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TECHNICAL BULLETIN 65-15

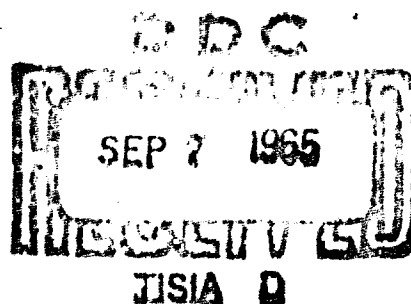
MAY 1965

A COMPUTER PROGRAM FOR THE MAXIMUM LIKELIHOOD ANALYSIS OF TYPES

John H. Wolfe

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BRIEF

This report contains a description of a computer program for estimating the parameters of a mixture of multivariate normal distributions with unknown frequencies, means, and covariances. The basic equations for the procedure are presented for the first time here, with their derivation omitted. An example with the results of the computer printout is described for an artificially constructed mixture of three bivariate normal distributions. The method of using the program and the Fortran listing are detailed in this report.

CONTENTS

BRIEF		PAGE
I	Identification	1
II	Purpose.	1
III	Restrictions	1
IV	Method	1
V	Usage.	10
VI	Example - Artificial Clusters.	12
	A. Listing of Input for Example.	17
	B. Computer Printout for Example	23
VII	Fortran Program Listing	30

A COMPUTER PROGRAM FOR THE
MAXIMUM LIKELIHOOD ANALYSIS OF TYPES

I Identification

A. TYPE

B. Written by John H. Wolfe, January 1963. Revised June, 1964.

C. U.S. Naval Personnel Research Activity, San Diego, California.

D. Coded entriely in FORTRAN II.

II Purpose

Given m scores on each of N individuals drawn from an unknown mixture of multivariate normal distributions, the program gives maximum-likelihood estimates of the means and covariances within each type, the relative frequency of each type, and the maximum likelihood of the sample for a given number of types.

III Restrictions

A. The program uses 14,554₈ words in the main program and 47,256₈ words in COMMON. One input magnetic tape (Unit #3), one output magnetic tape (Unit #2) and one tape for temporary binary storage (Unit #4) are required.

B. Restrictions on Parameters

Numbers of Variables ≤ 5

Number of Types ≤ 6

IV Method

The program solves the maximum-likelihood equations by one of four alternative iteration schemes, depending on a control card option.

Suppose that m measurements have been made on N individuals. Let x_{ik} be the i^{th} variable for individual k . Suppose the population from which the sample of N individuals is drawn is a mixture of r multivariate normal distributions. That is, the probability density, $f(\mathbf{x})$, is given by

$$f(\mathbf{x}) = \sum_{s=1}^r \lambda_s \alpha_s(\mathbf{x}), \text{ where } \sum_{s=1}^r \lambda_s = 1, \lambda_s \geq 0, \text{ and}$$

$$\alpha_s(\mathbf{x}) = \frac{1}{(2\pi)^m} \left(\frac{\sigma_s}{| \sigma_s^{ij} |} \right)^{1/2} e^{-1/2 \sum_{ij} (x_i - M_i^s) \sigma_s^{ij} (x_j - M_j^s)}$$

Here λ_s is the relative proportion of type s in population, $| \sigma_s^{ij} |$ is the determinant of the inverse of the covariance matrix for type s , and M_i^s is the mean of the i^{th} variable for type s .

Let us define $g_s(x) = \alpha_s(x) / f(x)$. SUBROUTINE DENSITY calculates α_s , f , and g_s for each individual. Define the "generalized sample moments" $\{\mu_{ijab}^{ps}\}$ as follows:

$$\mu_{ijab}^{ps} = \frac{1}{N} \sum_{k=1}^N x_{ik}^{p-1} x_{jk} x_{ak} x_{bk} g_s(x_k) g_p(x_k)$$

If a subscript is omitted or set to 0, the corresponding term on the right hand side of the equation is omitted. For example

$$\mu_{ijao}^{po} = \mu_{ija}^p = \frac{1}{N} \sum_{k=1}^N x_{ik} x_{jk} x_{ak} g_p(x_k)$$

$$\text{and } \mu_o^o = \frac{1}{N} \sum_{k=1}^N 1 = 1.$$

SUBROUTINE MOMENT computes a table of $\{\mu_{ijab}^{ps}\}$ with $b \leq a \leq j \leq i$ and $p \leq s$.

FUNCTION U (II, IJ, IA, IB, IP, IS) looks up the correct μ_{ijab}^{ps} from the two-dimensional table created by MOMENT, even when the inequalities on the indices are not satisfied.

The maximum likelihood estimates of the parameters are those values which maximize the function

$$L = \sum_{k=1}^N \log f(x_k) - \omega \left(\sum_{s=1}^r \lambda_s - 1 \right).$$

Setting the partial derivatives of the likelihood to zero results in the following equations:

$$f_{oo}^s = \mu_o^s - 1 = 0$$

$$f_{oi}^s = \mu_i^s - \mu_o^s M_i^s = 0$$

$$f_{ij}^s = \mu_{ij}^s - M_i^s \mu_j^s - M_j^s \mu_i^s - \mu_o^s (\sigma_{ij}^s - M_i^s M_j^s) = 0$$

The moments $\{\mu_{ij}^s\}$ can also be differentiated as follows:

$$\frac{\partial \mu_{ij}^s}{\partial \lambda_p} = -\mu_{ij}^{ps}$$

$$\frac{\partial \mu_{ij}^s}{\partial M_a^p} = -\lambda_p \sum_{b=1}^m \sigma_p^{ab} (\mu_{ijb}^{ps} - M_b^p \mu_{ij}^{ps}) + \delta_{ps} \sum_{b=1}^m \sigma_p^{ab} (\mu_{ijb}^p - M_b^p \mu_{ij}^p)$$

$$\begin{aligned} \frac{\partial \mu_{ij}^s}{\partial \sigma_p^{ab}} &= (1 - \frac{\delta_{ab}}{2}) \{-\lambda_p [\mu_{ij}^{ps} (\sigma_{ab}^p - M_a^p M_b^p) + M_a^p \mu_{ijb}^{ps} + M_b^p \mu_{ija}^{ps} - \mu_{ijab}^{ps}] \\ &\quad + \delta_{ps} [\mu_{ij}^p (\sigma_{ab}^p - M_a^p M_b^p) + M_a^p \mu_{ijb}^p + M_b^p \mu_{ija}^p - \mu_{ijab}^p]\} \end{aligned}$$

Where $\delta_{ps} = 1$ if $p=s$
 0 otherwise

The derivatives of the moments are computed by

$$\text{FUNCTION DERV (II, IJ, IS, IA, IB, IP)} = \frac{\partial \mu_{ij}^s}{\partial \theta_{ab}^p},$$

Where $\theta_{oa}^p = M_a^p$, $\theta_{oo}^p = \lambda_p$, and $\theta_{ab}^p = \sigma_p^{ab}$ for $a \neq o \neq b$.

DERV calls on three functions, ONE, EM, SIG defined as follows:

$$\text{ONE (IA, IB)} = \delta_{ab}$$

$$\text{EM (II, IJ, IA, IS, IP)} = \sum_{b=1}^m \sigma_p^{ab} (\mu_{ijb}^{ps} - M_b^p \mu_{ij}^{ps})$$

$$\text{SIG (II, IJ, IA, IB, IS, IP)} = \mu_{ij}^{ps} (\sigma_{ab}^p - M_a^p M_b^p) + M_a^p \mu_{ijb}^{ps} \\ + M_b^p \mu_{ija}^{ps} - \mu_{ijab}^{ps}$$

The function EM used with $s=0$ gives the second term in $\frac{\partial \mu_{ij}^s}{\partial M_a^p}$

and SIG used with $s=0$ gives the second term in $\frac{\partial \mu_{ij}^s}{\partial \sigma_{ab}^p}$.

The maximum likelihood equations $\{f_{ij}^s = 0\}$ can be solved in several ways. One iterative scheme is

Newton-Raphson iteration, which solves the linear equations

$$\sum_{p=1}^r \sum_{b=1}^m \sum_{a=b}^m \frac{\partial f_{ij}^s}{\partial \theta_{ab}^p} \Delta \theta_{ab}^p = -f_{ij}^s$$

for $\Delta \theta_{ab}^p$.

On the next iteration $\theta_{ab}^p = \theta_{ab}^p + \Delta \theta_{ab}^p$.

SUBROUTINE NEWTON computes the vector

$$B(IAT) = -f_{ij}^s \text{ and the matrix of coefficients}$$

$$A(IAT, JAT) = \partial f_{ij}^s / \partial \theta_{ab}^p.$$

SUBROUTINE MATINV, a standard SHARE routine, solves the linear equations for $\Delta \theta_{ab}^p$ and stores the result in the vector B (IAT).

SUBROUTINE RAPHSON computes the values of the parameters for the next iteration. First it determines if any of the increments are so large that $\lambda_p + \Delta \lambda_p < 0$ or > 1 . If so, the increment vector B is shortened until no $\Delta \lambda_p$ moves λ_p more than half the interval between λ_p and a boundary point.

That is, if $\Delta \lambda_p < 0$, $\Delta \lambda_p > -\frac{\lambda_p}{2}$, and if $\Delta \lambda_p > 0$, $\Delta \lambda_p < \frac{1 - \lambda_p}{2}$. After the $\Delta \theta_{ab}^p$

are shortened, they are added to the old estimates of the parameters to obtain new estimates for the next iterations.

The main routine also plays a role in shortening the increment vector. If the new likelihood is less than the previous one, or if the determinant of one of the covariance matrices as determined by RAPHSON is negative, then the increment vector B is shortened to half its previous value and subroutine RAPHSON is entered again.

Several alternative versions of $\{f_{ij}^S\}$ can be written.

At the maximum-likelihood points

$$f_{00}^S = \mu_0^S - 1 = 0, \text{ hence } \mu_0^S = 1.$$

$$f_{oi}^S = \mu_i^S - \mu_0^S M_i^S = \mu_i^S - M_i^S = 0, \text{ hence } \mu_i^S = M_i^S.$$

Substituting for μ_i^S and μ_0^S in f_{ij}^S , we have

$$f_{ij}^S = \mu_{ij}^S - M_i^S M_j^S.$$

When METH = 1 on the control card, the subroutine NEWTON iteratively solves the equations

$$1f_{00}^S = \mu_0^S - 1 = 0.$$

$$1f_{oi}^S = \mu_i^S - M_i^S = 0.$$

$$1f_{ij}^S = \mu_{ij}^S - M_i^S M_j^S = 0.$$

When METH = 2, subroutine NEWTON solves the original set of equations, hereafter referred to as $\{2f_{ij}^S = 0\}$

When METH = 3, the equations used are

$$3f_{00}^S = 1/\mu_0^S (2f_{00}^S) = 1 - 1/\mu_0^S = 0.$$

$$3f_{oi}^S = 1/\mu_0^S (2f_{oi}^S) = \mu_i^S/\mu_0^S - M_i^S = 0.$$

$$3f_{ij}^S = 1/\mu_0^S (2f_{ij}^S) = \mu_{ij}^S - M_i^S \mu_j^S/\mu_0^S - M_j^S \mu_i^S/\mu_0^S - (\sigma_{ij}^S - M_i^S M_j^S) = 0.$$

All three types of equations have the same solutions but their radii of convergence may differ.

For METH = 2,

the function $\{2f_{ij}^S\}$ can be summarized in the following formula:

$$2f_{ij}^S = \mu_{ij}^S - \delta_{io} - (1 - \delta_{io})\{\theta_i^S \mu_j^S + (1 - \delta_{jo})[\mu_o^S(\sigma_{ij}^S - \theta_i^S \theta_j^S) + \theta_j^S \mu_i^S]\}.$$

The partial derivatives of f_{ij}^S are then easily written as

$$\begin{aligned} \frac{\partial 2f_{ij}^S}{\partial \theta_{ab}^P} &= \frac{\partial \mu_{ij}^S}{\partial \theta_{ab}^P} - (1 - \delta_{io})\left\{\theta_i^S \frac{\partial \mu_j^S}{\partial \theta_{ab}^P} + (1 - \delta_{jo})\left[\frac{\partial \mu_o^S}{\partial \theta_{ab}^P}(\sigma_{ij}^S - \theta_i^S \theta_j^S) + \theta_j^S \frac{\partial \mu_i^S}{\partial \theta_{ab}^P}\right]\right\} \\ &\quad - \delta_{ps}(1 - \delta_{io})\{\delta_{bo}[\delta_{ia} \mu_j^S + (1 - \delta_{jo})(\delta_{ja} \mu_i^S - \delta_{ia} \mu_o^S \theta_j^S - \delta_{ja} \mu_o^S \theta_i^S)]\} \\ &\quad - (1 - \frac{\delta_{ab}}{2})(1 - \delta_{bo})(1 - \delta_{jo}) \mu_o^S (\sigma_{ia}^P \sigma_{jb}^P + \sigma_{ib}^P \sigma_{ja}^P) \end{aligned}$$

where $o < j < i < m$ and $o < b < a < m$.

