

THE DETERMINATION OF FROZEN GROUND PROBABILITIES
FROM CLIMATIC AND HYDROLOGIC DATA

A Thesis

Presented in Partial Fulfillment of the Requirements for the
DEGREE OF MASTER OF SCIENCE
Major in Agricultural Engineering

in the
UNIVERSITY OF IDAHO GRADUATE SCHOOL
by

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July, 1975

The work reported in this thesis was performed as part of project A-045-IDA of the Idaho Water Resources Research Institute under the allotment program from the Office of Water Research and Technology, US Department of the Interior. The Palouse Conservation Field Station, USDA,ARS helped extensively with collection of data and interpretation of results.

The overall coordination of the project is in the Agricultural Engineering Department under an Idaho Agricultural Experiment Station project entitled Erosion Research for Northern Idaho.

ACKNOWLEDGEMENTS

First, I wish to express my thanks to my Major Professor Myron Molnau for his help and instruction, which guided me to finishing my thesis smoothly. Special thanks are due Professor P. T. Sun and Professor G. L. Bloomsburg for their suggestions and for reviewing my thesis. Especially I wish to thank Professor John S. Gladwell for guidance through the maze of statistics and for reviewing the thesis. Dr. D. K. McCool of the Agricultural Research Service, USDA, Pullman, Washington, provided much guidance and help in the hydrologic significance of the results.

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LIST OF SYMBOLS

- ρ : The air density near the ground surface. g/cm^3 .
 ρ_d : The population autocorrelation coefficient at lag d .
 ρ_i : Canonical correlation or autocorrelation coefficient at lag i .
 δ_{ij} : Normalized eigenvector.
 λ : Dimensionless correction coefficient or latent root.
 Σ : The population of a variance-covariance matrix.
 μ : The mean of the population.
 A : Among group of sum of square and cross products matrix.
 C_p : The specific heat of air at constant pressure. 0.24 cal/g-C .
 D_a : The estimator of among-group variance-covariance matrix.
 D_{ij}^2 : Mahalanobis' distance.
 D_w : The estimator of within-group variance-covariance matrix.
 E : Expectation.
 F : The surface freeze index. Degree days F .
 $F(n_1, n_2)$: F-value with degrees of freedom n_1, n_2 .
 g : Number of groups.
 K : The turbulence coefficient for thermal conductivity. BTU.hr/ft/F .
 K_f : Thermal conductivity of frozen ground. BTU/hr/ft/F .
 L : Latent heat of soil. BTU/ft^3 .
 M : Box's test criterion.
 n_1 : The number of observation in group 1 or degrees of freedom.
 n_2 : The number of observation in group 2 or degrees of freedom.

- N_i : The number of observation in group i .
- $P(H_j | X_i)$: A posteriori probability under the given observation X_i .
- $P(H_j)$: A priori probability in a given group H_j .
- p : The number of variables.
- r_k : The estimator of the autocorrelation coefficient at lag k .
- S_j : A pooled within-group variance-covariance matrix of group j .
- T : The heat exchange between ground and air. $\text{cal/cm}^2 \cdot \text{sec}$.
or total sum of square and cross products matrix.
- Δt : The conventional vertical air temperature difference, C.
- t : The latent heat released as pore water freezes to a depth X in time t , hour.
- U : Wilk's $U = |W|/|T|$, $|W|$: Determinant of "within-group" sum of square and cross products matrix. $|T|$: Determinants of "total" sum of square and cross products matrix.
- Var: Variance.
- V_i : Coefficient vector of discriminant variables.
- V_s : Temperature of soil surface F .
- W : Within-group sum of square and cross products matrix.
- X : The depth of frost penetration. ft.
- \bar{X}_j : The mean of the observations of the variable j .
- X_{mjk} : The value of the observation of m discriminant function, k observation, j variable.

y_{mk} : The m -th discriminant function evaluated for observation k . y_{mk} is the first discriminant function.

z_{mk} : The m -th discriminant function evaluated for observation k . z_{mk} is second discriminant function.

ABSTRACT

The prediction of frozen ground occurrences can help the local planner regulate development in areas which are subject to floods. Discriminant analysis was used in the Missouri Flat Creek case in order to study past frozen or unfrozen ground runoff events using eight meteorological factors. These eight factors are number of freezing days in a sequence, freeze index (sum of degree-days below 32°F), precipitation two days before the start of the freeze period, precipitation during the freeze period, precipitation four days after the freeze period, the depth of snowfall during the freeze period, the depth of snow on the ground before and total stream flow 4 days after.

After a series of analyses, the eight meteorological factors used were determined to be sufficient to classify the runoff events into frozen or unfrozen ground events. The freeze index was found to be the single most important factor for estimating the frozen or unfrozen ground condition in the Missouri Flat Creek Basin. Ten of 89 unknown samples are probably frozen ground and the rest of them probably belong to the unfrozen ground class.

CHAPTER I
INTRODUCTION

GENERAL

Frozen ground is one of the important factors causing winter floods in the Pacific Northwest during the winter and early spring months. On February 26, 1948, a flood in the South Fork Palouse River Basin caused extensive damage in Pullman, especially in the mobile home court located southeast of town. This flood was probably triggered by rainfall on frozen ground. In 1962 and 1964, there were two other very severe floods in Idaho. The USGS and SCS reported that the cause of flooding was prolonged rainfall on snow and frozen soil. Johnson and McArthur (1973), in a study of winter flooding in Idaho and surrounding areas for 1955-1972, found that the amount and intensity of the rainfall, the amount of snowmelt, and the imperviousness of the frozen soil combined to affect the flood severity. Because of extensive damage from each flood, millions of dollars have been spent on flood control every year.

Climatological data has been found to be important in studying frozen ground. McCool and Molnau (1974) used the historical climatological record to study both the frequency and the severity of frozen soil conditions. They concluded that climatological data can be very useful in assessing the overall frozen ground hazard for a given location.

The Missouri Flat Creek drainage basin was selected as a test case. The period of record for this investigation is 1960 to 1973.

OBJECTIVE

The prediction of frozen ground occurrences under certain conditions will help local planners regulate development in areas which are subject to floods. For the purposes of this thesis, frozen ground floods or runoff events are defined as rainfall or snowmelt runoff events which occur when there is a crust of frozen soil covering the watershed to such an extent such that the infiltration rate of the watershed is decreased and the runoff is more than that which would occur under conditions of no frozen soil. Discriminant analysis was chosen to help provide a simple technique which can show the probability of occurrence of frozen ground runoff events in the past for a sequence of meteorological events. The number of freezing days, freezing index, precipitation two days before the start of the freeze period, precipitation during the freeze period, precipitation four days after the freeze period, the depth of snowfall during the freeze period, the depth of snow on ground before and stream flow four days after are chosen as discriminators.

The objective of this study is to:

1. Identify some hydrologic parameters that are important in distinguishing frozen from unfrozen ground runoff events.
2. Identify historical frozen ground runoff events using these parameters.

CHAPTER II

LITERATURE REVIEW

One of the major results of the transformation of water into ice and vice versa in soil is the erosion of the soil. Heavy rainfall over frozen or partially frozen soil, or on snow, often causes drastic soil erosion and winter floods.

Many factors govern whether or not soil is frozen. Bouyoucos (1916) grouped the influencing factors into two categories, (a) intrinsic and (b) external. He considered the intrinsic factors to be those governed by the properties of the soil and its cover, and the external factors to be basically meteorological in nature (Moulton, 1968). Similar classifications have been presented by Pelton, Campbell and Nicholaichuk (1968) and by Linell and Kaplar (1965). Since only meteorological factors are to be considered here, only external factors will be discussed in this literature review.

INFLUENCES OF METEOROLOGICAL FACTORS ON FROZEN GROUND

The meteorological factors affecting the freezing of soil are considered to include the air temperature, solar radiation, wind and precipitation (Moulton, 1968).

The temperature factors exert a direct influence on the transfer of heat to or from the earth. Molga (1962) mentioned that the air layers near the ground are heated during the day by the active surface warmed by solar

rays (termed "active surface" by Molga), the thin (several mm thick) air layer adhering to the active surface, called the boundary layer, is warmed by the surface. On winter nights, its thickness usually does not even reach 1 mm. The vertical temperature gradients are the greatest in the vicinity of the active surface and decline with increasing height, rather rapidly at first and more gently higher above the ground. This is due to the increase in the turbulence coefficient with height since the temperature gradient, Δt , is inversely proportional to the turbulence coefficient, K , according to the formula (Molga, 1962):

$$T = C_p \cdot \rho \cdot K \cdot \Delta t$$

Where: T : The heat exchange between ground and air.

C_p : The specific heat of air under constant pressure, equal to 0.24 cal/g.C.

ρ : The air density near the ground surface, equal to 0.0012-0.0013 g/cm³.

K : The turbulence coefficient.

Δt : The conventional vertical air temperature gradient, °C.

Frankline (1920) found that wind caused the surface temperature to be lower than the prevailing air temperature, and that this effect became more pronounced as the wind velocity increased or as the relative humidity decreased. In other words, high wind of low relative humidity can have a considerable effect upon subsurface temperatures.

Eno (1929) discussed the important influence of climate and the resultant frost action upon the performance of highways in the United States. Sourwine (1929) in a discussion of Eno's paper suggested a method to relate quantitatively air temperature data to the relative danger of ground freezing in any given locality. He defined three controlling variables that could be evaluated using data available from climatological records. These were:

1. The average minimum temperature.
2. The duration of the cold period.
3. The frequency of occurrence of a cold period of given intensity.

Sourwine concluded that an air temperature of 3°F represented a "critical absolute minimum value" coefficient with a 5 percent frequency of occurrence of ground freezing at the 3 inch level. He was able to relate this absolute minimum air temperature to monthly average minimum temperatures and the duration of the cold period to represent the southerly limits of serious danger to highways from ground freezing. This strongly suggests that the combined effect of low temperature and the duration of the cold period on ground freezing can be an essential variable for estimating frozen ground. He also introduced "degree-hours", with the degrees being measured in °F below the critical initial value of air temperature causing freezing at the surface and the duration being measured in hours of cold period below the critical initial value of air temperature.

Sourwine also suggested that some relationship might exist between this "degree-hour index" and the depth of frost penetration. Casagrande (1931) conducted field tests near the Massachusetts Institute of Technology and concluded that the relationship suggested by Sourwine existed. Later Casagrande found a good correlation between cumulative degree-days of below freezing temperature and the depth of frost penetration. Aldrich (1956) used the Stefan formula and the modified Berggren formula to predict the depth of frost penetration. The Stefan formula is:

$$X = \sqrt{48K_f F/L}$$

Where: X: Depth of frost penetration in ft.
 K_f : Thermal conductivity of frozen soil.
 BTU/hr/ft/F.
 F: The surface freezing index, degree days F.
 L; Latent heat of pore water, BTU/ft.³

Since this neglects the volumetric heat of the frozen and unfrozen soil and, hence, overestimates frost penetration, it has been modified by Berggren to :

$$X = \lambda \sqrt{2KVst/L}$$

Where: X: Depth of frost penetration in ft.
 K: Thermal conductivity of frozen or unfrozen soil, BTU/hr/ft/F.
 Vs: Temperature of soil surface °F.
 t: Hours.
 L: Latent heat of pore water, BTU/ft.³
 λ : Dimensionless correction coefficient.

Corrects the Stefan formula for the effects of volumetric heat which it neglected.

Aldrich (1956) also used air temperature in determining the freezing index because data on surface temperatures are almost nonexistent. Both the Stefan and the modified Berggren equations may also be used to predict the depth of thaw of a frozen soil. The "thawing index" used is calculated in degree-days as the mean daily air temperature in excess of 32°F.

Aldrich (1956) described the relationship between air and pavement temperatures and concluded that it was best to use the air freezing index with a correction factor to adjust it to an equivalent pavement surface freezing index. The value of freezing index for a specific location might differ greatly from the general value for the area, and that freezing index might vary widely over a small area (Straub and Wegmann, 1965).

McCool and Molnau (1974) investigated the freezing index as a parameter to predict frozen soil in the Palouse Basin. From an investigation of the 1953 through 1973 data, they found that individual freezing events with freeze index values of less than 100 were generally not associated with large runoff events, and values of 150 and over seldom left any doubt about a frozen ground flood if rain occurred at the end of the freezing event. They also concluded that the freezing index and precipitation combinations can be

used, with some discretion, in the prediction of frozen soil runoff events.

They also examined the Missouri Flat Creek stream-flow data. They found that many high volume runoff events occur independently of frozen soil conditions. During water years 1953-1973 for periods from 1 to 19 days in length and for runoff totalling greater than 0.5 inch, only 7 of 23 events were definitely associated with rain on frozen soil and one more event was probably associated with runoff on frozen soil. The remaining fifteen were caused by rain and rain on snow. Therefore, we know that runoff is not a single process. It is the result of several separate processes.

McArthur (1971) studied flood runoff from rainfall and snowmelt over a frozen soil mantle in Idaho and concluded that winter flooding is caused by:

1. Antecedent moisture and temperature conditions that produce frozen soil relatively impermeable to water infiltration.
2. A sudden warming period with daily minimum temperatures above 32°F.
3. An accompanying storm producing rainfall for a period of three or more days.

He also presented the following table that shows the hydrologic conditions associated with winter flooding in Idaho.

TABLE 1: HYDROLOGICAL CONDITIONS ASSOCIATED WITH WINTER
FLOODING IN IDAHO (McARTHUR, 1971)

Factor	Effect upon flood severity
Excess runoff from	
a. Rainfall	Proportional to total volume
b. Snowmelt	and to the rainfall intensity
c. Both	
Frozen soil	
a. Concrete frost	Increases runoff
b. Porous concrete frost	Can reduce runoff
c. Stalactite frost	Reduces runoff
Snow cover	
a. Deep	Decreases runoff
b. Shallow	Increases runoff
Vegetative cover	
a. Tree, larger type vegetation	Decreases runoff
b. Rangeland, sagebrush and grass	Decreases runoff
c. Cultivated lands	Generally increases runoff
Atmospheric temperature	
Lapse rate	
High	Decreases runoff
Low	Increases runoff (Snowmelt and contributing area)

In Figure 1, the precipitation occurred as snow during January because of the severe cold period. During this period the ground was frozen. A concrete frost was not prevalent because of the antecedent conditions; that is, warmer than usual temperatures and below normal precipitation in November and December. Soil moisture was not enough to produce a concrete, impermeable frost. During January there was a shallow snow cover. Toward the last of January there was a warming period and a storm moved through this area depositing a large volume of rain. So, there was a peak flow probably caused by rainfall on snow.

Merrell (1964) used frequency methods to examine rainfall, snowmelt, and frozen ground conditions in order to compute joint probabilities associated with combinations of these three parameters. He compared joint probabilities with the qualitative record of flooding. As a result, he found that the severity of flooding is not closely related to the joint probability of all three parameters in the Sand Creek watershed. According to the author's study, he did not prove this conclusion is valid for other places.

