

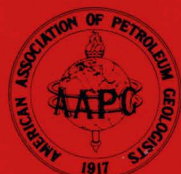
DANIEL F. MERRIAM, Editor

**MINIMUM ENTROPY
CRITERION FOR
ANALYTIC ROTATION**

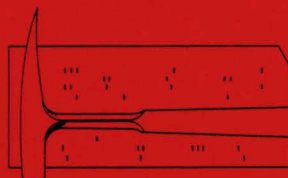
By

RICHARD B. McCAMMON

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in cooperation with the
American Association of Petroleum Geologists
Tulsa, Oklahoma



COMPUTER CONTRIBUTION 43

State Geological Survey

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Editor's Remarks

COMPUTER CONTRIBUTION 43, "Minimum entropy criterion for analytic rotation", by R.B. McCammon is a program for structuring data for both principal component and factor analysis. The fact that these two related techniques are used widely now makes this contribution an important one. The program should be of interest to many geologists and other earth scientists involved with classification and correlation.

The program can be purchased for a limited time from the Geological Survey for \$15.00 (US). The program is supplied on magnetic tape along with operating instructions. If punched cards are desired an extra \$10.00 is necessary to cover cost of handling and postage.

A complete list of COMPUTER CONTRIBUTIONS can be obtained by writing, Editor, COMPUTER CONTRIBUTIONS, Kansas Geological Survey, The University of Kansas, Lawrence, Kansas 66044.

Computer Contribution

1. Mathematical simulation of marine sedimentation with IBM 7090/7094 computers, by J.W. Harbaugh, 1966 (out of print)
2. A generalized two-dimensional regression procedure, by J.R. Dempsey, 1966 \$0.50
3. FORTRAN IV and MAP program for computation and plotting of trend surfaces for degrees 1 through 6, by Mont O'Leary, R.H. Lippert, and O.T. Spitz, 1966 \$0.75
4. FORTRAN II program for multivariate discriminant analysis using an IBM 1620 computer, by J.C. Davis and R.J. Sampson, 1966 \$0.50
5. FORTRAN IV program using double Fourier series for surface fitting of irregularly spaced data, by W.R. James, 1966 \$0.75
6. FORTRAN IV program for estimation of cladistic relationships using the IBM 7040, by R.L. Bartcher, 1966 \$1.00
7. Computer applications in the earth sciences: Colloquium on classification procedures, edited by D.F. Merriam, 1966 \$1.00
8. Prediction of the performance of a solution gas drive reservoir by Muskat's equation, by Apolonio Baca, 1967 \$1.00
9. FORTRAN IV program for mathematical simulation of marine sedimentation with IBM 7040 or 7094 computers, by J.W. Harbaugh and W.J. Wahlstedt, 1967 \$1.00
10. Three-dimensional response surface program in FORTRAN II for the IBM 1620 computer, by R.J. Sampson and J.C. Davis, 1967 \$0.75

(continued on inside back cover)

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MINIMUM ENTROPY CRITERION FOR ANALYTIC ROTATION

by

Richard B. McCammon

ABSTRACT

Minimum entropy is described as an analytic criterion for rotation to simple structure for both principal component and factor analysis data matrices. Minimum entropy rotated matrices come closer to achieving the ideal simple structure than is possible using the varimax method in the sense that a greater proportion of absolute values of the coefficients in the rotated matrix lie closer to zero. This allows greater ease of recognition of the underlying structure in the original data array. The concept of rotation is extended to include rotation of principal components. Numerical examples are given to illustrate the application of minimum entropy rotation in both principal component analysis and factor analysis.

INTRODUCTION

Factor analysis is recognized as an effective statistical tool for extracting meaning from large arrays of multivariate data. The computer programs for factor analysis that are available have reduced the necessary calculations to routine operations (Imbrie, 1953; Cooley and Lohnes, 1962; Manson and Imbrie, 1964; Sampson, 1968; Klován, 1968; IBM, 1968). This has resulted in an increased use of factor analysis in geology (for a list of applications, see Harbaugh and Merriam, 1968). Factor analysis, however, is not as many would think one single operation, but rather a sequence of statistical procedures in which each procedure entering into the calculations is considered on an individual basis in formulating the final result. It is with one of these procedures that the present paper is concerned - analytic rotation.

Analytic rotation has persisted as a challenge to those engaged in developing factor analytic methods. The reason stems from the desire for a simple factor structure which, as a rule, direct factor solutions do not provide. Although the problem has long been recognized (Thurstone, 1947), it has only been since computers became generally available that objective analytic criteria for rotation to simple structure have been devised for practical usage. Of the several analytic criteria proposed, the varimax method due to Kaiser (1958) is by far the best known and most widely used procedure for rotating an initial factor matrix to a position of simple structure. The computer programs that have been made available have incorporated this method.

The purpose of this paper is to make available a computer program for an analytic criterion of factor rotation proposed earlier (McCammon, 1966), which approaches more closely the intuitive concept of simple structure. The method is based on the entropy concept as it is defined in information theory and is used to describe the state of a given rotated factor matrix derived from an initial factor matrix. Although the results are similar to those obtained by

the varimax method, the basic difference is that for the minimum entropy criterion, a greater proportion of factor loading values are closer to zero. For large matrices, this can amount to a significant difference.

The second purpose of the paper is to indicate how the concept of analytic rotation can be extended to the method of principal components, a near relative of factor analysis. Principal component analysis differs from factor analysis in that the extracted components explain the total variance of a given set of variables rather than the intercorrelations. Principal component analysis finds useful application in reducing the number of variables in a study and also in problems of classification. The concept of simple structure can be utilized to interpret the principal components of a system of variables in terms of the individual variables. A given subset of principal components can be rotated into a position of simple structure while preserving the total variance. This is analogous in factor analysis to the rotation of initial factors to a position of simple structure while preserving the total correlation. The minimum entropy concept is applicable to both types of rotation.

Acknowledgments. - The program was written during my employment with Gulf Research and Development Company. I wish to thank Chester Peltó for suggesting that the entropy function be used to define a criterion for rotation and for the advice offered during the course of the study. Tables 4 and 5 are reproduced with permission of the University of Chicago Press. The subroutine to invert a matrix in the computer program described is reproduced with the permission of IBM.

ROTATION OF PRINCIPAL COMPONENTS

The principal components of a set of observations which involves n variables are defined as the eigenvectors of the transformation which reduces the covariance matrix of the variables to diagonal form. The diagonal elements in the reduced matrix

are the eigenvalues corresponding to the eigenvectors and represent the variances of each of the principal components. It is usual to arrange the eigenvalues in order of decreasing value along the diagonal. Let C represent the covariance matrix of n given variables, and D the diagonal matrix with the eigenvalues arranged in order of decreasing value. Because C is a positive definite symmetric matrix, P in the transformation

$$D = P'CP \quad (1)$$

is an orthogonal matrix (Murdoch, 1957, p. 145) and contains the set of eigenvectors as column vectors. The geometrical interpretation of principal components is that of a rotation of axes about an origin in a space defined by the coordinate axes of the n variables to a new position in which axes defined as linear combinations of the original variables represent variables which are uncorrelated. The origin is taken as the point defined by the means of the variables. Each newly defined variable is termed a principal component. The vector of principal components is defined for the i^{th} sample by a $n \times 1$ column vector Z_i expressed in terms of the original variables as

$$Z_i = P'X_i \quad (2)$$

where X_i is the $n \times 1$ column vector denoting the original variables, and P is the matrix containing the eigenvectors as column vectors. Without loss of generality, we can assume that the original variables have been previously adjusted to have zero means. Thus, combining (1) and (2) and remembering that the $C = XX'$, the covariance matrix of Z is given by

$$C_z = D \quad (3)$$

(see Scheffe, 1959, p. 8) where D is the diagonal matrix containing the eigenvalues of C . The variables defined as principal components, therefore, are uncorrelated and each has a variance corresponding to the eigenvalue given in the diagonal matrix.