The above formulas are used in METH = 2. If METH = 1, the functions

$$\begin{aligned} 1f_{oo}^S &= \mu_o^S - 1 \\ 1f_{oi}^S &= \mu_i^S - M_i^S \\ 1f_{ij}^S &= \mu_{ij}^S - (\sigma_{ij}^S + M_i^S M_j^S) \end{aligned}$$

are summarized by the formulas

$$\begin{aligned} 1f_{ij}^S &= \mu_{ij}^S - \delta_{io} - (1 - \delta_{io})\{\delta_{jo} \theta_i^S + (1 - \delta_{jo})(\sigma_{ij}^S + \theta_i^S \theta_j^S)\}, \text{ and} \\ \frac{\partial 1f_{ij}^S}{\partial \theta_{ab}^P} &= \frac{\partial \mu_{ij}^S}{\partial \theta_{ab}^P} - (1 - \delta_{io})\delta_{ps}\{\delta_{bo} \delta_{jo} \delta_{ia} + (1 - \delta_{jo})(\delta_{ia} \theta_j^S + \delta_{ja} \theta_i^S)\} \\ &\quad - (1 - \frac{\delta_{ab}}{2})(1 - \delta_{bo})[1 - \delta_{jo}](\sigma_{ia}^P \sigma_{jb}^P + \sigma_{ib}^P \sigma_{ja}^P). \end{aligned}$$

The functions $\{f_{ij}^s\} = \{f_{ij}^s/\mu_0^s\}$.

$$\text{Hence } \frac{\partial f_{ij}^s}{\partial \theta_{ab}^p} = \frac{1}{\mu_0^s} \frac{\partial f_{ij}^s}{\partial \theta_{ab}^p} - \frac{f_{ij}^s}{(\mu_0^s)^2} \frac{\partial \mu_0^s}{\partial \theta_{ab}^p}.$$

The maximum likelihood Newton-Raphson iteration equations give:

$$\sum_{abp} \frac{\partial f_{ij}^s}{\partial \theta_{ab}^p} \Delta \theta_{ab}^p = - f_{ij}^s.$$

Multiplying by μ_0^s and substituting, we have

$$\sum_{abp} \left[\frac{\partial f_{ij}^s}{\partial \theta_{ab}^p} - f_{ij}^s \left(\frac{1}{\mu_0^s} \frac{\partial \mu_0^s}{\partial \theta_{ab}^p} \right) \right] \Delta \theta_{ab}^p = - f_{ij}^s.$$

Thus the equations for METH = 3 are readily calculated from the equations for METH = 2. Only the matrix of coefficients has to be

changed, simply by subtracting $f_{ij}^s \left(\frac{1}{\mu_0^s} \frac{\partial \mu_0^s}{\partial \theta_{ab}^p} \right)$.

Instead of Newton-Raphson iteration, a method of successive substitutions may be used for finding M_i^s and σ_{ij}^s .

Since $f_{oi}^s = 0 = \mu_i^s - \mu_0^s M_i^s$, the value of M_i^s for the next iteration is defined as $M_i^s = \mu_i^s / \mu_0^s$.

Since $f_{ij}^s = 0 = \mu_{ij}^s - M_i^s \mu_i^s - M_j^s \mu_j^s - \mu_0^s (\sigma_{ij}^s - M_i^s M_j^s)$ we can solve for σ_{ij}^s after first substituting $M_i^s \mu_0^s$ for μ_i^s :

$$\sigma_{ij}^s = \mu_{ij}^s / \mu_0^s - M_i^s M_j^s, \text{ where } M_i^s \text{ are the new values} = \mu_i^s / \mu_0^s.$$

The new values of λ_p can be determined by Newton-Raphson iteration using only the equations $\{f_{oo}^s = \mu_0^s - 1 = 0\}$.

Differentiation of these equations leads to the system

$$\sum_{p=1}^r \frac{\partial \mu_0^s}{\partial \lambda_p} \Delta \lambda_p = - \mu_0^s + 1.$$

$$\text{or } \sum_{p=1}^r \mu_o^{ps} \Delta \lambda_p = \mu_o^s - 1$$

When the control card METH = 0, subroutine NEWTON determines the increments associated with successive substitutions.

Experience with the program seems to indicate that the NEWTON-Raphson iteration schemes have very small radii of convergence as compared with the successive substitution methods. Theoretically, however, the Newton-Raphson iteration should converge more rapidly once within its radius. Therefore, provision has been made in the program for running a fixed number of iterations with successive substitutions so as to get improved initial estimates and then switching to Newton-Raphson methods for the remaining iterations. If the control card option is -1, -2, or -3, then a certain number of iterations by successive substitution will be used before Newton-Raphson iteration by methods 1, 2, 3 respectively. The control card number IDUMP specifies the number of preliminary iterations.

The subroutine INITIAL determines the initial values of parameters preliminary to iteration. The proper determination of initial values is crucial to the successful convergence of any iteration method. The initial values determined by INITIAL are quite crude, and the researcher may wish to write his own version of INITIAL after some experimentation. The present version also allows the user to specify his own guesses of the initial values of the parameters of certain types.

The subroutine INITIAL takes a sample of 100 individuals and subjects them to a crude clustering procedure. For each individual, a count is made of the number of other individuals within a "box" two standard deviations on a side around it. That is, for individual k, the number of individuals j is counted such that $|x_{ik} - x_{ij}| \leq \sigma_i$ for all $i = 1, 2, \dots, m$. The individual with the highest count is made the centroid of the first cluster--that is, his scores are the initial estimates of the means for the first type.

The individuals in the first cluster are erased from the sample and the procedure is repeated with the remaining individuals.

The initial estimates of the λ_s are all equal to $1/r$, where r is the number of types.

The initial estimates of the covariances are the same for each type and are equal to the covariances computed from the sample of 100 taken as a whole.

After each iteration the subroutine RESULT prints the current estimates of the parameter for each type. At the end of the last iteration, the program PLACE gives the probabilities of membership in each type for each individual.

These probabilities are:

$$P(\text{individual } k \text{ \& Type } S) = \lambda_S g_S(x_k).$$

A few words should be said about the indexing used within the program. First of all, the indices of the moments do not range from 0 to m and 0 to r as in our equations, but from 1 to m+1 and 1 to r+1. Thus the value of $U(2, 3, 1, 1, 1, 4) = \mu_{12}^3$. The moments are conveniently calculated by

$$\text{setting } z_{i+1} = x_i \text{ for each } x_{ik}$$

$$\text{and } z_1 = 1.0. \text{ Similarly } G(1) = 1.0$$

$$\text{and } G(2) \text{ is the relative density for type 1, } g_1(x_k).$$

$$\text{The values for PERS } (K) = \lambda_{k-1}$$

$$\text{COV}(I, J, K) = \sigma_{ij}^{k-1} \text{ and COVIN } (I, J, K) = \sigma_{k-1}^{ij}.$$

$$\text{also AV}(I, K) = M_i^{k-1}.$$

The routine MOMENT collapses the $\{\mu_{ijab}^{ps}\}$ into a two dimensional array.

The single index KL is uniquely related to p and s and the single index IJMN is uniquely related to the indices i, j, a, and b.

In general, suppose we have an array indexed as follows:

There are M indices. The first index varies from 1 to N. Each succeeding index varies from 1 to the preceding index.

$$1 \geq IX(1) \geq \dots \geq IX(M)$$

Let $S(N, M)$ = number of elements in this array.

$$\text{Then } S(N, M) = \sum_{I=1}^N S(I, M-1)$$

$$\text{and } S(N, M) = \frac{(N+M-1)!}{M! (N-1)!} = \binom{N+M-1}{M}.$$

Let $IX(1) \geq IX(2) \geq \dots \geq IX(M)$ be a sequence of indices for a particular element of the array. The one-dimensional index of the element is

$$K = 1 + \sum_{I=1}^M \sum_{J=1}^I (IX(M-I+1) - 2 + J) / J$$

$$\text{or } K = 1 + \sum_{I=1}^M S(IX(M-I+1) - 1, I).$$

These formulas are used by the Function U to look up values of μ_{ijab}^{ps} .

V Usage

- A. Input (TAPE Unit 3, BCD-card images)
1. Title Card in columns 1-72, any alphanumeric characters.
 2. Control Card

COLS:	NAME	DEFINITION
1-4	MX	Number of variables
5-12	NX	Sample size
13-16	IRM	Number of Types Assumed. If IRM = 0, 6 analyses will be done assuming 1, 2, 3, 4, 5, 6 types.
17-20	ITERM	Maximum Number of iterations. If blank, ITERM is set to 50.
21-28	CONV	Criterion of Convergence which all parameters must satisfy between successive iterations. If blank, CONV is set to .0001.
29-32	IRUN	=1 if every iteration is printed, = 0 if only the last iteration is printed
33-36	METH	=0 if successive substitutions is used <u>+1</u> , <u>+2</u> , <u>+3</u> . if various Newton-Raphson methods are used.
37-40	IDUMP	The number of preliminary iterations by successive substitutions before Newton-Raphson iteration for METH = -1, -2, or -3.

3. Variable Format Card. This is an ordinary FORTRAN Variable Format Card according to which the data will be read.
4. Data Deck
N sets of one or more cards per individual.
5. Initial Estimates of Parameters (optional)
 - (a) Estimate control card
Cols 1-4 = K = TYPE # (1 through 6)
Cols 5-8 = λ_k = Proportion of population of type K (all 4 digits assumed after decimal point)
 - (b) Means for type K
8 digits per mean, last 4 digits assumed after decimal point.
 - (c) Standard deviations for type K (same format as (b))
 - (d) Correlation matrix for type K (1 row per card, same format as (b))
 - (e) Estimate control card for another type, etc.

A blank card terminates the reading of initial estimates. If no initial estimates are to be read, one blank card must be read. The sets of initial estimates do not have to be present for all types. For example, a set of initial estimates for type 5 may be followed by a set for type 3 followed by a blank card. If any initial parameters are read

for a given type, all parameters must be read for that type. For example, initial estimates of means for type 3 must be followed by initial estimates of standard deviations and correlations for type 3.

If no initial estimates of parameters are read in, the computer will generate its own by a clustering procedure.

Multiple runs may be made at one time by placing one batch of input cards [A(1) to A(5)] followed by another batch. The last batch of data must be followed by two blank cards.

B. Recommended typical usage.

Under most conditions, the control card A-2 will be blank in all but the first 12 columns. The composition of the input will be:

- Title Card
- Control Card (Cols 1-12 only)
- Variable Format Card
- Data Cards
- 3 blank cards.

Another highly successful set of control parameters is METH = -1 or -2 with IDUMP = 40.

C. Output (Tape No 2)

The natural logarithm of the likelihood is printed for each

iteration: $\sum_{k=1}^N \log f(x_k) - N \left(\sum_{s=1}^r \lambda_s - 1 \right)$. If the likelihood on an

iteration is not greater than that on all previous iterations, the computer prints out "Iteration--diverges". The parameters for each type are printed out on either the last iteration or on every iteration, depending on the value of IRUN. At the end of the last iteration, the probabilities of type membership for each individual, $\lambda_{sg_s}(x_k)$, are printed. If sense switch 3 is down, the computer dumps out the moments $\{\mu_{ijab}^{ps}\}$ and the matrices $\{\sigma_{ij}^s\}$ and $\{\sigma_s^{ij}\}$ following subroutine MOMENT, the matrix A of coefficients of the Newton-Raphson iteration and the vector $\{-f_{ij}^s\}$ following subroutine NEWTON, and the matrices A^{-1} and the vector $\{\Delta \theta_{ab}^p\}$ following the subroutine MATINV.

D. Time estimates.

For METH = 0, time per iteration in seconds is

$$T = .14 N (m+1)(m+2)(r+1)$$

For METH ≠ 0, time per iteration in seconds is

$$T = .0075 (m+1)(m+2)(m+3)(m+4)(r+1)(r+2)N + .00002 [(m+1)(m+2)r]^3 N$$

A typical run with N = 225, r = 1, 2, 3, 4, 5, 6, m = 2 required 50 minutes by METH = 0 on the CDC 1604.

