

*Personality Correlates
To
Performance Under Stress
In Simulated
Chemical Plant Emergencies*

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PERSONALITY CORRELATES TO PERFORMANCE UNDER STRESS
IN SIMULATED CHEMICAL PLANT EMERGENCIES

Submitted in partial fulfilment of the requirements
for the degree of Master of Arts of the Department for
Psychology at the University of Natal, Durban.

HANS PETER LEHMANN
DURBAN/SASOLBURG 1978

DECLARATION.

I declare this thesis, except where otherwise indicated in the text and acknowledgements, to be my own original work.

A handwritten signature in cursive script, appearing to read 'H.P. Lehmann', written in black ink.

H.P. LEHMANN

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2. AIM OF THE RESEARCH

When a fault develops in a chemical plant process, the plant operator must identify the fault rapidly and take immediate corrective action.

The interaction of process factors varies from fairly to highly complex (in extreme cases, this interaction is not yet fully understood even by chemical engineers) and consequences of faults can occur in chainreactions.

The operators task is to control all process parameters until the plant is brought back to normal conditions. Doing this, he is fully aware of the fact, that the consequences of wrong corrective action or failure to bring the plant under control can be grave in economic terms, extremely serious (lethal) in terms of the operator's hazards and potentially catastrophic.

Thus a considerable amount of stress can be built up, which is potentially interfering with the "cool" required to succeed in overcoming the emergency as quickly and efficiently as possible.

This project attempts to explore presumed correlations between personality factors and performance under stress.

Where such correlations exist in significant manifestation, their rank of magnitude was established and their predicitive value investigated.

3. PERSONALITY FACTORS AND COPING IN STRESSFUL SITUATIONS:
A BRIEF REVIEW WITH PARTICULAR REFERENCE TO THIS
INVESTIGATION.

In LAZARUS' model of coping with stress (Lazarus 1966) he found that personality factors influence the appraisal of threat as well as determining the coping process and the secondary appraisal of threat. He states:

"Three main classes of factors in the individual psychological structure influence threat appraisal. They do so by providing the capacity and dispositional bases for interpreting the significance of the stimulus cues.

The three classes are:

- 1 motivational characteristics of the individual;
- 2 belief systems concerning transactions with the environment;
- 3 intellectual resources, education, and sophistication".

These personality factors shape the inferences the individual can and will make about the stimulus configuration.

In discussing and specifying these three categories LAZARUS summarises that

- 1 The pattern and the strength of the motivation determines stress reactions. Situations will be appraised as threatening to the extent that they communicate with important goals;
- 2 Belief systems that deal with the confrontation of threat, e.g. beliefs that the environment was hostile and dangerous, could shape threat appraisal in any situation.

He argues that such general beliefs about the environment and one's ability to deal with it probably underlie (chronic) anxiety, and that thus anxiety is the original dispositional trait, if interpreted as the disposition to be threatened in a wide (or narrow) variety of situations;

- 2 Intellectual resources play an important part in influencing threat appraisal, but they are nondirectional with respect to whether they yield greater or less threat.

The conclusion is drawn that although the influence is non-directional, it does appear that lack of sophistication should increase the prospects of incorrect evaluations of the situation.

How personality traits may affect coping with stress can be divided into three main categories. LAZARUS states:

"The factors within the psychological structure that we shall discuss as influencing coping include some which do so by affecting secondary appraisal, some which do so directly because they refer to capacities, and some which are treated as dispositional variables which are neutral with respect to this issue of how they work.

We shall consider four classes of factors within the psychological structure that influence coping:

- 1 pattern of motivation,
- 2 ego resources,
- 3 defensive dispositions,
- 4 general beliefs about the environment and one's resources."

These four classes can be summarized with respect to the direction of influence as follows:

- Motivation determines which kind of action poses additional threat via a psycho-economic decision within the secondary appraisal and therefore influences the coping process indirectly.
- Certain ego resources, especially the capacity for impulse control and ego strength, influence coping directly rather than via appraisal.
- Those defensive dispositions (i.e. personality traits comprising the tendency to use one or another type of defense when the individual is threatened) that influence the reaction to threat in a particular way, are:
 - tendency to cope or avoid
 - the trait of "defensiveness" (tendency to deny weakness in oneself)
 - perceptual defence tendencies (e.g. sensitisation vs. repression)
 - consistency and/or generality of defence
- Beliefs about one's resources influence the individual's choice how to cope via secondary appraisal. LAZARUS however emphasises that this cognitive determination does not imply that the coping process is conscious, rational, or adaptive; any irrationality does not come from the interposition of emotion in thought, but is a reflection of the particular cognitive structure of the individual which disposes him to interpret situations in particular ways.

LAZARUS' theorem of interaction between personality factors and the coping process may be summarized as follows:

CLASSES OF FACTORS	INFLUENCE ON STRESS REACTION
1. <u>Motivation</u>	
Motivation pattern	primary threat appraisal
Motive strength	determines coping via secondary appraisal
1 <u>Personality traits</u>	
Anxiety (as the disposition to be threatened by situations consequential to and/or implied by belief systems concerning transactions with the environment)	primary threat appraisal
Defense dispositions	coping process via secondary threat appraisal
Ego resources	coping process and coping strategy
2 <u>Cognitive belief systems</u>	
These include the cognitive aspects of such belief systems that underlie anxiety and general beliefs about aspects of the environment (irrespective of their rationality) and concerning one's ability.	coping process via secondary threat appraisal.
4 <u>Intellectual resources</u>	
(intelligence and sophistication)	primary and secondary threat appraisal, non-directional to the process.

Research since then (1966) seems to have accepted this theorem in general.

The research activities in relation to derivatives and components of the overall theory can be divided into two principle parts, according to the breadth of approach.

Two selected studies which are of relevance to this investigation may be taken as examples of the broader approach, i.e. where it was attempted to establish connections between a battery of personality factors, several stress reactions and a variety of performance criteria, and the aim was the establishment of syndromes with respect to predictive quality rather than an in-depth exploration of the interaction of single variables.

JENNINGS, KREUZ & ROSE (1974) related military performance to measures of personality and stress. The design they used was a parallel one, i.e. personality measures and stress measures were equally and unilaterally valued as one-dimensional variables for the prediction of performance as the dependent variable. Ss were 69 candidates of an Officer Candidate School (OCS). The hypothesis put under test was that certain personality factors and styles of coping with stress should be predictor variables for the performance of candidates/officers during and after OCS.

Their findings distinguished between performance as officers (after graduation from OCS) and performance during the course (expressed as class-standing). Certain personality-factors and "maturational" variables seemed to correlate favourable with performance as officers, while no relation was found between coping with stress and the performance on duty as officers.

The data presented indicates that there may be a cross-correlation between personality traits, the occurrence of stress during the OCS and later officer-performance.

(The correlation-coefficients found, however, were so small and mostly insignificant, that, in a strict statistical sence, no conclusion whatsoever should have been drawn).

The second study, that uses a similar approach, is the investigation of J. DANIELS (1973) about personality traits and the adaption to "psychological" stress. *)

DANIELS supposes that previous research has implied that through the adaption to stress conditions, certain psychological and physiological indicators in the individual's stress reaction show a tendency to change and then differ from those observed under short-term stress. The objective of his study was to examine how and if personality traits, which are considered to be relatively stable, become altered under the effect of adaption to longer-term stress-conditions. Scores for personality traits (measured by the 16PF) were obtained from 85 parachutists, 40 experienced and 45 novice (first-time) jumpers.

DANIELS found that experienced parachutists showed higher scores in Factor A (Affectothymia), in Factor B (Scholastic Mental Ability) in Factor C (Ego Strength), in Factor E (Dominance) in Factor G (Superego Strength), in Factor Q1 (Radicalism) and in Factor Q2 (Self-sufficiency).

Expressing these findings in Cattells second and third stratum factors (Cattell et al. 1970), this means that experienced jumpers are higher on "Strength" (3rd-stratum factor), lower on "Anxiety" (2nd - stratum factor) and higher on "Independence" (2nd stratum factor).

*) = Adopting the distinction between psychological and physiological stress after Lamb (1976), as mentioned below, one would rather classify the stress experienced in parachute jumping as physiological.

From data presented however, it is not conclusive that these differences are in any correlation to any adaptation to stress that may be present in experienced parachutists. What is shown is that experienced parachutists differ in certain personality traits from individuals who try parachute jumping for the first time. It is equally likely that there is a selection process: only individuals with the configuration of personality traits conducive to endure stress, as shown by Daniel, become experienced parachute jumpers, the remaining proportion of novices drop out due to lack of sufficient coping "ability". *)

However, for the question, which set of personality traits correlates favourably with criteria for performance under stress, the establishment of significant differences in personality factors is sufficient. It may only be noted here, that DANIEL's results do conform with other findings in this field.

Where recent research work has concentrated more on the exploration of specific personality traits and their connection with stress and the coping process, an increasing emphasis was put on distinguishing between transitory emotional states and relatively stable personality traits, the latter regarded as primary, or basic, dispositions of the individual. Of particular interest, especially regarding the subject of this investigation was the development of a state-trait-anxiety - theory, and much of this research has employed the STATE - TRAIT ANXIETY INVENTORY (STAI), developed by SPIELBERGER et al.

*) = Due to the authors personal experience as a moderately experienced parachutist the average drop-out rate of student jumpers is as high as 96% during the first ten jumps, i.e. only one in twentyfive novices reaches the eleventh jump (on average the first "free-fall").

This inventory, showing good internal consistency (STAI-Manual, SPIELBERGER et. al 1970), defines TRAIT - ANXIETY as the individual's proneness to experience anxiety in a particular situation as a relatively stable dispositional variable. STATE - ANXIETY is the corresponding transitory condition, characterised by feeling of apprehension, tension, heightened activity of the automatic nervous system as a direct consequence to stress-inducing stimuli.

With regard to the character of the stress-inducing stimulus, a distinction has been made between psychological (threats to self-esteem) and physiological stress (anticipation of physical harm). Given the same situational stress, persons with high A-trait tend to be higher in A-state, and individuals with low anxiety proneness - low A-trait - tend to be lower in A-state. While the absolute magnitude of A-state varies according to A-trait levels, the amount of change in A-state scores during stress does not differ. This was shown, in particular for the case of physical stressors (dental treatment) by LAMB (1976).

A factor-analytical investigation by KENDALL et al (1976) added to the distinction between A-trait and A-state. They found the A-trait scale unidimensional and quite homogeneous, with factor loadings of +.50 in one factor in 14 out of 20 items. This factor was labelled "Cognitive Anxiety", operational to the description of ego-threats (in the 14 items) rather than physical threat situations.

For A-state two factors were found, which seem to correspond to the descriptors used in the scale (and may represent artefacts of the scoring system).

To validate their findings, KENDALL et al (1976) further investigated the reliability of the two A-state factors in stressful situations, using both psychological and physiological stressors. The results supported the above outlined trait-state distinction and underlined LAMB's (1976) findings in general. Some interesting indications were also given: A-Trait scores showed a tendency to fluctuate with time and situation.

Using a physical stressor, it was found that both A-state factors reflected stress well, and in addition, that both low and high A-trait levels tend to converge to the same overall A-state score, which seems to indicate that the mere amount of stress should be considered an additional variable. This could be explained by the fact that, as the factor loadings above showed, A-trait is orientated to reflect the reaction to psychological threat rather than to physical harm - stimuli.

In concluding this brief review it may be summarized that LAZARUS's theorem of interaction of personality traits and coping has been supported on the whole. Especially the importance of the group of "ego resources" has been highlighted and the development of the State-Trait-Anxiety - model seems to open a new field for refined investigation of the influence of anxiety on the coping process.

This consolidation may be considered as the initial encouragement for the project on hand. In the investigation it is attempted to establish a directly predictive link between a syndrome of primary personality factors and anxiety traits and coping as a highly specialised operational performance under stress.

The next section will cover the underlying hypothesis and the experimental layout in more detail.

4. DESIGN OF THE EXPERIMENT

4.1. BASIC HYPOTHESIS

Following LAZARUS' theorem, personality factors, especially what he labelled 'ego resources' and anxiety play a vital role in the coping process, partly directly and partly via appraisals. Research since then has generally confirmed these facts and on the part of anxiety, the phenomenon has been researched in greater detail.

The basic hypothesis underlying this project can be formulated as follows:

Coping in a task-performance-situation under stress is (also) determined by and can be directly predicted from a quantifiable syndrome of measurable personality factors.

The detailed definitions of the main components of the hypothesis are:

- Coping is understood as achieving the objective of the task satisfactorily, i.e. performing well according to set standards.
- Stress is mainly understood as task-based stress, increased by eventual failure-stress.
- Direct prediction implies that it is possible to establish a predictive link between personality-structure and performance in the task without any reference to or dependence on the internal mechanics of the actual coping process.
- Quantification of the syndrome imposes the constraint that the syndrome's structure can be defined by computable variables.
- Measurability of the personality factors restricts the investigation to the use of recognised personality-tests.

It might be helpful for better understanding of the detailed derivatives of the basic hypothesis and the set of supporting assumptions, if a brief anticipatory description of the experimental layout is given here.

The task given was the control of a Chemical-Plant-Simulator. Task-based stressor was a series of 'emergencies', i.e. simulated breakdowns. The required performance was a combination of fault-finding and corrective action. The pattern of emergencies was such that complete success within the allowed time was impossible and furthermore, failure to cope with each emergency in time increased the complexity and magnitude of the overall problem.

The objective of the task was explicitly defined.

HACKMANN (1970) introduced a framework for the description of tasks and task performance in research on stress, which was found quite useful for detailing and operationalizing the basic hypothesis.

HACKMANN distinguishes between the objective stimuli input and the psychological interactions in assessing the effects of tasks. The objective task input is 'redefined' by the individual corresponding to his personality structure. Then a 'hypothesis' on how to cope is formed, followed by the 'process' of coping with the task which leads to a 'trial outcome'. 'System evaluation', i.e. task-inherent feed back, and 'personal evaluation' might result in adapting the 'hypothesis' until a 'final outcome' is produced.

Using the terms of HACKMANN's framework, the hypothesis, incorporating the essentials of the experimental layout, can be described as follows (an illustrating diagram is given in Fig. 1, p.18):

The task stimuli were the objectives of the task and the actual indications of the 'emergency'. The redefinition introduces task-based-stress, which influences the formulation of the hypothesis, i.e. the intended coping strategy. Also of influence is an increment in the task based stress, initially only determined by anxiety. The process of action taken results in indications of the system, which are interpreted in the personal evaluation of this 'trial outcome'. Depending on the outcome of this evaluation, the hypothesis is adapted or confirmed. If the personal evaluation was positive, no additional stress occurs. If the trial outcome was taken as unsatisfactory, additional stress in the form of task-based-stress (through the now increasing complexity of the task) and failure stress occurs, which results in a consequent redefinition of the task. This feed-back-loop continues until the time limit has expired and/or the experiment is terminated.

The similarity of 'redefinition' and 'personal evaluation' in HACKMANN's framework to LAZARUS'S primary and secondary threat appraisal is obvious. Following the summary of LAZARUS's theorem of interaction between personality factors and the coping mechanisms, ego resources would be of determining influence on 'hypothesis' and 'process', while anxiety would determine the stress increment in or due to its influence on the threat appraisal. (The layout being a continuous loop blurs the distinction between primary and secondary appraisal to some extent).

It was furthermore expected, that the very condition of increasing stress and increasing complexity of the task and the thus decreasing probability of solving the task problem consequential to coping failure in the outlined experimental design should amplify and 'purify' the influence of personality factors and lessen the importance of cognitive or intellectual resources considerably.

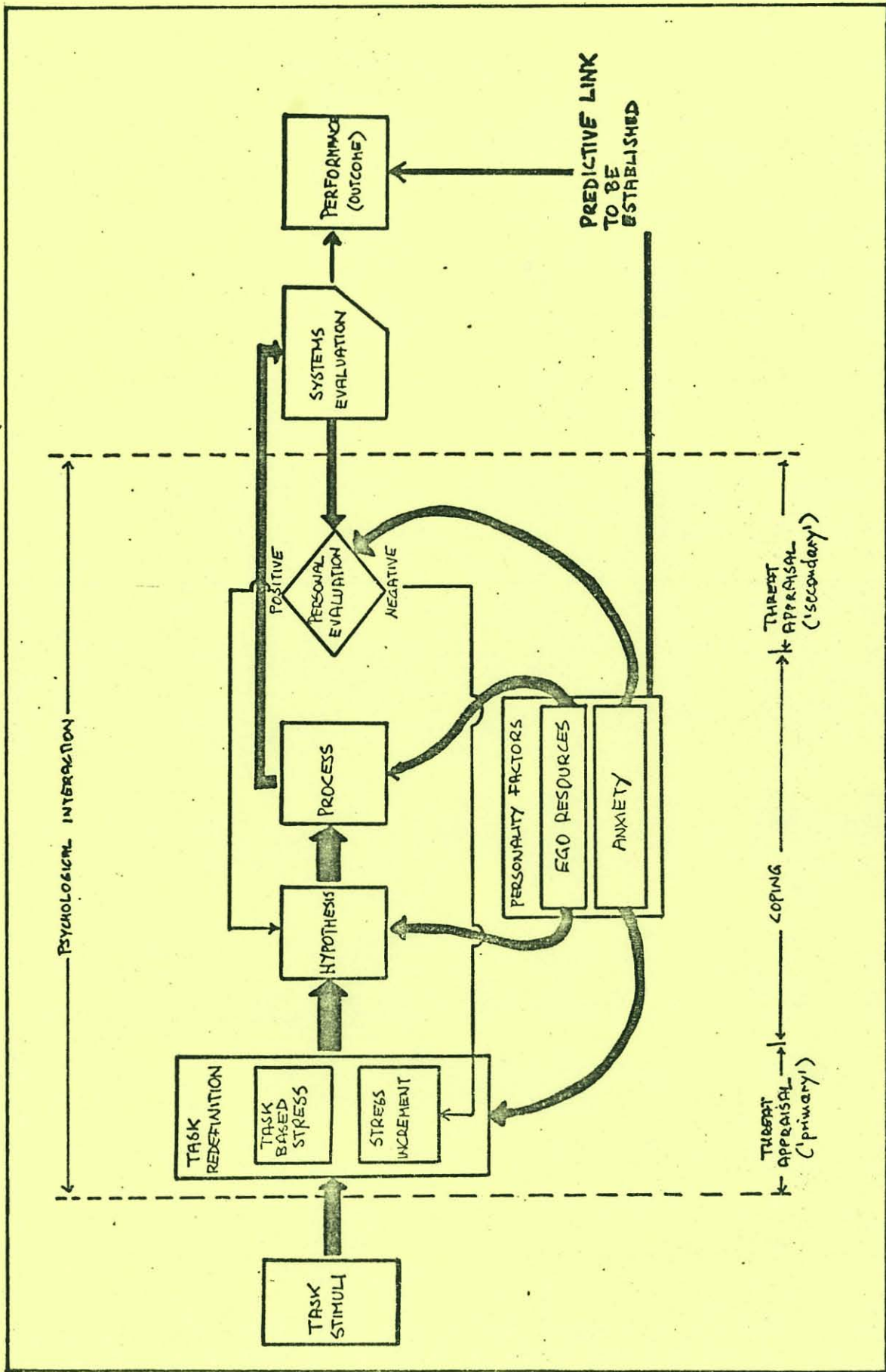


Fig. 1. Framework of the basic hypothesis (after HACKMANN)

4.2. EXPERIMENTAL LAYOUT

The participants in the experiment were 35 Plant-operators and 17 Plant-operator trainees, all white males and employees of AECI LTD, Umbogintwini Factory.

All 52 S's underwent a two weeks training, resp, re-training course at the factory training centre, consisting of one week theoretical training on technical equipment and second week of intensive training on operating a chemical-process-plant, using the CHEMICAL PLANT SIMULATOR.

During the first week the S's completed the personality and intelligence tests and the A-Trait questionnaire in group administration.

The actual experiment on the Simulator was carried out during the last day of the second week. Each participant was given the A-State questionnaire to complete immediately before his 'turn'.

The 'outcome' of the simulator task, i.e. the level of performance for every subject was assessed individually during task-completion by an experienced trainer. To increase the accuracy of the assessment, the task was broken down into a series of logical components and a rating score was issued at every step.

Thus a set of data, comprising a profile of personality factors, an estimate of the intelligence level, scalars for State- and Trait-Anxiety and a rating score of task-performance, complemented by records of age and previous experience as plant-operator were obtained for further statistical analysis.

Details of the psychological tests used, the simulator-task and the statistical analysis are to follow further below.

The 'field work' was carried out from July 1976 to early November 1976 at Umbogintwini.

4.2.1. Selection of psychological tests

The selection of the personality inventory to use followed in general the selection criteria listed below:

- The inventory had to be comprehensive. Since in many ways this project can be considered a 'pilot-study', and no particular cluster of personality factors was to be assumed influentially a priori, the inventory used was to have a fair degree of generality, covering as wide a spectrum as possible.
- The test would be established and proven. The small size of the sample used would result in quite broad confidence limits. To minimise an error-variance introduced by a test lacking the desired levels of reliability and validity, this constraint was imposed.
- Also to ensure a minimum quality level of the ensuing statistical analysis the measurements attained with the test should be clear and statistically sound.
- Considering possible further use, the test should be economical in use.

Evaluating several inventories according to these criterias, it was decided to use the CATELL SIXTEEN PERSONALITY FACTORS test.

This test has been standardised on vast populations (CATELL et al. 1970), is reasonably factorially valid (as recently shown by KARSON et al, 1974) and is in wide use for a wide variety of projects. (A scan through recent literature revealed the following articles: BOWMANN et al. 1974; BARTON et al. 1972, 1973, BACHTOLD et al. 1973; MYRICK et al. 1972; ASHTON et al. 1972, 1972)

The STATE TRAIT ANXIETY INVENTORY (SPIELBERGER et al. 1970) was used because it seems to be at present the best inventory to measure A-State and A-Trait with an optimum in accuracy of the score-scalar and economy of use.

It was felt that some measure of intelligence levels was needed. Since a reasonably reliable unifactorial measurement with good discrimination in the medium to lower intelligence range was felt sufficient, RAVEN's PROGRESSIVE MATRICES TEST was chosen.

4.2.2. The simulator task

The Chemical Plant Simulator used is a single three tank-three pump System, but equipped with most of the controls and features of a real chemical process plant. A flow diagram of the CPS is enclosed in Appendix A.1.

The operation simulated in the experiment was control of an acid plant, and the task objective was explicitly defined in terms of flow rates, pressure rates and tank levels.

The 'plant' was handed over to the subject in perfect (i.e. fully conforming to the above definitions) running conditions. It was then explained to him that he was to hand over the plant in these conditions at the end of the 'shift' (i.e. 30 minutes). Eventual faults have to be diagnosed and corrected.

After a 'warm-up' of two minutes a series of faults were introduced by the Training Officer in charge from a remote control console. A set of altogether five fault-conditions, standardized in time and character, was used. (A detailed description is enclosed in App. A.2.)

The timing was chosen such that correcting the fault was just not quite possible, resulting in a constant state of emergency. The sequence of fault conditions was furthermore selected in a way that failure to cope with each emergency fairly rapidly increased the magnitude of the overall problem.

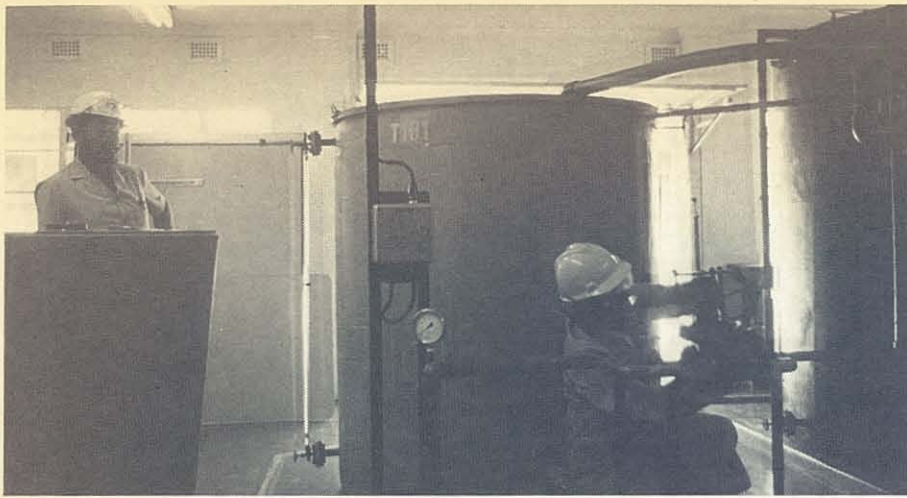
The fault finding and correcting for each set of conditions was broken down into logical steps. The performance in each step was rated by the Training Officer. Fault finding and correction were uniquely defined for each step.

The task-based stressors were:

- Overall time limit of thirty minutes, whereby the elapsed time was visually displayed. The Training Officer also gave a verbal 'count-down' in five minute intervals.
- The increasing complexity of the task itself.
- High achievement pressure, introduced by enforced self-reliance: The subjects only alternatives were to either overcome the emergency without help or to initiate a complete shutdown of the plant (which had been reinforced during the previous weeks of training as the greatest possible evil ever to befall any operator). In the case of gross failure, the Training officer would declare 'catastrophic conditions' (i.e. uncontrollable overflow of acid and escape of highly toxic fumes) and terminate the experiment.
- Noise, caused by the simulator itself and two very loud alarm sirens, which were turned hypersensitive for the experiment.
- Uncomfortable and hindering protective clothing (hard-hat, dust-coat, gum boots, goggles, earmuffs, rubbergloves and respirator).

A considerable potential of failure stress can be assumed.

An illustration of the simulator is given in Fig. 2.



'Corrective action' in a 'hazard area'. The back of the remote-control-console can be seen on the left.



View of the main circuit.



View of the control panel.

5. STATISTICAL ANALYSIS

The objective of the statistical analysis was to establish a directly predictive link between a set of personality factors and the performance in the simulator-task.

The direct way seemed to be the use of Multiple Regression Analysis as the main analytical tool.

The applicability of Multiple Regression Analysis depends mainly on two conditions:

- A. the measurements for all variables have to have at least interval-scale-properties.
 - B. the sample-population dealt with should be homogeneous with respect to its representativeness of the universe to be inferred.
- Ad. A. While interval-scale-properties may be assumed to some degree of validity for the test-scores obtained by the psychological tests, the quality of the measurement applied to performance in the simulator-task had to be investigated. Applying criteria outlined by Guilford (1954) sufficient approximation to an interval-scale was concluded.
- Ad. B. The sample dealt with consisted of two different groups: Operators and trainee operators. Several facts indicated that there should not be a significant difference between the two groups. An Analysis of Variance supported this assumption.

The Multiple Regression Analysis was carried out in several steps: first a linear model was fitted, secondly the fitting of a second-order-equation was attempted. The hyperplanes created by the quadratic model were finally investigated in a Response Surface Analysis.

In the following section the methods of Multiple Linear and Quadratic Regression as well as the principles of Response Surface Analysis are described in more detail. The chapter is then concluded with a scrutiny of the basic statistics concerning the statistical analysis.

5.1. Multiple Regression Analysis and Response Surface Analysis

The basic model to be fitted to a set of observations in a Multiple Regression Analysis is of the form

$$Y = b_0 + b_1 x_1 + b_2 x_2 \dots \dots \dots + b_n x_n$$

whereby Y is called the dependent variable and $x_1, x_2 \dots x_n$ are the independent variables. The 'constant term' b_0 and the coefficients $b_1, b_2 \dots b_n$ determine the form of the equation.

In a Multiple Regression it is thus attempted to construct an equation such that for any particular set of values of the independent variables an expected value for the dependent variable can be estimated.

According to the basic hypothesis the variable to be predicted, i.e. the 'dependent' variable, is to be the performance in the simulator-task. The hypothesis further states that the dependent variable is to be predicted directly from the personality factors that characterise the performing individual, i.e. the 'independent' variables.

The above regression equation provides the tool for this direct prediction.

To obtain an estimate for the 'real' relationship of variables as determined by constant term and coefficients, the Multiple Regression method 'fits' a model to the observational data. The 'fit' is achieved by obtaining a solution for the least squares "best" values for the coefficients for the particular set of equations defined by the sample of observations. The solution also provides a measure of the reliability of each of the coefficients so that inferences can be made regarding the parameters of the population from which the sample was taken.

There is a multitude of techniques available to obtain the multiple regression solution (DAVIES, 1968 gives a brief summary of the major techniques).

The algorithm selected for this project was developed by EFROYMSON, 1967.

This technique adopts a stepwise procedure and uses the Gaussian elimination method to solve the Normal equations.

In the stepwise procedure one variable at a time is added and thus the following intermediate equations are obtained:

$$\text{STEP 1} \quad Y = {}_0b_0 + {}_0b_1 x_1$$

$$\text{STEP 2} \quad Y = {}_1b_0 + {}_1b_1 x_1 + {}_1b_2 x_2$$

$$\text{STEP } n \quad Y = {}_nb_0 + {}_nb_1 x_1 + {}_nb_2 x_2 + \dots + {}_nb_n x_n$$

The variable added is the one which makes the greatest improvement in the overall 'Goodness of Fit' (i.e. the multiple correlation coefficient of the variables included in the particular sub-equation). The coefficients represent the best values at each stage the equation is fitted by the specific variables included in the equation.

The selection of variables to enter the equation is governed by F-values, or 'F-levels' as they will be called further on, which are supplied by the experimenter, obviously to suit the specific conditions of the observational data and the level of significance to be achieved.

The criterion used to select the x_i variable to enter or leave the regression equation is thus:

- If the variance contribution of a variable in the equation is insignificant at the specified 'F-level to leave', this variable is removed from the regression.

If no variable is to be removed, then the following criterion is used:

- If the variance reduction obtained by adding a variable to the regression is significant at a prespecified 'F-level to enter', this variable is entered into the regression.

In the technique used this decision of adding or removing is made at each stage in the elimination procedure, so that not only the final solution is of interest, but at every stage a "partial regression equation" is produced, whereby the - significant - variables already eliminated are in the equation, others are not. According to the add-or-delete decision, the elimination procedure is then "reset" at each stage.

This clever combination achieves an optimum of automatic procedure on one side and a great deal of experimenter-control on the other.

Multiple regression can also be used for nonlinear models of the general form:

$$Y = b_0 + b_1 z_1 + b_2 z_1^2 + b_3 z_1 z_2 \dots + b_n^f (z_1, z_2, \dots, z_m)$$

In the specific case of a quadratic model with n basic (linear) terms the regression equation takes the form:

$$Y = b_0 + \sum_{i=1}^n b_i x_i + \sum_{i=n+1}^{2n} b_i x_{i-n}^2 + \sum_{i=2n+1}^{\frac{n}{2}(3+n)} b_i x_{i-2n} x_i$$

This equation is made equivalent to the basic regression model of

$$Y = B_0 + \sum_{i=1}^M B_i V_i$$

by the substitutions:

$$V_i = X_i \text{ for } n \geq i \geq 1$$

$$\text{and } V_i = X_{i-n}^2 \text{ for } 2n \geq i \geq n+1$$

$$\text{and } V_i = X_{i-2n} \text{ for } \frac{n}{2}(3+n) \geq i \geq 2n+1$$

which means that each power or product entering the regression equation is treated as if it were a separate independent variable in setting up the least-squares equations for the regression coefficients, i.e.:

$$V_1 = X_1$$

$$V_2 = X_2$$

$$\vdots$$

$$V_n = X_n$$

$$\text{and } V_{n+1} = X_1^2$$

$$V_{n+2} = X_2^2$$

$$\vdots$$

$$V_{2n} = X_n^2$$

$$\text{and hence } V_{2n+1} = X_1 X_2$$

$$V_{2n+2} = X_1 X_2$$

$$\vdots$$

$$V_M = V_{\frac{n}{2}(3+n)} = X_n X_{n-1}$$

The maximum value of the subscript M of the transformed variables out of n basic variables thus consists of n linear terms

+ n square terms

+ $\frac{n}{2}$ (n-1) cross products

This steep increase in (secondary) variables somehow limits the applicability of Multiple Quadratic Regression Analysis. On the other hand, the inclusion of cross products widens the scope of the analysis considerably as interactions can be of particular interest especially with psychological variables.

A common difficulty experienced with Multiple Regression Analysis is that generally the full implications of the relations expressed in the regression equation are hard to comprehend and interpret, the more so, when the number of independent variables increases.

The problem escalates drastically if a higher order model is fitted.

Several techniques have been developed to aid in the interpretation of complex regression equations.

A very helpful technique especially with smaller samples and in pilot studies, is the construction of a profile based on the coefficients obtained for the set of independent variables conducive to a specified level of response. Since the independent variables are personality factors which are generally displayed in the form of a profile this technique is particularly applicable and was used in the interpretation of the linear regression equation.

A further technique used to investigate complex relationship of variables is Response Surface Analysis.

Any equation can be interpreted geometrically according to the number of variables it contains, i.e. two variables define a curve, three or more a plane or a hyperplane.

The regression equation containing only the dependent variable (the 'response') and one independent variable therefore would yield a two dimensional curve. If two independent variables are included in the equation, the relationship may be represented as a three-dimensional surface, called the Response Surface (RA).

The common way of a graphical display of the RA is the drawing of lines of equal response in a two dimensional graph whose co-ordinates denote the values of the independent variables. These contour lines (termed Response Contours) may then be read like a geographical map or a weather chart: Contour lines which are close together in any region imply a rapid change in the response for relatively small changes in the variables, and vice versa. Thus the direction and steepness of ascent/descent in response for sets of variable-configuration can be investigated. A linear regression equation produces a set of straight-line contours.

For a quadratic relationship, contour lines generally consist of either concentric ellipses, concentric hyperbolas or open-ended parabolas. Ellipses indicate a positive minimum or maximum, hyperbolas imply a 'saddle' and parabolas can be interpreted as a 'ridge'.

The technique of Response Surface Analysis had been originally developed for the chemical industry. (BOX, 1960; DAVIES, 1968; BOX AND DRAPER, 1969), but at least one application in psychological research could be found in the literature (CLARK & WILLIGES, 1973, MILLS AND WILLIGES, 1973).

In this project Response Surface Analysis was used as an additional investigation to explore the regression equation obtained by the Multiple Quadratic Regression Analysis.

5.2. Basic Statistics

In this section three topics will be discussed:

- Sample characteristics
- Quality of the rating-scores for performance in the Simulator-task
- Procedure and F-levels for the MRA's.

Appendix B contains all the relevant graphs, tables and illustrations for this section.

Sample characteristics

Overall sample size was 52 Ss.

The sample contained two groups:

- NEW STARTERS, i.e. Ss who were newly employed and were undergoing the compulsory induction training
- RETAINED OPERATORS, i.e. operators on a retraining session

To clarify the question of sample homogeneity, an Analysis of Variance was carried out. The analysis principally indicated no significant differences between the two groups, although not quite conclusively: Due to differences in magnitude between the scores (STENS vs. untransformed Performance-scores) high SS's of ERROR were experienced, which could method-inherently suppress significance in extreme cases.

Further findings, however, support the 'no-difference' hypothesis:

- Considering the mean profile in the 16PF-scores it appears that the experimental group represents a "good mix" with no characteristics in the profile. With three exceptions (B+, L+, Q2+) the mean scores lie within the range of STENS 5 and 6, which represents 38,2% of the underlying score distribution, and is the "average" range.

The Intelligence- and Anxiety-mean scores are also fairly close to the average range, although Intelligence represents the highest mean-score of all the psychological variables.

- 'Experience', the factor mainly expected to influence the performance-score and thus distort a relation between personality factors and performance, had a highly skewed distribution with the main bias towards 'lesser experience': nearly 50% of all participants had less than one year of experience as a plant operator.
- Another factor, supporting the hypothesis of a homogeneous group, was the relative simplicity of the actual operation of the simulator together with the extensive specific training given. Thus the initial training for 'new starters' proved to be enough counterweight to offset the advantage of previous experience on the side of the 'retrained operators'.

Considering these additional facts, it was felt that the hypothesis of sample-homogeneity was sufficiently supported.

The age distribution is also fairly skewed towards the age group of 19 to 25.

(These highly skewed frequency-polygons of 'AGE' and 'EXPERIENCE' were the main reason that these two variables were later excluded from the further analysis).

To standardise the range of numerical magnitude of the test scores for the psychological variables, all test scores were transformed to STENS.

The score for performance in the Simulator-task, although showing a good approximation of a Normal distribution, was not transformed into STENS, with its confidence interval being 17, 24 and a STEN-width of only 18,5, the probability of an 'overlap', i.e. the 'true' value might lie in any one of two neighboured STENS would introduce an actually uncontrollable second-order error.

Quality of the rating scores for performance in the simulator task

The total score for performance here was the sum of ratings issued in each of the logical steps of the task. The applied rating was numerical.

In strict terms of measurement theory therefore, the scale thus constructed should principally be considered an ordinal scale, and therefore certain restrictions in the choice of methods of statistical analysis apply.

As GUILFORD (1954) points out, however, a numerical rating-scale can be considered to hold at least interval-properties, if certain qualitative and quantitative criteria are applicable:

- Sufficient experience of the rating individual
- Simplicity and directness of the allocation of number properties to the observations

The rating was done by an experienced training officer. The procedure of rating was simplified by comparing the rating (from one to ten) to giving "schoolmarks" in full percentages only.

To increase the directness of the rating, the performance to be judged was broken down into as many little units as logically possible.

GUILFORD (1954) suggests two empirical checks, mainly to ensure rating-consistency:

- a. the frequency - distribution of the rating scale should not be different from the distribution of the rated phenomenon,
- b. the ratings should be considerably reliable (in terms of test-reliability)

Ad a. Performance was rated positively, i.e. the rating depends on the appearance of a certain reaction (corrective action) to a standardized stimulus (the "fault"). The reaction "chosen" stems from a finite set of possible reactions.

Thus the phenomenon is primarily binomially and consequently normally distributed.

A simple check (using the Chi-square method, Pfanzagl 1966) showed that the obtained distribution of rating - scores is not significantly different from a normal distribution (95% significance-level).

Ad b. The parameter chosen for estimating the consistency of the rating-scores was a split-half-correlation.

The observations were thoroughly mixed, to achieve a quasi-random sequence, separated in a one-for-one mode and rearranged into two distributions. Then the correlation-coefficient was computed. Simultaneously a regression equation for the "first" half on the "second" half was established.

The correlation-coefficient is high (0.9683) and the regression-coefficient close to 1.0, so that a fair consistency of ratings can be assumed.

The analysis of the data thus sufficiently supports the assumption that the performance score can be considered a measurement from an interval-scale. Hence its use in a regression analysis can be considered statistically justified.

The analysis of the distribution of the simulator-scores also served its purpose as a check against any existing constant errors (commonly expected in any rating).

The rather slim distribution (with a coefficient of variation *) of 19.9 the narrowest of all) indicates the presence of the error of "central tendency" (GUILFORD, 1954). The reason for this can probably be found in too markedly a formulation of the end-of-scale statements as "extreme poor-", resp. "extreme good performance".