The point to be made is that the principal component having the largest variance, called the first principal component, represents that linear combination of the variables with the greatest possible variance. The first principal component has a variance at least as large as the greatest variance of any single variable among the original variables. It represents the best single variable for characterizing the original data. The second principal component is defined as the linear combination of variables with the second largest variance but which, also, is uncorrelated with the first principal component. This can be viewed geometrically as choosing an axis which makes a right angle with the axis defined as the first principal component. Chosen in this manner, the two principal component axes represent axes along which the scatter of points representing the samples is at a minimum. By choosing the remaining principal components in the same manner, the result is a set of mutually orthogonal

axes. The coordinates of each principal component with respect to the original variables are the eigenvectors. Because the principal component transformation is an orthogonal transformation, the total variance remains unchanged.

Usually, a large part of the total variance is accounted for by the first few principal components. This reflects the correlation that exist among the original variables. Thus the variability contained in the original samples can be represented by fewer variables in a lower dimensional space by plotting samples along the axes defined by the few significant principal components. In this manner it is possible to achieve a substantial compression of the data.

For problems involving classification, it is helpful to be able to assign a physical meaning to each principal component that is chosen to represent a group of samples. Ideally, this would mean that some physical measure could be found which was highly correlated with each principal component. More commonly principal components are interpreted in terms of the variables they represent. The dominant variables of any principal component are found by examining the coefficient defined for each variable. The coefficients, being elements of a set of orthonormal eigenvectors, form an orthonormal vector basis. Thus the coefficient denoted by p_{ij} for the i^{th} variable of the j^{th} coefficient is bounded in the interval

$$-1 \leq p_{ij} \leq 1 ;$$

moreover,

$$\sum_{i=1}^n p_{ij}^2 = 1 \quad j = 1, \dots, n.$$

The principal components of a set of data are associated most closely with those variables whose coefficient values are close to plus or minus one. With respect to the geometry, the coordinate axes of the principal components are aligned or nearly so with these same variables.

Suppose that for a set of data which involves n variables the first k ($k < n$) principal components account for the greater part of the total variance. This means that the data can be represented in a k -dimensional space without much loss of information. Suppose further we wish to interpret the principal components we have chosen to represent the n original variables. If we examine the absolute values of the coefficients, we will find usually that in addition to values close to one and zero, there will be values intermediate between these limits making the interpretation more difficult. We can eliminate this difficulty by rotating the principal components within the given k -dimensional subspace to a position where the new axes are aligned as nearly as possible with the axes representing the original variables. For $k = n$, this would be

simply a rotation to the position defined by the original variables. For $k < n$, however, the rotation to be performed is constrained within a lower dimensional space which may or may not contain one or more of the axes representing the original variables. In matrix form,

$$B = PT \quad (4)$$

where B represents the transformed $n \times k$ matrix whose columns are the new coordinate axes representing the first k principal components, and T is a $k \times k$ orthogonal matrix which represents the orthogonal transformation. Because P is composed of orthonormal column vectors and T is an orthogonal matrix, B likewise will be composed of orthonormal column vectors. The rotation preserves the total variance contained within the k -dimensional subspace (McCammon, 1966, p. 728).

Recognizing that components are more readily interpretable if the absolute values of the coefficients are close to either one or zero, the entropy function can be used to characterize the nonuniform distribution of coefficient values. The entropy expression is given by

$$H = - \sum_{i=1}^n \sum_{j=1}^k b_{ij}^2 \ln b_{ij}^2. \quad (5)$$

To achieve a state of absolute minimum entropy requires that the squared coefficient values all have a value of either one or zero. Because of the constraints imposed by the dimension of the subspace and the condition of orthogonality for the column vectors of B , however, the minimum entropy will be some greater value. We wish to find a $k \times k$ orthogonal transformation matrix T such that the coefficients, b_{ij} , of the transformed components matrix B minimize the entropy expression H given by (5). The result is a set of k components which are more interpretable in terms of the original variables but yet which will account for the same total variance as the first k principal components. The difference will be that the components will no longer be uncorrelated.

Before considering how to find T , it is of interest to note that the varimax method due to Kaiser (1958) can be formulated to fit the present situation. By imposing the same conditions of orthogonal column vectors for the transformed matrix B in (4), Kaiser's "raw" varimax is given by

$$V = \sum_{i=1}^k \left[\sum_{j=1}^n (b_{ij}^2)^2 - \frac{\left(\sum_{j=1}^n b_{ij}^2 \right)^2}{n} \right], \quad (6)$$

which is the variance one wishes to maximize. How-

ever, because

$$\sum_{i=1}^n b_{ij}^2 = 1 \text{ for all } j,$$

the maximum found for (6) is identical to the maximum found for

$$Q = \sum_{i=1}^n \sum_{j=1}^k b_{ij}^4, \quad (7)$$

which is the expression used to define the quartimax method (Harman, 1967, p. 298). Thus, for component rotation, the varimax and quartimax methods give identical solutions.

The methods for finding the maximum of expressions as (6) and (7) which involve orthogonal transformations make use of the relation that a rotation matrix T can be represented as a product of elementary rotation matrices

$$T = \prod_{j>i} T_{ij},$$

where T_{ij} has for its elements

$$\begin{aligned} \delta_{rs} & \quad r \neq i, s \neq j \\ \cos \theta_{rs} & \quad r = i, s = i, \text{ or } r = j, s = j \\ \sin \theta_{rs} & \quad r = i, s = j \\ -\sin \theta_{rs} & \quad r = j, s = i \end{aligned}$$

and

$$-\pi/4 \leq \theta_{ij} \leq \pi/4.$$

For each elementary plane rotation, it is necessary only to solve for that value of θ_{ij} which maximizes the expression. The rotations are performed in cyclic pairs (i, j) for i less than j . The iterative procedure is continued over successive cycles until no further improvement in the criterion to be maximized is obtained. For both the quartimax and the varimax method, it is possible to obtain a closed analytic expression for θ_{ij} at each iteration which will maximize the expression given in (6) or (7) (Harman, 1967, p. 300, 307). However, for the entropy function, this is not possible. Consider an elementary plane rotation defined by θ_{ij} , and (5) becomes

$$H = - \sum_{r=1}^n (p_{ri} \cos \theta_{ij} - p_{rj} \sin \theta_{ij})^2 \ln (p_{ri} \cos \theta_{ij} - p_{rj} \sin \theta_{ij})^2$$

$$\begin{aligned}
& - \sum_{r=1}^n (p_{ri} \sin \theta_{ij} + p_{rj} \cos \theta_{ij})^2 \\
& \quad \ln (p_{ri} \sin \theta_{ij} + p_{rj} \cos \theta_{ij})^2 \\
& - \sum_{r=1}^n \sum_{s=1}^k b_{rs}^2 \ln b_{rs}^2 \quad . \quad (8) \\
& \quad r, s \neq i, j
\end{aligned}$$

