

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 \mathbf{X} in which column

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

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

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 Σ , whereas the vector of mean values $\bar{\mathbf{x}}$ 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:

$$\mathbf{V} = \mathbf{A}'\mathbf{X} \quad (1)$$

where

\mathbf{V} is a matrix of the new variables,

\mathbf{A} is a matrix of orthonormal eigenvectors of matrix \mathbf{S} , and

\mathbf{X} is the observation matrix.

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

$$|\mathbf{S} - l\mathbf{I}| = 0 \quad (2)$$

where

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

\mathbf{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

$$l_1 \geq l_2 \geq l_3 \geq \dots \geq l_p \geq 0.$$

Because there is an orthonormal column eigenvector \mathbf{A}_i corresponding to each root l_i , the variable V_i derived from Equation (2) has the maximum value l_i (maximum variance) and is termed the first principal component.

Because the sum $l_1 + l_2 + \dots + l_p = \text{tr } \mathbf{S}$ and is equal to the sum of the variances of matrix \mathbf{S} (i.e. $\sigma_{11} + \sigma_{22} + \dots + \sigma_{pp}$), l_1, l_2, \dots, 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 \mathbf{S} . The algorithm for the calculation of principal components is such that this is a decreasing sequence, which indicates that $l_1/\text{tr } \mathbf{S} 100$ 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 \mathbf{A}_i such that

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

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

Because vectors $\mathbf{A}_1, \mathbf{A}_2, \dots, \mathbf{A}_p$ are orthonormal, that is

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

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

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

$$\mathbf{I} = \mathbf{A}_1\mathbf{A}_1' + \dots + \mathbf{A}_p\mathbf{A}_p' \quad (7)$$

and

$$\mathbf{S} = l_1\mathbf{A}_1\mathbf{A}_1' + l_2\mathbf{A}_2\mathbf{A}_2' + \dots + l_p\mathbf{A}_p\mathbf{A}_p'. \quad (8)$$

Expression (8) is termed a spectral decomposition of a matrix \mathbf{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

$$\text{Var}(\mathbf{A}_i'\mathbf{X}) = l_i, \quad (9)$$

whereas

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

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 $l_1/\text{tr } \mathbf{S} 100$ 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 $l_i/\text{tr } \mathbf{S} 100$ 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 \mathbf{S} or \mathbf{R} is used.

Interpretation of principal components derived from variance-covariance matrix \mathbf{S}

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

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

Hence the coefficient of determination has the form:

$$r_{ij}^2 = \frac{a_{ij}^2 l_i}{S_j^2}, \quad (11)$$

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_j^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 $l_i A_i A_i'$ has the greatest share in this reconstruction. Moreover, in for example matrix $l_i A_i A_i'$ the elements on the main diagonal are estimates (supplied with the help of the first component) of the variance of the j th initial variable, which can be computed from a general expression:

$$\hat{\text{Var}} X_j = l_i a_{ij}^2, \quad (12)$$

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

$$r_{ij}^2 = \frac{\hat{\text{Var}} X_j}{\text{Var} X_j} = \frac{\hat{S}_j^2}{S_j^2}. \quad (13)$$

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 (i th) matrix from the spectral decomposition, then, by summing up the elements on the main diagonal

$$\sum_{j=1}^p \hat{S}_j^2 = \sum_{j=1}^p l_i a_{ij}^2$$

$$\sum_{j=1}^p \hat{S}_j^2 = l_i \sum_{j=1}^p a_{ij}^2$$

we get

$$\sum_{j=1}^p \hat{S}_j^2 = l_i. \quad (14)$$

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

$$\frac{\sum_{j=1}^p \hat{S}_j^2}{\text{tr } S} 100. \quad (15)$$

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:

$$w = \frac{\sum_{i,j} r_{ij}^2}{p} 100 \quad (16)$$

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 i th 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 i th 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:

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

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

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

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):

$$r_{ij} = a_{ij} \sqrt{l_i}, \quad (18)$$

$$r_{ij}^2 = a_{ij}^2 l_i, \quad (19)$$

$$\sum_{i,j} r_{ij}^2 = \sum_{i,j} a_{ij}^2 l_i,$$

$$\sum_{i,j} r_{ij}^2 = l_i. \quad (20)$$

Or,

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

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 \mathbf{S} and \mathbf{R} are only estimators of matrices Σ and \mathbf{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 \mathbf{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 \mathbf{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:

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

against an alternative hypothesis:

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

where δ is assumed known.