The slight "overweight" on the high-score side of the distribution suggests the existence of the "error of positive leniency" (GUILFORD 1954), i.e. a one-sided bias in form of a tendency to allocate higher scores.

This is partly underlined by the fact, that the regression equation obtained for the two "half" - distributions contains a positive constant term (although here it is equally probable that the quasi-random sequence of the observations was not randomised sufficiently before the split).

Both errors, although recognisable, were not considered serious enough to have a significantly biasing impact on the obtained results.

There were 28 logical steps in the task at which ratings were issued. The ratings ranged from 1 ('extremely poor performance') to 10 ('extremely good performance').

The maximum score was therefore 280; the minimum, however was 9 and not 28: This was the point at which, with no corrective action taken, the accumulated effects of the faults had reached "catastrophe" - level and the Training Officer had to initiate a shutdown of the simulator.

*) = This statistic is usually not applicable to an interval scale because it assumes an absolute origin of zero. Since this condition, however, applies in a very common sense to test-scores (if we equate 'not measurable' with zero) the coef. o.v. is used for demonstration purposes only throughout this thesis.

Procedure and F-levels for the MRA

Every MRA was initially carried out as a survey analysis without any reference to significance of included variables. This was done to check mainly for algorithm validity and to gain a first impression of variable configurations.

The consequent MRA's were carried out with applying the appropriate F-values, which are defined as follows:

- The F-level for a variable to enter the equation is determined by the F-ratio within the total number of variables.
- The critical F-value for a variable to be removed from the equation is determined by sample-size versus the number of variables.

The applied F-levels were:

MRA	F-Level ENTER	F-Level LEAVE	No. of variables
MLRA III	2,12	1,74	19
MLRA IV	3,18	2,07	9
MQRA II	1,88	1,69	27
MQRA V	2,48	1,89	14

Throughout the statistical analysis a significance level of 95% was applied.

6. FINDINGS

6.1. MULTIPLE LINEAR REGRESSION ANALYSIS (MLRA)

6.1.1. SUMMARY OF FINDINGS

Four sets of variables were analysed in MLRA I to MLRA V.

The first analysis MLRA I included all basic variables, i.e. all 16-PF, STATE- and TRAIT-ANXIETY, INTELLIGENCE (RAVEN PM), AGE and Experience.

To gather a first overall impression of the possible structure of the regression equation and in order to carry out certain checks to determine the validity of the algorithm, all variables without reference to their significance were allowed to enter the equation.

The thus fitted model achieved a fairly high Goodness of Fit, i.e. it can be expected that a high proportion of variance would be "explained" by the model. The structure of the model indicated that the variables 16PF-Q3 (SELF-CONCEPT-CONTROL), 16PF-L (ALAXIA), 16PF-L (EGO-STRENGTH) and TRAIT-ANXIETY would probably be of importance in the final model. Due to their skewed distribution, the variables AGE and EXPERIENCE were excluded from further analysis. Otherwise this analysis confirmed the validity of the algorithm used.

The exclusion of two variables, however, made a second survey - analysis recommendable. MLRA II confirmed and stabilised the structure of the model indicated by MLRA I. Additional variables such as INTELLIGENCE (RAVEN PM) and STATE-ANXIETY gained importance. The exclusion of AGE and EXPERIENCE had only a minor effect on the Goodness of Fit.

In MLRA III only significant variables

N.B.: A COMPENDIUM OF BRIEF DESCRIPTIONS OF THE PERSONALITY VARIABLES USED IS INCLUDED AS A FOLD-OUT ON THE LAST PAGE OF THIS THESIS.

were included into the fitted model. The model included eleven significant variables and achieved a Goodness of Fit of 82,5%. The configuration of the eleven variables conducive to good performance can be clustered into a syndrome of three main groups of personality traits and two clusters of lesser importance:

1. EGO STABILITY CONTROL (low 16PF-L, low A-TRAIT, high 16PF-L)
2. SELF ORIENTATION (high 16PF-Q2, low 16PF-Q3, low 16PF-G, low 16PF-I)
3. ACTIVATION POTENTIAL (high A-STATE, high 16PF-Q4)
4. CYCLOTHYMIA (high 16PF-A)
5. INTELLIGENCE (below average).

Since this construct follows roughly the relation between Cattell's first order traits to second-stratum factors, in MLRA IV it was attempted to fit a model using these second-stratum-factors as basic variables.

MLRA IV was completely unsuccessful: the survey-analysis (including all variables regardless of their significance) achieved a fit of only 41,8%, to which TRAIT-ANXIETY (being substituted for Cattell's QII) was a major contributor. Considering only significant variables, it proved impossible to fit an equation at all.

The following is a more detailed description of MLRA I to MLRA IV.

All relevant illustrations (print-outs, etc) can be found in Appendix C.

6.1.2. MLRA I: ALL VARIABLES/F-LEVELS SET TO 0.0

Variable 16PF-Q3 shows by far the highest significant contribution.

The positive coefficient (high, small variance) indicates that the Q3-high-score manifestation ("High self concept control") is of relevance. Having only entered in the fourth step with a fit contribution of only 8.2%, Q3 gained importance during the steps.

Trait-anxiety, supplying the greatest fit-contribution, is the second highly significant variable. Its negative coefficient (high, small variance) emphasises the low - score range of the variable. A-Trait has also gained significance during the selection procedure.

The third variable with a two-digit F-value is 16PF-L in the low-score - manifestation ("Alaxia"). The coefficient is fairly stable with a variation of roughly 19% *). Entering in Step 2, 16PF-L is second in fit contribution (23%).

"High Ego Strength" (16PF-C+) is ranked next, following by "State-Anxiety", which is interestingly represented in the high - score - range.

The coefficients for both variables show a variation around 20%.

("Experience", which showed also high contribution, will be dealt with later).

The least significant variables were:

- 16PF-D (Untroubled Adequacy vs. Guilt Proneness)
- Intelligence (Progr. Matr. Test), which is surprising, because 16PF-B is fairly significant contributor.

*) = STD. DEV expressed as percentage of the estimated coefficient.

- 16PF-N (Artlessness vs. Shrewdness)
- 16PF-E (Submissiveness vs. Dominance)
- and - 16PF-F (Desurgency vs. Surgency)

The equation achieved a Goodness of Fit of 90, 99%, with a standard error of the derived estimate of 14.478 (i.e. a confidence interval of $\pm 24,32$ on a significance level of 95% applies).

As an additional check, the residuals (i.e. the deviations between observed values for the simulator score and estimates, calculated via the obtained equations) were plotted against each variable in turn, including the dependent variable. Davies (1968) suggests this visual, but powerful technique for the following reasons: a major assumption underlying MLRA is, that the residuals, i.e. errors unexplained by the independent variables are independent with zero mean, constant variance and consequently Normally distributed. If these assumptions are fulfilled, the residuals should lie roughly in a horizontal band around the mean of the independent variable, looking "randomly".

With two exceptions, all variables indicated Normally distributed residuals, although in some cases curvatures were suspected.

Age and Experience, however, due to their highly skewed distribution, gave a fairly distorted picture.

Since "freak" observation---distributions can easily produce algorithm---inherent artefacts (Davies, 1968), it was decided to eliminate these two variables from the further analysis: the relatively high fit contribution of Experience is considered to be mainly due to such method particulars, and Age proved to be rather insignificant in any case.

The analysis of the variables in rank order of their contribution to the Goodness of Fit showed the following picture:

STEP NO.	VARIABLE *)	CONTRIBUTION TO G/F%	G/F% OF SUB-EQUA- TION
1	TRAIT-ANXIETY	29,3280	29,3280
2	ALAXIA (L)	22,9574	52,2854
3	EXPERIENCE	7,6938	59,9792
4	INTEGRATION (Q3)	8,1580	68,1372
5	GROUP ADHERENCE (Q2)	5,5127	73,6499
6	SIZE/CYCLOTHYMIA (A)	3,0597	76,7096
7	STATE-ANXIETY	3,8142	80,5238
8	THRECTIA (H)	1,3450	81,8688
9	EGO STRENGTH (C)	2,3929	84,2647
10	SCHOL. M. ABILITY (B)	1,8070	86,0717

The other 11 variables together increased the G/F by a mere 4,9205% to 90,992% G/F of the final equation.

This represents an average G/F contribution of only 0,4473%.

*) 16-PF variables are sometimes named in their low-score manifestation only.

G/F = Goodness of Fit.

6.1.3. MLRA II: VARIABLES: 16PF, A-TRAIT/STATE, INTELLIGENCE/F-LEVELS SET TO 0.0

The exclusion of the variables and Experience had mainly a moderating effect on the equation. Also, the structure of the included variables changed.

16PF-Q3 high - score is still the most significant contributor, while 16PF-L (low - score) holds now second rank before Trait-Anxiety (low-score) and 16PF - C (high - score).

Intelligence (measured with the Progr. Matr. - Test) has slightly gained importance, against 16PF-B, which has declined in significance. Interestingly the coefficients have contrary signs: 16PF-B is positively influencing, while the coefficient for Intelligence is negatively represented (i.e. a lower score here is conducive to satisfactory performance).

Another interesting contrast is seen in the configuration of A-Trait and A-State, with A-State being positively and A-Trait - fairly markedly - negatively "loaded".

The traits in rank-order of least significance in this equation are:

16PF-Q1 (Conservatism vs. Radicalism)

16PF-E (Submissiveness vs. Dominance)

16PF1M (Praxernia vs. Autia)

16PF-B (Scholastic Mental Capacity)

The equation achieved a Goodness of Fit) of 85.9% and showed a standard error of estimate of 17.54, which implies confidence limits of +/- 29.47.

The exclusion of the variables Age and Experience has decreased the Goodness of Fit by only 5%. The confidence interval for estimating the simulator - score from the equation has widened from F24.32 to F29.47 (i.e. a percentage increase in width of 8.25%).

An analysis of the variables in rank-order of their contribution to the Goodness of Fit showed the following picture:

<u>STEP</u>	<u>VARIABLE</u>	<u>CONTRIBUTION TO G/F%</u>	<u>G/F% OF SUB- EQUATION</u>
1	TRAIT ANXIETY	29,3280	29,3280
2	ALAXIA (L)	22,9574	52,2854
3	INTEGRATION (Q3)	5,6745	57,9599
4	GROUP ADHERENCE (Q2)	9,6179	67,5778
5	SUPEREGO STRENGTH (G)	3,3845	70,9623
6	EGO STRENGTH (C)	2,9920	73,9543
7	SIZO/CYCLOTHYMIA (A)	2,2855	76,2398
8	INTELLIGENCE (RAVEN)	1,7283	77,9681

The remaining 11 variables together increased the G/F only 7,928% to a final 85,8968%, which represents an average G/F contribution per variable of a mere 0,7208%.

6.1.4. MLRA III: VARIABLES: 16PF, A-STATE/TRAIT, INTELLIGENCE

The equation, containing only significant variables, consists of eleven variables and achieves a Goodness of Fit of 82.4438%, only 3.4538% less than the equation which also included insignificant variables.

The rank order of variables by significance remained unchanged as the following listing shows:

1. 16PF-Q3 (high-score) "High Self-concept Control"
2. 16PF-L (low-score) "Alaxia"
3. A-TRAIT (low-score) low level of trait-anxiety
4. 16PF-Q2 (high-score) "Self-Sufficiency"
5. 16PF-C (high-score) "Higher Ego-Strength".
6. 16PF-A (high-score) "Cyclothymia" (Affectothymia)
7. 16PF-G (low-score) "Weaker Superego Strength"
8. A-STATE (high-score) high level of transient anxiety.
9. 16PF-I (low-score) "Harria"
10. 16PF-Q4 (high-score) "Ergic Tension"
11. Intelligence (low-score)

To gain a better overall picture of the variable configuration, Fig. 3 shows a plot of coefficients against variables, the middle line representing zero and the coefficients directed towards the low or high-score range of the variables according to the sign of the coefficients.

(Coefficients for non-significant variables are set to zero)

Grouping the significant traits, one can separate five clusters of related traits (roughly guided by similar loadings of first - order - traits in second - stratum factors of the 16PF, Catell et al., 1970). Taking into account the particular configuration found, these can be labelled:

- Ego- Stability Control (a-neuroticism, represented by low A-TRAIT, C+, L-, Q3+),
 - Self - Orientation (Q2+, G-, underlined by I-),
 - Hyperactivation - Potential (Q4, A-STATE, underlined by I-)
- and
- Cyclothymia (A+)
 - Low Intelligence

The importance of these five groups of variables for achieving optimum performance seems to follow the above order.

It should be emphasized here that the clustering is merely of a syndromatic nature, i.e. it does not imply any underlying factorial structure of any nature. The "label" given is purely descriptive.

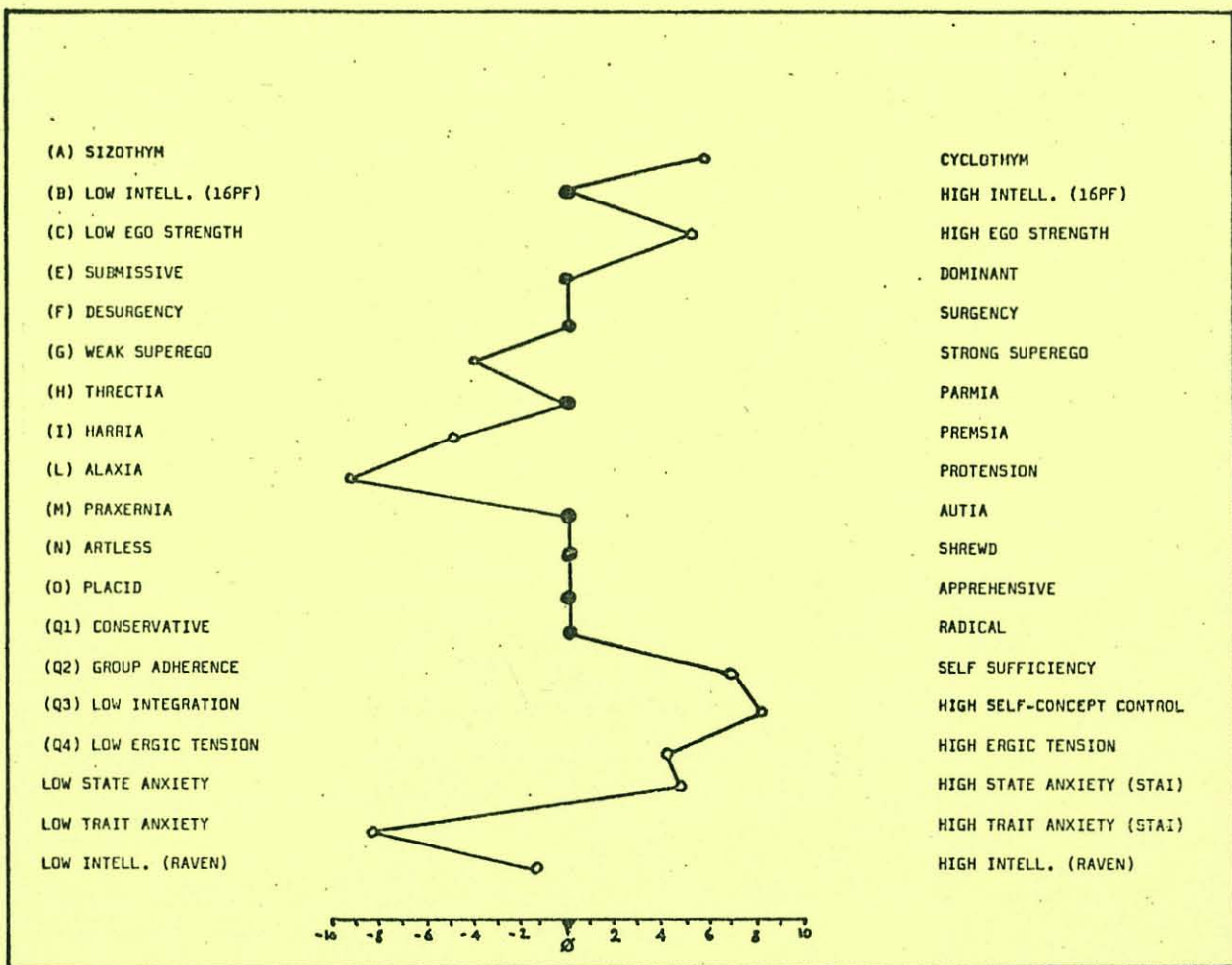


Fig.3. MLRA III-coefficient values in profile.

Since, however, consideration of Catell's second-stratum traits proved helpful to arrange the variables in a psychologically plausible cluster-pattern, it was decided to investigate the predictive quality of second stratum - trait - scores further.

The scores obtained for the first-order-factors were therefore transformed into STENS for second-stratum-traits (using weights and constants of Catell et. al. 1970).

The results of this analysis are described further below.

App. C4 shows the scattergram and the histogram of residuals plotted against the actually obtained simulator score. There is a small bias towards higher scores, i.e. the higher the actual score the more the model over-estimates.

Since a bias in the same direction was found in the basic frequency-polygon of the simulator-scores (possibly due to a "leniency-error" in the rating), the consistent deviation of estimates produced by the equation seems to simply reflect an extrapolation - effect within the model.

The analysis of variables in rank order of their contribution to the Goodness of fit showed the following picture: (see next page).

<u>STEP NO.</u>	<u>VARIABLE</u>	<u>CONTRIBUTION TO GOODN/FIT</u>	<u>GOODNESS OF FIT OF EQUA- TION</u>
1	TRAIT ANXIETY (STAI)	29,3280	29,3280
2	ALAXIA/PROTENSION	22,9574	52,2854
3	SELF CONCEPT CONTROL	5,6745	57,9599
4	GROUP ADHERENCE/SELF-SUFF	9,6179	67,5778
5	SUPEREGO STRENGTH	3,3845	70,9623
6	EGO STRENGTH	2,9920	73,9543
7	SIZOTHYMIA/CYCLOTHMIA	2,2885	76,2398
8	INTELLIGENCE (RAVEN/PROG. MATR.)	1,7283	77,9681
9	HARRIA-PREMSIA	1,3358	79,3039
10	STATE-ANXIETY (STAI)	1,4827	80,7857
11	ERGIC TENSION	1,6581	82,4438

The rank order is the same as in the previous analysis (with Zero-F-Levels). No variables have been deleted.

6.1.5. MLRA IV: VARIABLES: 16PF-SECOND-STRATUM-FACTORS
EXCLUDING: ANXIETY, INTELLIGENCE

F-LEVELS SET TO ZERO

To avoid the amplifying effect of having two correlated variables amongst the independent variables, the 16PF-second-stratum trait "anxiety" was excluded in favour of A-TRAIT (STAI) and 16PF - second-stratum factor "intelligence" was left out because of the presence of the score for intelligence measured by the Progressive - Matrices-Test, which was felt to be a more consolidated and reliable score for this variable.

App. C5 shows the transformed data matrix and the results of this analysis.

This set of variables only achieved a fit of 41.8487%.

The level of significance is fairly low throughout all variables.

However, the important role of A-Trait was supported here again: it is the most significantly contributing variable, and adds 29.328% to the fit. Since the overall fit was only 41.8, the six 16PF - second-stratums together contributed only 12.521% or not even a third of the total fit-percentage.

Two explanations for this surprising result seem plausible:

- The relatively high impurity of Cattell's second-stratum-factors and the particular configuration of their intercorrelations might counteract and suppress the effects attained by the analysis of the first-order-traits.

- Only eight first-order-traits showed a satisfactorily significant fit-contribution in the previous analysis, while the second-stratum-factors were derived from all sixteen primaries.

A further analysis showed that no variable in this configuration is significant enough to model an equation (F-level enter for 9/9 of = 3,18, F-Level for leave for 50/9 of = 2,07).

Applying the F-levels appropriate for the underlying variables (i.e. 2,1244 and 1,7444 respectively), which is statistically dubious and stretches the significance level to an estimated 78%, only produced two variables in the equation.

Expectedly one was A-Trait. Surprisingly the second one was QV i.e. "Discreetness", a second-order-trait that is nearly solely loaded by the primary N, which in turn was found to be highly insignificant in MLRA III.

A possible explanation may again refer to the above mentioned high impurities: QV not only has the highest intercorrelations amongst the 16-PF primaries, it is also highly intercorrelated with nearly all other second-stratum-factors.

Summarizing, it can be said that no significant relation between 16PF-second-stratum-factors and the dependent variable (simulator-score) could be found.

6.2. MULTIPLE QUADRATIC REGRESSION ANALYSIS (MQRA)

6.2.1. INTRODUCTION

The scrutiny of the scattergrams of residuals, as mentioned in the description of the findings of MLRA I, serves two purposes:

- a) To provide a visual check if the underlying assumptions for any MRA are fulfilled, i.e. if the residuals are normally distributed with zero-mean and constant variance. (The inherent Analysis of Variance for the multiple regression demands both criterias, t-tests and F-tests of significance only require Normal distribution). Thus, ill-distributed variables, or outlying observations may be recognised and consequently omitted to obtain an improved fit of the model.

Furthermore, systematic errors in the analysis might be diagnosed.

- b) If the 'band' of plotted residuals describes a curved path rather than a straight line, this could indicate the need for a polynomially expressed underlying relationship.

In MLRA I the examination of the residuals was utilised to delete two variables, and a systematic error (the model's extrapolatory reflection of the 'error of leniency' inherent to the rating) was detected.

In order to investigate if the linear model achieved so far is yielding the optimal degree of fit-exhaustion obtainable through the available observation-data, the scattergram of residuals was scrutinized with respect to curvatures.

N.B.: Appendix D contains all the relevant print-outs for this section.

A curved path was suspected for seven of the eleven variables that were included in the MLRA III - model.

The variables whose residuals indicated a curvature were:

SIZE/CYCLOTHYMIA (A)
EGO STRENGTH (C)
SUPEREGO STRENGTH (G)
ALAXIA (L)
GROUP ADHERENCE (Q2)
INTEGRATION (Q3)
TRAIT-ANXIETY

Despite the fact that the suspected curvature was not distinct in any of the seven variables (and the selection was therefore somewhat subjective) it was decided to attempt to fit a quadratic model.

Two additional considerations supported this decision:

Davies (1968) suggests that whenever a set of optimum conditions for the independent (response) variable is thought possible, consideration should be given to fitting a second-order-model, especially to multivariable data.

The given experimental layout by definition implies 'optimum conditions'. The test-score-range clearly limits the feasible set of values for the independent variables, and by defining the highest observed performance as the 'maximum response', a set of optimum conditions is formulated eo ipso.

Furthermore, the equation arrived at by the MLRA III does not allow conclusions as to the rank order of importance of variables other than by magnitude of coefficients.

The inclusion of square terms and cross-products in a quadratic model could, however, provide an increased insight into the interdependencies of variables beyond the unidimensional boundaries of the linear model.

Since the amount of terms (t) in a quadratic model constructed of p basic variables increases with $t = 0,5 p (3 + p)$ and, as a general rule (Davies 1968) the number of observations should at least be twice as great as the number of coefficients to be estimated, only six variables could be included in a second order model.

Also the consequences of a quadratic transformation to the significance of variables included into the regression equations had to be considered. The unavoidable increase of artificial intercorrelations between linear terms, square terms and crossproducts can have an uncontrollable influence on variable selection - in particular in the stepwise method, and especially since the algorithm used provides an automatic check against the possibility of 'degeneracy' (Efroymsen, 1967). To counteract this only the six most significant variables of MLRA III were to be included in a MQRA.

These were:

SIZO/CYCLOTHYMIA (A)
EGO STRENGTH (C)
ALAXIA (L)
GROUP ADHERENCE (Q2)
INTEGRATION (Q3)
TRAIT ANXIETY

6.2.2. SUMMARY OF FINDINGS

Three attempts were undertaken to fit a second order model.

MQRA I was carried out as the initial survey-analysis to obtain an overall picture of the possible structure of the regression equation and to check algorithm-validity.

The high number of variables taken into the equation and the unavoidable lack of variable independence (i.e. increased correlations between linear and square terms and cross-products) resulted in an levelling effect on the importance of individual variables. It could be seen, however, that crossproducts of the basic variables seem to have the main influence.

The achieved Goodness of Fit was quite high with 86,7%.

Applying the significance - 'filters' in MQRA II reduced the number of variables included in the equation to five and only resulted in a drop of the 'explained' variance to 77%. All six basic variables were represented in the model:

- 16PF-L and TRAIT-ANXIETY in their crossproduct term
- 16PF-Q2 in the square term
- 16PF-Q3 in linear form
- 16PF-A and 16PF-C as crossproduct
- and - the crossproduct of 16PF-Q2 and 16PF-Q3.

The dominance of crossproducts is a secure indicator for the fact that obviously the interaction of variables is the main determining factor in predicting performance. Furthermore the importance of 16PF-Q2 and 16PF-Q3 is underlined by the individual appearance of both variables.

The configuration of the variables conforms in principle with the clustered syndrome constructed in following the linear model.

MQRA III was conducted as an experiment with subsets of the basic variables. A survey-analysis with the subset 16PF-A, 16PF-C and 16PF-Q2 only achieved a Goodness of Fit ratio of 30%. Analysing the steps of assembling the final model, Q2 was found to be the main contributor.

The second subset consisted of 16PF-L, 16PF-Q3 and TRAIT-ANXIETY. The fitted model showed a G/F-Ratio of 64%, whereby the crossproduct of 16PF-L and A-TRAIT was the major contributor. It was then decided to form a third subset including 16PF-L, 16PF-Q2, 16PF-Q3 and A-TRAIT. In MQRA IV the survey-analysis was carried out. Allowing only significant variables to enter the model, MQRA V fitted an equation consisting of only two variables and still a Goodness of Fit of 65% could be achieved. Both variables in the equation were crossproducts:

- 16PF-L and A-TRAIT
- 16PF-Q2 and 16PF-Q3

Referring to the syndrome derived from the linear model, this result clearly puts a strong emphasis on the predictive value of the clusters 'EGO STABILITY CONTROL' and 'SELF ORIENTATION'.

The following is a more detailed description of MQRA I to MQRA V.

6.2.3. MQRA I: 16PF PRIMARIES: A, C, L, Q2, Q3, A-TRAIT

In this first survey-analysis a Goodness of Fit of 86,7% was obtained.

The fact that only seven variables - significantly - contributed to approximately 90% of the overall fit (78,1%) characterized this analysis.

The twenty insignificant contributor-variables very obviously levelled the overall significance of the equation out to a fairly low level.

Extreme distortions in contribution-significance occurred: the crossproduct of L x A-TRAIT, contributing 55% to the overall fit, plummeted from an initial F-Ratio on entering the equation of 61,24 to a barely significant 1,99 and so did the Crossproduct of Q2 x Q3: It dropped from a highly significant 13,8 to a final - insignificant - 0,9.

The experienced levelling effect seems to be mainly due to the fact that this analysis is probably testing the limits of the method by introducing a number of variables more than 50% of the total number of observations.

The algorithm used, however, combines the elimination-procedure with the stepwise decision of including or excluding variables. The application of the proper significance-filters should therefore overcome the distorting effect of the high number of variables.

6.2.4. MQRA II: 16PF-PRIMARIES: A,C,L,Q2,Q3,
A-TRAIT

Only five variables entered the final equation. These five variables, however, are highly significant and a fairly high percentage fit could be achieved: 77,0079%.

This is only 9,6425% less than the equation with all 27 variables achieved, and a mere 5,4359% less than the equation obtained in MLRA 111.

Included variables are:

"Crossproduct Alaxia - A-TRAIT", which contributed 55% to the overall fit. The coefficient is slightly greater than one, which indicates a rather "pure" contribution of the score-products. The sign of the coefficient is negative, hence low-scores in "Alaxia" and "Trait-Anxiety" lead to a high performance - score on the simulator.

This corresponds with the findings for linear terms only (MLRA 111), where both "ALAXIA" and "A-TRAIT" showed negative coefficients.

The next variable in the rank order of significance is the quadratic term of 16PF-Q2, i.e. "SELF-SUFFICIENCY". Its coefficient is also only slightly greater than one and positive, i.e. high-scores contribute most to a good performance-score.

The linear term of 16PF-Q3, "SELF-CONCEPT CONTROL" also has a positive coefficient, which makes high-scores in this variable conducive to good simulator-performance.

The high numerical value of the coefficient further suggests that already low scores in Q3 may contribute markedly.

The crossproducts of Q2 and Q3 as well as A and C are also included in the equation.

The analysis of variables in rank order of their contribution to Goodness of Fit showed the following.

<u>STEP NO.</u>	<u>VARIABLE</u>	<u>CONTRIBUTION TO GOODN/FIT</u>	<u>GOODN/OF FIT OF EQUATION</u>
1	CRP: ALAXIA x A-TRAIT	55,0511	55,0511
2	CRP: GROUP ADH x SELF CONC/C	9,5109	64,9562
3	CRP: SIZO/CYCLO x EGO STRENGTH	3,9392	68,8954
(4)	(CRP: EGO STRENGTH x ALAXIA)	(1,6760	(70,5714)
5	SQUARE: GROUP ADHERENCE	1,5995	72,1709
6	SELF CONCEPT CONTROL	5,1903	77,3612
7	DELETION OF VARIABLE: CRP: EGO STRENGTH x ALAXIA	0,3533	77,0079

The CRP: 16PF-C/16PF-L was taken in with a F-Ratio of 2,677. The addition of square term 16PF-Q2 increased the ratio to 2,903, the inclusion of 16PF-Q3 (linear term) caused the ratio to drop to 0,7021, well below the minimum F-level.

All six variables are represented in the final equation in cross-products. Group Adherence appears in the quadratic transformation, only Self-Concept-Control is included as the linear term.

The high significance together with the relatively high overall-fit achieved indicates a fairly stringent discriminatory quality of the quadratic transformation.

The rank order of contribution follows quite closely the rank order found for the linear terms. Additional emphasis is placed on factors 16PF-Q2 and 16PF-Q3.

6.2.5. MQRA III: 16-PF PRIMARIES A,C,Q2 and
16-PF PRIMARIES L,Q3, A-TRAIT (STAI)

To enable a further distinction of variable-importance, the set of six basic variables was divided into two subsets for further analysis. These subsets were:

1st subset

16-PF A (SIZO/CYCLOTHYMIA)
16-PF C (EGO STRENGTH)
16-PF Q2 (GROUP ADHERENCE/SELF SUFFICIENCY)

2nd subset

16-PF L (ALAXIA/PROTENSION)
16-PF Q3 (LOW/HIGH SELF CONCEPT CONTROL)
A-TRAIT (STAI)

Since only a first survey was intended, the F-Levels to enter/leave were set to ZERO in both runs.

The analysis of the first subset showed a comparatively low overall Goodness of Fit of the obtained equation: a mere 30,4% was achieved.

16PF - primary A proved to be the most significant contributor, being represented in the square term and both crossproducts. Q2 and C were only included in crossproduct terms.

The low fit is also reflected in the relatively high variances of the coefficient-estimates: the confidence interval for the coefficient includes zero for six variables. Only the three most significant variables (SQU:A, CRP: A x Q2, CRP:A x C) show a coefficient of variation less than 100%.

The insignificance of Q2 in this analysis is slightly surprising, especially after MQRA II had included Q2 both in the square term and in the crossproduct Q2 x Q3.

A possible explanation could be that A and C were represented in Crossproduct form in MQRA II, thus indicating a strong interrelation, in which A seems to dominate. This, together with the levelling effect caused by the inclusion of insignificant terms might have suppressed a possible influence of Q2.

An indication of the validity of this explanation can be seen in the fact that Q2 (square term) entered the equation in step 2 with a fit-contribution of 7,5% and an F-ratio of 4,750. The successive inclusion of non-significant variables then caused its F-ratio to fall to a final insignificant 0,5287.

The analysis of variables in rank-order of fit-contribution showed the following:

<u>STEP</u>	<u>VARIABLE</u>	<u>CONTRIBUTION TO G/F%</u>	<u>G/F OF SUBEQUA- TION</u>
1	CRP: A x C	15,3806	15,3806
2	SQU: Q2	7,4782	22,8588
3	SQU: A	1,5964	24,4552
4	CRP: A x Q2	4,0218	28,4770
5	LIN: A	0,8222	29,2992
6	LIN: C	0,7777	30,0769

The other 3 variables could increase the Goodness of Fit by only 0,3432% to 30,4201%. This is equivalent to an average variable contribution of 0,1144%.

The fitting of the quadratic model including the second subset showed a good fit with 63,6594%.

The significant variables in the equation were (in order of significance)

- Crossproduct L x A-TRAIT
- Square term of Q3
- Square term of L

Excessive variance of coefficients of insignificant variables (i.e. including zero in the confidence interval) was experienced in this analysis too.

The stepwise induction of variables into the equation followed the above rank order of final significance, thus indicating a 'genuine' rather than a spurious significance:

<u>STEP</u>	<u>VARIABLE</u>	<u>CONTRIBUTION TO G/F%</u>	<u>G/F% OF SUB-EQUA- TION</u>
1	CRP: L x A-TRAIT	55,0511	55,0511
2	SQU: Q3	6,5693	61,6204
3	SQU: L	0,2386	61,8590

The other six variables increased the overall Goodness of fit only marginally by 1,8004% to a final 63,6594%. This represents an average fit-contribution of 0,3001%.

The main contributor, the crossproduct L x A-TRAIT, can be considered as a relatively 'pure' variable, as a check on correlations revealed: Both linear and square terms are virtually uncorrelated with coefficients not greater than 0,06. The intercorrelations of the crossproducts with its components in linear and square form are - expectedly - higher with 0,6 to L and 0,8 to A, which may indicate increased importance of A-TRAIT by itself.

6.2.6. MQRA IV: 16-PF PRIMARIES L, Q2, Q3, A-TRAIT
(STAI)

Although Q2 did not contribute significantly in the analysis of subset I in MQRA III, on account of the strength shown in MQRA II, it was included in this analysis.

The first survey of significance (both F-levels set to ZERO) showed a high Goodness of Fit with 77,0299%.

The inclusion of insignificant variables brought the expected leveling effect.

Despite this artificial balance, however, significance seemed to spread nearly equally over all variables included in this subset. The relative homogeneity of this combination of variables seems to be further underlined by the fact that crossproducts dominate amongst the more significant variables.

This predominant role of crossproducts was confirmed as the analysis was repeated with the appropriate F-level-filter to induce only significant variables.

Only two variables were included in the final equation;

- crossproduct L x A-TRAIT
- crossproduct Q2 x Q3

The achieved Goodness of Fit was 64,9562% - only a mere 17,5% less than the linear fit derived from eleven variables achieved.

Both variables are highly significant with relatively stable coefficients. A scrutiny of the residuals showed a slight positive slope of the band - this model shows the same tendency to overestimate high scores and underestimate low scores as the linear equation.

6.3. RESPONSE SURFACE ANALYSIS (RSA)

6.3.1. INTRODUCTION

It was felt that the dramatic reduction of variables in the quadratic model, together with the high percentage of fit, could not be excepted without further analysis.

Response Surface Analysis was chosen as the main tool for this analysis.

The main objectives for this further investigation were:

- To explore in depth the interrelations between the variables included in the model and the connexion of model and predicted response.
- to determine the existence of optimum conditions.
- to confirm if the fitting is genuine, i.e. if the underlying 'true' relation between the included variables can in fact be best described with a quadratic model, or if the high fit was a mere 'arithmetical' artefact.

If there is a 'true' underlying second order relation then it can be expected that the system of contour-curves (the 'response surface', RS) has its centre within or very near the feasible region (Davies, 1968; Box 1960). The curves produced by a quadratic model are conics, i.e. have a strictly defined vertex.

There is some argument in the literature, whether a final quadratic model should have a balanced form. (DAVIES, 1968, Box 1960, BOX & DRAPER 1969).

This means, that despite certain individual terms having been discarded by the regression algorithm as insignificant, the final equation should always include a complete set of all terms concerned: if a crossproduct remains, both linear and square terms should be included; if square terms remain, than crossproducts should also appear in the final model, etc.

The disadvantages of an unbalanced model appear to be:

- the suppression of a non-significant linear term in one of the variables of a quadratic model might result in the implication that the vertex of the hyperplane concerned occurs at the origin of that variable, which
 - depending on the nature of the variable - can be an artefact.
- the inclusion of interaction terms only might result in a degree of symmetry of the final model which is probably spurious.

The advantage of an unbalanced model is, however, that the confidence limits of the predictions are kept to a minimum.

It was decided to include solely the unbalanced equations in the RSA.

The reasons for this decision were:

- having the vertex of the resulting conic correspond with the origin of the variables concerned can be a perfectly valid result. Since the test-scores have not been scaled in any way, the origin (i.e. a test-score of zero) is a 'true' value and not an arbitrarily chosen working origin.

- The argument of a probable 'spurious degree of symmetry' is not really applicable to this analysis. What it means is that successive layers of contour-curves might be artificially brought into a position of concentric vertices, whereby the 'true' surface might show a skew centre-line of vertices, thus showing different degrees of slope in different sections of the RS (whereas concentric vertices result in a constant slope in all sections of the RS)

This is of importance only in an environment of continuous variables which can be set by the experimenter at his sole discretion, e.g. to approach an optimum region via the path of steepest ascent of the RS.

This is not the case in this analysis: the variables are discrete and cannot be influenced by the experimenter.

- a scrutiny of the 'balanced' model, i.e. the equation containing all variables, showed that due to the excessive variability of coefficients of insignificant variables, the resulting picture of the RS would be extremely distorted. And principally: to analyse a hyperplane, which, by definition, can have virtually any slope, with the present one just being the one that happened to suggest itself by the data, seems a somewhat futile exercise.

In terms of the RS created, the levelling effect observed in the balanced models produced a singular pattern for all combinations of variables: flat parabolas (indicating a near-linear stationary ridge) with positive slope, consistently and indiscriminately showing high test scores to produce high responses.

The sets of variables included in the RSA were:

RSA I: 16-PF A, C, L, Q2, Q3 and
A-TRAIT (STAI)

RSA II: 16-PF L, Q2, Q3 and A-TRAIT
(STAI)

The analysis was concluded in two steps. The first step restricted the RS to the feasible region, i.e. variable-values from 1 to 10 only. To obtain a picture of the more complete structure of RS, the boundaries of the investigated region were extended to include -10 to +20 for variable values in the second step.

The following nomenclature is being used:

FR - Feasible Region

RS - Response surface

RL - Response levels ('contour heights'):

A = 25 (i.e. simulator score)

B = 50

C = 75

D = 100

E = 125

F = 150

G = 175

H = 200

I = 225

J = 250

Direction of Slopes:

NW - towards top-left-hand-corner;

opt. response: high ordinate-variable
low abscissa - variable

NE - towards top-right-hand corner;

opt. response: high ordinate -
variable
high abscissa - variable

68.

SE - towards bottom - right-hand-cover;
opt. response: low ordinate -
variable.
high abscissa - variable.

SW - towards bottom - left-hand-corner
opt. response: low ordinate-
variable.
low abscissa - variable.

6.3.2. SUMMARY OF FINDINGS

This analysis of the response surfaces yielded by the quadratic models was carried out to determine and explore the relevance of strengths of the personality-variables to the expected performance level in more depth.

In RSA I the contour diagrams of the model fitted in MQRA II were investigated.

