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6 HIER-GRP:

A COMPUTER PROGRAM FOR THE HIERARCHICAL  
GROUPING OF REGRESSION EQUATIONS

By

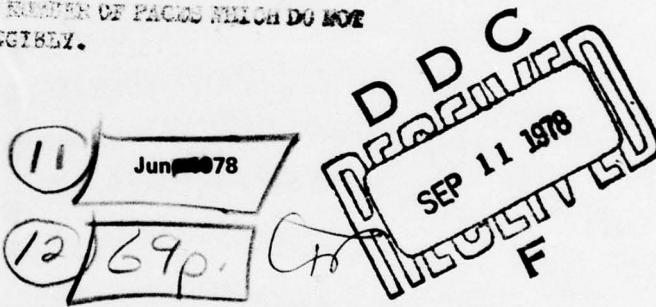
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COMPUTATIONAL SCIENCES DIVISION

Brooks Air Force Base, Texas 78235

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This final report was submitted by Computational Sciences Division, Air Force Human Resources Laboratory, Brooks Air Force Base, Texas 78235, under project 6323, with HQ Air Force Human Resources Laboratory (AFSC), Brooks Air Force Base, Texas 78235.

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This technical report has been reviewed and is approved for publication.

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20. ABSTRACT (Continue on reverse side if necessary and identify by block number) This report describes the technical details required for using the HIER-GRP computer program as it is currently operational on the Univac 1108 computer system at the Computational Sciences Division, Air Force Human Resources Laboratory, Brooks Air Force Base, Texas. HIER-GRP (or one of the earlier versions of the program) has been used extensively by the Air Force in the past, especially in conjunction with policy-capturing applications, and many of those applications are referenced herein. The report contains a discussion of the basic algorithm, an outline of the essential steps, specifications of the computer system requirements, descriptions of necessary control cards, and explanations of the program output. Also, appendices are included that contain the mathematical formulas used, some mathematical background helpful for understanding the algorithm, sample output, and a complete source card listing.		

## PREFACE

This research was completed under project 6323, Personnel Data Analyses; task 632305, Development of Analytic Methodology for Air Force Personnel Research Data.

In addition to the acknowledgments expressed in the introduction section of this report, the author wishes to give special credit to Mr. William S. Mathon for his numerous and valuable contributions to this project. Mathon performed the majority of the necessary programming tasks and prepared the basic text for Appendix B. Others who deserve mention include Mr. Larry K. Whitehead and Ms. Deana J. Olden for programming modifications and A1C Susan E. Tobey and Ms. Doris E. Black for technical editing. Finally, appreciation goes to Ms Dorothy M. Cobern and Ms. Laurel J. Betz for typing and proofreading the draft report.

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## HIER-GRP: A COMPUTER PROGRAM FOR THE HIERARCHICAL GROUPING OF REGRESSION EQUATIONS

### I. INTRODUCTION

HIER-GRP, an acronym for hierarchical grouping, is a computer program which was developed for various Air Force research purposes at the Computational Sciences Division, Air Force Human Resources Laboratory, Brooks AFB, Texas. Given a starting set of  $k$  regression equations, each of which contains the same criterion and predictor variables, the basic objective of the HIER-GRP algorithm is to group or to cluster the equations in a stepwise or iterative manner so as to minimize the overall loss of predictive efficiency at each iteration. Initially there are  $k$  separate groups; i.e., each of the  $k$  equations is considered as a group by itself, and a measure of overall predictive efficiency is computed. At the first iteration all possible ways of combining any two of the equations from the total  $k$  equations are examined, and that combination providing the minimum loss of overall predictive efficiency is selected to form a "new group." Formation of the new group reduces the number of equations to  $k-1$  for the start of the second iteration. The process continues until only one final group remains and is "hierarchical" in the sense that the pattern of the number of groups from start to finish is  $k, k-1, k-2, \dots, 1$ .

The mathematical theory upon which HIER-GRP is based is documented in an Air Force publication entitled *An Iterative Technique for Clustering Criteria Which Retains Optimum Predictive Efficiency* by Robert A. Bottenberg and Raymond E. Christal (3). Early developmental work was also accomplished by Joe H. Ward, Jr., (16), and some of the original programming was done by Daniel D. Rigney.

HIER-GRP or one of the earlier versions of the program has been used extensively by the Air Force in the past, especially in conjunction with "policy-capturing applications." Policy-capturing is a methodology composed of multiple linear regression analysis and hierarchical grouping procedures (1, 3, 4, 6, 7, 14, 16, 17, and 18). In this context, HIER-GRP was used in the development of the Weighted Airman Promotion System (WAPS) (10) and later in the reevaluation of WAPS (12 and 13). The program was also used in developing officer grade requirements (9), a promotion system for airman basics (2), a screening system for the Air Reserve Forces (8), and a senior NCO promotion system (11).

This report describes the technical details that are required for the use of the HIER-GRP program as it is currently operational on the Univac 1108 computer system at the Computational Sciences Division. The basic algorithm is first discussed, and the essential steps are outlined. Details of the computer system requirements and descriptions of necessary control cards are then presented. Next, the output of HIER-GRP is explained. Appendices are included that contain the mathematical formulas used in the program, some mathematical background helpful for understanding the algorithm, sample output, and a complete source card listing of the program.

Partly as a result of the research studies referenced above, requests for copies of the HIER-GRP computer program and associated documentation from different Air Force agencies, other governmental organizations, colleges, and universities have been numerous. Since 1969, approximately twenty copies of HIER-GRP have been provided to different requesters and implemented on a variety of different computer systems. One purpose of this report is to provide a document which can be used to satisfy any future requests for HIER-GRP.

### II. BASIC ALGORITHM

This section describes the basic structure of the HIER-GRP algorithm. The reader is referred to Appendix A for computational formulas mentioned in the various steps and to Appendix B for more detailed mathematical considerations.

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The basic steps of the HIER-GRP algorithm can be summarized as the following five phases: (a) data input and program termination, (b) computation of the overlap matrix, (c) determination of the order of clustering, (d) computation of the statistics for the initial k criteria, and (e) iteration to reduce the number of criteria. Each of these phases is described in the following steps. The steps are to be followed in numeric order unless indicated otherwise.

#### Steps 1-2. Data Input and Program Termination

1. Read "Problem Definition Card." This card defines  $k$ , the number of criteria or regression equations to be grouped and the number of standardized regression (beta) weights in each equation. If no Problem Definition Card is read, terminate the program.
2. Read in the number of cases, the criterion means and standard deviations, the standardized regression weights, the validities, and the predictor means and standard deviations for each equation. Assign each equation the identification numbers 1 through  $k$ , respectively, according to the order in which the equations were read.

#### Step 3. Computation of the Overlap Matrix

3. Compute the overlap matrix  $A$ , where each element  $a_{ij}$  denotes the decrease in overall predictive efficiency if equation  $i$  is combined with equation  $j$ ; for  $i = 1, 2, \dots, k, j = 1, 2, \dots, k$ , and  $i \neq j$ . The diagonal elements of  $A$  are undefined and the elements above the diagonal are symmetric with those elements below the diagonal.

#### Steps 4-8. Determination of the Order of Clustering

4. Set  $NGRPS$ , the index denoting the current number of groups, equal to  $k$ . Initially each criterion (equation) belongs to a separate cluster.
5. Considering all clusters present at the  $NGRPS$  stage, select two of the clusters denoted by  $i$  and  $j$  such that:
  - a.  $a_{ij} \leq a_{\ell m}$  where  $\ell$  and  $m$  are the identification numbers of any cluster present at the  $NGRPS$  stage and  $\ell \neq m$ , and
  - b.  $i < j$ . This can be accomplished by examining the elements above the diagonal of the overlap matrix and selecting the smallest element.
6. Form a new criterion cluster from the old clusters  $i$  and  $j$  identified in Step 5. Record the identifications of the two clusters  $i$  and  $j$  in the storage areas  $IU_{NGRPS}$  and  $JU_{NGRPS}$ , respectively. Assign the new cluster the identification number  $i$ .
7. Decrement  $NGRPS$  by 1. If  $NGRPS > 1$ , go to Step 8; otherwise proceed to Step 9.
8. Update the overlap matrix as follows. For each  $\ell$ ,  $\ell \neq i$  of Step 6 where  $\ell$  is the identification number of a criterion cluster present at the  $NGRPS$  stage, compute the decrease in overall predictive efficiency if equation  $\ell$  is combined with equation  $i$ . Since  $NGRPS$  was reduced by 1 in Step 7, the dimension of the updated overlap matrix will be reduced by 1. Return to Step 5.

#### Step 9. Computation of the Statistics for the Initial $k$ Criteria

9. Compute the squared multiple correlation coefficient for each of the initial  $k$  regression equations and, also,  $ORU_k$ , the overall squared multiple correlation coefficient obtained by considering a regression model with no grouping of initial equations.

### Steps 10-15. Iteration to Reduce the Number of Criteria

10. Form an initial grouping of the  $k$  equations by assigning each equation to a group by itself. This is the "k groups" stage. Set NGRPS equal to  $k$ .
11. Form a new grouping of the  $k$  equations by following the grouping order established in Steps 4-8. This is accomplished by combining the groups identified by  $IU_{NGRPS}$  and  $JU_{NGRPS}$  and assigning the new group (criterion cluster) the identification number in  $IU_{NGRPS}$ .
12. Compute the least squares regression equation which can be used to predict the new group and decrement NGRPS by 1.
13. Print all statistics concerning the new grouping including:
  - a. the identification numbers of the two equations combined at this iteration,
  - b. An F value testing the difference between the prediction equations for the two clusters in (a),
  - c. An F value testing the difference between the  $k$  initial prediction equations and the smaller set of NGRPS equations (one for each cluster) used at the "NGRPS groups" stage, and
  - d. the overall squared multiple correlation coefficient obtained using the NGRPS equations at this stage.
14. Print a summary of all groups (clusters) present at the NGRPS stage. Also, print the prediction equation for the new group (including standardized and raw score weights).
15. If  $NGRPS > 1$ , loop back to Step 11; otherwise, return to Step 1 and begin the next problem.

### **III. DESCRIPTIONS OF THE HIER-GRP PROGRAM**

#### Systems Requirements

The HIER-GRP program is composed of seven routines—the main or driver routine and six subroutines. The entire program, with the exception of the Univac Assembly Language subroutine START, is written in FORTRAN IV. The assembly subroutine START is called once at the beginning of the driver routine and is never called again. Its only function is to reset the margin control on the Univac 1108 printer.

The Univac version of FORTRAN has a special statement, the Parameter statement, which is used in the driver routine and which may not be available on other computers. The Parameter statement is used to define the dimensions of arrays at compilation time. The Parameter statement can be removed if each array is dimensioned to its required size.

The complete HIER-GRP program requires approximately 10,000 36-bit words of core storage in addition to the number of words required for arrays. If  $P$  is the number of predictors and  $E$  is the number of equations, then the amount of storage required for arrays is  $12E+3P+[2\cdot E\cdot P]+[E\cdot(E-1)/2]+14$ . For example, if  $P = 50$  and  $E = 50$ , then 6,989 words of storage are required for arrays.

There are a total of 1,121 cards in the HIER-GRP program deck. Of these, only 601 are source language cards and the remainder are comments cards. The number of cards and the intrinsic system routines required in each of the seven routines which make up HIER-GRP are listed in Table 1.

*Table 1. Characteristics of the HIER-GRP Routines*

Program Name	Source Language	Number of Source Language Cards	Number of Comment Cards	Intrinsic System Routines Required
DRIVER (MAIN)	FORTRAN IV	100	311	None
START	ASSEMBLY	7	0	None
OVRLP	FORTRAN IV	36	36	None
GROUP	FORTRAN IV	76	48	None
STAGE	FORTRAN IV	81	42	None
PRINTG	FORTRAN IV	218	82	SQRT
PLEVEL	FORTRAN IV	83	1	ATAN, SQRT, ALOG, EXP, SIN

#### Data Requirements

A HIER-GRP user must supply the following data for each regression equation:

1. The number of cases (N) which were used to compute the equation
2. The criterion mean and standard deviation (SD)
3. The standardized regression weights
4. The validity coefficients (correlations of predictor or independent variables with the criterion or dependent variable)
5. The predictor means and standard deviations.

The computational formulas developed by Bottenberg and Christal (3) and used within the program assume that the predictor sums-of-squares and cross-products matrices are proportional; i.e., that the ratios of the corresponding elements of the sums-of-squares and cross-products matrices for any two equations to be clustered are equal to the ratio of the corresponding numbers of the cases within each equation. This assumption of proportionality is discussed in detail by Bottenberg and Christal (1961, see pages 8 through 11) and also addressed in Appendix B (see equation 9b) of this report. In practice this assumption is met by selecting items (1) and (5) of the previous paragraph to be identical for each equation.

#### Run-Stream Organization

The following card sequence is required to use the HIER-GRP program as it is operational on a Univac 1108 computer:

Order	Card Type
1.	@RUN
2.	@XQT T*T.HIER-GRP
3.	Problem Definition Card
4.	Header Card(s)
5.	Format Card for Equation Ns
6.	Data Card(s) - Equation Ns
7.	Format Card for Criterion Means and SDs
8.	Data Card(s) - Criterion Means and SDs
9.	Format Card for Beta Weights
10.	Data Card(s) - Beta Weights
11.	Format Card for Validities

12. Data Card(s) - Validities
13. Format Card for Predictor Means and SDs
14. Data Card(s) - Predictor Means and SDs
15. The sequence of cards 3-14 is required for each run.  
As many problems as desired may be run by stacking one problem after another.
16. Blank Card to Terminate Run
17. @FIN

The Univac 1108 System Cards (1, 2, and 17) are described in the Univac Exec 8 Reference Manual (15). Descriptions of cards 3-16 are presented in the next section. See Appendix C for sample run-stream and sample control cards.

### Control Cards

#### Problem Definition Card

Card Columns	FORTRAN Format	Description
1-3	I3	NEQS, the number of criteria (systems, regression equations) in this problem. NEQS must be less than or equal to 50.
4-6	I3	NPREDS, the number of beta weights (standardized regression weights) in each equation. NPREDS must be less than or equal to 100.
7	I1	IOPT, the grouping (clustering) option desired. Normally a "6" is specified which causes the grouping to be done based on the iterative technique developed by Bottenberg and Christal (3). Other options are included in the program and comments cards, but are for future developmental purposes only.
8	I1	NHDRS, the number of header (label, title) cards that follow this control card. Header cards can be omitted (NHDRS = 0) or up to 9 cards may be specified.
9	I1	IREAD, the data read option. IREAD = 0 means read the beta weights and validities NPREDS items at a time. IREAD = 1 means read them NEQS*NPREDS items at a time. IREAD allows flexibility in the format of input data. However, IREAD is normally set equal to zero.
10-80		These card columns are not read.

#### Header Cards

Each header card will be printed only once at the beginning of the grouping report. Exactly NHDRS header cards must be present.

### **Format and Data Cards**

1. *Format Card for Equation Ns.* This card supplies the FORTRAN variable format by which the number of cases used in the computation of each equation is to be read. Only the F and X editing codes are permitted.
2. *Data Card(s) – Equation Ns.* These cards are read according to the previous format card. The number of cards required depends on the format specifications.
3. *Format Card for Criterion Means and SDs.* This card provides the FORTRAN variable format by which the criterion mean and standard deviation for each equation are to be read. Only the F and X editing codes are permitted.
4. *Data Card(s) – Criterion Means and SDs.* These cards are read according to the previous format card. The number of cards required depends on the format specifications.
5. *Format Card for Beta Weights.* This card supplies the FORTRAN variable format by which the beta weights (NPREDS weights per equation) are to be read. Only the F and X editing codes are permitted.
6. *Data Card(s) – Beta Weights.* These cards are read according to the previous format card. Exactly NEQS sets of cards are required if IREAD = 0. The first set contains the beta weights for equation 1; the second set contains the beta weights for equation 2; and so on. The number of cards within each set depends on the format specifications.
7. *Format Card for Validities.* This card provides the FORTRAN variable format by which the validity coefficients for each equation are read. Only the F and X editing codes are permitted.
8. *Data Card(s) – Validities.* These cards are read according to the previous format card. Exactly NEQS sets of cards are required if IREAD = 0. The first set contains the validities for equation 1; the second set contains the validities for equation 2; and so on. The number of cards within each set depends on the format specifications.
9. *Format Card for Predictor Means and SDs.* This card supplies the FORTRAN variable format by which the predictor means and standard deviations for each equation are to be read. Only the F and X editing codes are permitted.
10. *Data Card(s) – Predictor Means and SDs.* These cards are read according to the previous format card. The number of cards required depends on the format specifications.

### **Output**

The printed output of HIER-GRP is divided into five parts – the monogram and version date, the control card parameters, the problem header label, the format and input data cards, and the criterion grouping results. Each of these divisions is described in the following paragraphs. Refer to Appendix C for sample output.

#### **Monogram and Version Date**

The program title "Hierarchical Grouping Program HIER-GRP," the AFHRL monogram, and the program version date are printed at the beginning of each problem. The program version date is the last time the program was updated or modified.

#### **Control Card Parameters**

The parameters specified on the Problem Definition card are printed under the heading CONTROL CARD PARAMETERS. This includes the number of regression equations (criteria), the number of beta weights in each equation, the grouping option desired, and the number of header cards for this problem.

#### **Problem Header Label**

The problem header label, if header cards were specified on the Problem Definition Card, is printed under the heading PROBLEM HEADER LABEL.

#### **Format and Input Data Cards**

All format cards and all input data are printed under the heading FORMAT CARDS AND INPUT DATA. First, the format statements used to read the number of cases and the criterion means and standard deviations for each equation are printed. A table listing the equation numbers, the number of cases, the criterion means, and the criterion standard deviations is printed next. Third, the format statement used to read the beta weights and a table listing the equation number and the beta weights (15 per line) for each equation are printed. Fourth, the format statement used to read the validity coefficients, and a table listing the equation number and the validities (15 per line) for each equation are printed. Finally, the format statement used to read the predictor means and standard deviations and a table listing the predictor variable number and predictor means and standard deviations (one each per line) are printed.

#### **Criterion Grouping Results**

The results of the clustering process are printed under the heading HIERARCHICAL GROUPING RESULTS. The output in this division can be separated into three parts – the grouping option description, the R-square (RSQ) summary for the NEQS initial criteria, and the results of each iteration. Each of these sections is described as follows.

1. *Grouping Option Description.* The grouping option and a verbal description of the grouping option specified on the Problem Definition Card are printed.
2. *RSQ Summary for the NEQS Initial Criteria.* The number, NEQS, of initial criteria; the overall RSQ, ORU<sub>NEQS</sub>, achieved by using the beta weights specified on the input data cards; and a table listing the equation number and the RSQ for each equation are printed.
3. *Results of Each Iteration.* The statistics and tables printed at each iteration, i.e., the information printed below each row of asterisks is listed as the following in Table 2.