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

$$2 \left(\frac{\delta}{\text{tr } \Sigma} \right)^2 (\lambda_1^2 + \dots + \lambda_k^2) + 2 \left(\frac{1 - \delta}{\text{tr } \Sigma} \right)^2 (\lambda_{k+1}^2 + \dots + \lambda_p^2). \quad (22)$$

Hypothesis H_0 is rejected if $\sqrt{n} [f(\mathbf{I}) - \delta]$ 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:

$$2 \left(\frac{\delta}{\text{tr } \mathbf{S}} \right)^2 (l_1^2 + \dots + l_k^2) + 2 \left(\frac{1 - \delta}{\text{tr } \mathbf{S}} \right)^2 (l_{k+1}^2 + \dots + l_p^2) \quad (23)$$

whereas

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

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:

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

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

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

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

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

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

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

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:

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

Hypothesis H_0 usually is tested sequentially, taking $k = 0, k = 1$, etc. If $\chi_{\text{obl}}^2 \geq \chi_{\alpha, df}^2$, then H_0 is rejected.

Testing principal components derived from correlation matrix \mathbf{R}

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

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

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

where

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

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

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:

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

If $\chi_{\text{obl}}^2 \geq \chi_{\alpha, df}^2$, 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 $1 \leq P \leq 50$.


```

***** VAR.-COVARIANCE MATRIX *****
      COL( 1)      COL( 2)      COL( 3)      COL( 4)      COL( 5)
1  0.486468D-02  -0.158013D-01  -0.229941D-02  -0.128019D-01  0.160151D-01
2  -0.158013D-01  0.130006D+01  0.777635D-01  0.779941D-01  0.124907D+01
3  -0.229941D-02  0.777635D-01  0.868086D-02  0.365850D-02  0.652537D-01
4  -0.128019D-01  0.779941D-01  0.365850D-02  0.927149D-01  -0.205957D-02
5  0.160151D-01  0.124907D+01  0.652537D-01  -0.205957D-02  0.141626D+01
6  0.151046D-02  -0.728109D-02  -0.670036D-03  -0.289727D-02  0.335006D-02
      COL( 6)
1  0.151046D-02
2  -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)
--- NO. CRITICAL DELTA LEFTCONFID. RIGHTCONFID.
1  100 89.20 96.00
2  4 97.60 99.10
3  1 99.80 99.90
4  1 100.00 100.00
5  0 0.00 0.00
6  0 0.00 0.00
BARTLET'S TEST REJECTS EIGENVECTORS NO.
5
6
***** NON-ZERO EIGENVALUES *****
NR. LAMBDA PERCENTAGE OF TRACE
1  0.2613593905742508E+01 92.57
2  0.1631374931478640E+00 5.78
3  0.4304858866144624E-01 1.52
4  0.3247796870052141E-02 0.12
5  0.3270209024478446E-03 0.01
6  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
4  -0.0208237984  -0.6376345788
5  -0.7221949295  0.5291913710
6  0.0010306523  0.0478813818
***** VALUES OF THE PRINCIPAL COMPONENTS *****
      V( 1)      V( 2)
1  -0.1095584883804461D+02  1  -0.3684137088632902D+00
2  -0.1277621780691318D+02  2  -0.5711549374131692D+00
3  -0.1088636639158975D+02  3  -0.1311308911594027D+00
4  -0.9154394532615441D+01  4  -0.8119405490473295D+00
5  -0.1024152023331932D+02  5  -0.6714900809364176D+00
6  -0.117353222750920D+02  6  0.2681910528735984D+00
7  -0.9597303384400675D+01  7  -0.1193822078561360D+01
8  -0.1309397415020718D+02  8  -0.7065554935458379D+00
9  -0.1478552145730417D+02  9  -0.8358495501355532D+00
10 -0.9928464507290487D+01  10 0.8389599924975855D-01
11 -0.9445589847792558D+01  11 -0.1117015714194618D+00
12 -0.9693031858935081D+01  12 -0.1607771123805059D+00
13 -0.1047145550469394D+02  13 -0.4560731083644219D+00
14 -0.1062207319044447D+02  14 0.9132649503541419D-01
15 -0.1033337592670940D+02  15 -0.3558322408807868D+00
16 -0.1050531742480956D+02  16 -0.5230999872863084D+00
17 -0.1257441980887520D+02  17 -0.5034207331926865D+00
18 -0.8857152675013044D+01  18 -0.4782461294909233D+00
19 -0.1146614062732514D+02  19 -0.9664908816771722D+00
20 -0.8292967215995477D+01  20 0.1062143760071039D+00
21 -0.9369962434875720D+01  21 -0.1082235996258010D+01
22 -0.9587472449555461D+01  22 -0.8175395451033177D-01
23 -0.7796465582402760D+01  23 -0.1046868945368204D+01
24 -0.1020611846632740D+02  24 -0.1327259443636310D+01
25 -0.9952227459998097D+01  25 -0.2837886044528224D+00
26 -0.1198613878838233D+02  26 -0.4960124751256308D+00