Turner and Jumikis (1956) found that, when the precipitation was in the form of snow, more of the precipitation ultimately percolated into the ground than if the precipitation was in the form of rain. Snow cover exercises a decisive influence on the frozen ground. The albedo of snow is high, amounting to 75-88% and more on fresh snow, and to approximately 45% on older and moist snow. Thus snow

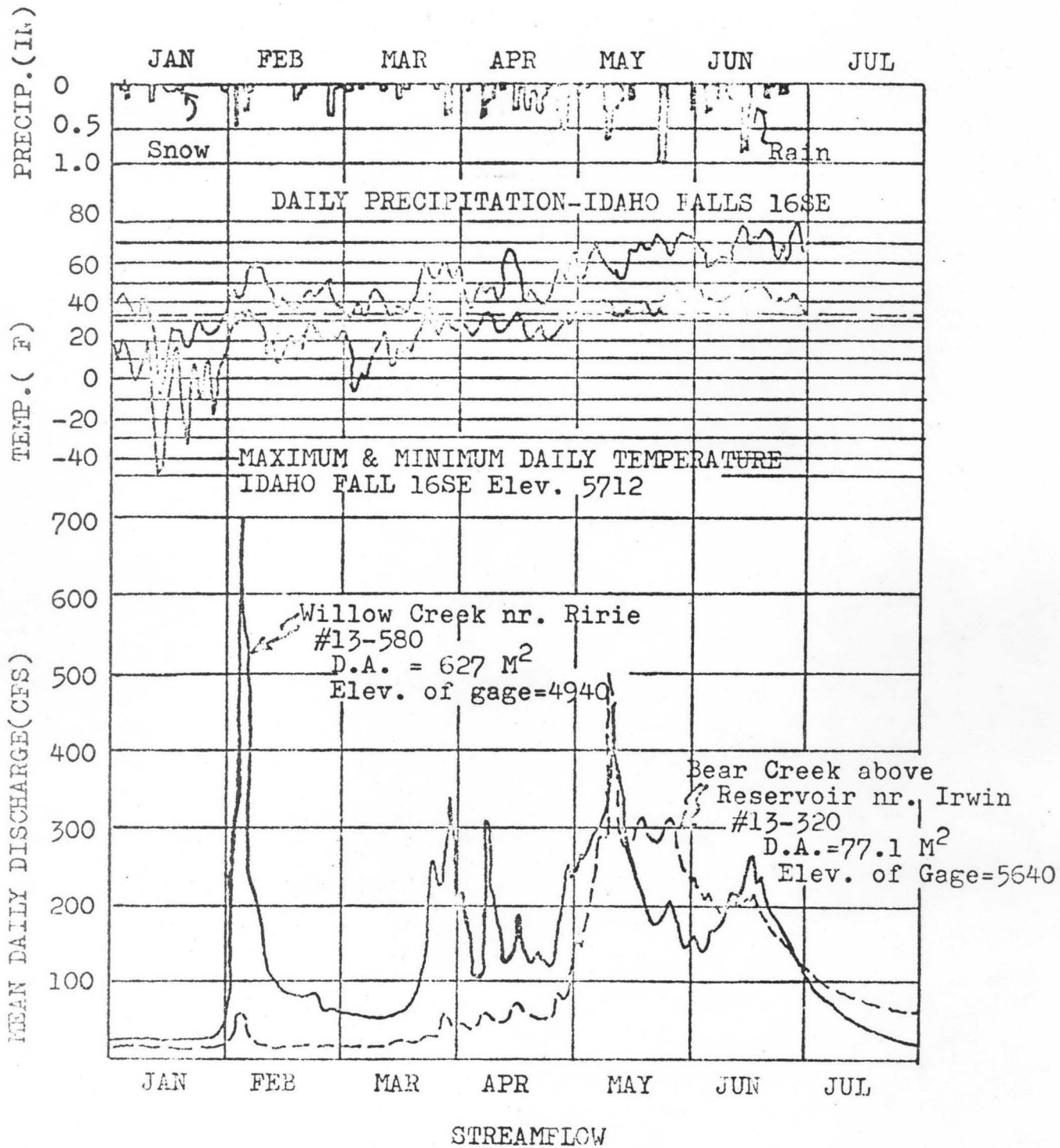


FIGURE 1 Comparison of the runoff hydrograph of Willow Creek with that of Bear Creek

warms up slightly and does not melt quickly under the influence of solar radiation. Even with the high albedo of snow cover, some of the solar rays penetrate its depth and the snow melts from below. If the soil is covered by a snow layer of over 60 cm depth, the soil is completely insulated against the influence of solar radiation (Molga, 1962). The insulating properties of a snow cover also exerts an influence on the runoff in springtime. By protecting the soil from freezing, the snow cover permits water to percolate into the soil and reduces surface runoff. These environmental factors affecting soil freezing can be summarized as follows:

1. Solar radiation: The amount of radiation received by the soil surface depends on:
 - a. The angle with which the soil faces the sun, due to latitude, season, time of day, steepness and direction of slope, and the altitude of the location.
 - b. The insulation of air, water vapor, clouds, dust, smog, snow, plants, or mulch.
2. Radiation from the sky: The sun's energy is absorbed by the atmosphere and is radiated in all directions. Then soil absorbs from sky indirectly.
3. Conduction and convection of heat from the atmosphere: Convection or wind is necessary to affect the soil temperature significantly since

the conductivity of air is very small.

4. Condensation: Condensation is an exothermic process. Freezing of water generates heat.
5. Evaporation: It is an endothermic process. The greater the rate of evaporation, the more the soil is cooled down. Also thawing of ice absorbs heat.
6. Rainfall: Rainfall, depending on its temperature, can cool or warm the soil.
7. Insulation: The soil can be insulated by plant cover, mulch, snow, and also clouds and fog. In winter, insulated soil is warmer than soil that is directly exposed to the environment.

INFLUENCES OF SOIL FACTORS ON FROZEN GROUND

The permeability and storage of frozen soil depend on the type of soil frost. Soil frost can be divided into four types (Trimble, Sartz, and Pierce, 1958):

1. Concrete frost: This type of frost usually appears in bare agricultural land. It is composed of a great many thin ice lenses, with the ground surface being very hard, like concrete.
2. Granular frost: It is always found in woodland soils which contain organic matter. This type of frozen soil contains small frost crystals mixed with the soil particles, but remain separate from each other and are easily broken.

3. Honeycomb frost: It is often found in highly aggregated soils and it has a loose, porous structure which looks like honeycomb.
4. Stalactite frost: It forms in bare saturated ground and consists of loosely fused, columnar ice crystals and absorbs rainfall-snowmelt like a sponge because of its rough and very porous microrelief. As stalactite frost melts, the infiltration rate decreases.

Haupt (1967) reported that concrete frost reduces the infiltration capacity so that overland flow is increased on burned or sparsely vegetated sites if the rainfall intensity exceeds 3 ins./hr. He also found that snow cover, by cooling rain water, tends to preserve soil frost and keep it visibly intact and that cover, such as plants, litter, and snow, can absorb raindrop energy and increase infiltration. But exposed rock usually accelerates overland flow and erosion. Pelton, Campbell and Nicholaichuk (1968) found that infiltration rates of frozen soil vary widely and depend largely upon the soil moisture content at the time of freezing. They found that the trapping of snow on the soil surface and its subsequent melting and infiltration appear to be dominant factors in conserving winter moisture. When moisture accumulates and freezes near the soil surface, thermally induced-moisture translocation is also a factor contributing to the reduction of infiltration rate. Bloomsburg and Wang (1969) in laboratory studies found that

below, then frost depth prediction methods based on an assumed 32°F freezing point will tend to underpredict the depth of the 32°F isotherm. Thermal conductivities of soil-forming minerals, while that of air is about 100 times smaller than that of soil-forming materials (Kohnke, 1968). Kersten (1949) found that the difference between the thermal conductivities of frozen and unfrozen soils was dependent upon their moisture content. As the water content increased the thermal conductivity of the frozen soil became considerably higher than that of the unfrozen soil. Kersten's tests showed that as the quartz content decreased and the clay mineral content increased there was a general decrease in thermal conductivity. He also found that the thermal conductivity decreased with decreasing grain size at a given density and moisture content, and an increase in soil density resulted in an increase in thermal conductivity. Patten (1909) showed that the thermal conductivity of soils increased continuously as the soil moisture content was increased to saturation.

A soil in undisturbed condition has a higher thermal conductivity than when it is recently disturbed. A blocky structure has a higher thermal conductivity than organic soil. But the differences of thermal conductivity due to structure and texture are small compared with those due to moisture changes (Kohnke, 1968).

The ice content of the soil has a very strong effect on meltwater absorption in a basin. The higher the ice

the parameter $\phi(1-s)$ (ϕ is porosity and s is saturation) at the start of freezing is directly related to the permeability after freezing by an exponential relationship. This means that the permeability decreases as $\phi(1-s)$ decreases. Also, they found that the value of $\phi(1-s)$ must be less than about 0.13 in silts and sands for the permeability to be zero when the soil is frozen.

The most significant thermal properties to be considered with respect to the determination of frost depth are (1) volumetric heat, (2) latent heat of fusion, and (3) thermal conductivity. The soil temperature is determined by the interaction of numerous factors. All soil heat comes from two sources; radiation from the sun and sky and conduction from the interior of the earth.

The thermal capacity of soil is determined by the specific heat of mineral soil and solution in soil. The specific heat of water is larger than many substances (1.0 cal/gm). The specific heat of ice is 0.5 (Kohnke, 1968). The magnitude of volumetric heat capacity of the soil exerts a relatively small influence on the depth of frost penetration. Also the latent heat of fusion of a soil depends only upon the amount of water in a unit volume of the soil. Practically, in the prediction of the depth of frost penetration, it is assumed that all the soil water freezes at 32°F and that all the latent heat of fusion is liberated at that temperature. If substantial portions of the soil water do not freeze at a temperature of 32°F and

content in the soil, the narrower these pores and passages are and, the lower the permeability of the soil. Komarov and Makarovo (1973) reported that the main factors which influence the permeability of the soil prior to snow melting are its ice content and depth of freezing. They also found that the spatial variations of these factors, are very large and are an important factor in the infiltration of meltwater in a basin. Borisovskiy (1973) showed that the temperature of drained soils is much lower throughout the winter than that of undrained soils. In his studies, in the warm period, the temperature of undrained soils is 1.5-2.5°C higher than in drained soils. Conversely, in the cold period, drained soils average 1.5-2° colder than undrained soils.

MULTIVARIATE STATISTICAL ANALYSIS

Multivariate analysis is concerned with analyzing multiple measurements that have been made on N individuals and are considered in combination, as systems. So m measurements which have been made on N individuals can be represented as N points in the m-dimensional space. Therefore, in order to analyze the relative importance of various meteorological factors on frozen soil prediction, it was decided to use some type of multivariate analysis.

Multivariate statistical analysis as applied to hydrologic problems is a new field and still in its developmental and trial stage. Wallis (1965) used the

characteristics of the hollow cylinder as an example to discuss and compare the effectiveness of multivariate analyses. He concluded that the different methods of generating prediction equations do not give identical answers, even when the same data are used. Therefore the method that will be most suitable for each specific problem should be selected. He recommended a combination of principal component regression with varimax rotation of the factor weight matrix for an initial analysis of multifactor hydrologic problems. He also introduced cluster analysis in hydrology which can be used if many observations are available.

Hastay and Gladwell (1969) used a combination of least squares regression with principal component analysis to evaluate a cloud-seeding program at the streamflow control level. They found that using principal components has the following advantages.

1. Conserving degrees of freedom in the estimate of residual variance.
2. Computational convenience, especially in making statistical tests to identify significant predictors.
3. Superior conditioning of covariance matrices against the effects of multicollinearity, thus increasing the precision of statistical tests.

Rice (1967) said that multivariate methods free the hydrologist from the need to describe a phenomenon as a single number. So if the hydrologist uses multivariate

methods, he can cope with the highly correlated "independent" variables common in hydrology. Also these methods provide a flexible approach that is more in harmony with nature of many hydrologic problems. Wallis (1967) used a discriminant function for testing the identification of marine versus nonmarine sediments based upon microelements. He reported that the prediction of the behavior of one or more variables for one or more of a new population of watersheds were often found to be associated with large errors. According to Wallis, the errors come from the specification of the original model and unwarranted assumptions concerning the degree of similarity between the original and the new group of watersheds. He also discussed whether a prediction equation developed from the watersheds of group 1 is applicable to all or any of the watersheds of group 2. Before answering this question, the following questions must first be answered:

1. Is there a significant difference between the two groups of watersheds?
2. Is a significant difference in the groups relevant to the specific criterion being predicted?

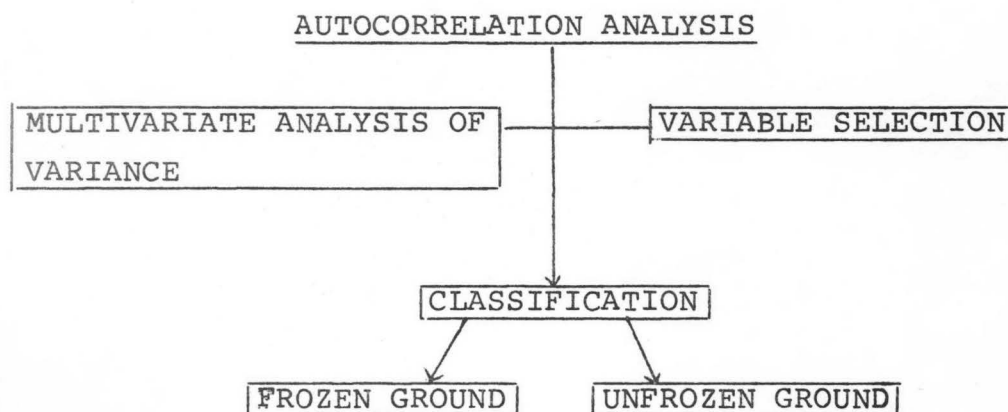
He recommended discriminant analysis to solve this type of uncertainty problem. DeCoursey (1973) used discriminant analysis to analyze factors related to the adequacy of standard slope protection on small dams. His analysis showed that a linear combination of the fetch length, the surface area to length ratio (average width), and the

plasticity index of the surface material on the dam could be used to distinguish between the two groups about 85% of the time. Thus a discriminant equation can be used to assign structures to the two groups (those sites with no indication of damage as in group 1 and those sites with observed damage as group 2) such that the additional protection can be most economically provided. Miller (1962) used discriminant analysis to forecast airfield ceiling conditions up to two hours in advance.

But the application of discriminant analysis to determination of frozen soil or unfrozen soil is a new challenge. Although many assumptions in discriminant analysis do not fit the real world perfectly, this method still may be one possible way to classify soil into frozen or unfrozen condition groups.

CHAPTER III
EXPERIMENTAL METHOD

The computation procedures used in this thesis are shown below and described in the following sections.



AUTOCORRELATION ANALYSIS

Autocorrelation is used for investigating dependence of one observation on another observation. The population autocorrelation coefficient of an observation series, X_i , is defined as

$$\rho_d = \text{cov}(X_i, X_{i+d}) / \text{var}(X_i) \quad (1)$$

Where $\text{cov}(X_i, X_{i+d})$ is the covariance between X_i and X_{i+d} , $\text{var}(X_i)$ is the variance of X_i and d is the lag.

For a discrete series, the value of ρ_d is estimated from a sample of size N and lag k and the estimator is

$$r_k = \text{cov}(X_i, X_{i+k}) / [\text{var}(X_i) * \text{var}(X_{i+k})]^{1/2} \quad (2)$$

For an uncorrelated series, the sampling distribution of r_k has an expected value $E(r_k)$ and a variance $\text{var}(r_k)$ given as

$$E(r_k) = -1/(N-k+1) \quad (3)$$

$$\text{var}(r_k) = \frac{[(n-k+1)^3 - 3(N-k+1)^2 + 4]/(N-k+1)^2}{[(N-k+1)^2 - 1]} \quad (4)$$

If a value of N is larger than 30, the sampling distribution of r_k may be approximated by a normal distribution (Torrainin, 1972). The 95% confidence limits of the uncorrelated series can be computed by

$$\rho(k) = E(r_k) \pm 1.96[\text{var}(r_k)]^{1/2} \quad (5)$$

If the sample correlogram lies within the confidence band, and/or if only a small percentage of r_k values defined by the confidence band lies outside these limits, this series can be considered to be serially uncorrelated.

MULTIVARIATE ANALYSIS OF VARIANCE

After the data are divided into two or more groups, it must be determined whether or not these groups have different population means among them and whether or not these groups all come from the same population. The mathematical model of multivariate analysis of variance is

$$X_{ki} = \mu + (\mu_k - \mu) + (X_{ki} - \mu_k) \quad (6)$$

Where X_{ki} is the dependent vector variable for the i -th observation in the k -th group.

μ is the vector of total samples means.

μ_k is the centroid for group k .

($k = 1, 2, \dots, g$. $i = 1, 2, \dots, N_k$. $N = N_1 + N_2 + \dots + N_g$)

From multivariate analysis of variance,

$$T = A + W \quad (7)$$

where T is total sum of square and cross products matrix of $p \times p$ order (p is the number of variables). A is the among-group sum of square and cross products matrix. W is the within-group sum of square and cross products matrix. The estimator of W based on the pooled within-groups deviations is

$$D_w = W/(N-g) \quad (8)$$

The estimator of A based on the group means vector is

$$D_a = A/(g - 1) \quad (9)$$

Test of Homoscedasticity of Dispersion Matrices

Box (1949) defined the test criterion M for

$H_0: \Sigma = \Sigma_k, k = 1, 2, \dots, g$

$$M = (N - g) * \ln |D_w| - \sum_{k=1}^g (N_k - 1) * \ln |D_k| \quad (10)$$

Required functions of the design parameters are

$$f_1 = \left[\sum_{k=1}^g \frac{1}{(N_k - 1)} - \frac{1}{(N-g)} \right] * (2 * p^2 + 3 * p - 1) / 6 (g-1) (p+1) \quad (11)$$

$$f_2 = \left[\sum_{k=1}^g \frac{1}{(N_k - 1)^2} - \frac{1}{(N-g)^2} \right] * (p-1)(p-2) / 6 (g-1) \quad (12)$$

If $f_2 - f_1^2 > 0$, then

$$\begin{aligned} n_1 &= (g-1)p(p+1)/2 \\ n_2 &= (n_1+2)/(f_2 - f_1^2) \\ b &= n_1 / [1 - f_1 - (n_1/n_2)] \\ F(n_1, n_2) &= M/b \end{aligned} \quad (13)$$

If $f_2 - f_1^2 \leq 0$, then

$$\begin{aligned} n_1 &= (g-1)p(p+1)/2 \\ n_2 &= (n_1+2)/(f_1^2 - f_2) \end{aligned}$$

$$b = n_2/[1-f_1+(2/n_2)]$$

$$F(n_1, n_2) = n_2 * M / n_1 (b - M) \quad (14)$$

If $F(n_1, n_2) < F_0$ at $\alpha\%$ significance level and d.f. = n_1, n_2 , we accept the null hypothesis. Otherwise we reject the null hypothesis.