**Table 2. Output for Each Iteration**

Computer Output Label	Meaning
Stage = $\ell$	$\ell$ is the number of criterion clusters present at the end of this iteration.
OVERALL RSQ = ORU $_{\ell}$	This is the RSQ obtained by using $\ell$ equations (one for each criterion cluster present at this stage) to predict the NEQS initial criteria.
SYSTEMS GROUPING THIS STAGE Table	
SYS IDENT	The identification (ID) numbers of the two criterion clusters combined at this iteration.
NO. MEMBERS	The number of members in each criterion cluster. The ID numbers of the members of each cluster can be obtained by referring to the summary roster for stage $\ell+1$ .
NO. CASES	The number of cases used in the computation of the prediction equation for each criterion cluster. This number is the sum of the number of cases used in the prediction equation for each member of the cluster.
RSQ	The squared multiple correlation coefficient which is obtained by predicting each criterion within a cluster from the same compromise regression equation.
DECISION VALUE	The loss associated with replacement of the two clusters combined at this stage.
F-TEST FOR THE EQUALITY OF REGRESSION PARAMETERS FOR SYS'S COMBINED AT THIS STAGE Table	This table outlines a test of the hypothesis that the prediction equations for the two criterion clusters combined at this stage are identical. Equivalently, it is a test of the loss in predictive efficiency when each criterion within the two clusters combined at this stage are predicted from the same compromise equation.
CHANGE FROM $\ell+1$ SYSTEMS	
RSQ = ORU $_{\ell+1}$ - ORU $_{\ell}$	The decrease in OVERALL RSQ from stage $\ell+1$ .
DF = NPRED $S+1$	The decrease in the number of parameters estimated from stage $\ell+1$ .
RESIDUAL	
RSQ = 1 - ORU $_{\ell+1}$	The proportion of the criterion variance attributable to error at stage $\ell+1$ .
DF = N - ( $\ell+1$ ) (NPRED $S+1$ )	The total number of cases less the number parameters estimated at stage $\ell+1$ . Equivalently, DF is the number of degrees of freedom associated with the error portion of the criterion variance at stage $\ell+1$ .
FSTAT = $\frac{[(ORU_{\ell+1} - ORU_{\ell}) / (NPREDS+1)]}{[(1 - ORU_{\ell+1}) / (N - (\ell+1)(NPREDS+1))]}$	The F statistic testing the hypothesis described in the preceding paragraph (FOR SYS'S COMBINED AT THIS STAGE)

Table 2. (Continued)

Computer Output Label	Meaning
SIG LVL	The probability that a value of the F statistic greater than FSTAT would occur by chance. A value of SIG LVL equal to $\alpha$ means that if the hypothesis being tested is true, then a value of the F statistic greater than FSTAT would have occurred 100 $\alpha$ percent of the time by chance. Therefore, small values of $\alpha$ tend to reject the hypothesis being tested.
F-TEST FOR THE EQUALITY OF REGRESSION PARAMETERS FOR SYS'S COMBINED UP TO THIS STAGE Table	This table outlines a test of the hypothesis that the prediction equations for all members of criterion cluster number 1 are identical, the prediction equations for all members of criterion cluster 2 are identical, and so on for the $\ell$ criterion clusters present at the end of this iteration. Equivalently, this tests the loss in predictive efficiency when $\ell$ equations (one for each criterion cluster) are used to predict the NEQS initial criteria instead of the original NEQS equations.
CHANGE FROM NEQS SYSTEMS	
RSQ = ORU <sub>NEQS</sub> - ORU <sub><math>\ell</math></sub>	The decrease in OVERALL RSQ from stage NEQS.
DF = (NEQS - $\ell$ )(NPREDS+1)	The decrease in the number of parameters estimated from stage NEQS.
RESIDUAL	
RSQ = 1 - ORU <sub>NEQS</sub>	The proportion of the criterion variance attributable to error at stage NEQS.
DF = N - (NEQS)(NPREDS+1)	The total number of cases less the number of parameters estimated at stage NEQS. Equivalently, DF is the number of degrees of freedom associated with the error portion of the criterion variance at stage NEQS.
FSTAT = [(ORU <sub>NEQS</sub> - ORU <sub><math>\ell</math></sub> )/(NEQS - $\ell$ )(NPREDS+1)] / [(1 - ORU <sub>NEQS</sub> )/(N - (NEQS)(NPREDS+1))]	The F statistic testing the hypothesis described in the preceding paragraph (FOR SYS'S COMBINED UP TO THIS STAGE)
SIG LVL	The probability that a value of the F statistic greater than FSTAT would occur by chance. A value of SIG LVL equal to $\alpha$ means that if the hypothesis being tested is true, then a value of the F statistic greater than FSTAT would have occurred 100 $\alpha$ percent of the time by chance. Therefore, small values of $\alpha$ tend to reject the hypothesis being tested.
SYSTEMS SUMMARY ROSTER Table	The summary roster is a listing of all the criterion clusters present at the end of the current iteration. The members and the RSQ for each cluster are also printed. In addition, the prediction equation and the system mean and standard deviation for the new criterion cluster formed at the present iteration are printed. The compromise equation for each criterion cluster present at a given iteration can be obtained by referring to the summary roster for the stage at which the cluster was formed.

Table 2. (Continued)

Computer Output Label	Meaning
STAGE IDENT	The stage at which each criterion cluster was formed.
SYS LOSS	The contribution of each criterion cluster to the decrease in OVERALL RSQ from stage NEQS. Equivalently, this is the amount by which the OVERALL RSQ would increase if each of the criteria within this cluster were predicted from their individual regression equations rather than from the compromise equation for the cluster.
NO. MEMBERS	The number of criteria within each criterion cluster. The ID numbers of the members of each cluster are listed under the headings SYS IDENT and IDENTIFICATION OF OTHER MEMBERS in this table.
RSQ	The squared multiple correlation coefficient which is obtained by predicting each criterion within a cluster from the same compromise regression equation.
NO. CASES	The number of cases used in the computation of the compromise equation for a criterion cluster. This number is the sum of the number of cases used to compute the regression equation for each criterion within the cluster.
SYS IDENT	The ID number of a criterion cluster. This is also the smallest ID number of the criteria within this cluster.
IDENTIFICATION OF OTHER MEMBERS	The ID numbers of the remaining criteria within a cluster.
NEW SYS CRITERION MEAN	The criterion mean for the cluster formed at this iteration.
NEW SYS CRITERION SD	The criterion standard deviation for the cluster formed at this iteration.
BETA WEIGHTS FOR THE NEW SYSTEM S	The values (10 per line) of the least squares standardized regression coefficients for the regression equation which is the best single predictor for all the criteria in the new cluster where S is the ID number of the new cluster. Equivalently, these are the beta weights which would be obtained by pooling the observations for all the criteria in the new cluster and computing the regression of the pooled criteria on the NPREDs predictor variables.
RAW SCORE WEIGHTS FOR THE NEW SYSTEM S	The values (5 per line) of the raw score weights for the regression equation which is the best single predictor for all the criteria in the new cluster S.
REGRESSION CONSTANT	The regression constant for the regression equation which is the best single predictor of all the criteria in the new cluster.
Y SINGLE MEMBER SYSTEMS	A list of the identification numbers of the "Y" single criteria which have not been combined with any system up to this stage.

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## APPENDIX A: NOTATION AND COMPUTATIONAL FORMULAS

- The transpose of the associated matrix.
- $k$ , The initial number of criteria.
- $p$ , The number of variables.
- $n_i$ , The number of cases used in the computation of the regression equation for criterion  $i$ .
- $m_i$ , The mean for criterion  $i$ .
- $\sigma_i^2$ , The variance for criterion  $i$ .
- $\alpha_i$ , The constant term in the regression equation for criterion  $i$ .
- $b_i$ , The  $p \times 1$  vector of regression weights for criterion  $i$ .
- $\beta_i$ , The  $p \times 1$  vector of standard regression weights for criterion  $i$ .
- $c_i$ , The  $p \times 1$  vector of validities (intercorrelations between the criterion and the  $p$  independent variables) for criterion  $i$ .
- $N$ , The total number of cases  $N = n_1 + n_2 + \dots + n_k$
- $m_o$ , The pooled criterion mean  $Nm_o = n_1 m_1 + n_2 m_2 + \dots + n_k m_k$
- $\sigma_o^2$ , The pooled criterion variance  

$$N\sigma_o^2 = n_1(\sigma_1^2 + m_1^2) + \dots + n_k(\sigma_k^2 + m_k^2) - Nm_o^2$$
- $g_I$ , The number of criteria in cluster  $I$ .
- $I$ , The set of criteria in cluster  $I$ .  $I = \{i_1, i_2, \dots, i_{g_I}\}$ . In the succeeding definitions, let  $I$  be the union of clusters  $J$  and  $L$ ,  $J \cup L$ .
- $N_I$ , The number of cases used in the computation of the composite equation for cluster  $I$ .
- $$N_I = \sum_{i \in I} n_i = N_J + N_L$$
- $M_I$ , The criterion mean for cluster  $I$ .
- $$N_I M_I = \sum_{i \in I} n_i m_i = N_J M_J + N_L M_L$$
- $\sigma_I^2$ , The criterion variance for cluster  $I$ .
- $$N_I \sigma_I^2 = \sum_{i \in I} n_i (\sigma_i^2 + m_i^2) - N_I M_I^2 = N_J (\sigma_J^2 + M_J^2) + N_L (\sigma_L^2 + M_L^2) - N_I M_I^2$$
- $\hat{\alpha}_I$ , The constant term in the regression equation for cluster  $I$ .
- $$N_I \hat{\alpha}_I = \sum_{i \in I} n_i \hat{\alpha}_i = N_J \hat{\alpha}_J + N_L \hat{\alpha}_L$$
- $\hat{b}_I$ , The  $p \times 1$  vector of regression weights for cluster  $I$ .
- $$N_I \hat{b}_I = \sum_{i \in I} n_i \hat{b}_i = N_J \hat{b}_J + N_L \hat{b}_L$$

$\hat{\beta}_I$ , The pxi vector of standard regression weights for cluster I.

$$N_I \sigma_I \beta_I = \sum_{i \in I} n_i \sigma_i \hat{\beta}_i = N_J \sigma_J \beta_J + N_L \sigma_L \hat{\beta}_L$$

$c_I$ , The pxi vector of validities for cluster I.

$$N_I \sigma_I c_I = \sum_{i \in I} n_i \sigma_i c_i = N_J \sigma_J c_J + N_L \sigma_L c_L$$

$R_i^2$ , The squared multiple correlation coefficient for the regression on criterion i.

$$R_i^2 = \hat{\beta}_i' c_i$$

$R_I^2$ , The squared multiple correlation coefficient for the regression on cluster I.

$$R_I^2 = \hat{\beta}_I' c_I = \frac{1}{N_I^2 \sigma_I^2} \left[ N_J^2 \sigma_J^2 R_J^2 + N_L^2 \sigma_L^2 R_L^2 + N_J N_L \sigma_J \sigma_L (\hat{\beta}_J' c_L + \hat{\beta}_L' c_J) \right]$$

$G_s$ , The set of s criterion clusters present at the s cluster stage.

$$G_s = \{ I_1, I_2, \dots, I_s \}$$

$s^2 R^2$ , The squared multiple correlation coefficient for the criterion grouping,  $G_s$ , at the s cluster stage.

$$N \sigma_0^2 s^2 R^2 = \sum_{I \in G_s} N_I (\sigma_I^2 R_I^2 + M_I^2) - N m_0^2$$

Let  $G_s = \{ J, L, K_3, \dots, K_s \}$  and

$G_{s-1} = \{ J \cup L, K_3, \dots, K_s \}$  then

$$s^2 R^2 - s-1^2 R^2 = \frac{N_J N_L}{N \sigma_0^2 (N_J + N_L)} \left[ \sigma_J^2 R_J^2 + \sigma_L^2 R_L^2 + (M_J - M_L)^2 - \sigma_J \sigma_L (\hat{\beta}_J' c_L + \hat{\beta}_L' c_J) \right]$$

## APPENDIX B: MATHEMATICAL BACKGROUND

### Mathematical Model for the Clustering Algorithm

Suppose that a set of  $p$  independent variables,  $v' = (v_1, \dots, v_p)$ , are linearly related to the expected values of each of  $k$  criteria,  $Y_1, \dots, Y_k$ ; that is,

$$(1) \quad E(Y_i|v) = v'b_i + \alpha_i \quad \text{for } i=1, \dots, k$$

where  $b_i$  is a  $px1$  vector of unknown population parameters and  $\alpha_i$  is an unknown population constant. Let  $y_i$  be an  $n_i \times 1$  vector of independent observations on criterion  $Y_i$ , let  $X_i$  be an  $n_i \times p$  matrix of observations on the set of  $p$  independent variables  $v$ , where the  $j$ -th element of  $y_i$  corresponds to the  $j$ -th row of  $X_i$ , and let  $u_i$  be an  $n_i \times 1$  vector in which each element is 1. Then from (1),

$$(1a) \quad E(y_i|X_i) = X_i b_i + u_i \alpha_i \quad \text{for } i=1, \dots, k.$$

Let  $N = n_1 + \dots + n_k$ ; let  $Y = [y_1, \dots, y_k]$ , the  $1 \times N$  vector obtained by pooling all the criterion observations; let

$$X = \begin{bmatrix} u_1 X_1 & 0 & 0 & \cdot & \cdot & \cdot & 0 \\ 0 & u_2 X_2 & 0 & \cdot & \cdot & \cdot & 0 \\ 0 & \cdot & \cdot & \cdot & \cdot & \cdot & u_k X_k \end{bmatrix}$$

the  $N \times k(p+1)$  block diagonal matrix obtained by placing the  $n_i \times (p+1)$  matrix of observations  $[u_i X_i]$  in the  $i$ -th block diagonal position, and let  $b' = [\alpha_1 b'_1 \dots \alpha_k b'_k]$ , the  $k(p+1)$  vector of unknown parameters. Under the assumption that the criterion observations are independent and have common variance, the mathematical model for the clustering algorithm is

$$(1b) \quad E(Y|X) = Xb \text{ with } D(Y|X) = \sigma^2 I,$$

where  $D(Y|X)$  is the dispersion matrix of the criterion observations,  $\sigma^2$  is the common variance, and  $I$  is the  $N \times N$  identity matrix.

### Minimum Variance Unbiased Estimation and Hypothesis Testing

The  $k(p+1) \times 1$  vector  $b$  of unknown parameters in (1b) correspond to the  $k$  equations in (1a). The minimum variance unbiased estimates (mvue),  $\hat{\alpha}_i$  and  $\hat{b}_i$ , of  $\alpha_i$  and  $b_i$  are obtained from (1b) by the method of least squares, where

$$(2) \quad \begin{aligned} \hat{b}_i &= [X_i' X_i - \frac{1}{n_i} X_i' u_i u_i' X_i]^{-1} [X_i' y_i - \frac{1}{n_i} X_i' u_i u_i' y_i] \quad \text{for } i=1, \dots, k. \\ \hat{\alpha}_i &= \frac{1}{n_i} u_i' y_i - \frac{1}{n_i} u_i' X_i \hat{b}_i \end{aligned}$$

These are the estimates that would be obtained by the method of least squares from the  $k$  separate models

$$(3) \quad E(y_i|X_i) = X_i b_i + u_i \alpha_i \text{ with } D(y_i|X_i) = \sigma^2 I \quad \text{for } i=1, \dots, k$$

where the error variance,  $\sigma^2$ , is the same for each model. It might be that some or all of the equations in (1) are identical. The technique of homogeneity of regression can be used to-test the equality of vectors of regression parameters across several criteria. Chipman and Rao (1964) and Theil (1970) have developed methods for obtaining mvue under general linear restrictions and for testing general linear hypotheses. Rao (1965, pp 189–190) shows that in the case

$$(4) \quad E(Y|X) = Xb \text{ with } D(Y|X) = \sigma^2 I,$$

where  $X$  is  $n \times s$  of rank  $s$  and  $b$  is  $s \times 1$ , the mvue,  $\hat{b}_{\Psi}$  for  $b$  under the linear restriction

$$(4a) \quad \Psi b = 0 \text{ is}$$

$$(4b) \quad \hat{b}_{\Psi} = B(B' X' X B)^{-1} B' X' Y$$

where  $\Psi$  is  $r \times s$  of rank  $r$ ,  $B$  is  $s \times (s-r)$  of rank  $(s-r)$ , and  $\Psi B = 0$ . Rao obtains this result by introducing the general solution,  $B\theta$ , where  $\theta$  is an  $(s-r) \times 1$  vector of new parameters, of (4a) into (4) to obtain the model

$$(5) \quad E(Y|X) = XB\theta \text{ with } D(Y|X) = \sigma^2 I$$

and no restrictions on  $\theta$ . The mvue,  $\hat{B}\theta$ , of  $B\theta$  is  $B\hat{\theta}$  (see Rao, 1965, pp. 181–182), where  $\hat{\theta}$  is the mvue of  $\theta$  in (5). If, in addition to (4),  $Y$  has a multivariate normal distribution, then Chipman and Rao develop an expression for an unbiased critical region of size  $\theta$  for the following hypothesis:

$$(6) \quad \Psi_1 b = 0 \text{ given that } \Psi_0 b = 0$$

where  $\Psi_1$  is  $r_1 \times s$  of rank  $r_1$ ,  $\Psi_0$  is  $r_0 \times s$  of rank  $r_0$ , and  $\Psi' = [\Psi_0' \Psi_1']'$  is  $s \times (r_0 + r_1)$  of rank  $(r_0 + r_1)$ . The expression for the unbiased critical region of size  $\theta$  is

$$(7) \quad \{F | F = \left( \frac{n-s+r_0}{r_1} \right) \left( \frac{EXSS}{ESSH} \right) = \left( \frac{n-s+r_0}{r_1} \right) \frac{\left( R_{\Psi_0}^2 - R_{\Psi}^2 \right)}{1 - R_{\Psi_0}^2} > F_{\theta} (r_1, n-s+r_0) \},$$

where  $F_{\theta} (r_1, n-s+r_0)$  is the upper 100  $(1-\theta)\%$  point of the central F distribution with  $r_1$  and  $n-s+r_0$  degrees of freedom, and

$$ESSH = (Y - X\hat{b}_{\Psi_0})'(Y - X\hat{b}_{\Psi_0}),$$

$$EXSS = (Y - X\hat{b}_{\Psi})'(Y - X\hat{b}_{\Psi}) - ESSH,$$

$\hat{b}_{\Psi_0}$  is the mvue of  $b$  under the restriction  $\Psi_0 b = 0$ ,

$\hat{b}_{\Psi}$  is the mvue of  $b$  under the restriction  $\Psi b = 0$ ,

$R_{\Psi_0}^2$  is the squared multiple correlation under the restriction

$\Psi_0 b = 0$ , and

$R_{\Psi}^2$  is the squared multiple correlation under the restriction

$\Psi b = 0$ .