The variables 16PF-A and 16PF-C were found to be of importance mainly for higher levels of performance. The interaction of A-TRAIT and 16PF-L spreads over a broader spectrum of performance levels. This spectrum, however, covers only the area of mediocre performance.

16PF-Q2 and 16PF-Q3 was found to be the only combination of variables that produced optimum performance. High scores in Q2 are of more influence than high scores in Q3.

RSA II, carried out on the surfaces yielded by the model developed in MQRA V, confirmed the findings of RSA I, although the importance of the interaction of Q2 and Q3 was found to be more evenly spread between both variables.

6.3.3. RSA I: 16-PF PRIMARIES A, C, L, Q2, Q3
and A-TRAIT (STAI)
 (Unbalanced Model)

Interaction A x C: Only one response level (RL) is manifest in the feasible region (FR). Its curve indicates that either high A with only 50% of C or vice versa leads to a upper-region simulator score.

Extending the boundaries, it was found that the RS created is a hyperbolic plane with a distinct saddle about G-level. The centre of the system is approximately at the origin of both variables. The low levels are sloped NW/SE, the ascending slope runs NE/SW.

The RS obtained seems to indicate that both variables come to importance at high performance level only.

Interaction L x A-TRAIT: four contours are crossing the FR: C to G. The slope is to SW and flattening out with increasing RL.

Extending the RS beyond the FR, a hyperbolic plane is found, again centred at the origin of both variables and with a saddle at G-level. The axes of the system are nearly congruent with the axes found for A x C, the slopes, however, are inversed: Low-RL's are on NE/SW, higher responses ascend along NW/SE.

In accordance with the magnitude and sign of the coefficients for L x A-TRAIT (around 1,0 and negative) the RS indicates clearly increasing response with decreasing variable-values.

The range of manifest RL's seems to indicate that both variables are of importance rather in the low-performance-region.

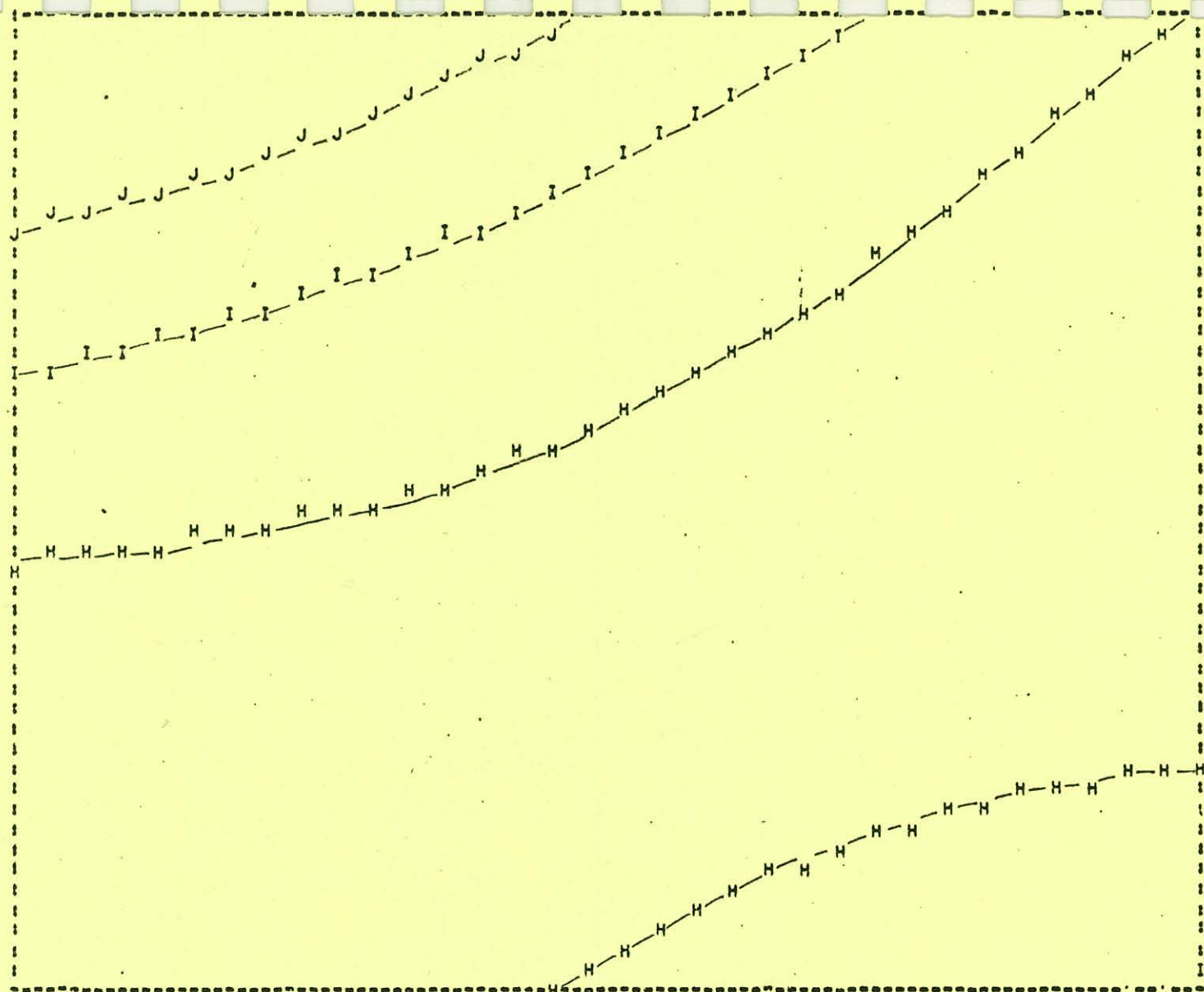
The near perfect symmetry of the conic system is clearly a result of the fact that both variable-combinations were represented in the model in crossproduct form only.

Interaction Q2 x Q3: The obtained hyperplane within the FR is obviously the saddle of a hyperbolic surface. Only high RL's are represented: H to J, and the high level slope runs NW/SE. The region SW/NE appears as a flattened saddle about, or just below H-level.

In exceeding the FR-boundaries, the RS becomes a longstretched hyperbolic plane, with the longitudinal axis in NE/SW (descending RL's), and the axis of ascending RL's at right angles to it. The saddle, around H-level, covers quite a large area, with steep inclines towards I and J in NW, resp. SE.

The specific character of the hyperplane, especially the dominating flat saddle at high RL seems to indicate that virtually all combinations of Q2 and Q3 would predict high performance. Q2, due to the J-contour in NW, would be the variable, whose high-score carries more weight. This is somehow underlined by the fact that Q2 was introduced into the model as square term, Q3 as linear term, and their crossproduct had a negative coefficient (as a levelling effect this probably caused the flat saddle).

Generally therefore, any combinations of Q2 and Q3 should yield a good performance-level. The optimum however could be seen in a combination of high Q2 and not too low Q3.

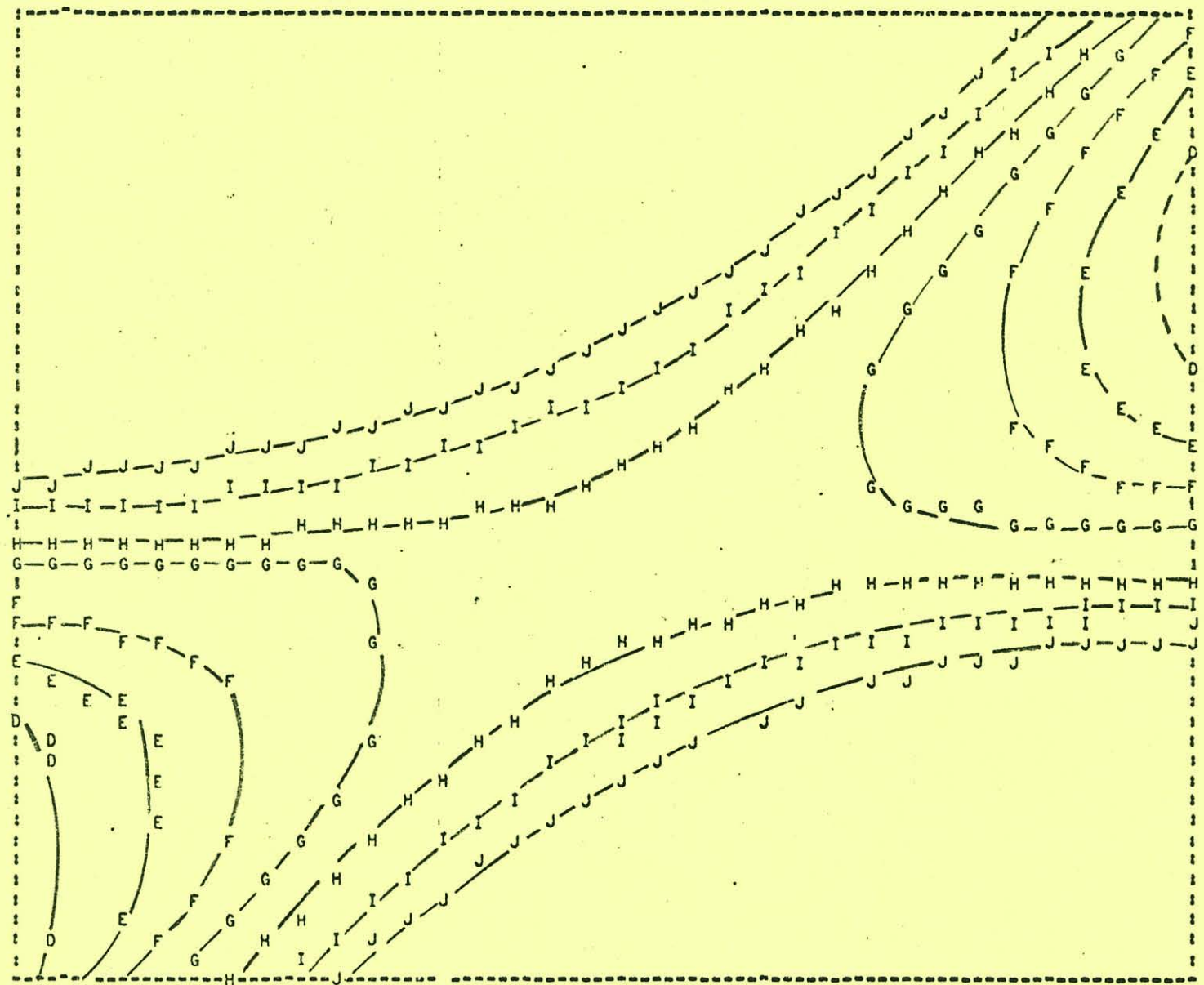


RESPONSE SURFACE OF VARIABLE 4 ON THE Y-AXIS
AND VARIABLE 5 ON THE X-AXIS

KEY TO CONTOUR HEIGHTS (RESPONSE LEVELS):

SYMBOLS:	A	B	C	D	E	F	G	H	I	J
RESP. L.:	25	50	75	100	125	150	175	200	225	250

Fig. 8: 16PF-Q2 (Var. 4) versus 16PF-Q3 (Var. 5) in the FS



RESPONSE SURFACE OF VARIABLE 4 ON THE Y-AXIS
AND VARIABLE 5 ON THE X-AXIS

KEY TO CONTOUR HEIGHTS (RESPONSE LEVELS):
 SYMBOLS: A B C D E F G H I J
 RESP. L.: 25 50 75 100 125 150 175 200 225 250

Fig. 9: 16PF-Q2 (Var. 4) versus 16PF-Q3 (Var. 5) in the extended FS

Since A-TRAIT was included into the model in crossproduct with L only, the RS of Q2 x A-TRAIT and Q3 x A-TRAIT is meaningless.

6.3.4. RSA II: 16-PF PRIMARIES L, Q2, Q3 and A-TRAIT

(Unbalanced model)

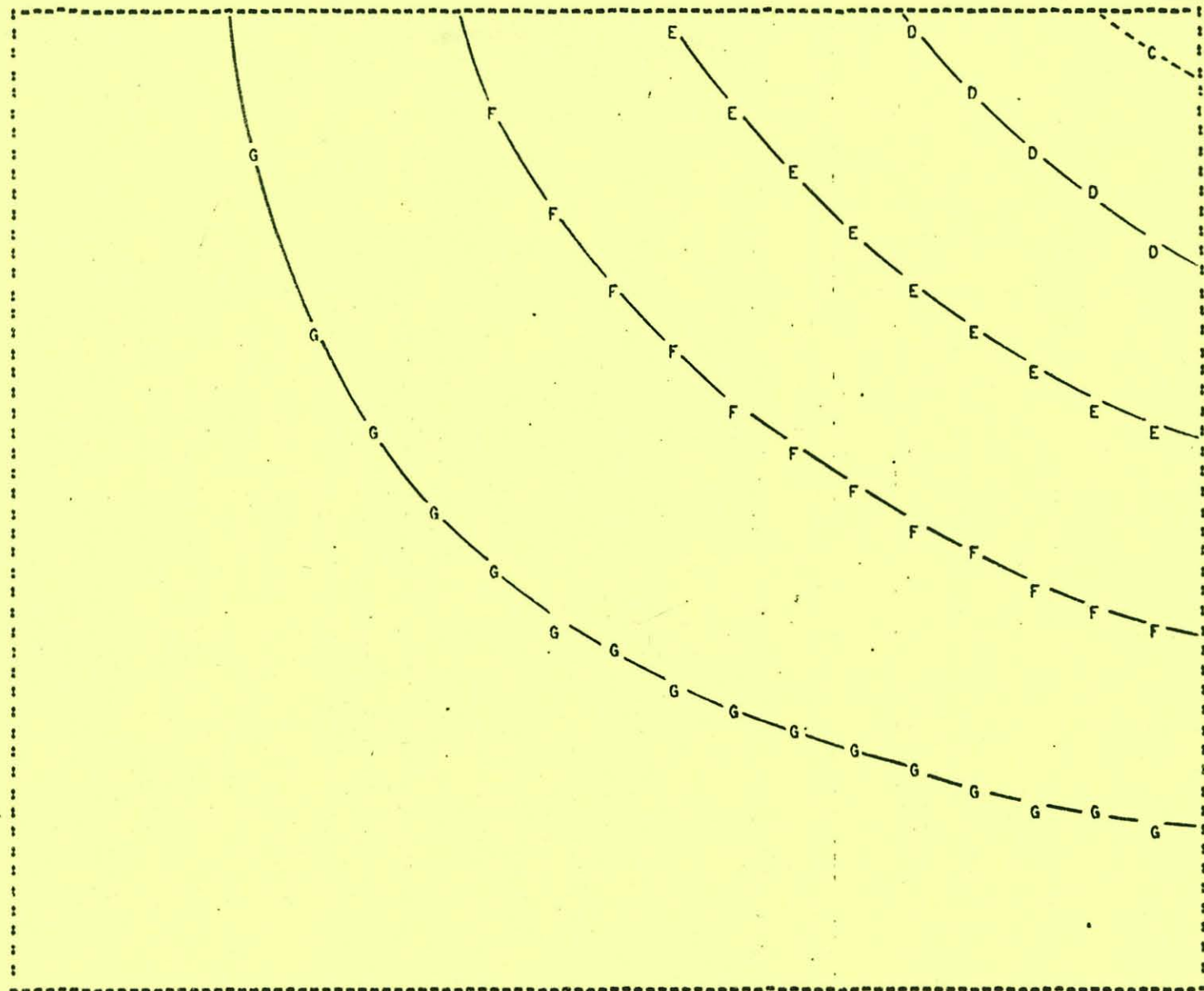
Interaction L x A-TRAIT: The RS contains the RL-range from C up to G, the slope is SW. The contour heights are equally spaced. The FR covers a higher range of RL's than the same interaction in RSA I, but the actual slope of the hyperplane seem to be virtually the same.

This is confirmed by the extension of the RS beyond the FR to include a range from -10 to +20. The resulting surface is a hyperbolic plane with the centre of the system approximately coinciding with the origin of both variables. The low-level slope is NE/SW, the high-level axis runs through NW/SE. Ascents and descents seem to be rather steep, with a high saddle point at H-level.

Interaction Q2 x Q3: Only three RL's (H to J) lie within the FR. The slope is NE, with the ascent steepening towards the optimum contour-line. The extended RS shows again a system of hyperbolics, with a saddle (H) coinciding roughly with the centre of the system in the origin of both variables. The descending slope is NW/SE, with the ascending axis joining at right angles.

The high degree of symmetry of both systems is again explained by the fact that no linear or square terms were included in the model.

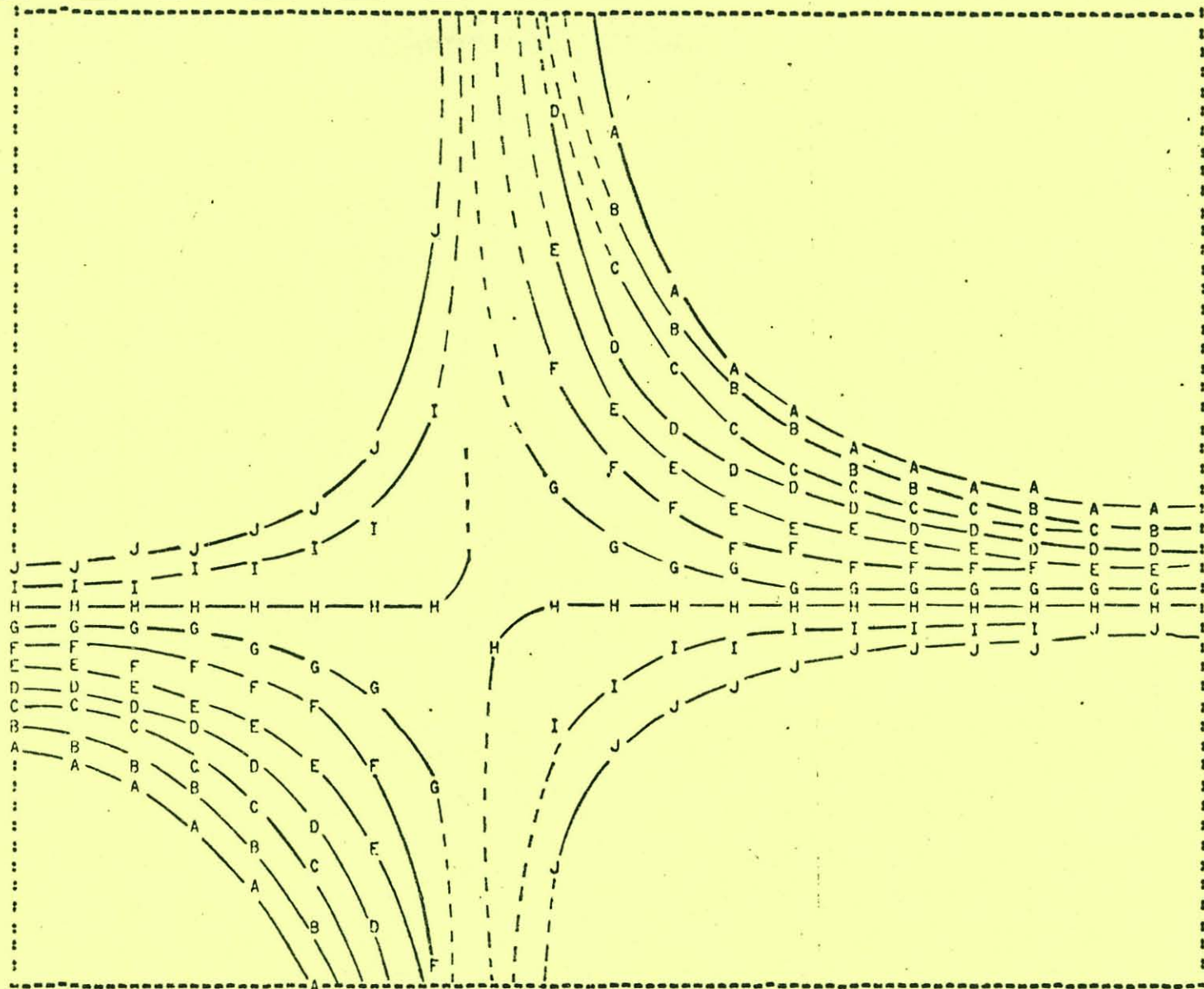
While the RS of L x A-TRAIT remained unchanged against RSAI, the interaction of Q2 x Q3 in this model points unmistakably in the direction of high performance due to high variable values in both variables. The optimum contour, however, also recognises 50% values in both variables, provided the other is at maximum.



RESPONSE SURFACE OF VARIABLE 1 ON THE Y-AXIS
AND VARIABLE 4 ON THE X-AXIS

KEY TO CONTOUR HEIGHTS (RESPONSE LEVELS):
 SYMBOLS: A B C D E F G H I J
 RESP. L.: 25 50 75 100 125 150 175 200 225 250

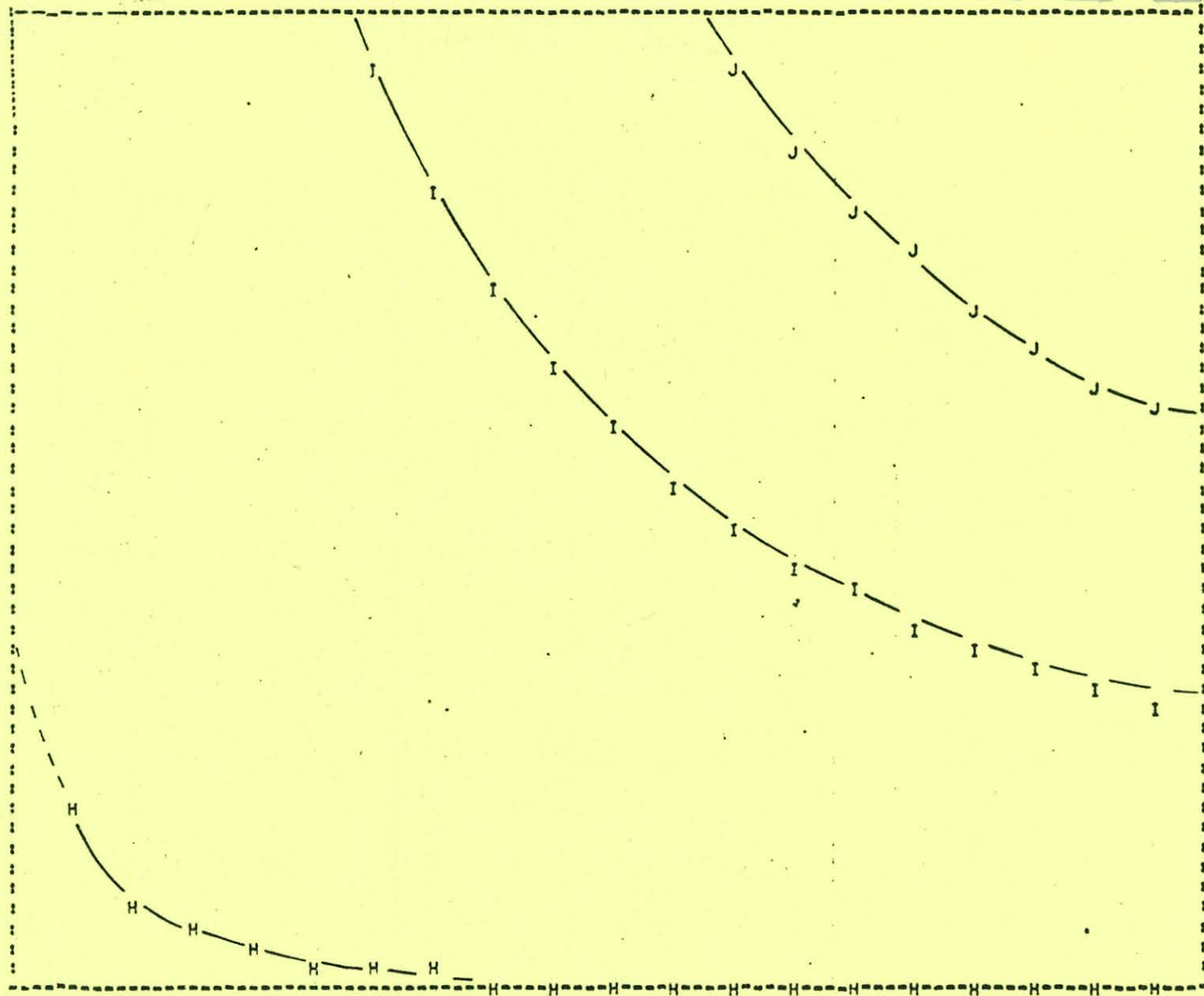
Fig. 10: 16PF-L (Var. 1) versus A-Trait (Var. 4) in the FS



RESPONSE SURFACE OF VARIABLE 1 ON THE Y-AXIS
AND VARIABLE 4 ON THE X-AXIS

KEY TO CONTOUR HEIGHTS (RESPONSE LEVELS):
 SYMBOLS: A B C D E F G H I J
 RESP·L·1 25 50 75 100 125 150 175 200 225 250

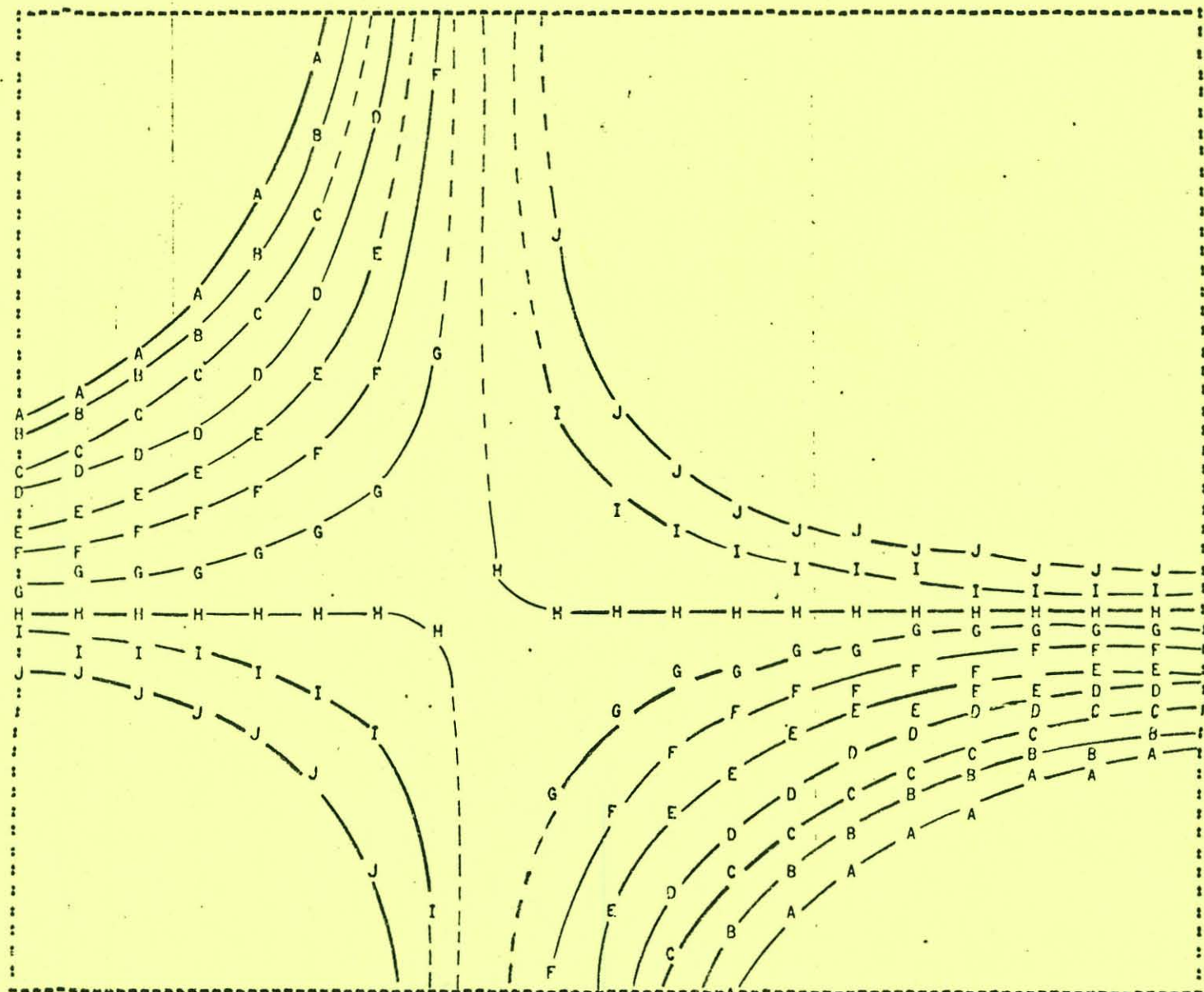
Fig. 11: 16PF-L (Var. 1) versus A-Trait (Var. 4) in the extended FS.



RESPONSE SURFACE OF VARIABLE 2 ON THE Y-AXIS
AND VARIABLE 3 ON THE X-AXIS

KEY TO CONTOUR HEIGHTS (RESPONSE LEVELS):
 SYMBOLS: A B C D E F G H I J
 RESP. L.: 25 50 75 100 125 150 175 200 225 250

Fig. 12: 16PF-Q2 (Var. 2) versus 16PF-Q3 (Var. 3) in the FS



RESPONSE SURFACE OF VARIABLE 2 ON THE Y-AXIS
AND VARIABLE 3 ON THE X-AXIS

KEY TO CONTOUR HEIGHTS (RESPONSE LEVELS):
 SYMBOLS: A B C D E F G H I J
 RESP. L.: 25 50 75 100 125 150 175 200 225 250

Fig. 13: 16PF-Q2 (Var. 2) versus 16PF-Q3 (Var. 3) in the extended FS

A comparison of the extended RS's shows that the RS yielded in this model is rotated by approximately 90° against the conic system obtained in the previous analysis.

Although, due to the dominant saddle area in the hyperplane of RSA II, which saturated virtually all of the FR with high-level response, there is no direct contradiction in the interpretation of the contour-systems, it was decided to disregard the RS of $Q2 \times Q3$ obtained in this analysis in any further interpretation. The corresponding RS of RSA I, yielded by a model which also includes linear and square terms, is regarded as more valid.

7. SUMMARY AND CONCLUSION

The Multiple Linear Regression Analysis has yielded a prediction equation with a reasonable Goodness of Fit and, furthermore, a fairly comprehensive set of psychologically plausible personality factors conducive to good operational performance under stress conditions could be derived.

The attempt to fit a second-order-model to the observational data did not prove successful with respect to predictive qualities. It was felt that the relatively small sample in conjunction with the drastic reduction of predictor variables could easily represent "too much coincidence", i.e. there might be a considerable probability that an unmeasurable amount of value configuration particular to the sample and amplified by the quadratic transformation influenced the outcome beyond the boundaries of routine significance considerations.

This suspicion was further underlined by the obtained RS's, which tended to be for one rather meagre and secondly possessed a fairly high degree of obviously artificial symmetry. Thirdly, the dominating form of hyperplane found turned out to be a saddle-plane, which is basically the least conclusive form of RS possible.

It was therefore decided to accept the model yielded by MLRA III as the main predictive link obtained from this project.

This linear model suggested a syndrome of personality-factors conducive to good operational performance which can be characterised by the following descriptive factor-clusters:

- EGO STABILITY CONTROL (low A-TRAIT, C+, L-, Q3)
- SELF ORIENTATION (Q2+, G-, I-)
- HYPER ACTIVATION-POTENTIAL (Q4+, high A-STATE) and, or minor importance
- CYCLOTHYMIA (A+)
- LOW INTELLIGENCE (RAVEN PM)

Using, and merging, CATELL's verbalizations of personality factors, the following - first - picture of the "emergency resistant operator" (in general terms) emerges:

He can be socially precise, and he tries to live up to his self-image. This, combined with the tendency to be rather resolute and being basically not an anxious person outlines a personality mainly characterised by a-neuroticism. Adding further to this impression is a certain capacity for emotional control with particular reference to exercising pragmatism where the realities of a situation may demand it.

Normally he is easy going and shows a fair degree of participating relaxedness.

The "rough edges" of this personality come through in a strong sense of self-reliance, expressed in a pronounced "no-nonsense" attitude. There is further a strong tendency to evade obligations and rules, the more when they collide with his own ideas. Also "underneath" the generally controlled, relaxed balance and the only thinly covered strong egocentrism, there is a moment of "hidden" tension and irritability, combined with a certain ability for arriving fairly rapidly at a relatively high level of (basically anxious) alertness in situations of a threatening character.

He need not be intelligent.

Although the findings of the MQRA and the RSA did not directly contribute to the main objective of this project, both methods proved very valuable as diagnostic tools, i.e. both methods of further analysis helped considerably in obtaining more insight into the structure of the above syndrome.

The findings of the Quadratic Regression principally underlined the outlines of the above syndrome.

In addition the component of EGO STABILITY CONTROL was clearly emphasised and in its infrastructure the interaction of L- and A-TRAIT was assigned a dominating weight above the other two factors indicated by the linear model.

The importance that the interaction of Q2 and Q3 was given by the MQRA results in a slight shift of character of the cluster SELF ORIENTATION: Under the aspect of interacting with Q3-, the predominantly positive character of self-sufficiency, rather strengthened by an element of ego-centrism, could now be seen as one solution of the "undisciplined self-conflict" that Q3 suggests.

The Response Surface Analysis served mainly to differentiate between the effects that the clusters have on performance levels, and further added to the understanding of the infrastructure of the syndrome.

The interaction of relaxedness and low anxiety(L x A-TRAIT) was again confirmed to be the basic cluster conducive to good operational performance. The response surface yielded by these two variables covered the largest section of response levels of all other variables.

Q2 and Q3, the two main components of the cluster SELF ORIENTATION, were found to be mainly indicators of the higher performance levels: interestingly, the RS suggested a combination of low Q3 and high Q2 as a set-variation conducive to favourable performance. The response surface created by A (Cyclothymia) and C (Ego Strength) indicated also some importance in the upper performance range only.

(Q4 and A-STATE were not included in the variables selected for the MQRA and RSA)

With respect to the dimensions Performance Level, Variable Influence and Score-Range, the findings of the MQRA and RSA can be summarized as follows:

- L and A-TRAIT, both low-score, are the broad foundation of the syndrome - of relatively high influence, but only for the low-to-medium-level performance range.

- The interaction of A and C, both in the above-average-score range, is interpreted as the enhancement of L x A-TRAIT (C+ is part of the cluster EGO STABILITY CONTROL) which effects the upper performance levels. Its importance is classed as minor to medium.
- Q2 x Q3, with Q2+ and Q3 rather in the low score area, are of minor importance as variables, but, since their syndrome seems to play a major part in the actual capacity of stress-tolerance, are predominantly effective in the medium-to-high performance class only.

Figure 14 is a graphic approximation of the assumed influence-spaces of the above variable combinations.

Incorporating the additional findings obtained by the MQRA and the RSA, the original syndrome as derived from the linear model can be reduced to three main clusters with supporting subsyndromes.

The main and basic set of personality factors needed to resist the operational stress in this experiment seems to be EGO STABILITY CONTROL, now defined as being basically consistent of low score L and low A-TRAIT. The extrovert element in A+ is to be seen as a strengthening agent to L-, as the overall emotional stability represented in C+ reinforces low anxiety. The interaction of A+ and C+ then would form the presupposition to achieve higher-performance levels.

The secondary condition needed to capacitate stress-resistance (at least in this experiment) is described as SELF ORIENTATION. It is based mainly on the interaction of Q2 and Q3, whereby the combination Q2+/Q3- seems to be preferable. Due to this particular configuration, the combination may be termed "unstable self-sufficiency", implying that Q2+ is at least partly a compensation for the low degree of social integration represented by Q3-. The supporting structure of the syndrome may be found in the subsyndrome of Ego-Centrism, represented by G- and I-.

The cluster termed HYPERACTIVATION POTENTIAL, consisting of Q4+ and high A-STATE, stays unchanged, since neither of its variables were included in the MQRA or RSA.

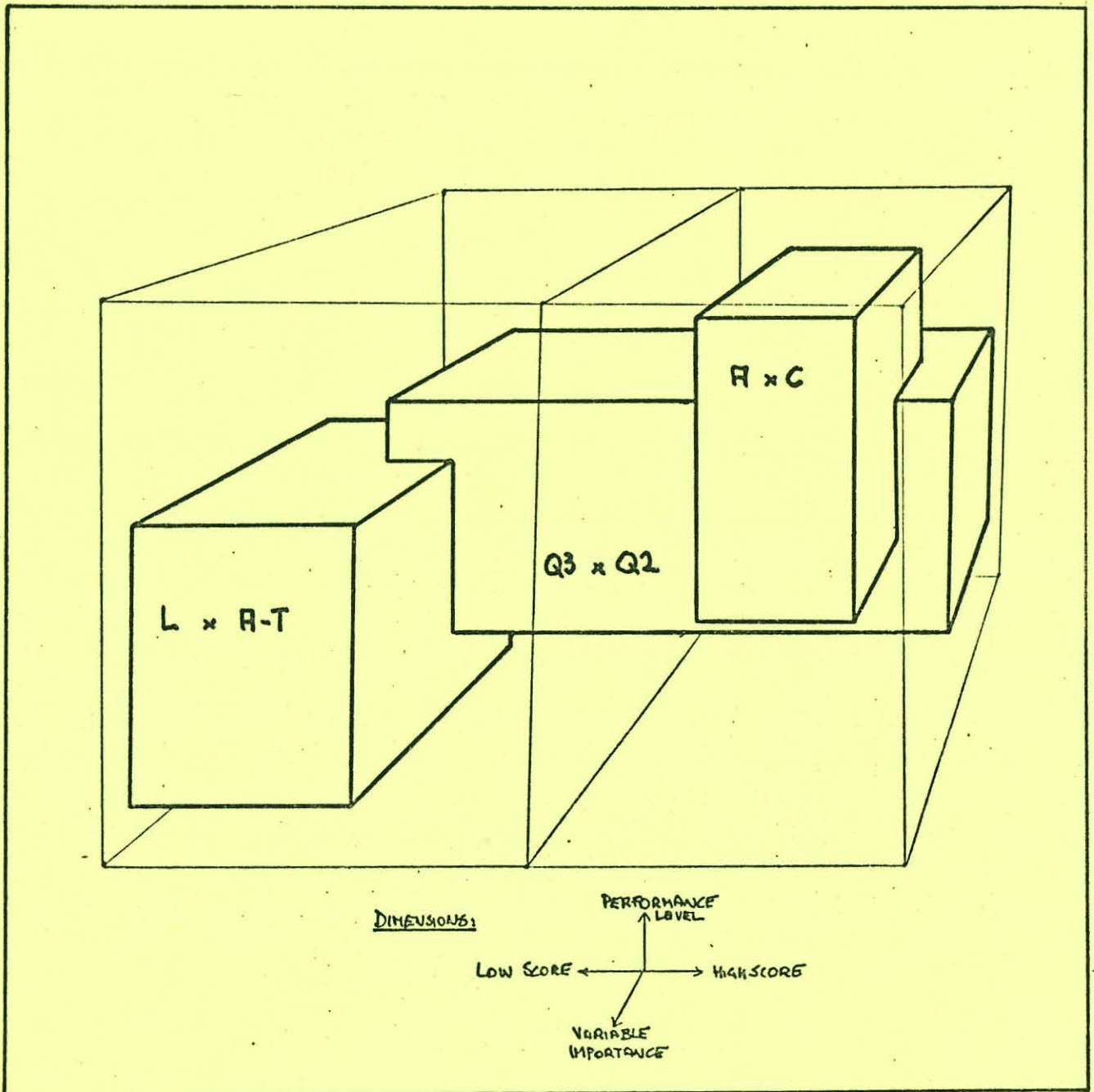


Fig.14. Graphic approximation of influence spaces of variable combinations

(Since it could be rightly argued that the fact that a lower degree of Intelligence was found to be conducive to higher performance in this experiment could be mainly due to the simplicity of the simulated operation, Intelligence is not included in this final clustering.)