```

27	-0.1337860765538985D+02	27	-0.2071740963255488D+00
28	-0.1313025519510916D+02	28	-0.5008657100114776D+00
29	-0.1006250738816002D+02	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
X(3)	R ² = 0.45155994	R= -0.67198210	SIGNIFICANT FOR ALPHA	< 0.001
X(4)	R ² = 0.01222387	R= -0.11056160	SIGNIFICANT FOR ALPHA	> 0.1
X(5)	R ² = 0.96250418	R= -0.98107298	SIGNIFICANT FOR ALPHA	< 0.001
X(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)

X(1)	R ² = 0.85289210	R= 0.92352158	SIGNIFICANT FOR ALPHA	< 0.001
X(2)	R ² = 0.03537160	R= -0.18807339	SIGNIFICANT FOR ALPHA	> 0.1
X(3)	R ² = 0.07088949	R= -0.26625080	SIGNIFICANT FOR ALPHA	> 0.1
X(4)	R ² = 0.71539870	R= -0.84581245	SIGNIFICANT FOR ALPHA	< 0.001
X(5)	R ² = 0.03225781	R= 0.17960461	SIGNIFICANT FOR ALPHA	> 0.1
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

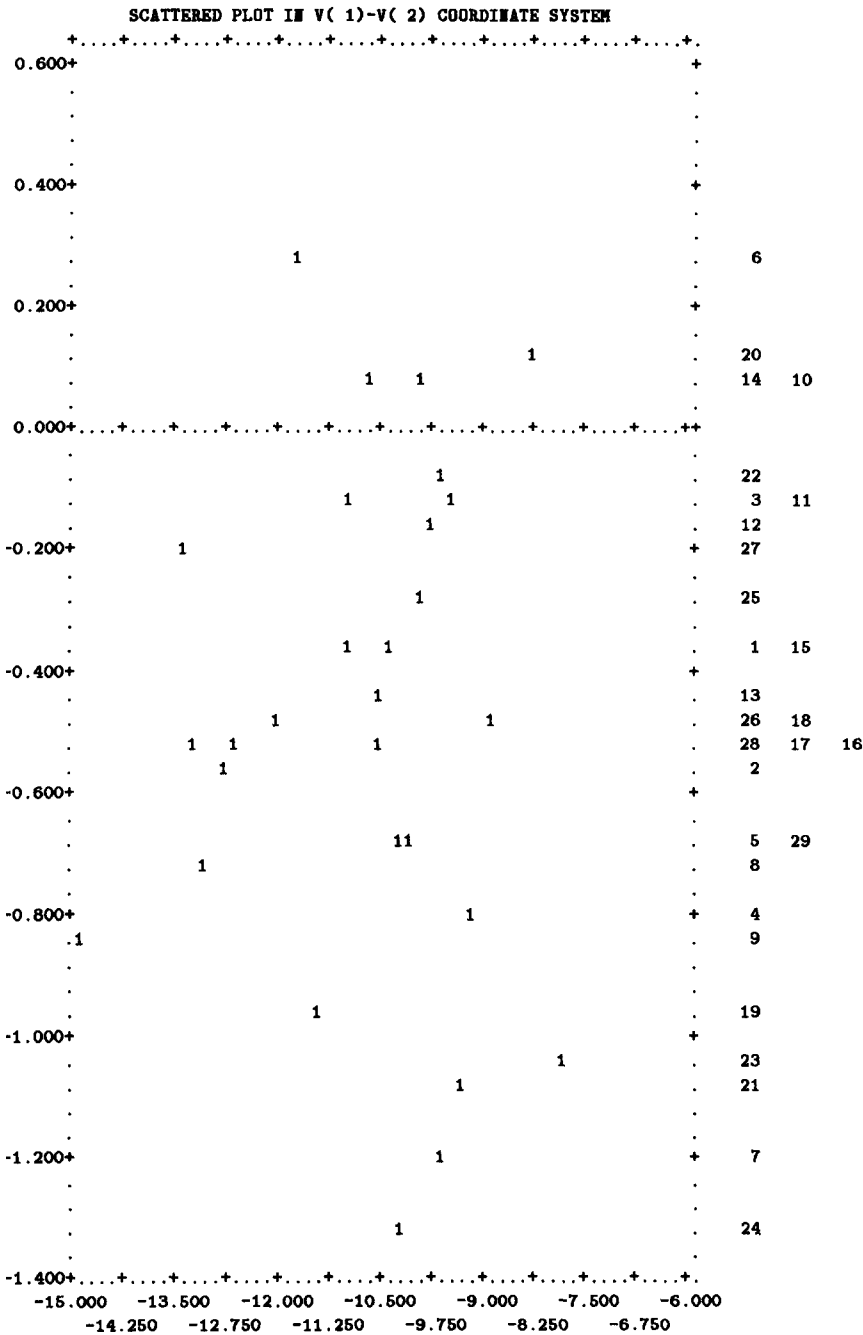
..D(1, 3) =	0.307083
...D(3, 6) =	0.955405
....D(6, 26) =	0.815476
.....D(26, 17) =	0.604401
.....D(17, 2) =	0.502295
.....D(17, 28) =	0.564624
.....D(28, 8) =	0.257682
.....D(28, 27) =	0.478748
.....D(27, 9) =	1.60577
..D(3, 14) =	0.397367
..D(1, 13) =	0.516817
..D(13, 5) =	0.386523
...D(5, 24) =	0.672532
...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) =	0.695918
.....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
MEAN =	0.55868870
STD. DEV=	0.32723257
MEAN + 2*(STD.DEV) =	1.21315384