Test of Equal-Group-Mean Vectors

The null hypothesis is $H_0: \mu_1 = \mu_2 = \dots = \mu_g$.

Wilk's determinant ratio test statistic is defined as

$$U = |W|/|T| \quad (15)$$

For Rao's F approximation it is necessary to compute a set of functions of the design parameters.

$$s = \sqrt{[p^2 (g-1)^2 - 4] / [p^2 + (g-1)^2 - 5]}$$

$$n_1 = (p * g - 1)$$

$$n_2 = s[(N-1) - (p + (g-1) + 1)/2] - [p(g-1) - 2]/2$$

Now let $y = U^{1/s}$

$$F(n_1, n_2) = [(1-y)/y] (n_2/n_1) \quad (16)$$

If $F(n_1, n_2) < F_0$ at a significance level of $\alpha\%$ and d.f. = n_1, n_2 , we accept the null hypothesis, H_0 at a certain level, and vice versa.

DISCRIMINANT ANALYSIS AND CLASSIFICATION OF DATA

Discriminant analysis is used to find out whether there is a set of the variables that can differentiate between the groups and use these variables to assign future individuals to a group.

In this thesis, the BMD-Biomedical Computer Program (Dixon, 1965) and the Statistical Analysis System (Barr and Goodnight, 1972) were used to analyze the frozen ground events. SAS-Discriminant Analysis was chosen for the analysis of frozen events because it can handle the conditions of unequal within group variance-covariance matrices and unequal priori probabilities between frozen and unfrozen events.

The BMD-Stepwise Discriminant Analysis can be used to select variables which have capability to discriminate the groups and also it can compute canonical correlations and coefficients for discriminant functions so that we can use these first two discriminant functions to plot two-dimensional pictures of the dispersion. The equations for the discriminant equation are as follows

$$(A - \lambda_i W) * V_i = 0 \quad (17)$$

A and W are as defined previously. V_i is the eigenvector (coefficient vector for discriminant function) corresponding to the eigenvalue λ_i . If the eigenvector is normalized we obtain

$$V_i W^{-1} V_j = \delta_{ij} \quad (18)$$

The canonical correlations of r_1, r_2, \dots, r_p relative to the groups are

$$r_i = \sqrt{\lambda_i / (1 + \lambda_i)} \quad (19)$$

The BMD program was to obtain first two discriminant functions for plotting a two-dimensional scattergram. The equations are:

$$\begin{aligned}
 Y_{mk} &= \sum_{j=1}^r v_{j1} * (X_{mkj} - \bar{X}_j) \\
 z_{mk} &= \sum_{j=1}^r v_{j2} * (X_{mkj} - \bar{X}_j)
 \end{aligned}
 \tag{20}$$

The SAS-Discriminant Analysis uses a Bayesian conditional probability interpretation of discriminant analysis and did not reduce the problem to 2 dimension but used the 8 variables directly to classify our data.

The classification of data is concerned with quantitative methods for the assignment of an individual observation to one of several groups on the basis of multiple measurements on that observation. It is assumed that each observation belongs to one of several specified groups or populations. Several relevant characteristics of the individual observation are observed, and on the basis of this information, a classification decision is made. The method used to assign individuals to these groups must have a minimum probability of error.

Let the conditional probability of an observation X_i (a $p \times 1$ vector) under the given group H_j be $P(X_i | H_j)$. According to the Bayesian conditional-probability model, we get

$$P(H_j | X_i) = P(X_i | H_j) * P(H_j) / \sum_{j=1}^g [P(X_i | H_j) * P(H_j)]
 \tag{21}$$

$i=1, 2, \dots, N$ (Number of observation).

$j=1, 2, \dots, g$ (Numbers of group).

$P(H_j)$ is called a priori probability in a given group H_j . $P(H_j | X_i)$ is called a posteriori probability

under the given observation X_i . The probability that the observation X_i belongs to group H_j is equal to the probability of the observation under group H_j times a priori probability divided by total probability.

Let it be assumed that the a priori probabilities exist and are estimable from the group relative frequencies, that is

$$\hat{P}(H_j) = n_j/N \quad (22)$$

Where: N is the total sample size.

n_j is the sample size in the group j .

Suppose a posteriori probability $P(X_i | H_j)$ has a multivariate normal distribution. Then

$$\hat{P}(X_i | H_j) = (|S_j|^{-1}/2\pi^p)^{1/2} \exp\left[-\frac{1}{2}(X_i - \bar{X}_j)' S_j^{-1} (X_i - \bar{X}_j)\right]$$

Where S_j is a within group variance-covariance matrix from samples in group j . Then

$$P(H_j | X_i) = \exp(-\chi^2_{ij}/2) / \sum_{j=1}^g \exp(-\chi^2_{ij}/2) \quad (24)$$

$$\text{Where: } \chi^2_{ij} = (X_i - \bar{X}_j)' S_j^{-1} (X_i - \bar{X}_j) + \ln |S_j| - 2 \ln [P(H_j)]$$

If the model of the a posteriori probability $P(X_i | H_j)$ is unknown, a nonparametric method (Pelto, 1969) could be used. Here, it is assumed that the data itself has a multivariate normal distribution, but the author did not verify it.

OTHER TOPICS IN DISCRIMINANT ANALYSIS

Lachenbruch (1968) introduced a method for the determination of the sample size for two groups in discriminant analysis. He said that for the sample discriminant

function to have an error rate within some given tolerance limit, n_1 and n_2 observations are necessary. How n_1 and n_2 are obtained depends on p (number of variables), tolerance limit, and $(\mu_1 - \mu_2)' \Sigma^{-1} (\mu_1 - \mu_2)$, where μ_1 is the population mean in group 1, μ_2 is the population mean in group 2, and Σ is population within-group variance-covariance matrix. Trial and error must be used to make the probability of misclassification less than some specified tolerance. Detail procedures can be obtained from Lachenbruch (1968).

A logical method to select variables in discriminant analysis is desirable because the set of variables available for measurement may be large and may contain redundancies which do not help the discriminant power and would increase the cost of measurement. We would like to select only those variables which have the capability to discriminate between groups in order to proceed to classification analysis. If there are p variables for our selection, then there are 2^p combinations. For each combination of variables a Wilk's determinant ratio U (equation 15) can be calculated. Clearly $0 \leq U \leq 1$ and small values of U indicate "good" discrimination. Thus only those combinations which have U values below a certain value would be used.

For the case of missing observations for some of the variables, Jackson (1968) introduced an iterative method in which an unknown variable is treated as the dependent variable for a regression analysis and all other variables are used as independent variables with

mean values substituted for unknowns from the regression equation. A new estimate of each unknown value is estimated and substituted.

Another problem is that the derivation of the discriminant functions is under the assumption of equal variance-covariance matrices for the groups. If these matrices are not all equal, a method proposed by Anderson and Bahadur (1962) can be used for the two group situation. Further discussion of this problem is beyond the scope of this thesis.

CHAPTER IV
DRAINAGE BASIN DESCRIPTION

Missouri Flat Creek is located from $46^{\circ}48'37''$ to $46^{\circ}43'47''$ north latitude and $116^{\circ}58'48''$ to $117^{\circ}10'30''$ west longitude. It originates on the northwestern corner of Latah County, Idaho and flows through Pullman, Washington. The total length of this watershed is approximately 12 miles. The drainage area is 27.1 square miles. A map of the basin is shown in Figure 2.

The basin consists of an irregular plateau with rolling topography averaging 2800 ft in elevation. It ranges from 2360 ft in elevation in Pullman to 3300 ft in the headwaters. The slope of this area is from 3% or less to 40% or over. Nearly all of the basin is in cultivation, wheat and peas being the principal crops. The cropping pattern has not changed over the years and cropland occupies 92% of the watershed, pasture 3% and others 5% (Potter and Love, 1942).

The acreage of each soil group and type are shown on Table 3. Dark, medium-textured, deep prairie-land soils with almost no stones cover most of the area. Shallow rocky soils are found in the higher elevations.

The mean annual maximum temperature is 57.4°F and mean annual minimum temperature is 36.4°F . The average monthly maximum and minimum temperatures are listed in Table 2. The mean annual snowfall of this basin is 22.4 inches.

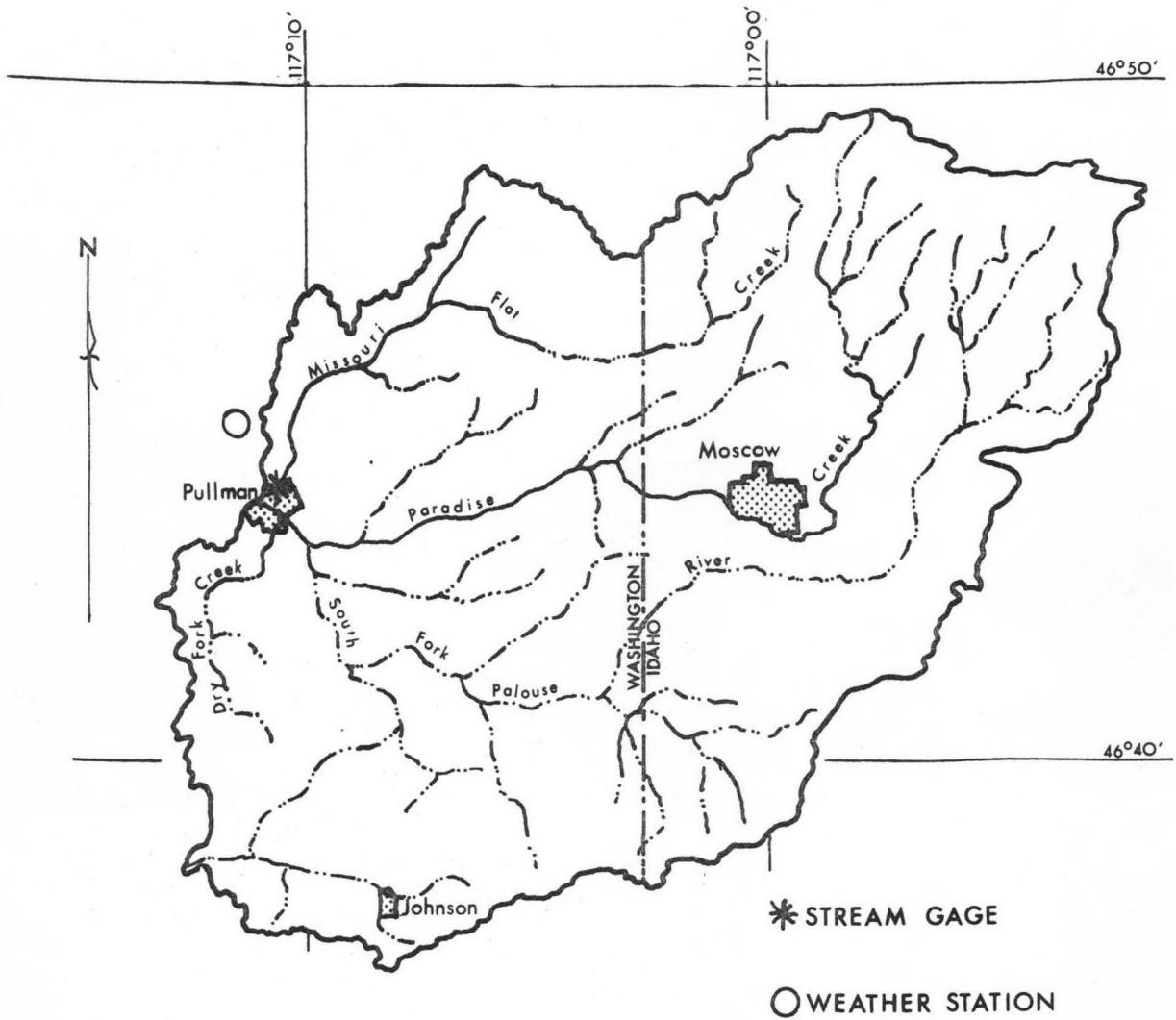


FIGURE 2. Map of Missouri Flat Creek and South Fork Palouse River Drainage Basins.

The total monthly snowfall and precipitation at Pullman from 1940-1971 is listed in Table 2.

TABLE 2 MEAN MONTHLY CLIMATIC FACTORS FOR PULLMAN, WASHINGTON (1940-1971)

Month	Temperature		Snowfall (in.)	Precipitation (in.)
	Maximum	Minimum		
January	34.4	21.7	7.1	3.13
February	40.6	27.0	3.6	2.16
March	46.0	29.6	3.1	2.01
April	55.8	5.2	0.5	1.54
May	64.6	41.0	0.0	1.55
June	71.0	46.3	0.0	1.70
July	82.2	49.2	0.0	0.46
August	81.4	49.2	0.0	0.75
September	72.3	44.6	0.0	1.15
October	59.7	38.0	0.2	1.94
November	43.9	30.4	2.0	2.94
December	36.3	25.5	5.9	3.05

TABLE 3 SOIL TYPES IN THE MISSOURI FLAT CREEK BASIN

Soil	Area (acres)	Percent
Dark-colored upland soil	16805	0.885
Light-colored upland soil	68	0.004
Soil on terraces	347	0.018
Soil on flood plains	1655	0.087
Rough stony land	<u>111</u>	<u>0.006</u>
Total	18986	100.00

Most of the precipitation is distributed from October through March. Rainfall intensities are usually very low. The 24-hour, 10-year precipitation value is 1.9 inches. This area has moderately cold winters dominated by Pacific maritime air masses with occasionally outbursts of colder continental air from Canada.

The city of Pullman is located at the confluence of Missouri Flat Creek and South Fork of the Palouse River. The population is increasing in this area and as a result, new areas in the flood plain are being developed. Flood induced damages have been increasing at a rapid rate. Basin flood plain zoning has been enacted to help decrease the damages.

CHAPTER V
RESULTS AND DISCUSSION

DATA COLLECTION

The available climatological and stream flow records were obtained from the National Weather Service and the U.S. Geological Survey. Data from October to April, covering 143 runoff events from 1960 to 1973 were used. Fifty-four of the 143 samples were determined to be either frozen or unfrozen ground events. The remaining 89 events could not be classified by an examination of the historical records. These 54 events were then used as a basis for deriving the prediction equation.

On the basis of the literature review, eight readily available variables were chosen as discriminators for a discriminant analysis. These eight variables are listed in Table 4.

DATA ANALYSIS

In this section, the derivation of the classification procedure is shown and applied to three of the unknown observations. This is done to show the procedures used to arrive at the ultimate classification of all 89 unknown observations. The methods used are outlined in Chapter III.