The conclusions drawn above are - to some degree - supported by other findings in the field of research on performance under stress. DANIELS (1973) in his above cited article on parachutists found A+, C+ and Q2+ of importance in the endurance of stress. BUCKY & SPIELBERGER (1973) in investigating correlates to withdrawal from flight training found an interesting connection between the relative degree of A-STATE and different kinds of "flight training drop outs": a very high degree of A-STATE seemed to lead to voluntary withdrawals (because of 'physical' reasons) and a low degree of State-Anxiety was correlated to failure-drop out. This seems to underline the 'breaking point theory': too high a degree of A-STATE tends to support escape-reactions, while too low a degree of A-STATE represents an insufficient degree of alertness to cope with sudden changes.

The relevance of the cluster of self-orientation is underlined by findings of BIERSNER & RYMAN (1974) in their research on performance-prediction for scuba-divers. They found that successful divers are high in Q2, low in Q3 and low in G. (Their term for this syndrome was "asocial").

No support could be found in the literature for the combination of L- and low A-TRAIT.

Due to the fact that in the literature there is only narrow support to be found for the conclusions come to in this project, and considering the small sample this project was based on, it seems only fair to classify the above described findings and derivations as pilot results only, and a guide to future research in this field.

Some recommendations for future research are given in the next section.

8. RECOMMENDATIONS FOR FUTURE RESEARCH

If the findings and conclusions in this study should be investigated further, the following list of recommendations may prove helpful:

- To verify the validity of the obtained linear model (in order to determine any inherent weaknesses) in a controlled experiment.
- To use a larger sample to improve the quality of the statistics obtained.
- To use a sufficiently large sample to be able to construct a Central Composite Design of the experiment in order to deploy the full capacity of Response Surface Methodology.
- To use specialised scales to measure the personality factors indicated in this project.
- To use sophisticated scaling methods in trying to fit a higher-order-model.
- To develop a better method for measuring performance in the task.

APPENDICES

APPENDIX A

A.1. Flow diagram of the CHEMICAL PLANT SIMULATOR

A.2. Description of Simulator-Task

A.2. Description of Simulator-Task

The following instruction was issued (verbally) to the S's:

"You are handed this plant over in perfect running condition. We expect you to hand over the plant to your successor in exactly the same condition. Your shift is 30 min. long. During the shift faults and defects will occur.

You have to fix them all up within the 30 min., which is quite possible if you keep your cool. The clock on the wall gives you the elapsed time, and the training officer will tell you the elapsed time in five minute intervals. He will also keep a score of how you are doing. These are the safety regulations; any violation will bring your score down. If you make a lethal mistake, the experiment will be terminated immediately".

Then followed detailed instructions on what pressure-, flow-rates and tank-level-readings were to be maintained and in which areas of the 'plant' to wear specific safety equipment, e.g. goggles, etc.

The table overleaf shows a detailed breakdown of the standardized series of faulty conditions:

<u>Time*)</u>	<u>Fault</u>	<u>Indication of Fault</u>	<u>Corrective action</u>
2	Pump 2 stopped PAV I inoperable FAV I inoperable	Main circulation comes to a standstill, no flow-readings available	Isolate valves a to f, start up pump, establish bypasses around PAV I, FAV I
8	PAV I reset FAV I reset LAV II fully open	Pressure rates drop (ALARM), flow rates increase (ALARM), level in T103 increases (ALARM)	Open LAV III and establish bypass, isolate valve h, close LAV II - adjust flow, pressure/levels
11	Pump I stopped LAV II inoperable	Level in 103 drops (ALARM), level in 101 increases (ALARM)	Adjust flows/levels, open valve h, close bypass around LAV III, adjust LAV III, isolate valve i, Start up pump, adjust LAV I
16	FAV I inoperable	Level in 101 drops (slowly), pressure increases rapidly (ALARM)	Establish FAV I bypass, adjust flow rates
20	Pump 2 stopped PAV I inoperable	Level in 103 drops (ALARM), level in 101 drops, level in 102 increases (ALARM)	Isolate valves a to f, start up pump, establish bypass around PAV I, balance and adjust flows

*) = Legend: See overleaf

Legend: TIME : elapsed time in minutes; at this point in time the fault-conditions were set via a remote-control-console

INDICATION OF FAULT :

where 'ALARM' is mentioned the change in conditions occurred so rapidly that on average the appropriate alarm would go off.

The alarms were set to very narrow limits around the levels to be maintained and were set to maximum loudness (siren) and increased flash-rate (visual indication).

APPENDIX B

- B.1. Original Data Matrix
- B.2. Analysis of Variance
- B.3. Mean profile of 16PF scores
- B.4. Distribution of the variables EXPERIENCE and AGE
- B.5. STEN - Transformation of STAI- and RAVEN (PM) - scores
- B.6. Distribution of the simulator - score
- B.7. Scattergram and split-half-regression of the simulator - score
- B.8. Distribution of the sample split

ORIGINAL DATA MATRIX

OBS	NO	PF/A	PF/B	PF/C	PF/E	PF/F	PF/G	PF/H	PF/I	PF/L	PF/M	PF/N	PF/O	P/01	PF02	P/03	P/04	STATE	TRAIT	INT	AGE	EXP	SCORE
1	6	10	4	6	2	6	6	6	6	8	3	6	5	7	7	6	4	5	6	42	0	188	
2	8	8	5	6	8	7	5	3	6	4	4	4	8	4	6	3	5	3	6	26	0	223	
3	8	9	2	5	5	3	4	7	9	6	4	6	2	7	1	9	8	4	6	23	0	169	
4	5	8	5	6	5	4	3	5	10	6	7	6	4	8	3	9	5	2	6	33	4	193	
5	7	8	4	7	8	3	5	7	8	7	4	6	6	5	5	5	7	5	6	21	0	190	
6	6	8	7	8	3	6	7	4	6	3	8	3	9	5	7	4	5	2	7	36	1	195	
7	6	8	4	9	7	3	10	6	6	6	3	7	7	10	3	9	5	4	5	39	15	231	
8	5	8	4	4	3	4	2	5	6	5	3	8	7	8	3	5	7	5	6	39	3	170	
9	5	9	2	4	3	2	3	4	3	5	6	8	5	9	2	7	7	8	8	40	2	187	
10	4	6	8	6	9	7	8	5	5	3	4	6	6	4	5	6	2	3	9	24	0	161	
11	5	8	2	4	3	7	3	5	6	7	7	8	4	8	7	6	4	6	5	29	3	193	
12	5	4	7	6	6	4	7	8	9	5	4	7	3	5	2	8	6	7	5	31	4	138	
13	6	3	8	5	5	4	6	7	8	6	3	6	4	6	4	8	6	6	6	28	3	142	
14	5	3	4	5	6	5	4	7	7	5	5	6	5	5	4	7	6	5	7	22	0	155	
15	6	8	4	3	6	6	2	4	4	5	8	3	7	5	8	2	3	4	7	32	0	228	
16	7	7	3	4	7	8	5	4	3	6	7	4	8	6	8	4	2	3	9	27	3	207	
17	6	6	2	5	6	7	4	5	4	5	7	5	7	7	7	3	6	5	8	23	4	222	
18	5	5	4	6	10	7	7	6	6	5	5	3	5	3	5	6	8	6	4	24	0	174	
19	7	6	7	5	7	2	7	3	9	6	1	5	5	5	7	5	3	1	6	24	0	205	
20	5	9	6	7	7	6	7	8	8	2	6	5	5	4	3	3	5	5	7	21	0	101	
21	5	9	3	9	6	6	3	6	6	4	6	6	3	8	5	5	3	5	6	26	0	165	
22	3	5	4	1	4	4	4	5	2	2	10	5	5	5	9	3	5	5	7	36	3	235	
23	6	4	3	2	2	4	3	7	8	3	7	6	5	7	4	7	7	8	7	24	0	115	
24	6	8	5	7	5	8	7	4	5	4	6	6	6	4	9	5	6	7	8	23	0	196	
25	6	7	4	4	5	3	5	6	6	4	8	5	6	3	6	6	7	3	24	3	159		
26	7	7	8	6	6	7	5	1	3	4	9	1	8	9	5	2	1	9	26	2	251		
27	7	8	6	4	6	7	6	4	9	4	8	8	5	8	9	6	3	2	7	24	2	231	
28	5	5	5	6	5	6	3	4	9	8	4	8	7	8	4	7	7	7	9	25	1	140	
29	7	7	5	2	4	7	7	5	6	5	8	4	3	5	5	4	3	5	37	2	182		
30	7	9	4	6	3	3	1	6	8	5	2	9	6	7	3	10	6	7	23	0	180		
31	6	8	3	5	5	4	3	5	6	8	8	7	4	7	2	6	3	5	7	35	7	163	
32	4	3	4	8	6	4	5	6	10	3	6	7	2	3	7	7	7	7	6	32	0	121	
33	3	6	8	4	4	9	4	4	7	6	7	5	8	8	6	5	6	5	7	23	0	172	
34	7	6	7	6	5	6	8	3	6	3	3	5	5	7	7	5	1	3	9	39	13	230	
35	7	8	5	6	5	4	6	6	6	5	4	6	2	7	3	5	4	4	10	20	0	179	
36	9	9	9	5	6	6	6	7	5	6	9	3	7	7	5	4	4	6	9	27	0	223	
37	6	7	1	4	4	5	4	7	2	8	5	5	7	7	9	6	5	6	6	28	0	199	
38	8	8	7	5	6	5	6	8	6	5	5	4	7	5	9	7	6	7	5	49	3	210	
39	7	6	5	5	4	4	4	6	7	7	4	8	6	8	7	5	5	5	7	23	3	205	
40	5	7	6	7	8	4	7	7	6	7	6	4	6	8	5	5	7	6	9	23	0	189	
41	6	8	2	7	8	6	4	6	9	7	1	9	3	4	5	9	5	8	9	32	0	95	
42	5	5	5	3	3	6	2	6	6	7	6	9	5	9	3	9	8	9	4	35	8	181	
43	6	5	5	7	5	3	4	4	6	7	6	6	5	3	5	6	4	6	3	33	9	188	
44	6	8	4	3	3	7	6	6	6	5	4	6	5	7	5	8	5	8	6	46	4	139	
45	7	8	3	8	7	4	4	4	4	2	6	6	5	4	9	7	5	7	5	27	6	218	
46	6	5	3	4	3	5	4	6	6	6	3	6	5	5	6	7	6	7	7	22	0	162	
47	5	7	7	5	7	6	4	6	8	4	4	3	8	7	7	8	5	3	9	32	6	241	
48	3	7	5	6	6	4	2	6	7	4	6	9	6	8	5	9	6	6	6	32	8	219	
49	3	10	3	8	5	4	5	6	7	9	5	5	5	6	5	10	3	4	7	23	0	169	
50	6	5	5	4	8	4	3	7	7	4	7	4	4	6	8	6	5	7	8	26	0	173	
51	7	7	10	6	3	10	6	5	4	8	7	3	7	5	9	2	2	1	7	31	0	259	
52	4	8	5	8	10	6	3	9	6	3	6	7	5	7	5	6	7	5	6	23	0	178	

98.

ANALYSIS OF VARIANCE - NEW STARTERS VS. RETRAINED OPERATORS
ORIGINAL DATA MATRIX

OBS.NO. PF/A PF/B PF/C PF/E PF/F PF/G PF/H PF/I PF/L PF/M PF/N PF/O P/Q1 PFQ2 P/Q3 P/Q4 STATE TRAIT INT. AGE EXP. SCORE

GROUP ONE: NEW STARTERS

1	6.	5.	5.	4.	8.	4.	3.	7.	7.	4.	7.	4.	4.	6.	8.	6.	5.	7.	8.	173.
2	4.	6.	8.	6.	9.	7.	8.	5.	5.	3.	4.	6.	6.	4.	5.	6.	2.	3.	9.	161.
3	6.	4.	3.	2.	2.	4.	3.	7.	8.	3.	7.	6.	5.	7.	4.	7.	7.	8.	7.	115.
4	6.	7.	4.	4.	5.	3.	5.	6.	6.	6.	4.	8.	5.	6.	3.	6.	6.	7.	3.	159.
5	3.	10.	3.	8.	5.	4.	5.	6.	7.	9.	5.	5.	5.	6.	5.	10.	3.	4.	7.	169.
6	7.	8.	5.	6.	5.	4.	6.	6.	6.	5.	4.	6.	2.	7.	3.	5.	4.	4.	10.	179.
7	5.	8.	2.	4.	3.	7.	3.	5.	6.	7.	7.	8.	4.	8.	7.	6.	4.	6.	5.	193.
8	6.	8.	5.	7.	5.	8.	7.	4.	5.	4.	6.	6.	6.	4.	9.	5.	6.	7.	8.	106.
9	3.	7.	5.	6.	6.	4.	2.	6.	7.	4.	6.	9.	6.	8.	5.	9.	6.	6.	6.	219.
10	9.	9.	9.	5.	6.	6.	6.	7.	5.	6.	9.	3.	7.	7.	5.	4.	4.	6.	9.	223.
11	7.	6.	7.	6.	5.	6.	8.	3.	6.	3.	3.	5.	5.	7.	7.	5.	1.	3.	9.	230.
12	3.	5.	4.	1.	4.	4.	4.	5.	2.	2.	10.	5.	5.	5.	9.	3.	5.	5.	7.	235.
13	4.	3.	4.	8.	6.	4.	5.	6.	10.	3.	6.	7.	2.	3.	7.	7.	7.	7.	6.	121.
14	6.	5.	3.	4.	3.	5.	4.	6.	6.	6.	3.	6.	5.	5.	6.	7.	6.	7.	7.	162.
15	3.	6.	8.	4.	4.	9.	4.	4.	7.	6.	7.	5.	8.	8.	6.	5.	6.	5.	7.	172.
16	6.	7.	1.	4.	4.	5.	4.	7.	2.	8.	5.	5.	7.	7.	9.	6.	5.	6.	6.	190.
17	5.	7.	7.	5.	7.	6.	6.	4.	8.	4.	4.	3.	8.	7.	7.	8.	5.	3.	9.	241.

GROUP TWO: RETRAINED OPERATORS

18	5.	9.	6.	7.	7.	6.	7.	8.	8.	2.	6.	5.	5.	4.	3.	3.	5.	5.	7.	101.
19	5.	4.	7.	6.	6.	4.	7.	8.	9.	5.	4.	7.	3.	5.	2.	8.	6.	7.	5.	138.
20	6.	8.	4.	3.	3.	7.	6.	6.	6.	5.	4.	6.	5.	7.	5.	8.	5.	8.	6.	139.
21	6.	3.	8.	5.	5.	4.	6.	7.	8.	6.	3.	6.	4.	6.	4.	8.	6.	6.	6.	142.
22	5.	3.	4.	5.	6.	5.	4.	7.	7.	5.	5.	6.	5.	5.	4.	7.	6.	5.	7.	155.
23	8.	9.	2.	5.	5.	3.	4.	7.	9.	6.	4.	6.	2.	7.	1.	9.	8.	4.	6.	169.
24	5.	5.	4.	6.	10.	7.	7.	6.	6.	6.	5.	5.	3.	5.	6.	6.	8.	6.	4.	174.
25	7.	8.	4.	7.	8.	3.	5.	7.	8.	7.	4.	6.	6.	5.	5.	5.	7.	5.	6.	190.
26	5.	8.	5.	6.	5.	4.	3.	5.	10.	6.	7.	6.	4.	8.	3.	9.	5.	2.	6.	193.
27	6.	8.	7.	8.	3.	6.	7.	4.	6.	3.	8.	3.	9.	5.	7.	4.	5.	2.	7.	195.
28	7.	6.	7.	5.	7.	2.	7.	3.	9.	6.	1.	5.	5.	5.	7.	5.	3.	1.	6.	205.
29	8.	8.	7.	5.	6.	5.	6.	8.	6.	5.	5.	4.	7.	5.	9.	7.	6.	7.	5.	210.
30	7.	8.	3.	8.	7.	4.	4.	4.	4.	2.	6.	6.	5.	4.	9.	7.	5.	7.	5.	218.
31	7.	7.	10.	6.	3.	10.	6.	5.	4.	8.	7.	3.	7.	5.	9.	2.	2.	1.	7.	259.
32	6.	8.	2.	7.	8.	6.	4.	6.	9.	7.	1.	9.	3.	4.	5.	9.	5.	8.	9.	95.
33	5.	5.	5.	6.	5.	6.	3.	4.	9.	8.	4.	8.	7.	8.	4.	7.	7.	7.	9.	140.
34	6.	8.	3.	5.	5.	4.	3.	5.	6.	8.	8.	7.	4.	7.	2.	6.	3.	5.	7.	163.
35	5.	9.	3.	9.	6.	6.	3.	6.	6.	4.	6.	6.	3.	8.	5.	5.	3.	5.	6.	165.
36	5.	8.	4.	4.	3.	4.	2.	5.	6.	5.	3.	8.	7.	8.	3.	5.	7.	5.	6.	170.
37	4.	8.	5.	8.	10.	6.	3.	9.	6.	3.	6.	7.	5.	7.	5.	6.	7.	5.	6.	178.
38	7.	9.	4.	6.	3.	3.	1.	6.	8.	5.	2.	9.	6.	7.	3.	10.	6.	7.	7.	180.
39	5.	5.	5.	3.	3.	6.	2.	6.	6.	7.	6.	9.	5.	9.	3.	9.	8.	9.	4.	181.
40	7.	7.	5.	2.	4.	7.	7.	5.	6.	5.	8.	4.	3.	5.	5.	5.	4.	3.	5.	182.
41	5.	9.	2.	4.	3.	2.	3.	4.	3.	5.	6.	8.	5.	9.	2.	7.	7.	8.	8.	187.
42	6.	10.	4.	6.	2.	6.	6.	6.	6.	8.	3.	6.	5.	7.	7.	6.	4.	5.	6.	188.
43	6.	5.	5.	7.	5.	3.	4.	4.	6.	7.	6.	6.	5.	3.	5.	6.	4.	6.	3.	188.
44	5.	7.	6.	7.	8.	4.	7.	7.	6.	7.	6.	4.	6.	8.	5.	5.	7.	6.	9.	189.
45	7.	6.	5.	5.	4.	4.	4.	6.	7.	7.	4.	8.	6.	8.	7.	5.	5.	5.	7.	205.
46	7.	7.	3.	4.	7.	8.	5.	4.	3.	6.	7.	4.	8.	6.	8.	4.	2.	3.	9.	207.
47	6.	6.	2.	5.	6.	7.	4.	5.	4.	5.	7.	5.	7.	7.	7.	3.	6.	5.	8.	222.
48	8.	8.	5.	6.	8.	7.	5.	3.	6.	4.	4.	8.	4.	6.	3.	5.	3.	6.	6.	223.
49	6.	8.	4.	3.	6.	6.	2.	4.	4.	5.	8.	3.	7.	5.	8.	2.	3.	4.	7.	228.
50	6.	8.	4.	9.	7.	3.	10.	6.	6.	6.	3.	7.	7.	10.	3.	9.	5.	4.	5.	231.
51	7.	8.	6.	4.	6.	7.	6.	4.	9.	4.	8.	8.	5.	8.	9.	6.	3.	2.	7.	231.
52	7.	7.	8.	6.	6.	7.	5.	1.	3.	4.	9.	1.	8.	9.	5.	2.	2.	1.	9.	251.

SOURCE	SS	DF	MS	F
BETWEEN SUBJ	3599.162			
FACTOR A	1.108	1	1.108	0.02
SUBJ W GRP	3598.054	50	71.961	
WITHIN SUBJ	*****			
FACTOR B	*****	19	84015.256	*****
FACTOR AB	55.412	19	2.916	0.04
B X SS W GRP	69396.619	950	73.049	

SIMPLE EFFECTS ANALYSIS

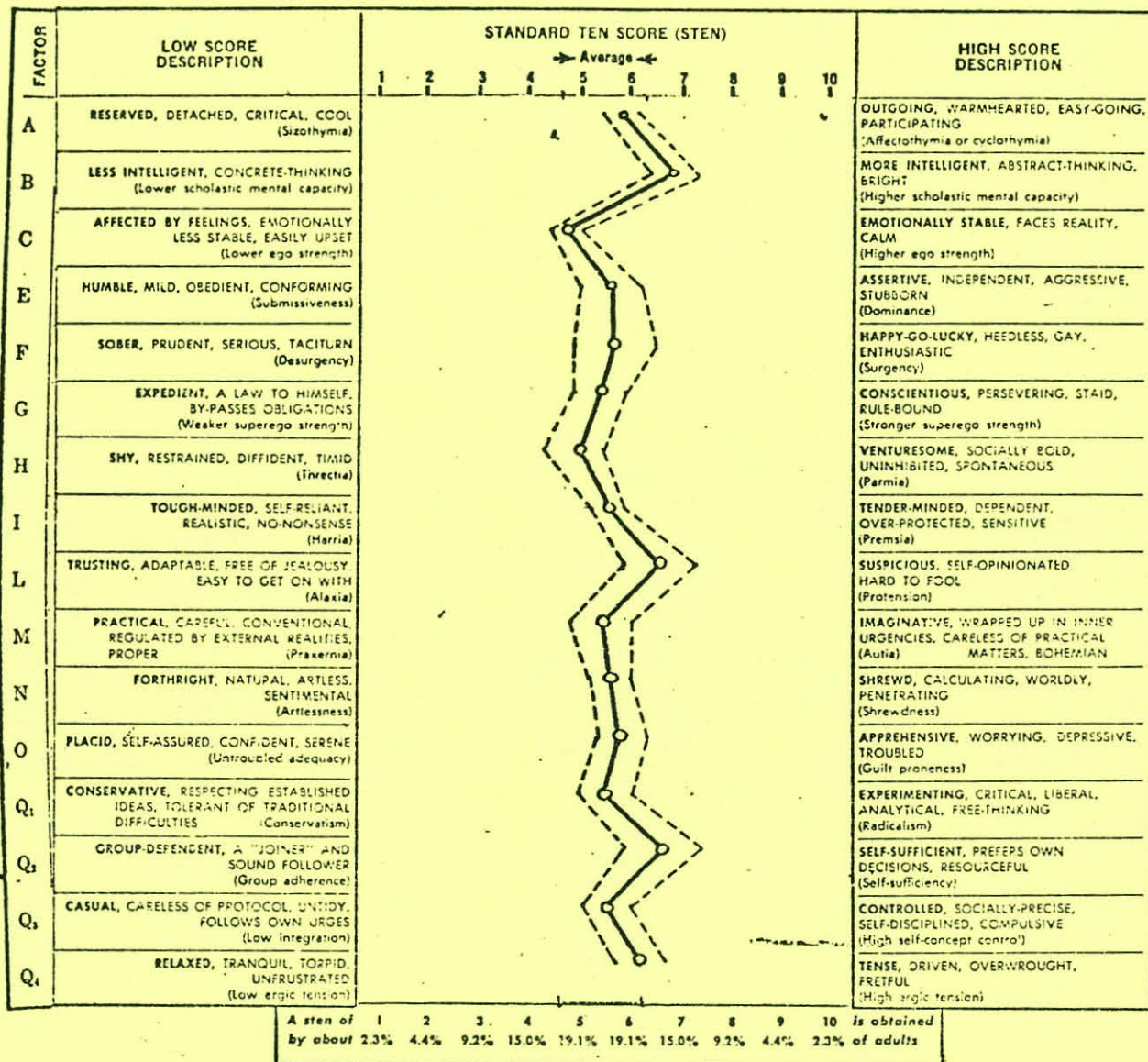
FACTOR A AT LEVEL:	SS	DF	MS	F
B 1	8.28	1	8.28	0.11
B 2	4.31	1	4.31	0.06
B 3	0.08	1	0.08	0.00
B 4	5.87	1	5.87	0.08
B 5	2.66	1	2.66	0.04
B 6	0.10	1	0.10	0.00
B 7	0.08	1	0.08	0.00
B 8	0.06	1	0.06	0.00
B 9	1.56	1	1.56	0.02
B10	4.17	1	4.17	0.06
B11	2.30	1	2.30	0.03
B12	0.26	1	0.26	0.00
B13	0.21	1	0.21	0.00

B14	0.43	1	0.43	0.01
B15	11.56	1	11.56	0.16
B16	0.62	1	0.62	0.01
B17	1.17	1	1.17	0.02
B18	4.33	1	4.33	0.06
B19	6.93	1	6.93	0.09
B20	1.55	1	1.55	0.02
ERROR	72994.67	1000	72.99	

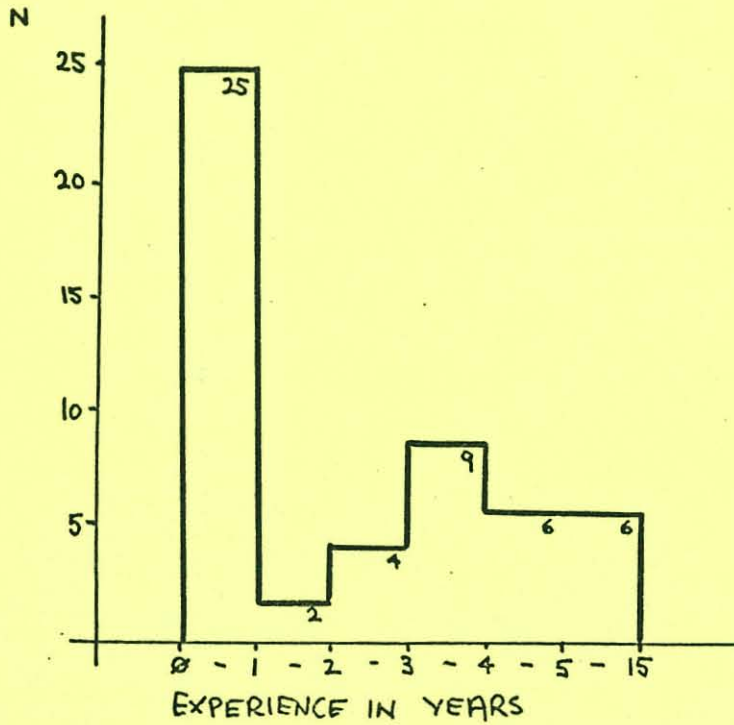
FACTOR B AT LEVEL:	SS	DF	MS	F
A 1	520662.05	19	27403.27	*****
A 2	1075683.23	19	56614.91	*****
ERROR	69396.62	950	73.05	

100.

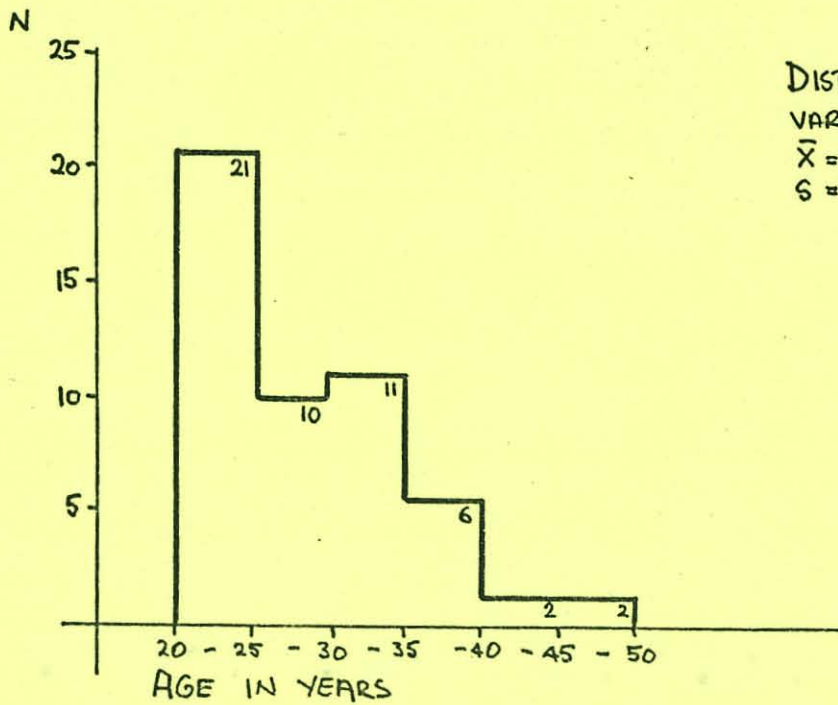
APP. B.2. / 2



.B.3. Mean profile of 16PF-scores

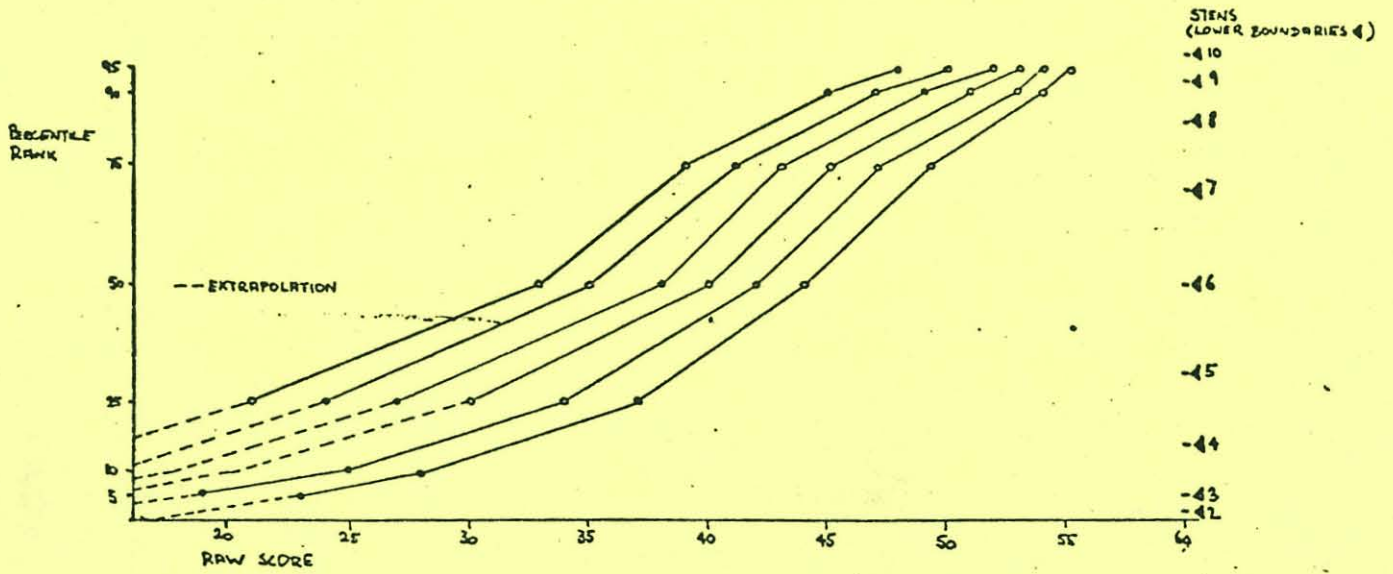


DISTRIBUTION
VAR. EXPERIENCE
 $\bar{X} = 2,35$
 $S = 3,41$

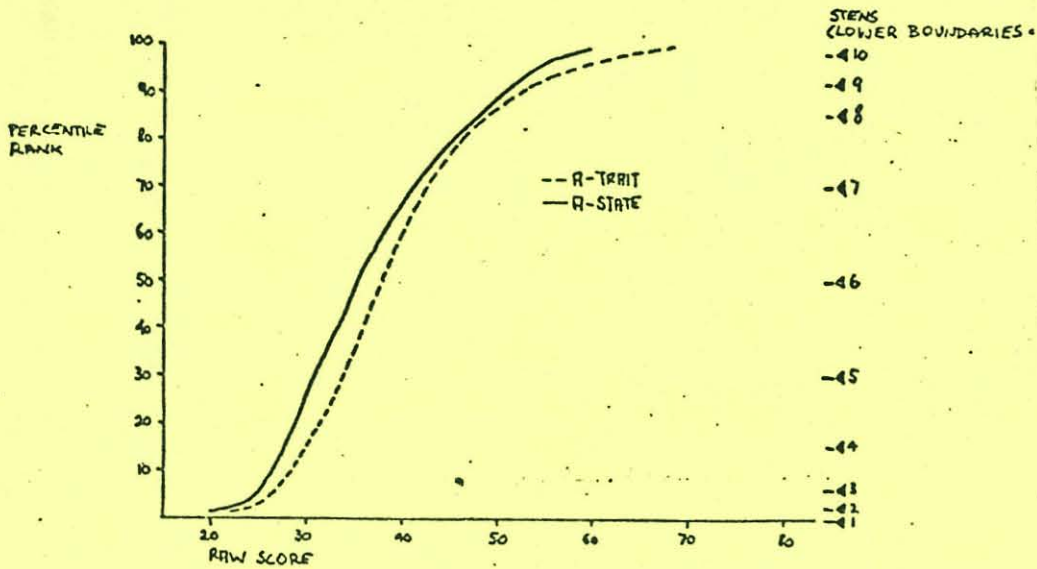


DISTRIBUTION
VAR. AGE
 $\bar{X} = 29,10$
 $S = 6,92$

B.4. Distribution of the variables EXPERIENCE and AGE



STEN-TRANSFORMATION OF RAW-SCORES OF RAVEN'S PMT



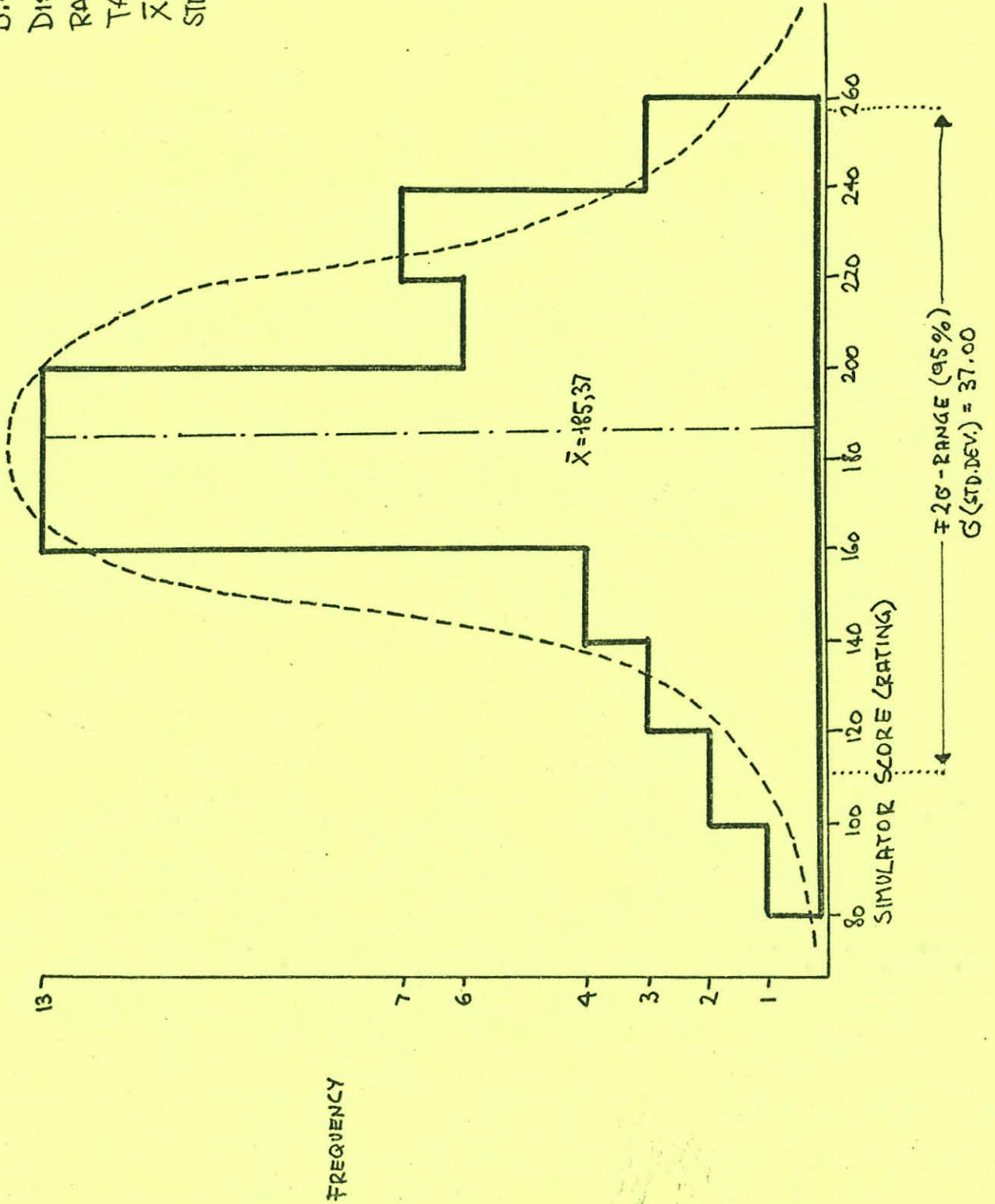
STEN-TRANSFORMATION OF RAW-SCORES OF THE STAI