CONSTRUCTING A DENDRITE FOR THE EUCLIDEAN DISTANCE MATRIX

DISTANCE MATRIX IN A COORDINATE SYSTEM WITH AXES V(1)-V(2) IS CONSIDERED

..D(1, 3) =	0.247247
..D(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
MEAN = 0.46439481
STD. DEV= 0.33063257
MEAN + 2*(STD.DEV) = 1.12565994




```

*   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 0 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),NSTEPPY
*   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          ST
*   CHARACTER        INSET*60,OUTSET*60,STR*9,YON*1
*   EQUIVALENCE      (Y,WDIS),(DD,X),(E,U),(S,V)
*   COMMON           RINFM,XINF,XMIN
*   RINFM= -RXXXM
*   XINF= XXXF
*   XMIN= XXXN
*   WRITE(*, '(12X, ''PRINCIPAL COMPONENTS ANALYSIS-INTERACTION PROGRAM
& ''//)')
*   WRITE(*, '(12X, ''*****
& '',///)')
*   WRITE(*, '( '' ***** GIVE THE PATH NAME OF THE INPUT FILE - AT
& MOST 60 CHARACTERS *****')')
*   READ(*,5) INSET
*   WRITE(*, '( '' ***** GIVE THE NAME OF THE OUTPUT FILE - AT MOST
& 60 CHARACTERS *****')')
*   READ(*,5) OUTSET
5  FORMAT (A)
OPEN(2,STATUS='UNKNOWN',FILE=OUTSET)
WRITE(2, '(22X, ''*****

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WRITE(2,'(22X, ''* PRINCIPAL COMPONENTS ANALYSIS *')')
WRITE(2,'(22X, ''*****')')
WRITE(2,'(/, ''* ', ' FOR DATA FILE ',A60,'* ')') INSET
ST=.FALSE.
10 IF (ST) THEN
    OPEN(1,STATUS='OLD',FILE=INSET)
ELSE
    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 = ',I3,33X, ''NUMBER OF CASES=',
&I3)') N,M
IF (M.LE.300) THEN
    NSTEPPY=50
ELSE IF (M.LE.600) THEN
    NSTEPPY=100
ELSE
    NSTEPPY=200
ENDIF
WRITE(2,*)
WRITE(2,*) ('<>',I=1,39)
IF(ST) THEN
    WRITE(2,'(/19X, ''CORRELATION MATRIX WAS TAKEN INTO ACCOUNT''/')')
ELSE
    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,

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```

&'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 #####')')
        ELSE
        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=MIN0(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(',I2,')',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.0) 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,'(/10X,'' ***** NON-ZERO 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(100DO*D(L)/TRACE)
        ENDIF
      150 CONTINUE
      160 FORMAT(5X,I3,10X,E22.16,7X,F6.2)
*-----
*   PRINTING EIGENVALUES ON THE SCREEN
*-----
      170 IF(ST) THEN
        WRITE(*,'(/10X,''+++ CORRELATION MATRIX IS CONSIDERED +++'')')
      ELSE
        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
        ELSE
          WRITE(*,160) I,D(L),SNGL(100DO*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.0) 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 ' ', I1,
&          ' ' TO ' ', I3, ' ' ) OF THE VAR.-COVARIANCE MATRIX' )' ) 1, K
        ENDIF
*-----
*      PRINTING THE EIGENVECTORS
*-----
      DO 230 L=0, K-1, 5
        WRITE(2, *)
        J1=MIN0(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=OD0
          DO 260 LA=1, N
            BX=BX+X(LA)*A(LA, N-I+1)
260      CONTINUE
          WRITE(2, '(8X, I3, 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.0) 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 '
    ENDIF
    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'))
    WRITE(2,(' ACCOUNTED FOR BY V('',I3,'') MEASURE W('',I3,'')
&)='',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,M
        X(K)=Y(K,I)
340    CONTINUE
      DO 360 J=I+1,M
        SX1=0.0
        DO 350 K=1,M
          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 DISTANCE 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 COORDINATE 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.EQ.'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)=SMGL(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=RXMX*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
&NATE 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
  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..0 .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)*N
  X=D*P
  Y=X**(2.0/N)
  IF(Y.GT..05+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)