E. Suggestions for reducing running time.

1. By intuition, cluster analysis of variables, factor analysis, or any other means, reduce the number of variables to those that are really important for discriminating types.

2. Use any good prior information from theoretical hypotheses or any other classification-clustering procedures to provide initial estimates.

3. If you have a large sample, say 10,000, don't run it all at once. Take a small sample, say 100 to 200 cases at random and run an analysis using a weak convergence criterion, say .01. Get an idea of how many types there are and some initial estimates from the small sample. Run the entire set of data using these initial estimates on specified numbers of types. The hypothesis H_r that there are r types can be tested against the alternative, H_{r-1} that there are $r-1$ types by the χ^2 test

$$\chi^2_{d.f.} = -2 \log \lambda = 2(L_r - L_{r-1})$$

$$\text{with d.f.} = (m+1)(m+2)/2$$

where L_r = likelihood printed for r types. Thus the

χ^2 test of the likelihood ratio on the small sample may give a good guess as to how many types there are. A final test on the entire sample could be performed specifying r , $r+1$, and $r-1$ types on the runs, where r is the hypothesized number of types.

VI Example - Artificial Clusters

To test this program, an example was constructed consisting of three artificial clusters in two dimensions. The points in each cluster were generated by a pseudo-random normal deviate generator. The characteristics of each cluster are given in Table 1 below. The results of the computer printout are summarized in Table 2. The points are plotted in Figure 1. The 75 points in Cluster 1 are designated by triangles in Figure 1, the 50 points in Cluster 2 are designated by squares, and the 100 points in Cluster 3 are designated by circles. Drawn around some of the squares, circles, and triangles are larger squares, circles and triangles. The larger symbols give the classification assigned by the computer program with 3 specified types. If a point does not have two symbols around it, it was correctly classified by the computer. It is evident from the figure that most points were correctly classified, and the computer's types have clear cut boundaries whereas the actual clusters overlap to some degree.

TABLE 1
 Characteristics of Artificial Clusters

	Cluster 1		Cluster 2		Cluster 3	
Number/225	.333		.222		.444	
MEANS	.05	-1.33	-.83	1.64	1.41	.85
S.D. 's	1.26	.49	.87	1.04	.92	.97
r_{12}	.1590		.4040		.4743	

TABLE 2
 Characteristics of Types from Computer Program

	Type 2		Type 3		Type 1	
λ_p	.346		.170		.484	
Means	.20	-1.31	-1.11	1.79	1.19	.93
S.D. 's	1.28	.50	.83	1.12	1.04	.91
r_{12}	.2462		.7168		.5231	

FIGURE 1

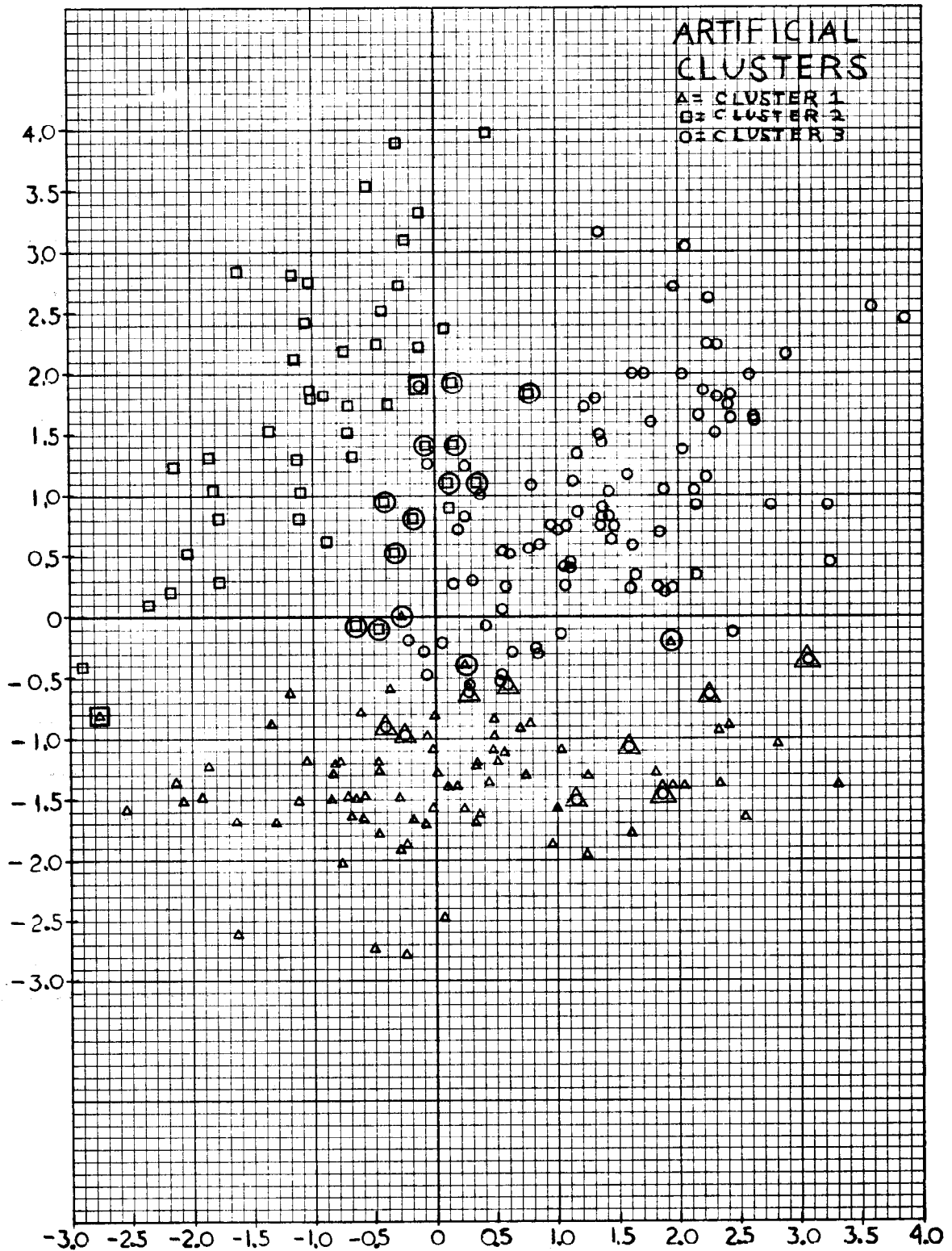


TABLE 3
Likelihoods and χ^2 for Numbers of Types

Number of Types	Natural Logarithm of Likelihood	χ^2 (with 6d.f.) =2(L _R - L _{R-1})	P
1	-380.48930		
2	-358.96468	43.04924	.114 X 10 ⁻⁶
3	-340.36400	37.20136	.161 X 10 ⁻⁵
4	-334.83078	11.06644	.863 X 10 ⁻¹
5	-325.63847	18.38462	.534 X 10 ⁻²
6	-318.02872	15.21950	.186 X 10 ⁻¹

The data cards for input are listed in Section B and the output is given in Section C.

A previous run with METH = 0, IRM = 0 (not given here) gave the likelihoods for 1 to 6 types. These are presented in Table 3, along with associated χ^2 values. The results indicate that the hypothesis that there are only two types can be rejected against the hypothesis that there are three types; but the hypothesis that there are three types cannot be rejected against the alternative that there are four types.

In order to save space in this report, only a run with IRM = 3 is given in Sections B and C.

The method used was METH = -1, which has 10 preliminary iterations by successive substitutions followed by Newton-Raphson iteration. The previous unpublished run with METH = 0 took 45 iterations to converge, while the one presented here took only 17. The difference can be attributed to the superior convergence rate of Newton-Raphson methods. However, other computer runs with IRM = 0, METH = -1 and IDUMP = 10, 30, or 40 sometimes failed to converge at all, once the Newton-Raphson procedure was started. If the likelihoods have not converged to 0.1 by the successive substitutions, then Newton-Raphson iteration often fails. Thus the initial estimates must be quite accurate if Newton-Raphson iteration is to work. So far the various methods -1, -2, -3, appear to work equally well. All three converged in exactly 17 iterations in the present example. The run reported here took 13 minutes on the CDC 1604.

SECTION VI

A. Listing of Input for Example

ARTIFICIAL CLUSTERS		METHOD-1	10PRELIMINARY ITERATIONS
2	225	3	-1 10
(2F4.2)			
-058-141			001
-008-167			002
193-017			003
-027 002			004
-002-151			005
-016-165			006
024-157			007
-070-165			008
-110-150			009
000-075			010
-214-131			011
006-243			012
281-100			013
232-090			014
-047-173			015
-082-118			016
-251-157			017
204-134			018
240-089			019
330-138			020
001-125			021
-083-129			022
-207-149			023
103-108			024
183-122			025
034-114			026
035-160			027
-194-146			028
-084-148			029
082-088			030
-077-200			031
198-134			032
-136-085			033
-130-168			034
121-191			035
-048-113			036
122-129			037
253-161			038
-036-055			039
-080-119			040
-069-162			041
-003-104			042
-072-141			043
-276-077			044
049-107			045
050-115			046
011-139			047

-162-257	048
-186-120	049
034-116	050
-164-166	051
-024-272	052
077-122	053
043-144	054
-030-144	055
-049-121	056
-108-115	057
078-086	058
-061-073	059
-050-270	060
016-135	061
232-133	062
-066-146	063
-120-060	064
-024-184	065
032-169	066
049-093	067
025-034	068
056-107	069
048-080	070
094-185	071
100-152	072
161-174	073
-007-096	074
-026-189	075
-070 134	076
-047 229	077
-030 278	078
077 186	079
-073 177	080
-107 245	081
-235-012	082
-219 025	083
-059 357	084
-286 167	085
-179 081	086
-111 105	087
013 145	088
-074 220	089
-071 155	090
-046 261	091
-068-004	092
-014 337	093
-115 292	094
-114 215	095
-179 031	096
-025 312	097

-041 098	098
-035 055	099
-103 280	100
-292-038	101
-092 184	102
-205 053	103
-186 108	104
-018 084	105
-215 128	106
045 400	107
-049-007	108
-161 288	109
-007 143	110
-111 082	111
011 111	112
-090 066	113
034 112	114
-113 133	115
013 196	116
011 091	117
-186 130	118
106 240	119
-103 187	120
-040 178	121
-138 157	122
-011 226	123
-032 391	124
-101 185	125
063-027	126
131 180	127
062 052	128
275 092	129
107 042	130
208 306	131
038 105	132
218 137	133
113-150	134
139 147	135
027 110	136
-014 190	137
060-054	138
240 179	139
161 200	140
141 084	141
086 061	142
-007 129	143
134 151	144
172 200	145
230 153	146
160 025	147

013 030	148
017 071	149
214 107	150
135 078	151
224-063	152
204 224	153
103-012	154
-009-049	155
056 056	156
181 025	157
163 060	158
189 106	159
261 163	160
205 139	161
193 126	162
359 255	163
118 187	164
109 074	165
211 038	166
023 083	167
322 093	168
011 109	169
054-042	170
111 112	171
-010-028	172
112 040	173
106 029	174
211 093	175
148 118	176
023 128	177
065 027	178
136 083	179
079 059	180
002-020	181
-031-089	182
139 092	183
287 217	184
044-002	185
241 185	186
179 160	187
096 079	188
058 009	189
221 188	190
232 180	191
222 117	192
083-027	193
148 078	194
135 316	195
197 270	196
-025-095	197

203 200
080 110
175 122
233 225
245 167
261 161
-023-018
189 022
218 169
226 261
096 018
259 197
031 032
112 141
385 246
306-035
242-010
142 106
158-102
107 076
024-054
144 064
324 046
082-022
163 037
051-050
182 071
175-143

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225
BLANK001
BLANK002
BLANK003

SECTION VI

B. Computer Printout for Example

```
ITERATION 1, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.42371292E+03
ITERATION 2, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.37195046E+03
ITERATION 3, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.36744158E+03
ITERATION 4, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.36440563E+03
ITERATION 5, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.36094806E+03
ITERATION 6, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.35693192E+03
ITERATION 7, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.35270142E+03
ITERATION 8, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.34901014E+03
ITERATION 9, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.34638251E+03
ITERATION 10, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.34476925E+03
ITERATION 11, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.34381568E+03
ITERATION 12, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.34657528E+03
ITERATION 11 DIVERGES
ITERATION 13, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.34173161E+03
ITERATION 14, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.34064823E+03
ITERATION 15, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.34037172E+03
ITERATION 16, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.34036401E+03
ITERATION 17, LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.34036400E+03
```