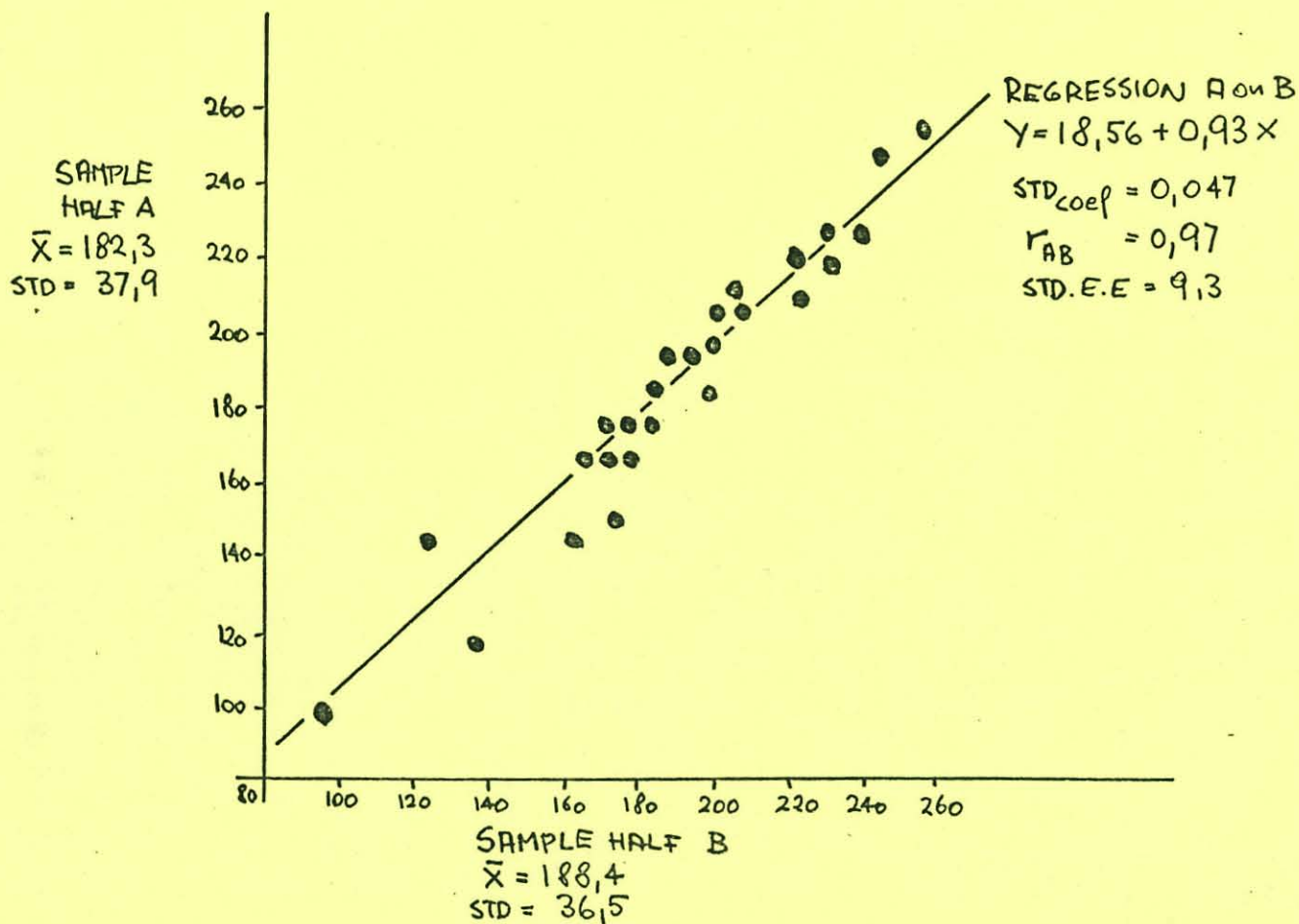
B.5. STEN - Transformation of STAI- and RAVEN (PM) - scores

B.6.
 DISTRIBUTION OF
 RATING-SCORES FOR
 TASK PERFORMANCE
 $\bar{X} = 185,37$
 STD = 37,00

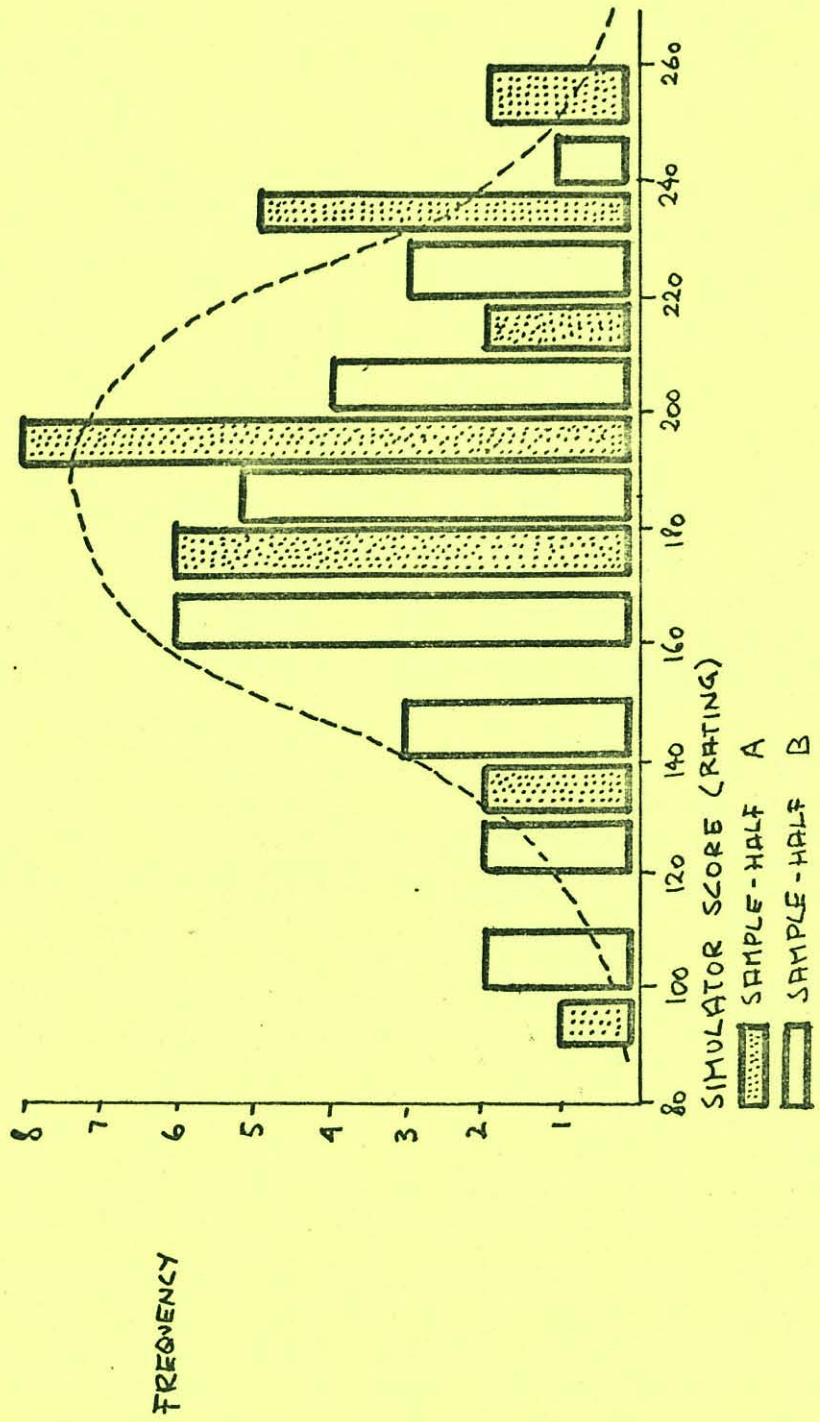
104.

$$\chi^2 = 9,26 < \chi^2_{95}(6df) = 12,16$$





B.7. Scattergram and split-half-regression of the simulator-score



B.8. Distribution of the sample-split

APPENDIX C

- C.1. Results of MLRA I
- C.2. Results of MLRA II
- C.3. MLRA III - full analysis
- C.4. MLRA III - Plot and histogram of residuals
- C.5. MLRA IV - Transformed data matrix and results.

STP NUMBER 21 ENTER VARIABLE 12
 STANDARD ERROR OF ESTIMATE= 14.478
 MULTIPLE CORRELATION COEFFICIENT = 0.954
 GOODNESS OF FIT: 90.9922%

CONSTANT TERM= 127.0743

VAR	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	VAR. CONTR		
1	.4682E 01	.1868E 01	.6283E 01	.1700E 00	
2	.5069E 01	.1753E 01	.8360E 01	.2418E 00	
3	.6913E 01	.1554E 01	.1978E 02	.3732E 00	
4	-.1100E 01	.1551E 01	.5034E 00	-.5279E -01	
5	.1203E 01	.1367E 01	.7728E 00	.6356E -01	
6	-.3429E 01	.1487E 01	.5313E 01	-.1648E 00	
7	-.3876E 01	.1582E 01	.6001E 01	-.1976E 00	
8	-.2978E 01	.1828E 01	.2654E 01	-.1240E 00	
9	-.8884E 01	.1729E 01	.2640E 02	-.4624E 00	
10	.2058E 01	.1385E 01	.2208E 01	.2685E -01	
11	.9671E 00	.1481E 01	.4263E 00	.5336E -01	
12	-.1956E 00	.2033E 01	.9256E -02	-.9655E -02	
13	-.2236E 01	.1851E 01	.1459E 01	-.1031E 00	
14	.2891E 01	.1810E 01	.2557E 01	.1318E 00	
15	.9259E 01	.1275E 01	.5274E 02	.5485E 00	
16	.2380E 01	.1723E 01	.1921E 01	.1331E 00	
17	.8494E 01	.2133E 01	.1586E 02	.3962E 00	
18	-.9902E 01	.1861E 01	.2833E 02	-.5344E 00	
19	.3683E 00	.1680E 01	.4802E -01	.1605E -01	
20	-.7079E 00	.4263E 00	.2757E 01	-.1325E 00	
21	.4470E 01	.1090E 01	.1681E 02	.4115E 00	

STEP NUMBER 19 ENTER VARIABLE 13
 STANDARD ERROR OF ESTIMATE= 17.540
 MULTIPLE CORRELATION COEFFICIENT = 0.927
 GOODNESS OF FIT: 85.8968%

CONSTANT TERM= 108.8429

VAR	COEFF	STD DEV COEFF	F-RATIO VAR. CONTR	BETA COEFF
1	.6396E 01	.2194E 01	.8498E 01	.2323E 00
2	.1707E 01	.1809E 01	.8902E 00	.8140E -01
3	.5972E 01	.1845E 01	.1048E 02	.3224E 00
4	.3522E 00	.1798E 01	.3838E -01	.1690E -01
5	.2429E 01	.1561E 01	.2422E 01	.1283E 00
6	-.4482E 01	.1773E 01	.6391E 01	-.2154E 00
7	-.1756E 01	.1785E 01	.9679E 00	-.8951E -01
8	-.4703E 01	.2142E 01	.4821E 01	-.1959E 00
9	-.9720E 01	.2074E 01	.2197E 02	-.5059E 00
10	.8475E 00	.1628E 01	.2710E 00	.3989E -01
11	.2305E 01	.1748E 01	.1738E 01	.1272E 00
12	.2208E 01	.2335E 01	.8943E 00	.1090E 00
13	.2336E 00	.2122E 01	.1212E -01	.1077E -01
14	.5997E 01	.1991E 01	.9077E 01	.2734E 00
15	.8239E 01	.1508E 01	.2984E 02	.4881E 00
16	.4313E 01	.1974E 01	.4774E 01	.2404E 00
17	.4900E 01	.2345E 01	.4366E 01	.2286E 00
18	-.9252E 01	.2196E 01	.1772E 02	-.4993E 00
19	-.2220E 01	.1815E 01	.1497E 01	-.9675E -01

APP. C. 2.
 RESULTS OF MLRA II

PERSONALITY CORRELATES TO PERFORMANCE UNDER STRESS

H.P. LEHMANN

ORIGINAL DATA MATRIX

OBS.	NO.	PF/A	PF/B	PF/C	PF/E	PF/F	PF/G	PF/H	PF/I	PF/L	PF/M	PF/N	PF/O	P/O1	PF02	P/O3	P/O4	STAT	TRAIT	INT.	AGE	EXP.	SCORE
1	6.	5.	5.	4.	8.	4.	3.	7.	7.	4.	7.	4.	4.	6.	8.	6.	5.	7.	8.				173.
2	4.	6.	8.	6.	9.	7.	8.	5.	5.	3.	4.	6.	6.	4.	5.	6.	2.	3.	9.				181.
3	6.	4.	3.	2.	2.	4.	3.	7.	8.	3.	7.	6.	5.	7.	4.	7.	7.	8.	7.				115.
4	6.	7.	4.	4.	5.	3.	5.	6.	6.	6.	4.	8.	5.	6.	3.	6.	6.	7.	3.				150.
5	3.	10.	3.	8.	5.	4.	5.	6.	7.	9.	5.	5.	5.	6.	5.	10.	3.	4.	7.				169.
6	7.	8.	5.	6.	5.	4.	6.	6.	6.	5.	4.	6.	6.	2.	7.	3.	5.	4.	4.	10.			179.
7	5.	8.	2.	4.	3.	7.	3.	5.	6.	7.	7.	8.	4.	8.	7.	6.	4.	6.	5.				193.
8	6.	8.	5.	7.	5.	8.	7.	4.	5.	4.	6.	6.	6.	4.	9.	5.	6.	7.	8.				196.
9	3.	7.	5.	6.	6.	4.	2.	6.	7.	4.	6.	9.	6.	8.	5.	9.	6.	6.	6.				219.
10	9.	9.	9.	5.	6.	6.	6.	7.	5.	6.	9.	3.	7.	7.	5.	4.	4.	6.	9.				223.
11	7.	6.	7.	6.	5.	6.	8.	3.	6.	3.	3.	5.	5.	7.	7.	5.	1.	3.	9.				230.
12	3.	5.	4.	1.	4.	4.	4.	5.	2.	2.	10.	5.	5.	5.	9.	3.	5.	5.	7.				235.
13	4.	3.	4.	8.	6.	4.	5.	6.	10.	3.	6.	7.	2.	3.	7.	7.	7.	7.	6.				121.
14	6.	5.	3.	4.	3.	5.	4.	6.	6.	6.	3.	6.	5.	5.	6.	7.	6.	7.	7.				162.
15	3.	6.	8.	4.	4.	9.	4.	4.	7.	6.	7.	5.	8.	8.	6.	5.	6.	5.	7.				172.
16	6.	7.	1.	4.	4.	5.	4.	7.	2.	8.	5.	5.	7.	7.	9.	6.	5.	6.	6.				199.
17	5.	7.	7.	5.	7.	6.	6.	4.	8.	4.	4.	3.	8.	7.	7.	8.	5.	3.	9.				241.
18	5.	9.	6.	7.	7.	6.	7.	8.	8.	2.	6.	5.	5.	4.	3.	3.	5.	5.	7.				101.
19	5.	4.	7.	6.	6.	4.	7.	8.	9.	5.	4.	7.	3.	5.	2.	8.	6.	7.	5.				138.
20	6.	8.	4.	3.	3.	7.	6.	6.	6.	5.	4.	6.	5.	7.	5.	8.	5.	8.	6.				139.
21	6.	3.	8.	5.	5.	4.	6.	7.	8.	6.	3.	6.	4.	6.	4.	8.	6.	6.	6.				142.
22	5.	3.	4.	5.	6.	5.	4.	7.	7.	5.	5.	6.	5.	5.	4.	7.	6.	5.	7.				155.
23	8.	9.	2.	5.	5.	3.	4.	7.	9.	6.	4.	6.	2.	7.	1.	9.	8.	4.	6.				169.
24	5.	5.	4.	6.	10.	7.	7.	6.	6.	5.	5.	5.	3.	5.	6.	6.	8.	6.	4.				174.
25	7.	8.	4.	7.	8.	3.	5.	7.	8.	7.	4.	6.	6.	5.	5.	5.	7.	5.	6.				190.
26	5.	8.	5.	6.	5.	4.	3.	5.	10.	6.	7.	6.	4.	8.	3.	9.	5.	2.	6.				193.
27	6.	8.	7.	8.	3.	6.	7.	4.	6.	3.	8.	3.	9.	5.	7.	4.	5.	2.	7.				195.
28	7.	6.	7.	5.	7.	2.	7.	3.	9.	6.	1.	5.	5.	5.	7.	5.	3.	1.	6.				205.
29	8.	8.	7.	5.	6.	5.	6.	8.	6.	5.	5.	4.	7.	5.	9.	7.	6.	7.	5.				210.
30	7.	8.	3.	8.	7.	4.	4.	4.	4.	2.	6.	6.	5.	4.	9.	7.	5.	7.	5.				218.
31	7.	7.	10.	6.	3.	10.	6.	5.	4.	8.	7.	3.	7.	5.	9.	2.	2.	1.	7.				259.
32	6.	8.	2.	7.	8.	6.	4.	6.	9.	7.	1.	9.	3.	4.	5.	9.	5.	8.	9.				95.
33	5.	5.	5.	6.	5.	6.	3.	4.	9.	8.	4.	8.	7.	8.	4.	7.	7.	7.	9.				140.
34	6.	8.	3.	5.	5.	4.	3.	5.	6.	8.	8.	7.	4.	7.	2.	6.	3.	5.	7.				163.
35	5.	9.	3.	9.	6.	6.	3.	6.	6.	4.	6.	6.	3.	8.	5.	5.	3.	5.	6.				165.
36	5.	8.	4.	4.	3.	4.	2.	5.	6.	5.	3.	8.	7.	8.	3.	5.	7.	5.	6.				170.
37	4.	8.	5.	8.	10.	6.	3.	9.	6.	3.	6.	7.	5.	7.	5.	6.	7.	5.	6.				178.
38	7.	9.	4.	6.	3.	3.	1.	6.	8.	5.	2.	9.	6.	7.	3.	10.	6.	7.	7.				180.
39	5.	5.	5.	3.	3.	6.	2.	6.	6.	7.	6.	9.	5.	9.	3.	9.	8.	9.	4.				181.
40	7.	7.	5.	2.	4.	7.	7.	5.	6.	5.	8.	4.	3.	5.	5.	5.	4.	3.	5.				182.
41	5.	9.	2.	4.	3.	2.	3.	4.	3.	5.	6.	8.	5.	9.	2.	7.	7.	8.	8.				187.
42	6.	10.	4.	6.	2.	6.	6.	6.	6.	8.	3.	6.	5.	7.	7.	6.	4.	5.	6.				188.
43	6.	5.	5.	7.	5.	3.	4.	4.	6.	7.	6.	6.	5.	3.	5.	6.	4.	6.	3.				188.
44	5.	7.	6.	7.	8.	4.	7.	7.	6.	7.	6.	4.	6.	8.	5.	5.	7.	6.	9.				189.
45	7.	6.	5.	5.	4.	4.	4.	6.	7.	7.	4.	8.	6.	8.	7.	5.	5.	5.	7.				205.
46	7.	7.	3.	4.	7.	8.	5.	4.	3.	6.	7.	4.	8.	6.	8.	4.	2.	3.	9.				207.
47	6.	6.	2.	5.	6.	7.	4.	5.	4.	5.	7.	5.	7.	7.	7.	3.	6.	5.	8.				222.
48	8.	8.	5.	6.	8.	7.	5.	3.	6.	4.	4.	4.	8.	4.	6.	3.	5.	3.	6.				223.
49	6.	8.	4.	3.	6.	6.	2.	4.	4.	5.	8.	3.	7.	5.	8.	2.	3.	4.	7.				228.
50	6.	6.	4.	9.	7.	3.	10.	6.	6.	6.	3.	7.	7.	10.	3.	9.	5.	4.	5.				231.
51	7.	8.	6.	4.	6.	7.	6.	4.	9.	4.	8.	8.	5.	8.	9.	6.	3.	2.	7.				231.
52	7.	7.	8.	6.	6.	7.	5.	1.	3.	4.	9.	1.	8.	9.	5.	2.	2.	1.	9.				251.

PERS. CORR. / STRESS = FIN TWOTWELVE FOUT ONESEVENTEIGHT /

APP. C. 3.
FULL ANALYSIS - MLRA III

1	5.81	1.34
2	6.94	1.76
3	4.83	2.00
4	5.42	1.78
5	5.44	1.95
6	5.23	1.78
7	4.83	1.89
8	5.48	1.54
9	6.31	1.93
10	5.29	1.74
11	5.40	2.04
12	5.81	1.83
13	5.38	1.71
14	6.31	1.69
15	5.50	2.19
16	6.02	2.06
17	5.04	1.73
18	5.12	2.00
19	6.71	1.61
20	185.37	37.00

SIMPLE CORRELATION COEFFICIENTS

1	1	1.0000	1	2	0.2350	1	3	0.1189	1	4	-0.0557	1	5	-0.0267	1	6	0.0107
1	7	0.2032	1	8	-0.1155	1	9	-0.0600	1	10	0.0912	1	11	-0.1070	1	12	-0.2791
1	13	0.1014	1	14	-0.0772	1	15	0.1065	1	16	-0.2321	1	17	-0.2166	1	18	-0.1816
1	19	0.0644	1	20	0.2886	2	3	-0.1698	2	4	0.2583	2	5	-0.0436	2	6	0.0481
2	1	0.2350	2	2	1.0000	2	9	-0.1620	2	10	0.1204	2	11	0.0066	2	12	-0.0400
2	7	-0.0089	2	8	-0.0545	2	15	-0.0532	2	16	-0.0320	2	17	-0.2374	2	18	-0.2040
2	13	0.1509	2	14	0.2301	3	3	1.0000	3	4	0.0930	3	5	0.1255	3	6	0.3151
2	19	0.0974	2	20	0.2334	3	8	-0.1445	3	9	0.1059	3	10	-0.1545	3	11	0.1089
3	1	0.1189	3	2	-0.1698	3	14	-0.0770	3	15	0.1187	3	16	-0.2895	3	17	-0.2881
3	7	0.4551	3	8	-0.1445	4	3	0.0930	4	4	1.0000	4	5	0.4310	4	6	-0.0812
3	13	0.3193	3	14	-0.0770	4	8	0.2308	4	9	-0.0149	4	10	-0.2591	4	12	0.0316
3	19	0.1911	3	20	0.2734	4	15	-0.0857	4	16	0.1370	4	17	-0.0502	4	18	-0.1302
4	1	-0.0557	4	2	0.2583	5	3	0.1255	5	4	0.4310	5	5	1.0000	5	6	0.0265
4	7	0.2800	4	8	0.0748	5	8	0.1234	5	9	0.1090	5	10	-0.2226	5	11	-0.0800
4	13	-0.0289	4	14	-0.1360	5	14	-0.2800	5	15	-0.0848	5	16	-0.0110	5	17	-0.1590
4	19	0.0161	4	20	-0.0532	6	3	0.3151	6	4	-0.0812	6	5	0.0265	6	6	1.0000
5	1	-0.0267	5	2	-0.0436	6	8	-0.2732	6	9	-0.0409	6	10	0.3250	6	12	-0.3605
5	7	0.2818	5	8	0.1234	6	14	-0.0699	6	15	0.4377	6	16	-0.3224	6	18	-0.2617
5	13	-0.0462	5	14	-0.2800	7	3	0.4551	7	4	0.2900	7	5	0.2818	7	6	0.1817
5	19	0.1097	5	20	0.0232	7	8	-0.0383	7	9	0.0096	7	10	-0.1039	7	11	-0.1597
6	1	0.0107	6	2	0.0481	7	14	-0.2233	7	15	0.1399	7	16	-0.1604	7	17	-0.2810
6	7	0.1817	6	8	-0.2633	8	3	-0.1445	8	4	0.0748	8	5	0.1234	8	6	-0.2633
6	13	0.3323	6	14	-0.0699	8	8	1.0000	8	9	0.2599	8	10	-0.0862	8	11	-0.1627
6	19	0.2425	6	20	0.1924	8	14	-0.0505	8	15	-0.2701	8	16	0.3550	8	17	0.4797
7	1	0.2032	7	2	-0.0089	9	3	0.1059	9	4	0.2308	9	5	0.1090	9	6	-0.2732
7	7	1.0000	7	8	-0.0383	9	8	0.2599	9	9	1.0000	9	10	0.4164	9	11	-0.4164
7	13	0.0455	7	14	-0.2233	9	14	-0.0599	9	15	-0.3670	9	16	0.5218	9	17	0.2914
7	19	0.0864	7	20	0.1181	10	3	-0.1545	10	4	-0.0149	10	5	-0.2226	10	6	-0.0409
8	1	-0.1155	8	2	-0.0545	10	8	0.0862	10	9	1.0000	10	10	-0.2375	10	11	0.1473
8	7	-0.0383	8	8	1.0000	10	14	0.2362	10	15	-0.1567	10	16	0.0028	10	18	0.0635
8	13	-0.3779	8	14	-0.0505	10	20	-0.0303									
8	19	-0.2272	8	20	-0.4890												
9	1	-0.0600	9	2	-0.1620												
9	7	0.0096	9	8	0.2599												
9	13	-0.4010	9	14	-0.0599												
9	19	-0.1098	9	20	-0.5092												
10	1	0.0912	10	2	0.1204												
10	7	-0.1039	10	8	0.0862												
10	13	0.0015	10	14	0.2362												
10	19	-0.1234	10	20	-0.0303												

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C.3./2

11	1	0.1070	11	2	0.0066	11	3	0.1089	11	4	0.2591	11	5	0.0800	11	6	0.3250
11	7	0.1597	11	8	0.1627	11	9	0.4164	11	10	0.2375	11	11	1.0000	11	12	0.4101
11	13	0.1629	11	14	0.0714	11	15	0.2608	11	16	0.4722	11	17	0.1770	11	18	0.1656
11	19	0.0718	11	20	0.3262	12	2	0.0400	12	4	0.0316	12	5	0.1790	12	6	0.3665
12	1	0.2791	12	8	0.2566	12	9	0.3741	12	10	0.1473	12	11	0.4101	12	12	1.0000
12	7	0.3799	12	14	0.2233	12	15	0.4017	12	16	0.5996	12	17	0.4005	12	18	0.5278
12	13	0.3976	12	20	0.4319	13	3	0.3193	13	4	0.0289	13	5	0.0462	13	6	0.3323
12	19	0.2855	13	2	0.1509	13	9	0.4010	13	10	0.0015	13	11	0.1629	13	12	0.3976
13	1	0.1014	13	8	0.3779	13	15	0.3357	13	16	0.3756	13	17	0.1584	13	18	0.2640
13	7	0.0455	13	14	0.1626	14	3	0.0770	14	4	0.1360	14	5	0.2800	14	6	0.0699
13	13	1.0000	13	20	0.5220	14	9	0.0599	14	10	0.2362	14	11	0.0714	14	12	0.2233
13	19	0.2407	14	2	0.2301	14	15	0.3193	14	16	0.1955	14	17	0.0767	14	18	0.0166
14	1	0.0772	14	8	0.0505	15	3	0.1187	15	4	0.0657	15	5	0.0755	15	6	0.4377
14	7	0.2233	14	14	1.0000	15	9	0.3670	15	10	0.1567	15	11	0.2608	15	12	0.4017
14	13	0.1626	14	20	0.2348	15	15	1.0000	15	16	0.4229	15	17	0.3421	15	18	0.1971
14	19	0.1414	15	2	0.0532	16	3	0.2895	16	4	0.1370	16	5	0.0848	16	6	0.4450
15	1	0.1065	15	8	0.2701	16	9	0.5218	16	10	0.2168	16	11	0.4722	16	12	0.5996
15	7	0.1399	15	14	0.3183	16	15	0.4229	16	16	1.0000	16	17	0.3908	16	18	0.4422
15	13	0.3357	15	20	0.4951	17	3	0.2881	17	4	0.0502	17	5	0.0110	17	6	0.3224
15	19	0.1026	16	2	0.0320	17	9	0.2914	17	10	0.0028	17	11	0.1770	17	12	0.4005
16	1	0.2321	16	8	0.3550	17	15	0.3421	17	16	0.3908	17	17	1.0000	17	18	0.6246
16	7	0.1604	16	14	0.1955	18	3	0.4128	18	4	0.1302	18	5	0.1590	18	6	0.2617
16	13	0.3756	16	20	0.3919	18	9	0.0569	18	10	0.0635	18	11	0.1656	18	12	0.5278
16	19	0.2458	17	2	0.2374	18	15	0.1971	18	16	0.4422	18	17	0.6246	18	18	1.0000
17	1	0.2166	17	8	0.4797	19	3	0.1911	19	4	0.0161	19	5	0.1097	19	6	0.2425
17	7	0.2810	17	14	0.0767	19	9	0.1098	19	10	0.1234	19	11	0.0718	19	12	0.2855
17	13	0.1584	17	20	0.3970	20	3	0.2734	19	16	0.2458	19	17	0.3129	19	18	0.2269
17	19	0.3129	18	2	0.2040	20	9	0.5092	20	4	0.0532	20	5	0.0232	20	6	0.1924
18	1	0.1816	18	8	0.4790	20	15	0.4951	20	10	0.0303	20	11	0.3262	20	12	0.4319
18	7	0.3591	18	14	0.0166	20	16	0.3919	20	16	0.3919	20	17	0.3970	20	18	0.5416
18	13	0.2840	18	20	0.5416	19	2	0.0074									
18	19	0.2269	19	2	0.0074	19	8	0.2272									
19	1	0.0644	19	14	0.1414	19	15	0.1026									
19	7	0.0864	19	20	0.0892	20	2	0.2334									
19	13	0.2407	20	2	0.2334	20	8	0.4890									
19	19	1.0000	20	14	0.2348												
20	1	0.2886	20	20	1.0000												
20	7	0.1181															
20	13	0.5220															
20	19	0.0892															

STEP NUMBER 1 ENTER VARIABLE 18
 STANDARD ERROR OF ESTIMATE= 31.412
 MULTIPLE CORRELATION COEFFICIENT = 0.542
 GOODNESS OF FIT: 29.3280%
 CONSTANT TERM= 236.6983

VAR	COEFF	STD DEV	F-RATIO	BETA COEFF
		COEFF	VAR.CONTR	
18	-.1003E 02	.2203E 01	.2075E 02	-.5416E 00

STEP NUMBER 2 ENTER VARIABLE 9
 STANDARD ERROR OF ESTIMATE= 26.072
 MULTIPLE CORRELATION COEFFICIENT = 0.723
 GOODNESS OF FIT: 52.2854%
 CONSTANT TERM= 292.2763

VAR	COEFF	STD DEV	F-RATIO	BETA COEFF
		COEFF	VAR.CONTR	
9	-.9221E 01	.1899E 01	.2358E 02	-.4799E 00
18	-.9529E 01	.1832E 01	.2707E 02	-.5143E 00

STEP NUMBER 3 ENTER VARIABLE 15
 STANDARD ERROR OF ESTIMATE= 24.727
 MULTIPLE CORRELATION COEFFICIENT = 0.761
 GOODNESS OF FIT: 57.9599%
 CONSTANT TERM= 252.4038

VAR	COEFF	STD DEV	F-RATIO	BETA COEFF
		COEFF	VAR.CONTR	
9	-.7432E 01	.1933E 01	.1478E 02	-.3868E 00
15	.4403E 01	.1730E 01	.6479E 01	.2608E 00
18	-.8674E 01	.1769E 01	.2404E 02	-.4681E 00

STEP NUMBER 4 ENTER VARIABLE 14
 STANDARD ERROR OF ESTIMATE= 21.944
 MULTIPLE CORRELATION COEFFICIENT = 0.822
 GOODNESS OF FIT: 67.5778%
 CONSTANT TERM= 182.1369

VAR	COEFF	STD DEV	F-RATIO	BETA COEFF
		COEFF	VAR.CONTR	
9	-.6107E 01	.1752E 01	.1215E 02	-.3178E 00
14	.7355E 01	.1970E 01	.1394E 02	.3353E 00
15	.6727E 01	.1657E 01	.1649E 02	.3985E 00
18	-.8141E 01	.1577E 01	.2667E 02	-.4393E 00

STEP NUMBER 5 ENTER VARIABLE 6
 STANDARD ERROR OF ESTIMATE= 20.992
 MULTIPLE CORRELATION COEFFICIENT = 0.842
 GOODNESS OF FIT: 70.96237
 CONSTANT TERM= 203.6938

VAR	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	VAR.CONTR		
6	-.4391E 01	.1897E 01	.5362E 01	-.2110E 00	
9	-.6617E 01	.1691E 01	.1532E 02	-.3444E 00	
14	.7524E 01	.1886E 01	.1592E 02	.3430E 00	
15	.8036E 01	.1682E 01	.2281E 02	.4761E 00	
18	-.8851E 01	.1539E 01	.3308E 02	-.4777E 00	

STEP NUMBER 6 ENTER VARIABLE 3
 STANDARD ERROR OF ESTIMATE= 20.101
 MULTIPLE CORRELATION COEFFICIENT = 0.860
 GOODNESS OF FIT: 73.95438
 CONSTANT TERM= 187.6618

VAR	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	VAR.CONTR		
3	.3718E 01	.1635E 01	.5169E 01	.2007E 00	
6	-.5561E 01	.1887E 01	.8680E 01	-.2672E 00	
9	-.7352E 01	.1651E 01	.1983E 02	-.3826E 00	
14	.7784E 01	.1809E 01	.1851E 02	.3549E 00	
15	.8113E 01	.1611E 01	.2535E 02	.4807E 00	
18	-.7527E 01	.1584E 01	.2257E 02	-.4062E 00	

STEP NUMBER 7 ENTER VARIABLE 1
 STANDARD ERROR OF ESTIMATE= 19.416
 MULTIPLE CORRELATION COEFFICIENT = 0.873
 GOODNESS OF FIT: 76.23987
 CONSTANT TERM= 158.0364

VAR	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	VAR.CONTR		
1	.4280E 01	.2080E 01	.4232E 01	.1554E 00	
3	.3483E 01	.1584E 01	.4838E 01	.1480E 00	
6	-.5205E 01	.1831E 01	.8080E 01	-.2502E 00	
9	-.7148E 01	.1598E 01	.2001E 02	-.3720E 00	
14	.8002E 01	.1751E 01	.2089E 02	.3644E 00	
15	.7935E 01	.1559E 01	.2591E 02	.4701E 00	
18	-.7065E 01	.1547E 01	.2086E 02	-.3813E 00	

STEP NUMBER 8 ENTER VARIABLE 19
 STANDARD ERROR OF ESTIMATE= 18.912
 MULTIPLE CORRELATION COEFFICIENT = 0.883
 GOODNESS OF FIT: 77.9681%

CONSTANT TERM= 173.9691

VAR	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	VAR.CONTR		
1	.4417E 01	.2028E 01	.4744E 01	.1604E 00	
3	.3735E 01	.1549E 01	.5817E 01	.2016E 00	
6	-.4720E 01	.1803E 01	.6852E 01	-.2269E 00	
9	-.7271E 01	.1558E 01	.2179E 02	-.3784E 00	
14	.8514E 01	.1728E 01	.2428E 02	.3881E 00	
15	.7996E 01	.1519E 01	.2772E 02	.4737E 00	
18	-.7394E 01	.1517E 01	.2375E 02	-.3990E 00	
19	-.3217E 01	.1751E 01	.3373E 01	-.1402E 00	

STEP NUMBER 9 ENTER VARIABLE 8
 STANDARD ERROR OF ESTIMATE= 18.547
 MULTIPLE CORRELATION COEFFICIENT = 0.891
 GOODNESS OF FIT: 79.3039%

CONSTANT TERM= 189.0817

VAR	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	VAR.CONTR		
1	.4336E 01	.1989E 01	.4751E 01	.1575E 00	
3	.3704E 01	.1522E 01	.6579E 01	.2108E 00	
6	-.4893E 01	.1771E 01	.7628E 01	-.2351E 00	
8	-.3345E 01	.2032E 01	.2711E 01	-.1393E 00	
9	-.6858E 01	.1548E 01	.1962E 02	-.3569E 00	
14	.8332E 01	.1698E 01	.2407E 02	.3798E 00	
15	.7721E 01	.1499E 01	.2653E 02	.4574E 00	
18	-.6275E 01	.1635E 01	.1471E 02	-.3386E 00	
19	-.3498E 01	.1726E 01	.4108E 01	-.1525E 00	

STEP NUMBER 10 ENTER VARIABLE 17
 STANDARD ERROR OF ESTIMATE= 18.087
 MULTIPLE CORRELATION COEFFICIENT = 0.899
 GOODNESS OF FIT: 80.7857%

CONSTANT TERM= 178.6407

VAR	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	VAR.CONTR		
1	.4707E 01	.1951E 01	.5820E 01	.1709E 00	
3	.4014E 01	.1486E 01	.7299E 01	.2167E 00	
6	-.4791E 01	.1729E 01	.7683E 01	-.2303E 00	
8	-.3935E 01	.2009E 01	.3836E 01	-.1638E 00	
9	-.7469E 01	.1548E 01	.2327E 02	-.3887E 00	
14	.7498E 01	.1667E 01	.2302E 02	.3646E 00	
15	.7954E 01	.1468E 01	.2940E 02	.4715E 00	
17	.3782E 01	.2127E 01	.3162E 01	.1764E 00	
18	-.7799E 01	.1811E 01	.1854E 02	-.4209E 00	
19	-.2925E 01	.1714E 01	.2913E 01	-.1275E 00	

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C.B./6

STEP NUMBER 11 ENTER VARIABLE 16
 STANDARD ERROR OF ESTIMATE= 17.504
 MULTIPLE CORRELATION COEFFICIENT = 0.908
 GOODNESS OF FIT: 82.4438%
 CONSTANT TERM= 164.8749

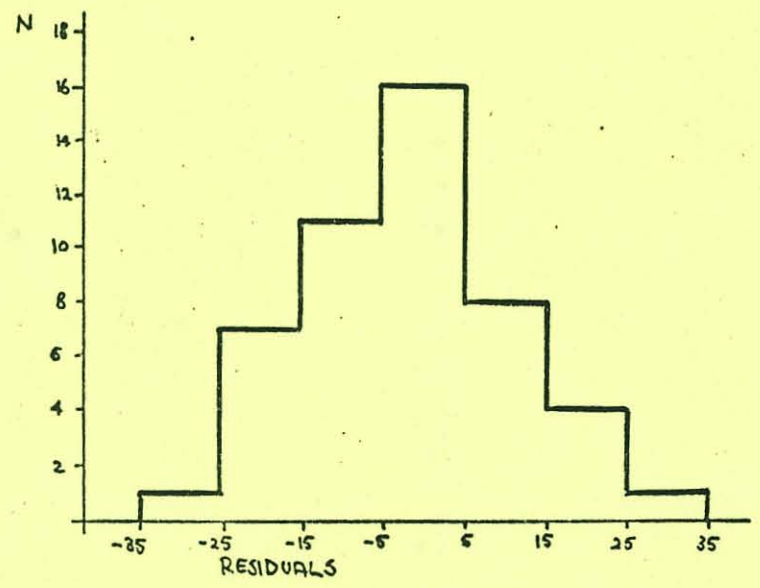
VAR	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	VAR. CONTR		
1	.5400E 01	.1921E 01	.7697E 01	.1961E 00	
3	.4497E 01	.1459E 01	.9500E 01	.2428E 00	
6	-.4122E 01	.1708E 01	.5824E 01	-.1981E 00	
8	-.4127E 01	.1947E 01	.4495E 01	-.1718E 00	
9	-.9352E 01	.1784E 01	.2747E 02	-.4867E 00	
14	.7043E 01	.1686E 01	.1744E 02	.3211E 00	
15	.8102E 01	.1422E 01	.3244E 02	.4800E 00	
16	.3586E 01	.1845E 01	.3778E 01	.1999E 00	
17	.4540E 01	.2095E 01	.4696E 01	.2118E 00	
18	-.9120E 01	.1880E 01	.2354E 02	-.4922E 00	
19	-.2413E 01	.1679E 01	.2064E 01	-.1052E 00	