```



```

        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-1E0) THEN
                SEL=SEL+C(I)
            ELSE
                SEL=INF-1E0
            ST=1
        ENDIF
40    CONTINUE
        IF (PR.EQ.0.OR.S.NE.1) RETURN
        DO 50 I=1,N
            H(I)=0
50    CONTINUE
        DO 60 I=2,N
            J=B(I)
            H(J)=H(J)+1
60    CONTINUE
        R(1)=1
        J=1
        K=1
        DO 100 I=2,N
70    IF (H(K).EQ.0) THEN
            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) THEN
                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 '
                ENDIF
                WRITE(2,'(133A)') SHORT
            ENDIF
        MEAN=MEAN+C(M1)

```


CAGEO 19/3—C

```

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

```


[illegible]

```

DOUBLE PRECISION C,C2,C3,DL1,EL1,F,G,H,P,R,S,S2,TST1,TST2,PYTHAG
IERR = 0
IF (N .EQ. 1) GO TO 1001
DO 100 I = 2, N
100 E(I-1) = E(I)
F = 0.0D0
TST1 = 0.0D0
E(N) = 0.0D0
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
120 IF (M .EQ. L) GO TO 220
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 + DSIGN(R,P))
D(L1) = E(L) * (P + DSIGN(R,P))
DL1 = D(L1)
H = G - D(L)
IF (L2 .GT. N) GO TO 145
DO 140 I = L2, N
140 D(I) = D(I) - H
145 F = F + H
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
        END
* <<<<<<<<<<<<<<<<<<<<<<<<<<<<*>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>*
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 LINPACK library                   *
*****
INTEGER I,J,K,L,M,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)
    D(I) = A(M,I)
100  CONTINUE
IF (N .EQ. 1) GO TO 510
DO 300 II = 2, N
  I = N + 2 - II
  L = I - 1
  H = 0.0D0
  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) = 0.0D0
135     CONTINUE
      GO TO 290
140     DO 150 K = 1, L
          D(K) = D(K) / SCALE
          H = H + D(K) * D(K)
150     CONTINUE
      F = D(L)
      G = -DSIGN(DSQRTH,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. 4.0D0) 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,XXM,XMIN,YMN,YMX,YMIN
      INTEGER J,MAYA,MAYYA,MAXB,MXYB,NSX1,NSY1
      INTEGER POM,LOGNB2,I,K,L,M,NN
      LOGNB2=INT(ALOG(FLOAT(NOBS)))*1.4426950+1E-5)
      DO 10 I=1,NOBS