MAXIMUM-LIKELIHOOD ANALYSIS OF TYPES
ARTIFICIAL CLUSTERS METHOD-1 10PRELIMINARY ITERATIONS

SAMPLE SIZE = 225
NUMBER OF VARIABLES = 2
NUMBER OF TYPES = 3

ITERATION NUMBER 17

LIKELIHOOD OF 3 TYPES IN THIS SAMPLE = -.34036400E+03

CHARACTERISTICS OF THE WHOLE SAMPLE

MEANS	
.46	.30
STANDARD DEVIATIONS	
1.38	1.47
CORRELATIONS	
1.0000	.1849
.1849	1.0000

CHARACTERISTICS OF TYPE 1

THE PROPORTION OF THE POPULATION FROM THIS TYPE = .484

MEANS	
1.19	.93
STANDARD DEVIATIONS	
1.04	.91

CORRELATIONS

1.0000	.5231
.5231	1.0000

CHARACTERISTICS OF TYPE 2

THE PROPORTION OF THE POPULATION FROM THIS TYPE = .346

MEANS

.20	-1.31
-----	-------

STANDARD DEVIATIONS

1.28	.50
------	-----

CORRELATIONS

1.0000	.2462
.2462	1.0000

CHARACTERISTICS OF TYPE 3

THE PROPORTION OF THE POPULATION FROM THIS TYPE = .170

MEANS

-1.11	1.79
-------	------

STANDARD DEVIATIONS

.83	1.12
-----	------

CORRELATIONS

1.0000	.7168
.7168	1.0000

PROBABILITIES OF TYPE MEMBERSHIP

	1	2	3
1	.041	.959	.000
2	.021	.979	.000
3	.739	.261	.000
4	.957	.040	.003
5	.029	.971	.000
6	.023	.977	.000
7	.022	.978	.000
8	.025	.975	.000
9	.031	.969	.000
10	.284	.716	.000
11	.023	.945	.032
12	.009	.991	.000
13	.006	.994	.000
14	.025	.975	.000
15	.021	.979	.000
16	.072	.928	.000
17	.010	.954	.037
18	.006	.994	.000
19	.023	.977	.000
20	.000	1.000	.000
21	.057	.943	.000
22	.054	.946	.000
23	.018	.971	.012
24	.056	.944	.000
25	.013	.987	.000
26	.071	.929	.000
27	.020	.980	.000
28	.021	.971	.008
29	.034	.966	.000
30	.138	.862	.000
31	.016	.984	.000
32	.007	.993	.000
33	.150	.834	.015
34	.021	.979	.000
35	.005	.995	.000
36	.087	.913	.000
37	.022	.978	.000
38	.001	.999	.000
39	.496	.504	.000
40	.070	.930	.000
41	.026	.974	.000
42	.111	.889	.000
43	.041	.959	.000
44	.016	.344	.640
45	.084	.916	.000
46	.064	.936	.000
47	.037	.963	.000
48	.015	.985	.000
49	.039	.942	.019
50	.067	.933	.000
51	.019	.980	.001
52	.013	.987	.000
53	.042	.958	.000
54	.028	.972	.000
55	.037	.963	.000
56	.069	.931	.000
57	.072	.927	.001
58	.153	.847	.000
59	.292	.707	.001
60	.014	.986	.000

61	.041	.959	.000
62	.004	.996	.000
63	.036	.964	.000
64	.342	.626	.031
65	.017	.983	.000
66	.016	.984	.000
67	.137	.863	.000
68	.746	.254	.000
69	.081	.919	.000
70	.217	.783	.000
71	.007	.993	.000
72	.014	.986	.000
73	.004	.996	.000
74	.144	.856	.000
75	.016	.984	.000
76	.186	.000	.814
77	.020	.000	.980
78	.006	.000	.994
79	.992	.000	.008
80	.042	.000	.958
81	.001	.000	.999
82	.023	.016	.961
83	.017	.001	.982
84	.000	.000	1.000
85	.000	.000	1.000
86	.015	.000	.985
87	.089	.000	.911
88	.927	.000	.073
89	.009	.000	.991
90	.094	.000	.906
91	.006	.000	.994
92	.906	.066	.028
93	.001	.000	.999
94	.000	.000	1.000
95	.003	.000	.997
96	.054	.001	.944
97	.002	.000	.998
98	.788	.000	.212
99	.963	.001	.036
100	.000	.000	1.000
101	.008	.051	.941
102	.016	.000	.984
103	.014	.000	.986
104	.006	.000	.994
105	.960	.000	.040
106	.002	.000	.998
107	.001	.000	.999
108	.919	.073	.008
109	.000	.000	1.000
110	.810	.000	.190
111	.169	.000	.831
112	.979	.000	.021
113	.477	.000	.523
114	.995	.000	.005
115	.036	.000	.964
116	.594	.000	.406
117	.991	.000	.009
118	.004	.000	.996
119	.984	.000	.016
120	.009	.000	.991
121	.160	.000	.840
122	.007	.000	.993
123	.106	.000	.894
124	.000	.000	1.000
125	.011	.000	.989
126	.800	.200	.000

127	1.000	.000	.000
128	.999	.001	.000
129	1.000	.000	.000
130	.997	.003	.000
131	1.000	.000	.000
132	.997	.000	.003
133	1.000	.000	.000
134	.013	.987	.000
135	1.000	.000	.000
136	.993	.000	.007
137	.304	.000	.696
138	.477	.523	.000
139	1.000	.000	.000
140	1.000	.000	.000
141	1.000	.000	.000
142	.999	.001	.000
143	.883	.000	.117
144	1.000	.000	.000
145	1.000	.000	.000
146	1.000	.000	.000
147	.987	.013	.000
148	.994	.006	.001
149	.997	.000	.003
150	1.000	.000	.000
151	1.000	.000	.000
152	.105	.895	.000
153	1.000	.000	.000
154	.893	.107	.000
155	.575	.425	.000
156	.999	.001	.000
157	.984	.016	.000
158	.999	.001	.000
159	1.000	.000	.000
160	1.000	.000	.000
161	1.000	.000	.000
162	1.000	.000	.000
163	1.000	.000	.000
164	1.000	.000	.000
165	1.000	.000	.000
166	.992	.008	.000
167	.997	.000	.003
168	1.000	.000	.000
169	.981	.000	.019
170	.638	.362	.000
171	1.000	.000	.000
172	.805	.195	.000
173	.997	.003	.000
174	.993	.007	.000
175	1.000	.000	.000
176	1.000	.000	.000
177	.980	.000	.020
178	.993	.007	.000
179	1.000	.000	.000
180	.999	.001	.000
181	.868	.132	.000
182	.184	.816	.000
183	1.000	.000	.000
184	1.000	.000	.000
185	.951	.049	.000
186	1.000	.000	.000
187	1.000	.000	.000
188	1.000	.000	.000
189	.975	.025	.000
190	1.000	.000	.000
191	1.000	.000	.000
192	1.000	.000	.000

193	.786	.214	.000
194	1.000	.000	.000
195	.897	.000	.103
196	1.000	.000	.000
197	.151	.849	.000
198	1.000	.000	.000
199	1.000	.000	.000
200	1.000	.000	.000
201	1.000	.000	.000
202	1.000	.000	.000
203	1.000	.000	.000
204	.876	.123	.001
205	.979	.021	.000
206	1.000	.000	.000
207	1.000	.000	.000
208	.985	.015	.000
209	1.000	.000	.000
210	.995	.005	.000
211	1.000	.000	.000
212	1.000	.000	.000
213	.135	.865	.000
214	.709	.291	.000
215	1.000	.000	.000
216	.041	.959	.000
217	1.000	.000	.000
218	.506	.494	.000
219	.999	.001	.000
220	.982	.018	.000
221	.833	.167	.000
222	.995	.005	.000
223	.538	.462	.000
224	1.000	.000	.000
225	.007	.993	.000

VII Fortran Program Listing

```

PROGRAM TYPE TYPE0000
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5TYPE0001
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1TYPE0002
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100) TYPE0003
DIMENSION FMT(12) TYPE0004
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SDTYPE0005
1,MX,MX1,IR,XIR,IR1,IJMN,KLX,NX,XXN,JATMX,CONV,ITERM,LX,XSAM,MAT,ETYPE0006
2PH,IX,ID,ITER,PROB,METH,IDUMP TYPE0007
EQUIVALENCE (XSAM,V),(A,MAT,X) TYPE0008
1 READ INPUT TAPE 3,101,(RECORD(I),I=1,12),MX,NX,IRM,ITERM,CONV,IRUNTYPE0009
1,MOTH,IDUMP TYPE0010
101 FORMAT(12A6/I4,I8,2I4,F8.8,3I4) TYPE0011
IF(MX) 23,23,2 TYPE0012
C MX=NUMBER OF VARIABLES. TYPE0013
C THE PROGRAM READS BATCHES OF DATA UNTIL 2 BLANK CARDS ARE ENCOUNTETYPE0014
C 1RED. TYPE0015
2 IF(ITERM)3,3,4 TYPE0016
C ITERM=MAXIMUM NUMBER OF PERMISSABLE ITERATIONS. TYPE0017
3 ITERM=50 TYPE0018
4 IF(CONV)5,5,6 TYPE0019
C CONV=CRITERION OF CONVERGENCE WHICH ALL PARAMETERS MUST SATISFY BETYPE0020
C 1TWEEN SUCCESSIVE ITERATIONS. TYPE0021
5 CONV= 0.0001 TYPE0022
6 MX1=MX+1 TYPE0023
REWIND 4 TYPE0024
XXN=NX TYPE0025
NEV=XINTF(XXN/100.)+1 TYPE0026
IJMN=(MX1*(MX1+1)*(MX1+2)*(MX1+3))/24 TYPE0027
KRMN=0 TYPE0028
LX=0 TYPE0029
READ INPUT TAPE 3,102,(FMT(I),I=1,12) TYPE0030
102 FORMAT(12A6) TYPE0031
7 CALL BLOCK(NX,3000,KRMN,KST,KEND,LONG) TYPE0032
DO 8 J=1,LONG TYPE0033
8 READ INPUT TAPE 3,FMT,(X(I,J),I=1,MX) TYPE0034
DO 9 J=1,LONG,NEV TYPE0035
LX=LX+1 TYPE0036
DO 9 I=1,MX TYPE0037
9 XSAM (I,LX)=X(I,J) TYPE0038
C XSAM=SAMPLE OF UP TO 100 CASES USED FOR INITIAL ESTIMATES OF PARAMTYPE0039
C 1ETERS. TYPE0040
WRITE TAPE 4,((X(I,J),I=1,MX),J=1,LONG) TYPE0041
IF (KRMN)10,10,7 TYPE0042
10 CALL INITIAL TYPE0043
IR=IRM TYPE0044
C IRM=NUMBER OF TYPES ASSUMED. IF IRM=0, 6 DIFFERENT ANALYSES ARE TYPE0045
C DONE ASSUMING 1 TO 6 TYPES. TYPE0046
IF (IRM)11,11,12 TYPE0047
11 IR=IR+1 TYPE0048
12 IR1=IR+1 TYPE0049
METH=MOTH TYPE0050