TABLE 4 DISCRIMINATORS FOR FROZEN OR UNFROZEN GROUND ANALYSIS

Discriminators	Notation	Unit
1) Number of days below freezing	X_1	Day
2) Total freezing index	X_2	Degree Day
3) Precipitation, 2 days before freeze period	X_3	Inch
4) Precipitation during freeze period	X_4	Inch
5) Precipitation, 4 days after freeze period	X_5	Inch
6) The depth of snowfall during freeze period	X_6	Inch
7) The depth of snow on ground at start of freeze period	X_7	Inch
8) Stream flow for 4 days after freeze period	X_8	Inch

Autocorrelation Analysis

If the coefficients of autocorrelation are near zero, the sequence being tested is assumed to be random and the notion of the i -th variate being independent of the $(i+k)$ -th variate (k is lag) is supported. The autocorrelation coefficients of the freezing index and stream flow is shown in Figures 3 and 4. It can be seen that most of the observations are within the 95% confidence band. So, it is assumed that the data are independent and random. All

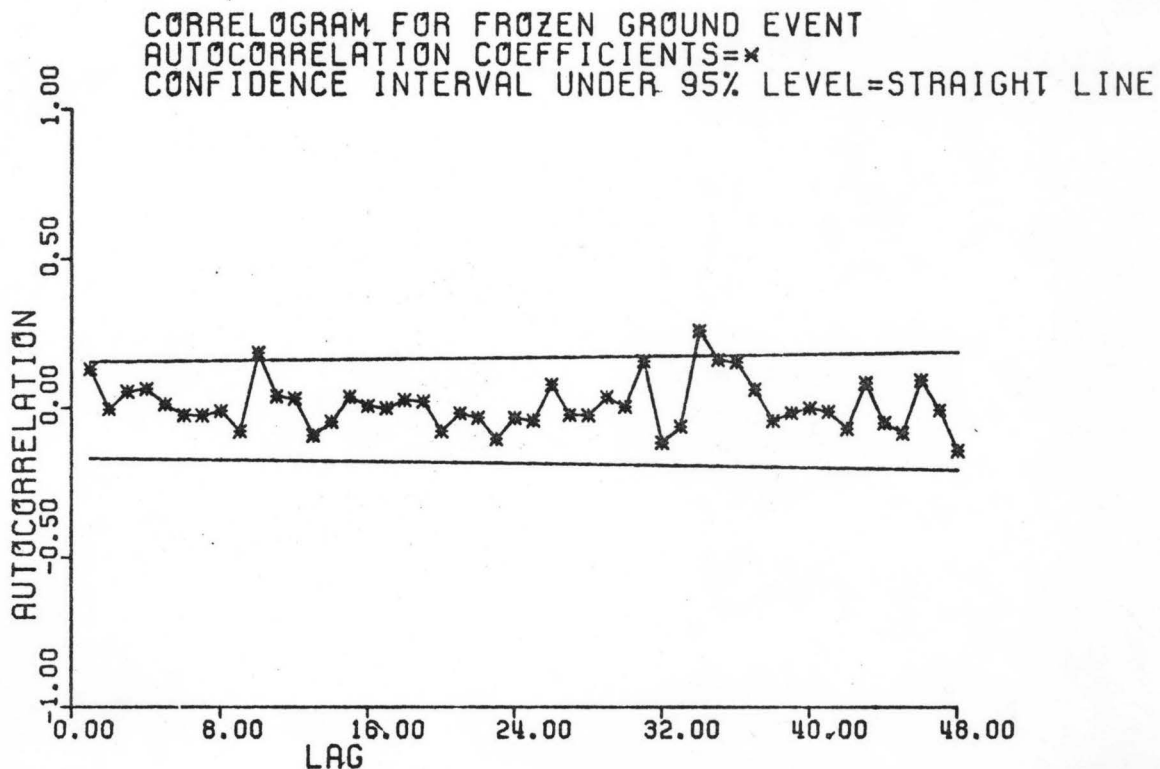
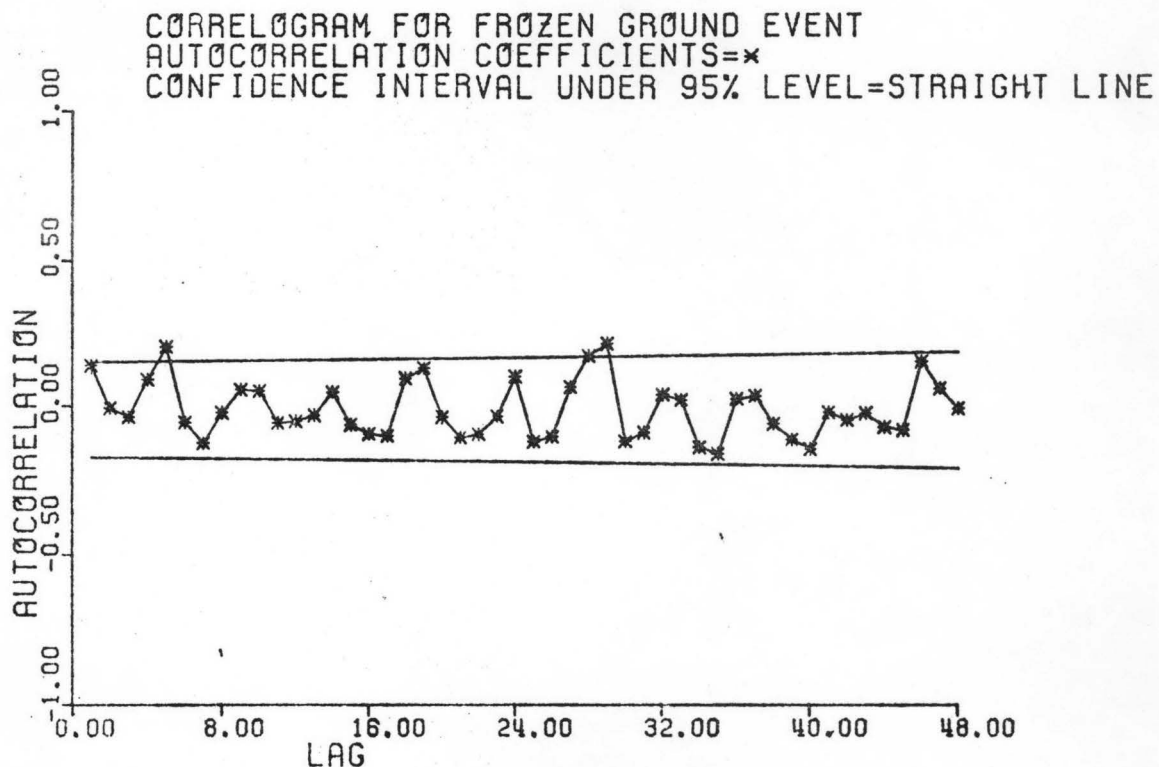


FIGURE 3 Autocorrelation analysis for total freeze index

FIGURE 4 Autocorrelation analysis for streamflow,
4 days after freeze period

143 observations are included in this analysis and all 8 variables were checked, but only 2 are shown here for reasons of clarity.

Multivariate Analysis of Variance and Variable Selection

The objective of multivariate analysis of variance in this thesis is to test group differences in a multi-dimensional measurement space after some combination of variables, which have a small value of Wilk's U, is chosen. Also, our program of a multivariate analysis of variance gives a test of the variance-covariance matrix differences between these two groups for use in classification.

To select the variables, we used the BMD-Stepwise Discriminant analysis program and McCabe and Pohl's optimal method. The BMD method used "F-value to enter or remove" to control variable selection. After the variable is included, Wilk's U is calculated. McCabe's (1975) method computed the Wilk's U-value for each of the $2^p - 1$ subsets of the variables X_1, X_2, \dots, X_p . The algorithm first determines the U-value of the optimum set of variables by calculating and comparing the U-values of all possible sets. The order in which the program choose the variables is shown in Table 5. Table 6 shows the BMD method. Next, multivariate analysis of variance was used to test the difference between these two groups for the chosen set of variables.

TABLE 5 MCCABE AND POHL'S STEPWISE SELECTION OF VARIABLES

Stage	Variable								U	F-value	D.F.	Theoretical F(1%)
	1	2	3	4	5	6	7	8				
1		*							0.522	47.62	1-52	7.14
2		*							0.399	38.41	2-51	5.04
3		*	*					*	0.396	25.42	3-50	4.20
4		*	*			*		*	0.394	18.84	4-49	3.73
5		**		*		*		*	0.390	15.02	5-48	3.43
6			*	*	*		*	*	0.388	12.36	6-47	3.21
7		*	*	*	*		*	*	0.382	10.63	7-46	3.06
8		*	*	*	*	*	*	*	0.379	9.22	8-45	2.94

The asterisk symbol in Tables 5 and 6 indicates the variables chosen in each stage.

TABLE 6 BMD STEPWISE SELECTION OF VARIABLES

Stage	Variable								F-value to remove	U-value
	1	2	3	4	5	6	7	8		
1		*							47.67	0.522
2		*						*	15.58	0.399
3		*	*					*	0.36	0.396
4		*	*				*	*	0.27	0.394
5		*	*		*		*	*	0.52	0.390
6		*	*		*	*	*	*	0.16	0.389
7		*	*	*	*	*	*	*	0.19	0.388
8		*	*	*	*	*	*	*	0.88	0.379

From Table 5 and 6, we can see that the variables entered in the first five stages are the same between these two methods but after that, the variables entered are different. There is nothing in the literature to indicate which method is better. For the purposes of this thesis, all 8 variables were included. An F test for all 8 variables is as follows.

$$\begin{aligned} s &= \sqrt{[p^2(g-1)^2-4]/[p^2+(g-1)^2-5]} \\ &= \sqrt{[8^2(2-1)^2-4]/[8^2+(2-1)^2-5]} \\ &= 1 \end{aligned}$$

$$n_1 = (p \cdot g - 1) = 8(2-1) = 8$$

$$\begin{aligned} n_2 &= s[(N-1) - (p + (g-1) + 1)/2] - [p(g-1) - 2]/2 \\ &= [(54-1) - (8 + (2-1) + 1)/2] - [8(2-1) - 2]/2 = 45 \end{aligned}$$

$$y = U^1/s = 0.379$$

$$F(8, 45) = (1-y)(n_2/n_1)/y = (1-0.379)(45/8)/0.379 = 9.22$$

The theoretical F-value is 2.94 under d.f.=8.45 and 1% level.

So, $F > F_0$, we reject the $H_0: \mu_1 = \mu_2$.

Classification of Data

The SAS method was used for the classification of frozen ground events rather than the BMD method because SAS can handle unequal priori probability and unequal within group variance-covariance matrices.

In order to determine whether or not the frozen and unfrozen ground events come from the same population, Box's method (1949) with M-statistic was used to test $H_0: \Sigma_1 = \Sigma_2$, where Σ_1 is the "within group" variance-covariance matrix of the population of frozen ground and that of Σ_2 is

unfrozen ground. Doing this, $M=189.54$ is obtained. This M value is converted to $F=3.72$, $n_1=36$ and $n_2=1413$ degree of freedom. Since $F(36,1413)$ is 1.60 at the 1% level, we reject the null hypothesis $H_0: \Sigma_1 = \Sigma_2$ and accept $H: \Sigma_1 \neq \Sigma_2$. This means that frozen ground and unfrozen ground events do not come from the same population. Thus the "within group" variance-covariance matrices are unequal between these two groups. This will have to be considered later in the classification analysis and may restrict use of the discriminant functions as a classification tool.

A Bayesian conditional-probability model was used to identify the probability of frozen or unfrozen ground occurring during a given freeze event. Suppose the population means and within group variance-covariance matrices of each group can be estimated from that of the samples. Prior probabilities can also be estimated from the group relative frequencies. Then, if three observations are obtained from historical data without knowing whether or not these observations are for frozen ground events, the following procedure may be used to assign a frozen or unfrozen probability to each observation.

The means and the "within group" variance-covariance matrices which were calculated from the original 54 observations (42 are unfrozen and 12 are frozen) are taken as given.

TABLE 7 THE MEANS OF EACH GROUP

Group	Variable							
	\bar{X}_1	\bar{X}_2	\bar{X}_3	\bar{X}_4	\bar{X}_5	\bar{X}_6	\bar{X}_7	\bar{X}_8
Frozen	10.83	143.13	0.23	0.63	1.00	5.54	0.42	0.54
Unfrozen	3.57	17.57	0.23	0.25	0.48	2.36	1.07	0.15

TABLE 8 THREE ADDITIONAL SAMPLE OBSERVATIONS

Observation	Variable							
	X_1	X_2	X_3	X_4	X_5	X_6	X_7	X_8
1	1.00	2.00	0.52	0.05	1.74	0.00	1.00	0.05
2	13.00	84.50	0.00	0.80	0.44	5.80	0.00	0.12
3	12.00	70.00	0.25	1.21	0.94	12.25	0.00	0.60

First, Mahalanobis' distance, $D_{ij}^2 = (X_i - \bar{X}_j)' S_j^{-1} (X_i - \bar{X}_j)$, must be calculated. Where the vector of observation 1, X_1 , is

$$(1.00 \ 2.00 \ 0.52 \ 0.05 \ 1.74 \ 0.00 \ 1.00 \ 0.05)'$$

$(X_i - \bar{X}_j)'$ is the transpose of $(X_i - \bar{X}_j)$. \bar{X}_1 , the mean vector of group 1 (frozen ground) is

$$(10.83 \ 143.13 \ 0.23 \ 0.63 \ 1.00 \ 5.54 \ 0.42 \ 0.54)'$$

and \bar{X}_2 , the mean vector of group 2 (unfrozen ground) is

$$(3.57 \ 17.57, \ 0.23, \ 0.25 \ 0.48 \ 2.36 \ 1.07 \ 0.15)'$$

Then, D_{11}^2 is the value of observation 1 in frozen ground

$$= [(1.00-10.83) \ (2.00-143.13) \ (0.52-0.23) \ (0.05-0.63) \\ (1.74-1.00) \ (0.00-5.54) \ (1.00-0.42) \ (0.05-0.54)]'*$$

0.65	-0.02	2.90	-0.21	0.69	-0.30	-0.67	-1.00
-0.02	0.00	-0.09	-0.01	-0.03	0.01	-0.67	0.06
2.90	-0.09	36.00	9.50	2.80	-2.60	-8.30	-1.60
-0.20	-0.01	9.50	33.00	-0.42	-2.00	-4.80	-14.00
0.69	-0.03	2.80	-0.42	2.40	-0.35	-0.59	-4.70
-0.30	0.01	-2.60	-2.00	-0.35	0.32	0.88	1.40
-0.67	0.02	-8.30	-4.80	-0.59	0.88	5.00	2.50
-1.00	0.06	-1.60	-14.00	-4.70	1.40	2.50	29.00

$$*[(1.00-10.83) (2.00-143.13) (0.52-0.23) (0.05-0.63) \\ (1.74-1.00) (0.00-5.54) (1.00-0.42) (0.05-0.54)]$$

So, $D^2_{11}=12.01$. D^2_{12} , D^2_{21} , D^2_{22} , D^2_{31} , and D^2_{32} are calculated similarly and are listed in Table 9.

TABLE 9 THE MAHALANOBIS' DISTANCE D^2_{ij}

Observation	Frozen ground	Unfrozen ground
1	12.01	18.16
2	18.47	23.02
3	5.51	24.36

$$\hat{P}(X_i | H_j) \hat{P}(H_j) = \hat{P}(H_j) \exp[-(X_i - \bar{X}_j)' S_j^{-1} (X_i - \bar{X}_j) / 2] / \sqrt{2\pi^p |S_j|} \\ = \exp\left[-\frac{(X_i - \bar{X}_j)' S_j^{-1} (X_i - \bar{X}_j)}{2} - \ln |S_j| + 2 \ln \hat{P}(H_j)\right]$$

$$\text{Then, let Chi-square} = [(X_i - \bar{X}_j)' S_j^{-1} (X_i - \bar{X}_j) + \ln |S_j| - 2 \ln \hat{P}(H_j)] \\ = [D^2_{ij} + \ln |S_j| - 2 \ln \hat{P}(H_j)]$$

Where $\ln|S_1|=3.18$, $\ln|S_2|=-1.02$ and $2*\ln\hat{P}(H_1)=2*\ln(0.22)=-3.03$,

$$2*\ln\hat{P}(H_2)=2*\ln 0.78=0.50$$

The Chi-square values are shown in Table 10.

TABLE 10 CHI-SQUARE VALUES

Observation	Frozen ground	Unfrozen ground
1	18.22	17.64
2	24.68	22.50
3	11.72	23.84

TABLE 11 CONDITIONAL PROBABILITIES FOR 3 OBSERVATIONS

Observation	Frozen ground	Unfrozen ground
1	0.430	0.570
2	0.250	0.750
3	0.998	0.002

From Table 10, it is seen that Chi-square for observation 1 for frozen ground is greater than that of observation 1 for unfrozen ground. This means that observation 1 is closer to the centroid of the known unfrozen ground observation than that of frozen ground. In Table 11, the probability of frozen ground in observation 1 is less than that of unfrozen ground in observation 1 since the probabilities are inversely proportional to the distance. The probabilities of frozen ground of 143 observations in the

data were calculated by a computer program using the above procedures. All the results are given in the appendix.

In the original data, forty-two of the 54 observations were unfrozen ground events and twelve were frozen ground. In order to test the discriminatory power on the basis of these eight variables, these 54 given observations were classified with the Bayesian method. It was found that 2 of the observations of frozen ground events which had been thought to be frozen ground events were classified into the unfrozen ground group and 1 of 42 observations of unfrozen ground were misclassified into the frozen ground group. This is not too surprising since the original classification was made using somewhat sparse data. In matrix notation this misclassification is indicated by the "confusion matrix".

$$C = \begin{bmatrix} 41 & 1 \\ 2 & 10 \end{bmatrix}$$

There is a method to test the discriminatory power on the basis of their variables for this confusion matrix. Massy (1965) used the Chi-square statistic for testing this discriminatory power. The null hypothesis is that the function does not have the ability to classify correctly. If we reject the null hypothesis, it means that we have a relatively high degree of confidence in its ability to discriminate properly. The statistic is

$$\text{Chi-square} = (N - nk)^2 / [N(k-1)]$$

Where N is total number of observations.

n is the number of correct classifications.

k is the number of populations.

So, Chi-square = $(54 - 51 * 2)^2 / 54(2 - 1) = 42.67$. Its degree of freedom is one. The theoretical Chi-square at the 1% level and 1 d.f. is 3.84. Since 42.67 is greater than the theoretical Chi-square, the null hypothesis is rejected. We are confident that the Bayesian method was able to distinguish between frozen and unfrozen ground events.

Results and Discussion

Using the 89 unknown observations in the Bayesian classification model, it was found that 10 of the 89 unknown observations were classified into frozen ground events and the remaining 79 were classified as unfrozen. Table 12 shows 20 observations sampled at random from the 89 events.