To minimize this expression, the partial derivative with respect to θ_{ij} would be set equal to zero, and then one would solve for the critical value, $\hat{\theta}_{ij}$, which can be done only by solving a nonlinear equation. Rather than solving for $\hat{\theta}_{ij}$ using nonlinear methods, it is more convenient to find $\hat{\theta}_{ij}$ by direct search because $\hat{\theta}_{ij}$ is contained within a known interval. Successive trial values of θ_{ij} may be substituted in (8) until a minimum value is attained. The complete minimization of (5) using elementary plane rotations is obtained in the same manner as for the varimax and quartimax method.

One may ask what advantage there is in the minimum entropy criterion if the varimax method can be adapted for principal component rotation. The answer is that the minimum entropy criterion comes closer to achieving simple structure in the sense that a greater proportion of coefficient values expressed in absolute value lie closer to zero following the transformation. This is illustrated in Figure 1. For large component matrices, this difference can be significant.

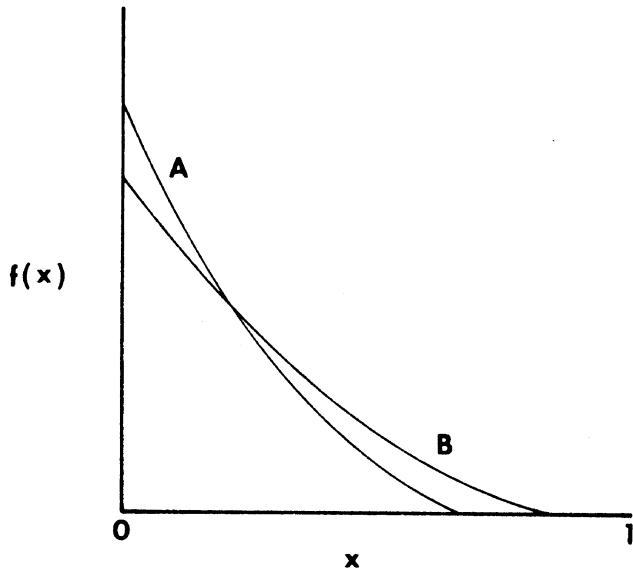


Figure 1.- Expected frequency distribution curves for coefficients of rotated data matrices. Coefficients are expressed as absolute values (A) minimum entropy rotation, and (B) varimax rotation.

FACTOR ANALYTIC ROTATION

In the previous section, we indicated how minimum entropy serves to define an analytic criterion for rotation of principal components. With some modification, we can apply the same concept to factor rotation.

In factor analysis, an initial factor matrix is obtained by any one of a number of methods (Harman, 1967, Pt. II, Chaps. 7-11). For the principal axis-solution, the initial factor matrix is made up of column vectors which are the eigenvectors multiplied by the square root of their associated eigenvalues derived from the correlation matrix of the original variables. If communalities rather than unities are placed in the diagonal elements of the correlation matrix, the resulting factor matrix is called the principal factor solution. In either situation, the initial factor matrix is the starting point for factor rotation.

In factor analysis, the interest lies in the observed correlations between variables. Each derived element in the initial factor matrix turns out to be the correlation coefficient between each variable and the corresponding factor. The goal of simple structure is to reduce to near zero as many of the correlations as possible while retaining a factor structure which accounts for the same total correlation observed for the original variables. One method for this is by an appropriate orthogonal transformation on the initial factor matrix.

The varimax method is designed to achieve simple structure. Given an initial $n \times k$ factor matrix A defined for n variables and the first k principal factors, a $k \times k$ rotation matrix T is found that defines a new $n \times k$ factor matrix B given by

$$B = AT \quad (9)$$

in such a manner that the expression

$$V = \sum_{j=1}^k \sum_{i=1}^n \left(\frac{b_{ij}}{h_i} \right)^4 - \frac{\sum_{j=1}^k \left(\sum_{i=1}^n \left(\frac{b_{ij}}{h_i} \right)^2 \right)^2}{n} \quad (10)$$

is a maximum (Harman, 1966, p. 306). The communality, h_i^2 , defined for each variable as

$$h_i^2 = \sum_{j=1}^k a_{ij}^2$$

remains invariant under the transformation.

To develop an analogous minimum entropy criterion, it is necessary to recognize the difference

that exists in the initial matrix to be rotated as compared with the initial matrix to be rotated in principal component analysis. For the initial factor matrix, the column vectors are no longer unit vectors, and, consequently, the length of the vectors will not remain invariant under an orthogonal transformation. Also, for the varimax method, the row vectors are normalized prior to rotation. The effect of this is to distribute the lengths of the column vectors of the rotated matrix as evenly as possible subject to the constraints imposed by the transformation. An expression which at its minimum value distributes the squared coefficient values within a column as unevenly as possible and the lengths of the column vectors as evenly as possible following an orthogonal transformation is given by

$$H^* = \frac{\sum_{i=1}^n \sum_{j=1}^k \frac{b_{ij}^2}{A_j} \ln \frac{b_{ij}^2}{A_j}}{\sum_{j=1}^k \frac{A_j}{A} \ln \frac{A_j}{A}} \quad (11)$$

$$\text{where } A_j = \sum_{i=1}^n b_{ij}^2, \quad \text{and } A = \sum_{j=1}^k A_j.$$

This expression defines the minimum entropy criterion for factor rotation. It has the property that, if $A_j = 1$ for all j , the solution in terms of the rotated matrix B is the same as the rotated matrix obtained in the situation of principal components. By substituting $A_j = 1$ for all j in (11),

$$H^* = \frac{H}{\ln k} \quad (12)$$

indicating that the derived solutions for both entropy functions give rise to the same minima.

The method for finding the rotation matrix T which minimizes the entropy expression in (11) is the same as before, that is, a direct search using elementary plane rotations. Convergence to the minimum value is attained in the same manner following an unspecified number of cyclic rotations.

The results obtained using the minimum entropy criterion may be compared with those obtained by the varimax method. Again, for the minimum entropy criterion, a higher proportion of absolute coefficient values in the transformed matrix lie closer

to zero. It is for this reason that the minimum entropy criterion comes closer to satisfying the ideal simple structure.

PROGRAM OPERATION

Program Dimensions

The program is dimensioned to rotate twelve components or factors for fifty variables. However the limits may be raised by increasing the dimensions for the appropriate program variables. No change in the logic of the program is necessary. To facilitate the changing of program variables, the main program and all subroutines have been written so that all double subscripted arrays are expressed as single vector arrays. In order to change the dimension of any variable, it is only necessary to make the change in the main program.

