO = 0
    IYZERO = 0
    IF (XMIN.NE.O.ODO) THEN
       IF (DABS((XMIN/XSTEP)-DINT(XMIN/XSTEP+DSIGN(0.5D0,XMIN))).
   & GE.O.OOOO5DO) ZDIFF = DMOD(XMIN, XSTEP)
       IF (ZDIFF.NE.O.ODO) THEN
          IF (ZDIFF.LT.O.ODO) ZDIFF = XSTEP + ZDIFF
          XMIN = XMIN - ZDIFF
          NSTEPX = NSTEPX + 1
       ENDIF
       IF (XMIN.NE.O.ODO)
   æ
          IXZERO = -IDINT(XMIN/XSTEP+DSIGN(0.5DO,XMIN))
    ENDIF
    IF (YMIN.NE.O.ODO) THEN
       IF (DABS((YMIN/YSTEP)-DINT(YMIN/YSTEP+DSIGN(0.5D0,YMIN))).
   & GE.O.OOOO5DO) YDIFF = DMOD(YMIN, YSTEP)
       IF (YDIFF.NE.O.ODO) THEN
          IF (YDIFF.LT.O.ODO) YDIFF = YSTEP + YDIFF
          YMIN = YMIN - YDIFF
          NSTEPY = NSTEPY + 1
```

```
ENDIF
    ENDIF
    IF (YMIN.NE.O.ODO)
        IYZERO = -IDINT(YMIN/YSTEP+DSIGN(0.5D0,YMIN))
    NSTPX2 = NSTEPX + 2
    DO 10 I=1,NSTPX2
       IXLINE(I) = ICODE(1)
       IF (MOD(IABS(IXZERO-I+1),5).EQ.0) THEN
          IXLINE(I) = ICODE(2)
          NEND = (I-1)/5 + 1
          XAXIS(NEND) = XMIN + XSTEP*DBLE(I-1)
10 CONTINUE
    IFOR = MOD(IXZERO,5)
    MAXYF = MAXYA + MAXYB + 2
    IF (MAXYF.LE.11) THEN
       IFMA(9) = NUMB(MAXYB+1)
       LYE = .FALSE.
    ELSE
       IFMA(5) = ICODE(17)
       IFMA(9) = NUMB(4)
       LYE = .TRUE.
    ENDIF
    J1 = NSTPX2/100
    J = J1 + 1
    IFMA(11) = NUMB(J)
    IFMB(6) = NUMB(J)
    J = NSTPX2/10 + 1 - J1*10
    IFMA(12) = NUMB(J)
    IFMB(7) = NUMB(J)
    J = MOD(NSTPX2, 10) + 1
    IFMA(13) = NUMB(J)
    IFMB(8) = NUMB(J)
    IF (MOD(IABS(IYZERO-(NSTEPY+1)),5).NE.0) THEN
       DO 20 I=1,22
           MB(I)=CHAR(IFMB(I))
       CONTINUE
20
       WRITE (2,MB) (IXLINE(I), I=1,NSTPX2)
    ELSE
       YAXIS = YMIN + YSTEP*DBLE(NSTEPY+1)
       DO 30 I=1,27
         MA(I)=CHAR(IFMA(I))
       CONTINUE
30
       WRITE (2, MA) YAXIS, (IXLINE(I), I=1, NSTPX2)
    ENDIF
    NCOUNT = 1
    DO 110 J=1,NSTEPY
       K = NSTEPY - J + 1
       DO 40 I=1, NSTEPX
          IA(I) = 3
      CONTINUE
40
      K1=NCOUNT
       IF (NCOUNT.GT.NOBS) GO TO 70
50
       IAY = (Y(NCOUNT)-YMIN)/YSTEP + 0.5D0
       IF (IAY.EQ.0) IAY = 1
```