The Chipman and Rao computational form for  $F$  is different from the form in (7), but the two are equivalent. (See Rao, 1965, pp. 199–200).

#### MVUE for a Criterion Cluster

The restriction  $\alpha_1 = \alpha_2 = \dots = \alpha_t$  and  $b_1 = b_2 = \dots = b_t$  can be expressed in the form  $\Psi b = 0$  as

$$(8) \quad (t-1)(p+1) \left\{ \begin{bmatrix} I & -I & 0 & \dots & & \\ . & -I & 0 & \dots & & \\ . & . & . & & 0 & \\ . & 0 & & & & \\ & & & 0 & & \\ I & 0 & \dots & 0 & -I & 0 \dots 0 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ b_1 \\ \vdots \\ \alpha_k \\ b_k \end{bmatrix} \right\} = 0$$

$t(p+1) \qquad (k-t)(p+1)$

where  $I$  is the  $(p+1) \times (p+1)$  identity matrix. To express model (1b) in a form similar to equation (5) under the above restriction (8), the  $k(p+1) \times (k-t+1)(p+1)$  matrix  $B$ , where

$$B' = \left[ \begin{array}{cc} t(p+1) & (k-t)(p+1) \\ \underbrace{\begin{matrix} I & \dots & I \\ 0 & \dots & 0 \end{matrix}}_{\text{t rows}} & \underbrace{\begin{matrix} 0 & \dots & 0 \\ I & & \\ & \ddots & 0 \\ & & I \end{matrix}}_{(k-t+1)(p+1)} \end{array} \right]$$

and the  $(k-t+1)(p+1)$  vector of new parameters  $\theta$ , where

$$\theta' = [\alpha_\Psi \ b_\Psi \ \alpha_{t+1} \ b_{t+1} \ \dots \ \alpha_k \ b_k],$$

is introduced into (1b) to yield the model

$$(9) \quad E(Y|X) = \begin{bmatrix} u_1 X_1 \\ \vdots \\ u_t X_t \\ u_{t+1} X_{t+1} \\ \vdots \\ 0 \\ u_k X_k \end{bmatrix} \quad 0 \quad \begin{bmatrix} \alpha_\Psi \\ b_\Psi \\ \alpha_{t+1} \\ b_{t+1} \\ \vdots \\ \alpha_k \\ b_k \end{bmatrix}$$

with  $D(Y|X) = \sigma^2 I$ .

The effect of  $B$  is to pool the observations for criteria 1, ...,  $t$ . The mvue  $\hat{\alpha}_\Psi$  and  $\hat{b}_\Psi$ , for the criterion cluster (1, 2, ...,  $t$ ) formed from criteria 1, ...,  $t$  can be calculated in either of two ways: pool the observations as in (9) and compute  $\hat{\alpha}_\Psi$  and  $\hat{b}_\Psi$  from the normal equations

$$(9a) \quad \left\{ \begin{bmatrix} n_1 u_1' X_1 \\ X_1' u_1 X_1' X_1 \end{bmatrix} + \dots + \begin{bmatrix} n_t u_t' X_t \\ X_t' u_t X_t' X_t \end{bmatrix} \right\} \begin{bmatrix} \hat{\alpha}_\Psi \\ \hat{b}_\Psi \end{bmatrix} = \left\{ \begin{bmatrix} u_1' y_1 \\ X_1' y_1 \end{bmatrix} + \dots + \begin{bmatrix} u_t' y_t \\ X_t' y_t \end{bmatrix} \right\}$$

or if the predictor sums-of-squares and cross-product matrices are proportional, i.e.,

$$(9b) \quad \frac{1}{n_1} \begin{bmatrix} n_1 u_1' X_1 \\ X_1' u_1 X_1' X_1 \end{bmatrix} = \frac{1}{n_2} \begin{bmatrix} n_2 u_2' X_2 \\ X_2' u_2 X_2' X_2 \end{bmatrix} = \dots = \frac{1}{n_t} \begin{bmatrix} n_t u_t' X_t \\ X_t' u_t X_t' X_t \end{bmatrix},$$

then  $\hat{\alpha}_\Psi$  and  $\hat{b}_\Psi$  can be calculated from  $\hat{\alpha}_1, \hat{b}_1, \dots, \hat{\alpha}_t$ , and  $\hat{b}_t$  given in (2) without forming the sum of matrices on the left hand side in (9a). Using (9b) this sum of matrices is

$$(9c) \quad \left\{ \begin{bmatrix} n_1 u_1' X_1 \\ X_1' u_1 X_1' X_1 \end{bmatrix} + \dots + \begin{bmatrix} n_t u_t' X_t \\ X_t' u_t X_t' X_t \end{bmatrix} \right\} = \frac{N_t}{n_i} \begin{bmatrix} n_i u_i' X_i \\ X_i' u_i X_i' X_i \end{bmatrix} \quad \text{for } i = 1, \dots, t$$

where  $N_t = n_1 + n_2 + \dots + n_t$ . Using (9c) the solution of (9a) is

$$\begin{aligned} \begin{bmatrix} \hat{\alpha}_\Psi \\ \hat{b}_\Psi \end{bmatrix} &= \sum_{i=1}^t \left( \begin{bmatrix} n_i u'_i X_i \\ X'_i u_i X'_i \end{bmatrix} + \dots + \begin{bmatrix} n_t u'_t X_t \\ X'_t u_t X'_t \end{bmatrix} \right)^{-1} \begin{bmatrix} u'_i y_i \\ X'_i y_i \end{bmatrix} \\ &= \sum_{i=1}^t \frac{n_i}{N_t} \begin{bmatrix} n_i u'_i X_i \\ X'_i u_i X'_i \end{bmatrix}^{-1} \begin{bmatrix} u'_i y_i \\ X'_i y_i \end{bmatrix} \end{aligned}$$

Thus, the mvue for a criterion cluster are

$$(10) \quad \begin{bmatrix} \hat{\alpha}_\Psi \\ \hat{b}_\Psi \end{bmatrix} = \frac{n_1}{N_t} \begin{bmatrix} \hat{\alpha}_1 \\ \hat{b}_1 \end{bmatrix} + \dots + \frac{n_t}{N_t} \begin{bmatrix} \hat{\alpha}_t \\ \hat{b}_t \end{bmatrix}$$

When (9b) holds, the formula for the standardized regression weights for a criterion cluster is easy to obtain. Let  $\hat{\beta}_\Psi, \hat{\beta}_1, \dots, \hat{\beta}_t$  be the standardized weights corresponding to the raw weights  $\hat{b}_\Psi, \hat{b}_1, \dots, \hat{b}_t$ ; let  $Q_i$  be the pxp diagonal matrix with its elements equal to the standard deviations calculated from the observation matrix  $X_i$  for the p independent variables; let  $Q_\Psi$  be the pxp diagonal matrix with its elements equal to the standard deviations calculated from the pooled observation matrix  $[X'_1 X'_2 \dots X'_t]'$  for the p independent variables; and let  $\sigma_\Psi^2, \sigma_1^2, \dots, \sigma_t^2$  be the sample variances for the vectors of criterion observations  $[y'_1 y'_2 \dots y'_t]', y_1, \dots, y_t$ , respectively. By definition the standardized weights are

$$\hat{\beta}_\Psi = \frac{Q_\Psi \hat{b}_\Psi}{\sigma_\Psi}, \hat{\beta}_1 = \frac{Q_1 \hat{b}_1}{\sigma_1}, \dots, \hat{\beta}_t = \frac{Q_t \hat{b}_t}{\sigma_t}.$$

From (9b),  $Q_\Psi = Q_1 = \dots = Q_t$ ; therefore using (10), the formula for the standardized weights for a criterion cluster is

$$(10a) \quad \hat{\beta}_\Psi = \frac{1}{N_t \sigma_\Psi} (n_1 \sigma_1 \hat{\beta}_1 + \dots + n_t \sigma_t \hat{\beta}_t).$$

#### Multiple Correlation Coefficient for a Criterion Cluster

Let  $R_\Psi^2, R_1^2, \dots, R_t^2$  be the squared multiple correlation coefficients for the criterion cluster formed from criteria 1, ..., t and for the t criteria  $y_1, \dots, y_t$ , respectively; let  $c_i$  be the pxl vector of intercorrelations calculated from the observations  $X_i$  and  $y_i$  between the p independent variables and the i-th criterion; and let  $c_\Psi$  be the pxl vector of intercorrelations calculated from the pooled observations  $[X'_1 X'_2 \dots X'_t]'$  and  $[y'_1 y'_2 \dots y'_t]'$  between the p independent variables and the criterion cluster (1, 2, ..., t). By definition,

$$n_i \sigma_i Q_i c_i = X'_i y_i - \frac{1}{n_i} X'_i u_i u'_i y_i \quad \text{for } i=1, \dots, k \text{ and}$$

$$N_t \sigma_\Psi Q_\Psi c_\Psi = (X'_1 y_1 + \dots + X'_t y_t) - \frac{1}{N_t} [X'_1 u_1 + \dots + X'_t u_t] [u'_1 y_1 + \dots + u'_t y_t].$$

From (9c),  $\frac{1}{N_t} |X'_1 u_1 + \dots + X'_t u_t| = \frac{1}{n_i} X'_i u_i$  for  $i=1, \dots, t$ . Therefore,

$$N_t \sigma_\Psi Q_\Psi c_\Psi = n_1 \sigma_1 Q_1 c_1 + \dots + n_t \sigma_t Q_t c_t.$$

But  $Q_\Psi = Q_1 = \dots = Q_t$  so the validity coefficients for a criterion cluster are

$$(10b) c_{\Psi} = \frac{1}{N_t \sigma_{\Psi}} (n_1 \sigma_1 c_1 + \dots + n_t \sigma_t c_t).$$

The squared multiple correlation coefficient for the cluster

(1, 2, ..., t) is

$$(10c) R_{\Psi}^2 = \hat{\beta}_{\Psi} c_{\Psi} = \frac{1}{N_t^2 \sigma_{\Psi}^2} (n_1 \sigma_1 \hat{\beta}_1 + \dots + n_t \sigma_t \hat{\beta}_t)' (n_1 \sigma_1 c_1 + \dots + n_t \sigma_t c_t).$$

#### Hypothesis Testing

The critical region given in (7) for the hypothesis (6) requires the calculation of the error sum of squares or the squared multiple correlation coefficient for model (1b) when restrictions are imposed on the unknown parameters. The error sum of squares, ESS, for model (1b) when there are no restrictions on the unknown parameters is equal to the sum of the error sum of squares, ESS<sub>i</sub>, for the k models (see (3)), i.e.,

$$ESS = ESS_1 + ESS_2 + \dots + ESS_k.$$

Let m<sub>0</sub> and σ<sub>0</sub><sup>2</sup> be the criterion mean and variance calculated from the pooled criterion observation vector Y, and let m<sub>1</sub>, ..., m<sub>k</sub> be the criterion means for y<sub>1</sub>, ..., y<sub>k</sub>, respectively. Then

$$ESS_i = n_i \sigma_i^2 (1 - R_i^2) \quad \text{for } i=1, \dots, k$$

$$Nm_0 = n_1 m_1 + n_2 m_2 + \dots + n_k m_k$$

$$N\sigma_0^2 = n_1 (\sigma_1^2 + m_1^2) + \dots + n_k (\sigma_k^2 + m_k^2) - Nm_0^2$$

Therefore the squared multiple correlation, R<sup>2</sup>, for (1b) is

$$(11) \quad R^2 = \frac{N\sigma_0^2 - ESS}{N\sigma_0^2} = \frac{\left[ n_1 (\sigma_1^2 R^2 + m_1^2) + \dots + n_k (\sigma_k^2 R^2 + m_k^2) - Nm_0^2 \right]}{\left[ n_1 (\sigma_1^2 + m_1^2) + \dots + n_k (\sigma_k^2 + m_k^2) - Nm_0^2 \right]}$$

The error sum of squares, ESSH, for (9) is

$$ESSH = ESS_{\Psi} + ESS_{t+1} + \dots + ESS_k$$

where ESS<sub>\Psi</sub> = N<sub>t</sub> σ<sub>\Psi</sub><sup>2</sup> (1 - R<sub>\Psi</sub><sup>2</sup>). Therefore the squared multiple correlation, R<sub>0</sub><sup>2</sup>, for (9) is

$$\frac{R^2}{N\sigma_0^2} - ESSH$$

The hypothesis (8) can be tested at the α significance level by computing

$$(11a) \quad F = \left( \frac{N-k(p+1)}{(t-1)(p+1)} \right) \left( \frac{R^2 - R_0^2}{1 - R^2} \right)$$

and rejecting (8) if F exceeds the 100(1-α)% point of the central F distribution with (t-1)(p+1) and N - k(p+1) degrees of freedom.

#### Application to a Four Criteria Model; A Worked Example

Given four criteria y<sub>1</sub>, y<sub>2</sub>, y<sub>3</sub>, and y<sub>4</sub>, where y<sub>i</sub> is an n<sub>i</sub> × 1 vector of observations, and the predictor matrices X<sub>1</sub>, X<sub>2</sub>, X<sub>3</sub>, and X<sub>4</sub>, where X<sub>i</sub> is an n<sub>i</sub> × p matrix of observations on p independent variables, the

greatest predictive power is attained when each criterion variable is predicted from its regression on the independent variables. The initial stage, i.e., Stage 4, employs the following model:

$$(12) E \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} = \begin{bmatrix} u_1 X_1 & 0 & 0 & 0 \\ 0 & u_2 X_2 & 0 & 0 \\ 0 & 0 & u_3 X_3 & 0 \\ 0 & 0 & 0 & u_4 X_4 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ b_1 \\ \alpha_2 \\ b_2 \\ \alpha_3 \\ b_3 \\ \alpha_4 \\ b_4 \end{bmatrix} = \begin{bmatrix} \alpha_1 u_1 + b_1 X_1 \\ \alpha_2 u_2 + b_2 X_2 \\ \alpha_3 u_3 + b_3 X_3 \\ \alpha_4 u_4 + b_4 X_4 \end{bmatrix} \text{ with } D \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} = \sigma^2 I .$$

$4 \times 4 \quad 4 \times 1 \quad 4 \times 1$

The mvue  $\hat{\alpha}_i$  and  $\hat{b}_i$ , for  $\alpha_i$  and  $b_i$  are obtained from (2) and the squared multiple correlation coefficient,  $R^2$ , for model (12) is obtained from (11).

For Stage 3, assume (9b) holds for  $X_1$ ,  $X_2$ ,  $X_3$ , and  $X_4$ . Under the linear hypothesis  $\alpha_1 = \alpha_2$  and  $b_1 = b_2$ , the mvue  $\hat{\alpha}_{12}$  and  $\hat{b}_{12}$ , for the criterion cluster (1,2) formed from criteria 1 and 2 are (see (10))

$$\begin{bmatrix} \hat{\alpha}_{12} \\ \hat{b}_{12} \end{bmatrix} = \frac{n_1}{(n_1+n_2)} \begin{bmatrix} \hat{\alpha}_1 \\ \hat{b}_1 \end{bmatrix} + \frac{n_2}{n_1+n_2} \begin{bmatrix} \hat{\alpha}_2 \\ \hat{b}_2 \end{bmatrix}$$

The standard weights,  $\hat{\beta}_{12}$ , and the validities,  $c_{12}$ , for the cluster (1,2) are (see (10a) and (10b))

$$\begin{aligned} \hat{\beta}_{12} &= \frac{1}{(n_1+n_2)\sigma_{12}} (n_1\sigma_1\hat{\beta}_1 + n_2\sigma_2\hat{\beta}_2), \text{ and} \\ c_{12} &= \frac{1}{(n_1+n_2)\sigma_{12}} (n_1\sigma_1 c_1 + n_2\sigma_2 c_2), \text{ where} \\ (n_1+n_2)\sigma_{12}^2 &= n_1(\sigma_1^2 + m_1^2) + n_2(\sigma_2^2 + m_2^2) - \frac{(n_1m_1 + n_2m_2)^2}{n_1 + n_2}. \end{aligned}$$

The model used to obtain these estimates is (see (9))

$$(13) E \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} = \begin{bmatrix} u_1 X_1 & 0 & 0 & 0 \\ u_2 X_2 & u_3 X_3 & 0 & 0 \\ 0 & 0 & u_4 X_4 & 0 \end{bmatrix} \begin{bmatrix} \alpha_{12} \\ b_{12} \\ \alpha_3 \\ b_3 \\ \alpha_4 \\ b_4 \end{bmatrix} = \begin{bmatrix} \alpha_{12} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} + b_{12} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} \\ \alpha_3 u_3 + b_3 X_3 \\ \alpha_4 u_4 + b_4 X_4 \end{bmatrix} \text{ with } D \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} = \sigma^2 I .$$

The squared multiple correlation coefficient,  $R^2$ , for (13) is (from (11) with  $k=3$ )

$$\begin{aligned} {}_3 R^2 &= \frac{[(n_1+n_2)(\sigma_{12}^2 R_{12}^2 + m_{12}^2) + n_3(\sigma_3^2 R_3^2 + m_3^2) + n_4(\sigma_4^2 R_4^2 + m_4^2) - Nm_0^2]}{[(n_1+n_2)(\sigma_{12}^2 + m_{12}^2) + n_3(\sigma_3^2 + m_3^2) + n_4(\sigma_4^2 + m_4^2) - Nm_0^2]}, \text{ where} \\ R_{12}^2 &= \hat{\beta}_{12} c_{12}, \quad m_{12} = \frac{(n_1m_1 + n_2m_2)}{(n_1+n_2)}, \quad N = n_1 + n_2 + n_3 + n_4, \text{ and} \end{aligned}$$

$$Nm_0 = n_1m_1 + n_2m_2 + n_3m_3 + n_4m_4 .$$

(11a) can now be used to test at the  $\alpha$  significance level the hypothesis  $H_1: \alpha_1 = \alpha_2$  and  $b_1 = b_2$  by computing

$$F = \left( \frac{N-4(p+1)}{(p+1)} \right) \left( \frac{\frac{4}{4} R^2 - \frac{3}{4} R^2}{\frac{1}{4}(1-\frac{3}{4} R^2)} \right)$$

and rejecting  $H_1$  if  $F$  exceeds  $F_{\alpha}(p+1, N-4(p+1))$ .