A frozen ground event is caused by the total effects of the eight factors. In Table 12, it can be seen that the freeze index of some observations is not very large and the soil still is assigned a frozen condition because the other seven variables carry enough weight to assign the event to frozen ground. A trait of multivariate analysis is to consider all eight variables as a system. A frozen ground event is caused by the combined effects of these eight correlated factors, not by a single variable. Also from the data in Table 12, it can be seen that McCool and Molnau (1974) could not correctly classify all runoff events

TABLE 12 THE PROBABILITY OF FROZEN GROUND FOR 20 RANDOM
EVENTS 1960 TO 1973; MISSOURI FLAT CREEK

Month/Date/ Year	X ₁	X ₂	X ₃	X ₄	X ₅	X ₆	X ₇	X ₈	Prob. of frozen ground events
1) 12/15/60	3.0	13.5	0.05	0.15	0.91	1.70	0	0.05	0.06%
2) 12/04/62	1.0	2.0	0.68	0.00	0.04	0	0	0.01	4.89%
3) 2/13/69 2/15/69	3.0	13.5	0.46	0.00	0.00	0	0	0.18	1.57%
4) 1/28/70 1/30/70	3.0	4.5	0.36	0.01	0.21	0	0	0.16	0.98%
5) 11/22/60	1.0	2.0	0.52	0.0.	1.74	0	1	0.05	42.70%
6) 12/23/60 1.04/61	13.0	84.5	0.00	0.80	0.44	5.8	0	0.12	25.09%
7) 11/24/61 11/26/61	3.0	12.0	1.45	0.86	0.11	8.0	2	0.01	40.00%
8) 12/5/61 12/16/61	11.0	137.5	0.23	1.10	1.31	10.1	0	0.02	100.00%
9) 11/15/61 11/21/61	7.0	59.0	0.03	0.71	2.29	3.7	0	0.02	99.45%
10) 1/09/62 1/24/62	16.0	247.0	0.37	0.70	0.01	8.8	0	0.02	100.00%
11) 2/21/62 3/04/62	12.0	123.0	0.00	0.96	0.05	11.6	0	0.06	98.52%
12) 1/10/63 1/14/63	5.0	115.5	0.00	0.05	0.03	0.5	0	0.00	100.00%
13) 12/25/69 1/08/70	15.0	144.0	0.17	0.31	1.11	4.9	2	0.02	99.35%
14) 1/10/70 1/13/70	4.0	13.0	0.48	1.08	0.60	1.5	2	0.16	100.00%
15) 1/17/70 1/19/70	3.0	14.5	0.18	1.59	1.16	6.2	0	0.73	100.00%
16) 2/26/71 3/09/71	12.0	70.0	0.25	1.21	0.94	12.5	0	0.60	99.77%
17) 12/27/68 1/04/69	9.0	199.5	0.04	1.44	1.42	21.5	0	0.90	100.00%
18) 12/22/62 12/26/62	5.0	46.5	0.04	0.00	0.09	0.0	0	0.01	1.19%
19) 1/20/73 1/23/73	4.0	5.0	0.04	0.00	0.00	0.0	0	0.03	0.54%
20) 11/28/62 11/29/62	2.0	6.5	0.34	0.00	1.12	0.0	0	0.01	0.74%

on the basis of freeze index along since several of the observations have a freeze index below 100.

The BMD method uses a certain number of discriminant functions with equal prior probabilities among groups and equal within-group variance-covariance matrices. The first discriminant function usually takes care of a major part of the variance, the second one takes care of a major part of the remainder, and so on. The first two discriminant variables for the fifty-four observations known to be frozen or unfrozen are listed as follows

$$\begin{aligned}
 y_{mk} = & -0.11*(X_{mk1}) - 5.19 + 0.01*(X_{mk2} - 45.47) + 0.91* \\
 & (X_{mk3} - 0.23) + 1.07*(X_{mk4} - 0.33) + 0.21*(X_{mk5} - 0.59) \\
 & - 0.13*(X_{mk6} - 3.07) - 0.10*(X_{mk7} - 0.93) + 1.79*(X_{mk8} - 0.24) \\
 z_{mk} = & -0.57*(X_{mk1} - 5.19) + 0.03*(X_{mk2} - 45.47) - 1.41*(X_{mk3} - 0.23) \\
 & - 2.16*(X_{mk4} - 0.33) - 0.20*(X_{mk5} - 0.59) + 0.27*(X_{mk6} - 3.07) \\
 & + 0.13*(X_{mk7} - 0.93) + 0.20*(X_{mk8} - 0.24)
 \end{aligned}$$

These two functions, when plotted in two dimensional scattergrams, result in Figure 5 for fifty-four observations and Figure 6 for all the observations. It is obvious that the two groups are well separated in general, but there are some overlapping. The resulting classification is also somewhat different from that by the SAS method. The actual classification of events was done using the SAS program for the initial 54 observations and a program based on the SAS method for all additional observations (appendix C) because it can handle unequal prior probabilities and unequal within group variance-covariance matrices.

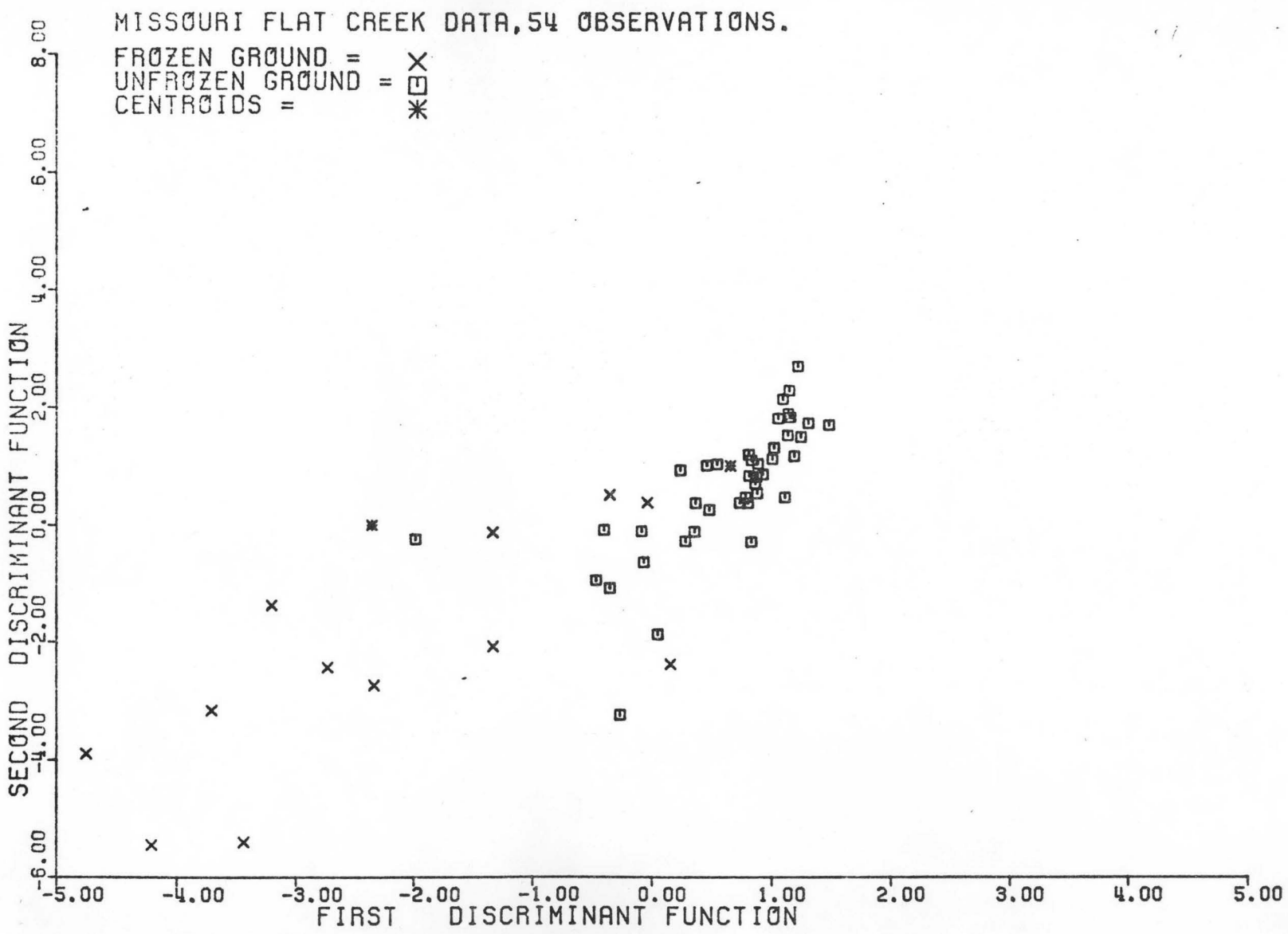


FIGURE 5 Scattergram of 54 observations

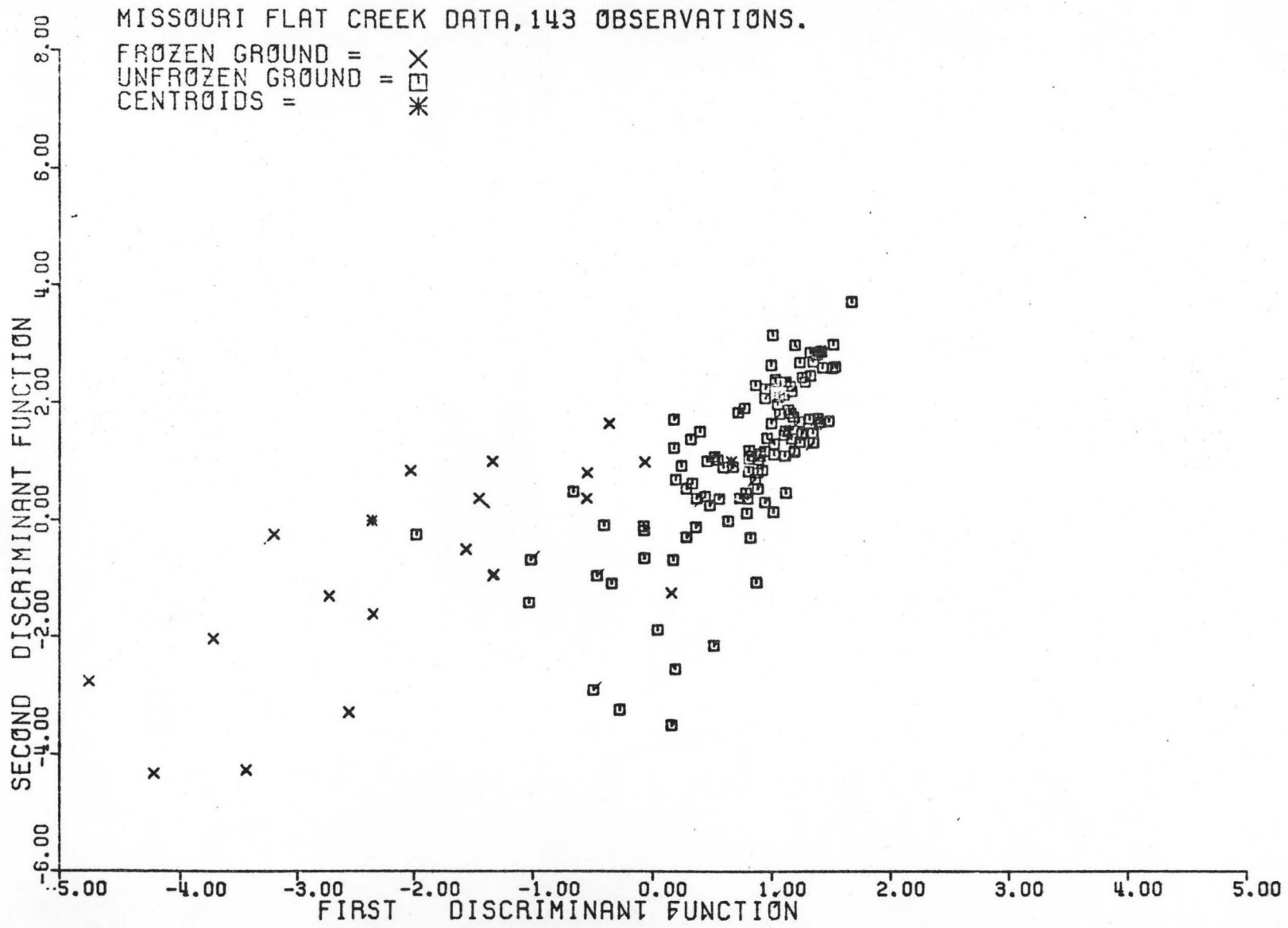


FIGURE 6 Scattergram of 143 observations

CHAPTER VI
CONCLUSIONS AND RECOMMENDATIONS

The main purposes of this thesis were to introduce a discriminant method to analyze the frozen ground problem in small watersheds and to identify historical frozen ground floods from past data. Although it is not known whether the data fits some of the assumptions of discriminant analysis or not, the combination of use of discriminant function and Bayesian classification appears to be a good way to classify frozen or unfrozen ground events.

The classification of observations can lead to results having widely different probabilities of error. For some observations, the decision of their assignment can be made with confidence. But in others the best decision will be subject to great uncertainty. Overall and Klett (1972) suggested:

1. Classify the observations into group I if the probability of its belonging to group II is smaller than some prespecified value, 0.10.
2. Classify the observation into group II if the probability of its belonging to group I is smaller than some prespecified value, 0.5.
3. Avoid making a definite classification pending examination of additional information.

There are several methods for variable selection. McCabe (1975) said that the selection of an optimal subset

of variables should not be considered as a final solution but rather as a starting point for further analysis.

Autocorrelation analysis can be used to show the independence or dependence of the data. Although some of the variables have dependent relations among the observations, their mutual independence is still assumed.

Mahalanobis' generalized distance is applicable only to groups in which the measurements are normally distributed. For our data we assume it is a multivariate normal distribution in order to simplify the analysis but much further analysis would have to be done in order to prove this.

Unequal within group variance-covariance matrices between these two groups cause an inaccurate discriminant function. This is why we used the 8 variables directly in the Bayesian classification model which can deal with unequal within group variance-covariance matrices.

CONCLUSIONS

The following conclusions can be drawn from this study:

1. The eight variables used in this thesis are sufficient to classify the data into frozen or unfrozen ground runoff events.
2. Using discriminant analysis, it was revealed that 10 of 89 unknown observations are probably frozen ground and the remaining 79 probably are not frozen ground runoff events.

3. The freeze index is the most important of the eight factors studied for estimating the probability of frozen ground in Missouri Flat Creek. This is seen by the stepwise selection of variables (Table 5).
4. The occurrence of a crust of frozen soil on a watershed at the start of a historical snowmelt or rainfall event cannot be studied by relying upon only one variable such as the freeze index, but several variables must be used to reflect the interaction that results in the physical world.
5. Precipitation on frozen ground does not necessarily cause high runoff.

RECOMMENDATIONS FOR FUTURE STUDIES

The following recommendations are given in the hope that the next step in this project will be able to build upon the results of this thesis. These recommendations are:

1. Check the classification of frozen ground with independent data from other sources to verify the classification.
2. Use other methods, such as a nonparametric method, to analyze the data and compare the events with the result of the method used in this thesis.

3. Use additional factors which may influence the occurrence of frozen ground events and try to use discriminant analysis again.
4. Use this method in another area where freeze-thaw cycles are prominent in the winter runoff region, such as Spokane or Pendleton.