```

	XIR=IR	TYPE0051
	KLX=(IR1*(IR1+1))/2	TYPE0052
	JATMX = (MX1*(MX1+1)*IR)/2	TYPE0053
C	PERS(K)=PROPORTION OF POPULATION OF TYPE K.	TYPE0054
	DO 913 K=2,IR1	TYPE0055
	IF(IRM)13,13,912	TYPE0056
912	IF(PERS(K))13,13,913	TYPE0057
13	PERS(K)=1./XIR	TYPE0058
913	CONTINUE	TYPE0059
	ITER=0	TYPE0060
	IDIV=0	TYPE0061
	PROBA= -10E30	TYPE0062
14	ITER=ITER+1	TYPE0063
	IF(DETERM(1)) 3500,3501,3501	TYPE0064
3500	DETERM(1)=0.0	TYPE0065
	GO TO 346	TYPE0066
3501	CALL MOMENT	TYPE0067
	WRITE OUTPUT TAPE 2,9916,ITER,IR,PROB	TYPE0068
9916	FORMAT(10H ITERATION I3,15H, LIKELIHOOD OF I3,23H TYPES IN THIS SAT	TYPE0069
	1MPLE =E18.8)	TYPE0070
	IF(PROB-PROBA)346,347,347	TYPE0071
346	IDIV=IDIV+1	TYPE0072
	ITERA=ITER-1	TYPE0073
	WRITE OUTPUT TAPE 2,9915,ITERA	TYPE0074
9915	FORMAT(10H ITERATION I3,9H DIVERGES)	TYPE0075
	DO 3445 K=1,JATMX	TYPE0076
3445	B(K)= 0.5*B(K)	TYPE0077
	IF(IDIV-1)115,3446,115	TYPE0078
3446	DO 3447 K=1,JATMX	TYPE0079
3447	B(K)= -B(K)	TYPE0080
	GO TO 115	TYPE0081
347	IDIV=0	TYPE0082
	IF(METH)344,348,348	TYPE0083
344	IF(ITER-IDUMP) 348,348,3444	TYPE0084
3444	METH=-METH	TYPE0085
C	IDUMP= THE NUMBER OF PRELIMINARY ITERATIONS BY SUCCESSIVE	TYPE0086
C	SUBSTITUTIONS BEFORE NEWTON-RAPHSON ITERATION FOR METH=-1,-2,OR -3	TYPE0087
348	PROBA=PROB	TYPE0088
	IF(SENSE SWITCH 3)349,249	TYPE0089
349	WRITE OUTPUT TAPE 2,9910	TYPE0090
9910	FORMAT(7H MOMENT)	TYPE0091
	WRITE OUTPUT TAPE 2,9914,((V(I,J),J=1,KLX),I=1,IJMNX)	TYPE0092
	WRITE OUTPUT TAPE 2,9914,(((COV(I,J,K),I=1,MX),J=1,MX),K=1,IR1)	TYPE0093
	WRITE OUTPUT TAPE 2,9914,((((COVIN(I,J,K),I=1,MX),J=1,MX),K=1,IR1)	TYPE0094
249	CONTINUE	TYPE0095
114	CALL NEWTON	TYPE0096
	IF(SENSE SWITCH3) 350,250	TYPE0097
350	WRITE OUTPUT TAPE 2,9911	TYPE0098
9911	FORMAT(7H NEWTON)	TYPE0099
	WRITE OUTPUT TAPE 2,9914,((A(I,J),I=1,JATMX),J=1,JATMX)	TYPE0100

	WRITE OUTPUT TAPE 2,9914,(B(I),I=1,JATMX)	TYPE0101
9914	FORMAT(6F12.6)	TYPE0102
250	CONTINUE	TYPE0103
	IF(METH) 115,115,248	TYPE0104
C	IF METH = 0,SUCCESSIVE SUBSTITUTION IS USED,OTHERWISE	TYPE0105
C	NEWTON-RAPHSON ITERATION.	TYPE0106
248	CALL MATINV(A,JATMX,B,1,DA)	TYPE0107
	IF(SENSE SWITCH3) 351,251	TYPE0108
351	WRITE OUTPUT TAPE 2,9912	TYPE0109
	WRITE OUTPUT TAPE 2,9914,((A(I,J),I=1,JATMX),J=1,JATMX)	TYPE0111
9912	FORMAT(7H MATINV)	TYPE0110
	WRITE OUTPUT TAPE 2,9914,(B(I),I=1,JATMX)	TYPE0112
251	CONTINUE	TYPE0113
115	CALL RAPHSON	TYPE0114
	IF(IRUN) 15,16,15	TYPE0115
C	IF IRUN=0,ONLY THE LAST ITERATION IS PRINTED.	TYPE0116
15	CALL RESULT	TYPE0117
16	IF(ITER-ITERM)17,19,19	TYPE0118
17	DO 18 K=1,JATMX	TYPE0119
	TA=B(K)	TYPE0120
	TA=ABSF(TA)	TYPE0121
	IF(TA-CONV)18,18,14	TYPE0122
18	CONTINUE	TYPE0123
19	IF(IRUN) 21,20,21	TYPE0124
20	CALL RESULT	TYPE0125
21	CALL PLACE	TYPE0126
	IF(IRM)22,22,1	TYPE0127
22	IF(IR-6)11,1,1	TYPE0128
23	END FILE 2	TYPE0129
	CALL EXIT	TYPE0130
	END(0,1,0,0,0)	TYPE0131

```

SUBROUTINE INITIAL
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5,
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100)
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SD
1,MX,MX1,IR,XIR,IR1,IJMNX,KLX,NX,XXN,JATMX,CONV,ITERM,LX,XSAM,MAT,E
2PH,IX,ID,ITER,PROB,METH,IDUMP
EQUIVALENCE (XSAM,V),(A,MAT,X)
C THIS SUBROUTINE DETERMINES INITIAL ESTIMATES OF THE MEANS AND COVARIANCES OF THE TYPES BY APPLYING A CLUSTERING PROCEDURE TO A SAMPLE
C 1RIANCES OF THE TYPES BY APPLYING A CLUSTERING PROCEDURE TO A SAMPLE
C 2E OF UP TO 100 CASES.
C FIRST THE SAMPLE MEANS AND COVARIANCES ARE DETERMINED.
XLX=LX
DO 22 I=1,MX
AV(I,1)=0.
DO 1 K=1,LX
1 AV(I,1)=AV(I,1)+XSAM(I,K)
22 AV(I,1)=AV(I,1)/XLX
DO 3 I=1,MX
DO 3 J=I,MX
COV(I,J,1)=0.
DO 2 K=1,LX
2 COV(I,J,1)=COV(I,J,1)+XSAM(I,K)*XSAM(J,K)
COV(I,J,1)=COV(I,J,1)/XLX-AV(I,1)*AV(J,1)
3 COV(J,I,1)=COV(I,J,1)
DO 4 I=1,MX
C AN LX BY LX MATRIX IS COMPUTED. AN ELEMENT CORRESPONDING TO A PAIR
C 1 OF POINTS IS 1 IF AND ONLY IF BOTH POINTS LIE WITHIN A BOX WHOSE
C 2 SIDES ARE ONE STANDARD DEVIATION LONG.
SDV=COV(I,I,1)
4 SD(I)=SQRTF(SDV)
DO 7 K=1,LX
DO 7 L=K,LX
DO 5 I=1,MX
R=XSAM(I,K)-XSAM(I,L)
R=ABSF(R)
IF(R-SD(I))5,6,6
5 CONTINUE
MAT(L,K)=1
MAT(K,L)=1
GO TO 7
6 MAT(L,K)=0
MAT(K,L)=0
7 CONTINUE
DO 14 M=2,7
C THE CENTROID OF A CLUSTER IS THE POINT WITH THE GREATEST NUMBER OF
C 1 OTHER POINTS IN A BOX AROUND IT.
C FIRST THE INDEX OF THE POINT OF GREATEST DENSITY IS FOUND. THE TY
C 1E MEANS ARE THE CO-ORDINATES OF THE DENSEST POINT.
MAX=0
DO 10 K=1,LX
LB=0
DO 8 L=1,LX

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      8 LB=LB+MAT(L,K)                                TYPE0185
      IF(LB-MAX)10,10,9                                TYPE0186
      9 MAX=LB                                          TYPE0187
      LAK=K                                            TYPE0188
      10 CONTINUE                                     TYPE0189
      DO 11 I=1,MX                                     TYPE0190
      AV(I,M)=XSAM(I,LAK)                             TYPE0191
      DO 11 J=I,MX                                     TYPE0192
      COV(J,I,M)=COV(J,I,1)                           TYPE0193
      11 COV(I,J,M)=COV(J,I,M)                         TYPE0194
      DO 14 K=1, LX                                    TYPE0195
      IF(MAT(K,LAK))14,14,12                           TYPE0196
      12 DO 13 L=1,LX                                  TYPE0197
      13 MAT(K,L)=0                                    TYPE0198
C      ALL PREVIOUSLY CLUSTERED POINTS ARE ERASED. THE CENTROID OF THE NETYPE0199
C      1XT CLUSTER THUS WILL BE THE POINT WITH THE GREATEST NUMBER OF PREVTYPE0200
C      2IOUSLY UNCLUSTERED POINTS WITHIN A BOX AROUND IT. TYPE0201
      14 CONTINUE                                     TYPE0202
      DO 15 I=1,MX                                     TYPE0203
      DO 15 J=1,MX                                     TYPE0204
      15 A(I,J)= COV(I,J,1)                             TYPE0205
      CALL MATINV(A,MX,B,0,DA)                         TYPE0206
      AD=ABSF(DA)/DA                                   TYPE0207
      DA=(SQRTF(AD/DA))*AD                             TYPE0208
      DO 16 K=1,7                                       TYPE0209
      PERS(K)=-1.0                                     TYPE0210
      DETERM(K)= DA                                    TYPE0211
      DO 16 I=1,MX                                     TYPE0212
      DO 16 J=1,MX                                     TYPE0213
      16 COVIN(I,J,K)=A(I,J)                           TYPE0214
C      THE ROUTINE ALSO READS IN INITIAL ESTIMATES UNTIL A BLANK CARD TYPE0215
C      IS ENCOUNTERED.                                TYPE0216
      17 READ INPUT TAPE 3,122,K,DA                   TYPE0217
      122 FORMAT(I4,F4.4)                               TYPE0218
      IF(K) 18,21,18                                   TYPE0219
      18 K=K+1                                          TYPE0220
      PERS(K)=DA                                       TYPE0221
      READ INPUT TAPE 3,123,(AV(I,K),I=1,MX)           TYPE0222
      READ INPUT TAPE 3,123,(SD(I),I=1,MX)            TYPE0223
      123 FORMAT(5F8.4)                                 TYPE0224
      DO 19 J=1,MX                                     TYPE0225
      READ INPUT TAPE 3,123,(COV(I,J,K),I=1,MX)       TYPE0226
      DO 19 I=1,MX                                     TYPE0227
      COV(I,J,K)=COV(I,J,K)*SD(I)*SD(J)              TYPE0228
      19 A(I,J)=COV(I,J,K)                             TYPE0229
      CALL MATINV(A,MX,B,0,DA)                         TYPE0230
      AD=ABSF(DA)/DA                                   TYPE0231
      DA=(SQRTF(AD/DA))*AD                             TYPE0232
      DETERM(K)=DA                                     TYPE0233
      DO 20 I=1,MX                                     TYPE0234
      DO 20 J=1,MX                                     TYPE0235
      20 COVIN(I,J,K)=A(I,J)                           TYPE0236
      GO TO 17                                          TYPE0237
      21 RETURN                                         TYPE0238
      END(0,1,0,0,0)                                  TYPE0239

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SUBROUTINE DENSITY
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100)
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SD
1,MX,MX1,IR,XIR,IR1,IJMN,KLX,NX,XXN,JATMX,CONV,ITERM,LX,XSAM,MAT,E
2PH,IX,ID,ITER,PROB,METH,IDUMP
EQUIVALENCE (XSAM,V),(A,MAT,X)
C FOR CURRENT ESTIMATES OF THE PARAMETERS, THIS ROUTINE COMPUTES-
C ALPHA(K)=MULTIVARIATE NORMAL DENSITY FOR TYPE K AT POINT Z.
C EPH=DENSITY AT Z FOR THE MIXTURE OF DISTRIBUTIONS
C G(K)=ALPHA/EPH AT POINT Z.
DO 3 K=2,IR1
AL=0.
DO 2 I=1,MX
IA=I+1
DO 1 J=I,MX
1 AL=AL-COVIN(J,I,K)*(Z(IA)-AV(I,K))*(Z(J+1)-AV(J,K))
2 AL=AL+0.5*COVIN(I,I,K)*(Z(IA)-AV(I,K))**2
3 ALPHA(K)=DETERM(K)*EXPF(AL)
EPH=0.
DO 4 K=2,IR1
4 EPH=EPH+PERS(K)*ALPHA(K)
DO 5 K=2,IR1
5 G(K)=ALPHA(K)/EPH
G(1)=1.
C THE ADDITION OF A DUMMY TYPE OF DENSITY G=1 SIMPLIFIES THE COMPUTAT
C ION OF LOWER ORDER MOMENTS WITH ONE OR MORE TYPES OMITTED.
RETURN
END(0,1,0,0,0)

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SUBROUTINE MOMENT                                     TYPE0270
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5TYPE0271
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1TYPE0272
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100)      TYPE0273
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SDTYPE0274
1,MX,MX1,IR,XIR,IR1,IJMNX,KLX,NX,XX,JATMX,CONV,ITERM,LX,XSAM,MAT,ETYPE0275
2PH,IX,ID,ITER,PROB,METH,IDUMP                             TYPE0276
EQUIVALENCE (XSAM,V),(A,MAT,X)                             TYPE0277
C THIS ROUTINE COMPUTES THE GENERALIZED MOMENTS V(IJMN,KL) FOR VARIATYPE0278
C 1BLES I,J,M, AND N, AND TYPES K AND L. THE MOMENTS REALLY HAVE SIX TYPE0279
C 2INDICES BUT ARE STORED AS A 2-DIMENSIONAL MATRIX. TRIANGULAR INDEXTYPE0280
C 3ING IS USED TO ELIMINATE DUPLICATION.                   TYPE0281
DO 1 IJMN=1,IJMNX                                         TYPE0282
DO 1 KL=1,KLX                                             TYPE0283
1 V(IJMN,KL)=0.                                          TYPE0284
KRMN=0                                                    TYPE0285
C INITIALIZE                                               TYPE0286
REWIND 4                                                  TYPE0287
PROB=XXN                                                 TYPE0288
DO 10 K=2,IR1                                           TYPE0289
10 PROB=PROB-XXN*PERS(K)                                 TYPE0290
2 CALL BLOCK(NX,3000,KRMN,KST,KEND,LONG)                 TYPE0291
READ TAPE 4,((X(I,J),I=1,MX),J=1,LONG)                 TYPE0292
C TAPE 4 IS READ IN BLOCKS OF 3000 CASES.                TYPE0293
DO 17 KA=1,LONG                                          TYPE0294
Z(1)=1.0                                                 TYPE0295
C THE ADDITION OF A DUMMY VARIABLE=1 SIMPLIFIES THE COMPUTATION OF LTYPE0296
C LOWER ORDER MOMENTS OMITTING ONE OR MORE VARIABLES.   TYPE0297
DO 3 I=1,MX                                             TYPE0298
3 Z(I+1)=X(I,KA)                                         TYPE0299
CALL DENSITY                                             TYPE0300
C PROB=LIKELIHOOD OF SAMPLE                               TYPE0301
IF(EPH)7,7,8                                             TYPE0302
7 INDEX=KST+KA-1                                         TYPE0303
WRITE OUTPUT TAPE 2,120,INDEX,EPH                       TYPE0304
120 FORMAT(45H NEGATIVE PROBABILITY DENSITY FOR OBSERVATION I5,3H = TYPE0305
1F16.9)                                                  TYPE0306
GO TO 9                                                  TYPE0307
8 PROB=PROB+LOGF(EPH)                                    TYPE0308
9 KL=0                                                    TYPE0309
IF(METH) 12,12,11                                       TYPE0310
11 DO 4 K=1,IR1                                          TYPE0311
DO 4 L=1,K                                               TYPE0312
KL=KL+1                                                  TYPE0313
GT1=G(K)*G(L)                                           TYPE0314
IJMN=0                                                    TYPE0315
DO 4 I=1,MX1                                             TYPE0316
GT2=GT1*Z(I)                                             TYPE0317
DO 4 J=1,I                                               TYPE0318
GT3=GT2*Z(J)                                             TYPE0319

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	DO 4 M=1,J	TYPE0320
	GT4 =GT3*Z(M)	TYPE0321
	DO 4 N=1,M	TYPE0322
	IJMN=IJMN+1	TYPE0323
	4 V(IJMN,KL)=V(IJMN,KL)+GT4*Z(N)	TYPE0324
	17 CONTINUE	TYPE0325
	15 IF(KRMN)5,5,2	TYPE0326
	5 DO 6 IJMN=1,IJMNX	TYPE0327
	DO 6 KL=1,KLX	TYPE0328
	6 V(IJMN,KL)=V(IJMN,KL)/XNX	TYPE0329
	RETURN	TYPE0330
	12 KL=1	TYPE0331
	DO 13 K=1,IR1	TYPE0332
C	FOR METH=0,WE DO NOT NEED ALL THE MOMENTS	TYPE0333
	IJMN=2	TYPE0334
	DO 14 I=2,MX1	TYPE0335
	GT1= Z(I)* G(K)	TYPE0336
	JS=0	TYPE0337
	DO 14 J=1,I	TYPE0338
	V(IJMN,KL)= V(IJMN,KL) + GT1* Z(J)	TYPE0339
	JS=JS+J	TYPE0340
	14 IJMN=IJMN+JS	TYPE0341
	DO 13 L=1,K	TYPE0342
	V(1,KL)=V(1,KL)+G(K)*G(L)	TYPE0343
	13 KL=KL+1	TYPE0344
	GO TO 17	TYPE0345
	END(0,1,0,0,0)	TYPE0346