```
IF (IAY.LT.K) GO TO 70
       II = KEY(NCOUNT)
       IXAX = (X(II)-XMIN)/XSTEP + 0.5D0
       IF (IXAX.EQ.0) IXAX = 1
       IF (IA(IXAX).GT.3) THEN
          IA(IXAX) = IA(IXAX) + 1
       ELSE
          IA(IXAX) = 4
          IF (NCOUNT.GT.K1) THEN
             L1=NCOUNT-1
             Z=Y(NCOUNT)
             Q=X(KEY(NCOUNT))
             P=KEY(NCOUNT)
               IF (L1.GE.K1) THEN
60
                 IF(Q.LT.X(KEY(L1))) THEN
                   Y(L1+1)=Y(L1)
                   KEY(L1+1)=KEY(L1)
                   L1=L1-1
                   GOTO 60
                 ENDIF
               ENDIF
               Y(L1+1)=Z
               KEY(L1+1)=P
          ENDIF
       ENDIF
       NCOUNT = NCOUNT + 1
       GO TO 50
       DO 80 I=1, NSTEPX
70
          IF (IA(I).EQ.3) THEN
             IF (K.EQ.IYZERO) THEN
                 IA(I) = 1
                 IF (MOD(IABS(IXZERO-I), 5).EQ.0) IA(I) = 2
             ELSE
                 IF (I.EQ.IXZERO) THEM
                    IA(I) = 1
                    IF (MOD(IABS(IYZERO-K), 5).EQ.O) IA(I) = 2
                 ENDIF
              ENDIF
           ENDIF
           II = MINO(IA(I),39)
           IA(I) = ICODE(II)
80
        CONTINUE
        IF (PRMODE.EQ.1) THEN
             NN=30
           ELSE
             NN=10
        ENDIF
        IF (MOD(IABS(IYZERO-K),5).NE.0) THEN
           DO 90 I=1,22
               MB(I)=CHAR(IFMB(I))
 90
           CONTINUE
           WRITE (2,MB) ICODE(1),(IA(I),I=1,NSTEPX),ICODE(1),
                 (KEY(I), I=K1, MINO(K1+NW, NCOUNT-1))
        ELSE
           YAXIS = YMIN + YSTEP*DBLE(K)
```

```
DO 100 I=1,27
             MA(I)=CHAR(IFMA(I))
100
          CONTINUE
          WRITE (2, MA) YAXIS, ICODE(2), (IA(I), I=1, WSTEPX), ICODE(2),
                (KEY(I), I=K1, MINO(K1+NN, NCOUNT-1))
       ENDIF
110 CONTINUE
    IF (MOD(IABS(IYZERO),5).NE.0) THEN
       DO 120 I=1,22
          MB(I)=CHAR(IFMB(I))
120
       CONTINUE
       WRITE (2,MB) (IXLINE(I), I=1, NSTPX2)
     RLSE
       YAXIS = YMIN
       DO 130 I=1,27
          MA(I)=CHAR(IFMA(I))
130
       CONTINUE
       WRITE (2,MA) YAXIS, (IXLINE(I), I=1, NSTPX2)
    ENDIF
    IFOR = MOD(IXZERO, 5)
    IF (IFOR.LT.O) IFOR = IFOR + 5
    ISP1 = 11 + IFOR
    IF (MAXA+MAXB+2.LE.9) ISP1 = ISP1 - MAXA
    I = ISP1/10 + 1
    IFMO(2) = NUMB(I)
    I = MOD(ISP1, 10) + 1
    IFMO(3) = NUMB(I)
    MAXF = MAXA + MAXB + 2
    IF (MAXF.LE.9) THEN
       ISPACE = 10 - MAXF
       LXE = .FALSE.
       IFMO(10) = NUMB(MAXF+1)
       IFMO(12) = NUMB(MAXB+1)
       IFMO(14) = NUMB(ISPACE+1)
    ELSE
       IFMO(9) = ICODE(17)
       IFMO(10) = NUMB(10)
       IFMO(12) = NUMB(3)
       IFMO(14) = NUMB(2)
       LXE = .TRUE.
    ENDIF
    DO 140 I=1,17
         MO(I)=CHAR(IFMO(I))
140 CONTINUE
    WRITE (2,MO) (XAXIS(II),II=1,NEND,2)
    ISP2 = ISP1 + 5
    I = ISP2/10 + 1
    IFMO(2) = NUMB(I)
    I = MOD(ISP2, 10) + 1
    IFMO(3) = NUMB(I)
   DO 150 I=1,17
        MO(I)=CHAR(IFMO(I))
150 CONTINUE
    WRITE (2,MO) (XAXIS(II),II=2,NEND,2)
    IF (LXE) IFMO(9) = ICODE(18)
```