ORS	ACTUAL	ESTIMATE	RESIDUAL
1	173.00	177.06	-4.06
2	161.00	176.40	-15.40
3	115.00	139.30	-24.30
4	159.00	166.26	-7.26
5	169.00	166.72	2.28
6	179.00	176.88	2.12
7	193.00	181.21	11.79
8	196.00	186.63	9.37
9	219.00	184.01	34.99
10	223.00	199.44	23.56
11	230.00	220.33	9.67
12	235.00	222.33	12.67
13	121.00	126.09	-5.09
14	162.00	164.71	-2.71
15	172.00	185.61	-13.61
16	199.00	230.80	-31.80
17	241.00	215.62	25.38
18	101.00	109.73	-8.73
19	138.00	121.12	16.88
20	139.00	159.29	-20.29
21	142.00	174.45	-32.45
22	155.00	152.37	2.63
23	169.00	166.68	2.32
24	174.00	177.42	-3.42
25	190.00	169.94	20.06
26	193.00	186.62	6.38
27	195.00	225.24	-30.24
28	205.00	229.24	-24.24
29	210.00	214.38	-4.38
30	218.00	218.74	-0.74
31	259.00	246.75	12.25
32	95.00	101.58	-6.58
33	140.00	149.02	-9.02
34	163.00	155.67	7.33
35	165.00	168.08	-3.08
36	170.00	186.90	-16.90
37	178.00	174.00	4.00
38	180.00	164.69	15.31
39	181.00	173.30	7.70
40	182.00	191.94	-9.94
41	187.00	192.26	-5.26
42	188.00	195.26	-7.26
43	188.00	174.12	13.88
44	189.00	187.49	1.51
45	205.00	209.63	-4.63
46	207.00	220.04	-13.04
47	222.00	198.48	23.52
48	223.00	201.61	21.39
49	228.00	204.06	23.94
50	231.00	223.18	7.82
51	231.00	229.39	1.61
52	251.00	266.92	-15.92



GRAPH OF VARIABLE 20 VS. RESIDUALS

X-AXIS BOUNDS ARE: 94.000000 260.000000 X INCREMENT = 2.101266
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



HISTOGRAM OF RESIDUALS

APP. C.4.
 PLOT AND HISTOGRAM OF RESIDUALS
 OF MLRA III

PERSONALITY CORRELATES TO PERFORMANCE UNDER STRESS

H.P. LEHMANN

TRANSFORMED DATA: SECOND-STRATUM STEN/SCORES FOR 16/PF-VALUES

	EXVIA	ANXIETY	CORTERTIA	INDEP.	DISCREET	PROD.S.	SUPEREGO	STATF/A.	TRAIT/A.	RAVEN/INT	AGE	EXPERIENCE	SIM.SCORE
1	5.07		3.89	4.60	7.01	4.84	5.23	5.00	7.00	8.00			173.00
2	8.25		8.08	4.94	3.73	6.16	6.32	2.00	3.00	9.00			161.00
3	2.16		2.07	3.45	7.50	4.75	4.79	7.00	8.00	7.00			115.00
4	4.63		5.14	4.91	4.14	6.86	2.96	6.00	7.00	3.00			159.00
5	5.60		8.13	7.88	4.64	8.50	4.29	3.00	4.00	7.00			169.00
6	5.37		5.22	5.51	4.21	6.22	3.89	4.00	4.00	10.00			172.00
7	2.63		4.25	4.72	7.20	6.10	7.11	4.00	6.00	5.00			193.00
8	6.49		5.15	5.49	6.08	4.00	8.46	6.00	7.00	8.00			196.00
9	3.75		6.48	5.30	5.83	6.45	4.36	6.00	6.00	6.00			219.00
10	5.73		3.07	6.26	8.86	7.11	5.81	4.00	6.00	9.00			223.00
11	6.43		6.13	6.12	3.35	2.94	6.65	1.00	3.00	9.00			230.00
12	2.63		4.97	1.86	10.06	4.89	5.87	5.00	5.00	7.00			235.00
13	6.07		4.96	5.20	6.07	3.50	5.31	7.00	7.00	6.00			121.00
14	4.09		4.03	4.70	3.43	5.63	5.53	6.00	7.00	7.00			162.00
15	3.65		6.61	6.11	6.97	6.05	8.10	6.00	5.00	7.00			172.00
16	3.71		3.94	5.50	5.16	7.79	5.95	5.00	6.00	6.00			194.00
17	6.46		7.14	6.92	4.00	3.99	6.44	5.00	3.00	9.00			241.00
18	6.82		4.75	5.69	6.00	5.93	5.13	5.00	5.00	7.00			101.00
19	6.47		5.34	5.41	4.00	7.50	3.91	6.00	7.00	5.00			138.00
20	4.39		4.43	4.40	4.36	6.24	7.05	5.00	8.00	6.00			139.00
21	5.68		5.01	5.69	3.17	6.82	4.48	6.00	6.00	6.00			142.00
22	5.14		4.70	4.73	5.06	6.47	4.72	6.00	5.00	7.00			155.00
23	4.86		3.90	5.37	4.25	6.25	2.76	8.00	4.00	6.00			169.00
24	7.98		6.76	5.33	4.73	6.78	6.35	8.00	6.00	4.00			174.00
25	6.60		4.66	7.37	3.96	6.27	2.99	7.00	5.00	6.00			190.00
26	4.17		6.01	6.47	6.90	5.62	4.01	5.00	2.00	6.00			193.00
27	5.62		5.11	7.36	8.05	3.55	6.63	5.00	2.00	7.00			195.00
28	7.01		6.73	7.23	1.28	2.84	3.32	3.00	1.00	6.00			205.00
29	6.29		2.74	5.92	5.14	5.81	6.55	6.00	7.00	5.00			210.00
30	6.37		5.21	4.84	5.99	2.73	5.51	5.00	7.00	5.00			218.00
31	5.27		4.01	6.66	7.09	6.45	9.69	2.00	1.00	7.00			259.00
32	6.98		5.60	5.66	1.09	5.83	5.34	5.00	8.00	9.00			45.00
33	4.23		6.15	7.23	4.06	5.35	5.00	7.00	7.00	9.00			140.00
34	3.82		5.46	5.34	7.82	7.50	3.16	3.00	5.00	7.00			163.00
35	4.73		5.75	6.40	5.99	5.70	5.46	3.00	5.00	6.00			165.00
36	2.45		5.39	5.40	3.38	5.30	3.46	7.00	5.00	6.00			170.00
37	6.11		5.52	5.61	5.67	7.68	5.22	7.00	5.00	6.00			174.00
38	3.31		4.10	5.65	2.39	5.33	3.23	6.00	7.00	7.00			180.00
39	2.27		4.65	4.09	6.10	8.04	5.44	8.00	9.00	4.00			181.00
40	5.05		3.86	3.40	8.19	5.32	7.07	4.00	3.00	5.00			182.00
41	2.39		6.75	4.42	6.05	6.82	2.08	7.00	8.00	8.00			187.00
42	4.37		4.77	7.03	3.35	6.82	6.51	4.00	5.00	6.00			188.00
43	5.54		5.93	5.87	5.89	5.38	3.41	4.00	6.00	3.00			188.00
44	6.53		6.73	7.95	5.67	8.15	3.93	7.00	6.00	9.00			189.00
45	3.72		3.45	6.48	4.35	5.67	4.79	5.00	5.00	7.00			205.00
46	5.65		4.88	5.15	6.96	5.26	7.46	2.00	3.00	9.00			207.00
47	4.47		4.43	5.36	7.09	5.23	6.47	6.00	5.00	8.00			222.00
48	7.15		5.41	6.13	4.13	2.77	5.99	5.00	3.00	6.00			223.00
49	3.89		4.61	4.52	3.06	4.17	6.04	3.00	4.00	7.00			228.00
50	7.56		7.76	8.96	2.78	7.89	3.02	5.00	4.00	5.00			231.00
51	5.04		3.61	5.19	0.15	2.98	8.05	3.00	2.00	7.00			231.00
52	5.17		7.39	6.83	8.82	3.75	6.02	2.00	1.00	9.00			251.00

STEP NUMBER 9 ENTER VARIABLE 2
 STANDARD ERROR OF ESTIMATE = 31.089
 MULTIPLE CORRELATION COEFFICIENT = 0.647
 GOODNESS OF FIT: 41.6487%

CONSTANT TERM = 191.5354

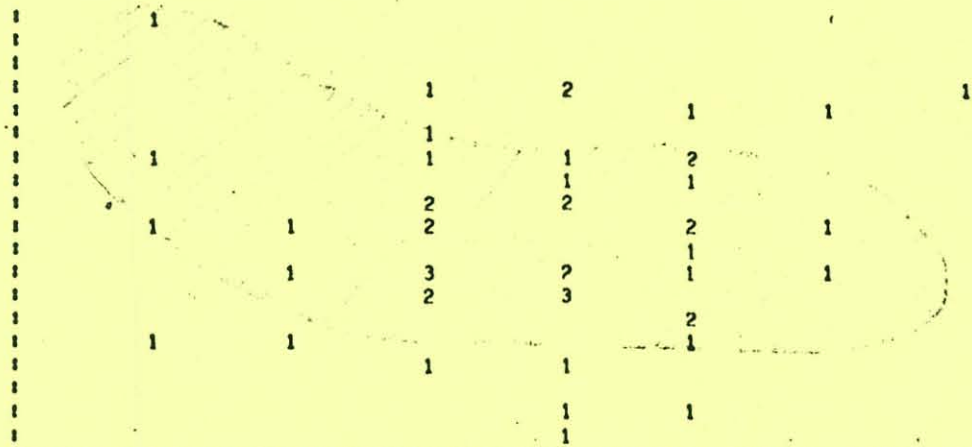
VAR	COEFF	STD DEV COEFF	F-RATIO VAR CONTR	BETA COEFF
1	-.4434E 01	.3608E 01	.1511E 01	-.1830E 00
2	.1618E 01	.4001E 01	.1635E 00	.5794E -01
3	.7729E 01	.4778E 01	.2617E 01	.2599E 00
4	.4498E 01	.2812E 01	.2558E 01	.2406E 00
5	-.3677E 01	.3386E 01	.1179E 01	-.1497E 00
6	.3214E 01	.3295E 01	.9510E 00	.1421E 00
7	-.1444E 01	.3456E 01	.1747E 00	-.6738E -01
8	-.6356E 01	.3315E 01	.3676E 01	-.3430E 00
9	-.2481E 01	.2910E 01	.7270E 00	-.1082E 00

STEP NUMBER 2 ENTER VARIABLE 4
 STANDARD ERROR OF ESTIMATE= 30.392
 MULTIPLE CORRELATION COEFFICIENT = 0.593
 GOODNESS OF FIT: 35.1642%
 CONSTANT TERM= 208.1580

VAR	COEFF	STD DEV	F-RATIO	BETA COEFF
		COEFF	VAR.CONTR	
4	.4570E 01	.2176E 01	.4411E 01	.2445E 00
8	-.9339E 01	.2157E 01	.1074E 02	-.5040E 00

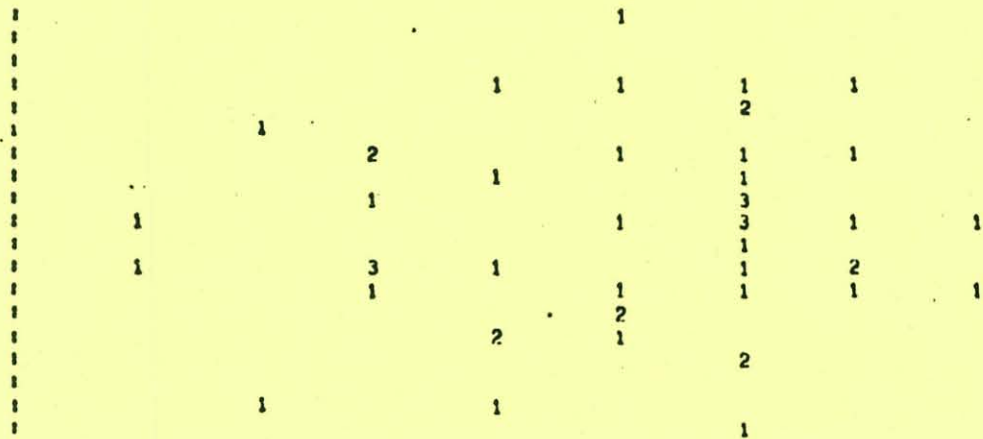
APPENDIX D

- D.1. Scattergram of residuals (MLRA III) - investigation for curvatures
- D.2. MQRA I - Results
- D.3. MQRA II - Results
- D.4. MQRA III - Results of both subsets
- D.5. MQRA IV - Results



GRAPH OF VARIABLE 1 VS. RESIDUALS

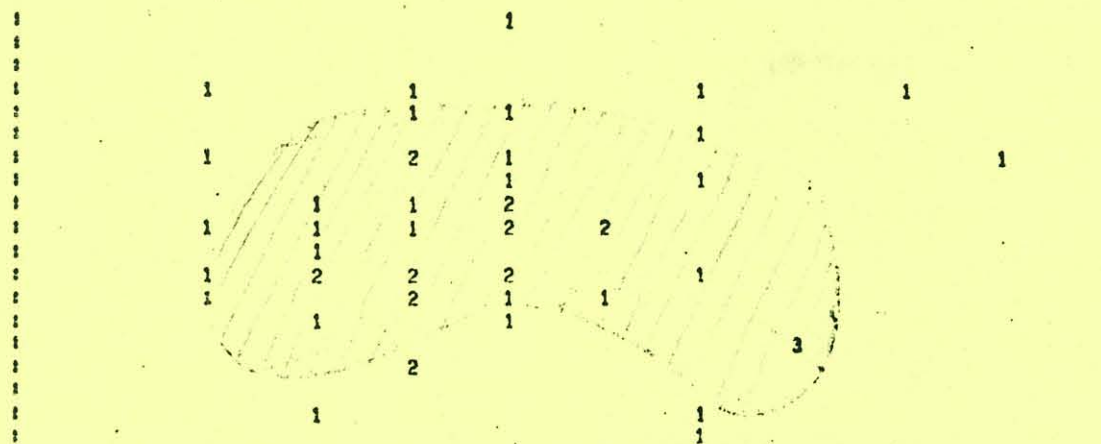
X-AXIS BOUNDS ARE: 2.000000 10.000000 X INCREMENT = 0.101266
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



GRAPH OF VARIABLE 2 VS. RESIDUALS

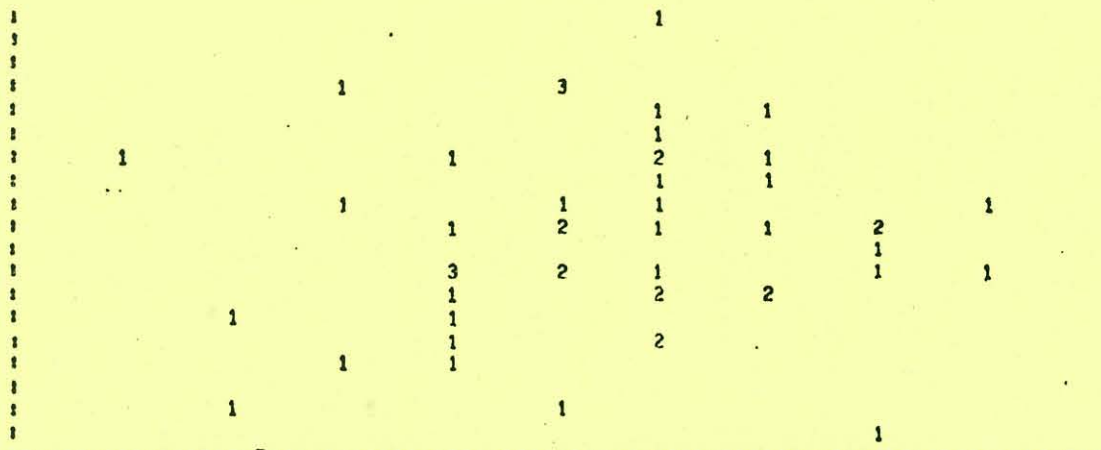
X-AXIS BOUNDS ARE: 2.000000 11.000000 X INCREMENT = 0.113924
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52

FPP. D.I.)))))
 SCATTERGRAM OF RESIDUALS
 (CHIA III)
 INVESTIGATION FOR
 CURVATURES



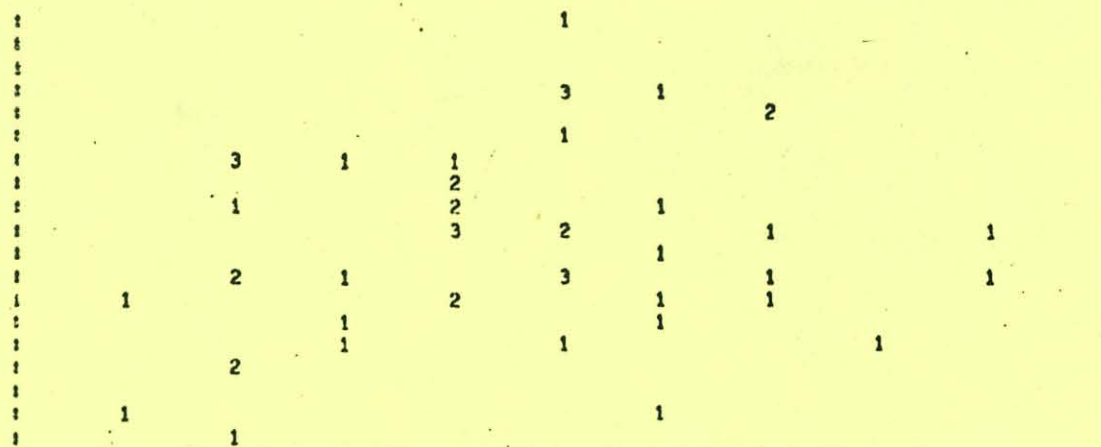
GRAPH OF VARIABLE 3 VS. RESIDUALS

X-AXIS BOUNDS ARE: 0.000000 11.000000 X INCREMENT = 0.139241
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



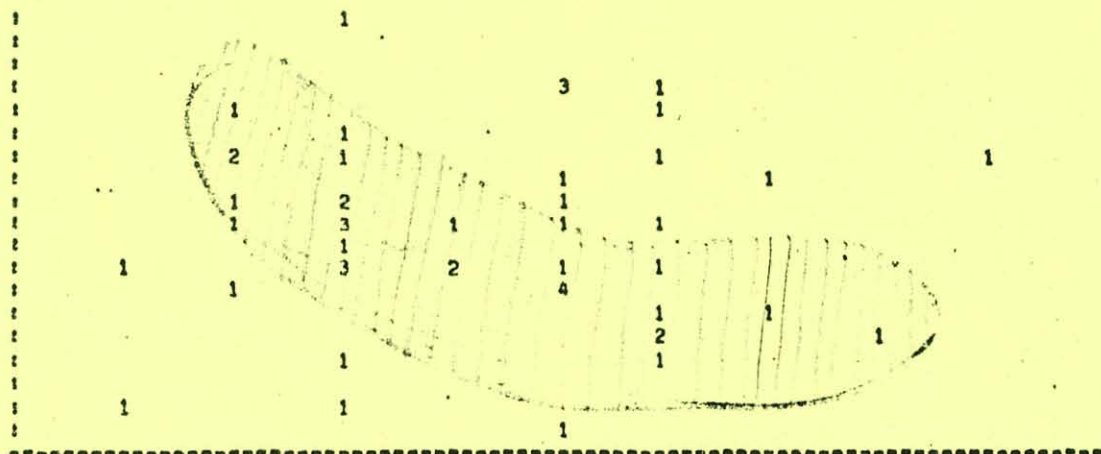
GRAPH OF VARIABLE 4 VS. RESIDUALS

X-AXIS BOUNDS ARE: 0.000000 10.000000 X INCREMENT = 0.126522
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



GRAPH OF VARIABLE 5 VS. RESIDUALS

X-AXIS BOUNDS ARE: 1.000000 11.000000 X INCREMENT = 0.126582
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



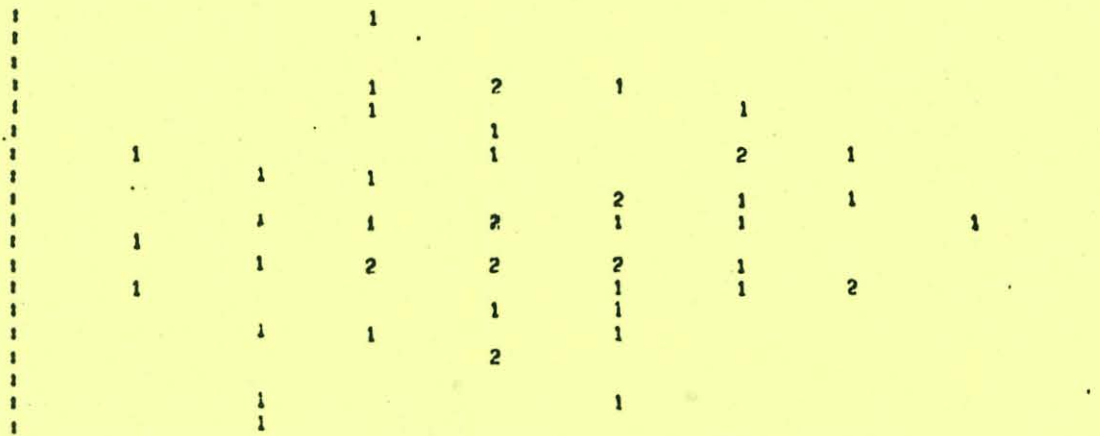
GRAPH OF VARIABLE 6 VS. RESIDUALS

X-AXIS BOUNDS ARE: 1.000000 11.000000 X INCREMENT = 0.126582
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



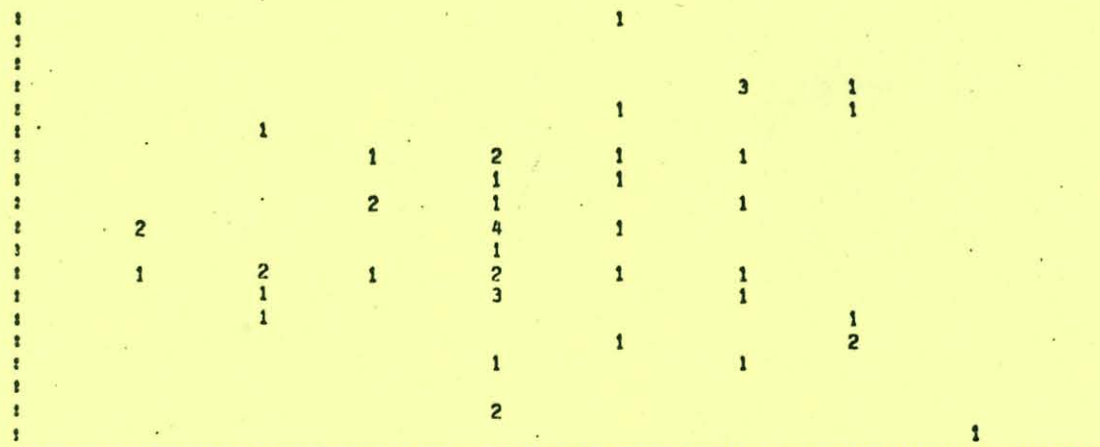
GRAPH OF VARIABLE 9 VS. RESIDUALS

X-AXIS BOUNDS ARE: 1.000000 11.000000 X INCREMENT = 0.126582
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



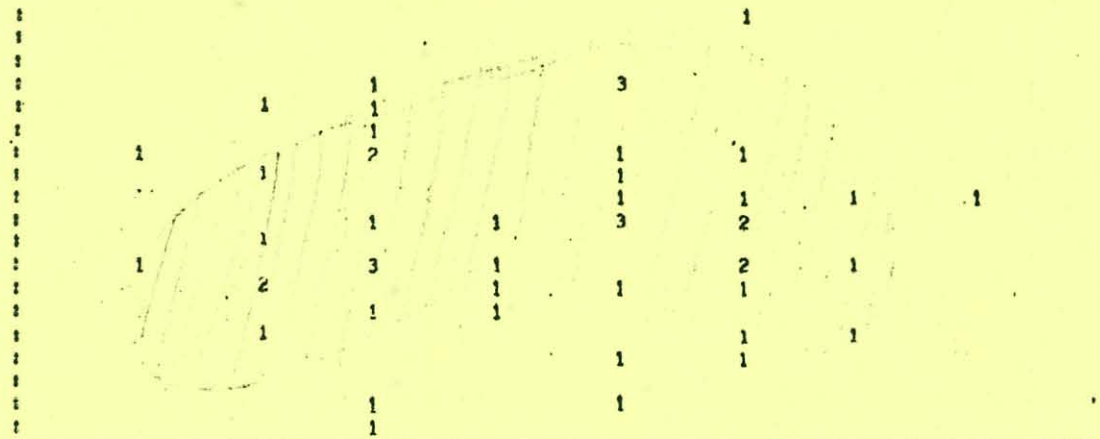
GRAPH OF VARIABLE 10 VS. RESIDUALS

X-AXIS BOUNDS ARE: 1.000000 10.000000 X INCREMENT = 0.113924
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



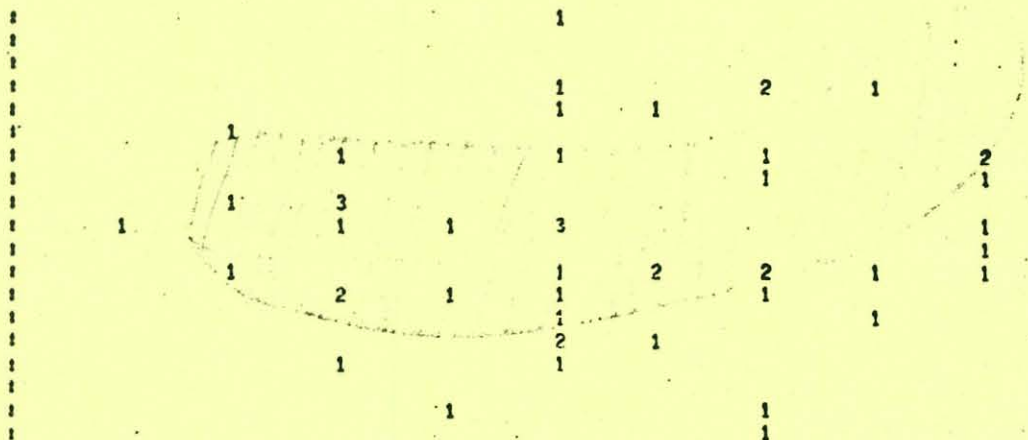
GRAPH OF VARIABLE 13 VS. RESIDUALS

X-AXIS BOUNDS ARE: 1.000000 10.000000 X INCREMENT = 0.113924
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



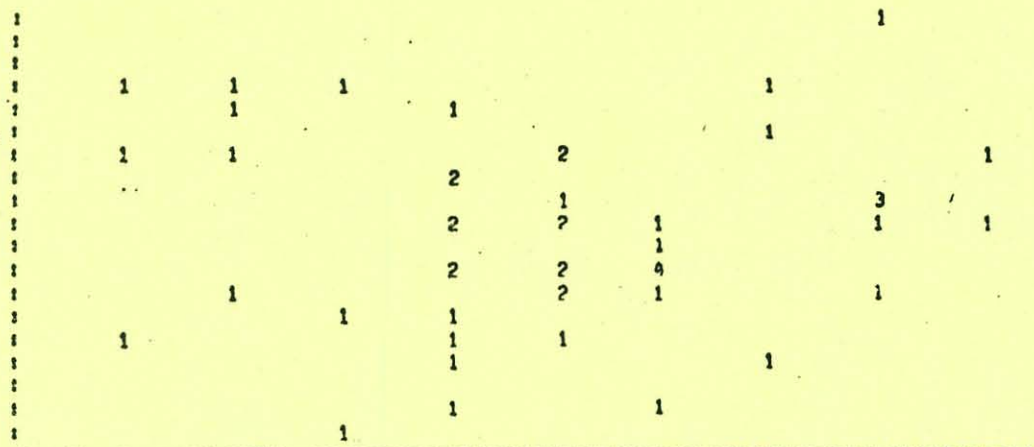
GRAPH OF VARIABLE 14 VS. RESIDUALS

X-AXIS BOUNDS ARE: 2.000000 11.000000 X INCREMENT = 0.113924
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



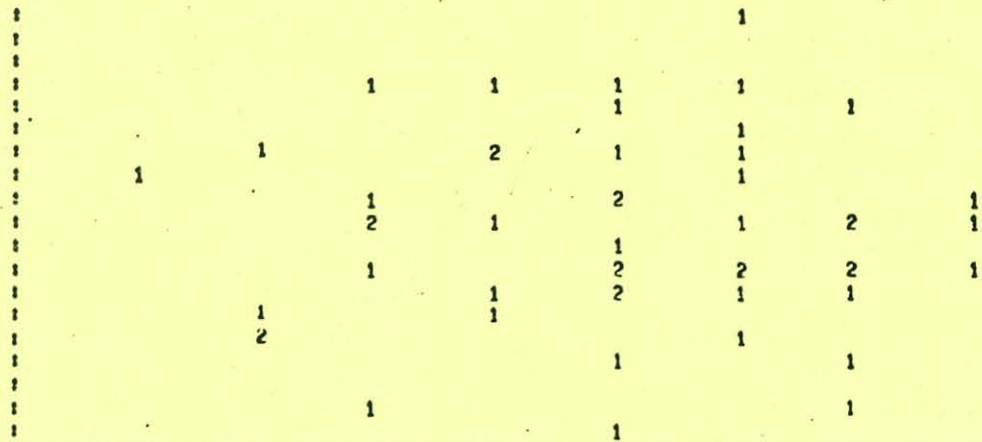
GRAPH OF VARIABLE 15 VS. RESIDUALS

X-AXIS BOUNDS ARE: 0.000000 10.000000 X INCREMENT = 0.126582
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



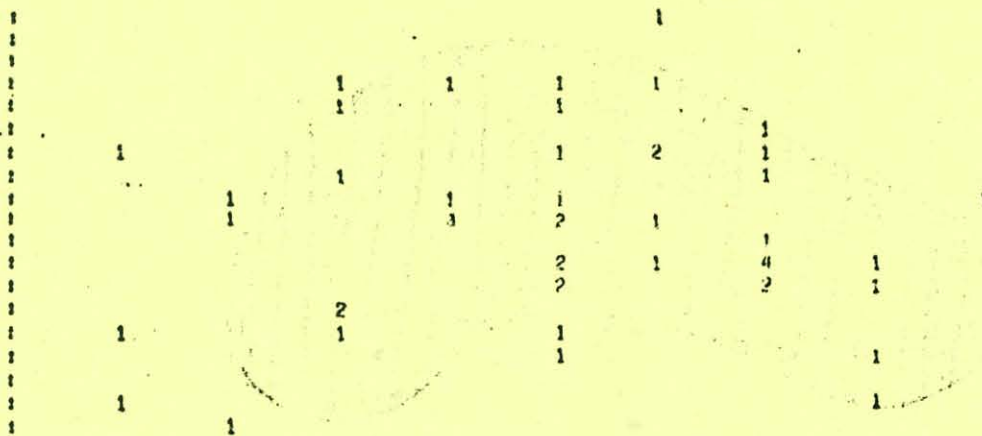
GRAPH OF VARIABLE 16 VS. RESIDUALS

X-AXIS BOUNDS ARE: 1.000000 11.000000 X INCREMENT = 0.126582
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



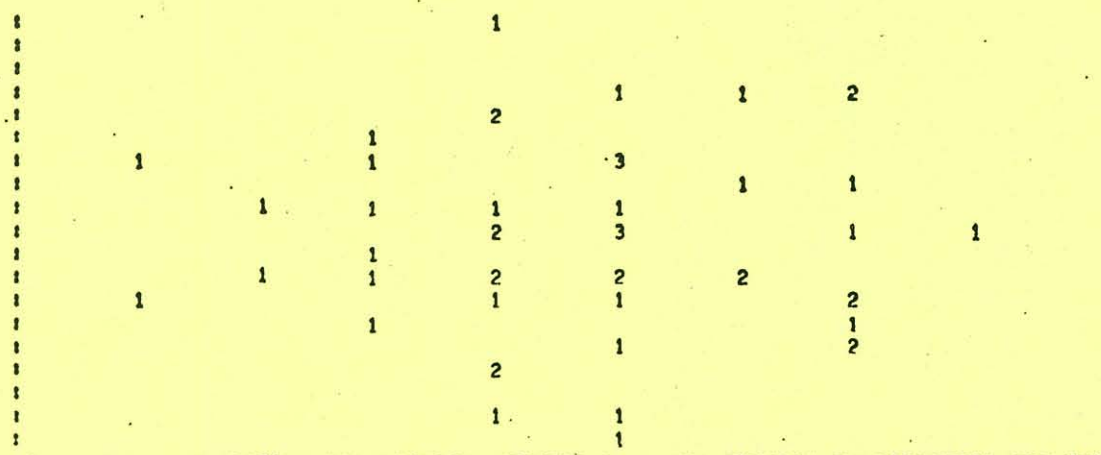
GRAPH OF VARIABLE 17 VS. RESIDUALS

X-AXIS BOUNDS ARE: 0.000000 9.000000 X INCREMENT = 0.113924
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



GRAPH OF VARIABLE 18 VS. RESIDUALS

X-AXIS BOUNDS ARE: 0.000000 10.000000 X INCREMENT = 0.126582
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52



 GRAPH OF VARIABLE 19 VS. RESIDUALS

X-AXIS BOUNDS ARE: 2.000000 11.000000 X INCREMENT = 0.113924
 Y-AXIS BOUNDS ARE: -33.000000 35.000000 Y INCREMENT = 3.578947
 NO. OF OBSERVATIONS = 52

STFP NUMBER 27 ENTER VARIABLE 17
 STANDARD ERROR OF ESTIMATE= 19.705
 MULTIPLE CORRELATION COEFFICIENT = 0.931
 GOODNESS OF FIT: 80.6504%
 CONSTANT TERM= -519.8912

VAR	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	VAR. CONTR		
1	.1896E 02	.3665E 02	.2676E 00	.6887E 00	
2	.4322E 02	.3838E 02	.1268E 01	.2333E 01	
3	.3140E 02	.3174E 02	.9783E 00	.1634E 01	
4	.4436E 02	.4310E 02	.1056E 01	.2022E 01	
5	.5422E 02	.2712E 02	.3998E 01	.3212E 01	
6	.4824E 02	.3630E 02	.1767E 01	.2604E 01	
7	.2247E 01	.2014E 01	.1245E 01	.9346E 00	
8	-.1815E 01	.1171E 01	.2403E 01	-.1063E 01	
9	-.3765E 00	.1319E 01	.8149E 01	-.2443E 00	
10	.1472E 01	.1513E 01	.9471E 00	.8500E 00	
11	.5716E 00	.1283E 01	.1987E 00	.3827E 00	
12	-.1298E 01	.1403E 01	.8568E 00	-.6877E 00	
13	.7840E 00	.1490E 01	.2770E 00	.3237E 00	
14	-.1215E 01	.1808E 01	.4513E 00	-.4446E 00	
15	-.4097E 01	.2645E 01	.2400E 01	-.1367E 01	
16	-.2705E 01	.2047E 01	.1746E 01	-.1192E 01	
17	.1169E 01	.2481E 01	.2220E 00	.4100E 00	
18	-.1333E 01	.2150E 01	.3847E 00	-.5524E 00	
19	-.1421E 01	.1999E 01	.5055E 00	-.5704E 00	
20	-.1011E 01	.1608E 01	.3949E 00	-.4648E 00	
21	-.7402E 00	.1563E 01	.2243E 00	-.2392E 00	
22	-.6286E 00	.1687E 01	.1358E 00	-.2703E 00	
23	-.9735E 00	.1343E 01	.5252E 00	-.3949E 00	
24	-.2060E 01	.1460E 01	.1992E 01	-.9351E 00	
25	-.1604E 01	.1644E 01	.9524E 00	-.6144E 00	
26	-.2365E 01	.1830E 01	.1669E 01	-.1020E 01	
27	-.3185E 01	.1497E 01	.4524E 01	-.1310E 01	

VAR.
DESCR.:
 A
 C
 L
 Q2
 Q3
 A-T
 A²
 C²
 L²
 Q2²
 Q3²
 A-T²
 AxC
 AxL
 AxQ2
 AxQ3
 Ax A-T
 CxL
 CxQ2
 CxQ3
 Cx A-T
 LxQ2
 LxQ3
 Lx A-T
 Q2xQ3
 Q2x A-T
 Q3x A-T

STPP NUMBER 7 DELETE VARIABLE 18
 STANDARD ERROR OF ESTIMATE = 18.679
 MULTIPLE CORRELATION COEFFICIENT = 0.878
 GOODNESS OF FIT: 77.0079%
 CONSTANT TERM = 118.2649

VAR	COEFF	STD DEV COEFF	F-RATIO VAR. CONTR	BETA	COEFF
5	.1575E 02	.4337E 01	.1319E 02	.9333E 00	
10	.1189E 01	.2951E 00	.1623E 02	.6865E 00	
13	.5693E 00	.1791E 00	.1010E 02	.2350E 00	
24	-.1135E 01	.1758E 00	.4169E 02	-.5152E 00	
25	-.1474E 01	.6534E 00	.5089E 01	-.5646E 00	

VAR.
DESCR.:
 Q3
 Q2²
 R x C
 L x R-T
 Q2 x Q3

STEP NUMBER 9 ENTER VARIABLE 9
 STANDARD ERROR OF ESTIMATE = 34.007
 MULTIPLE CORRELATION COEFFICIENT = 0.552
 GOODNESS OF FIT: 30.4201%
 CONSTANT TERM = 179.6998

VAR	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	VAR. CONTR		
1	-.1666E 02	.2873E 02	.3364E 00	-.6052E 00	
2	-.9829E 01	.2030E 02	.2345E 00	-.5306E 00	
3	.6605E 01	.2434E 02	.7362E-01	.3011E 00	
4	.2774E 01	.2181E 01	.1618E 01	.1154E 01	
5	.3320E 00	.1066E 01	.9695E-01	.1945E 00	
6	.1173E 01	.1613E 01	.5287E 00	.6771E 00	
7	.1802E 01	.1777E 01	.1028E 01	.7440E 00	
8	-.2649E 01	.2387E 01	.1232E 01	-.8839E 00	
9	-.1233E 00	.1699E 01	.5264E-02	-.4948E-01	

VAR.
DESCR.
 A
 C
 Q2
 R2
 C2
 Q2
 A x C
 A x Q2
 C x Q2

STEP NUMBER 9 ENTER VARIABLE 3
 STANDARD ERROR OF ESTIMATE= 24.577
 MULTIPLE CORRELATION COEFFICIENT = 0.798
 GOODNESS OF FIT: 63.6594%
 CONSTANT TERM= 217.3154

VAR	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	VAR. CONTR		
1	.7748E 01	.1497E 02	.2678E 00	.4032E 00	
2	-.6948E 01	.1448E 02	.2304E 00	-.4116E 00	
3	-.2409E 01	.1336E 02	.3248E 01	-.1300E 00	
4	-.8716E 00	.8710E 00	.1001E 01	-.5655E 00	
5	.1022E 01	.8667E 00	.1392E 01	.6845E 00	
6	.3536E 00	.9338E 00	.1434E 00	.1873E 00	
7	.3343E 00	.9105E 00	.1348E 00	.1356E 00	
8	-.1258E 01	.8585E 00	.2147E 01	-.5709E 00	
9	-.3316E 00	.9201E 00	.1298E 00	-.1363E 00	