Order of Input Cards

1. Program control card
2. Title card
3. Input - output data format card
Data cards
4. Blank card

Program Usage

Card 1

Columns

- | | |
|------|--------------------------------------------------------------------|
| 1-5 | NR = number of rows (variables) in initial matrix |
| 6-10 | NC = number of columns (components or factors) in initial matrix |
| 11 | MODE = 2 analytic factor rotation
1 analytic component rotation |

Card 2

Columns 1-72 may be used for a title

Card 3

Columns 1-24 are used to specify the format to read in the initial data matrix. The data in the matrix are to be read one column at a time starting with the first column.

Columns 25-48 are used to specify the format to write out the data matrix. The data are to be written one row at a time starting with the first row.

Data Cards

(if a new set of data is to read, go to Card 1)

Card 4

A blank card follows the last data card.

Subroutines Required

All subroutines required are provided along with the main program. The user need only check to see that all subroutine decks are present before program execution. The subroutines included are:

MINENT	subroutine to perform the minimum entropy rotation.
MINSEK	subroutine to find the minimum of a nonlinear function.
COLENT	subroutine to evaluate the partial entropy for one column of a matrix
SHOOT	subroutine to calculate the entropy function for an elementary plane rotation.
MINV	subroutine to invert a matrix.

GEOLOGIC EXAMPLES

Environmental Classification

In a recent paper, multiple component analysis was shown to be an effective method for classifying large numbers of samples taken in environmental studies (McCammon, 1968). Of the other methods of classification that were compared, a method which produced similar results was based on minimum entropy rotated principal components derived from the correlation matrix of the original variables. Listed in Table 1 are the first three principal components in this study which accounted for the greater part of the total variance among the twelve original variables (Purdy, 1960). The minimum entropy rotated components are given also in Table 1. This rotation provided a set of components which could be more readily interpreted in terms of the original var-

iables and produced a classification similar to the one obtained through multiple component analysis. In order that this example might serve as a test problem for the program on a different computer, Table 2 lists the input data for this problem and Table 3 lists the output from which Table 1 was constructed.

Biostratigraphic Correlation

In an earlier paper (McCammon, 1966), principal component analysis was used as a means for establishing a regional biostratigraphic marker horizon over an area in which the relative abundance for a large number of Foraminifera had been tabulated. Table 4 lists the first seven principal components of the covariance matrix obtained for 48 Foraminifera species. The minimum entropy rotated components are given in Table 5. After rotation, the number of coefficients whose absolute values exceeded 0.25 was reduced from 34 to 24, and the number of zero loadings (rounded to two decimals) was increased from 43 to 60. Thus the rotated components were easier to interpret in that fewer variables contributed to each component. Rotated component II was used subsequently to define a subsurface marker horizon which could be traced over the entire area of study (McCammon, 1966, fig. 5).

Brine Analysis

As a final example, we consider the minimum entropy rotation in factor analysis. In this instance, we can compare results with those obtained by the varimax method. Table 6 gives an initial factor matrix based on data for a large number of brine analyses (Sampson, 1968). The rotated matrix due to the varimax method is given in Table 7. The rotated matrix due to the minimum entropy criterion is given in Table 8. As can be seen, the results are almost identical. The minimum entropy criterion, however, does produce a matrix of coefficients whose average absolute value is 0.297 which is slightly less than the average value of 0.298 obtained for the varimax method. For larger size arrays, it is expected that the differences between these two methods will be significantly greater.

Table 1.-First three principal components obtained from correlation matrix of Bahamian sediment data taken from Purdy (1960) and rotated to position of simple structure using minimum entropy criterion.

Particle Constituents	<u>Components</u>					
	Initial			Rotated (Minimum Entropy)		
	I	II	III	I	II	III
Coralline algae	.095	.410	-.391	-.009	.001	-.574
Halimeda	.231	.262	-.245	.157	-.041	-.395
Corals	.095	.401	-.442	-.026	-.039	-.602
Oolite	-.321	-.298	-.342	-.429	-.351	.039
Grapestone	-.217	.293	.489	-.035	.587	.161
Cryptocryst. gr.	-.141	.417	.413	.020	.603	.008
Peneroplidae	.372	-.024	.126	.391	-.031	.035
Other foraminifera	.349	.177	.160	.391	.138	-.081
Mollusks	.333	.216	.041	.341	.090	-.187
Wgt. % < 1/8 mm	.428	-.215	.134	.435	-.174	.166
Fecal pellets	.300	-.336	.067	.286	-.269	.231
Mud aggregates	.337	-.125	-.010	.309	-.183	.021

Table 2.-Input for Bahamian sediment data.

```
0000000001111111112222222222333333333344444444445555555555666666666677777777778
1234567890123456789012345678901234567890123456789012345678901234567890
```

12 31

BAHAMIAN SEDIMENTS DATA - FIRST THREE PRINCIPAL COMPONENTS

(12F5.3)

(1H010X3F8.3)

```
095 231 095 -321 -217 -141 372 349 333 428 300 337
410 262 401 -298 293 417 -024 177 216 -215 -336 -125
-391 -245 -442 -342 489 413 126 160 041 134 067 -010
```

Table 3.-Results of minimum entropy rotation
for Bahamian sediments data.

INITIAL COMPONENT MATRIX		
0.095	0.410	-0.391
0.231	0.262	-0.245
0.095	0.401	-0.442
-0.321	-0.298	-0.342
-0.217	0.293	0.489
-0.141	0.417	0.413
0.372	-0.024	0.126
0.349	0.177	0.160
0.333	0.216	0.041
0.428	-0.215	0.134
0.300	-0.336	0.067
0.337	-0.125	-0.010
CRITERION VALUES		
1	0.29421E	01
2	0.26303E	01
3	0.25414E	01
4	0.25411E	01
5	0.25411E	01
TRANSFORMATION MATRIX		
0.947	-0.262	-0.182
0.058	0.698	-0.713
0.315	0.666	0.677
ROTATED MATRIX		
-0.009	0.001	-0.574
0.157	-0.041	-0.395
-0.026	-0.039	-0.602
-0.429	-0.351	0.039
-0.035	0.587	0.161
0.020	0.603	0.008
0.391	-0.031	0.035
0.391	0.138	-0.081
0.341	0.090	-0.187
0.435	-0.174	0.166
0.286	-0.269	0.231
0.309	-0.183	0.021

Table 4.-First seven principal components extracted from covariance matrix of Foraminifera data taken from McCammon (1966). Leading decimal of coefficients has been omitted for convenience. Coefficients whose absolute value is 0.25 or greater are underlined.