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APPENDIX A

DATA INPUT LIST

1	8	44.5	0	0.01	1.54	0.1	0	160
1	3	6.5	0.75	0.13	0.31	0.9	0	174
1	2	9.0	0.43	0.03	0.20	0.3	0	32.1
1	5	14.5	0.05	0.26	0	3.2	0	46.4
1	1	1.5	0.01	0	0.01	0	0	9.2
1	3	14.0	0.10	0.95	0.48	7.5	4	39.6
1	4	17.5	0.03	0.54	0.04	5.3	0	33.5
1	1	1.0	0	0	0.47	0	0	40.6
1	2	2.0	0.35	0.10	0.59	0	1	88.0
1	3	5.5	0.04	0.41	0.42	2.9	0	94.0
1	7	28.0	0.34	0.28	0	2.6	2	63.6
1	5	16.5	0	0.17	0.72	0	1	58.0
1	1	2.5	0.42	0	1.21	0	1	1178
1	3	4.0	0.20	0.02	0.08	0.3	0	122
1	4	16.5	0.30	0	0.66	0	0	137
1	2	1.5	0.26	0.16	0.26	0.6	0	216
1	1	1.0	0.28	0	0.46	0	0	114
1	3	8.0	0.07	1.42	1.62	14.7	0	23.0
1	1	0.5	1.75	0.25	0.97	0	18	63.0
1	11	98.0	0.02	0.49	0.03	5.0	0	7.5
1	9	53.5	0.12	0.67	0.29	5.7	0	16.5
1	4	18.5	0.04	0	0.52	0	0	25.0
1	2	4.0	0.20	0.27	0	0.3	0	60.5
1	3	10.0	0.12	0.09	1.19	0.5	1	43.4
1	2	2.0	0.11	0.17	1.05	1.3	0	330
1	4	15.0	0.25	0.07	0.88	0.8	0	298
1	2	2.0	0.11	0.18	0.27	1.8	0	38.4
1	2	1.5	0.26	0	0	0	0	20.9
1	3	5.5	0.08	0.14	0.63	1.5	1	23.3
1	4	9.5	0.10	0.11	0.25	0	4	36.9
1	2	10.5	0.15	0.12	0.80	0	0	8.5
1	7	58.5	0.25	0.53	0.66	3.50	0	61.0
1	4	6.0	0	0.07	0	0.3	0	87
1	1	4.0	0	0.25	0.14	3.0	0	174
1	1	4.5	0.26	0	0.26	0	2	54.2
1	2	4.5	0.03	0.14	1.67	3.8	0	80.3
1	10	88.5	1.49	2.15	0.29	26.5	4	24.6

1	7	65.5	0	0.19	0.14	3.8	0	42.9
1	3	40.5	0	0.14	0.15	3.0	0	37.6
1	3	27.5	0	0	0.67	0	0	175
1	1	2.0	0.20	0	0.10	0	0	170
1	4	12.0	0.45	0	0.06	0	6	76.6
2	16	205.0	0.03	1.04	1.29	6.6	0	814
2	9	123.0	0.12	0.95	4.18	9.5	0	739.9
2	11	146.5	0.01	0.90	1.20	3.2	0	353.6
2	5	22.0	1.06	0.52	0.48	11.1	0	213
2	25	323.0	0.38	1.50	0.16	23.2	0	54
2	5	20.0	0	0	0.28	0	0	475
2	8	90.5	0.49	0.30	0.43	0.30	2	227
2	4	22.0	0.12	0.79	1.08	1.8	1	586
2	6	20.5	0.01	0.11	0.84	0.5	0	228
2	12	234.0	0.07	0.38	0.14	4.5	0	165
2	14	306.5	0.43	0.34	0.87	1.2	1	231
2	15	204.5	0.06	0.73	1.05	4.6	1	523
1	1	2	0.52	0.05	1.74	0	1	33.7
1	4	11.5	0.99	0	0.39	0	1	7.1
1	7	51.0	0.19	0	0.05	0	0	2.9
1	3	13.5	0.05	0.15	0.91	1.7	0	38.5
1	13	84.5	0	0.80	0.44	5.8	0	82.3
1	1	6.0	0.08	0.14	0	0.8	0	1.3
1	3	12.0	1.45	0.86	0.11	8.0	2	6.1
1	1	5.0	0.55	0.04	0.97	0.5	5	27.5
1	1	1.5	0.32	0.08	0.49	0	0	37.5
1	1	0.5	0	0.06	0.22	0	0	1.9
1	2	2.0	0.09	0.16	0.56	0.7	0	4.8
1	2	6.5	0.34	0	1.12	0	0	8.4
1	1	2.0	0.68	0	0.04	0	0	5.5
1	2	5.0	0	0	1.06	0	0	13.2
1	5	46.5	0.04	0	0.09	0	0	4.9
1	2	7.0	0	0	0	0	0	2.5
1	4	29.5	0	0	1.04	0	0	4.7
1	8	67.5	0.02	0.63	0.76	5.4	0	4.1
1	2	6.0	0.76	0	0.19	0	4	6.4
1	2	4.0	0	0.19	0.16	0	3	5.4

1	1	2.0	0.09	0.14	0.88	1.5	0	16.0
1	9	40.5	0.84	0.56	0.85	5.5	1	6.6
1	2	3.0	0.37	0.48	0.16	3.5	3	10.3
1	1	2.0	1.37	0	0.26	0	4	38.5
1	1	0.5	0	0	0.36	0	2	44.9
1	1	2.0	0.10	0	0.07	0	1	46.0
1	2	6.5	0.07	0	0.03	0	0	38.2
1	1	3.0	0.01	0	0.02	0	0	34.0
1	1	3.5	0.02	0	0.20	0	0	23.0
1	3	14.5	0.01	0.06	0	0	0	113.0
1	6	29.5	0	0	1.83	0	0	8.2
1	1	0.5	0	0.10	1.88	0	0	10.6
1	2	4.0	0.15	0.27	1.06	2.5	1	21.7
1	12	108.0	2.36	0.84	0.53	6.9	0	39.0
1	2	6.0	0.39	0.34	0.49	2.8	1	272
1	2	4.5	0.46	0	0	0	1	44.6
1	3	26.5	0	0	0.23	0	0	19.1
1	1	1.0	0	0.05	0.22	0	0	18.3
1	6	33.0	0.23	0.32	0	2.2	0	20.4
1	1	3.5	0.26	0.13	0.19	0.8	0	0.4
1	8	45.5	0.10	0	0.10	0	0	0.4
1	6	37.5	0	0.21	0.20	3.0	0	2.1
1	1	0.5	0.01	0	0.02	0	2	13.0
1	1	2.5	0	0	0.32	0	0	19.1
1	1	3.0	0	0	0.36	0	0	5.6
1	1	3.5	0.04	0	0.49	0	0	2.0
1	1	1.5	0.36	0	0.04	0	0	5.2
1	4	7.5	0.04	0	0.48	0	0	16.1
1	1	0.5	0.11	0.11	0.43	0	0	22.8
1	3	11.0	0.34	0.09	0.09	1.2	0	11.2
1	2	2.5	0.05	0	0	0	0	9.3
1	1	0.5	0.15	0.29	0	4.0	0	33.0
1	6	20.5	0.15	0.28	0.23	1.8	0	1.3
1	3	7.0	0.17	0.03	0.03	0.3	1	1.8
1	1	2.5	0	0	0.31	0	0	6.9
1	8	56.5	0.19	0.51	0.21	6.0	2	5.5
1	7	47.5	0	0.02	0.50	0	0	36.2

1	2	6.0	0	0	1.21	0	0	43.2
1	1	1.5	0.11	0	0.54	0	1	29.9
1	2	2.0	0.31	0	0.45	0	0	24.2
1	3	13.5	0.46	0	0	0	0	126
1	15	75.5	0	0.68	0.20	4.1	0	6.5
1	3	4.5	0.36	0.01	0.21	0	0	114
1	6	32.5	0	0.53	0.31	5.0	0	197
1	2	26.5	0	0.11	0.69	0	0	9.6
1	2	3.5	0.44	0.25	0.14	0.7	3	11.5
1	3	10.0	0.04	0.21	0.02	0.80	2	19.8
1	1	4.5	0	0.02	0.25	0	0	5.9
1	3	6.0	0.04	0.23	0.08	2.5	0	5.8
1	12	102.5	0.08	0.37	0.58	3.6	2	14.6
1	1	1.0	0.42	0.02	0	0	3	152
1	3	5.5	0.02	0	0	0	2	170
1	3	22.0	0.11	0.12	0.51	1.5	1	6.6
1	1	1	0.22	0.05	0.35	0	1	3.7
1	1	1	0	0	0.43	0	0	5.9
1	4	5.0	0.04	0	0	0	0	18.6
1	3	14.5	0	0	0.08	0	0	11.1
1	1	0.5	0.08	0	0.01	0	1	10.9
1	5	18.5	0.01	0	0.45	0	0	10.3
2	7	59.0	0.03	0.71	2.29	3.7	0	15.3
2	11	137.5	0.23	1.10	1.31	10.1	0	12.0
2	16	247.0	0.37	0.70	0.01	8.8	0	10.9
2	12	123.0	0	0.96	0.05	11.6	0	43.5
2	5	115.5	0	0.05	0.03	0.5	0	2.1
2	9	199.5	0.04	1.44	1.42	21.5	0	640
2	15	144.0	0.17	0.31	1.11	4.9	2	12.1
2	4	13.0	0.48	1.08	0.60	1.5	2	113
2	3	14.5	0.18	1.59	1.16	6.2	0	523
2	12	70.0	0.25	1.21	0.94	12.5	0	432

APPENDIX B

MULTIVARIATE ANALYSIS OF VARIANCE OUTPUT

GROUP 1ND. OF OBS. 42 67
 MEANS FOR GROUP 1
 3.57 17.57 0.23 0.25 0.48 2.36 1.07 0.15
 STANDARD DEVIATIONS
 2.54 23.62 0.36 0.41 0.47 4.74 3.02 0.26
 DISPERSION DETERMINANT= 0.3681E+00

GROUP 2ND. OF OBS. 12
 MEANS FOR GROUP 2
 10.83 143.12 0.23 0.63 1.00 5.54 0.42 0.54
 STANDARD DEVIATIONS
 6.07 111.85 0.31 0.43 1.08 6.63 0.67 0.34
 DISPERSION DETERMINANT= 0.2426E+02

MEANS FOR TOTAL SAMPLE
 5.19 45.47 0.23 0.33 0.59 3.07 0.93 0.24
 3.59 55.56 0.35 0.42 0.65 5.20 2.70 0.28

T MATRIX
 ROW= 1 0.12E+04 0.18E+05 0.45E+00 0.66E+02 0.22E+02
 0.75E+03 -0.70E+02 0.16E+02
 ROW= 2 0.31E+06 0.15E+02 0.91E+03 0.46E+03 0.99E+04
 -0.93E+03 0.31E+03
 ROW= 3 0.63E+01 0.21E+01 -0.61E+00 0.32E+02 0.33E+02
 -0.24E+00
 ROW= 4 0.10E+02 0.39E+01 0.11E+03 0.41E+01 0.84E+00
 ROW= 5 0.24E+02 0.25E+02 0.91E+00 0.61E+01
 ROW= 6 0.15E+04 -0.20E+01 -0.24E+01
 ROW= 7 0.38E+03 -0.43E+01
 ROW= 8 0.54E+01

A MATRIX
 ROW= 1 0.49E+03 0.85E+04 0.18E+00 0.26E+02 0.35E+02
 0.22E+03 -0.44E+02 0.26E+02
 ROW= 2 0.15E+06 0.31E+01 0.45E+03 0.61E+03 0.37E+04
 -0.77E+03 0.45E+03
 ROW= 3 0.64E-04 0.93E-02 0.13E-01 0.78E-01 -0.16E-01
 0.94E-02
 ROW= 4 0.13E+01 0.18E+01 0.11E+02 -0.23E+01 0.14E+01
 ROW= 5 0.25E+01 0.15E+02 -0.32E+01 0.19E+01
 ROW= 6 0.94E+02 -0.19E+02 0.11E+02
 ROW= 7 0.40E+01 -0.24E+01
 ROW= 8 0.14E+01

W MATRIX
 ROW= 1 0.67E+03 0.90E+04 0.28E+00 0.40E+02 -0.13E+02
 0.54E+03 -0.26E+02 -0.10E+02
 ROW= 2 0.16E+06 0.12E+02 0.47E+03 -0.15E+03 0.62E+04
 -0.16E+03 -0.14E+03
 ROW= 3 0.63E+01 0.21E+01 -0.62E+00 0.31E+02 0.33E+02
 -0.25E+00
 ROW= 4 0.90E+01 0.21E+01 0.10E+03 0.64E+01 -0.52E+00
 ROW= 5 0.22E+02 0.93E+01 0.41E+01 0.42E+01
 ROW= 6 0.14E+04 0.17E+02 -0.14E+02
 ROW= 7 0.38E+03 -0.20E+01
 ROW= 8 0.40E+01

DISPERSION DETERMINANT= 0.3418E+02
 FOR TEST OF H1(EQUALITY OF DISPERSIONS),M= 189.541 AND F= 3.717
 FOR F, NDF1= 36 AND NDF2= 1413

UNIVARIATE F-RATIOS, WITH NDF1= 1 AND NDF2= 52
 VARIABLE AMONG MEAN SQ WITHIN MEAN SQ F-RATIO ETA SQUARE
 1 492.20 12.88 38.20 .4235
 2 147127.87 3086.56 47.67 .4783
 3 0.00 0.12 0.00 .0000
 4 1.35 0.17 7.74 .1296
 5 2.54 0.42 6.08 .1047
 6 94.37 26.99 3.50 .0630
 7 4.00 7.26 0.55 .0105
 8 1.38 0.08 17.99 .2570

WILKS LAMBDA= 0.3805 GENERALIZED CORRELATION RATIO , ETA SQUARE=.6195
 F-RATIO FOR H2, OVERALL DISCRIMINATION,= 9.16
 NDF1= 8 AND NDF2= 45

APPENDIX C

CLASSIFICATION OUTPUT

VAR-COVAR MATRIX= 1

6.444	54.129	0.042	0.500	-0.117	5.893	-0.505	-0.150
54.129	557.958	0.829	4.883	-1.235	61.010	-3.908	-1.155
0.042	0.829	0.126	0.052	0.005	0.572	0.801	0.007
0.500	4.883	0.052	0.170	0.014	1.890	0.174	-0.001
-0.117	-1.237	0.006	0.014	0.216	0.139	0.128	0.539
5.893	61.010	0.572	1.890	0.139	22.440	0.900	-0.210
-0.505	-3.908	0.801	0.174	0.123	0.900	9.092	-0.004
-0.150	-1.155	0.007	-0.001	0.033	-0.210	-0.064	0.007

VAR-COVAR MATRIX= 2

36.879	616.250	-0.130	1.798	-0.757	26.708	-0.470	-0.380
616.250	2511.367	-1.983	24.092	-9.203	331.703	-0.148	-8.139
-0.130	-1.983	0.099	-0.002	-0.073	0.729	0.039	-0.058
1.798	24.092	-0.002	0.168	0.137	2.226	-0.063	0.630
-0.757	-9.203	-0.078	0.137	1.168	0.327	-0.104	0.241
26.708	331.703	0.729	2.226	0.327	43.946	-1.773	-0.475
-0.470	-0.148	0.039	-0.063	-0.104	-1.773	0.447	-0.016
-0.380	-8.139	-0.048	0.030	0.241	-0.475	-0.016	0.114

INVERSE VAR-COVAR MATRIX= 1

0.92E+00	-0.50E-01	0.12E+00	-0.25E+01	-0.95E-01	0.22E+00	0.32E-01	0.46E+00
-0.90E-01	0.11E-01	0.39E-02	0.31E+00	0.26E-01	-0.34E-01	-0.34E-02	-0.31E-01
0.12E+00	0.38E-02	0.27E+02	0.10E+02	0.24E+01	-0.16E+01	-0.24E+01	-0.71E+01
-0.25E+01	0.31E+00	0.10E+02	0.12E+03	-0.62E-02	-0.10E+02	-0.21E+01	0.19E+01
-0.97E-01	0.27E-01	0.24E+01	-0.15E-01	0.56E+01	-0.17E+00	-0.29E+00	-0.40E+01
0.22E+00	-0.34E-01	-0.16E+01	-0.10E+02	-0.17E+00	0.57E+00	0.24E+00	0.17E+00
0.33E-01	-0.34E-02	-0.24E+01	-0.21E+01	-0.29E+00	0.24E+00	0.35E+00	0.74E+00
0.46E+00	-0.31E-01	-0.71E+01	0.19E+01	-0.39E+01	0.16E+00	0.74E+00	0.20E+02

INVERSE VAR-COVAR MATRIX= 2

0.65E+00	-0.23E-01	0.29E+01	-0.21E+00	0.69E+00	-0.30E+00	-0.67E+00	-0.10E+01
-0.23E-01	0.99E-03	-0.87E-01	0.82E-02	-0.25E-01	0.10E-01	0.19E-01	0.56E-01
0.29E+01	-0.87E-01	0.36E+02	0.95E+01	0.28E+01	-0.26E+01	-0.83E+01	-0.16E+01
-0.21E+00	-0.82E-02	0.95E+01	0.33E+02	-0.42E+00	-0.20E+01	-0.48E+01	-0.16E+02
0.69E+00	-0.25E-01	0.28E+01	-0.42E+01	0.24E+01	-0.35E+00	-0.59E+00	-0.47E+01
-0.30E+00	0.10E-01	-0.26E+01	-0.20E+01	-0.35E+00	0.32E+00	0.88E+00	0.14E+01
-0.67E+00	0.19E-01	-0.83E+01	-0.48E+01	-0.59E+00	0.88E+00	0.50E+01	0.25E+01
-0.10E+01	0.56E-01	-0.16E+01	-0.14E+02	-0.47E+01	0.14E+01	0.25E+01	0.28E+02

ABSOLUTE VALUE OF THE DETERMINANT FOR GROUP 1= 0.36 ABSOLUTE VALUE OF THE DETERMINANT FOR GROUP 2= 24.30

OBSERVATION NO.= 1

3.00	44.50	0.00	0.01	1.54	0.10	0.00	0.22	
CHI-SQUARE VALUE, GROUP 1=	12.959		GROUP 2=	24.468				
OBSERVATION NO.= 2	3.00	6.50	0.75	0.13	0.31	0.90	0.00	0.24
CHI-SQUARE VALUE, GROUP 1=	10.461		GROUP 2=	18.284				
OBSERVATION NO.= 3	2.00	9.00	0.43	0.03	0.20	0.30	0.00	0.04
CHI-SQUARE VALUE, GROUP 1=	3.840		GROUP 2=	13.468				
OBSERVATION NO.= 4	5.00	14.50	0.05	0.26	0.00	3.20	0.00	0.06
CHI-SQUARE VALUE, GROUP 1=	4.893		GROUP 2=	14.053				
OBSERVATION NO.= 5	1.00	1.50	0.01	0.00	0.01	0.00	0.00	0.01
CHI-SQUARE VALUE, GROUP 1=	2.711		GROUP 2=	20.328				
OBSERVATION NO.= 6	3.00	14.00	0.10	0.95	0.40	7.50	4.00	0.06
CHI-SQUARE VALUE, GROUP 1=	12.808		GROUP 2=	123.833				
OBSERVATION NO.= 7	4.00	17.50	0.03	0.54	0.04	5.30	0.00	0.05
CHI-SQUARE VALUE, GROUP 1=	2.449		GROUP 2=	21.649				
OBSERVATION NO.= 8	1.00	1.00	0.00	0.00	0.47	0.00	0.00	0.06
CHI-SQUARE VALUE, GROUP 1=	1.699		GROUP 2=	19.055				
OBSERVATION NO.= 9	2.00	2.00	0.35	0.10	0.59	0.00	1.00	0.12