VAR.
DESCR.:
 L
 Q3
 A-T
 L²
 Q3²
 A-T²
 L x Q3
 L x A-T
 Q3 x A-T

RPP. D.4.
 RESULTS OF HSRRA III
 SUBSET: 16 Pf - L
 16 Pf - Q3
 A-TRAIT

NSTEP NUMBER 14 ENTER VARIABLE 7 SQUARE SELF CONCEPT CONTROL
 STANDARD ERROR OF ESTIMATE = 20.818
 MULTIPLE CORRELATION COEFFICIENT = 0.878
 GOODNESS OF FIT: 77.0299%
 CONSTANT TERM = -145.9194

	VARIABLE	COEFF	STD DEV	F-RATIO BETA COEFF		VAR DESCR.
				COEFF	VAR CONTR	
1	ALAXIA*PROTENDION(TRUSTING/SP)	.2659E 02	.2585E 02	.1057E 01	.1384E 01	L
2	GROUP ADHERENI*SELF*SUFFICIENT	.1666E 02	.2464E 02	.4569E 00	.7594E 00	Q2
3	SELF CONCEPT CONTROL	.4140E 02	.2197E 02	.3551E 01	.2453E 01	Q3
4	TRAIT*ANXIETY (STAI)	.3158E 02	.1980E 02	.2545E 01	.1704E 01	A-T
5	SQUARE*ALAXIA	-.1543E 01	.9624E 00	.2569E 01	-.1001E 01	L ²
6	SQUARE*GROUP ADHERENCE	.1064E 01	.1239E 01	.7372E 00	.6141E 00	Q2 ²
7	SQUARE*SELF CONCEPT CONTROL	.3796E-01	.8226E 00	.2130E-02	.2542E-01	Q3 ²
8	SQUARE*TRAIT ANXIETY	-.5944E 00	.8796E 00	.4567E 00	-.3150E 00	A-T ²
9	CRP*ALAXIA X GROUP ADH*	.1811E 00	.1335E 01	.1840E-01	.7787E-01	L x Q2
10	CRP*ALAXIA X SELF CONC/C	-.1108E 01	.1131E 01	.9604E 00	-.4495E 00	L x Q3
11	CRP*ALAXIA X A*TRAIT	-.1544E 01	.9116E 00	.2869E 01	-.7008E 00	L x A-T
12	CRP*GROUP ADH* X SELF CONC/C	-.2651E 01	.1285E 01	.4261E 01	-.1016E 01	Q2 x Q3
13	CRP*GROUP ADH* X A*TRAIT	-.1994E 01	.1235E 01	.2606E 01	-.8598E 00	Q2 x A-T
14	CRP*SELF CONC/C X A*TRAIT	-.2185E 01	.1136E 01	.3700E 01	-.8984E 00	Q3 x A-T

NSTEP NUMBER 2 ENTER VARIABLE 12 CRP1GROUP ADH. X SELF CONC/C
 STANDARD ERROR OF ESTIMATE= 22.344
 MULTIPLE CORRELATION COEFFICIENT = 0.806
 GOODNESS OF FIT: 64.9562%
 CONSTANT TERM= 198.9462

	VARIABLE	COEFF	STD DEV	F-RATIO	BETA	COEFF
		COEFF	COEFF	VAR. CONTR		
11	CRP1ALAXIA X A-TRAIT	.1340E 01	.2025E 00	.4380E 02		.6082E 00
12	CRP1GROUP ADH. X SELF CONC/C	.8928E 00	.2399E 00	.1385E 02		.3420E 00

VAR
 DESCR:
 L x R-T
 Q2 x Q3

APPENDIX E

E.1. FORTRAN - program used for MLRA's

E.2. FORTRAN - program used for MQRA's

E.3. FORTRAN - program used for RSA's

MLINEAR/PSPHANS

```

3 SET TAPE LISTP
FILE 2 = INPUT          ,UNIT = READER          R 0000
FILE 3 = OUTPUT        ,UNIT = PRINTER        T 0000
FILE 1=LINEAR/REGRESS, UNIT=DISK, RANDOM, AREA=1001, RECORD=30 00002000 T 0000
C                               00003000 T 0000
C -----
C MULTIPLE LINEAR REGRESSION - INPUT PHASE 00004000 T 0000
C                               00005000 T 0000
C                               00006000 T 0000
                                START OF SEGMENT ***** 1
COMMON I1, R1(30,30), R2(30), R3(30), R4 00007000 T 0000
COMMON NPLOT 00007100 R 0000 PATCH
DIMENSION X(30), NAME(10) 00008000 T 0000
C                               00009000 T 0000
C==== READ "NUMBER OF OBSERVATIONS", "NUMBER OF VARIABLES", AND
C "NAME OF PROJECT", 00010000 T 0000
READ (2,100) NOBS, NVAR, NAME 00011000 T 0000
100 FORMAT ( 2I4, 2X, 10A6 ) 00012000 T 0000
WRITE(3,953) 00013000 T 0015
WRITE (3,958) 00014000 R 0015 PATCH
WRITE(3,954) R 0018 PATCH
95a FORMAT (3X,"UBS,NO.", "PF/A",1X,"PF/B",1X,"PF/C",1X,"PF/E",1X,"PF/F" R 0022 PATCH
-,1X,"PF/G",1X,"PF/H",1X,"PF/I",1X,"PF/L",1X,"PF/M",1X,"PF/N",1X,"PF/" R 0025 PATCH
-F/O",1X,"P/Q1",1X,"PFQ2",1X,"P/Q3",1X,"P/Q4",1X,"STATE", "TRAIT",1X R 0026 PATCH
-, "INT.",1X,"AGE",2X,"EXP.",1X,"SCORE") R 0026 PATCH
95b FORMAT (10X,"ORIGINAL DATA MATRIX",//) R 0026 PATCH
953 FORMAT(10X,"PERSONALITY CORRELATES TO PERFORMANCE UNDER STRESS",10 R 0026 PATCH
-X,"H.P.LEHMANN",//) R 0026 PATCH
DO 20 J=1, NUBS 00015000 T 0026
C==== READ THE DEPENDANT AND INDEPENDANT VARIABLES (DEPENDANT LAST) 00016000 T 0026
READ (2,200) ( X(K), K=1,NVAR) 00017000 T 0031
C==== THE FOLLOWING FORMAT MAY BE CHANGED TO SUIT YOUR REQUIREMENTS 00018000 T 0045
C 00019000 T 0045
200 FORMAT (22F3.0) 00020000 R 0049 PATCH
C 00021000 T 0049
C==== DO 10 K=1, NVAR PERFORM ANY TRANSFORMATIONS OF DATA HERE, 00022000 T 0049
C==== X(K) = ALOG( X(K) ) E.G. A NAPERIAN LOGARITHM, AS SHOWN, 00023000 T 0049
C==== X(K) = X(K)**X(K) OR SQUARE, ETC, OR NONE. 00024000 T 0049
WRITE (3,955) J, (X(K), K=1,NVAR) 00025000 R 0049 PATCH
955 FORMAT(I10, 22F5.0) R 0068 PATCH
20 WRITE (1"J) X 00026000 T 0068
C==== NOW CALL THE SUBROUTINE THAT DOES ALL THE LINEAR REGRESSING 00027000 T 0069
CALL REGR1 00028000 T 0079
C==== READ THE "F VALUES" FOR VARIABLES (1) TO ENTER, & (2) TO LEAVE 00029000 T 0079
C THE REGRESSION EQUATION, AND AN OPTION TO PRINT THE RESIDUALS. 00030000 T 0079
READ (2,300,END=50)FIN,FOUT,IRES,NPLOT 00031000 R 0079 PATCH
300 FORMAT(2F10.5,2I2) 00032000 R 0097 PATCH
CALL REGR2 ( FIN, FOUT, IRES ) 00033000 T 0098
STOP 00034000 T 0099
50 CALL REGR2 ( 0.0, 0.0, 1 ) 00035000 T 0101
STOP 00036000 T 0102
END 00037000 T 0104
C 00038000 T 0104
C 00039000 T 0104
SEGMENT 1 IS 115 LONG
    
```

140.

APP. E.1.
FORTRAN PROGRAM
USED FOR THE HLRA'S

```

SUBROUTINE REGR1
C--- MULTIPLE LINEAR REGRESSION - PHASE 1
COMMON NVAR, R(30,30), XBAR(30), SIGMA(30), OBS
DIMENSION X(30), NAME(10)
C
READ (1*0) NUBS, NVAR, NAME
OBS=NUBS
DO 2 J=1,NVAR
XBAR(J)=0
DO 2 K=1,NVAR
2 R(J,K)=0
DO 3 I=1,NUBS
READ (1*I) X
DO 4 J=1,NVAR
XBAR(J)=XBAR(J)+X(J)
DO 4 K=1,J
R(J,K)=R(J,K)+X(J)*X(K)
4 R(K,J)=R(J,K)
3 CONTINUE
DO 5 J=1,NVAR
5 SIGMA(J)=SQRT(R(J,J)-XBAR(J)*XBAR(J)/OBS)
DO 6 J=1,NVAR
DO 6 K=1,J
R(J,K)=(R(J,K)-XBAR(J)*XBAR(K)/OBS)/(SIGMA(J)*SIGMA(K))
6 R(K,J)=R(J,K)
DEN=SQRT(OBS*1.)
DO 7 J=1,NVAR
XBAR(J)=XBAR(J)/OBS
7 SIGMA(J)=SIGMA(J)/DEN
WRITE(3,200)NAME
200 FORMAT ( 11X, 10A6 // )
WRITE(3,201)
201 FORMAT(1H0" VAR          MEAN   STD DEV"/)
WRITE(3,202)(J,XBAR(J),SIGMA(J),J=1,NVAR)
202 FORMAT(1X, I4, 2F10.2)
WRITE(3,203)
203 FORMAT(1H0, 10X, "SIMPLE CORRELATION COEFFICIENTS")
DO 10 J=1,NVAR
10 WRITE(3,204)(J,K,R(J,K),K=1,NVAR)
204 FORMAT(6(2I4,F8.4,4X))
C
CALL LINK(REGR2)
RETURN
END
C
C
C
C
C
C
C
C
C
C

```

```

START OF SEGMENT ***** 2
00040000 T 0000
00041000 T 0000
00042000 T 0000
00043000 T 0000
00044000 T 0000
00045000 T 0000
00046000 T 0017
00047000 T 0018
00048000 T 0024
00049000 T 0026
00050000 T 0033
00051000 T 0036
00052000 T 0042
00053000 T 0055
00054000 T 0061
00055000 T 0065
00056000 T 0070
00057000 T 0078
00058000 T 0084
00059000 T 0084
00060000 T 0091
00061000 T 0098
00062000 T 0105
00063000 T 0111
00064000 T 0122
00065000 T 0127
00066000 T 0129
00067000 T 0135
00068000 T 0139
00069000 T 0142
00070000 T 0156
00071000 T 0156
00072000 T 0160
00073000 T 0160
00074000 T 0183
00075000 T 0183
00076000 T 0187
00077000 T 0187
00078000 T 0194
00079000 T 0217
00080000 T 0217
00081000 T 0217
00082000 T 0220
00083000 T 0220
00084000 T 0220
00085000 T 0220
00086000 T 0220
00087000 T 0220
00088000 T 0220
00089000 T 0220
SEGMENT 2 IS 230 LONG

```

141.

E.1/2

```

SUBROUTINE REGR2 ( FIN, FOUT, IRES )
C--- MULTIPLE LINEAR REGRESSION - PHASE 2
COMMON NVAR, RIJ(30,30), XBAR(30), SIGMA(30), OBS
COMMON NPLOT
DIMENSION SIGB(30), B(30), ID(30), DATA(30)
DIMENSION RPLLOT(1000)

C
NINDV=NVAR-1
NOBS=OBS
C PHASE 2, PERFORM STEPWISE CALCULATIONS AND PRINT RESULTS.
C DIMENSIONS
C INITIALIZE
DO 190 I=1,NVAR
SIGB(I)=0.0
190 B(I)=0.0
NENT=0
DF=OBS-1.0
NSTEP=1
C TRANSFORM SIGMA VECTOR FROM STANDARD DEVIATIONS TO SQUARE
C ROOTS OF SUMS OF SQUARES.
DO 310 I=1,NVAR
310 SIGMA(I)=SIGMA(I)*SQRT(OBS-1.0)
C
C BEGIN STEP NUMBER NSTEP.
200 NSTEP=NSTEP+1
WRITE (3,300)
3000 FORMAT("1")
STDEE=SQRT(RIJ(NVAR,NVAR)/DF) * SIGMA(NVAR)
DF=DF-1.0
IF(DF)1010,1010,205
205 VMIN=0.0
VMAX=0.0
NIN=0
C FIND MINIMUM VARIANCE CONTRIBUTION OF VARIABLES IN REGRESSION
C EQUATION, FIND MAXIMUM VARIANCE CONTRIBUTION OF VARIABLES
C NOT IN REGRESSION EQUATION.
DO 300 I=1,NINDV
IF(RIJ(I,I)=.001)300,300,210
210 VI=RIJ(I,NVAR)*RIJ(NVAR,I)/RIJ(I,I)
IF(VI)240,300,220
220 IF(VI-VMAX)300,300,230
230 VMAX=VI
NMAX=I
GO TO 300
240 NIN=NIN+1
ID(NIN)=I
C COMPUTE REGRESSION COEFFICIENT AND ITS STANDARD DEVIATION.
B(NIN)=RIJ(I,NVAR)*SIGMA(NVAR)/SIGMA(I)
SIGB(NIN)=STDEE*SQRT(RIJ(I,I))/SIGMA(I)
IF(VMIN)250,260,1000
250 IF(VI-VMIN)300,300,260
260 VMIN=VI
NMIN=I
300 CONTINUE
IF(NIN)1000,460,400
C COMPUTE CONSTANT TERM.
400 BSUB0=XBAR(NVAR)
DO 410 I=1,NIN
J=ID(I)
410 BSUB0=BSUB0-B(I)*XBAR(J)
IF(NENT)1000,480,420
C OUTPUT FOR VARIABLE ADDED

```

START OF SEGMENT ***** 3

```

00090000 T 0000
00091000 T 0000
00092000 T 0000.
00092100 R 0000 PATCH
00093000 T 0000
00093100 R 0000 PATCH
00094000 T 0000
00095000 T 0000
00096000 T 0001
00097000 T 0001
00098000 T 0001
00099000 T 0001
00100000 T 0003
00101000 T 0009
00102000 T 0012
00103000 T 0014
00104000 T 0015
00105000 T 0018
00106000 T 0018
00107000 T 0018
00108000 T 0019
00109000 T 0027
00110000 T 0032
00111000 T 0032
00112000 T 0034
00112100 R 0035 PATCH
R 0038 PATCH
00113000 T 0039
00114000 T 0044
00115000 T 0047
00116000 T 0050
00117000 T 0050
00118000 T 0051
00119000 T 0051
00120000 T 0051
00121000 T 0051
00122000 T 0052
00123000 T 0057
00124000 T 0065
00125000 T 0072
00126000 T 0078
00127000 T 0081
00128000 T 0081
00129000 T 0082
00130000 T 0084
00131000 T 0085
00132000 T 0086
00133000 T 0087
00134000 T 0093
00135000 T 0099
00136000 T 0104
00137000 T 0107
00138000 T 0107
00139000 T 0109
00140000 T 0109
00141000 T 0109
00142000 T 0115
00143000 T 0116
00144000 T 0122
00145000 T 0125
00146000 T 0128
00147000 T 0129

```

```

420 WRITE(3,57)NSTEP,K
57 FORMAT("OSTEP NUMBER ",I2,10X,"ENTER VARIABLE ",I2)
425 WRITE(3,58)STDEE
58 FORMAT(" STANDARD ERROR OF ESTIMATE=",F11.3)

R=SQRT(1.0*RIJ(NVAR,NVAR))
WRITE(3,59)R
59 FORMAT(" MULTIPLE CORRELATION COEFFICIENT =",F6.3)
IDFN=OBS-DF-2.0
IDFD=DF+1.0
-----
C EXPRESS GOODNESS OF FIT AS PERCENTAGE FIT, GIVING THE RATIO OF
C SUMS OF SQUARES DUE TO REGRESSION VERSUS THE TOTAL SUM OF SQUARES
C-----
FIT=(R**2)*100
WRITE(3,66)FIT
66 FORMAT(1X,"GOODNESS OF FIT:",F8.4,"%")
WRITE(3,60)BSUBD
60 FORMAT(" CONSTANT TERM=",F12.4)
WRITE(3,61)
61 FORMAT(3X,"VAR",5X,"COEFF",5X,"STD DEV",4X,"F-RATIO",1X,"BETA COE
FF")
WRITE(3,62)
62 FORMAT(23X,"COEFF",2X,"VAR,CONTR")
DO 430 I=1,NIN
J=ID(I)
T=8(I)/SIGB(I)
-----
C INSTEAD OF USING THE RAW T-VALUE USE T T-SQU AS F-RATIO FOR THE
C SIGNIFICANCE OF THE CONTRIBUTION OF VARIABLES TO THE OVERALL FIT
C-----
T=T**2
430 WRITE(3,63)ID(I),B(I),SIGB(I),T,RIJ(J,NVAR)
63 FORMAT(3X,I3,4E11.4)
C COMPUTE F LEVEL FOR MINIMUM VARIANCE CONTRIBUTION VARIABLE
C IN REGRESSION EQUATION.
FLEVL=VMIN*DF/RIJ(NVAR,NVAR)
IF(FOUT+FLEVL)460,460,450
C INITIALIZE FOR REMOVAL OF VARIABLE K FROM EQUATION.
450 K=NMN
NENT=0
DF=DF+2.0
GO TO 500
C COMPUTE F LEVEL FOR MAXIMUM VARIANCE CONTRIBUTION VARIABLE
C NOT IN EQUATION.
460 FLEVL=VMAX*DF/(RIJ(NVAR,NVAR)-VMAX)
IF(FLEVL-FIN)600,600,470
C INITIALIZE FOR ENTRY OF VARIABLE K INTO EQUATION.
470 K=NMAX
NENT=K
GO TO 500
C OUTPUT FOR VARIABLE DELETED
480 WRITE(3,64)NSTEP,K
64 FORMAT("OSTEP NUMBER ",I2,10X,"DELETE VARIABLE ",I2)
GO TO 425
C
C UPDATE MATRIX
500 STDEE=1.0/RIJ(K,K)
DO 540 I=1,NVAR
IF(I=K)510,540,510
510 DO 530 J=1,NVAR
IF(J=K)520,530,520
520 RIJ(I,J)=RIJ(I,J)-RIJ(I,K)*RIJ(K,J)*STDEE

```

```

00148000 T 0134
00149000 T 0146
00150000 T 0146
00151000 T 0156
SEGMENT A IS 126 LONG
00152000 T 0156
00153000 T 0161
00154000 T 0172
00155000 T 0172
00156000 T 0176
00157000 R 0176 PATCH
R 0176 PATCH
R 0176 PATCH
00158000 R 0176 PATCH
R 0180 PATCH
00159000 R 0182 PATCH
00160000 R 0191 PATCH
00161000 T 0192
00162000 T 0202
00163000 T 0202
00164000 R 0205 PATCH
R 0206 PATCH
00165000 T 0206
00166000 R 0209 PATCH
00167000 T 0210
00168000 T 0215
00169000 T 0217
00169050 R 0218 PATCH
00169100 R 0218 PATCH
R 0218 PATCH
R 0218 PATCH
R 0220 PATCH
00170000 T 0222
00171000 R 0245 PATCH
00172000 T 0245
00173000 T 0245
00174000 T 0245
00175000 T 0249
00176000 T 0249
00177000 T 0252
00178000 T 0252
00179000 T 0253
00180000 T 0256
00181000 T 0256
00182000 T 0256
00183000 T 0258
00184000 T 0262
00185000 T 0263
00186000 T 0266
00187000 T 0266
00188000 T 0267
00189000 T 0267
00190000 T 0269
00191000 T 0281
00192000 T 0281
00193000 T 0281
00194000 T 0281
00195000 T 0282
00196000 T 0286
00197000 T 0293
00198000 T 0297
00199000 T 0303
00200000 T 0307

```

5.1/4

530	CONTINUE				00201000	T	0317	
540	CONTINUE				00202000	T	0318	
	DO 560 J=1,NVAR				00203000	T	0318	
	IF(J=K)550,560,550				00204000	T	0324	
550	RIJ(K,J)=RIJ(K,J)*STDEE				00205000	T	0328	
560	CONTINUE				00206000	T	0333	
	DO 580 I=1,NVAR				00207000	T	0333	
	IF(I=K)570,580,570				00208000	T	0339	
570	RIJ(I,K)=RIJ(I,K)*STDEE				00209000	T	0343	
580	CONTINUE				00210000	T	0349	
	RIJ(K,K)=STDEE				00211000	T	0349	
	GO TO 200				00212000	T	0352	
					00213000	T	0352	
C					00214000	T	0353	
600	IF(IRES)610,640,610				00215000	T	0353	
C	PRINT RESIDUALS				00216000	T	0356	
610	IFA=2				00217000	T	0356	
	WRITE(3,67)				00218000	T	0361	
67	FORMAT('O OBS ACTUAL ESTIMATE RESIDUAL')				00219000	T	0361	
	DO 630 K=1,NOBS				00220000	T	0366	
	READ (1,K) (DATA(I), I=1,NVAR)				00221000	T	0384	
	EST=HSURO				00222000	T	0385	
	DO 620 I=1,NIN				00223000	T	0390	
	J=ID(I)				00224000	T	0393	
620	EST=EST+B(I)*DATA(J)				00225000	T	0396	
	RESID=DATA(NVAR)-EST				00225100	R	0399	PATCH
	RPLOT (K)=RESID				00226000	T	0400	
	WRITE(3,68)K,DATA(NVAR),EST,RESID				00227000	T	0418	
68	FORMAT(' I4,3F12.2')				00228000	T	0418	
630	CONTINUE				00228100	R	0418	PATCH
	IF (NPLOT,EQ,1) CALL PLOT(RPLOT,NVAR,NOBS)				00229000	T	0422	
C--	NORMAL END OF PROGRAM				00230000	T	0423	
640	STOP				00231000	T	0425	
1000	CALL ERROR(1)				00232000	T	0426	
1010	CALL ERROR(2)				00233000	T	0427	
1020	CALL ERROR(3)				00234000	T	0427	
	END							
					SEGMENT	3	IS	453 LONG

					START OF SEGMENT	*****	5	
	SUBROUTINE ERROR(I)				00235000	T	0000	
	GO TO (10, 20, 30), I				00236000	T	0000	
10	WRITE (3,100)				00237000	T	0007	
100	FORMAT ('OERRROR... NIN, NENT, VMIN, NCONS, OR NTRAM IS NEGATIVE. *CHECK FOR ERROR ON PARAMETER CARDS')				00238000	T	0011	
	STOP				00239000	T	0011	
20	WRITE (3,200)				00240000	T	0011	
200	FORMAT ('OERRROR... DEGREES OF FREEDOM = 0. EITHER ADD MORE OBSERVATIONS OR DELETE ONE OR MORE INDEPENDANT VARIABLE.' / 'SAMPLE SIZE *E MUST EXCEED NUMBER OF INDEPENDANT VARIABLES BY AT LEAST 2.')				00241000	T	0013	
	STOP				00242000	T	0017	
30	WRITE (3,300)				00243000	T	0017	
300	FORMAT ('OERRROR... F LEVEL FOR INCOMING VARIABLE IS LESS THAN F LEVEL FOR OUTGOING VARIABLE.')				00244000	T	0017	
	STOP				00245000	T	0017	
	END				00246000	T	0019	
					00247000	T	0023	
					00248000	T	0023	
					SEGMENT	6	IS	119 LONG
					00249000	T	0023	
					00250000	T	0024	

```

SUBROUTINE PLOT(Y,NVAR,NOBS)
DIMENSION Y(NOBS),X(1000),DUMMY(30),GRAPH(50,100)
NX=20
NY=20
C-----CALCULATE THE MAXIMUM AND MINIMUM OF ORDINATE Y
CALL MAXMIN(Y,NOBS,YMAX,YMIN)
YFACT=(YMAX-YMIN)/(NY-1)
C-----ONE LOOP FOR EACH GRAPH
DO 500 J=1,NVAR
WRITE(3,3000)
3000 FORMAT("1")
C-----READ IN VARIABLE J
DO 50 I=1,NOBS
READ(1,"I")(DUMMY(L),L=1,NVAR)
X(I)=DUMMY(J)
50 CONTINUE
C-----NOW FIND THE MAXIMUM AND MINIMUM OF X
CALL MAXMIN(X,NOBS,XMAX,XMIN)
XFACT=(XMAX-XMIN)/(NX-1)
C-----INITIALISE GRAPH
DO 40 I=1,NX+1
DO 40 K=1,NY+1
40 GRAPH(K,I)=0
C-----PLOT POINTS
DO 100 I=1,NOBS
NXPOS=IFIX((X(I)-XMIN)/XFACT+0.5)+1
NYPOS=IFIX((Y(I)-YMIN)/YFACT+0.5)+1
GRAPH(NYPOS,NXPOS)=GRAPH(NYPOS,NXPOS)+1
100 CONTINUE
C-----PUT IN X AND Y AXES
DO 60 I=1,NY
GRAPH(I,1)=13
60 CONTINUE
DO 70 I=1,NX
GRAPH(1,I)=44
70 CONTINUE
C-----NOW PRINT THE GRAPH
DO 120 II=1,NY+1
DO 120 JJ=1,NX+1
IF(GRAPH(II,JJ).EQ.0)GRAPH(II,JJ)=48
120 CONTINUE
DO 200 I=1,NY+1
II=NY-I+2
WRITE(3,1000)(GRAPH(II,K),K=1,NX+1)
1000 FORMAT(1X,130C1)
200 CONTINUE
C-----PRINT OUT GRAPH DATA
WRITE(3,2000)J,XMIN,XMAX,XFACT,YMIN,YMAX,YFACT,NOBS
2000 FORMAT(" GRAPH OF VARIABLE",I4," VS. RESIDUALS "//
" X-AXIS BOUNDS ARE1",F12.6,3X,F12.6," X INCREMENT =",F12.6/
" Y-AXIS BOUNDS ARE1",F12.6,3X,F12.6," Y INCREMENT =",F12.6/
" NO. OF OBSERVATIONS = ",I4)
500 CONTINUE

```

```

00300000 R 0029 PATCH
START OF SEGMENT ***** 7
R 0000 PATCH
R 0000 PATCH
R 0000 PATCH
R 0000 PATCH
R 0001 PATCH
R 0003 PATCH
R 0005 PATCH
R 0006 PATCH
R 0011 PATCH
R 0015 PATCH
R 0016 PATCH
R 0016 PATCH
R 0021 PATCH
R 0039 PATCH
R 0041 PATCH
R 0042 PATCH
R 0042 PATCH
R 0044 PATCH
R 0046 PATCH
R 0047 PATCH
R 0053 PATCH
R 0060 PATCH
R 0065 PATCH
R 0067 PATCH
R 0072 PATCH
R 0078 PATCH
R 0084 PATCH
R 0094 PATCH
R 0095 PATCH
R 0100 PATCH
R 0105 PATCH
R 0105 PATCH
R 0110 PATCH
R 0115 PATCH
R 0116 PATCH
R 0116 PATCH
R 0122 PATCH
R 0129 PATCH
R 0140 PATCH
R 0142 PATCH
R 0148 PATCH
R 0150 PATCH
R 0172 PATCH
R 0173 PATCH
R 0173 PATCH
R 0173 PATCH
R 0197 PATCH
R 0198 PATCH
R 0198 PATCH
R 0198 PATCH
R 0198 PATCH
R 0198 PATCH

```

END

SEGMENT R 0201 PATCH
7 IS 221 LONG

SUBROUTINE MAXMIN (A,N,RMAX,RMIN)

DIMENSION A(N)
RMAX=-1E40
RMIN=1E40
DO 100 J=1,N
IF(A(J).GT.RMAX)RMAX=A(J)
IF(A(J).LT.RMIN)RMIN=A(J)
100 CONTINUE
RMAX = IFIX(RMAX+1)
RMIN = IFIX(RMIN-1)
RETURN
END

00500000 R 0220 PATCH
START OF SEGMENT ***** 8
R 0000 PATCH
R 0000 PATCH
R 0002 PATCH
R 0005 PATCH
R 0010 PATCH
R 0015 PATCH
R 0020 PATCH
R 0021 PATCH
R 0023 PATCH
R 0025 PATCH
R 0028 PATCH
SEGMENT 8 IS 39 LONG

99999999 R 0038 PATCH
SEGMENT 9 IS 53 LONG
SEGMENT 10 IS 29 LONG
SEGMENT 11 IS 138 LONG
START OF SEGMENT ***** 12
SEGMENT 12 IS 16 LONG

NUMBER OF CARDS = 345
COMPILATION TIME = 17 SECS

CORE MEMORY ALLOCATION = 8064 WORDS
ELAPSED TIME = 55 SECS


```

C THE REGRESSION EQUATION, AND AN OPTION TO PRINT THE RESIDUALS. 00030000 T 0222
READ (2,300,END=50)FIN,FOUT,IRES,NPLOT 00031000 R 0223 PATCH
300 FORMAT(2F10.5,2I2) 00032000 R 0240 PATCH
CALL REGR2 ( FIN, FOUT, IRES ) 00033000 T 0241
STOP 00034000 T 0242
50 CALL REGR2 ( 0.0, 0.0, 1 ) 00035000 T 0244
STOP 00036000 T 0245
END 00037000 T 0247
C 00038000 T 0247
C 00039000 T 0247
SEGMENT 1 IS 261 LONG

```

```

SUBROUTINE REGR1
MULTIPLE LINEAR REGRESSION - PHASE 1
COMMON NVAR, R(30,30), XBAR(30), SIGMA(30), OBS
DIMENSION X(30), NAME(10)

READ (1"0) NUBS, NVAR, NAME
OBS=NUBS
DO 2 J=1,NVAR
XBAR(J)=0.
DO 2 K=1,NVAR
R(J,K)=0.
DO 3 I=1,NOBS
READ (1"1) X
DO 4 J=1,NVAR
XBAR(J)=XBAR(J)+X(J)
DO 4 K=1,J
R(J,K)=R(J,K)+X(J)*X(K)
4 R(K,J)=R(J,K)
3 CONTINUE
DO 5 J=1,NVAR
5 SIGMA(J)=SQRT(R(J,J)-XBAR(J)*XBAR(J)/OBS)
DO 6 J=1,NVAR
DO 6 K=1,J
R(J,K)=(R(J,K)-XBAR(J)*XBAR(K)/OBS)/(SIGMA(J)*SIGMA(K))
6 R(K,J)=R(J,K)
DEN=SQRT(OBS-1.)
DO 7 J=1,NVAR
XBAR(J)=XBAR(J)/OBS
7 SIGMA(J)=SIGMA(J)/DEN
WRITE (3,300)
3000 FORMAT ("1")
WRITE (3,3001)
3001 FORMAT (10X,"MULTIPLE QUADRATIC REGRESSION")
WRITE (3,3002)
3002 FORMAT (10X,"-----")
WRITE(3,200)NAME
200 FORMAT (11X,10A6 // )
WRITE(3,201)
201 FORMAT(11H" VAR MEAN STD DEV"/)
WRITE(3,202)(J,XBAR(J),SIGMA(J),J=1,NVAR)
202 FORMAT(1X,14,2F10.2)
WRITE(3,203)
203 FORMAT(11H,10X,"SIMPLE CORRELATION COEFFICIENTS")

```

START OF SEGMENT ***** 2

```

00040000 T 0000
00041000 T 0000
00042000 T 0000
00043000 T 0000
00044000 T 0000
00045000 T 0000
00046000 T 0017
00047000 T 0018
00048000 T 0024
00049000 T 0026
00050000 T 0033
00051000 T 0036
00052000 T 0042
00053000 T 0055
00054000 T 0061
00055000 T 0065
00056000 T 0070
00057000 T 0078
00058000 T 0084
00059000 T 0084
00060000 T 0091
00061000 T 0098
00062000 T 0105
00063000 T 0111
00064000 T 0122
00065000 T 0127
00066000 T 0129
00067000 T 0135
00068000 T 0139
00068100 R 0142 PATCH
R 0146 PATCH
R 0146 PATCH
R 0149 PATCH
R 0150 PATCH
R 0153 PATCH
00069000 T 0154
00070000 T 0167
00071000 T 0167
00072000 T 0171
00073000 T 0171
00074000 T 0194
00075000 T 0194
00076000 T 0198

```

```

DO 10 J=1,NVAR
10 WRITE(3,204)(J,K,R(J,K),K=1,NVAR)
204 FORMAT(6(2I4,F8.4,4X))
C CALL LINK(REGR2)
RETURN
END

```

```

C
C
C
C
C
C
C

```

```

SEGMENT 3 IS 126 LONG
00077000 T 0198
00078000 T 0205
00079000 T 0228
00080000 T 0228
00081000 T 0228
00082000 T 0231
00083000 T 0231
00084000 T 0231
00085000 T 0231
00086000 T 0231
00087000 T 0231
00088000 T 0231
00089000 T 0231
SEGMENT 2 IS 241 LONG

```

```

-----
C--- SUBROUTINE REGR2 ( FIN, FOUT, IRES )
MULTIPLE LINEAR REGRESSION - PHASE 2
COMMON NVAR, RIJ(30,30), XBAR(30), SIGMA(30), OBS
COMMON NPLUT
DIMENSION SIGB(30), B(30), ID(30), DATA(30)
DIMENSION RPLUT(1000)

C NINDV=NVAR-1
NORS=OBS
C PHASE 2. PERFORM STEPWISE CALCULATIONS AND PRINT RESULTS.
C DIMENSIONS
C INITIALIZE
DO 190 I=1,NVAR
190 SIGB(I)=0.0
B(I)=0.0
NENT=0
DF=OBS-1.0
NSTEP=-1
C TRANSFORM SIGMA VECTOR FROM STANDARD DEVIATIONS TO SQUARE
C ROOTS OF SUMS OF SQUARES.
DO 310 I=1,NVAR
310 SIGMA(I)=SIGMA(I)*SQRT(OBS-1.0)
C BEGIN STEP NUMBER NSTEP.
200 NSTEP=NSTEP+1
WRITE (3,3000)
3000 FORMAT("1")
STDFE=SQRT(RIJ(NVAR,NVAR)/DF) * SIGMA(NVAR)
DF=DF-1.0
IF(DF)1010,1010,205
205 VMIN=0.0
VMAX=0.0
NIN=0
C FIND MINIMUM VARIANCE CONTRIBUTION OF VARIABLES IN REGRESSION
C EQUATION. FIND MAXIMUM VARIANCE CONTRIBUTION OF VARIABLES
C NOT IN REGRESSION EQUATION.
DO 300 I=1,NINDV
IF(RIJ(I,1)-.001)300,300,210
210 VI=RIJ(I,NVAR)*RIJ(NVAR,I)/RIJ(I,I)

```

```

START OF SEGMENT ***** 4
00090000 T 0000
00091000 T 0000
00092000 T 0000
00092100 R 0000 PATCH
00093000 T 0000
00093100 R 0000 PATCH
00094000 T 0000
00095000 T 0000
00096000 T 0001
00097000 T 0001
00098000 T 0001
00099000 T 0001
00100000 T 0003
00101000 T 0009
00102000 T 0012
00103000 T 0014
00104000 T 0015
00105000 T 0018
00106000 T 0018
00107000 T 0018
00108000 T 0019
00109000 T 0027
00110000 T 0032
00111000 T 0032
00112000 T 0034
00112100 R 0035 PATCH
R 0038 PATCH
00113000 T 0039
00114000 T 0044
00115000 T 0047
00116000 T 0050
00117000 T 0050
00118000 T 0051
00119000 T 0051
00120000 T 0051
00121000 T 0051
00122000 T 0052
00123000 T 0057
00124000 T 0055

```



```

DF=DF+2.0
GO TO 500
C COMPUTE F LEVEL FOR MAXIMUM VARIANCE CONTRIBUTION VARIABLE
C NOT IN EQUATION.
460 FLEVEL=VMAX*DF/(RIJ(NVAR,NVAR)-VMAX)
IF(FLEVEL-FIN)600,600,470
C INITIALIZE FOR ENTRY OF VARIABLE K INTO EQUATION.
470 K=NMAX
NENT=K
GO TO 500
C OUTPUT FOR VARIABLE DELETED
480 WRITE(3,64)NSTEP,K
64 FORMAT("STEP NUMBER ",I2,10X,"DELETE VARIABLE ",I2)
GO TO 425
C
C UPDATE MATRIX
500 STDEF=1.0/RIJ(K,K)
DO 540 I=1,NVAR
IF(I-K)510,540,510
510 DO 530 J=1,NVAR
IF(J-K)520,530,520
520 RIJ(I,J)=RIJ(I,J)-RIJ(I,K)*RIJ(K,J)*STDEF
530 CONTINUE
540 CONTINUE
DO 560 J=1,NVAR
IF(J-K)550,560,550
550 RIJ(K,J)=RIJ(K,J)*STDEF
560 CONTINUE
DO 580 I=1,NVAR
IF(I-K)570,580,570
570 RIJ(I,K)=-RIJ(I,K)*STDEF
580 CONTINUE
RIJ(K,K)=STDEF
GO TO 200
C
C 600 IF(IRES)610,640,610
C PRINT RESIDUALS
610 IFA=2
WRITE(3,3004)
WRITE(3,67)
67 FORMAT("O OBS            ACTUAL    ESTIMATE    RESIDUAL")
DO 630 K=1,NOBS
READ(1,K)(DATA(I), I=1,NVAR)
FST=BSUBD
DO 620 I=1,NIN
J=ID(I)
620 FST=EST+B(I)*DATA(J)
RESID=DATA(NVAR)-EST
RPLLOT(K)=RESID
WRITE(3,68)K,DATA(NVAR),EST,RESID
68 FORMAT(" ",I4,3F12.2)
630 CONTINUE
IF(NPLOT,EQ,1) CALL PLOT(RPLOT,NVAR,NOBS)
C--- NORMAL END OF PROGRAM
640 STOP
1000 CALL ERROR(1)
1010 CALL ERROR(2)
1020 CALL ERROR(3)
END

```

```

00179000 T 0253
00180000 T 0256
00181000 T 0256
00182000 T 0256
00183000 T 0258
00184000 T 0262
00185000 T 0263
00186000 T 0266
00187000 T 0266
00188000 T 0267
00189000 T 0267
00190000 T 0269
00191000 T 0281
00192000 T 0281
00193000 T 0281
00194000 T 0281
00195000 T 0282
00196000 T 0286
00197000 T 0293
00198000 T 0297
00199000 T 0303
00200000 T 0307
00201000 T 0317
00202000 T 0318
00203000 T 0318
00204000 T 0324
00205000 T 0328
00206000 T 0333
00207000 T 0333
00208000 T 0339
00209000 T 0343
00210000 T 0349
00211000 T 0349
00212000 T 0352
00213000 T 0352
00214000 T 0353
00215000 T 0353
00216000 T 0356
00216100 R 0356 PATCH
00217000 T 0360
00218000 T 0364
00219000 T 0364
00220000 T 0369
00221000 T 0387
00222000 T 0388
00223000 T 0393
00224000 T 0396
00225000 T 0399
00225100 R C402 PATCH
00226000 T 0403
00227000 T 0421
00228000 T 0421
00228100 R 0421 PATCH
00229000 T 0425
00230000 T 0426
00231000 T 0429
00232000 T 0429
00233000 T 0430
00234000 T 0430
SEGLCNT 4 IS 456 LONG