For Stage 2, accepting  $H_1$  as true, the additional restrictions  $\alpha_3 = \alpha_4$  and  $b_3 = b_4$  are imposed and the mvue,  $\hat{\alpha}_{34}$  and  $\hat{b}_{34}$ , for the criterion cluster (3,4) formed from criteria 3 and 4 are

$$\begin{bmatrix} \hat{\alpha}_{34} \\ \hat{b}_{34} \end{bmatrix} = \frac{n_3}{n_3 + n_4} \begin{bmatrix} \hat{\alpha}_3 \\ \hat{b}_3 \end{bmatrix} + \frac{n_4}{n_3 + n_4} \begin{bmatrix} \hat{\alpha}_4 \\ \hat{b}_4 \end{bmatrix}$$

The standard weights,  $\hat{\beta}_{34}$ , and the validities,  $c_{34}$ , for the cluster (3,4) are

$$\hat{\beta}_{34} = \frac{1}{(n_3 + n_4)\sigma_{34}} (n_3 \sigma_3 \hat{\beta}_3 + n_4 \sigma_4 \hat{\beta}_4), \text{ and}$$

$$c_{34} = \frac{1}{(n_3 + n_4)\sigma_{34}} (n_3 \sigma_3 c_3 + n_4 \sigma_4 c_4), \text{ where}$$

$$(n_3 + n_4)\sigma_{34}^2 = n_3(\sigma_3^2 + m_3^2) + n_4(\sigma_4^2 + m_4^2) - \frac{(n_3 m_3 + n_4 m_4)^2}{(n_3 + n_4)}$$

The model used to obtain these estimates is

$$(14) E \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} = \begin{bmatrix} u_1 X_1 & & 0 \\ u_2 X_2 & & \\ 0 & & u_3 X_3 \\ & & u_4 X_4 \end{bmatrix} \begin{bmatrix} \alpha_{12} \\ b_{12} \\ \alpha_{34} \\ b_{34} \end{bmatrix} = \begin{bmatrix} \alpha_{12} & \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} + b_{12} & \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} \\ \alpha_{34} & \begin{bmatrix} u_3 \\ u_4 \end{bmatrix} + b_{34} & \begin{bmatrix} X_3 \\ X_4 \end{bmatrix} \end{bmatrix} \text{ with } D \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} = \sigma^2 I.$$

The squared multiple correlation coefficient,  $R^2$ , for (14) is (from (11) with  $k=2$ )

$$R^2 = \frac{[(n_1 + n_2)(\sigma_{12}^2 R_{12}^2 + m_{12}^2) + (n_3 + n_4)(\sigma_{34}^2 R_{34}^2 + m_{34}^2) - Nm_0^2]}{[(n_1 + n_2)(\sigma_{12}^2 + m_{12}^2) + (n_3 + n_4)(\sigma_{34}^2 + m_{34}^2) - Nm_0^2]}$$

where  $R_{34}^2 = \hat{\beta}_{34}' c_{34}$ ,  $(n_3 + n_4)m_{34} = n_3 m_3 + n_4 m_4$ . Equation (11a) can now be used to test at the  $\alpha$  significance level the hypothesis

$H_2: \alpha_3 = \alpha_4$  and  $b_3 = b_4$  given  $H_1$  is true by computing

$$F = \left( \frac{N-3(p+1)}{(p+1)} \right) \left( \frac{\frac{3}{3} R^2 - \frac{2}{3} R^2}{\frac{1}{3}(1-\frac{2}{3} R^2)} \right)$$

and rejecting  $H_2$  if  $F$  exceeds  $F_{\alpha}(p+1, N-3(p+1))$ .

Equation (11a) can also be used to test the hypothesis

$H_3: \alpha_1 = \alpha_2, b_1 = b_2, \alpha_3 = \alpha_4$ , and  $b_3 = b_4$  by computing

$$F = \left( \frac{N-4(p+1)}{2(p+1)} \right) \left( \frac{{}_4 R^2 - {}_2 R^2}{(1-{}_4 R^2)} \right)$$

and rejecting H3 if F exceeds  $F_\alpha(2(p+1), N-4(p+1))$ .

For Stage 1, accepting H2 as true, the additional restrictions  $\alpha_{12} = \alpha_{34}$  and  $b_{12} = b_{34}$  are imposed and the mvue,  $\hat{\alpha}_{1234}$  and  $\hat{b}_{1234}$ , for the criterion cluster (1,2,3,4) formed from all four criteria are

$$\begin{bmatrix} \hat{\alpha}_{1234} \\ \hat{b}_{1234} \end{bmatrix} = \frac{(n_1+n_2)}{N} \begin{bmatrix} \hat{\alpha}_{12} \\ \hat{b}_{12} \end{bmatrix} + \frac{(n_3+n_4)}{N} \begin{bmatrix} \hat{\alpha}_{34} \\ \hat{b}_{34} \end{bmatrix}$$

The standard weights,  $\hat{\beta}_{1234}$ , and the validities,  $c_{1234}$ , for the cluster (1,2,3,4) are

$$\hat{\beta}_{1234} = \frac{1}{N\sigma_{1234}} (n_1+n_2)\sigma_{12}\hat{\beta}_{12} + (n_3+n_4)\sigma_{34}\hat{\beta}_{34}, \text{ and}$$

$$c_{1234} = \frac{1}{N\sigma_{1234}} (n_1+n_2)\sigma_{12}c_{12} + (n_3+n_4)\sigma_{34}c_{34}, \text{ where}$$

$$N\sigma_{1234}^2 = (n_1+n_3)(\sigma_{12}^2 + m_{12}^2) + (n_3+n_4)(\sigma_{34}^2 + m_{34}^2) - Nm_{1234}^2, \text{ and}$$

$$Nm_{1234} = n_1m_1 + n_2m_2 + n_3m_3 + n_4m_4.$$

The model used to obtain the estimates for cluster (1,2,3,4) is

$$(15) E \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} \begin{bmatrix} u_1 X_1 \\ u_2 X_2 \\ u_3 X_3 \\ u_4 X_4 \end{bmatrix} \begin{bmatrix} \alpha_{1234} \\ b_{1234} \end{bmatrix} = \begin{bmatrix} \alpha_{1234} \\ u_1 \\ u_2 \\ u_3 \\ u_4 \end{bmatrix} + b_{1234} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \end{bmatrix} \text{ with } D \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} = \sigma^2 I.$$

The squared multiple correlation coefficient,  ${}_1 R^2$ , for (15) is

$${}_1 R^2 = \hat{\beta}_{1234}' c_{1234}.$$

Equation (11a) can now be used to test at the  $\alpha$  significance level the hypothesis

H4:  $\alpha_{12} = \alpha_{34}$  and  $b_{12} = b_{34}$ , given  $\alpha_1 = \alpha_2$ ,  $\alpha_3 = \alpha_4$ ,  $b_1 = b_2$  and  $b_3 = b_4$  by computing

$$F = \left( \frac{N-2(p+1)}{(p+1)} \right) \left( \frac{{}_2 R^2 - {}_1 R^2}{(1-{}_2 R^2)} \right)$$

and rejecting H4 if F exceeds  $F_\alpha(p+1, N-2(p+1))$ . The hypothesis

H5:  $\alpha_1 = \alpha_2 = \alpha_3 = \alpha_4$  and  $b_1 = b_2 = b_3 = b_4$

can be tested at the  $\alpha$  significance level by computing

$$F = \left( \frac{N-4(p+1)}{3(p+1)} \right) \left( \frac{{}_4 R^2 - {}_1 R^2}{(1-{}_4 R^2)} \right)$$

and rejecting H5 if F exceeds  $F_\alpha(3(p+1), N-4(p+1))$ .

#### ***APPENDIX C: HIER-GRP SAMPLE OUTPUT***

The HIER-GRP sample output provided in this appendix was produced by using the Univac 1108 runstream which precedes it. The sample analysis involves a grouping problem with nine regression equations having nine predictors.

## UNIVAC 1108 RUNSTREAM FOR HIER-GRP SAMPLE PROBLEM

BRUN T478,3659,T  
WATE T+T-HIER-GRP  
009009620

SAMPLE HIERARCHICAL GROUPING ANALYSIS FOR A PROBLEM WITH NINE  
REGRESSION EQUATIONS CONTAINING NINE PREDICTOR VARIABLES

(11F7.0)

498. 498. 498. 498. 498. 498. 498. 498. 498.

(1eF12.6)

56.451007	21.308644					
53.395562	21.914342					
68.744980	16.605052					
72.044176	18.446208					
60.363453	17.790080					
52.500000	15.220105					
53.965863	15.463798					
61.050200	18.499403					
59.132530	14.158818					
(1eF12.6/3F12.6)						
.210613	.084266	.690537	.044687	.022704	.019168	
.030065	.051014	.002466				
.497270	.253267	.146937	.111735	-.048070	.070620	
.023948	.067965	.002673				
.830726	.062409	.110464	.066591	-.017240	.013254	
.008630	-.008133	.001335				
.000000	.503307	.046976	.400246	-.098074	.081764	
.088532	.010101	.055075				
.501937	.223452	.241215	.079214	-.058422	.015964	
.027251	.071612	.017675				
.157068	.523549	.177268	.109043	-.119310	.033768	
.061973	.125317	.077995				
.025357	.140070	.179064	.078076	-.055456	.063479	
.007705	.036213	.039056				
.054619	.609892	-.032781	.111520	-.159781	.119546	
.065551	.132338	.072512				
.183373	.264566	.198929	.281333	-.088878	.067113	
.030136	.106099	.041553				
(1eF12.6/3F12.6)						
.851747	.403108	.943397	.330665	-.267945	.263114	
.260275	.748397	.061240				
.812347	.525137	.748172	.400812	-.306481	.277655	
.284223	.641246	.048847				
.963222	.380671	.830779	.326702	-.325629	.318978	
.230669	.666403	.082854				
.379912	.716074	.383850	.652332	-.200817	.054977	
.400267	.345576	.0165272				
.867424	.523466	.823604	.388907	-.337009	.251625	
.294623	.694826	.074963				
.641551	.734258	.649314	.466729	-.317681	.130361	
.383139	.589730	.095018				
.902620	.431595	.817043	.351056	-.347421	.323654	
.243765	.672775	.103907				
.442066	.726975	.420952	.427142	-.284123	.129945	
.355647	.429272	.062678				
.627808	.539934	.635186	.526863	-.292518	.184302	
.298642	.565390	.073218				

(5x,2F18.4)

1	1025.3594	160.2755
2	281.0361	.50.0347
3	43.6386	30.6446
4	2.9217	.6665
5	.6526	.4761
6	3.9398	1.3035
7	343.2791	44.7462
8	40.7771	30.2306
9	.7048	.4561

WF IN

## HIERARCHICAL GROUPING PROGRAM HIEK-GRP

COMPUTATIONAL SCIENCES DIVISION  
AF HUMAN RESOURCES LABORATORY  
AIR FORCE SYSTEMS COMMAND

VERSION DATE 11 JAN 1978

1. CONTROL CARD PARAMETERS

NUMBER OF REGRESSION EQUATIONS = 9 (NUMBER OF PREDICTOR VARIABLES = 9  
GROUPING OPTION = 6 NUMBER OF HEADER CARDS = 2)

PROBLEM HEADER LABEL

SAMPLE HIERARCHICAL GROUPING ANALYSIS FOR A PROBLEM WITH NINE  
REGRESSION EQUATIONS CONTAINING NINE PREDICTOR VARIABLES

II. FORMAT CARDS AND INPUT DATA

FORMAT TO READ SMI(1) = (11F7.0)

FORMAT TO READ SMD(1) AND SSD(1) = (2F12.6)

EQUATION	N	CRITERION MEAN	CRITERION SD
1	498.	56.45181	21.30867
2	498.	63.39558	21.91434
3	498.	60.74498	16.60505
4	498.	72.04418	18.44621
5	498.	65.86345	17.79008
6	498.	52.50000	15.22011
7	498.	53.94636	15.46360
8	498.	61.05020	18.49940
9	498.	57.13253	14.15882

FORMAT TO READ BETA WEIGHTS = (6F12.6/3F12.6)

BETA WEIGHTS	1	2	3	4	5	6	7	8	9
EQU 1	.2106	.0843	.6905	.0447	.0227	.0192	.0301	.0510	.0023
EQU 2	.4973	.2533	.1969	.0117	.0031	.0705	.0239	.0580	.0027
EQU 3	.8307	.0624	.1105	.0566	.0172	.0133	.0086	-.0081	.0013
EQU 4	.0000	.5033	.0470	.4002	-.0981	.0818	.0885	.0101	.0551
EQU 5	.5019	.2885	.2412	.0792	-.0584	.0160	.0273	.0716	.0177
EQU 6	.1570	.5235	.1773	.1090	-.1193	.0338	.0020	.253	.0780
EQU 7	.6254	.1011	.1791	.0781	-.0555	.0635	.0079	.0262	.0391
EQU 8	.0546	.6079	-.0328	.1115	-.1598	.1195	.00659	.1323	.0725
EQU 9	.1834	.2646	.1989	.2813	-.0889	.0671	.0301	.1061	.0416

VALIDITIES	1	2	3	4	5	6	7	8	9
EQU 1	.8517	.4031	.9434	.3307	-.2679	.2631	.2603	.7484	.0612
EQU 2	.8123	.5251	.7482	.4008	-.3065	.2777	.2842	.6412	.0488

## COMPUTER OUTPUT FROM HIERARCHICAL SAMPLE PROGRAM

DATE 020378 PAGE 2

EQU	4	*.9632	*.3807	*.3308	*.3267	-*.3256	*.3190	*.2309	*.6664	*.0822
EQU	5	*.3794	.7181	.3639	.6523	-.2008	.0550	.4003	.3456	.0653
EQU	6	*.674	*.235	*.6230	*.3889	-.3370	*.216	.2948	*.6948	*.0750
EQU	7	*.6416	*.7343	*.493	*.4667	-.3177	*.1304	.5897	*.0950	
EQU	8	*.9026	*.4316	*.6170	*.3511	-.3474	*.3237	.2438	*.6728	*.1039
EQU	9	*.4921	*.7270	*.4210	*.4271	-.2841	*.1299	.3558	*.4293	*.0627
EQU	10	*.6278	*.5399	*.6352	*.5269	-.2925	*.1843	.2988	*.5654	*.0732

FORMAT TO READ PM(L) AND PSU(L) = (5X,2F18.4)

PREDICTOR PREDICTOR MEAN SD

1		1.025	*.5941	1.60	*.2755	U				
2		2.31	*.0361	50	*.0347	U				
3		4.3	*.3360	30	*.6446	U				
4		2.0	*.2170	*.8665	U					
5		5	*.5260	*.4761	U					
6		3.93	*.9460	1.*.3035	U					
7		34.3	*.2791	44.*.7462	U					
8		40.7771	U	30.*.2306	U					
9		9	*.0480	*.4561	U					

## III. HIERARCHICAL GROUPING RESULTS

MINIMIZED DECISION VALUE = ORU(KU+1)-ORU(KU)

ORU(KU) = R SQUARED FOR KU SYSTEMS.

KU = NUMBER OF SYSTEMS REMAINING AT KU STAGE.

GROUPING OPTION 6

9 INITIAL SYSTEMS RSQ = \*823J

SYS	RSQ										
1	*.947	2	.7765	3	.9440	4	.7072	5	.8673	6	.7946
*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****	*****

STAGE = 3 OVERALL RSQ = \*.d203

F-TEST FOR THE EQUALITY OF REGRESSION PARAMETERS

SYSTEMS GROUPING THIS STAGE FOR SYS'S COMBINED UP TO THIS STAGE

SYS	NO.	NO.	CHANGE FROM	RESIDUAL	SYS	RSQ	SYS	RSQ	SYS	RSQ
IDENT	MEMBERS	CASES	KSYS	9 SYSTEMS	1770	R5Q	4392	DF	1U	*.0026
2	1	498	*.7765	•	•0026	•1770	•0026	•1770		
5	1	498	*.8673	•	10	4392	1U	4392		

DECISION VALUE = .0026 FSTAT = 6.53 SIG LVL = .0000

## SYSTEMS SUMMARY RUSTER

STAGE	SYS	NO.	NO.	SYS						
IDENT	LOSS	MEMBERS	KSYN	CASES	LuEN	IDENTIFICATION OF OTHER MEMBERS				
8	*.0026	2	*.8028	946	2	5				

BETA WEIGHTS FOR THE NEW SYSTEM 2  
NEW SYS CRITERION MEAN = 62.1295  
NEW SYS CRITERION SD = 19.9991



## COMPUTER OUTPUT FROM HIER-GRP SAMPLE PROBLEM

						DATE 020378	PAGE 4
IDENT LOSS MEMBERS RSQ			CASES	IDENT	IDENTIFICATION OF OTHER MEMBERS		
IDENT	LOSS	MEMBERS	RSQ	CASES	IDENT		
8	.0026	2	.8026	996	2	5	
6	.0094	2	.7707	996	6	7	
STAGE = 5	OVERALL RSQ = .7385						
BETA WEIGHTS FOR THE NEW SYSTEM 6							
						NEW SYS CRITERION MEAN = 53.2329	
						NEW SYS CRITERION SD = 15.3599	
						6	7
						8	9
						• 3926	.3299
						.1780	.0933
						.0870	.0487
						.0347	.0803
							.0583
RAW SCORE WEIGHTS FOR THE NEW SYSTEM 6							
						REGRESSION CONSTANT = -30.0948	
						5	
						1	
						2	
						3	
						4	
						• 0376	.1013
						.0119	.0892
						.0408	.1.6544
							-2.8077
							.1.9634
7	.0086	2	.5460	996	8	9	
3 SINGLE MEMBER SYSTEMS 1 3 4							
STAGE = 5	OVERALL RSQ = .7385						
F-TEST FOR THE EQUALITY OF REGRESSION PARAMETERS							
SYSTEMS GROUPING THIS STAGE							
						FOR SYS'S COMBINED AT THIS STAGE	FOR SYS'S COMBINED UP TO THIS STAGE
						CHANGE FROM	
						9 SYSTEMS	
						RESIDUAL	
						RSQ	
						.0138	.0345
						.1977	.1770
						4422	40
						DF	4392
DECISION VALUE= .0138	FSTAT= .0138					FSTAT= 21.38 SIG LVL= .0000	
SYSTEMS SUMMARY RUSTER							
STAGE	SYS NO.	NO.	SYS				
IDENT	LOSS	MEMBERS	RSQ	CASES	IDENT	IDENTIFICATION OF OTHER MEMBERS	
5	.0165	3	.8030	1494	2	5 3	
STAGE = 4	OVERALL RSQ = .7663						
BETA WEIGHTS FOR THE NEW SYSTEM 2							
						NEW SYS CRITERION MEAN = 64.3347	
						NEW SYS CRITERION SD = 19.1906	
						6	7
						8	9
						• 5d40	.1850
						.1623	.0862
						.0413	.0356
						.0200	.0457
							.0069
RAW SCORE WEIGHTS FOR THE NEW SYSTEM 2							
						REGRESSION CONSTANT = -42.6322	
						5	
						1	
						2	
						3	
						4	
						• 0699	.0710
						.5247	.0066
						.0117	.1.9094
						.0290	.2888
6	.0094	2	.7707	996	6	7	
7	.0086	2	.5460	996	8	9	
2 SINGLE MEMBER SYSTEMS 1 4							
STAGE = 4	OVERALL RSQ = .7663						
SYSTEMS GROUPING THIS STAGE							
						FOR SYS'S COMBINED AT THIS STAGE	FOR SYS'S COMBINED UP TO THIS STAGE
						CHANGE FROM	
						9 SYSTEMS	
						RESIDUAL	
						RSQ	
						.0215	.0562
						.0117	.0170