Foram- inifera Species	I	II	III	IV	V	VI	VII
1	09	22	-54	-27	14	16	57
2	00	00	<u>01</u>	<u>00</u>	00	00	<u>01</u>
3	03	11	04	-05	-03	09	-05
4	<u>32</u>	01	00	- <u>37</u>	-05	08	-07
5	<u>14</u>	<u>32</u>	-38	<u>29</u>	-22	-67	-11
6	02	<u>02</u>	<u>02</u>	<u>04</u>	04	- <u>01</u>	01
7	<u>32</u>	-07	-01	10	02	17	-15
8	<u>12</u>	04	02	-04	-07	27	-19
9	10	42	-20	09	-13	<u>23</u>	-24
10	00	<u>12</u>	01	-03	18	-01	-08
11	00	<u>36</u>	05	-06	<u>42</u>	-03	-31
12	06	<u>00</u>	-22	-03	<u>06</u>	-07	<u>13</u>
13	05	-17	-13	-14	06	-06	-06
14	24	14	20	-05	03	-15	-01
15	30	-04	04	-29	-18	-17	-30
16	<u>32</u>	04	07	- <u>14</u>	-48	08	<u>05</u>
17	<u>14</u>	-24	-25	03	<u>17</u>	-09	-10
18	02	-03	<u>04</u>	02	03	-05	02
19	04	08	-02	-10	25	-02	-07
20	02	-05	01	-03	<u>00</u>	-02	00
21	04	27	05	-18	05	16	08
22	01	<u>00</u>	-02	-02	-02	-03	00
23	30	-34	-14	19	14	06	-01
24	<u>16</u>	- <u>12</u>	-20	-08	15	07	-22
25	07	-12	-05	06	08	-01	-05
26	24	-07	06	-27	06	-15	11
27	11	23	08	<u>07</u>	08	-05	11
28	02	04	01	00	00	06	-01
29	19	00	32	-02	02	-16	18
30	07	-05	- <u>11</u>	00	01	-02	04
31	-01	14	00	-04	30	-06	-20
32	24	-28	-10	23	<u>17</u>	00	09
33	05	-04	13	02	00	-10	12
34	07	-04	14	07	-01	-09	15
35	22	12	-04	<u>48</u>	-11	39	-02
36	04	-02	-01	<u>09</u>	04	- <u>01</u>	03
37	02	03	02	-01	-04	02	00
38	09	00	01	-02	00	00	-02
39	03	-02	08	07	-05	-11	09
40	00	01	00	01	05	00	-03
41	01	00	00	02	00	02	-01
42	-01	02	01	00	09	01	-05
43	00	04	-02	-01	02	01	-01
44	00	00	00	00	01	00	-01
45	-02	05	01	00	17	-01	-10
46	26	14	23	07	28	-05	18
47	<u>19</u>	18	21	<u>25</u>	<u>16</u>	00	22
48	00	00	00	<u>01</u>	01	-02	00

Table 5.-Rotated components using minimum entropy criterion obtained from initial principal components in Table 4. Coefficients whose absolute value is 0.25 or greater have been underlined.

Foram- inifera Species	I	II	III	IV	V	VI	VII
1	00	00	00	00	-02	-01	89
2	01	00	01	00	00	00	00
3	01	09	-10	-08	04	05	00
4	00	-02	00	-48	03	11	12
5	00	00	-01	00	00	-92	00
6	05	-01	-01	02	03	-01	00
7	05	-26	-25	-18	02	06	-07
8	-09	-01	-26	-17	03	15	-06
9	-14	16	-50	-10	14	-16	08
10	04	03	-01	-01	22	-01	01
11	05	06	-05	-03	57	-01	-07
12	-02	-10	06	00	01	-11	23
13	-12	-15	15	-11	05	00	04
14	28	05	01	-21	07	-08	-07
15	-08	-02	06	-52	01	-12	-17
16	05	08	-17	-41	-40	-04	01
17	-13	-37	09	-03	11	-09	04
18	06	-03	04	00	00	-01	-03
19	04	-02	05	-05	27	02	06
20	00	-03	04	-03	-01	02	-01
21	10	22	-11	-11	11	12	19
22	00	-01	03	-02	02	-02	02
23	03	-53	-05	-02	-03	02	00
24	-18	-26	-04	-16	20	03	04
25	-01	-18	02	00	04	00	-03
26	16	-06	23	-29	00	02	10
27	26	08	-09	00	07	-08	08
28	00	02	-06	-02	01	03	01
29	39	02	14	-10	-08	02	-08
30	-02	-10	01	-02	-02	-04	08
31	02	01	01	00	39	-03	-04
32	12	-46	00	07	-04	00	02
33	17	-01	10	00	-08	00	-04
34	21	-02	07	02	-11	-01	-04
35	11	-18	-62	14	-13	02	-02
36	05	-07	-02	05	00	-03	00
37	01	03	-03	-03	-03	00	-01
38	00	00	01	-02	01	00	-01
39	12	-01	06	03	-10	-07	-04
40	00	-01	00	01	06	00	-01
41	00	-01	-03	01	00	01	00
42	00	-01	00	01	10	02	-02
43	-01	01	-02	00	04	00	02
44	00	00	00	00	01	00	00
45	01	-01	00	03	20	01	-03
46	48	-06	-02	-03	16	04	06
47	-46	-03	-14	12	03	-01	03
48	01	-01	01	01	00	-01	00

Table 6.-Initial factor matrix of Arbuckle brine data taken from Sampson (1968).

.8667	-.3192	.0506	-.0655
.7482	-.3418	-.1126	-.0811
.2827	.4587	.0834	.5449
-.2062	-.0498	-.6663	.4700
.2528	.6907	-.3526	-.0244
.9051	-.0967	-.0543	-.0043
.7919	.2839	.0184	.0244
.8451	-.0578	-.0491	-.0338
-.0369	-.3663	.3888	.7147
-.7214	-.3258	-.3670	-.0862
-.3923	.3040	.5243	-.1116

Table 7.-Varimax rotated factor matrix of Arbuckle brine data taken from Sampson (1968).

.9175	.0233	.0994	.0872
.8303	-.0611	-.0492	.0191
.0498	.7553	-.0721	.1264
-.1639	.0852	-.8214	.0315
.0388	.5408	-.1411	-.5933
.8823	.2235	.0249	-.0501
.6405	.4919	.1326	-.1973
.8158	.2159	.0445	-.0839
-.0167	.1704	-.0705	.8736
-.5241	-.5750	-.4041	.0027
-.5025	.0903	.5205	.0459

Table 8.-Minimum entropy rotated factor matrix of Arbuckle brine data.

.9143	.0931	.0901	.0850
.8319	.0091	-.0597	.0133
-.0144	.7422	-.0629	.1981
-.1810	.0719	-.8175	.0606
-.0151	.5957	-.1515	-.5365
.8598	.3019	.0145	-.0310
.5959	.5605	.1239	-.1549
.7941	.2917	.0340	-.0657
-.0235	.0868	-.0471	.8872
-.4784	-.6135	-.4023	-.0398
-.5010	.0394	.5286	.0416

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Listing of minimum entropy rotation program.