```



```

                                START OF SEGMENT ***** 5
SUBROUTINE ERROR( I )
GO TO (10, 20, 30), I
10 WRITE (3,100)
100 FORMAT ("OERROR... NIN, NENT, YMIN, NCONS, OR NYRAN IS NEGATIVE,
*CHECK FOR ERROR ON PARAMETER CARDS" )
STOP
20 WRITE (3,200)
200 FORMAT ("OERROR... DEGREES OF FREEDOM = 0, EITHER ADD MORE OBSERVATIONS OR
*DELETE ONE OR MORE INDEPENDANT VARIABLE." / "SAMPLE SIZE MUST EXCEED NUMBER OF INDEPENDANT VARIABLES BY AT LEAST 2." )
STOP
30 WRITE (3,300)
300 FORMAT ("OERROR... F LEVEL FOR INCOMING VARIABLE IS LESS THAN F L
*VLL FOR OUTGOING VARIABLE." )
STOP
END
                                SEGMENT 6 IS 118 LONG
                                SEGMENT 5 IS 30 LONG

```

```

SUBROUTINE PLOT(Y,NVAR,NOBS)
DIMENSION Y(NOBS),X(1000),DUMMY(30),GRAPH(50,100)
NX=80
NY=20
C-----CALCULATE THE MAXIMUM AND MINIMUM OF ORDINATE Y
CALL MAXMIN(Y,NOBS,YMAX,YMIN)
YFACT=(YMAX-YMIN)/(NY-1)
C-----ONE LOOP FOR EACH GRAPH
DO 500 J=1,NVAR
WRITE(3,3000)
3000 FORMAT("1")
C-----READ IN VARIABLE J
DO 50 I=1,NOBS
READ (1,"I")(DUMMY(L),L=1,NVAR)
X(I)=DUMMY(J)
50 CONTINUE
C-----NOW FIND THE MAXIMUM AND MINIMUM OF X
CALL MAXMIN(X,NOBS,XMAX,XMIN)
XFACT=(XMAX-XMIN)/(NX-1)
C----- INITIALISE GRAPH
DO 40 I=1,NX+1
DO 40 K=1,NY+1
40 GRAPH (K,I)=0
C----- PLOT POINTS
DO 100 I=1,NOBS
NXPOS=IFIX((X(I)-XMIN)/XFACT+0.5)+1
NYPOS=IFIX((Y(I)-YMIN)/YFACT+0.5)+1
GRAPH(NYPOS,NXPOS)=GRAPH(NYPOS,NXPOS)+1
100 CONTINUE
C----- PUT IN X AND Y AXES
DO 60 I=1,NY

```

```

                                00300000 R 0029
START OF SEGMENT ***** 7
R 0000 PATCH
R 0000 PATCH
R 0000 PATCH
R 0001 PATCH
R 0003 PATCH
R 0005 PATCH
R 0006 PATCH
R 0011 PATCH
R 0015 PATCH
P 0016 PATCH
R 0016 PATCH
P 0021 PATCH
R 0039 PATCH
R 0041 PATCH
R 0042 PATCH
R 0042 PATCH
R 0044 PATCH
R 0046 PATCH
R 0047 PATCH
R 0053 PATCH
R 0060 PATCH
R 0065 PATCH
R 0067 PATCH
R 0072 PATCH
R 0078 PATCH
R 0084 PATCH
R 0094 PATCH
R 0095 PATCH
R 0095 PATCH

```

```

    GRAPH(I,1)=13
60  CONTINUE
    DO 70 I=1,NX
        GRAPH(1,I)=44
70  CONTINUE
C---- NOW PRINT THE GRAPH
    DO 120 II=1,NY+1
        DO 120 JJ = 1,NX+1
            IF(GRAPH(II,JJ).EQ.0)GRAPH(II,JJ)=48
120  CONTINUE
    DO 200 I=1,NY+1
        II=NY-I+2
        WRITE(3,1000)(GRAPH(II,K),K=1,NX+1)
1000 FORMAT(1X,130C1)
200  CONTINUE
C---- PRINT OUT GRAPH DATA
    WRITE(3,2000)J,XMIN,XMAX,XFACT,YMIN,YMAX,YFACT,NOBS
2000  FORMAT(" GRAPH OF VARIABLE",I4," VS. RESIDUALS "//
    " X-AXIS BOUNDS ARE:",F12.6,3X,F12.6," X INCREMENT =",F12.6/
    " Y-AXIS BOUNDS ARE:",F12.6,3X,F12.6," Y INCREMENT =",F12.6/
    " NO. OF OBSERVATIONS = ",I4)
500  CONTINUE
    RETURN
    END

```

```

R 0100 PATCH
R 0105 PATCH
R 0105 PATCH
R 0110 PATCH
R 0115 PATCH
R 0116 PATCH
R 0116 PATCH
R 0122 PATCH
R 0129 PATCH
R 0140 PATCH
R 0142 PATCH
R 0148 PATCH
R 0150 PATCH
R 0172 PATCH
R 0173 PATCH
R 0173 PATCH
R 0173 PATCH
R 0197 PATCH
R 0198 PATCH
R 0198 PATCH
R 0198 PATCH
R 0198 PATCH
R 0198 PATCH
R 0201 PATCH

```

SEGMENT 7 IS 221 LONG

SUBROUTINE MAXMIN (A,N,RMAX,RMIN)

```

DIMENSION A(N)
RMAX=-1E40
RMIN=1E40
DO 100 J=1,N
    IF(A(J).GT.RMAX)RMAX=A(J)
    IF(A(J).LT.RMIN)RMIN=A(J)
100  CONTINUE
RMAX = IFIX(RMAX+1)
RMIN = IFIX(RMIN-1)
RETURN
END

```

```

00500000 R 0220 PATCH
START OF SEGMENT ***** 8
R 0000 PATCH
R 0000 PATCH
R 0002 PATCH
R 0005 PATCH
R 0010 PATCH
R 0015 PATCH
R 0020 PATCH
R 0021 PATCH
R 0023 PATCH
R 0025 PATCH
R 0028 PATCH

```

SEGMENT 8 IS 39 LONG

```

99999999 R 0038 PATCH
SEGMENT 9 IS 80 LONG
SEGMENT 10 IS 29 LONG
SEGMENT 11 IS 138 LONG
START OF SEGMENT ***** 12
SEGMENT 12 IS 16 LONG

```

MLINEAR/PSPHANS

```

& SET TAPE LISTP
FILE 2 = INPUT          UNIT = READER
FILE 3 = OUTPUT        UNIT = PRINTER
FILE 1=LINEAR/REGRESS, UNIT=DISK, RANDOM, AREA=1001, RECORD=30
C -----
C MULTIPLE LINEAR REGRESSION - INPUT PHASE
C -----
                                START OF SEGMENT ***** 1
COMMON I1, R1(30,30), R2(30), R3(30), R4
COMMON NPLOT, QUADRT(6,7), RSUBO
COMMON COEFF(30), NSQU, NDEP, DEP(20), DPGRPH(20), SQGRAF(50,100)
DIMENSION X(60,30), NAME(10), SQUVN(10)
READ (2,100) NUBS, NVAR, NSQU, NDEP
100 FORMAT (A15)
READ (2,150) NAME
150 FORMAT (10A6)
READ (2,200) (SQUVN(I), I=1, NSQU)
200 FORMAT (6I3)
READ (2,110) (DEP(I), I=1, NDEP)
110 FORMAT (20F3.0)
READ (2,115) (DPGRPH(I), I=1, NDEP)
115 FORMAT(20A1)
WRITE(3,953)
953 FORMAT(10X, "PERSONALITY" CORRELATES TO PERFORMANCE UNDER STRESS", 10
-x, "H. P. LEHMANN", //)
WRITE (3,958)
958 FORMAT (10X, "ORIGINAL DATA MATRIX", //)
WRITE(3,954)
954 FORMAT (3X, "UBS, ND, ", "PF/A", 1X, "PF/B", 1X, "PF/C", 1X, "PF/E", 1X, "PF/F",
-, 1X, "PF/G", 1X, "PF/H", 1X, "PF/I", 1X, "PF/L", 1X, "PF/M", 1X, "PF/N", 1X, "PF/
-F/O", 1X, "P/Q1", 1X, "P/Q2", 1X, "P/Q3", 1X, "P/Q4", 1X, "STATE", "TRAIT", 1X
-, "INT.", 1X, "AGE", 2X, "EXP.", 1X, "SCORE")
GO 10 K=1, NUBS
                                00014000 R 0103 PATCH
READ (2,250) (X(K,J), J=1, NVAR)
                                00015000 R 0108 PATCH
955 FORMAT (110, 19F5.0, 10X, 1F5.0)
WRITE (3,955) K, (X(K,J), J=1, NVAR)
                                R 0130 PATCH
250 FORMAT (19F3.0, 6X, 1F3.0)
DEPVAR = X(K, NVAR)
                                00016000 R 0153 PATCH
I=1
                                00017000 R 0153 PATCH
DO 20 J = 1, NVAR
                                00018000 R 0158 PATCH
IF (J.NE.SQUVN(I)) GO TO 20
                                00019000 R 0158 PATCH
X(K,I) = X(K,J)
                                00020000 R 0164 PATCH
X(K,1+NSQU) = X(K,J) **2
                                00021000 R 0167 PATCH
I=I+1
                                00022000 R 0177 PATCH
20 CONTINUE
                                R 0188 PATCH
J = 2*NSQU
                                00023000 R 0190 PATCH
DO 30 M=1, NSQU =1
                                00024000 R 0190 PATCH
DO 40 N = M+1, NSQU
                                00025000 R 0192 PATCH
J = J + 1
                                R 0207 PATCH
X(K,J) = X(K,M) * X(K,N)
                                R 0209 PATCH
40 CONTINUE
                                R 0223 PATCH
30 CONTINUE
                                R 0224 PATCH
X(K,J+1) = DEPVAR
                                R 0225 PATCH
10 CONTINUE
                                R 0231 PATCH
NVAR = J + 1
                                R 0232 PATCH
WRITE (1"0) NUBS, NVAR, NAME
                                R 0233 PATCH

```

APP. E.3.
FORTRAN PROGRAM
USED FOR THE R5A'S

```

DO 45 J = 1,NOBS
45 WRITE (1,J) (X(J,I),I=1,NVAR)
C----- NOW CALL THE SUBROUTINE THAT DOES ALL THE LINEAR REGRESSING
CALL REGR1
C===== READ THE "F VALUES" FOR VARIABLES (1) TO ENTER, & (2) TO LEAVE
C THE REGRESSION EQUATION, AND AN OPTION TO PRINT THE RESIDUALS,
READ (2,300,END=50)FIN,FOUT,IRES,NPLOT
300 FORMAT(2F10.5,2I2)
WRITE (3,3001)
3001 FORMAT ( 10X," MULTIPLE QUADRATIC REGRESSION")
WRITE (3,3002)
3002 FORMAT (10X,"-----")
CALL REGR2 ( FIN, FOUT, IRES )
STOP
50 CALL REGR2 ( 0.0, 0.0, 1 )
C----- NOW START PLUTTING
WRITE(3,9000) (COEFF(I),I=1,NVAR)
9000 FORMAT(F7.4,///)
MEIN=2*NSQU
DO 61 J=1,NSQU
QUADRT(J,J)=COEFF(J+NSQU)
DO 62 K=J,NSQU-1
MEIN = MEIN+1
QUADRT (J,K+1)=COEFF(MEIN)
62 CONTINUE
61 CONTINUE
DO 7000 J=U,NSQU
WRITE (3,9001)(QUADRT(J,I),I=1,NSQU)
7000 CONTINUE
9001 FORMAT (F7.4)
C----- DRAW ONE GRAPH FOR EACH COMBINATION OF VARIABLES
DO 63 M=1,NSQU-1
DO 64 N=M+1,NSQU
C----- NOW INITIATE FOR THE GRAPHS
DO 65 I=1,50
DO 66 J=1,100
SQGRAF(1,J)=" "
SQGRAF(1,J)=" "
SQGRAF(50,J)=" "
SQGRAF(1,1)=" "
SQGRAF(1,100)=" "
66 CONTINUE
65 CONTINUE
C----- DRAW ALL THE SELECTED RESPONSE LEVELS IN ONE GRAPH
DO 67 I=1,NDEP
C----- ONE LOOP FOR EACH GRAPH
DO 68 K = 1,100
X2 = FLOAY(K/10)
C----- FIND THE ROOTS; DEP,X2 IS GIVEN, FIND X1
C=(BSUBO-DEP(I)+COEFF(N)*X2+QUADRT(N,N)*(X2**2))/QUADRT(M,N)
B=(COEFF(M)+QUADRT(M,N)*X2)/QUADRT(M,H)
DISCR=B**2-4*C
C----- CHECK FOR COMPLEX ROOTS
IF (DISCR.LT.0) GO TO 68
TERM=-B/2
COMP=SQRT(DI-GR)/2
C----- CHECK IF THE ROOTS ARE OUTSIDE THE GRAPH BOUNDARIES
DO 69 Z=0.,2.,2.
YPOS=IFIX((TERM+COMP*(Z-1.))*5.+0.5)
IF ((YPOS.LT.0).OR.(YPOS.GE.50)) GO TO 69
C----- FILL GRAPH ARRAY (SQGRAF) WITH RESPONSE=SYMBOLS (DPGRPH)
SQGRAF(YPOS,K)=DPGRPH(I)
69 CONTINUE

```

```

R 0248 PATCH
R 0253 PATCH
00027000 T 0271
00028000 T 0275
00029000 T 0275
00030000 T 0275
00031000 R 0276 PATCH
00032000 R 0293 PATCH
R 0294 PATCH
R 0297 PATCH
R 0298 PATCH
R 0301 PATCH
00033000 T 0302
00034000 T 0303
00035000 T 0305
00035100 R 0306 PATCH
R 0306 PATCH
R 0327 PATCH
R 0327 PATCH
R 0329 PATCH
R 0335 PATCH
R 0345 PATCH
R 0353 PATCH
R 0355 PATCH
R 0364 PATCH
R 0364 PATCH
R 0365 PATCH
R 0371 PATCH
R 0394 PATCH
R 0394 PATCH
R 0395 PATCH
R 0395 PATCH
R 0402 PATCH
R 0404 PATCH
R 0409 PATCH
R 0415 PATCH
R 0421 PATCH
R 0430 PATCH
R 0438 PATCH
R 0446 PATCH
R 0454 PATCH
R 0462 PATCH
R 0463 PATCH
R 0464 PATCH
R 0464 PATCH
R 0465 PATCH
R 0470 PATCH
R 0476 PATCH
R 0477 PATCH
R 0478 PATCH
R 0500 PATCH
R 0514 PATCH
R 0515 PATCH
R 0516 PATCH
R 0519 PATCH
R 0520 PATCH
R 0521 PATCH
R 0522 PATCH
R 0527 PATCH
R 0533 PATCH
R 0535 PATCH
R 0536 PATCH
R 0548 PATCH

```

```

68 CONTINUE
67 CONTINUE
  WRITE (3,3000)
3000 FORMAT("1")
C-----PRINT THE GRAPH
  DO 70 K=1,50
  KK=50-K+1
  WRITE (3,750) (S0GRAF(KK,L),L=1,100)
750 FORMAT(15X,100A1)
70 CONTINUE
C-----WRITE THE GRAPH-COMMENT
  WRITE (3,751) M
751 FORMAT(15X,"RESPONSE SURFACE OF VARIABLE "I2" ON THE Y-AXIS")

  WRITE(3,752) N
752 FORMAT(31X,"AND VARIABLE "I2" ON THE X-AXIS"/)
  WRITE (3,753)
753 FORMAT(15X,"KEY TO CONTOUR HEIGHTS (RESPONSE LEVELS)I")
  WRITE (3,754)(DPGRPH(I),I=1,NDEP)
754 FORMAT (15X,"SYMBOLS",20A5)
  WRITE (3,755)(DEP(I),I=1,NDEP)
755 FORMAT (15X,"RESP.L.I",20I5)
64 CONTINUE
63 CONTINUE
  STOP
  END

```

```

R 0549 PATCH
R 0550 PATCH
R 0551 PATCH
R 0555 PATCH
R 0555 PATCH
R 0555 PATCH
R 0560 PATCH
R 0562 PATCH
R 0585 PATCH
R 0585 PATCH
R 0585 PATCH
R 0585 PATCH
R 0595 PATCH
SEGMENT 2 IS 125 LONG
R 0596 PATCH
R 0605 PATCH
R 0606 PATCH
R 0609 PATCH
R 0610 PATCH
R 0632 PATCH
R 0632 PATCH
R 0654 PATCH
R 0654 PATCH
R 0654 PATCH
R 0654 PATCH
00036000 T 0655
00037000 T 0657
00038000 T 0657
00039000 T 0657
SEGMENT 1 IS 674 LONG

```

C
C

```

SUBROUTINE REGR1
C--- MULTIPLE LINEAR REGRESSION - PHASE 1
COMMON NVAR, R(30,30), XBAR(30), SIGMA(30), OBS
DIMENSION X(30), NAME(10)
C
  READ (1*0) NOBS, NVAR, NAME
  OBS=NOBS
  DO 2 J=1,NVAR
  XBAR(J)=0.
  DO 2 K=1,NVAR
  R(J,K)=0.
2  R(J,K)=0.
  DO 3 I=1,NOBS
  READ (1*I) X
  DO 4 J=1,NVAR
  XBAR(J)=XBAR(J)+X(J)
  DO 4 K=1,J
  R(J,K)=R(J,K)+X(J)*X(K)
4  R(K,J)=R(J,K)
3  CONTINUE
  DO 5 J=1,NVAR
5  SIGMA(J)=SQRT(R(J,J)-XBAR(J)*XBAR(J)/OBS)
  DO 6 J=1,NVAR
  DO 6 K=1,J
  R(J,K)=(R(J,K)-XBAR(J)*XBAR(K)/OBS)/(SIGMA(J)*SIGMA(K))
6  R(K,J)=R(J,K)

```

```

START OF SEGMENT ***** 3
00040000 T 0000
00041000 T 0000
00042000 T 0000
00043000 T 0000
00044000 T 0000
00045000 T 0000
00046000 T 0017
00047000 T 0019
00048000 T 0025
00049000 T 0030
00050000 T 0037
00051000 T 0043
00052000 T 0048
00053000 T 0061
00054000 T 0067
00055000 T 0076
00056000 T 0081
00057000 T 0094
00058000 T 0105
00059000 T 0105
00060000 T 0112
00061000 T 0130
00062000 T 0136
00063000 T 0142
00064000 T 0169

```

156.

E.S./3

```

DEN=SQRT(OBS-1.0)
DO 7 J=1,NVAR
XBAR(J)=XBAR(J)/OBS
7 SIGMA(J)=SIGMA(J)/DEN
WRITE (3,3000)
3000 FORMAT ("1")
WRITE(3,200)NAME
200 FORMAT ( 11X, 10A6 // )
WRITE(3,201)
201 FORMAT(1H0" VAR MEAN STD DEV"/)
WRITE(3,202)(J,XBAR(J),SIGMA(J),J=1,NVAR)
202 FORMAT(1X,I4,2F10.2)
WRITE(3,203)
203 FORMAT(1H0,10X,"SIMPLE CORRELATION CO-EFFICIENTS")
DO 10 J=1,NVAR
10 WRITE(3,204)(J,K,R(J,K),K=1,NVAR)
204 FORMAT(6(2I4,F8.4,4X))
C CALL LINK(REUR2)
RETURN
END
C
C
C
C
C
C
C

```

```

00065000 T 0179
00066000 T 0182
00067000 T 0186
00068000 T 0198
00068100 R 0206 PATCH
R 0210 PATCH
00069000 T 0210
00070000 T 0223
00071000 T 0223
00072000 T 0227
00073000 T 0227
00074000 T 0256
00075000 T 0256
00076000 T 0260
00077000 T 0260
00078000 T 0267
00079000 T 0294
00080000 T 0294
00081000 T 0294
00082000 T 0297
00083000 T 0297
00084000 T 0297
00085000 T 0297
00086000 T 0297
00087000 T 0297
00088000 T 0297
00089000 T 0297
SEGMENT 3 IS 307 LONG

```

```

SUBROUTINE REGR2 ( FIN, FOUT, IRES )
C--- MULTIPLE LINEAR REGRESSION - PHASE 2
COMMON NVAR, RIJ(30,30), XBAR(30), SIGMA(30), OBS
COMMON NPLUT, COEFF(30)
DIMENSION SIGB(30), B(30), ID(30), DATA(30)
DIMENSION RPLUT(1000)
C
NINDV=NVAR-1
NOBS=OBS
C PHASE 2, PERFORM STEPWISE CALCULATIONS AND PRINT RESULTS.
C DIMENSIONS
C INITIALIZE
DO 190 I=1,NVAR
SIGB(I)=0.0
190 B(I)=0.0
NENT=0
DF=OBS-1.0
NSTEP=-1
C TRANSFORM SIGMA VECTOR FROM STANDARD DEVIATIONS TO SQUARE.
C ROOTS OF SUMS OF SQUARES.
DO 310 I=1,NVAR
310 SIGMA(I)=SIGMA(I)*SQRT(OBS-1.0)
C
C BEGIN STEP NUMBER NSTEP.
200 NSTEP=NSTEP+1
WRITE (3,3000)

```

```

START OF SEGMENT ***** 4
00090000 T 0000
00091000 T 0000
00092000 T 0000
00092100 R 0000 PATCH
00093000 T 0000
00093100 R 0000 PATCH
00094000 T 0000
00095000 T 0000
00096000 T 0002
00097000 T 0002
00098000 T 0002
00099000 T 0002
00100000 T 0004
00101000 T 0010
00102000 T 0013
00103000 T 0015
00104000 T 0016
00105000 T 0019
00106000 T 0019
00107000 T 0019
00108000 T 0020
00109000 T 0028
00110000 T 0039
00111000 T 0039
00112000 T 0041
00112100 R 0042 PATCH

```

3000	FORMAT('1')	R	0045	PATCH
	STDEE=SQRT(RIJ(NVAR,NVAR)/DF) * SIGMA(NVAR)	00113000	T	0046
	DF=DF-1.0	00114000	T	0050
	IF(DF)1010,1010,205	00115000	T	0061
205	VMIN=0.0	00116000	T	0064
	VMAX=0.0	00117000	T	0064
	NIN=0	00118000	T	0065
C	FIND MINIMUM VARIANCE CONTRIBUTION OF VARIABLES IN REGRESSION	00119000	T	0065
C	EQUATION, FIND MAXIMUM VARIANCE CONTRIBUTION OF VARIABLES	00120000	T	0065
C	NUT IN REGRESSION EQUATION.	00121000	T	0065
	DO 300 I=1,NINDV	00122000	T	0066
	TF(RIJ(I,I))=.001)300,300,210	00123000	T	0071
210	VI=RIJ(I,NVAR)*RIJ(NVAR,I)/RIJ(I,I)	00124000	T	0081
	TF(VI)240,300,220	00125000	T	0097
220	IF(VI-VMAX)300,300,230	00126000	T	0103
230	VMAX=VI	00127000	T	0106
	NMAX=I	00128000	T	0106
	GO TO 300	00129000	T	0107
240	NIN=NIN+1	00130000	T	0109
	ID(NIN)=I	00131000	T	0110
C	COMPUTE REGRESSION COEFFICIENT AND ITS STANDARD DEVIATION.	00132000	T	0111
	B(NIN)=RIJ(I,NVAR)*SIGMA(NVAR)/SIGMA(I)	00133000	T	0112
	SIGB(NIN)=STDEE*SQRT(RIJ(I,I))/SIGMA(I)	00134000	T	0127
	IF(VMIN)250,260,1000	00135000	T	0138
250	IF(VI-VMIN)300,300,260	00136000	T	0143
260	VMIN=VI	00137000	T	0146
	NMIN=I	00138000	T	0146
300	CONTINUE	00139000	T	0148
	IF(NIN)1000,460,400	00140000	T	0148
C	COMPUTE CONSTANT TERM.	00141000	T	0148
400	RSUBD=XBAR(NVAR)	00142000	T	0154
	DO 410 I=1,NIN	00143000	T	0159
	J=ID(I)	00144000	T	0164
410	RSUBD=RSUBD-B(I)*XBAR(J)	00145000	T	0167
	IF(NENT)1000,480,420	00146000	T	0173
C	OUTPUT FOR VARIABLE ADDED	00147000	T	0173
420	WRITE(3,57)N,STEP,K	00148000	T	0178
57	FORMAT('0STEP NUMBER ',I2,10X,'ENTER VARIABLE ',I2)	00149000	T	0190
425	WRITE(3,58)STDEE	00150000	T	0190
58	FORMAT(' STANDARD ERROR OF ESTIMATE=',F11,3)	00151000	T	0200
	R=SQRT(1.0-RIJ(NVAR,NVAR))	00152000	T	0200
	WRITE(3,59)R	00153000	T	0209
59	FORMAT(' MULTIPLE CORRELATION COEFFICIENT ',F6,3)	00154000	T	0220
	IDFN=OBS-DF-2.0	00155000	T	0220
	IDFD=DF+1.0	00156000	T	0225
C	-----	00157000	R	0225
C	EXPRESS GOODNESS OF FIT AS PERCENTAGE FIT, GIVING THE RATIO OF	R	0225	PATCH
C	SUMS OF SQUARES DUE TO REGRESSION VERSUS THE TOTAL SUM OF SQUARES	R	0225	PATCH
C	-----	00158000	R	0225
	FIT=(R**2)*100	R	0229	PATCH
	WRITE(3,66)FIT	00159000	R	0231
66	FORMAT(1X,'GOODNESS OF FIT:',F8,4,'%')	00160000	R	0240
	WRITE(3,60)RSUBD	00161000	T	0241
60	FORMAT(' CONSTANT TERM=',F12,4)	00162000	T	0251
	WRITE(3,61)	00163000	T	0251
61	FORMAT(3X,'VAR',5X,'COEFF',5X,'STD DEV',4X,'RATIO',1X,'BETA COE	00164000	R	0254
	'FF')	R	0255	PATCH
		SFGHENT	5	15 117 LONG
	WRITE(3,62)	00165000	T	0255
62	FORMAT(23X,'COEFF',2X,'VAR, CONTR')	00166000	R	0258
	DO 430 I=1,NIN	00167000	T	0259
	J=ID(I)	00168000	T	0264
	T=B(I)/SIGB(I)	00169000	T	0266

```

C-----
C INSTEAD OF USING THE RAW T-VALUE USE T T*SQ AS F-RATIO FOR THE
C SIGNIFICANCE OF THE CONTRIBUTION OF VARIABLES TO THE OVERALL FIT
C-----
T=T**2
430 WRITE(3,63)ID(I),B(I),SIGB(I),T,RIJ(J,NVAR)
63 FORMAT (3X,I3,4E11.4)
COEFF(J)=B(I)
C COMPUTE F LEVEL FOR MINIMUM VARIANCE CONTRIBUTION VARIABLE
C IN REGRESSION EQUATION.
FLEV=VMIN*DF/RIJ(NVAR,NVAR)
IF(FOUT+FLEV)460,460,450
C INITIALIZE FOR REMOVAL OF VARIABLE K FROM EQUATION.
450 K=NMN
NENT=0
DF=DF+2.0
GO TO 500
C COMPUTE F LEVEL FOR MAXIMUM VARIANCE CONTRIBUTION VARIABLE
C NOT IN EQUATION.
460 FLEV=VMAX*DF/(RIJ(NVAR,NVAR)-VMAX)
IF(FLEV=FIN)600,600,470
C INITIALIZE FOR ENTRY OF VARIABLE K INTO EQUATION.
470 K=NMAX
NENT=K
GO TO 500
C OUTPUT FOR VARIABLE DELETED
480 WRITE(3,64)NSTEP,K
64 FORMAT("0STEP NUMBER ",I2,10X,"DELETE VARIABLE ",I2)
GO TO 425
C
C UPDATE MATRIX
500 STDEE=1.0/RIJ(K,K)
DO 540 I=1,NVAR
IF(I=K)510,540,510
510 DO 530 J=1,NVAR
IF(J=K)520,530,520
520 RIJ(I,J)=RIJ(I,J)-RIJ(I,K)*RIJ(K,J)*STDEE
530 CONTINUE
540 CONTINUE
DO 560 J=1,NVAR
IF(J=K)550,560,550
550 RIJ(K,J)=RIJ(K,J)*STDEE
560 CONTINUE
DO 580 I=1,NVAR
IF(I=K)570,580,570
570 RIJ(I,K)=-RIJ(I,K)*STDEE
580 CONTINUE
RIJ(K,K)=STDEE
GO TO 200
C
600 IF(IRES)610,640,610
C PRINT RESIDUALS
610 IFA=2
WRITE (3,3000)
WRITE(3,67)
67 FFORMAT("0 OBS ACTUAL ESTIMATE RESIDUAL")
DO 630 K=1,NUBS
READ (1,K) (DATA(I), I=1,NVAR)
EST=BSUB0
DO 620 I=1,NIN
J=ID(I)
620 FST=EST+B(I)*DATA(J)
RESID=DATA(NVAR)-EST

```

```

00169050 R 0267 PATCH
00169100 R 0267 PATCH
R 0267 PATCH
R 0267 PATCH
R 0269 PATCH
00170000 T 0271
00171000 R 0298 PATCH
00171100 R 0298 PATCH
00172000 T 0301
00173000 T 0301
00174000 T 0303
00175000 T 0311
00176000 T 0311
00177000 T 0314
00178000 T 0314
00179000 T 0315
00180000 T 0318
00181000 T 0318
00182000 T 0318
00183000 T 0320
00184000 T 0328
00185000 T 0329
00186000 T 0332
00187000 T 0332
00188000 T 0333
00189000 T 0333
00190000 T 0335
00191000 T 0347
00192000 T 0347
00193000 T 0347
00194000 T 0347
00195000 T 0348
00196000 T 0355
00197000 T 0361
00198000 T 0365
00199000 T 0371
00200000 T 0375
00201000 T 0395
00202000 T 0398
00203000 T 0396
00204000 T 0402
00205000 T 0406
00206000 T 0416
00207000 T 0416
00208000 T 0422
00209000 T 0426
00210000 T 0437
00211000 T 0437
00212000 T 0442
00213000 T 0442
00214000 T 0444
00215000 T 0444
00216000 T 0447
00216100 R 0447 PATCH
00217000 T 0451
00218000 T 0455
00219000 T 0455
00220000 T 0460
00221000 T 0479
00222000 T 0479
00223000 T 0485
00224000 T 0488
00225000 T 0491

```



```

RPLLOT (K)=RESID
WRITE(3,68)K*DATA(NVAR),EST,RESID
68 FORMAT(" *I4*3F12.2)
630 CONTINUE
IF (NPLLOT.EQ.1) CALL PLOT(RPLLOT,NVAR,NOBS)
C--- NORMAL END OF PROGRAM
640 RETURN
1000 CALL ERROR(1)
1010 CALL ERROR(2)
1020 CALL ERROR(3)
      END

```

```

00225100 R 0494 PATCH
00226000 T 0496
00227000 T 0515
00228000 T 0515
00228100 R 0515 PATCH
00229000 T 0520
00230000 R 0521 PATCH
00231000 T 0524
00232000 T 0525
00233000 T 0526
00234000 T 0526
SEGMENT 4 IS 552 LONG

```

```

-----
                                START OF SEGMENT ***** 6
SUBROUTINE ERROR( I )
GO TO (10, 20, 30), I
10 WRITE (3,100)
100 FORMAT ("OERROR... NIN, NENT, VMIN, NCONS, OR NTRAN IS NEGATIVE.
*CHECK FOR ERROR ON PARAMETER CARDS" )
STOP
20 WRITE (3,200)
200 FORMAT ("OERROR... DEGREES OF FREEDOM = 0. EITHER ADD MORE OBSERV
*ATIONS OR DELETE ONE OR MORE INDEPENDANT VARIABLE." / " SAMPLE SIZE
*E MUST EXCEED NUMBER OF INDEPENDANT VARIABLES BY AT LEAST 2." )
STOP
30 WRITE (3,300)
300 FORMAT ("OERROR... F LEVEL FOR INCOMING VARIABLE IS LESS THAN F LE
*VEL FOR OUTGOING VARIABLE." )
STOP
END
                                00235000 T 0000
                                00236000 T 0000
                                00237000 T 0007
                                00238000 T 0011
                                00239000 T 0011
                                00240000 T 0011
                                00241000 T 0013
                                00242000 T 0017
                                00243000 T 0017
                                00244000 T 0017
                                00245000 T 0017
                                00246000 T 0019
                                00247000 T 0023
                                00248000 T 0023
                                00249000 T 0023
                                00250000 T 0024
SEGMENT 6 IS 30 LONG

```

```

-----
SUBROUTINE PLOT(Y,NVAR,NOBS)
DIMENSION Y(NOBS)*X(1000),DUMMY(30),GRAPH(50*100)
NX=80
NY=20
C-----CALCULATE THE MAXIMUM AND MINIMUM OF ORDINATE Y
CALL MAXMIN(Y,NOBS,YMAX,YMIN)
YFACT=(YMAX-YMIN)/(NY-1)
C-----ONE LOOP FOR EACH GRAPH
DO 500 J=1,NVAR
WRITE(3,3000)
3000 FURHAT("1")
C-----READ IN VARIABLE J
DO 50 I=1,NOBS
READ (1,"I)(DUMMY(L),L=1,NVAR)
X(I)=DUMMY(J)
50 CONTINUE
                                00300000 R 0029 PATCH
START OF SEGMENT ***** 7
                                R 0000 PATCH
                                R 0000 PATCH
                                R 0000 PATCH
                                R 0000 PATCH
                                R 0001 PATCH
                                R 0003 PATCH
                                R 0005 PATCH
                                R 0006 PATCH
                                R 0011 PATCH
                                R 0015 PATCH
                                R 0016 PATCH
                                R 0016 PATCH
                                R 0021 PATCH
                                R 0039 PATCH
                                R 0041 PATCH

```

160.

E.3./7

```

C-----NOW FIND THE MAXIMUM AND MINIMUM OF X
CALL MAXMIN(X,NOBS,XMAX,XMIN)
XFACT=(XMAX-XMIN)/(NX-1)
C----- INITIALISE GRAPH
DO 40 I=1,NX+1
  DO 40 K=1,NY+1
40  GRAPH(K,I)=0
C----- PLOT POINTS
DO 100 I=1,NOBS
  NXPOS=FIX((X(I)-XMIN)/XFACT+0.5)+1
  NYPOS=FIX((Y(I)-YMIN)/YFACT+0.5)+1
  GRAPH(NYPOS,NXPOS)=GRAPH(NYPOS,NXPOS)+1
100  CONTINUE
C----- PUT IN X AND Y AXES
DO 60 I=1,NY
  GRAPH(I,1)=13
60  CONTINUE
DO 70 I=1,NX
  GRAPH(1,I)=44
70  CONTINUE
C----- NOW PRINT THE GRAPH
DO 120 II=1,NY+1
  DO 120 JJ = 1,NX+1
    IF(GRAPH(II,JJ).EQ.0)GRAPH(II,JJ)=48
120  CONTINUE
DO 200 I=1,NY+1
  II=NY-I+2
  WRITE(3,1000)(GRAPH(II,K),K=1,NX+1)
1000 FORMAT(1X,130C1)
200  CONTINUE
C----- PRINT OUT GRAPH DATA
WRITE(3,2000)J,XMIN,XMAX,XFACT,YMIN,YMAX,YFACT,NOBS
2000  FORMAT(" GRAPH OF VARIABLE",I4," VS. RESIDUALS "//
  " X-AXIS BOUNDS ARE",F12.6,3X,F12.6," X INCREMENT =",F12.6/
  " Y-AXIS BOUNDS ARE",F12.6,3X,F12.6," Y INCREMENT =",F12.6/
  " NO. OF OBSERVATIONS = ",I4)
500  CONTINUE
RETURN
END

```

```

R 0042 PATCH
R 0042 PATCH
R 0044 PATCH
R 0046 PATCH
R 0047 PATCH
R 0053 PATCH
R 0060 PATCH
R 0065 PATCH
R 0067 PATCH
R 0072 PATCH
R 0078 PATCH
R 0084 PATCH
R 0094 PATCH
R 0095 PATCH
R 0095 PATCH
R 0100 PATCH
R 0105 PATCH
R 0105 PATCH
R 0110 PATCH
R 0115 PATCH
R 0116 PATCH
R 0116 PATCH
R 0122 PATCH
R 0129 PATCH
R 0140 PATCH
R 0142 PATCH
R 0148 PATCH
R 0150 PATCH
R 0172 PATCH
R 0173 PATCH
R 0173 PATCH
R 0173 PATCH
R 0197 PATCH
R 0198 PATCH
R 0198 PATCH
R 0198 PATCH
SEGMENT 8 IS 110 LONG
R 0198 PATCH
R 0198 PATCH
R 0201 PATCH
SEGMENT 7 IS 221 LONG

```

SUBROUTINE MAXMIN (A,N,RMAX,RMIN)

```

DIMENSION A(N)
RMAX=-1E40
RMIN=1E40
DO 100 J=1,N
  IF(A(J).GT.RMAX)RMAX=A(J)
  IF(A(J).LT.RMIN)RMIN=A(J)
100  CONTINUE
RMAX = IFIX(RMAX+1)
RMIN = IFIX(RMIN-1)
RETURN
END

```

```

00500000 R 0220 PATCH
START OF SEGMENT ***** 9
R 0000 PATCH
R 0000 PATCH
R 0002 PATCH
R 0005 PATCH
R 0010 PATCH
R 0015 PATCH
R 0020 PATCH
R 0021 PATCH
R 0023 PATCH
R 0025 PATCH
R 0028 PATCH
SEGMENT 9 IS 39 LONG

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Personality Variables

	<u>Var.</u>
A - RESERVED/OUTGOING	(
B - SCHOLASTIC MENTAL CAPACITY	(
C - EGO STRENGTH	(
E - DOMINANCE	(
F - SOBER/IMPULSIVE	(
G - SUPEREGO STRENGTH	(
H - SHY/VENTURESOME	(
I - TOUGH/TENDER-MINDED	(
L - TRUSTING/SUSPICIOUS	(
M - PRACTICAL/IMAGINATIVE	(
N - ARTLESS/SHREWD	(
O - CONFIDENT/APPREHENSIVE	(
Q1 - CONSERVATIVE/RADICAL	(
Q2 - SELF-SUFFICIENCY	(
Q3 - SELF-CONCEPT CONTROL	(
Q4 - RELAXED/OVERWROUGHT	(
A-T - TRAIT-ANXIETY	(
A-S _t - STATE-ANXIETY	(
INTELLIGENCE (RAVEN'S PMT)	(