```


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,NH
      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',' ',' ')/
      DATA MA /'(','1','X',' ',' ','F','1','1',' ',' ',' ',' ',' ',' ',' '
& ' ',' ','A','1',' ',' ','1','X',' ',' ','3','0','(','I','5',' ',' ',' ')/
      DATA MB /'(','1','2','X',' ',' ',' ',' ',' ','A','1',' ',' ','1','
& '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
1      IFMO(I)=ICHAR(MO(I))
      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.0.0D0) THEN
          IF (DABS((XMIN/XSTEP)-DINT(XMIN/XSTEP+DSIGN(0.5D0,XMIN))))
& GE.0.00005D0) ZDIFF = DMOD(XMIN,XSTEP)
          IF (ZDIFF.NE.0.0D0) THEN
              IF (ZDIFF.LT.0.0D0) ZDIFF = XSTEP + ZDIFF
              XMIN = XMIN - ZDIFF
              NSTPX = NSTPX + 1
          ENDIF
          IF (XMIN.NE.0.0D0)
& IXZERO = -IDINT(XMIN/XSTEP+DSIGN(0.5D0,XMIN))
      ENDIF
      IF (YMIN.NE.0.0D0) THEN
          IF (DABS((YMIN/YSTEP)-DINT(YMIN/YSTEP+DSIGN(0.5D0,YMIN))))
& GE.0.00005D0) YDIFF = DMOD(YMIN,YSTEP)
          IF (YDIFF.NE.0.0D0) THEN
              IF (YDIFF.LT.0.0D0) YDIFF = YSTEP + YDIFF
              YMIN = YMIN - YDIFF
              NSTEPY = NSTEPY + 1
          ENDIF
      ENDIF

```



```

      ENDIF
ENDIF
IF (YMIN.NE.0.0D0)
&   IYZERO = -IDINT(YMIN/YSTEP+DSIGN(0.5D0,YMIN))
NSTPX2 = NSTPX + 2
DO 10 I=1,NSTPX2
  IXLIN(I) = ICODE(1)
  IF (MOD(ABS(IYZERO-I+1),5).EQ.0) THEN
    IXLIN(I) = ICODE(2)
    NEND = (I-1)/5 + 1
    XAXIS(NEND) = XMIN + XSTEP*DBLE(I-1)
  ENDIF
10 CONTINUE
  IFOR = MOD(IYZERO,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(ABS(IYZERO-(NSTEPI+1)),5).NE.0) THEN
    DO 20 I=1,22
      MB(I)=CHAR(IFMB(I))
20    CONTINUE
    WRITE (2,MB) (IXLIN(I),I=1,NSTPX2)
  ELSE
    YAXIS = YMIN + YSTEP*DBLE(NSTEPI+1)
    DO 30 I=1,27
      MA(I)=CHAR(IFMA(I))
30    CONTINUE
    WRITE (2,MA) YAXIS,(IXLIN(I),I=1,NSTPX2)
  ENDIF
  NCOUNT = 1
  DO 110 J=1,NSTEPI
    K = NSTEPI - J + 1
    DO 40 I=1,NSTPX
      IA(I) = 3
40    CONTINUE
    K1=NCOUNT
50    IF (NCOUNT.GT.NOBS) GO TO 70
    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)
60      IF (L1.GE.K1) THEN
            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
70  DO 80 I=1,NSTEPX
    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) THEN
                IA(I) = 1
                IF (MOD(IABS(IYZERO-K),5).EQ.0) IA(I) = 2
            ENDIF
        ENDIF
    ENDIF
    II = MIN0(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,MIN0(K1+NN,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,NSTEPX),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)
        ELSE
          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.0) 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)

```


[illegible]

```

*                               -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.0.0D0) THEN
            DGAMMA = -XINF
            RETURN
        ENDIF
    ENDIF
ENDIF
IF (T.LE.0.0D0) THEN
    MFLAG = .TRUE.
    T = -T
    R = DINT(T)
    SIGN = 1.0D0
    IF (DMOD(T,2.0D0).EQ.0.0D0) SIGN = -1.0D0
    R = T-R
    IF (R.NE.0.0D0) THEN
        R = PI/DSIN(R*PI)*SIGN
    T = T+1.0D0
    ELSE
        DGAMMA = XINF
        IF (SIGN.EQ.-1.0D0) 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
20      CONTINUE
    ELSE

```

CAGEO 19/3—D

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