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C	MINIMUM ENTROPY ROTATION PROGRAM	2
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C		4
C	A COMPUTER PROGRAM TO PERFORM AN ANALYTIC ROTATION BASED	5
C	ON THE MINIMUM ENTROPY CRITERION ON AN INITIAL DATA MATRIX. THE	6
C	ROTATION MAY BE PERFORMED ON EITHER A PRINCIPAL COMPONENT MATRIX	7
C	OR A FACTOR MATRIX. THE PRESENT PROGRAM IS DIMENSIONED TO ROTATE	8
C	UP TO A 50 BY 12 MATRIX. THESE LIMITS CAN BE EXTENDED BY	9
C	INCREASING THE DIMENSIONS OF THE APPROPRIATE PROGRAM VARIABLES.	10
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C	ORDER OF INPUT CARDS	13
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C	PROGRAM CONTROL PARAMETERS	21
C		22
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C	NC - NUMBER OF COLUMNS IN INITIAL MATRIX	24
C	MODE - MODE = 1, COMPONENT ROTATION	25
C	MODE = 2, FACTOR ROTATION	26
C		27
C		28
C	PROGRAM VARIABLES	29
C		30
C	X - INPUT DATA VECTOR OF LENGTH NR*NC	31
C	C - WORKING VECTOR OF LENGTH NR*NC	32
C	D,D1 - WORKING VECTORS OF LENGTH NC**2	33
C	S - WORKING VECTOR OF LENGTH NR	34
C	EV - CRITERION VECTOR OF LENGTH 10*NC	35
C	AC,AE,LC,MC - WORKING VECTORS OF LENGTH NC	36
C		37
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C		50
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C	*****	53
C	MAIN PROGRAM	54

C	*****	55
C		56
	DIMENSION FMT(18),FMI(6),FMD(6),X(600),C(600),S(50),D(144),D1(144)	57
	1,EV(120),LC(12),MC(12),AE(12),AC(12)	58
	1 READ(5,100) NR,NC,MODE	59
100	FORMAT(2I5,I1)	60
	IF(NR.EQ.0) STOP	61
	READ(5,101) (FMT(I),I=1,18)	62
101	FORMAT(18A4)	63
	READ(5,102) (FMI(I),I=1,6),(FMD(I),I=1,6)	64
102	FORMAT(12A4)	65
	I2=0	66
	DO 10 J=1,NC	67
	I1=I2+1	68
	I2=I1+NR-1	69
10	READ(5,FMI) (X(I),I=I1,I2)	70
	WRITE(6,200) (FMT(I),I=1,18)	71
200	FORMAT(1H1/1H T30,'MINIMUM ENTROPY ROTATION' //1H 5X18A4 /)	72
	GO TO (8,9),MODE	73
8	WRITE(6,203)	74
203	FORMAT(1H ' INITIAL COMPONENT MATRIX')	75
	GO TO 11	76
9	WRITE(6,205)	77
205	FORMAT(1H ' INITIAL FACTOR MATRIX')	78
11	I2=(NC-1)*NR	79
	DO 12 I=1,NR	80
	I2=I2+1	81
12	WRITE(6,FMD) (X(J),J=I,I2,NR)	82
	IF(MODE.EQ.1) GO TO 15	83
	DO 17 I=1,NR	84
	K=I-NR	85
	S(I)=0.	86
	DO 17 J=1,NC	87
	K=K+NR	88
17	S(I)=S(I)+X(K)**2	89
	DO 18 I=1,NR	90
	S(I)=SQRT(S(I))	91
	K=I-NR	92
	DO 18 J=1,NC	93
	K=K+NR	94
18	X(K)=X(K)/S(I)	95
15	K=C	96
	DO 40 I=1,NR	97
	DO 40 J=1,NC	98
	K=K+1	99
40	C(K)=X(K)	100
	CALL MINENT(NR,NC,X,NITN,EV,AE,AC)	101
	WRITE(6,207) (I,EV(I),I=1,NITN)	102
207	FORMAT(1H0' CRITERION VALUES' //(1H 10X13,E20.5))	103
	IF(NITN.LT.(10*NC)) GO TO 13	104
	WRITE(6,203) NITN	105
208	FORMAT(1H0' ERROR MESSAGE: NUMBER OF CYCLES EQUALS SET LIMIT 0	106
	IF 10*NC = ',13 //)	107
13	CONTINUE	108
	DO 41 I=1,NC	109
	K1=I-NC	110
	DO 41 J=1,NC	111
	K1=K1+NC	112
	D(K1)=C.	113
	K2=(I-1)*NR	114

K3=(J-1)*NR	115
DO 41 K=1,NR	116
K2=K2+1	117
K3=K3+1	118
41 D(K1)=D(K1)+C(K2)*X(K3)	119
IF(MODF,EQ,1) GO TO 60	120
DO 62 I=1,NC	121
K1=I-NC	122
DO 62 J=1,NC	123
K1=K1+NC	124
D1(K1)=0.	125
K2=(I-1)*NR	126
K3=(J-1)*NR	127
DO 62 K=1,NR	128
K2=K2+1	129
K3=K3+1	130
62 D1(K1)=D1(K1)+C(K2)*C(K3)	131
CALL MINV(D1,NC,DET,LC,MC)	132
DO 64 I=1,NC	133
K1=I-NC	134
DO 64 J=1,NC	135
K1=K1+NC	136
C(K1)=0.	137
K2=I-NC	138
K3=(J-1)*NC	139
DO 64 K=1,NC	140
K2=K2+NC	141
K3=K3+1	142
64 C(K1)=C(K1)+D1(K2)*D(K3)	143
K=C	144
DO 65 I=1,NC	145
DO 65 J=1,NC	146
K=K+1	147
65 D(K)=C(K)	148
DO 28 I=1,NR	149
K=I-NR	150
DO 28 J=1,NC	151
K=K+NR	152
28 X(K)=X(K)*S(I)	153
60 CONTINUE	154
WRITE(6,209)	155
209 FORMAT(1H0' TRANSFORMATION MATRIX')	156
I2=(NC-1)*NC	157
DO 16 I=1,NC	158
I2=I2+1	159
16 WRITE(6,FMC) (D(J),J=I,I2,NC)	160
WRITE(6,211)	161
211 FORMAT(1H0' ROTATED MATRIX')	162
I2=(NC-1)*NR	163
DO 14 I=1,NR	164
I2=I2+1	165
14 WRITE(6,FMC) (X(J),J=I,I2,NR)	166
GO TO 1	167
END	168

C		169
C		170
C	*****	171
	SUBROUTINE MINENT(NR,NC,X,NITN,EV,AE,AC)	172
C	*****	173
C		174
C	SUBROUTINE TO PERFORM MINIMUM ENTROPY ROTATION	175
	DIMENSION X(1),AE(1),AC(1),EV(1),T(4),H(10)	176
	A1=-.785398	177
	A2=.785398	178
	H(1)=A2/10.	179
	DO 14 J=2,5	180
14	H(J)=H(J-1)/10.	181
	SA=0.	182
	K=0	183
	DO 32 J=1,NC	184
	XS=0.	185
	AE(J)=0.	186
	DO 52 I=1,NR	187
	K=K+1	188
	XC=X(K)**2	189
	XS=XS+XC	190
	IF(XC)52,52,34	191
34	AE(J)=AE(J)-XC*ALOG(XC)	192
52	CONTINUE	193
	SA=SA+XS	194
32	AC(J)=-XS*ALOG(XS)	195
	CN=0.	196
	CD=0.	197
	ASA=ALOG(SA)	198
	ASA=SA*ASA	199
	DO 33 J=1,NC	200
	CN=CN+AE(J)	201
33	CD=CD+AC(J)	202
	NITN=1	203
	FENT=(CN+ASA)/(CD+ASA)	204
	EV(NITN)=FENT	205
	NTOL=10*NC	206
	NC1=NC-1	207
22	IF(NITN.GE.NTOL) GO TO 50	208
23	NITN=NITN+1	209
	ENT=0.	210
	DO 25 II=1,NC1	211
	II=II	212
	CN=CN-AE(II)	213
	CD=CD-AC(II)	214
	IJ=II+1	215
	DO 20 JJ=IJ,NC	216
	JJ=JJ	217
	K1=(II-1)*NR	218
	K2=(JJ-1)*NR	219
	CN=CN-AE(JJ)	220
	CD=CD-AC(JJ)	221
	CALL MINSEK(B,H,A1,A2,NR,NC,II,JJ,CN,CD,ASA,X)	222
	I(1)=COS(B)	223
	T(4)=T(1)	224
	I(3)=SIN(B)	225
	T(2)=-T(3)	226
	XN=0.	227
	XS=0.	228