DECISION VALUE =	•.0217	FSTAT =	45.16	SIG LVL =	•0000	FSTAT =	27.87	SIG LVL =	•0000
------------------	--------	---------	-------	-----------	-------	---------	-------	-----------	-------

## SYSTEMS SUMMARY RASTER

STAGE	SYS NO.	NO.	SYS
IDENT LOSS MEMBERS RSQ	CASES	IDENTIFICATION OF OTHER MEMBERS	
5 •0165	3 8130	1494 2 5 3	
4 •0397	4 •6258	1942 6 7 8 9	
			BETA WEIGHTS FOR THE NEW SYSTEM 6
			NEW SYS CRITERION MEAN = 56.0621
			NEW SYS CRITERION SD = 16.3052
			REGRESSION CONSTANT = -23.9405
			CHANGE FROM
			9 SYSTEMS RESIDUAL
			.0823 .1770
			DF .60
			RSQ .0531
			2.0395
			4
			5
			-3.0167
			2.5717
			4
			3
			6
			7
			8
			9
			-----
			RAW SCORE WEIGHTS FOR THE NEW SYSTEM 6
			REGRESSION CONSTANT = -35.6910
			CHANGE FROM
			9 SYSTEMS RESIDUAL
			.0307 .0464
			DF .2431
			RSQ .0055
			4
			5
			-9952
			1.7068
			2.431
			-----
			1 SINGLE MEMBER SYSTEMS 4
			-----
			2 OVERALL RSQ = •.6828
			-----

## COMPUTER OUTPUT FROM HIER-GNP SAMPLE PROBLEM

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SYSTEMS GROUPING THIS STAGE		FOR SYS'S COMBINED AT THIS STAGE		FOR SYS'S COMBINED UP TO THIS STAGE	
SYS NO.	NO.	CHANGE FROM		CHANGE FROM	
IDENT MEMBERS CASES RSQ	CASES	3 SYSTEMS RESIDUAL		9 SYSTEMS RESIDUAL	
1 4 1992 .707	.707	.0578	.2594	.1402	.1770
6 4 1992 .6258	.6258	10 DF	4452	70 DF	4392
DECISION VALUE= .0578	FSTAT= 99.23 SIG LVL= .0000			FSTAT= 49.67 SIG LVL= .0000	

## SYSTEMS SUMMARY ROSTER

STAGE	SYS NO.	NO.	SYS
IDENT LOSS MEMBERS RSQ	CASES	1 IDENTIFICATION OF OTHER MEMBERS	
2 .1402 8 .6636	3984	1 2 5 3 6 7 8 9	
		BETA WEIGHTS FOR THE NEW SYSTEM 1	
		1 2 3 4 5 6 7 8 9	
		• .3636 • .2543 • .2146 • .1003 • .0594 • .0481 • .0305 • .0282	
		RAW SCORE WEIGHTS FOR THE NEW SYSTEM 1	
		1 2 3 4 5	
		• .0419 • .0940 • .1295 • .1392 • .0419 • .1413	
1 SINGLE MEMBER SYSTEMS 4			
STAGE = 1 OVERALL RSQ = .6184			

## F-TEST FOR THE EQUALITY OF REGRESSION PARAMETERS

SYSTEMS GROUPING THIS STAGE		FOR SYS'S COMBINED AT THIS STAGE		FOR SYS'S COMBINED UP TO THIS STAGE	
SYS NO.	NO.	CHANGE FROM		CHANGE FROM	
IDENT MEMBERS CASES RSQ	CASES	2 SYSTEMS RESIDUAL		9 SYSTEMS RESIDUAL	
1 8 3984 .6636	.6636	.0745	.3172	.2146	.1770
4 1 498 .7072	.7072	10 DF	4462	80 DF	4392
DECISION VALUE= .0745	FSTAT= 104.75 SIG LVL= .0000			FSTAT= 66.56 SIG LVL= .0000	

## SYSTEMS SUMMARY ROSTER

STAGE	SYS NO.	NO.	SYS
IDENT LOSS MEMBERS RSQ	CASES	1 IDENTIFICATION OF OTHER MEMBERS	
1 .2146 9 .6083	4982	1 2 5 3 6 7 8 9 4	
		BETA WEIGHTS FOR THE NEW SYSTEM 1	
		1 2 3 4 5 6 7 8 9	
		• .3162 • .2757 • .1917 • .1306 • .0623 • .0507 • .0361	
		RAW SCORE WEIGHTS FOR THE NEW SYSTEM 1	
		1 2 3 4 5	
		• .0373 • .1041 • .1162 • .8483	
		• .7353 • .0152 • .0379 • .2620	

END OF JNL

***APPENDIX D: HIER-GRP SOURCE LISTING***

SOURCE LISTING FOR HIER-GRP DRIVER (MAIN) PROGRAM H6-MAIN

1:C !\$!# ABCDEFGHIJKLMNOPQRSTUVWXYZ+-+<-&>(\*?!,/0123456789\*;\n0"

2:C THE ABOVE LINE IS THE UNIVAC LINE CHARACTER SET (OCTAL 00 - 77)

3:C -----

4:C SECTION 1)

HIER-GRP

5:C -----

6:C -----

7:C A COMPUTER PROGRAM FOR THE HIERARCHICAL GROUPING OF REGRESSION EQUATIONS

8:C -----

9:C SECTION 2)

TABLE OF CONTENTS

- |      |   |
|------|---|
| 10:C | SECTION 1) HIER-GRP TITLE                   |
| 11:C | SECTION 2) TABLE OF CONTENTS                |
| 12:C | SECTION 3) ALGORITHM DESCRIPTION            |
| 13:C | SECTION 4) REQUIRED INFORMATION FOR USERS   |
| 14:C | SECTION 5) HOW TO USE HIER-GRP              |
| 15:C | SECTION 6) PROGRAMS IN THE HIER-GRP PACKAGE |
| 16:C | SECTION 7) VARIABLE DESCRIPTIONS            |
| 17:C | SECTION 8) HIER-GRP DRIVER PROGRAM          |
| 18:C |   |
| 19:C |   |
| 20:C |   |

21:C SECTION 3)

ALGORITHM DESCRIPTION

22:C -----

23:C -----

24:C THE BASIC IDEA OF THE HIER-GRP ALGORITHM IS TO REDUCE A SET OF REGRES-

25:C SION EQUATIONS (ALSO CALLED SYSTEMS), COMPUTED FROM PROPORTIONAL PREDICTION

26:C THE SUMS OF CROSS PRODUCTS MATRICES, TO A SINGLE EQUATION. AT EACH STEP

27:C THE NUMBER OF EQUATIONS IS REDUCED BY ONE. THIS IS ACCOMPLISHED BY RE-

28:C PLACING TWO EQUATIONS WITH A COMPROMISE EQUATION. THE REMAINING EQUATIONS

29:C TOGETHER WITH THE COMPROMISE EQUATION FORM THE POOL FOR THE NEXT REPLACE-

30:C MENT. THE TWO EQUATIONS REPLACED ARE REFERRED TO AS THE MEMBERS OF THE

31:C COMPROMISE SYSTEM. IF TWO COMPROMISE EQUATIONS ARE REPLACED, THEN THE

32:C MEMBERS OF EACH COMPROMISE EQUATION BECOME MEMBERS OF THE NEW COMPROMISE

33:C SYSTEM. THE CRITERIA FOR SELECTING THE TWO EQUATIONS TO BE REPLACED IS

34:C THE SAME AT EVERY STEP. THE PROGRAM HAS SIX GROUPING CRITERIA, ONE OF

35:C WHICH MUST BE SPECIFIED BY THE USER. ALL THE GROUPING CRITERIA MAY BE  
 36:C VIEWED AS BEING CONCEPTUALLY THE SAME. IN EACH CASE THE REPLACEMENT OF  
 37:C TWO EQUATIONS BY A COMPROMISE EQUATION IS ASSOCIATED WITH A LOSS; THOUGH  
 38:C THE LOSS AND THE METHOD USED TO COMPUTE THE LOSS ARE DIFFERENT FOR EACH.  
 39:C THE PAIR OF EQUATIONS TO BE REPLACED IS THE PAIR HAVING THE SMALLEST OR  
 40:C LARGEST ASSOCIATED LOSS DEPENDING ON WHETHER IT IS DESIRED TO GROUP TO-  
 41:C GETHER SIMILAR OR DISSIMILAR EQUATIONS. THE SIX CRITERIA ARE LISTED IN  
 42:C THE FOLLOWING TABLE.

	GROUPING CRITERIA AVAILABLE IN HIER-GRP	
	GROUPING	LOSS ASSOCIATED WITH REPLACEMENT OF SYS'S I AND J
43:C	CRITERIA	CRITERIA
44:C	OPTION	OPTION
45:C	LOSS	LOSS
46:C		
47:C	1 ** LOSS = DECAYFASE IN THE SQUARED MULTIPLE CORRELATION	LOSS = WEIGHTED AVERAGE OF LOSSES FOR ALL COMBINATIONS
48:C	COEFFICIENT, RSQ, UNDER THE HYPOTHESIS THAT THE BETA	OF A MEMBER OF SYS I WITH A MEMBER OF SYS J, WHERE FOR
49:C	WEIGHTS FOR EQU I EQUAL THE CORRESPONDING WEIGHTS FOR	A GIVEN COMBINATION THE WEIGHT IS THE PRODUCT OF THE
50:C	EQU J.	NUMBER OF CASES IN EACH MEMBER OF THE COMBINATION AND
		THE LOSS IS THE DECREASE IN RSQ UNDER THE HYPOTHESIS
51:C	2 ** LOSS = SAME AS FOR 1.	THAT THE CORRESPONDING BETA WEIGHTS FOR EACH MEMBER
52:C	3 ** LOSS = WEIGHTED AVERAGE OF LOSSES FOR ALL COMBINATIONS	OF THE COMBINATION ARE EQUAL.
53:C	OF A MEMBER OF SYS I WITH A MEMBER OF SYS J, WHERE FOR	LOSS = MINIMUM OF LOSSES FOR ALL COMBINATIONS
54:C	A GIVEN COMBINATION THE WEIGHT IS THE PRODUCT OF THE	OF A MEMBER OF SYS I OR SYS J WITH A DIFFERENT MEMBER
55:C	NUMBER OF CASES IN EACH MEMBER OF THE COMBINATION AND	OF EITHER SYS I OR SYS J, WHERE FOR A GIVEN COMBINATION
56:C	THE LOSS IS THE DECREASE IN RSQ UNDER THE HYPOTHESIS	THE LOSS IS THE DECREASE IN RSQ UNDER THE HYPOTHESIS
57:C	THAT THE CORRESPONDING BETA WEIGHTS FOR EACH MEMBER OF	(MAX LOSS)
58:C	THE COMBINATION ARE EQUAL.	THE COMBINATION ARE EQUAL.
59:C	4 ** LOSS = SAME AS FOR 2 BUT WITH MAXIMUM REPLACING	LOSS = SAME AS FOR 2 BUT WITH MAXIMUM REPLACING
60:C	MINIMUM AS THE FIRST WORD OF THE DESCRIPTION.	MINIMUM AS THE FIRST WORD OF THE DESCRIPTION.
61:C		
62:C		
63:C		
64:C		
65:C		
66:C		
67:C		
68:C		
69:C	SECTION 4)	
70:C	REQUIRED INFORMATION FOR USERS	

71:C FOR EACH EQUATION, THE FOLLOWING INFORMATION IS REQUIRED:

72:C 1. THE NUMBER OF CASES(OBSERVATIONS)

74:C 2. THE CRITERION MEAN AND STANDARD DEVIATION

75:C 3. THE STANDARDIZED REGRESSION WEIGHTS

76:C 4. THE VALIDITY COEFFICIENTS

77:C 5. THE MEANS AND STANDARD DEVIATIONS OF THE PREDICTORS.

78:C IT IS IMPORTANT TO REALIZE THAT THE HIER-GRP ALGORITHM REQUIRES ALL THE REGRESSION EQUATIONS TO BE LEAST SQUARES SOLUTIONS DERIVED FROM

79:C PROPORTIONAL PREDICTOR SUMS OF CROSS PRODUCTS MATRICES (I.E., THE MEAN

80:C OF A PREDICTOR FOR ONE EQUATION MUST EQUAL THE MEAN OF THE CORRESPONDING

81:C PREDICTORS IN ALL OTHER EQUATIONS. IN ADDITION THE COVARIANCE MATRICES

82:C FOR EACH EQUATION MUST BE IDENTICAL.). THIS IS THE PROPORTIONALITY

83:C ASSUMPTION OF BUTTERBERG AND CHRISTAL (WADD-TN-61-30).

84:C

85:C

86:C SECTION 5)

87:C HOW TO USE HIER-GRP

88:C

89:C TO EXECUTE THIS PROGRAM THE USER MUST PREPARE THE FOLLOWING SEQUENCE

90:C OF CONTROL CARDS AND DATA CARDS:

91:C 1. CONTROL CARD.

92:C CARD FORTKAN

93:C COLUMNS FORMAT DESCRIPTION

94:C 1-3 13 NEQS, NUMBER OF REGRESSION EQUATIONS(SYSTEMS) TO BE

95:C GROUPED

96:C NEQS MUST BE •LE. 50

97:C NPREDs, NUMBER OF PREDICTOR VARIABLES IN EACH EQUATION, I.E., NUMBER OF BETA WEIGHTS(STANDARDIZED)

98:C

99:C REGRESSION(WEIGHTS) IN EACH EQUATION.

100:C NPREDs MUST BE •LE. 100

101:C 101 11 101, THE GROUPING OPTION DESIRED. NORMALLY OPTION

102:C •6• IS SPECIFIED WHICH CAUSES THE GROUPING TO BE

103:C DONE BASED ON THE ITERATIVE TECHNIQUE DEVELOPED BY

104:C BOTTERBERG AND CHRISTAL, WADD-TN-61-30.

105:C SEE SECTION 3 FOR MORE INFORMATION

106:C NIDS, NUMBER OF HEADER(LABEL,TITLE) CARDS THAT

107:C FOLLOW THIS CONTROL CARD. THE NUMBER OF HEADER  
108:C CARDS CAN RANGE FROM 0 TO 9.

109:C 9 11 READ, THE DATA READ OPTION. IREAD = 0 MEANS  
110:C READ THE BETA WEIGHTS AND VALIDITIES NPREDS  
111:C ITEMS AT A TIME. IREAD = 1 MEANS READ THEM  
112:C NEQS•NPREDS ITEMS AT A TIME.

113:C 2. • HEADER CARD(S)  
114:C CARD FORTRAN  
115:C COLUMNS FORMAT DESCRIPTION  
116:C 1-8U 13A6,A2 EACH HEADER CARD CONTAINS UP TO 80 ALPHANUMERIC  
117:C CHARACTERS WHICH ARE PRINTED(80 CHARACTERS/LINE) AT  
118:C THE BEGINNING OF THE GROUPING REPORT. HEADER CARDS  
119:C CAN BE OMITTED, OR UP TO 9 CARDS CAN BE USED,  
120:C DEPENDING UPON THE NUMBER(NHDRS) SPECIFIED IN THE  
121:C INITIAL CONTROL CARD. THE NUMBER OF HEADER CARDS  
122:C PRESENT MUST EQUAL NHDRS.  
123:C 3. FORMAT CARD TO READ THE NUMBER OF CASES FOR EACH EQUATION  
124:C CARD FORTRAN  
125:C COLUMNS FORMAT DESCRIPTION  
126:C 1-8U 13A6,A2 THE FORTRAN VARIABLE FORMAT BY WHICH THE SN(I) ARE  
127:C TO BE READ. SN(I) DENOTES THE NUMBER OF CASES THAT  
128:C WERE USED IN THE COMPUTATION OF EQUATION I. FOR  
129:C EXAMPLE, IF 10 EQUATIONS ARE TO BE GROUPED(NEQS=10)  
130:C THE FORMAT CARD MIGHT BE •(10F5.0)•. THE F EDIT  
131:C CODE MUST BE USED SINCE SN(I) IS A REAL VARIABLE.  
132:C 4. SN(I) DATA CARD(S)  
133:C CARD FORTRAN  
134:C COLUMNS FORMAT DESCRIPTION  
135:C \*\*\* DATA IS READ ACCORDING TO PREVIOUS FORMAT  
136:C 5. FORMAT CARD TO READ CRITERION MEANS AND STANDARD DEVIATIONS  
137:C CARD FORTRAN  
138:C COLUMNS FORMAT DESCRIPTION  
139:C 1-80 13A6,A2 THE FORTRAN VARIABLE FORMAT BY WHICH THE SM(I) AND  
140:C THE SS(I) DATA ARE TO BE READ. SM(I) DENOTES THE  
141:C CRITERION MEAN FOR EQUATION I. SSD(I) DENOTES THE  
142:C CRITERION STANDARD DEVIATION FOR EQUATION I. FOR

143:C  
 144:C  
 145:C        EXAMPLE, IF 16 EQUATIONS ARE TO BE GROUPED (NEQS=10)  
 146:C        THE 10 PAIRS OF SM(1) AND SSD(1) MIGHT BE READ WITH  
 147:C        THE FOLLOWING FORMAT CARD: \*(IUX,F10.4,5X,F10.4)\*.  
 6• SM(1) AND SSD(1) DATA CARDS.  
 148:C        COLUMNS FORMAT DESCRIPTION  
 149:C        CARD FORTRAN  
 150:C        7• FORMAT CARD TO READ BETA WEIGHTS FOR EACH EQUATION.  
 151:C        CARD FORTRAN  
 152:C        COLUMNS FORMAT DESCRIPTION  
 153:C        1-80    13A6,A2    THE FORTRAN VARIABLE FORMAT BY WHICH THE BETA  
 154:C        WEIGHTS FOR EACH EQUATION ARE READ. THE BETA  
 155:C        WEIGHTS(NPRED'S WEIGHTS PER EQUATION) ARE STORED IN  
 156:C        THE B(1) ARRAY. THE FIRST NPRED'S ELEMENTS OF B  
 157:C        CONTAIN THE BETA WEIGHTS FOR EQUATION 1, THE NEXT  
 158:C        NPRED'S ELEMENTS CONTAIN THE BETA WEIGHTS FOR  
 159:C        EQUATION 2, AND SO ON. THE WEIGHTS FOR EACH  
 160:C        EQUATION ARE STORED IN VARIABLE NUMBER ORDER.  
 8• BETA WEIGHT DATA CARDS.  
 161:C        CARD FORTRAN  
 162:C        COLUMNS FORMAT DESCRIPTION  
 163:C        9• FORMAT CARD TO READ VALIDITY COEFFICIENTS FOR EACH EQUATION.  
 164:C        \*\*\* THE BETA WEIGHTS ARE READ ACCORDING TO THE PREVIOUS FORMAT.  
 165:C        CARD FORTRAN  
 166:C        COLUMNS FORMAT DESCRIPTION  
 167:C        1-80    13A6,A2    THE FORTRAN VARIABLE FORMAT BY WHICH THE VALIDITY  
 168:C        COEFFICIENTS (CORRELATIONS BETWEEN THE PREDICTOR  
 169:C        VARIABLES AND THE KREITERION VARIABLE) FOR EACH  
 170:C        EQUATION ARE TO BE READ. THE VALIDITY COEFFICIENTS  
 171:C        (NPRED'S COEFFICIENTS PER EQUATION) ARE STORED IN  
 172:C        THE V(1) ARRAY. THE FIRST NPRED'S ELEMENTS OF V  
 173:C        CONTAIN THE VALIDITIES FOR EQUATION 1, THE NEXT  
 174:C        NPRED'S ELEMENTS CONTAIN THE VALIDITIES FOR EQUATION  
 175:C        2, AND SO ON. THE VALIDITIES FOR EACH EQUATION ARE  
 176:C        STORED IN VARIABLE NUMBER ORDER.  
 177:C        1.J• THE VALIDITY COEFFICIENT DATA CARDS.