CHI-SQUARE VALUE, GROUP 1=	1.577	GROUP 2=	12.522				
OBSERVATION NO.= 10							
3.00 5.50	0.04	0.41	0.42	2.90	0.00	0.13	
CHI-SQUARE VALUE, GROUP 1=	1.865	GROUP 2=	17.131				
OBSERVATION NO.= 11							
7.00 28.00	0.34	0.28	0.00	2.60	2.00	0.09	
CHI-SQUARE VALUE, GROUP 1=	6.373	GROUP 2=	19.493				
OBSERVATION NO.= 12							
5.00 16.50	0.00	0.17	0.72	0.00	1.00	0.08	
CHI-SQUARE VALUE, GROUP 1=	3.109	GROUP 2=	17.155				
OBSERVATION NO.= 13							
1.00 2.50	0.42	0.00	1.21	0.00	1.00	1.65	
CHI-SQUARE VALUE, GROUP 1=	33.834	GROUP 2=	55.352				
OBSERVATION NO.= 14							
3.00 4.00	0.20	0.02	0.08	0.30	0.00	0.17	
CHI-SQUARE VALUE, GROUP 1=	1.886	GROUP 2=	11.253				
OBSERVATION NO.= 15							
4.00 16.50	0.30	0.00	0.66	0.00	0.00	0.19	
CHI-SQUARE VALUE, GROUP 1=	1.948	GROUP 2=	12.549				
OBSERVATION NO.= 16							
2.00 1.50	0.26	0.16	0.26	0.60	0.00	0.30	
CHI-SQUARE VALUE, GROUP 1=	1.812	GROUP 2=	11.032				
OBSERVATION NO.= 17							
1.00 1.00	0.28	0.00	0.46	0.00	0.00	0.16	
CHI-SQUARE VALUE, GROUP 1=	1.901	GROUP 2=	12.759				
OBSERVATION NO.= 18							
3.00 8.00	0.07	1.42	1.62	14.70	0.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	19.128	GROUP 2=	61.643				
OBSERVATION NO.= 19							
1.00 0.50	1.75	0.25	0.97	0.00	18.00	0.09	
CHI-SQUARE VALUE, GROUP 1=	33.371	GROUP 2=	1187.650				
OBSERVATION NO.= 20							
11.00 28.00	0.02	0.49	0.03	5.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	14.807	GROUP 2=	14.631				
OBSERVATION NO.= 21							
9.00 53.50	0.12	0.57	0.29	5.70	0.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	6.640	GROUP 2=	17.552				
OBSERVATION NO.= 22							
4.00 18.50	0.04	0.00	0.52	0.00	0.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	0.734	GROUP 2=	15.010				
OBSERVATION NO.= 23							
2.00 4.00	0.20	0.27	0.00	0.30	0.00	0.08	
CHI-SQUARE VALUE, GROUP 1=	6.684	GROUP 2=	16.322				
OBSERVATION NO.= 24							
3.00 10.00	0.12	0.00	1.19	0.50	1.00	0.06	
CHI-SQUARE VALUE, GROUP 1=	3.219	GROUP 2=	18.296				
OBSERVATION NO.= 25							
2.00 2.00	0.11	0.17	1.05	1.30	0.00	0.46	
CHI-SQUARE VALUE, GROUP 1=	2.142	GROUP 2=	10.997				
OBSERVATION NO.= 26							
4.00 15.00	0.25	0.07	0.38	0.30	0.00	0.42	
CHI-SQUARE VALUE, GROUP 1=	2.155	GROUP 2=	9.522				
OBSERVATION NO.= 27							
2.00 2.00	0.11	0.18	0.27	1.80	0.00	0.05	
CHI-SQUARE VALUE, GROUP 1=	0.679	GROUP 2=	16.591				
OBSERVATION NO.= 28							
2.00 1.50	0.26	0.00	0.00	0.00	0.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	2.756	GROUP 2=	13.574				
OBSERVATION NO.= 29							
3.00 5.50	0.08	0.14	0.63	1.50	1.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	1.172	GROUP 2=	19.794				
OBSERVATION NO.= 30							
4.00 9.50	0.10	0.11	0.25	0.00	4.00	0.05	
CHI-SQUARE VALUE, GROUP 1=	4.868	GROUP 2=	78.272				
OBSERVATION NO.= 31							
2.00 10.50	0.15	0.12	0.80	0.00	0.00	0.01	

CHI-SQUARE VALUE, GROUP 1 =	3.988	GROUP 2 =	16.843					
OBSERVATION NO. = 32								
7.00 58.50	0.25	0.53	0.66	3.50	0.00	0.09		
CHI-SQUARE VALUE, GROUP 1 =	11.008	GROUP 2 =	15.582					
OBSERVATION NO. = 33								
4.00 6.00	0.00	0.07	0.00	0.30	0.00	0.12		
CHI-SQUARE VALUE, GROUP 1 =	3.559	GROUP 2 =	13.415					
OBSERVATION NO. = 34								
1.00 4.00	0.00	0.25	0.14	3.00	0.00	0.24		
CHI-SQUARE VALUE, GROUP 1 =	4.044	GROUP 2 =	25.819					
OBSERVATION NO. = 35								
1.00 4.50	0.26	0.00	0.26	0.00	2.00	0.08		
CHI-SQUARE VALUE, GROUP 1 =	1.972	GROUP 2 =	30.742					
OBSERVATION NO. = 36								
2.00 4.50	0.03	0.14	1.67	3.80	0.00	0.11		
CHI-SQUARE VALUE, GROUP 1 =	14.554	GROUP 2 =	21.492					
OBSERVATION NO. = 37								
10.00 88.50	1.49	2.15	0.29	26.50	4.00	0.03		
CHI-SQUARE VALUE, GROUP 1 =	32.609	GROUP 2 =	119.396					
OBSERVATION NO. = 38								
7.00 65.50	0.00	0.19	0.14	3.80	0.00	0.06		
CHI-SQUARE VALUE, GROUP 1 =	8.855	GROUP 2 =	13.588					
OBSERVATION NO. = 39								
3.00 40.50	0.00	0.14	0.15	3.00	0.00	0.05		
CHI-SQUARE VALUE, GROUP 1 =	9.704	GROUP 2 =	22.055					
OBSERVATION NO. = 40								
3.00 27.50	0.00	0.00	0.67	0.00	0.00	0.24		
CHI-SQUARE VALUE, GROUP 1 =	3.129	GROUP 2 =	13.149					
OBSERVATION NO. = 41								
1.00 2.00	0.20	0.00	0.10	0.00	0.00	0.24		
CHI-SQUARE VALUE, GROUP 1 =	1.959	GROUP 2 =	13.776					
OBSERVATION NO. = 42								
4.00 12.00	0.45	0.00	0.06	0.00	6.00	0.11		
CHI-SQUARE VALUE, GROUP 1 =	7.559	GROUP 2 =	145.603					
OBSERVATION NO. = 43								
1.00 2.00	0.57	0.05	1.74	0.00	1.00	0.05		
CHI-SQUARE VALUE, GROUP 1 =	17.636	GROUP 2 =	18.224					
OBSERVATION NO. = 44								
4.00 11.50	0.99	0.00	0.39	0.00	1.00	0.01		
CHI-SQUARE VALUE, GROUP 1 =	20.404	GROUP 2 =	26.239					
OBSERVATION NO. = 45								
7.00 51.00	0.19	0.00	0.05	0.00	0.00	0.00		
CHI-SQUARE VALUE, GROUP 1 =	4.683	GROUP 2 =	16.361					
OBSERVATION NO. = 46								
3.00 13.50	0.05	0.15	0.91	1.70	0.00	0.05		
CHI-SQUARE VALUE, GROUP 1 =	1.505	GROUP 2 =	16.286					
OBSERVATION NO. = 47								
13.00 84.50	0.00	0.80	0.44	5.80	0.00	0.12		
CHI-SQUARE VALUE, GROUP 1 =	22.496	GROUP 2 =	24.684					
OBSERVATION NO. = 48								
1.00 6.00	0.08	0.14	0.00	0.80	0.00	0.00		
CHI-SQUARE VALUE, GROUP 1 =	3.540	GROUP 2 =	21.005					
OBSERVATION NO. = 49								
3.00 12.00	1.45	0.86	0.11	6.00	2.00	0.01		
CHI-SQUARE VALUE, GROUP 1 =	33.451	GROUP 2 =	34.262					
OBSERVATION NO. = 50								
1.00 5.00	0.55	0.04	0.97	0.50	5.00	0.04		
CHI-SQUARE VALUE, GROUP 1 =	5.629	GROUP 2 =	106.515					
OBSERVATION NO. = 51								
1.00 1.50	0.32	0.08	0.49	0.00	0.00	0.05		
CHI-SQUARE VALUE, GROUP 1 =	3.904	GROUP 2 =	14.921					
OBSERVATION NO. = 52								
1.00 0.50	0.00	0.06	0.22	0.00	0.00	0.00		
CHI-SQUARE VALUE, GROUP 1 =	2.000	GROUP 2 =	20.507					
OBSERVATION NO. = 53								
2.00 2.00	0.09	0.16	0.56	6.70	0.00	0.01		

CHI-SQUARE VALUE, GROUP 1=	1.165	GROUP 2=	17.337				
OBSERVATION NO.= 54							
2.00 6.50	0.34	0.00	1.12	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	7.197	GROUP 2=	17.003				
OBSERVATION NO.= 55							
1.00 2.00	0.68	0.00	0.04	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	11.287	GROUP 2=	17.221				
OBSERVATION NO.= 56							
2.00 5.00	0.00	0.00	1.06	0.00	0.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	3.482	GROUP 2=	18.723				
OBSERVATION NO.= 57							
5.00 46.50	0.04	0.00	0.09	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	5.090	GROUP 2=	13.932				
OBSERVATION NO.= 58							
2.00 7.00	0.00	0.00	0.00	0.00	0.00	0.00	
CHI-SQUARE VALUE, GROUP 1=	1.844	GROUP 2=	18.081				
OBSERVATION NO.= 59							
4.00 29.50	0.00	0.00	1.04	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	4.813	GROUP 2=	17.790				
OBSERVATION NO.= 60							
8.00 67.50	0.02	0.63	0.76	5.40	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	8.824	GROUP 2=	17.747				
OBSERVATION NO.= 61							
2.00 6.00	0.76	0.00	0.19	0.00	4.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	4.889	GROUP 2=	54.272				
OBSERVATION NO.= 62							
2.00 4.00	0.00	0.19	0.16	0.00	3.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	4.863	GROUP 2=	61.696				
OBSERVATION NO.= 63							
1.00 2.00	0.09	0.14	0.88	1.50	0.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	3.110	GROUP 2=	19.715				
OBSERVATION NO.= 64							
9.00 40.50	0.84	0.56	0.85	5.50	1.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	21.248	GROUP 2=	31.279				
OBSERVATION NO.= 65							
2.00 3.00	0.37	0.48	0.16	3.50	3.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	2.850	GROUP 2=	53.557				
OBSERVATION NO.= 65							
1.00 2.00	1.37	0.00	0.26	0.00	4.00	0.05	
CHI-SQUARE VALUE, GROUP 1=	24.644	GROUP 2=	46.305				
OBSERVATION NO.= 67							
1.00 0.50	0.00	0.00	0.36	0.00	2.00	0.06	
CHI-SQUARE VALUE, GROUP 1=	3.422	GROUP 2=	43.478				
OBSERVATION NO.= 68							
1.00 2.00	0.10	0.00	0.07	0.00	1.00	0.06	
CHI-SQUARE VALUE, GROUP 1=	2.310	GROUP 2=	23.120				
OBSERVATION NO.= 69							
2.00 6.50	0.07	0.00	0.03	0.00	0.00	0.05	
CHI-SQUARE VALUE, GROUP 1=	1.317	GROUP 2=	15.667				
OBSERVATION NO.= 70							
1.00 3.00	0.01	0.00	0.02	0.00	0.00	0.05	
CHI-SQUARE VALUE, GROUP 1=	2.727	GROUP 2=	19.957				
OBSERVATION NO.= 71							
1.00 3.50	0.02	0.00	0.20	0.00	0.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	2.232	GROUP 2=	19.497				
OBSERVATION NO.= 72							
3.00 14.50	0.01	0.06	0.00	0.00	0.00	0.16	
CHI-SQUARE VALUE, GROUP 1=	1.289	GROUP 2=	13.885				
OBSERVATION NO.= 73							
6.00 29.50	0.00	0.00	1.83	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	15.296	GROUP 2=	28.625				
OBSERVATION NO.= 74							
1.00 0.50	0.00	0.10	1.88	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	14.735	GROUP 2=	24.206				
OBSERVATION NO.= 75							
2.00 4.00	0.15	0.27	1.06	2.50	1.00	0.03	

CHI-SQUARE VALUE, GROUP 1=	2.502	GROUP 2=	22.298				
OBSERVATION NO.= 76							
12.00 108.00	2.36	0.84	0.53	6.90	0.00	0.05	
CHI-SQUARE VALUE, GROUP 1=	161.253	GROUP 2=	216.094				
OBSERVATION NO.= 77							
2.00 6.00	0.39	0.34	0.49	2.80	1.00	0.38	
CHI-SQUARE VALUE, GROUP 1=	1.659	GROUP 2=	14.215				
OBSERVATION NO.= 78							
2.00 4.50	0.46	0.00	0.00	0.00	1.00	0.06	
CHI-SQUARE VALUE, GROUP 1=	3.066	GROUP 2=	12.095				
OBSERVATION NO.= 79							
3.00 26.50	0.00	0.00	0.23	0.00	0.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	2.830	GROUP 2=	16.167				
OBSERVATION NO.= 80							
1.00 1.00	0.00	0.05	0.22	0.00	0.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	1.905	GROUP 2=	20.068				
OBSERVATION NO.= 81							
6.00 33.00	0.23	0.32	0.00	2.20	0.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	2.940	GROUP 2=	14.406				
OBSERVATION NO.= 82							
1.00 3.50	0.26	0.13	0.19	0.80	0.00	0.00	
CHI-SQUARE VALUE, GROUP 1=	3.215	GROUP 2=	16.914				
OBSERVATION NO.= 83							
8.00 45.50	0.10	0.00	0.10	0.00	0.00	0.00	
CHI-SQUARE VALUE, GROUP 1=	6.017	GROUP 2=	20.131				
OBSERVATION NO.= 84							
6.00 37.50	0.00	0.21	0.20	3.00	0.00	0.00	
CHI-SQUARE VALUE, GROUP 1=	2.398	GROUP 2=	14.941				
OBSERVATION NO.= 85							
1.00 0.50	0.01	0.00	0.02	0.00	2.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	4.526	GROUP 2=	44.587				
OBSERVATION NO.= 86							
1.00 2.50	0.00	0.00	0.32	0.00	0.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	2.039	GROUP 2=	19.856				
OBSERVATION NO.= 87							
1.00 3.00	0.00	0.00	0.36	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	2.198	GROUP 2=	20.184				
OBSERVATION NO.= 88							
1.00 3.50	0.04	0.00	0.49	0.00	0.00	0.00	
CHI-SQUARE VALUE, GROUP 1=	2.306	GROUP 2=	19.194				
OBSERVATION NO.= 89							
1.00 1.50	0.36	0.00	0.04	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	4.151	GROUP 2=	14.701				
OBSERVATION NO.= 90							
4.00 7.50	0.04	0.00	0.48	0.00	0.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	2.508	GROUP 2=	15.758				
OBSERVATION NO.= 91							
1.00 0.50	0.11	0.11	0.43	0.30	0.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	2.392	GROUP 2=	17.698				
OBSERVATION NO.= 92							
3.00 11.00	0.34	0.09	0.09	1.20	0.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	2.145	GROUP 2=	13.101				
OBSERVATION NO.= 93							
2.00 2.50	0.05	0.00	0.00	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	1.687	GROUP 2=	16.565				
OBSERVATION NO.= 94							
1.00 0.50	0.15	0.29	0.00	4.00	0.00	0.05	
CHI-SQUARE VALUE, GROUP 1=	4.177	GROUP 2=	24.594				
OBSERVATION NO.= 95							
6.00 20.50	0.15	0.28	0.23	1.80	0.00	0.00	
CHI-SQUARE VALUE, GROUP 1=	4.122	GROUP 2=	16.700				
OBSERVATION NO.= 96							
3.00 7.00	0.17	0.03	0.08	0.30	1.00	0.00	
CHI-SQUARE VALUE, GROUP 1=	1.302	GROUP 2=	16.254				
OBSERVATION NO.= 97							
1.00 2.50	0.00	0.00	0.31	0.00	0.00	0.01	