DO 5 I=1,NR	229
K1=K1+1	230
K2=K2+1	231
Y=X(K1)*T(1)+X(K2)*T(2)	232
X(K2)=X(K1)*T(3)+X(K2)*T(4)	233
XC=X(K2)**2	234
IF(XC)53,53,54	235
54 XN=XN-XC*ALOG(XC)	236
53 XS=XS+XC	237
5 X(K1)=Y	238
AE(JJ)=XN	239
AC(JJ)=-XS*ALOG(XS)	240
CN=CN+AE(JJ)	241
CD=CD+AC(JJ)	242
20 CONTINUE	243
XN=0.	244
XS=0.	245
K1=(II-1)*NR	246
DO 55 I=1,NR	247
K1=K1+1	248
XC=X(K1)**2	249
XS=XS+XC	250
IF(XC)55,55,35	251
35 XN=XN-XC*ALOG(XC)	252
55 CONTINUE	253
AE(II)=XN	254
AC(II)=-XS*ALOG(XS)	255
CN=CN+AE(II)	256
CD=CD+AC(II)	257
CALL COLENT(TENT,II,X,NR)	258
ENT=ENT+TENT	259
25 CONTINUE	260
CALL COLENT(TENT,NC,X,NR)	261
ENT=ENT+TENT	262
38 FNT=(CN+ASA)/(CD+ASA)	263
EV(NITN)=ENT	264
TI=(FENT-ENT)/FENT	265
IF(TI-.000001)50,50,29	266
29 FENT=ENT	267
GO TO 22	268
50 CONTINUE	269
RETURN	270
END	271

C		272
C		273
C	*****	274
C	SUBROUTINE MINSEK (B,H,A1,A2,NR,NC,II,JJ,CN,CD,ASA,X)	275
C	*****	276
C		277
C	SUBROUTINE TO FIND THE MINIMUM OF A NONLINEAR FUNCTION	278
	DIMENSION X(1),H(1)	279
	TMB=0.	280
	ICY=0	281
2	I=C	282
	MIN=2	283
	CALL SHOOT(TMF,TMB,II,JJ,NR,NC,X,ASA,CN,CD)	284
20	B2=TMB	285
	I=I+1	286
	IF(I-5)21,21,40	287
21	B1=B2-H(I)	288
	IF(B1-A1)24,22,22	289
22	CALL SHOOT(F,B1,II,JJ,NR,NC,X,ASA,CN,CN)	290
	IF(F-TMF)23,24,24	291
23	TMF=F	292
	TMB=B1	293
	MIN=1	294
24	B3=B2+H(I)	295
	IF(B3-A2)25,25,16	296
25	CALL SHOOT(F,B3,II,JJ,NR,NC,X,ASA,CN,CN)	297
	IF(F-TMF)26,16,16	298
26	TMF=F	299
	TMB=B3	300
	MIN=3	301
	GO TO 16	302
10	MIN=2	303
	B1=TMB-H(I)	304
	IF(B1-A1)16,11,11	305
11	CALL SHOOT(F,B1,II,JJ,NR,NC,X,ASA,CN,CN)	306
	IF(F-TMF)12,16,16	307
12	TMF=F	308
	TMB=B1	309
	MIN=1	310
	GO TO 16	311
30	MIN=2	312
	B3=TMB+H(I)	313
	IF(B3-A2)31,31,16	314
31	CALL SHOOT(F,B3,II,JJ,NR,NC,X,ASA,CN,CN)	315
	IF(F-TMF)32,16,16	316
32	TMF=F	317
	TMB=B3	318
	MIN=3	319
16	GO TO(10,20,30),MIN	320
40	IF(ICY)41,41,43	321
41	IF(ABS(TMB)-A2+.00001)44,44,42	322
42	ICY=1	323
	B=TMB	324
	IF=TMF	325
	IF(B)46,47,47	326
46	TMB=A2	327
	GO TO 2	328
47	TMB=A1	329
	GO TO 2	330
43	IF(TM-F-TF)44,45,45	331
44	B=TMB	332
45	RETURN	333
	END	334

C		335
C		336
C	*****	337
	SUBROUTINE COLENT(E,J,X,NR)	338
C	*****	339
C		340
C	SUBROUTINE TO EVALUATE THE ENTROPY CRITERION FOR ONE COLUMN OF A	341
C	MATRIX	342
	DIMENSION X(1)	343
	T=0.	344
	E=0.	345
	K=(J-1)*NR	346
	DO 4 I=1,NR	347
	K=K+1	348
	XC=X(K)**2	349
	IF(XC)4,4,3	350
3	E=E-XC*ALCG(XC)	351
4	T=T+XC	352
	E=E/T+ALOG(T)	353
	END	354

C		355
C		356
C	*****	357
	SUBROUTINE SHOUT(F,B,II,JJ,NR,NC,X,ASA,CN,CD)	358
C	*****	359
C		360
C	SUBROUTINE TO EVALUATE THE ENTROPY CRITERION FOR AN ELEMENTARY	361
C	PLANE ROTATION	362
	DIMENSION X(1),T(4),IJ(2)	363
	IJ(1)=II	364
	IJ(2)=JJ	365
	T(1)=COS(B)	366
	T(4)=T(1)	367
	T(3)=SIN(B)	368
	T(2)=-T(3)	369
	F=0.	370
	CN1=0.	371
	CD1=0.	372
	DO 5 J=1,2	373
	TOT=0.	374
	TE=0.	375
	DO 7 I=1,NR	376
	Y=0.	377
	K2=(J-1)*2	378
	DO 6 K=1,2	379
	L=IJ(K)	380
	K1=I+(L-1)*NR	381
	K2=K2+1	382
6	Y=Y+X(K1)*T(K2)	383
	Y=Y**2	384
	IF(Y)7,7,8	385
8	TE=TE-Y*ALOG(Y)	386
7	TOT=TOT+Y	387
	CN1=CN1+TE	388
	CD1=CD1-TOT*ALOG(TOT)	389
5	CONTINUE	390
	F=(CN+CN1+ASA)/(CD+CD1+ASA)	391
	RETURN	392
	END	393