CARD	FORTRAN	COLUMNS	FORMAT	DESCRIPTION
179:C 180:C				
181:C 182:C	*** THE VALIDITIES ARE READ ACCORDING TO THE PREVIOUS FORMAT.			
11:C	FORMAT CARD TO READ PREDICTOR MEANS AND STANDARD DEVIATIONS.			
183:C	CARD FORTRAN			
184:C	COLUMNS FORMAT			
185:C 1-80	13A6,A2	THE FORTRAN VARIABLE FORMAT BY WHICH THE PM(L)		
186:C	AND PSD(L) DATA ARE TO BE READ. PM(L) DENOTES			
187:C	THE MEAN OF THE PREDICTOR VARIABLE L. PSD(L)			
188:C	DENOTES THE STANDARD DEVIATION OF PREDICTOR			
189:C	VARIABLE L.			
190:C	12. PM(L) AND PSD(L). DATA CARDS.			
191:C	CARD FORTRAN			
192:C	COLUMNS FORMAT			
193:C	*** DATA IS READ ACCORDING TO PREVIOUS FORMAT.			
194:C	13. REPEAT 1 - 12 ABOVE AS OFTEN AS DESIRED, THEN INCLUDE A BLANK			
195:C	CARD AS THE LAST CARD IN THE DECK.			
196:C				
197:C SECTION 6)	PROGRAMS IN THE HIER-GRP PACKAGE			
198:C				
199:C 200:C	6.1 HG-MAIN. FORTRAN MAIN PROGRAM. PACKAGE DRIVER - READS			
201:C	CONTROL CARD, INPUT DATA, CALLS REMAINING ROUTINES.			
202:C	6.2 START. ASSEMBLER SUBROUTINE. RESETS PAGE MARGINS ON			
203:C	THE PRINTER TO 0 LINES AT TOP, 0 LINES AT BOTTOM.			
204:C	6.3 OVRLP. FORTRAN SUBROUTINE. COMPUTES THE OVERLAP			
205:C	VECTOR, A, FOR USE IN GROUP.			
206:C	6.4 GROUP. FORTRAN SUBROUTINE. PERFORMS THE GROUPING			
207:C	PROCEDURE.			
208:C	6.5 STAGE. FORTRAN SUBROUTINE. COMPUTES THE GROUPING			
209:C	OUTPUT.			
210:C	6.6 PRINTG. FORTRAN SUBROUTINE. PRINTS THE GROUPING			
211:C	REPORT.			
212:C	6.7 PLEVEL. FORTRAN SUBROUTINE. CALCULATES THE PROB-			
213:C	ABILITY OF THE F-RATIO TEST STATISTIC.			
214:C				

215:C SECTION 7)

VARIABLE DESCRIPTIONS

216:C  
217:C  
218:C NEQS = NUMBER OF REGRESSION EQUATIONS TO BE GROUPED.  
219:C NPREDs = NUMBER OF BETA WEIGHTS IN EACH EQUATION.  
220:C IOPT = GROUPING OPTION SWITCH. (SEE SECTION 3, AND 5 ABOVE FOR DETAILS)  
221:C IREAD = THE DATA READ OPTION. (SEE SECTION 5)  
222:C NHDRs = NUMBER OF CARDS IN HEADER LABEL. (MUST BE LESS THAN 10)  
223:C FMT = FORMAT AREA.  
224:C SN = VECTOR OF N'S. SN(I) IS THE NUMBER OF CASES IN SYSTEM I.  
225:C REQUIRED SIZE = NEQS.  
226:C SM = VECTOR OF CRITERION MEANS. SM(I) IS THE CRITERION MEAN FOR  
227:C SYSTEM I.  
228:C REQUIRED SIZE = NEQS.  
229:C SSD = VECTOR OF CRITERION STANDARD DEVIATIONS. SSD(I) IS THE CRITERION  
230:C STANDARD DEVIATION FOR SYSTEM I.  
231:C REQUIRED SIZE = NEQS.  
232:C B = VECTOR OF BETA WEIGHTS. B(I,L) IS THE BETA WEIGHT FOR PREDICTOR L  
42 233:C IN SYSTEM I, WHERE IL=NPREDs\*(I-1)+L.  
234:C REQUIRED SIZE = NEQS\*NPREDs.  
235:C V = VECTOR OF VALIDITIES. V(I,L) IS THE CORRELATION BETWEEN PREDICTOR  
236:C L AND THE CRITERION VARIABLE FOR SYSTEM I, WHERE IL IS AS IN B.  
237:C REQUIRED SIZE = NEQS\*NPREDs.  
238:C PM = VECTOR OF PREDICTOR MEANS. PM(L) IS THE MEAN OF PREDICTOR  
239:C VARIABLE L.  
240:C REQUIRED SIZE = NPREDs.  
241:C PSD = VECTOR OF STANDARD DEVIATIONS OF PREDICTOR VARIABLES.  
242:C PSD(L) IS THE STANDARD DEVIATION FOR PREDICTOR VARIABLE L.  
243:C REQUIRED SIZE = NPREDs.  
244:C A = THE OVERLAP VECTOR (COMPUTED IN SUBROUTINE OVRLP). A(I,J) IS THE  
245:C OVERLAP OR DECISION VALUE IF SYSTEMS I AND J ARE GROUPED, WHERE  
246:C I=NEQS\*(I-1) - 1\*(I-1)/2 + (J-1). ON RETURN FROM SUBROUTINE  
247:C OVRLP A(I,J) IS THE DROP IN RSQ WHEN SYSTEM I IS COMBINED WITH  
248:C SYSTEM J. SUBROUTINE GROUP USES THE A VECTOR TO PERFORM THE  
249:C GROUPING PROCEDURE.  
250:C REQUIRED SIZE = NEQS\*(NEQS-1)/2.

```

251:C   Z   WORK AREA USED IN OVRLP, GROUP, AND PRINTG. ON RETURN FROM GROUP
252:C   Z(KU) IS THE DECISION VALUE FOR THE GROUPING THAT OCCURED AT THE
253:C   KU SYSTEM STAGE.
254:C   REQUIRED SIZE = NEQS.
255:C   IU = VECTOR OF ABSORBING SYSTEM ID'S (COMPUTED IN GROUP). IU(KU) IS
256:C   THE ID OF THE ABSORBING SYSTEM SELECTED AT THE KU SYSTEM STAGE.
257:C   REQUIRED SIZE = NEQS.
258:C   JU = VECTOR OF ABSORBED SYSTEM ID'S (COMPUTED IN GROUP). JU(KU) IS THE
259:C   IU OF THE ABSORBED SYSTEM SELECTED AT THE KU SYSTEM STAGE.
260:C   REQUIRED SIZE = NEQS.
261:C   SR = VECTOR OF RSQ'S (COMPUTED IN STAGE). SR(I) IS THE RSQ ASSOCIATED
262:C   WITH REGRESSION EQUATION I.
263:C   REQUIRED SIZE = NEQS.
264:C   ORU = VECTOR OF RSQ'S (COMPUTED IN STAGE). ORU(KU) IS THE OVERALL RSQ
265:C   AT THE KU SYSTEM STAGE.
266:C   REQUIRED SIZE = NEQS.
267:C   SRU = VECTOR OF RSQ'S (COMPUTED IN STAGE). SRU(KU) IS THE RSQ FOR THE
268:C   PAIR OF SYSTEMS, IU(KU) AND JU(KU), GROUPING AT THE KU SYSTEM
269:C   STAGE.
270:C   REQUIRED SIZE = NEQS.
271:C   KP = VECTOR OF SYSTEM ID'S (COMPUTED IN STAGE). KP(I) IS THE ID OF THE
272:C   SYSTEM FOLLOWING SYSTEM I IN THE FINAL HIERARCHICAL ORDERING.
273:C   REQUIRED SIZE = NEQS.
274:C   KS = INVERSE OF JU (COMPUTED IN PRINTG). IF JU(KU)=J THEN KS(J)=KU.
275:C   REQUIRED SIZE = NEQS.
276:C   KO = VECTOR OF SYSTEM ID'S USED IN PRINTG TO STORE THE MEMBERS OF EACH
277:C   SYSTEM AT EACH STAGE.
278:C   REQUIRED SIZE = NEQS.
279:C   bL( ) = WORK AREA, USED TO STORE BETA WEIGHTS FOR STEPWISE PRINTING
280:C   IN SUBROUTINE PRINTG.
281:C   REQUIRED SIZE = NPREDs.
282:C   -----
283:C SECTION 8)
284:C   -----
285:C   HIER-GRP PACKAGE DRIVER
286:C   -----
THE PURPOSE OF THIS PROGRAM IS:

```

```

287: C 1. READ AND PRINT CONTROL CARD.
288: C 2. READ AND PRINT INPUT DATA (NUMBER OF CASES, CRITERION MEANS AND
289: C STANDARD DEVIATIONS, BETA WEIGHTS, VALIDITIES, AND PREDICTOR MEANS
290: C AND STANDARD DEVIATIONS).
291: C 3. CALL ROUTINES TO: A) COMPUTE UVPLAP MATRIX, B) PERFORM GROUPING
292: C PROCEDURE, C) COMPUTE GROUPING OUTPUT, AND D) PRINT GROUPING REPORT.
293: C
294: C
295: C PARAMETER EQS=50,PKDS=100,VDT=EQS*PKDS,ADIM=(EQS*(EQS-1))/2
296: C MAXIMUM OF 50 EQUATIONS AND 100 PREDICTORS ALLOWED BY PARAMETER STATEMENT
297: C DIMENSION SN(EQS),SI(EQS),SSD(EQS),H(BVDIM),V(BVDIM),PM(PKDS),
298: C PSD(PKDS),A(ADIM),/EQS),IUE(EQS),JUE(EQS),SR(EQS),OKU(EQS),
299: C ZSU(EQS),KP(EQS),KS(EQS),RU(EQS),FMT(14),OE(PKDS)
300: C
301: C RESET PAGE MARGINS TO 0 LINES AT TOP, 0 LINES AT BOTTOM
302: C
303: C READ AND PRINT INITIAL CONTROL CARD
304: I READIS,BUU,ENU=1000)EQS,NPRED5,IOPT,NHRS,IREAD
305: SUU FORMAT(213,311)
306: IF(INEQS.EQ.0)GO TO 900
307: C
308: C CALL START
309: C PRINT 598
310: I SY5 FORMAT(11/20X,*HIERARCHICAL GROUPING PROGRAM HIER-GRP*/,
311: 1 80X,*COMPUTATIONAL SCIENCES DIVISION*/
312: 2 80X,* AF HUMAN RESOURCES LABORATORY*/
313: 3 80X,* AIR FORCE SYSTEMS COMMAND*/
314: 4 80X,* VERSION DATE 11 JAN 1978*)
315: C PRINT 600,NEQS,IPHEUS,IOPT,NHRS
316: C CONTROL CARD PARAMETERS*/6X,23(1H-),
317: 1// NUMBER OF REGRESSION EQUATIONS = *,12,5X,
318: 1* NUMBER OF PREDICTOR VARIABLES = *,12/1X,
319: 2* GROUPING OPTION = *,11,5X,
320: 4* NUMBER OF HEADER CARDS = *,11)
321: C IF((IOPT.LT.1).OR.(IOPT.GT.6))GO TO 901
322: C IF(NHRS.EQ.0)GO TO 3

```

```

323:C      READ AND PRINT HEADER CARDS;
324:      PRINT 597
325: 597 FORMAT//6X,*PROBLEM HEADER LINES//6X,2*(1H-/)
326:  DO 2 I=1,NHDRS
327:      READ S01,FMT
328:  S01 FORMAT(13A6,A2)
329:      PRINT 604,FMT
330:  604 FORMAT(1H ,13A6,A2)
331: 2  CONTINUE
332:C      DATA INPUT
333: 3  PRINT 596
334: 596 FORMAT//* 11* FORMAL CARDS AND INPUT DATA*/6X,2/(1H-/)
335:      READ S01,FMT
336:      PRINT 605,FMT
337: 605 FORMAT/* FORMAT TO READ SN(I) = *,13A6,A2)
338:      READ FMT,(SN(I),I=1,NEQS)
339:C
45
340:      READ S01,FMT
341:      PRINT 606,FMT
342: 606 FORMAT/* FORMAT TO READ SM(I) AND SS0(I) = *,13A6,A2)
343:      PRINT 607
344: 607 FORMAT//* EQUATION N CRITERION MEAN CRITERION SU*)
345:      READ FMT,(SM(I),SSD(I),I=1,NEQS)
346:      PRINT 608,(I,SN(I),SM(I),SSD(I),I=1,NEQS)
347: 608 FORMAT(1H ,18,F12.0,F16.5,F15.5)
348:C
349:  IEP = NEQS * NPREDS
350:  MIN = MIN(15,NPREDS )
351:      READ S01,FMT
352:      PRINT 609,FMT
353: 609 FORMAT//* FORMAT TO READ BETA WEIGHTS = *,13A6,A2)
354:      PRINT 610,( 1, 1 = 1, MIN )
355: 610 FORMAT* BETA WEIGHTS*,15,1418)
356:      IF ( IREAD .NE. 0 ) READ FMT,( B(I), I = 1,IEP )
357:  IE=0
358:      DO 4 I=1,NEQS

```

```

359:      IS=IE+1
360:      IE=IE+NPREDS
361:      IF ( IREAD .EQ. 0 ) READ FMT,(B(J),J=IS,IE)
362:      PRINT 611,1,(B(J),J=IS,IE)
363:      FORMAT( EQU *,13,2x,15F8.4,14/10X,15F8.4 )
364:      CONTINUE
365: C
366:      READ S01,FMT
367:      PRINT 612,FMT
368:      612   FORMAT(//,FORMAT TO READ VALIDITIES = *,13A6,A2)
369:      PRINT 613,( T, T = 1, MIN )
370:      FORMAT(* VALIDITIES*,17,14I8)
371:      IF ( IREAD .NE. 0 ) READ FMT,( V(I), I = 1,IEP )
372:      JE=0
373:      DO 5 I=1,NEGS
374:      IS=IE+1
375:      IE=IE+NPREDS
376:      IF ( IREAD .EQ. 0 ) READ FMT,(V(J),J=IS,IE)
46      PRINT 611,1,(V(J),J=IS,IE)
377:      CONTINUE
378:      S
379: C
380:      READ S01,FMT
381:      PRINT 614,FMT
382:      FORMAT(//,FORMAT TO READ PA(L) AND PSU(L) = *,13A6,A2)
383:      PRINT 615
384:      PRINT 615, PREDICTOR MEAN*,10X,*PREDICTOR SU*,10X,*PREDICTOR SU*
385:      READ FMT,(PA(L),PSU(L),L = 1,NPREDS)
386:      DO 617 L = 1,NPREDS
387:      PRINT 616,L,PA(L),PSU(L)
388:      FORMAT(1H ,16,3X,4X,F18.5,4X,F18.5)
389: C      COMPUTE OVERLAP MATRIX
390:      CALL OVERLAPS,NPREDS,SN,SSU,B,V,A,7)
391: C      PERFORM GROUPING PROCEDURE
392:      CALL GROUP(TOPT,NEUS,SN,A,Z,LU,JU)
393: C      COMPUTE STAGE VALUES
394:      CALL STAGE(NEWS,NPREDS,SN,SSU,S,V,LU,JU,SR,ORU,ERU,KP)

```

395: C PRINT GEOFULAS REPORT  
396: CALL PRINTS  
397: 1 (I0PT,NEWS,IPRENS,SU,SM,SSD,B,PH,PSD,Z,IU,JU,SK,ORU,SRU,  
398: 2 KS,KP,KU,A,SE)  
399: C RETURN TO START

400: GO TO 1 END OF JOB  
401: C  
402: 9UU PRINT OUT  
403: OUT FORMAT(1EEND OF JOB,0)  
404: STOP  
405: C CONTROL CARD ERROR  
406: 9U1 PRINT 6U2  
407: 6U2 FORMAT(1CONTROL CARD ERROR)  
408: STOP  
409: C  
410: 1000 PRINT 603  
411: 603 FORMAT(1END OF CARD FILE)  
412: STOP  
413: END

SOURCE LISTING FOR HIER-GNP SUBROUTINE START

```

1: AXIVD *          ASSEMBLER LANGUAGE ROUTINE
2: IMAGEL *M,66,J,U,*   RECALLS PAGE MARGINS
3: START*          TU U LINES FROM TOP
4: LA   AU,(2,IMAGE) *
5: ER   PRTCNS *
6: J    I,XII *
7: END*          AND U LINES FROM BOTTOM

```

SOURCE LISTING FOR HIER-GNP SUBROUTINE OVRFLP

```

1:C -----
2:C -----
3:C -----
4:C -----
5:C -----
6:C      THIS ROUTINE COMPUTES THE INITIAL VALUES OF THE OVERLAP VECTOR,
7:C      A. IT IS ASSUMED THAT THE PROPORTIONALITY ASSUMPTION (SEE SECTION 3
8:C      OF HG-MAIN) IS SATISFIED BY THE PREDICTOR VARIABLES; OTHERWISE THE
9:C      EQUATION FOR A(IJ) FOLLOWING STATEMENT 210 IS NOT VALID.
10:C -----
11:C -----
12:C -----
13:C      NEQS = NUMBER OF SYSTEMS.
14:C      NPREDS = NUMBER OF PREDICTOR VARIABLES.
15:C      SN (I) = NUMBER OF CASES (OBSERVATIONS)
16:C      SM (I) = CRITERION MEAN
17:C      SSD(I) = CRITERION STANDARD DEVIATION
18:C      B (IL) = STANDARD REGRESSION WEIGHT FOR PREDICTOR L IN SYSTEM I.
19:C      WHERE IL = (I-1)*NPREDS + L
20:C      V (IL) = CRITERION CORRELATION WITH PREDICTOR L IN SYSTEM I.