```
IF (LYE) IFMA(5) = ICODE(18)
     RETURN
     END
SUBROUTINE CDTR (X.G.P.IER)
***************
     COMPUTES P(X) = PROBABILITY THAT RANDOM VARIABLE
     DESCRIBED ACCORDING TO THE CHI-SQUARE DISTRIBUTION WITH
     G DEGREES OF FREEDOM, IS LESS THAN OR EQUAL TO X
********************
      REAL
                      X,G,P
      INTEGER
                      I,IER
      REAL
                      PT2,X1,NORMAL,RINFM
      DOUBLE PRECISION A, Z, DGAM, EPS, W, W1, B, Z1, HALF, ONE, THRTEN, THRD
      DOUBLE PRECISION DGAMMA, FUNC, XINF, XMIN
      COMMON RINFM, XINF, XMIN
             EPS/1.OD-6/, HALF/5.D-1/, THRTEN/13.DO/, ONE/1.DO/
      DATA
      DATA
             THRD/.333333333333333300/,PT2/.222222E0/
      FUNC(W,A,Z)=W*DEXP(A*DLOG(Z)-Z)
      IF (G.LT..5. OR. G.GT.2.E5 .OR. X.LT.O.O) THEN
        WRITE (2,*) 'DATA ERROR IN THE SUBROUTINE CDTR '
        IER=1
        RETURN
      ENDIF
      IER=0
      IF (X .LE. 1.E-12) THEN
        P=0.0
        RETURN
      ENDIF
      IF(G.GT.66.) THEN
        IF(X.LT.2.0) THEN
      P=0.0
      RETURN
        ENDIF
        X1=((X/G)**THRD-(ONE-PT2/G))/SQRT(PT2/G)
        IF (X1 .GT. 5.0) THEN
      P=1.0
      RETURN
        ENDIF
        IF (X1 .LT. -18.8055) THEN
      IER=1
      P=0.0
      RETURN
        ENDIF
        P=NORMAL(X)
        RETURN
      IF (X .GT. 200.) THEN
        P=1.0
        RETURN
      ELSE
        A=HALF*G
        Z=HALF*X
        DGAM = DGAMMA(A)
```

```
W=DMAX1(HALF*A, THRTEN)
       IF (Z.LT.W) THEN
     IF (G.GT.25. .AND. X.LT.2.) THEN
          P=0.0
          RETURN
       ENDIF
       W=ONE/(DGAM*A)
       W1=W
          DO 10 I=1,50
        B=I
        W1=W1*Z/(A+B)
        IF (W1 .LE. EPS+W) GO TO 20
        W=W+W1
  10
             CONTINUE
  20
           P=FUNC(W,A,Z)
       RETURN
        ENDIF
         Z1=ONE/Z
         B=A-ONE
        W1=B*Z1
         W=ONE+W1
       DO 30 I=2,50
          B=B-ONE
          W1=W1*B*Z1
          IF (W1 .LE. EPS*W) GO TO 40
          W=W+W1
  30
          CONTINUE
  40
        W=Z1*FUNC(W,A,Z)
        P=ONE-W/DGAM
      ENDIF
      END
DOUBLE PRECISION FUNCTION DGAMMA (X)
**************
     EVALUATES GAMMA FUNCTION
**************
      DOUBLE PRECISION X
      REAL
                       RINFM
      DIMENSION
                      P(9),Q(8),Z(7)
      DOUBLE PRECISION A,B,DEN,P,Z,Q,PI,T,TOP,BIG1,XINF,XMIN,Y,SIGN,R
      INTEGER
                      I,J
                      MFLAG
      LOGICAL
      COMMON
                      RINFM, XINF, XMIN
                       P/-5.966047488753637D01, 5.864023793062003D01,
      DATA
                        -1.364106217165365D03,-8.117569271425580D02,
                        -1.569414683149179D04,-1.525979925758372D04,
    Ł
                        -7.264059615964330D04,-8.972275718101010D-01,
                         3.349618189847578D00/
                       Q/ 4.103991474182904D02,-2.262590291514875D03,
      DATA
                         2.494325576714903D03, 2.362106244383048D04,
                         -5.741873227396418D04,-7.257239715408240D04,
                         -9.491399521686949D00,-3.255006939455704D01/
                       Z/8.40596949829D-04, -5.9523334141881D-04,
      DATA
                         7.9365078409227D-04, -2.777777777769526D-03,
                          8.33333333333333D-02,9.189385332046727D-01,
```

20