```

FUNCTION U(IA,JA,MA,NA,KA,LA )                                TYPE0347
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5TYPE0348
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1TYPE0349
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100)        TYPE0350
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SDTYPE0351
1,MX,MX1,IR,XIR,IR1,IJMN,KLX,NX,XXN,JATMX,CONV,ITERM,LX,XSAM,MAT,ETYPE0352
2PH,IX,ID,ITER,PROB,METH,IDUMP                                TYPE0353
EQUIVALENCE (XSAM,V),(A,MAT,X)                                TYPE0354
C THIS ROUTINE FINDS THE MOMENT CORRESPONDING TO 6 INDICES BY COMPUTTYPE0355
C LING THE 2 INDICES OF THE MOMENT IN THE MATRIX V.            TYPE0356
IX(1)=IA                                                       TYPE0357
IX(2)=JA                                                       TYPE0358
IX(3)=MA                                                       TYPE0359
IX(4)=NA                                                       TYPE0360
C THE FIRST 4 INDICES ARE PUT IN ORDER FROM SMALLEST TO LARGEST TYPE0361
DO 3 J=1,4                                                     TYPE0362
KAT=100                                                         TYPE0363
DO 2 I=1,4                                                     TYPE0364
IF(KAT-IX(I)) 2,2,1                                           TYPE0365
1 KAT=IX(I)                                                    TYPE0366
K=I                                                             TYPE0367
GO TO 2                                                         TYPE0368
2 CONTINUE                                                     TYPE0369
IX(K)=100                                                       TYPE0370
3 ID(J)=KAT                                                     TYPE0371
C THE LAST 2 INDICES ARE PUT IN ORDER FROM LARGER TO SMALLER TYPE0372
IF (KA-LA)4,5,5                                               TYPE0373
4 KAT=LA                                                         TYPE0374
LAT=KA                                                         TYPE0375
GO TO 6                                                         TYPE0376
5 KAT=KA                                                         TYPE0377
LAT=LA                                                         TYPE0378
6 IJMN=1                                                         TYPE0379
C THE APPROPRIATE INDICES OF THE MOMENT MATRIX V ARE COMPUTED. TYPE0380
DO 8 I=1,4                                                     TYPE0381
LPR=1                                                           TYPE0382
DO 7 J=1,I                                                     TYPE0383
7 LPR=(LPR*(ID(I)-2+J))/J                                     TYPE0384
8 IJMN=IJMN+LPR                                               TYPE0385
KL=(KAT*(KAT-1))/2+LAT                                       TYPE0386
U=V(IJMN,KL)                                                  TYPE0387
RETURN                                                         TYPE0388
END(0,1,0,0,0)                                               TYPE0389

```

	FUNCTION ONE(IW,IU)	TYPE0390
C	THIS IS JUST THE KRONECKER DELTA	TYPE0391
	IF(IW-IU)1,2,1	TYPE0392
1	ONE=0.	TYPE0393
	GO TO 3	TYPE0394
2	ONE=1.	TYPE0395
3	RETURN	TYPE0396
	END	TYPE0397

```

FUNCTION EM(II,IJ,IA,IS,IP)                                TYPE0398
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5TYPE0399
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1TYPE0400
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100)      TYPE0401
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SDTYPE0402
1,MX,MX1,IR,XIR,IR1,IJMNX,KLX,NX,XNX,JATMX,CONV,ITERM,LX,XSAM,MAT,ETTYPE0403
2PH,IX,ID,ITER,PROB,METH,IDUMP                               TYPE0404
EQUIVALENCE (XSAM,V),(A,MAT,X)                              TYPE0405
C   EM IS A TERM WHICH APPEARS IN THE PARTIAL DERIVATIVE OF THE MAXIMUTYPE0406
C 1M LIKELIHOOD EQUATIONS WITH RESPECT TO THE TYPE MEANS.   TYPE0407
EM=0.                                                         TYPE0408
IA1=IA-1                                                      TYPE0409
DO 1 I=1,MX                                                  TYPE0410
IB=I                                                          TYPE0411
1 EM=EM+COVIN(IA1,IB,IP)*(U(1,II,IJ,IB+1,IP,IS)-AV(IB,IP)*U(1,1,II,ITYPE0412
1J,IP,IS))                                                  TYPE0413
RETURN                                                       TYPE0414
END(0,1,0,0,0)                                             TYPE0415

```

```

FUNCTION SIG(II,IJ,IA,IB,IS,IP)                                TYPE0416
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5TYPE0417
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1TYPE0418
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100)        TYPE0419
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SDTYPE0420
1,MX,MX1,IR,XIR,IR1,IJMNX,KLX,NX,XNX,JATMX,CONV,ITERM,LX,XSAM,MAT,ETTYPE0421
2PH,IX,ID,ITER,PROB,METH,IDUMP                                TYPE0422
EQUIVALENCE (XSAM,V),(A,MAT,X)                                TYPE0423
C   SIG IS A TERM WHICH APPEARS IN THE PARTIAL DERIVATIVE OF THE MAXIMTYPE0424
C   LUM LIKELIHOOD EQUATIONS WITH RESPECT TO THE ELEMENTS OF THE INVERSTYPE0425
C   2E OF THE TYPE COVARIANCE MATRIX.                          TYPE0426
IA1=IA-1                                                       TYPE0427
IB1=IB-1                                                       TYPE0428
SIG=U(1,1,II,IJ,IP,IS)*(COV(IA1,IB1,IP)-AV(IA1,IP)*AV(IB1,IP))+AV(TYPE0429
1IB1,IP)*U(1,II,IJ,IA,IP,IS)+AV(IA1,IP)*U(1,II,IJ,IB,IP,IS)-U(II,IJTYPE0430
2,IA,IB,IP,IS)                                               TYPE0431
RETURN                                                         TYPE0432
END(0,1,0,0,0)                                               TYPE0433

```



```

FUNCTION DERV(II,IJ,IS,IA,IB,IP)                                TYPE0434
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5TYPE0435
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1TYPE0436
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100)          TYPE0437
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SDTYPE0438
1,MX,MX1,IR,XIR,IR1,IJMNX,KLX,NX,XNX,JATMX,CONV,ITERM,LX,XSAM,MAT,ETYPE0439
2PH,IX,ID,ITER,PROB,METH,IDUMP                                  TYPE0440
EQUIVALENCE (XSAM,V),(A,MAT,X)                                  TYPE0441
C THIS ROUTINE COMPUTES THE DERIVATIVE OF THE IJS MOMENT WITH   TYPE0442
C RESPECT TO THE ABP PARAMETER.                                  TYPE0443
IF(IA-1)1,1,2                                                   TYPE0444
1 DERV = -U(II,IJ,1,1,IS,IP)                                    TYPE0445
GO TO 10                                                         TYPE0446
2 IF(IB-1)3,3,6                                                  TYPE0447
3 DERV=-PERS(IP)*EM(II,IJ,IA,IS,IP)                             TYPE0448
IF(IP-IS)5,4,5                                                  TYPE0449
4 DERV=DERV+EM(II,IJ,IA,1,IP)                                   TYPE0450
5 GO TO 10                                                       TYPE0451
6 DERV=-PERS(IP)*SIG(II,IJ,IA,IB,IS,IP)                        TYPE0452
IF(IP-IS)8,7,8                                                  TYPE0453
7 DERV=DERV+SIG(II,IJ,IA,IB,1,IP)                              TYPE0454
8 IF(IA-IB)10,9,10                                              TYPE0455
9 DERV=0.5*DERV                                                 TYPE0456
10 RETURN                                                        TYPE0457
END                                                                TYPE0458