```

```

21:C      WHERE IL = (I-1)* NPREDS + L
22:C      A(IJ) = ORU(NENS)-ORU(NES-I) OBTAINED BY PAIRING SYSTEMS I AND J.
23:C      WHERE IJ = NENS*(I-1) - I*(I-1)/2 + (J-I) AND I .LT. J
24:C      Z( ) = WORK AREA
25:C
26:C      ORU(NENS) = R SQUARED FOR NENS SYSTEMS.
27:C      ORU(NES-I) = R SQUARED FOR NES-I SYSTEMS.
28:C
29:C
30:      SUBROUTINE UVRLP
31:      I ( NENS, NPREDS, SN, SM, SSD, B, V, A, Z )
32:C      INPUT   *   *   *   *   *
33:C      OUTPUT   *   *   *   *   *   *   W
34:C
35:      DIMENSION SN(IL), SM(IL), SSD(IL), B(IL), V(IL), A(IL), Z(IL)
36:C
37:      100  ON = U*U
38:      SUMX = U*U
49
39:      SUMXY = U*U
40:      IL = 0
41:      DO 120  I= 1,IL
42:C      COMPUTE COMPOSITE VARIANCE, Z(I), FOR SYSTEM I .
43:      SUMBV = 0.0
44:      DO 110  L = 1,NPREDS
45:      IL = IL + 1
46:      SUMBV = SUMBV + B(IL)* V(IL)
47:      Z(I) = SSD(I)*Z + SUMBV
48:      ON = OH + SS(I)
49:      SUMX = SUMX + SS(I)* SM(I)
50:      SUMXY = SUMXY + SS(I)* (SS(I)*Z + SM(I)*Z )
51:C
52:      ONSSW = SUMXY - SUMX*Z/ON
53:      IJ = 1
54:      LASTI = NENS-1
55:C
56:      260  DO 220  I = 1, LASTI

```

```

57:      IN    = (I - 1) * NPREDS
58:      JN    = IN
59:      JFIRST = I + 1
60:      DO 220 J = JFIRST, NWS
61:      SUMBV = 0.0
62:      JN = JN + NPREDS
63:      DO 210 L = 1, NPREDS
64:          JL = L + IN
65:          JL = L + JN
66:      SUMBV = SUMBV + B(IL) * V(JL) + V(IL)
67:      A(IJ) = ((SN(1)*SN(J)) / (SN(1) + SN(J)))
68:          * (SSD(1)*SSD(J)*SUMBV - Z(1) - Z(J))
69:      1      -(SN(1) - SN(J)) ** 2 ) ) / (-ONSSQ)
70:      220   IJ = IJ + 1
71:      RETURN
72:      END

```

SOURCE LISTING FOR HIER-GRP SUBROUTINE GROUP

```

1:C-----
2:C
3:C      SUBROUTINE GROUP
4:C
5:C      THIS ROUTINE PERFORMS THE GROUPING PROCEDURE ACCORDING TO THE
6:C      VALUE OF LOPT. IF LOPT = 1 OR 6 IT IS ASSUMED THAT THE PREDICTOR
7:C      VARIABLES SATISFY THE PROPORTIONALITY ASSUMPTION. IF LOPT = 2, 3, 4,
8:C      OR 5, THE PROPORTIONALITY ASSUMPTION DOES NOT HAVE TO HOLD. THE
9:C      GROUPING OPTIONS ARE BRIEFLY DESCRIBED BELOW (ALSO SEE SECTION 3 OF
10:C      HG-MAIN).
11:C      If LOPT = 1, MAXIMIZED Z(KU) = CKU(KU+1)-CKU(KU)
12:C      If LOPT = 6, MINIMIZED Z(KU) = ABOVE

```

```

13:C IF IOPT = 2, MAXIMIZED Z(KU) = AVG. OF ALL NI(KU)*NJ(KU) VALUES OF
14:C ALL JU(KU) BETWEEN IJ(KU), JU(KU)
15:C IF IOPT = 4, MINIMIZED Z(KU) = ABOVE
16:C IF IOPT = 3, MAXIMIZED Z(KU) = MINIMUM OF ALL NI(KU)*(2-NU(KU))/2
17:C VALUES OF ALL JU(KU) WITHIN IJ(KU), JU(KU)
18:C IF IOPT = 5, MINIMIZED Z(KU) = MAXIMUM OF ABOVE
19:C
20:C
21:C      VARIABLE DESCRIPTIONS
22:C
23:C      NEQS = NUMBER OF SYSTEMS BEFORE GROUPING (AT STAGE) *
24:C      SN(1) = NUMBER OF CASES IN SYSTEM 1 AT NEWS STAGE *
25:C      SN( ) VECTOR NOT USED IF IOPT = 3 OR 5
26:C      AI(J) = Z(NEQS-1) VALUE HYPOTHESIZING IJ(NEQS-1) = 1, JU(NEQS-1)=J.
27:C      Z(KU) = DECISION VALUE CAUSING THE SELECTION OF IJ(KU), JU(KU)
28:C      IJ(KU) = IDENT OF SYSTEM ABSORBING SYSTEM JU(KU) AT KU STAGE.
29:C      JU(KU) = IDENT OF SYSTEM GROUPING WITH SYSTEM IJ(KU) AT KU STAGE
30:C      KU = NUMBER OF SYSTEMS AT KU STAGE, RANGES FROM NEQS-1 TO 1.
31:C      ORU(KU) = KU SYSTEM RSQ
32:C      NU(KU) = NUMBER OF MEMBERS IN SYSTEM IJ(KU) AT KU STAGE.
33:C      NI(KU) = NUMBER OF MEMBERS IN SYSTEM IJ(KU) AT KU +1 STAGE *
34:C      NJ(KU) = NUMBER OF MEMBERS IN SYSTEM JU(KU) AT KU +1 STAGE *
35:C      IJ = N*I-N-(I*I+1)/2 +J, WHERE I IS ALWAYS LESS THAN J.
36:C
37:C
38:C      SUBROUTINE GROUP
39:      1 ( IJPT, NEQS, SN, A, Z, IJ, JU )
40:C      INPUT   *   *   *
41:C      OUTPUT  ---UNCHANGE--- C   *   *
42:C
43:      DIMENSION SN(1), A(1), Z(1), IJ(1), JU(1)
44:C
45:      DO 120 K = 1, NEQS
46:      Z(K) = SN(K)
47:      120 IJ(K) = K
48:      NGRPS = NEQS

```

49:c

50: 200 BEST = -1.0E35

51: IF ( IOPT \* LE \* 3 ) GO TO 215

52: BEST = 1.0E35

53: 215 DO 280 KI = 1, NGRPS

54: 1 = IU(KI)

55: DO 280 KJ = KI, NGRPS

56: J = IU(KJ)

57: IF ( I = J) 220, 280, 230

58: 220 IJ = NEQS\*I-NEQS - (I\*I+1)/2 + J

59: 60 T0 240

60: 230 IJ = NEQS\*J-NEQS -(J\*J+J)/2 + 1

61: 240 IF ( IOPT \* GT \* 3 ) GO TO 260

62: IF ( BEST - A(IJ) ) 270, 270, 280

63: 260 IF ( BEST - A(IJ) ) 260, 270, 270

64: 270 BEST = A(IJ)

65: IBEST = KI

66: JBEST = KJ

67: 280 CONTINUE

68: C

69: 300 IF ( IU(JBEST) - IU(JBEST) ) 320, 9000, 310

70: 310 KBEST = IBEST

71: IBEST = JBEST

72: JBEST = KBEST

73: SNI = Z(JBEST)

74: 1 = IU(JBEST)

75: SNI = Z(JBEST)

76: J = IU(JBEST)

77: Z(JBEST) = SNI + 2NJ

78: Z(JBEST) = Z(NGRPS)

79: IU(JBEST) = IU(NGRPS)

80: Z(NGRPS) = BEST

81: IU(NGRPS) = 1

82: JU(NGRPS) = J

83: NGRPS = NGRPS - 1

84:c

```

85:      DO   460      KK = 1, NGRPS
86:      K = IU(KK)
87:      SINK = Z(KK)
88:      IF (K - 1) 410, 460, 420
89:      KI = NEQS•K - NEQS = (K•K+K)/2 + 1
90:      KJ = KI - 1 + J
91:      GO TO 450
92:      KI = NEQS•I - NEQS = (I•I+I)/2 + K
93:      IF ( K •GT. J ) GO TO 440
94:      NJ = NEQS•K - NEQS = (K•K+K)/2 + J
95:      GO TO 450
96:      NJ = NEQS•J - NEQS = (J•J+J)/2 + K
97:      GO TO ( 451, 452, 453, 454, 455, 456 ), IOPT
98: C
99: 451      A(KI) = ( (SNI+SINK)*A(KI) + (SNJ+SNK)*A(KJ) - SNK•BEST )
100:     / (SNI + SNJ + SNK )
101:     GO TO 460
102: 452      A(KI) = (SNI•A(KI) + SNJ•A(KJ) ) / (SNI + SNJ)
53 103:     GO TO 460
104: 453      A(KI) = AMIN( A(KI), A(KJ) )
105:     GO TO 460
106: 455      A(KI) = AMAX( A(KI), A(KJ) )
107: 460      CONTINUE
108: C
109:      IF ( NGRPS •GT. 1 ) GO TO 200
110: C
111:      DO   610      K = 2, NEQS
112:      Z(K-1) = Z(K)
113:      IU(K-1) = IU(K)
114:      JU(K-1) = JU(K)
115:      Z(NEQS) = 0.0
116:      IU(NEQS) = 0
117:      JU(NEQS) = 0
118:      RETURN
119: C
120: 9000 PRINT 800

```

```
121: BUU FORMAT
122: 110 SUBROUTINE GROUP. ALL VALUES OF OVERLAP VECTOR EXCEED 1.E35*)
123: STOP
124: END
```

SOURCE LISTING FOR HIER-GRP SUBROUTINE STAGE

```
1:C
2:C
3:C
4:C
5:C THIS ROUTINE COMPUTES THE STAGE VALUES FOR PRINTING IN SUBROUTINE
6:C PRINTG. IT IS ASSUMED THAT THE PROPORTIONALITY ASSUMPTION IS SATISIFIED
7:C BY THE PREDICTOR VARIABLES, OTHERWISE THE FORMULAS USED TO COMPUTE THE
8:C VALUES OF THE SRU AND ORU VECTORS ARE NOT VALID.
9:C
10:C
11:C VARIABLE DESCRIPTIONS
12:C
13:C NEMS = NUMBER OF SYSTEMS.
14:C UPREDS = NUMBER OF PREDICTOR VARIABLES.
15:C SJ(I) = NUMBER OF CASES OBSERVATIONS.
16:C SM(I) = CRITERION MEAN
17:C SD(I) = CRITERION STANDARD DEVIATION
18:C B(IL) = STANDARD REGRESSION WEIGHT FOR PREDICTOR L
19:C WHERE IL = UPREDS*(I-1) + L
20:C V(IL) = CRITERION CORRELATION WITH PREDICTOR L IN SYSTEM I.
21:C WHERE IL = UPREDS*(I-1) + L
22:C IUT(KU) = IDENT OF SYSTEM ABSORBING SYSTEM JUKU AT STAGE KU.
```

```

23:C JU(KU) = IDENT OF SYSTEM GROUPING WITH I(U(KU)) AT STAGE KU *
24:C SK(11) = R SQUARED
25:C ORU(KU) = R SQUARED FOR N SYSTEMS
26:C SKU(KU) = R SQUARED OF COMBINED SYSTEMS I(U(KU)), J(U(KU)) AT STAGE KU *
27:C KP(11) = IDENT OF SYSTEM FOLLOWING SYSTEM 1 IN FINAL
28:C HIERARCHICAL ORDERING.
29:C KU = NUMBER OF SYSTEMS AT KU STAGE, RANGES FROM NEQS-1 TO 1.
30:C
31:C
32:
33:   1  ( NEQS,INPRES,SN,SM,SSD,V,IU,JU,SK,ORU,SRU, KP )
34:C INPUT   *   *   *   *   *   *   *
35:C OUTPUT   *   *   *   *   *   *   *
36:C
37: DIMENSION SN(11),SSU(11),S(11),IU(11),SR(11),ORU(11),KP(11)
38:   1 ,S(11),V(11),JU(11),SRU(11)
39:C
40: ON   = 0.0
41: SUMX = 0.0
42: SUMXY = 0.0
43: SUMR = 0.0
44: IL   = 0
45: DO 120 1 = 1, NEQS
46:   KP(11) = 0
47:   ON   = ON + SN(1)
48:   SUMX = SUMX + SN(1) * SM(1)
49:   SUMXY = SUMXY + SN(1) * ( SSU(1)*2 + SM(1)*2 )
50:   SUMRY = 0.0
51: DO 110 L = 1, MPREDS
52:   IL   = IL + 1
53:   SUMsV = SUMsV + D(IL) * V(IL)
54:   SK(11) = SUMBV
55:   120  SUMR = SUMR + SN(1) * ( SSU(1)*2 + SUMBV + SM(1)*2 )
56:   0MSQ = ( SUMX / ON ) * 2
57:   OSqY = SUMY / ON - ONsq
58:   ORU(NEQS) = ( SUMR / ON - 0MSQ ) / OSS

```

```

59: SRU(NEQS) = U•IJ
60: KU = NEQS - 1
61: C
62: 200
63: SUMRIJ = 0•U
64: IJ = JU(KU)
65: JK = TU(KU)
66: C
67: 300
68: K54I = SR(IK)
69: IF ( KP(IK) • LU. 0 ) GO TO 400
70: KI = KU + 1
71: 320
72: CONTINUE
73: 330
74: C
75: 400
76: SUMI = 0•U
77: SUMIS = 0•U
78: I = IK
79: SNI = SNI + SN(IL)
80: SUMI = SUMI + SN(IL) * SM(IL)
81: SUMIS = SUMIS + SN(IL) * (SSU(IL) * (SM(IL)*2 + SM(IL)*2 )
82: J = JK
83: 420
84: SUMEV = 0•U
85: IS = NPREDS * (I-1)
86: JN = NPREDS * (J-1)
87: DU 430 L= 1,NPREDS
88: IL = L + IN
89: 430
90: SUMBV = SUEBV + S(IL) * V(JL)
91: SUMRIJ = SUMRIJ + SN(IL) * SSD(IL) * SSD(J) * SUMBV
92: J = KP(J)
93: IF ( KP(IL) • LU. 0 ) GO TO 420
94: I = KP(IL)

```

```

45:      GO TO 410
46:c
47: 500  IF ( JK •EQ• JU(KU) ) GO TO 600
48:  SNJ = SNI
49:  SUMJ = SUMI
50:  SUMJSN = SUMISN
51:  RSQJ = RSQI
52:  IK = IUIKU
53:  JK = JUJKU
54:  GO TO 300
55:c
56:  600  KP(1) = JK
57:  SUMKI = RSQI • ( SNI • SUMISQ - SUMI••2 )
58:  SUMKJ = RSQJ • ( SNJ • SUMJSQ - SUMJ••2 )
59:  SUMR = SUHR + ( SUMKJ + 2•0 • SUMI • SUMJ
60:  1 - SNJ • ( SUMRI + SUMI••2 ) / SNI
61:  2 - SJI • ( SUHRJ + SUMJSQ••2 ) / SNI + SNJ)
62:  ORU(KU) = ( SUMR / 0.1 - 0.05 ) / 0.59
63:  SRU(KU) = ( SUMKI + SURI + SUMRJ /
64:  1 - ( SNI+SNJ ) • ( SUMISQ + SUMJSQ ) - ( SUMI+SUMJ )••2 )
65:  KU = KU - 1
66:  IF ( KU • GT. 0 ) GO TO 200
67:c
68:  RETURN
69:c
70:  900 PRINT 800
71:  800 FORMAT (• SUBROUTINE STAGE • ASSURING SYSTEM NOT FOUND •)
72:  STOP
73:  END

```

SOURCE LISTING FOR HIER-GRP SUBROUTINE PRINTS

```

1:C
2:C
3:C
4:C
5:C      THIS PROGRAM PRINTS THE GROUPING REPORT. OUTPUT INCLUDES A LIST
6:C      OF THE ORIGINAL SYSTEMS TOGETHER WITH THEIR RSQ'S. AT EACH STEP OF THE
7:C      GROUPING PROCEDURE, THE SYSTEMS GROUPED, THE DECISION VALUE AND THE F-
8:C      STATISTIC ARE PRINTED. IN ADDITION A SUMMARY ROSTER GIVING ALL THE
9:C      SYSTEMS AND THEIR HEADERS IS PRINTED. THE INFORMATION PRINTED BY THIS
10:C     SUBROUTINE DEPENDS HEAVILY ON THE PROPORTIONALITY ASSUMPTION FOR ITS
11:C     CORRECTNESS. IN PARTICULAR, ALL SYSTEMS RSQ'S EXCEPT THE NEQS SYSTEMS
12:C     RSQ AND THE NEQS INITIAL SYSTEMS RSQ'S ARE INCORRECT IF THE PROPORTION-
13:C     ALITY ASSUMPTION DOES NOT HOLD. THIS IN TURN INVALIDATES THE F-TESTS,
14:C     AND THE ORDER OF GROUPING IF GROUPING OPTIONS 1 OR 6 ARE USED.
15:C
16:C
17:C
18:C
19:C      IOPT = GROUPING OPTION NO. (1 THRU 6) USED TO PRINT OBJECTIVE
20:C      FUNCTION DESCRIPTION AFTER REPORT HEAD.
21:C      NEQS = NUMBER OF SYSTEMS BEFORE GROUPING (AT NEQS STAGE).
22:C      NPRED = NUMBER OF PREDICTORS IN EACH SYSTEM.
23:C      SN(I) = NUMBER OF CASES IN SYSTEM I      AT NEQS STAGE .
24:C      R(I,L) = BETA WEIGHT FOR PREDICTOR L    IN SYSTEM I.
25:C      WHERE RL = NPRED*(I-1) + L.
26:C      PM(L) = MEAN FOR PREDICTOR L.
27:C      PSD(L) = STANDARD DEVIATION FOR PREDICTOR L.
28:C      IU (KU) = IDENT OF SYSTEM ABSORBING SYSTEM JU(KU) AT KU STAGE .
29:C      JU (KU) = IDENT OF SYSTEM GROUPING WITH IU(KU) AT KU STAGE .
30:C      Z (KU) = DECISION VALUE CAUSING THE SELECTION OF IU(KU), JU(KU).
31:C      SR (I) = R SQUARED OF SYSTEM I .
32:C      GRU (KU) = R SQUARED FOR KU SYSTEMS      AT KU STAGE .
33:C      SPU (KU) = R SQUARED OF COMBINED SYSTEMS IU(KU), JU(KU)      AT KU STAGE .
34:C      KS (J) = KU WHERE J = JU(KU)