C		394
C		395
C	*****	396
	SUBROUTINE MINV(A,N,D,L,M)	397
C	*****	398
C		399
C	SUBROUTINE TO INVERT A MATRIX TAKEN FROM IBM SYSTEM 360 SCIENTIFIC	400
C	SUBROUTINE PACKAGE (360A-CM-03X) VERSION III PROGRAMMER'S MANUAL	401
C	APPL PROG H 20-0205-3 ,TECH. PUBL. DEPT.,WHITE PLAINS,NEW YORK,	402
C	454 P.	403
	DIMENSION A(1),L(1),M(1)	404
	D=1.0	405
	NK=-N	406
	DO 80 K=1,N	407
	NK=NK+N	408
	L(K)=K	409
	M(K)=K	410
	KK=NK+K	411
	BIGA=A(KK)	412
	DO 20 J=K,N	413
	IZ=N*(J-1)	414
	DO 20 I=K,N	415
	IJ=IZ+I	416
	10 IF(ABS(BIGA)- ABS(A(IJ))) 15,20,20	417
	15 BIGA=A(IJ)	418
	L(K)=I	419
	M(K)=J	420
	20 CONTINUE	421
C		422
C	INTERCHANGE ROWS	423
C		424
	J=L(K)	425
	IF(J=K) 35,35,25	426
	25 KI=K-N	427
	DO 30 I=1,N	428
	KI=KI+N	429
	HOLD=-A(KI)	430
	JI=KI-K+J	431
	A(KI)=A(JI)	432
	30 A(JI) =HOLD	433
C		434
C	INTERCHANGE COLUMNS	435
C		436
	35 I=M(K)	437
	IF(I=K) 45,45,33	438
	38 JP=N*(I-1)	439
	DO 40 J=1,N	440
	JK=NK+J	441
	JI=JP+J	442
	HOLD=-A(JK)	443
	A(JK)=A(JI)	444
	40 A(JI) =HOLD	445
C		446
C	DIVIDE COLUMN BY MINUS PIVOT (VALUE OF PIVOT ELEMENT IS	447
C	CONTAINED IN BIGA)	448
C		449
	45 IF(BIGA) 48,46,48	450
	46 D=0.0	451
	RETURN	452
	48 DO 55 I=1,N	453
	IF(I=K) 50,55,50	454
	50 IK=NK+I	455

A(IK)=A(IK)/(-BIGA)	456
55 CONTINUE	457
C	458
C REDUCE MATRIX	459
C	460
DO 65 I=1,N	461
IK=NK+I	462
HOLD=A(IK)	463
IJ=I-N	464
DO 65 J=1,N	465
IJ=IJ+N	466
IF(I-K) 60,65,60	467
60 IF(J-K) 62,65,62	468
62 KJ=IJ-I+K	469
A(IJ)=HOLD*A(KJ)+A(IJ)	470
65 CONTINUE	471
C	472
C DIVIDE ROW BY PIVOT	473
C	474
KJ=K-N	475
DO 75 J=1,N	476
KJ=KJ+N	477
IF(J-K) 70,75,70	478
70 A(KJ)=A(KJ)/BIGA	479
75 CONTINUE	480
C	481
C PRODUCT OF PIVOTS	482
C	483
D=D*BIGA	484
C	485
C REPLACE PIVOT BY RECIPROCAL	486
C	487
A(KK)=1.0/BIGA	488
80 CONTINUE	489
C	490
C FINAL ROW AND COLUMN INTERCHANGE	491
C	492
K=N	493
100 K=(K-1)	494
IF(K) 150,150,105	495
105 I=L(K)	496
IF(I-K) 120,120,108	497
108 JQ=N*(K-1)	498
JR=N*(I-1)	499
DO 110 J=1,N	500
JK=JQ+J	501
HOLD=A(JK)	502
JI=JR+J	503
A(JK)=-A(JI)	504
110 A(JI) =HOLD	505
120 J=M(K)	506
IF(J-K) 100,100,125	507
125 KI=K-N	508
DO 130 I=1,N	509
KI=KI+N	510
HOLD=A(KI)	511
JI=KI-K+J	512
A(KI)=-A(JI)	513
130 A(JI) =HOLD	514
GO TO 100	515
150 RETURN	516
END	517

MINIMUM ENTROPY ROTATION

ARBUCKLE BRINE DATA - FIRST FOUR PRINCIPAL FACTORS

INITIAL FACTOR MATRIX

0.8667	-0.3192	0.0506	-0.0655
0.7482	-0.3418	-0.1126	-0.0811
0.2827	0.4587	0.0834	0.5449
-0.2062	-0.0498	-0.6663	0.4700
0.2528	0.6907	-0.3526	-0.0244
0.9051	-0.0967	-0.0543	-0.0043
0.7919	0.2839	0.0184	0.0244
0.8451	-0.0578	-0.0491	-0.0338
-0.0369	-0.3663	0.3888	0.7147
-0.7214	-0.3258	-0.3670	-0.0862
-0.3923	0.3040	0.5243	-0.1116

CRITERION VALUES

1	0.25163E	C1
2	0.22956E	C1
3	0.22612E	C1
4	0.22546E	C1
5	0.22539E	C1
6	0.22539E	C1
7	0.22539E	C1

TRANSFORMATION MATRIX

0.9018	0.4201	0.0855	-0.0543
-0.3984	0.7557	0.1795	-0.4878
-0.0809	0.0576	0.8734	0.4768
-0.1465	0.4991	-0.4446	0.7293

ROTATED MATRIX

0.9143	0.0931	0.0901	0.0850
0.8319	0.0091	-0.0597	0.0133
-0.0144	0.7422	-0.0629	0.1981
-0.1810	0.0719	-0.8175	0.0606
-0.0151	0.5957	-0.1515	-0.5365
0.8598	0.3019	0.0145	-0.0310
0.5959	0.5605	0.1239	-0.1549
0.7941	0.2917	0.0340	-0.0657
-0.0235	0.0868	-0.0471	0.8872
-0.4784	-0.6135	-0.4023	-0.0398
-0.5010	0.0394	0.5286	0.0416

KANSAS GEOLOGICAL SURVEY COMPUTER PROGRAM
THE UNIVERSITY OF KANSAS, LAWRENCE

PROGRAM ABSTRACT

Title (If subroutine state in title):

MINIMUM ENTROPY CRITERION FOR ANALYTIC ROTATION

Date: June, 1969

Author, organization: Richard B. McCammon

Department of Geological Sciences, University of Illinois at Chicago

Direct inquiries to:

Name: Richard B. McCammon

Address: Department of Geological Sciences

University of Illinois at Chicago

Purpose/description: To reduce a factor analytic or principal component data matrix to a state
of minimum entropy using an orthogonal transformation

Mathematical method: A modified Jacobi procedure

Restrictions, range: The program is currently dimensioned to rotate twelve principal factors
or components for fifty variables.

Computer manufacturer: IBM

Model: 360/50

Programming language: FORTRAN IV

Memory required: 3 K Approximate running time:

Special peripheral equipment required:

Remarks (special compilers or operating systems, required word lengths, number of successful runs, other machine versions, additional information useful for operation or modification of program)

(continued from inside front cover)

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12. Computer applications in the earth sciences: Colloquium on trend analysis, edited by D.F. Merriam and N.C. Cocke, 1967	\$1.00
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