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SUBROUTINE NEWTON
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100)
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SD
1,MX,MX1,IR,XIR,IR1,IJMN,KLX,NX,XXN,JATMX,CONV,ITERM,LX,XSAM,MAT,E
2PH,IX,ID,ITER,PROB,METH,IDUMP
EQUIVALENCE (XSAM,V),(A,MAT,X)
C THIS ROUTINE SETS UP THE NEWTON-RAPHSON MATRIX EQUATION FOR THE CH
C LANGES IN THE PARAMETERS FOR THE NEXT ITERATION.
MEK=XABSF(METH)
C METH = +1 OR -1 IMPLIES SIMPLIFIED MAXIMUM-LIKELIHOOD EQUATIONS
C METH = +2 OR -2 IMPLIES COMPLETE MAXIMUM-LIKELIHOOD EQUATIONS
C METH= +3 OR -3 IMPLIES COMPLETE EQUATIONS DIVIDED BY UOS
IF(METH)42,42,8
8 IAT=0
DO 27 K=2,IR1
UOS=U(1,1,1,1,K,1)
DO 27 J=1,MX1
UJS=U(1,1,1,J,K,1)
DO 27 I=J,MX1
UIS=U(1,1,1,I,K,1)
JAT=0
IAT=IAT+1
DO 27 L=2,IR1
DO 27 N=1,MX1
DO 27 M=N,MX1
JAT=JAT+1
TEMP=DERV(I,J,K,M,N,L)
IF(I-1)27,27,9
9 GO TO (12,10,10),MEK
10 TEMP=TEMP-AV(I-1,K)*DERV(J,1,K,M,N,L)
IF(J-1)12,12,11
11 TEMP=TEMP-(COV(I-1,J-1,K)-AV(I-1,K)*AV(J-1,K))*DERV(1,1,K,M,N,L)
1-AV(J-1,K)*DERV(I,1,K,M,N,L)
12 IF(K-L)27,13,27
13 IF(N-1)14,14,17
14 GO TO (16,15,15),MEK
15 TEMP=TEMP-ONE(I,M)*UJS-(1.0-ONE(J,1))*(ONE(J,M)*UIS-UOS*
1(ONE(I,M)*AV(J-1,K)+ONE(J,M)*AV(I-1,K)))
GO TO 27
16 TEMP=TEMP-ONE(J,1)*ONE(I,M)-(1.0-ONE(J,1))*(ONE(I,M)*AV(J-1,K)+
1ONE(J,M)*AV(I-1,K))
GO TO 27
17 IF(J-1)27,27,18
18 TEM=COV(I-1,M-1,L)*COV(J-1,N-1,L)+COV(I-1,N-1,L)*COV(J-1,M-1,L)
GO TO (20,19,19),MEK
19 TEM=TEM*UOS
20 IF(M-N)22,21,22
21 TEM=0.5*TEM
22 TEMP=TEMP+TEM
27 A(IAT,JAT)=TEMP
L=0
DO 33 K=2,IR1
L=L+1

```

UOS=U(1,1,1,1,K,1)	TYPE0514
B(L)=1.0-UOS	TYPE0515
DO 29 I=2,MX1	TYPE0516
L=L+1	TYPE0517
B(L)=AV(I-1,K)	TYPE0518
GO TO (29,28,28),MEK	TYPE0519
28 B(L)=B(L)*UOS	TYPE0520
29 B(L)=B(L)-U(1,1,1,I,K,1)	TYPE0521
DO 33 J=2,MX1	TYPE0522
DO 33 I=J,MX1	TYPE0523
L=L+1	TYPE0524
GO TO (31,30,30),MEK	TYPE0525
30 B(L)=UOS*(COV(I-1,J-1,K)-AV(I-1,K)*AV(J-1,K))+AV(I-1,K)*	TYPE0526
1U(1,1,1,J,K,1)+AV(J-1,K)*U(1,1,1,I,K,1)	TYPE0527
GO TO 33	TYPE0528
31 B(L)=COV(I-1,J-1,K)+AV(I-1,K)*AV(J-1,K)	TYPE0529
33 B(L)=B(L)-U(1,1,I,J,K,1)	TYPE0530
GO TO (37,37,34),MEK	TYPE0531
34 IAT=0	TYPE0532
DO 36 K=2,IR1	TYPE0533
JAT=0	TYPE0534
DO 35 L=2,IR1	TYPE0535
DO 35 N=1,MX1	TYPE0536
DO 35 M=N,MX1	TYPE0537
JAT=JAT+1	TYPE0538
35 BV(JAT)= DERV(1,1,K,M,N,L)/U(1,1,1,1,K,1)	TYPE0539
DO 36 J=1,MX1	TYPE0540
DO 36 I=J,MX1	TYPE0541
IAT=IAT+1	TYPE0542
JAT=0	TYPE0543
DO 36 L=2,IR1	TYPE0544
DO 36 N=1,MX1	TYPE0545
DO 36 M=N,MX1	TYPE0546
JAT=JAT+1	TYPE0547
36 A(IAT,JAT)= A(IAT,JAT)+B(IAT)*BV(JAT)	TYPE0548
37 RETURN	TYPE0549
42 DO 44 I=1,IR	TYPE0550
DO 43 J=1,I	TYPE0551
A(I,J)=U(1,1,1,1,I+1,J+1)	TYPE0552
43 A(J,I)=A(I,J)	TYPE0553
44 B(I)= U(1,1,1,1,I+1)-1.0	TYPE0554
CALL MATINV(A,IR,B,1,DA)	TYPE0555
L=IR	TYPE0556
DO 52 K=2,IR1	TYPE0557
PERT= U(1,1,1,1,1,K)	TYPE0558
DO 52 I=1,MX	TYPE0559
L=L+1	TYPE0560
B(L)=U(1,1,1,I+1,1,K)/PERT -AV(I,K)	TYPE0561
BV(I)= AV(I,K)+B(L)	TYPE0562
DO 52 J=1,I	TYPE0563
L=L+1	TYPE0564
52 B(L)=U(1,1,J+1,I+1,1,K)/PERT -BV(I)*BV(J)-COV(I,J,K)	TYPE0565
GO TO 37	TYPE0566
END	TYPE0567

```

SUBROUTINE RAPHSON                                         TYPE0568
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5TYPE0569
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1TYPE0570
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100)     TYPE0571
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SDTYPE0572
1,MX,MX1,IR,XIR,IR1,IJMNX,KLX,NX,XNX,JATMX,CONV,ITERM,LX,XSAM,MAT,ETYPE0573
2PH,IX,ID,ITER,PROB,METH,IDUMP                             TYPE0574
EQUIVALENCE (XSAM,V),(A,MAT,X)                             TYPE0575
C THIS ROUTINE ADDS THE APPROPRIATE CHANGES TO THE PARAMETERSTO OBTTYPE0576
C IAIN THEIR VALUES FOR THE NEXT ITERATION.               TYPE0577
AKAT=2.0                                                    TYPE0578
IF(METH)30,30,23                                           TYPE0579
C NEWTON-RAPHSON ITERATION                                TYPE0580
23 IAT=1                                                    TYPE0581
MXT=(MX1*(MX1+1))/2                                        TYPE0582
C SHORTEN INCREMENT VECTOR UNTIL ALL PERCENTAGES ARE WITHIN BOUNDSTYPE0583
DO 10 K=2,IR1                                             TYPE0584
AKAZ=-PERS(K)/(2.0*B(IAT))                                TYPE0585
IF(AKAZ)7,7,8                                             TYPE0586
7 AKAZ =AKAZ + 0.5/B(IAT)                                  TYPE0587
8 IF(AKAZ-AKAT)9,10,10                                     TYPE0588
9 AKAT=AKAZ                                               TYPE0589
10 IAT=IAT+MXT                                           TYPE0590
IF(AKAT-1.0)11,13,13                                      TYPE0591
11 DO 12 K=1,JATMX                                       TYPE0592
12 B(K)=AKAT*B(K)                                         TYPE0593
13 IAT =0                                                 TYPE0594
DO 26 K=2,IR1                                             TYPE0595
IAT=IAT+1                                                 TYPE0596
PERS(K)=PERS(K)+B(IAT)                                    TYPE0597
DO 24I=1,MX                                               TYPE0598
IAT=IAT+1                                                 TYPE0599
24 AV(I,K)=AV(I,K)+B(IAT)                                 TYPE0600
DO 25 J=1,MX                                              TYPE0601
DO 25I=J,MX                                               TYPE0602
IAT=IAT+1                                                 TYPE0603
COVIN(I,J,K)=COVIN(I,J,K)+B(IAT)                         TYPE0604
COVIN(J,I,K)=COVIN(I,J,K)                                TYPE0605
A(I,J)=COVIN(I,J,K)                                       TYPE0606
25 A(J,I)=A(I,J)                                          TYPE0607
CALL MATINV(A,MX,BV,0,DA)                                  TYPE0608
AD=ABSF(DA)/DA                                            TYPE0609
DETERM(K)=(SQRTF(AD*DA))*AD                                TYPE0610
IF(AD)27,27,28                                           TYPE0611
27 INDEX=K-1                                              TYPE0612
WRITE OUTPUT TAPE 2,121,INDEX,DA                          TYPE0613
121 FORMAT(44H DETERMINANT OF COVARIANCE INVERSE FOR TYPE I2,1H=F16.9)TYPE0614
DETERM(1)= -1.0                                          TYPE0615
28 DO 26 J=1,MX                                           TYPE0616
DO 26 I=1,MX                                              TYPE0617
26 COV(I,J,K)=A(I,J)                                      TYPE0618
29 RETURN                                                  TYPE0619

```

C	SUCCESSIVE SUBSTITUTIONS.	TYPE0620
30	L=IR	TYPE0621
	MXT=1	TYPE0622
C	SHORTEN INCREMENT VECTOR UNTIL ALL PERCENTAGES ARE WITHIN BOUNDS	TYPE0623
	DO 36 K=2,IR1	TYPE0624
	AKAZ=-PERS(K)/(2.0*B(K-1))	TYPE0625
	IF(AKAZ)33,33,34	TYPE0626
33	AKAZ = AKAZ +0.5/B(K-1)	TYPE0627
34	IF(AKAZ-AKAT)35,36,36	TYPE0628
35	AKAT=AKAZ	TYPE0629
36	CONTINUE	TYPE0630
	IF(AKAT-1.0)37,39,39	TYPE0631
37	DO 38 K=1,IR	TYPE0632
38	B(K)=AKAT*B(K)	TYPE0633
39	DO 55 K= 2,IR1	TYPE0634
	PERS(K)=PERS(K)+ B(K-1)	TYPE0635
	DO 52 I=1,MX	TYPE0636
	L=L+1	TYPE0637
47	AV(I,K)=AV(I,K)+B(L)	TYPE0638
	DO 52 J=1,I	TYPE0639
	L=L+1	TYPE0640
49	COV(I,J,K)=COV(I,J,K)+B(L)	TYPE0641
51	COV(J,I,K)=COV(I,J,K)	TYPE0642
	A(I,J)=COV(I,J,K)	TYPE0643
	A(J,I)=A(I,J)	TYPE0644
52	CONTINUE	TYPE0645
53	CALL MATINV(A,MX,BV,0,DA)	TYPE0646
	AD=ABSF(DA)/DA	TYPE0647
	DETERM(K)=(SQRTF(AD/DA))*AD	TYPE0648
	DO 54 I=1,MX	TYPE0649
	DO 54 J=1,MX	TYPE0650
54	COVIN(I,J,K)=A(I,J)	TYPE0651
55	CONTINUE	TYPE0652
	GO TO 29	TYPE0653
	END	TYPE0654

```

SUBROUTINE RESULT
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100)
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SD
1,MX,MX1,IR,XIR,IR1,IJMNX,KLX,NX,XXN,JATMX,CONV,ITERM,LX,XSAM,MAT,E
2PH,IX,ID,ITER,PROB,METH,IDUMP
EQUIVALENCE (XSAM,V),(A,MAT,X)
C THIS ROUTINE PRINTS OUT THE PARAMETERS DESCRIBING EACH TYPE.
1 FORMAT(1H1,15X,36HMAXIMUM-LIKELIHOOD ANALYSIS OF TYPES//4X,11A6)
2 FORMAT(1H1,15X,35HCHARACTERISTICS OF THE WHOLE SAMPLE//3X)
3 FORMAT(3X/////16X,23HCHARACTERISTICS OF TYPE I4//3X)
4 FORMAT(11X,49HTHE PROPORTION OF THE POPULATION FROM THIS TYPE =F6.
13/3X)
5 FORMAT(30X,5HMEANS/5F12.2)
6 FORMAT(3X/23X,19HSTANDARD DEVIATIONS /5F12.2)
7 FORMAT(3X/28X,12HCORRELATIONS)
8 FORMAT (5F12.4)
9 FORMAT(3X/////10X,13HSAMPLE SIZE =I10/10X,21HNUMBER OF VARIABLES =
1I2/10X,17HNUMBER OF TYPES =I6//25X,16HITERATION NUMBER I3//3X,
213HLIKELIHOOD OF I2,23H TYPES IN THIS SAMPLE =E18.8)
DO 10 J=1,MX
JA=J+1
10 AV(J,1)=U(1,1,1,JA,1,1)
DO 20 J=1,MX
JA=J+1
DO 20 I=J,MX
IA=I+1
COV(I,J,1)=U(1,1,JA,IA,1,1)-AV(J,1)*AV(I,1)
20 COV(J,I,1)=COV(I,J,1)
WRITE OUTPUT TAPE 2,1,(RECORD(I),I=1,11)
WRITE OUTPUT TAPE 2,9,NX,MX,IR,ITER,IR,PROB
DO 16 K=1,IR1
K1=K-1
IF (K1) 11,11,12
11 WRITE OUTPUT TAPE 2,2
GO TO 13
12 WRITE OUTPUT TAPE 2,3,K1
WRITE OUTPUT TAPE 2,4,PERS(K)
13 WRITE OUTPUT TAPE 2,5,(AV(I,K),I=1,MX)
DO 14 I=1,MX
SDV=COV(I,I,K)
AD=ABSF(SDV)/SDV
SDV=(SQRTF(AD*SDV))*AD
14 SD(I)=SDV
DO 15 I=1,MX
DO 15 J=I,MX
A(I,J)=COV(I,J,K)/(SD(I)*SD(J))
15 A(J,I)=A(I,J)
WRITE OUTPUT TAPE 2,6,(SD(I),I=1,MX)
WRITE OUTPUT TAPE 2,7
DO 16 J=1,MX
16 WRITE OUTPUT TAPE 2,8,(A(I,J),I=1,MX)
RETURN
END(0,1,0,0,0)

```