```

```

35:C KP(1) = IDENT OF SYSTEM FOLLOWING SYSTEM 1 IN FINAL
36:C HIERARCHICAL ORDERING.
37:C K0() = SYSTEM IDENT'S IN FINAL HIERARCHICAL ORDERING.
38:C A() = WORK AREA
39:C BE() = WORK AREA, USED TO STORE BETA WEIGHTS FOR STEPWISE PRINTING
40:C KU = NUMBER OF SYSTEMS AT KU STAGE, RANGES FROM NEQS-1 TO 1.
41:C RCONST = REGRESSION CONSTANT.
42:C
43:C
44:
45: I(TOPT,NEQS,NPREQS,SN,SM,SSD,B,PH,PSD,Z,TU,JU,SR,ORU,SRU,
46:C 1n * * * * * * * * * * * * * * * * * * * * * *
47:C OUT ----- UNCHANGED -----
48:C
49: 2 KS,KP,K0,A,BE)
50:C OUT * * * * W
51: DIMENSION SN(I), SM(I), SSD(I), Z(I), JU(I), ORU(I), KST(I), K0(I),
52: 1 TU(I), SR(I), SKU(I), KP(I), A(I), BE(I)
53: 2 ,RU(I), PH(I), PSD(I)
54:C
55: 1 FORMAT( / 45H MAXIMIZED DECISION VALUE = ORU(KU+1)-ORU(KU) )
56:C
57: 2 FORMAT(/125H MAXIMIZED DECISION VALUE = AVERAGE OF ALL SNI(KU)*SNJ
58: 1(KU) VALUES OF ORU(NEQS)-ORU(NEQS-1) BETWEEN SYSTEMS TU(KU) AND JU
59: 1(KU) )
60:C
61: 3 FORMAT( /123H MINIMIZED DECISION VALUE = MINIMUM OF ALL (NU(KU)*
62: 12-NU(KU))/2 VALUES OF ORU(NEQS)-ORU(NEQS-1) WITHIN NEW SYSTEM TU(
63: 2(KU)) )
64:C
65: 4 FORMAT(/125H MINIMIZED DECISION VALUE = AVERAGE OF ALL SNI(KU)*SNJ
66: 1(KU) VALUES OF ORU(NEQS)-ORU(NEQS-1) BETWEEN SYSTEMS TU(KU) AND JU
67: 2(KU) )
68:C
69: 5 FORMAT( /123H MINIMIZED DECISION VALUE = MAXIMUM OF ALL (NU(KU)*
70: 12-NU(KU))/2 VALUES OF ORU(NEQS)-ORU(NEQS-1) WITHIN NEW SYSTEM TU(KU)

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```

11:      20).*
12:C
13:      6 FORMAT ( / 45H MINIMIZED DECISION VALUE = CRU(KU+1)-ORU(KU) )
14:C
15:      7 FORMAT( 8X 116H
16:          1KU) AT KU+1 STAGE.
17:C
18:      b FORMAT( 8X 116H
19:          1 TU(KU) AT KU STAGE.
20:C
21:      9 FORMAT( 8X 116H
22:          1M JU(KU) AT KU STAGE.
23:C
24:      1G FORMAT( 7X 52H
25:          1   /8X 116H
26:          2KU STAGE.
27:C
28:      11 FORMAT ( / 43H
29:          1   // 15, 22H INITIAL SYSTEMS RSQ = F7.4
30:          2   // 131 12H SYS PSQ 1)
31:C
32:      12 FORMAT(1X,126(1H*))* STAGE = * 13, * OVERALL RSQ = * F7.4,12X,* * /
33:      146X,* * * 14X,* F-TEST FOR THE EQUALITY OF REGRESSION PARAMETERS*/*
34:      127X,* SYSTEMS* 12X,* * * /22X,* GROUPING THIS STAGE * * ,
35:      2* FOR SYS*'S COMBINED UP AT THIS STAGE*,*
36:      3* FOR SYS*'S COMBINED UP TO THIS STAGE* /22X,19(1H-),* * ,
37:      432(1H-),3X,35(1H-)/20X,* SYS NO.    NO.    * * ,10X,
38:      5* CHANGE FROM* 25X,* CHANGE FROM* 19X,* IDENT MEMBERS CASES RSQ*,
39:      6* * ,1UX,13,* SYSTEMS RESIDUAL*,13X,13,* SYSTEMS RESIDUAL* /
40:      719X,1.3,17,19,F6.4,* RSQ* * 12.4,F13.4,8X,* RSQ*,F12.4,F13.4/
41:      519X,1.3,17,19,F6.4,* 2F*,112,113,9X,* DF*,112,113/46X,* * /*
42:      919X,* DECISION VALUE=*,F10.4,* * FSTAT=*,F8.2,* SIG LVL=*,F8.4,
43:      17X,* FSTAT=*,FD.2,* SIG LVL=*,F6.4//36X,* SYSTEMS SUMMARY RSQ* /*
44:      236X,22(1H-)/14X,* STAGL SYS NO.    NO.    * * SYS* /
45:      3   14X,* LOAN LOSSES MEMBERS RSQ CASES * IDENT * ,
46:      4* IDENTIFICATION OF OTHER MEMBERS* ,

```

```

107:C
108: 13 FORMAT(14X,13,F7.4,15,F7.4,19,*,*,14,1X,2014,9(/40X,*,*,5X,2014))
109:C
110: 14 FORMAT(46X,*,*/46X,*,*,5X,*,SETA WEIGHTS FOR THE NEW SYSTEM,
111: 1 14,8X,*NEW SYS CRITERION MEAN =*,F9.4/46X,*,*,5X,35(1H-),
112: 2 8X, *NEW SYS CRITERION SD =*,F9.4/46X,*,*,1018)
113:C
114: 15 FORMAT(46X,*,*,1UF4.4, 2U(14X,*,*,*,1UF6.4) )
115:C
116: 16 FORMAT(46X,*,*/46X,*,* RAW SCORE WEIGHTS FOR THE NEW SYSTEM,
117: 1 14,10X,*REGRESSION CONSTANT =*,F9.4/46X,*,*, ,
118: 2 40(1H-)46X,*,*,5116)
119:C
120: 117 FORMAT(46X,*,*,5F16.4, 20(/40X,*,*,5F16.4))
121:C
122: 17 FORMAT(14X,13, * SINGLE MEMBER SYSTEMS*,2014, 9(/39X,2014) )
123:C
124:911 FORMAT( 1U(15,F7.4) )
61
125:C
126: TOT =0.0
127: DO 50 I=1,NEQS
128: 50 TOT = TOT + SN(I)
129: NTOT =TOT
130:C
131: PRINT 90
132: 90 FORMAT( 1H2 / 1H1 )
133: PRINT 91
134: 91 FORMAT( /* 111. HIERARCHICAL GROUPING RESULTS */6X,29(1H-))
135:C
136: 115 GO TO ( 121, 122, 123, 124, 125, 126 ), IOPT
137:C
138: 121 PRINT 1
139: 139:          GO TO 160
140: 122 PRINT 2
141: 141:          PRINT 7
142: 142:          PRINT 9

```

```

143: 123 PRINT 3
144: 123 PRINT 3
145: 123 PRINT 3
146: 123 PRINT 3
147: 124 PRINT 4
148: 124 PRINT 4
149: 124 PRINT 4
150: 125 PRINT 5
151: 125 PRINT 5
152: 125 PRINT 5
153: 126 PRINT 6
154: 126 PRINT 6
155: 126 PRINT 6
156: 126 PRINT 6
157:C
158: 166 PRINT 10
159:C
62 160: 200 DO 210 K = 1, NEW$, KS(K) = K+1
161: KP(K) = J
162: A(K) = SR(K)
163: 210 KS(NEW$) = 0
164: PRINT 11, 10PT, NEWS, ORU(NEWS)
165: PRINT 911, 1, SK11, 1 = 1, NEWS )
166: KS = NEWS - 1
167: DF2X = NOT - NEWS * (NEWS + 1)
168: IF ( ORU(NEWS) * GT. 1. ) ORU(NEWS) = 1.
169: RES = 1. - ORU(NEWS)
170: C
171:C
172:C
173: 300 KUP1 = KU + 1
174: IF ( ORU(KU) * GT. ORU(KUP1) ) ORU(KU) = ORU(KUP1)
175: DI = ORU(KUP1) - ORU(KU)
176: ODI = ORU(NEWS) - ORU(KU)
177: REST = 1. - ORU(KUP1)
178: DFIX = (NEWS - KU) * (PREDS + 1)

```

```

179:      ID = I0(KU)
180:      JD = J0(KU)
181:      1 = ID
182:      310  IF (KS(1)*EJ*JD) GO TO 330
183:      1 = KS(1)
184:      90  TO 310
185:      330  KS(1) = KS(JD)
186:      KS(JD) = KU
187:      1 = ID
188:      N1 = 1
189:      SII = SH(1)
190:      340  IP (KP(1)*EJ*JD) GO TO 360
191:      1 = KP(1)
192:      NI = NI + 1
193:      SII = SII + SI(1)
194:      60  TO 340
195:      360  KP(1) = JD
196:      J = JD
197:      NJ = 1
198:      370  IP (KP(J)*EJ*JD) GO TO 390
199:      SJ = SJ(J)
200:      J = KP(J)
201:      NJ = NJ + 1
202:      SNJ = SNJ + SJ(J)
203:      60  TO 370
204:      390  NSI = SNI
205:      NSJ = SJ(J)
206:      DFI = IPRED$+1
207:      DF2 = NTOT - KP1*(IPRED$+1)
208:      F = (DI*DF2)/(DF1*NTOT)
209:      F1 = (DI*DF2X)/(DF1*XRES)
210:      IF1=DF1
211:      IF2=DF2
212:      IF1X=DF1X
213:      IF2X=DF2X
214:      CALL PLEVEL( DFI, DF2, F, PROG)

```

```

215:      CALL PLEVEL( OFIX, OFIX, F1, PROB1)
216:      PRINT 12,
217:      1      KU, ORUKU), KUPI, NEUS, ID, NI, NSI, A(ID), DI, RES1, ODI, RES,
218:      2      JD, IJ, NSJ, AJU),
219:      3      IF1, IF2, IFIX, IF2X,
220:      A(ID) = SRU(KU)
221:      MIT = NEQS - KU + 2
222:      KOW = MIT - 1
223:      I = 1
224: C
225: 400      IF ( KPI(I) * GT. U ) GO TO 410
226:      KOW = KOW + 1
227:      KU(KOW) = I
228:      GO TO 440
229: 410      N1 = 1
230:      NU(N1) = 1
231:      SN1 = SN(I)
232:      SLUSS = U*U
233:      IGRUP = NEUS
234:      J = KF(I)
235:      IF ( I * NE * ID ) GO TO 420
236: C          INITIALIZE BETA WEIGHT COMPUTATION
237:      SUMX = SN(I) * SM(I)
238:      SUMX2 = SN(I) * (SSD(I)*2 + SN(I)*2)
239:      CONST = SN(I) * SSU(I)
240:      IK = (I-1) * NPIDS
241:      DO 412 K = 1, NPIDS
242:      IK = JK + 1
243: 412      RE(K) = B(IK) * CG,SI
244: 420      KUJ = KS(J)
245:      SLUSS = SLUSS + (KU(KUJ + 1) - ORU(KUJ))
246:      IF ( KUJ * GT. IGRUP ) GO TO 426
247:      IGRUP = KUJ
248: 426      NI = I + 1
249:      SN1 = SN1 + SN(J)
250:      K0(I) = J

```

```

251: IF ( I .LT. 1D ) GO TO 429
      ACCUMULATE PTA WEIGHTS FOR THIS STAGE
252: C
253: SUMX = SUMX + S1(J) * S1(J)
254: SUMX2 = SUMX2 + S1(J) * ( S50(J)**2 + SM(J)**2 )
255: CONST = S1(J) * S50(J)
256: IK = (J-1) * NPREDS
257: DO 428 K = 1,NPREDS
258: 426 BE(K) = RE(K) + B(IK)*CONST
259: 429 J = KP(J)
260: C
261: IF( J .GT. 0 ) GO TO 420
262: 430 RSI = SNI
263: PRINT 13, LKUP, SLOSS,
      NI, A(1), NSI, ( R0(K), K = 1, NI )
264: 1
265: IF ( I .LT. 1D ) GO TO 440
266: C
      COMPUTE AND PRINT SYSTEM MEAN, SD, BETA WEIGHTS
267: SUMX = SUMX / SNI
268: SUMX2 = SUMX2 / SNI - SUMX**2
269: CONST = 1. / (SNI * SUMX2)
270: DO 435 K = 1,NPREDS
271: 435 BE(K) = RE(K) * CONST
272: MIN = MIN0 ( 10, NPREDS )
273: PRINT 14, 1D, SUMX, SUMX2, ( K, K = 1, MIN )
274: PRINT 15, ( RE(K), K = 1, NPREDS )
275: C
      COMPUTE AND PRINT RAW SCORE WEIGHTS AND REGRESSION CONSTANT
276: SUM = 0
277: DO 437 K = 1,NPREDS
278: 437 BE(K) = RE(K) * SUMX2 / PSD(K)
279: 437 SUM = SUM + BE(K) * PSD(K)
280: RCONST = SUMX - SUM
281: MIN = MIN0 ( 5,NPREDS )
282: PRINT 16, 1D, RCONST, ( K, K = 1, MIN )
283: PRINT 117, ( RE(K), K = 1, NPREDS )
284: 440 I = KS(1)
285: IF ( I .GT. 0 ) GO TO 430
266: C

```

```

287:      IF ( KOW • L1 • MIT ) GO 10 520
288:      510      NKOW = KOW - MIT + 1
289:      PRINT 17
290:      1      * NKOW, ( KOK ), K = MIT, NOW )
291:      520      KU = KU - 1
292:      1F ( KU • GT • 0 ) GO TO 300
293: C
294:      PRINT 421
295:      421      FORMAT ( / 4H*,***** )
296:      1      58H***** / ***** / ****
297:      2      27H***** / ***** / ****
298:      RETURN
299: C
300: END

```

#### SOURCE LISTING FOR HIER-GRP SOURCELINE LEVEL

```

1: C PLEVEL 31 OCTOBER 1966
2: SUBROUTINE PLEVEL(X,DF1,ADDF2,F,P)
3: DIMENSION Y(6),ARG(3),SIGMA(3)
4: SASHLN(Z,X)=(X+.5)*ALOG(X)-X+.918938534+(.833333333E-1-L*(.2777777
5: 178E-2-Z*(.793651794E-5-Z*(.595238095E-3-Z*.841750842E-3)))/X
6: DF1=X*DF1
7: DF2=X*DF2
8: IF(DF1+.5T+.100D+DF1)=1000.
9: IF((DF2+.5T+.100D)+DF2)=1000.
10: IF((DF1*L1+1.0R+DF2*L2+1.0R+F*L3+0.0)GO TO 14
11: IF(ARG(1).EQ.0.01 AND ARG(2).EQ.0.01 AND ARG(3).EQ.0.01)GO TO 7
12: ARG(1)=DF1
13: ARG(2)=DF2
14: ARG(3)=DF1+DF2

```

```

15:      GO TO 6   I=1,3
16:      IF (ARG(1)*1.4*1.0) GO TO 5
17:      T = ARG(1)-2.
18:      J = ABSD(ARG(1),2.*)+1.
19:      GO TO (Z,J)
20:      U = (T-1.)*5
21:      GAMMA(1) = .572354943-T*.693147181
22:      IF ((J+L1)*2.) GO TO 25
23:      GO TO 3
24:      Z T = T*0.5
25:      GAMMA(1) = J*
26:      IF (T*LT*Z.) GO TO 6
27:      GO TO 4
28:      J Z = 1./((U*U))
29:      GAMMA(1) = GAMMA(1)-.5*SIN(U*(Z,U))
30:      U Z = 1./((T*T))
31:      GAMMA(1) = GAMMA(1)+.5*SIN(U*(Z,T))
32:      GO TO 6
33:      D GAMMA(1) = .572354943,
34:      CONTINUE
35:      C = GAMMA(1)+GAMMA(2)-GAMMA(3)-6*--.693147181
36:      Y(1) = .495462181
37:      Y(2) = -1.*203972804
38:      Y(3) = .587786654
39:      Y(4) = Y(2)
40:      Y(5) = Y(1)
41:      AX = DF2/(F*DF1+DF2)
42:      IF (AX*GT*.99999999) GO TO 136
43:      H = ATAN(SQRT(AX/(1.-AX))/765.)
44:      IF (H*LE*.130899694E-1) GO TO 6
45:      H = .261794387E-1-H
46:      CN = DF1-1.
47:      CM = (DF2-1.)*0.
48:      P = -1.*d
49:      XA = -H
50:      GO TO 7

```

```

>1:      S CN = UF2-1.
>2:      CM = (UF1-1.)•.5
>3:      P = 0.
>4:      XH = H
>5:      Y IF (CN•NE•0..) GO TO 95
>6:      XX = Y(2)-C
>7:      IF (XX•LT•0..) GO TO 95
>8:      P=P+EXP(XX-69.)
>9:      Y(6) = -•5106256230
>10:     K = 0.
>11:     DO 13 I=1,10
>12:     S0 = F(11,11,11,11,11,11,11,11,11,11)
>13:     Y(6) = Y(2)
>14:     DO 13 J=1,5
>15:     A = X+A
>16:     CS = S14(X)
>17:     Z = Y(J)-C+CN•ALOG(CS)+CM•ALOG(1.-CS•XS)
>18:     IF (Z) 15,12,12
>19:     P = P+EXP(Z-69.)
>20:     13 CONTINUE
>21:     P = X4•P
>22:     IF (P•LT•1.0E-05) P = 0.
>23:     RETURN
>24:     130 P = 1.
>25:     130 P = 1.35
>26:     130 P = 1.
>27:     PRINTS,OF1,OF2,F
>28:     13 FORMAT(74I1)F1 = F5•J,DF2 = F5•J,5H,F = F10•5//'
>29:     148H P DOES NOT EXIST IF OF1 OR OF2 IS LESS THAN 1 /
>30:     249H OR IF F IS LESS THAN ZERO.
>31:     343H P FOR THIS PROJECT HAS BEEN ARBITRARILY SET /
>32:     448H EQUAL TO 1. AND A NORMAL RETURN HAS OCCURED.
>33:     50 TO 135
>34:     END

```