```
-1.7816839846D-03/
   DATA
             PI/3.141592653589793D0/,BIG1/34.844D0/
   MFLAG = .FALSE.
   T = X
   IF (DABS(T).GT.XMIN) THEN
      IF (DABS(T).GE.BIG1) THEN
    DGAMMA = XINF
    RETURN
      ELSE
    DGAMMA = XINF
    IF (T.LE.O.ODO) THEN
       DGAMMA = -XINF
       RETURN
    ENDIF
      ENDIF
   ENDIF
   IF (T.LE.O.ODO) THEN
      MFLAG = .TRUE.
      T = -T
      R = DINT(T)
      SIGN = 1.0D0
      IF (DMOD(T,2.0D0).EQ.0.0D0) SIGN = -1.0D0
      IF (R.NE.O.ODO) THEN
    R = PI/DSIN(R*PI)*SIGN
    T = T+1.0D0
       ELSE
    DGAMMA = XINF
     IF (SIGN.EQ.-1.ODO) THEN
       DGAMMA = -XINF
       RETURN
    ENDIF
       ENDIF
    ENDIF
    IF (T.GT.12.0D0) THEN
       TOP = DLOG(T)
       TOP = T*(TOP-1.0D0)-.5D0*TOP
       T = 1.0D0/T
       B = T*T
       A = Z(7)
       DO 10 J = 1,5
10
          A = A*B+Z(J)
       Y = A*T+Z(6)+TOP
       Y = DEXP(Y)
       IF (MFLAG) Y = R/Y
      DGAMMA = Y
   ELSE
       I = T
       A = 1.0D0
      IF (I.GT.2) THEN
        DO 20 J=3,I
      T = T-1.0D0
      A = A * T
         CONTINUE
      ELSE
```

```
I = I+1
      IF (I .EQ.1) THEN
         A = A/(T*(T+1.0D0))
         T = T+2.0D0
      ENDIF
      IF (I .EQ. 2) THEN
         A = A/T
         T = T+1.0D0
      ENDIF
        ENDIF
        TOP = P(8)*T+P(9)
        DEN = T+Q(8)
        DO 30 J=1,7
      TOP = TOP*T+P(J)
      DEN = DEN*T+Q(J)
  30
         CONTINUE
        Y = (TOP/DEN)*A
        IF (MFLAG) Y = R/Y
        DGAMMA = Y
      ENDIF
     END
REAL FUNCTION NORMAL(X)
  *******************
     EVALUATES THE LOWER TAIL AREA OF THE STANDARIZED NORMAL CURVE *
*****************
     REAL X
     REAL Y, W
     DOUBLE PRECISION Z
     IF (X.EQ.O.O) THEN
       Z=0D0
     ELSE
        Y=ABS(X)*0.5
        IF (Y.GT.3.0) THEN
          Z=1.0
        ELSE
          IF (Y.LT.1) THEN
             W=Y*Y
             Z=(((((((((0.000124818987D0*W
              -0.001075204047D0) *W + 0.005198775019D0) *W
    Ł
              -0.019198292004D0)*W + 0.059054035642D0)*W
    k
              -0.151968751364D0) +W + 0.319152932694D0) +W
    æ
              -0.531923007300D0)*\ + 0.797884560593D0)*\ Y*2D0
    &
          ELSE
             Y=Y-2.0
             Z=((((((((((( -0.00045255659D0*Y
              +0.000152529290D0)*Y - 0.000019538132D0)*Y
    Æ
              -0.000676904986D0)*Y + 0.001390604284D0)*Y
    æ
              -0.000794620820D0)*Y - 0.002034254874D0)*Y
    k
              +0.006549791214D0) +Y - 0.010557625006D0) +Y
    æ
              +0.011630447319D0)*Y - 0.009279453341D0)*Y
              +0.005353579108D0) *Y - 0.002141268741D0) *Y
    æ
              +0.000535310849D0)*Y + 0.999936657524D0
          ENDIF
        ENDIF
     ENDIF
```

IF(X.GT.0) THEN
NORMAL=REAL((Z+1D0)*0.5D0)
ELSE
NORMAL=REAL((1D0-Z)*0.5D0)
ENDIF
END