```

SUBROUTINE PLACE
DIMENSION X(5,3000),Z(6),PER(7),PERS(7),AV(5,7),COV(5,5,7),COVIN(5
1,5,7),ALPHA(7),DETERM(7),G(7),IX(4),ID(4),V(126,28),A(126,126),B(1
226),BV(126),RECORD(12),SD(5),XSAM(5,100),MAT(100,100)
COMMON X,Z,PER,PERS,AV,COV,COVIN,ALPHA,DETERM,G,V,A,B,BV,RECORD,SD
1,MX,MX1,IR,XIR,IR1,IJMNX,KLX,NX,XNX,JATMX,CONV,ITERM,LX,XSAM,MAT,E
2PH,IX,ID,ITER,PROB,METH,IDUMP
EQUIVALENCE (XSAM,V),(A,MAT,X)
C THIS ROUTINE PRINTS OUT THE PROBABILITIES OF TYPE MEMBERSHIP FOR
C EACH OBSERVATION
1 FORMAT(1H1,9X,32HPROBABILITIES OF TYPE MEMBERSHIP//7X,7I8)
2 FORMAT(I9,7F8.3)
WRITE OUTPUT TAPE 2,1,(I,I=1,IR)
KB=0
KRMN=0
REWIND 4
3 CALL BLOCK(NX,3000,KRMN,KST,KEND,LONG)
READ TAPE 4,((X(I,J),I=1,MX),J=1,LONG)
DO 6 K=1,LONG
DO 4 I=1,MX
4 Z(I+1)=X(I,K)
CALL DENSITY
KB=KB+1
DO 5 I=2,IR1
5 G(I)=G(I)*PERS(I)
6 WRITE OUTPUT TAPE 2,2,KB,(G(I),I=2,IR1)
IF(KRMN)7,7,3
7 RETURN
END(0,1,0,0,0)

```

	SUBROUTINE BLOCK(NX,LBLK,KRMN,KST,KEND,LONG)	TYPE0739
C	THIS ROUTINE COMPUTES NUMBERS USEFUL IN THE CONTROL OF THE INPUT	TYPE0740
C	AND OUTPUT OF LISTS OF LENGTH NX IN BLOCKS OF LENGTH LBLK.	TYPE0741
C	KRMN=NUMBER OF ITEMS REMAINING IN THE LIST	TYPE0742
C	LONG=LENGTH OF CURRENT BLOCK	TYPE0743
C	KST AND KEND ARE THE STARTING AND ENDING INDEXES FOR THE ITEMS IN	TYPE0744
C	THE CURRENT BLOCK.	TYPE0745
	IF (KRMN) 2, 1, 2	TYPE0746
1	KRMN=NX	TYPE0747
	KEND=0	TYPE0748
2	LONG=LBLK	TYPE0749
	KST=KEND+1	TYPE0750
	IF (KRMN-LBLK) 3, 4, 4	TYPE0751
3	LONG=KRMN	TYPE0752
4	KEND=KEND+LONG	TYPE0753
	KRMN=KRMN-LONG	TYPE0754
	RETURN	TYPE0755
	END	TYPE0756


```

SUBROUTINE MATINV(A,N,B,M,DETERM)                                TYPE0757
DIMENSION IPIVOT(126),A(126,126),B(126,1),INDEX(126,2),PIVOT(126) TYPE0758
EQUIVALENCE (IROW,JROW), (ICOLUM,JCOLUM), (AMAX, T, SWAP)      TYPE0759
C                                                                 TYPE0760
C     PROGRAMMED BY BURTON S. GARBOW, ARGONNE NATIONAL LABORATORY, TYPE0761
C AND REPORTED IN IBM 704-709 SHARE LIBRARY AS AN F402.        TYPE0762
C     THIS SUBROUTINE COMPUTES THE INVERSE AND DETERMINANT OF   TYPE0763
C MATRIX A, OF ORDER N, BY THE GAUSS-JORDAN METHOD. A-INVERSE  TYPE0764
C REPLACES A, AND THE DETERMINANT OF A IS PLACED IN DETERM. IF TYPE0765
C M = 1 THE VECTOR B CONTAINS THE CONSTANT VECTOR WHEN MATINV IS TYPE0766
C CALLED, AND THIS IS REPLACED WITH THE SOLUTION VECTOR. IF M = 0, TYPE0767
C NO SIMULTANEOUS EQUATION SOLUTIONS ARE CALLED FOR, AND B IS NOT TYPE0768
C PERTINENT. N IS NOT TO EXCEED 50.                             TYPE0769
C A, N, B, M, AND DETERM IN THE ARGUMENT LIST ARE DUMMY VARIABLES. TYPE0770
C                                                                 TYPE0771
C                                                                 TYPE0772
C     INITIALIZATION                                           TYPE0773
10 DETERM=1.0                                                    TYPE0774
15 DO 20 J=1,N                                                    TYPE0775
20 IPIVOT(J)=0                                                    TYPE0776
30 DO 550 I=1,N                                                    TYPE0777
C     SEARCH FOR PIVOT ELEMENT                                  TYPE0778
40 AMAX=0.0                                                       TYPE0779
45 DO 105 J=1,N                                                    TYPE0780
50 IF (IPIVOT(J)-1) 60, 105, 60                                   TYPE0781
60 DO 100 K=1,N                                                    TYPE0782
70 IF (IPIVOT(K)-1) 80, 100, 740                                  TYPE0783
80 IF (ABSF(AMAX)-ABSF(A(J,K))) 85, 100, 100                     TYPE0784
85 IROW=J                                                         TYPE0785
90 ICOLUM=K                                                        TYPE0786
95 AMAX=A(J,K)                                                    TYPE0787
100 CONTINUE                                                       TYPE0788
105 CONTINUE                                                       TYPE0789
110 IPIVOT(ICOLUM)=IPIVOT(ICOLUM)+1                              TYPE0790
C     INTERCHANGE ROWS TO PUT PIVOT ELEMENT ON DIAGONAL      TYPE0791
130 IF (IROW-ICOLUM) 140, 260, 140                               TYPE0792
140 DETERM=-DETERM                                                TYPE0793
150 DO 200 L=1,N                                                  TYPE0794
160 SWAP=A(IROW,L)                                                TYPE0795
170 A(IROW,L)=A(ICOLUM,L)                                         TYPE0796
200 A(ICOLUM,L)=SWAP                                             TYPE0797

```

205	IF(M) 260, 260, 210	TYPE0798
210	DO 250 L=1, M	TYPE0799
220	SWAP=B(IROW,L)	TYPE0800
230	B(IROW,L)=B(ICOLUM,L)	TYPE0801
250	B(ICOLUM,L)=SWAP	TYPE0802
260	INDEX(I,1)=IROW	TYPE0803
270	INDEX(I,2)=ICOLUM	TYPE0804
310	PIVOT(I)=A(ICOLUM,ICOLUM)	TYPE0805
320	DETERM=DETERM*PIVOT(I)	TYPE0806
C	DIVIDE PIVOT ROW BY PIVOT ELEMENT	TYPE0807
330	A(ICOLUM,ICOLUM)=1.0	TYPE0808
340	DO 350 L=1,N	TYPE0809
350	A(ICOLUM,L)=A(ICOLUM,L)/PIVOT(I)	TYPE0810
355	IF(M) 380, 380, 360	TYPE0811
360	DO 370 L=1,M	TYPE0812
370	B(ICOLUM,L)=B(ICOLUM,L)/PIVOT(I)	TYPE0813
C	REDUCE NON-PIVOT ROWS	TYPE0814
380	DO 550 L1=1,N	TYPE0815
390	IF(L1-ICOLUM) 400, 550, 400	TYPE0816
400	T=A(L1,ICOLUM)	TYPE0817
420	A(L1,ICOLUM)=0.0	TYPE0818
430	DO 450 L=1,N	TYPE0819
450	A(L1,L)=A(L1,L)-A(ICOLUM,L)*T	TYPE0820
455	IF(M) 550, 550, 460	TYPE0821
460	DO 500 L=1,M	TYPE0822
500	B(L1,L)=B(L1,L)-B(ICOLUM,L)*T	TYPE0823
550	CONTINUE	TYPE0824
C	INTERCHANGE COLUMNS	TYPE0825
600	DO 710 I=1,N	TYPE0826
610	L=N+1-I	TYPE0827
620	IF (INDEX(L,1)-INDEX(L,2)) 630, 710, 630	TYPE0828
630	JROW=INDEX(L,1)	TYPE0829
640	JCOLUM=INDEX(L,2)	TYPE0830
650	DO 705 K=1,N	TYPE0831
660	SWAP=A(K,JROW)	TYPE0832
670	A(K,JROW)=A(K,JCOLUM)	TYPE0833
700	A(K,JCOLUM)=SWAP	TYPE0834
705	CONTINUE	TYPE0835
710	CONTINUE	TYPE0836
740	RETURN	TYPE0837
	END	TYPE0838
	END(0,1,0,0,0)	TYPE0839

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1. ORIGINATING ACTIVITY (Corporate author) U. S. NAVAL PERSONNEL RESEARCH ACTIVITY SAN DIEGO, CALIFORNIA, 92152		2a. REPORT SECURITY CLASSIFICATION UNCLASSIFIED	
		2b. GROUP	
3. REPORT TITLE A COMPUTER PROGRAM FOR THE MAXIMUM LIKELIHOOD ANALYSIS OF TYPES			
4. DESCRIPTIVE NOTES (Type of report and inclusive dates) FINAL			
5. AUTHOR(S) (Last name, first name, initial) WOLFE, JOHN H.			
6. REPORT DATE MAY 1965	7a. TOTAL NO. OF PAGES 52	7b. NO. OF REFS NONE	
8a. CONTRACT OR GRANT NO.	9a. ORIGINATOR'S REPORT NUMBER(S) TECHNICAL BULLETIN 65-15		
b. PROJECT NO.			
c.	9b. OTHER REPORT NO(S) (Any other numbers that may be assigned this report)		
d.			
10. AVAILABILITY/LIMITATION NOTICES Release to CFSTI authorized			
11. SUPPLEMENTARY NOTES		12. SPONSORING MILITARY ACTIVITY CHIEF OF NAVAL PERSONNEL (Pers-A3) NAVY DEPARTMENT WASHINGTON, D. C., 20370	
13. ABSTRACT This report contains a description of a computer program for estimating the parameters of a mixture of multivariate normal distributions with unknown frequencies, means, and covariances. The basic equations for the procedure are presented for the first time here, with their derivation omitted. An example with the results of the computer printout is described for an artificially constructed mixture of three bivariate normal distributions. The method of using the program and the Fortran listing are detailed in this report.			

4 KEY WORDS	LINK A		LINK B		LINK C	
	ROLE	WT	ROLE	WT	ROLE	WT
Computer Techniques Maximum Likelihood Type Analysis Multivariate Analysis Cluster Analysis Multivariate Normal Distribution Mixtures						

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It is highly desirable that the abstract of classified reports be unclassified. Each paragraph of the abstract shall end with an indication of the military security classification of the information in the paragraph, represented as (TS), (S), (C), or (U).

There is no limitation on the length of the abstract. However, the suggested length is from 150 to 225 words.

14. **KEY WORDS:** Key words are technically meaningful terms or short phrases that characterize a report and may be used as index entries for cataloging the report. Key words must be selected so that no security classification is required. Identifiers, such as equipment model designation, trade name, military project code name, geographic location, may be used as key words but will be followed by an indication of technical context. The assignment of links, roles, and weights is optional.