CHI-SQUARE VALUE, GROUP 1=	2.142	GROUP 2=	20.168				
OBSERVATION NO.= 98							
8.00 56.50	0.19	0.51	0.21	6.00	2.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	4.356	GROUP 2=	28.052				
OBSERVATION NO.= 99							
7.00 47.50	0.00	0.02	0.50	0.00	0.00	0.05	
CHI-SQUARE VALUE, GROUP 1=	3.728	GROUP 2=	16.848				
OBSERVATION NO.=100							
2.00 6.00	0.00	0.00	1.21	0.00	0.00	0.06	
CHI-SQUARE VALUE, GROUP 1=	4.344	GROUP 2=	18.181				
OBSERVATION NO.=101							
1.00 1.50	0.11	0.00	0.54	0.00	1.00	0.04	
CHI-SQUARE VALUE, GROUP 1=	1.718	GROUP 2=	21.872				
OBSERVATION NO.=102							
2.00 2.00	0.31	0.00	0.45	0.00	0.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	2.374	GROUP 2=	13.881				
OBSERVATION NO.=103							
3.00 13.50	0.46	0.00	0.00	0.00	0.00	0.18	
CHI-SQUARE VALUE, GROUP 1=	3.744	GROUP 2=	12.027				
OBSERVATION NO.=104							
15.00 75.50	0.00	0.68	0.20	4.10	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	39.428	GROUP 2=	47.558				
OBSERVATION NO.=105							
3.00 4.50	0.36	0.01	0.21	0.00	0.00	0.16	
CHI-SQUARE VALUE, GROUP 1=	2.618	GROUP 2=	11.844				
OBSERVATION NO.=106							
6.00 32.50	0.00	0.53	0.31	5.00	0.00	0.28	
CHI-SQUARE VALUE, GROUP 1=	3.366	GROUP 2=	13.002				
OBSERVATION NO.=107							
2.00 26.50	0.00	0.11	0.69	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	9.013	GROUP 2=	19.256				
OBSERVATION NO.=108							
2.00 3.50	0.44	0.25	0.14	0.70	3.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	3.053	GROUP 2=	39.858				
OBSERVATION NO.=109							
3.00 10.00	0.04	0.21	0.02	0.80	2.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	2.642	GROUP 2=	34.298				
OBSERVATION NO.=110							
1.00 4.50	0.00	0.02	0.25	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	2.435	GROUP 2=	20.503				
OBSERVATION NO.=111							
3.00 6.00	0.04	0.23	0.08	2.50	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	1.780	GROUP 2=	17.849				
OBSERVATION NO.=112							
12.00 102.50	0.08	0.37	0.58	3.60	2.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	18.835	GROUP 2=	26.774				
OBSERVATION NO.=113							
1.00 1.00	0.42	0.02	0.00	0.00	3.00	0.21	
CHI-SQUARE VALUE, GROUP 1=	2.704	GROUP 2=	46.266				
OBSERVATION NO.=114							
3.00 5.50	0.02	0.00	0.00	0.00	2.00	0.24	
CHI-SQUARE VALUE, GROUP 1=	5.276	GROUP 2=	34.194				
OBSERVATION NO.=115							
3.00 22.00	0.11	0.12	0.51	1.50	1.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	1.416	GROUP 2=	20.095				
OBSERVATION NO.=116							
1.00 1.00	0.22	0.05	0.35	0.00	1.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	1.689	GROUP 2=	18.928				
OBSERVATION NO.=117							
1.00 1.00	0.00	0.00	0.43	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	1.968	GROUP 2=	19.965				
OBSERVATION NO.=118							
4.00 5.00	0.04	0.00	0.00	0.00	0.00	0.03	
CHI-SQUARE VALUE, GROUP 1=	4.205	GROUP 2=	14.633				
OBSERVATION NO.=119							
3.00 14.50	0.00	0.00	0.08	0.00	0.00	0.02	

CHI-SQUARE VALUE, GROUP 1=	1.098	GROUP 2=	16.020				
OBSERVATION NO.=120							
1.00 0.50	0.08	0.00	0.01	0.00	1.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	2.682	GROUP 2=	24.390				
OBSERVATION NO.=121							
5.00 18.50	0.01	0.00	0.45	0.00	0.00	0.01	
CHI-SQUARE VALUE, GROUP 1=	2.316	GROUP 2=	16.384				
OBSERVATION NO.=122							
16.00 205.00	0.03	1.04	1.29	6.60	0.00	1.14	
CHI-SQUARE VALUE, GROUP 1=	186.802	GROUP 2=	13.918				
OBSERVATION NO.=123							
9.00 123.00	0.12	0.95	4.18	9.50	0.00	1.04	
CHI-SQUARE VALUE, GROUP 1=	127.397	GROUP 2=	16.099				
OBSERVATION NO.=124							
11.00 146.50	0.01	0.90	1.20	3.20	0.00	0.50	
CHI-SQUARE VALUE, GROUP 1=	138.755	GROUP 2=	13.972				
OBSERVATION NO.=125							
5.00 27.00	1.06	0.52	0.48	11.10	0.00	0.30	
CHI-SQUARE VALUE, GROUP 1=	37.763	GROUP 2=	15.630				
OBSERVATION NO.=126							
25.00 323.00	0.38	1.50	0.16	23.20	0.00	0.08	
CHI-SQUARE VALUE, GROUP 1=	241.886	GROUP 2=	15.561				
OBSERVATION NO.=127							
5.00 20.00	0.00	0.00	0.28	0.00	0.00	0.66	
CHI-SQUARE VALUE, GROUP 1=	8.644	GROUP 2=	12.273				
OBSERVATION NO.=128							
8.00 90.50	0.49	0.30	0.43	0.30	2.00	0.32	
CHI-SQUARE VALUE, GROUP 1=	36.295	GROUP 2=	13.075				
OBSERVATION NO.=129							
4.00 22.00	0.12	0.79	1.08	1.80	1.00	0.82	
CHI-SQUARE VALUE, GROUP 1=	49.438	GROUP 2=	14.293				
OBSERVATION NO.=130							
6.00 20.50	0.01	0.11	0.84	0.50	0.00	0.32	
CHI-SQUARE VALUE, GROUP 1=	5.034	GROUP 2=	12.684				
OBSERVATION NO.=131							
12.00 234.00	0.07	0.38	0.14	4.50	0.00	0.23	
CHI-SQUARE VALUE, GROUP 1=	262.328	GROUP 2=	12.803				
OBSERVATION NO.=132							
14.00 306.50	0.43	0.34	0.87	1.20	1.00	0.32	
CHI-SQUARE VALUE, GROUP 1=	557.989	GROUP 2=	13.304				
OBSERVATION NO.=133							
15.00 204.50	0.06	0.73	1.05	4.60	1.00	0.73	
CHI-SQUARE VALUE, GROUP 1=	168.368	GROUP 2=	8.999				
OBSERVATION NO.=134							
7.00 59.00	0.03	0.71	2.29	3.70	0.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	42.586	GROUP 2=	32.197				
OBSERVATION NO.=135							
11.00 137.50	0.23	1.10	1.31	10.10	0.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	64.493	GROUP 2=	20.913				
OBSERVATION NO.=136							
16.00 247.00	0.37	0.70	0.01	8.80	0.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	205.888	GROUP 2=	10.953				
OBSERVATION NO.=137							
12.00 123.00	0.00	0.96	0.05	11.60	0.00	0.06	
CHI-SQUARE VALUE, GROUP 1=	26.279	GROUP 2=	17.879				
OBSERVATION NO.=138							
5.00 115.50	0.00	0.05	0.03	0.50	0.00	0.00	
CHI-SQUARE VALUE, GROUP 1=	87.463	GROUP 2=	19.435				
OBSERVATION NO.=139							
9.00 199.50	0.04	1.44	1.42	21.50	0.00	0.90	
CHI-SQUARE VALUE, GROUP 1=	215.266	GROUP 2=	118.227				
OBSERVATION NO.=140							
15.00 144.00	0.17	0.31	1.11	4.90	2.00	0.02	
CHI-SQUARE VALUE, GROUP 1=	44.924	GROUP 2=	34.853				
OBSERVATION NO.=141							
4.00 13.00	0.48	1.08	0.60	1.50	2.00	0.16	

CHI-SQUARE VALUE, GROUP 1=	93.538	GROUP 2=	36.764				
OBSERVATION NO.=142							
3.00	14.50	0.18	1.59	1.16	6.20	0.00	0.73
CHI-SQUARE VALUE, GROUP 1=	132.204	GROUP 2=	46.685				
OBSERVATION NO.=143							
12.00	70.00	0.25	1.21	0.94	12.50	0.00	0.60
CHI-SQUARE VALUE, GROUP 1=	23.842	GROUP 2=	11.716				
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
1	0.99684	0.00316					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
2	0.98038	0.01962					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
3	0.99195	0.00805					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
4	0.98985	0.01015					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
5	0.99985	0.00015					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
6	1.00000	0.00000					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
7	0.99993	0.00007					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
8	0.99983	0.00017					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
9	0.99582	0.00418					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
10	0.99952	0.00048					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
11	0.99859	0.00141					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
12	0.99911	0.00089					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
13	0.99998	0.00002					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
14	0.99984	0.00016					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
15	0.99504	0.00496					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
16	0.99014	0.00985					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
17	0.99563	0.00437					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
18	1.00000	0.00000					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
19	1.00000	0.0					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
20	0.47802	0.52198					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
21	0.99575	0.00425					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
22	0.99921	0.00079					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
23	0.99199	0.00801					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
24	0.99947	0.00053					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
25	0.98820	0.01180					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
26	0.97543	0.02457					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
27	0.99965	0.00035					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
28	0.99653	0.00347					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					
29	0.99991	0.00009					
OBSERVATION PROB. OF GROUP 1		PROB. OF GROUP 2					

20	1.00000	0.00000
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
31	0.99339	0.00161
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
32	0.90781	0.09219
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
33	0.99281	0.00719
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
34	0.99998	0.00002
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
35	1.00000	0.00000
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
36	0.66980	0.33020
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
37	1.00000	0.00000
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
38	0.91426	0.08574
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
39	0.99792	0.00208
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
40	0.99337	0.00663
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
41	0.69729	0.00271
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
42	1.00000	0.00000
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
43	0.57301	0.42700
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
44	0.94870	0.05130
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
45	0.99710	0.00290
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
46	0.99938	0.00062
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
47	0.74915	0.25085
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
48	0.99984	0.00016
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
49	0.60001	0.39999
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
50	1.00000	0.00000
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
51	0.99596	0.00404
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
52	0.99990	0.00010
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
53	0.99969	0.00031
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
54	0.99263	0.00737
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
55	0.95106	0.04894
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
56	0.99951	0.00049
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
57	0.98813	0.01188
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
58	0.99970	0.00030
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
59	0.99843	0.00157
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
60	0.98856	0.01144
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
61	1.00000	0.00000
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
62	1.00000	0.00000
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2

63	0.99975	0.00025
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
64	0.99341	0.00659
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
65	1.00000	0.00000
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
66	0.99998	0.00002
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
67	1.00000	0.00000
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
68	0.99997	0.00003
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
69	0.99924	0.00076
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
70	0.99982	0.00018
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
71	0.99982	0.00018
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
72	0.99816	0.00184
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
73	0.99873	0.00127
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
74	0.99130	0.00870
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
75	0.99995	0.00005
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
76	1.00000	0.00000
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
77	0.99813	0.00187
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
78	0.98917	0.01083
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
79	0.99873	0.00127
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
80	0.99989	0.00011
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
81	0.99677	0.00323
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
82	0.99894	0.00106
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
83	0.99914	0.00086
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
84	0.99811	0.00189
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
85	1.00000	0.00000
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
86	0.99986	0.00014
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
87	0.99988	0.00012
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
88	0.99979	0.00021
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
89	0.99491	0.00509
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
90	0.99867	0.00133
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
91	0.99953	0.00047
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
92	0.99584	0.00416
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
93	0.99941	0.00059
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
94	0.99996	0.00004
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2
95	0.99815	0.00185
OBSERVATION	PRCB. OF GROUP 1	PROB. OF GROUP 2

96	0.99943	0.00057		129	0.00000	1.00000	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
97	0.99988	0.00012		130	0.97865	0.02135	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
98	0.99999	0.00001		131	0.00000	1.00000	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
99	0.99859	0.00141		132	0.0	1.00000	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
100	0.99901	0.00099		133	0.00000	1.00000	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
101	0.99996	0.00004		134	0.00552	0.99448	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
102	0.99684	0.00316		135	0.00000	1.00000	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
103	0.98435	0.01565		136	0.00000	1.00000	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
104	0.98312	0.01688		137	0.01477	0.98523	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
105	0.99018	0.00982		138	0.00000	1.00000	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
106	0.99198	0.00802		139	0.00000	1.00000	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
107	0.99407	0.00593		140	0.00646	0.99354	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
108	1.00000	0.00000		141	0.00000	1.00000	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
109	1.00000	0.00000		142	0.00000	1.00000	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2	
110	0.99988	0.00012		143	0.00232	0.99768	
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2		"			
111	0.99968	0.00032					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
112	0.98147	0.01853					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
113	1.00000	0.00000					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
114	1.00000	0.00000					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
115	0.99991	0.00009					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
116	0.99982	0.00018					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
117	0.99988	0.00012					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
118	0.99459	0.00541					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
119	0.99943	0.00057					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
120	0.99998	0.00002					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
121	0.99912	0.00088					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
122	0.00000	1.00000					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
123	0.00000	1.00000					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
124	0.00000	1.00000					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
125	0.00002	0.99998					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
126	0.00000	1.00000					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
127	0.85991	0.14009					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					
128	0.00001	0.99999					
OBSERVATION	PRCB. OF GROUP 1	PRCB. OF GROUP 2					

APPENDIX D

STATISTICAL ANALYSIS SYSTEM-DISCRIMINANT

ANALYSIS PROGRAM OUTPUT

DATA FRT;
INPUT GROUP 1 X1 2-6 X2 7-12 X3 13-18 X4 19-24 X5 25-30 X6 31-36
X7 37-42 X8 43-48;
CARDS

54 OBSERVATIONS IN DATA SET FRT 9 VARIABLES

PROC DISCRIM S WCOV WCOVR PCOV PCORR LIST PROP POOL=TEST SLPOOL=0.05;
CLASSES GROUP;
TITLE 'DISCRIMINANT ANALYSIS';

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS

GROUP	FREQUENCY	PRIOR PROBABILITY
1	42	0.77777778
2	12	0.22222222
-----	---	-----
TOTAL	54	1.00000000

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS SIMPLE STATISTICS

VARIABLE	N	SUM	GROUP = 1		
			MEAN	VARIANCE	STANDARD DEVIATION
X1	42	150.0000000	3.57142857	6.44599303	2.53889603
X2	42	738.0000000	17.57142857	557.95818815	23.62113659
X3	42	9.62000000	0.22904762	0.12620803	0.35525682
X4	42	10.51000000	0.25023810	0.16395848	0.41226021
X5	42	20.09000000	0.47833333	0.21623374	0.46500940
X6	42	99.20000000	2.36190476	22.43948897	4.73703377
X7	42	45.00000000	1.07142857	9.09233449	3.01534981
X8	42	6.37000000	0.15166667	0.06712154	0.25937826

			GROUP = 2		
X1	12	130.0000000	10.83333333	36.87878788	6.07279076
X2	12	1717.5000000	143.12500000	12511.36931818	111.35423246
X3	12	2.78000000	0.23166667	0.09835152	0.31443661
X4	12	7.56000000	0.63000000	0.18825455	0.43306310
X5	12	12.00000000	1.00000000	1.16739091	1.35067002
X6	12	66.50000000	5.54166667	43.94626788	6.62919964
X7	12	5.00000000	0.41666667	0.44696970	0.66855792
X8	12	6.45000000	0.53833333	0.11372424	0.33723025

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS		WITHIN COVARIANCE MATRICES				
DF = 41		GROUP = 1				
VARIABLE	X1	X2	X3	X4	X5	X6
X1	6.44599303	54.12891986	0.04153310	0.50034843	-0.11707317	5.89790941
	X7	X8				
	-0.50522648	-0.15073171				

X2	54.12891986	557.95818815	0.82872822	4.88839721	-1.23646341	61.01010453
	X7	X8				
	-3.90766551	-1.15817073				

X3	0.04153310	0.82372822	0.12620383	0.05238560	0.00586423	0.57159698
	X7	X8				
	0.80055749	0.00698943				

X4	0.50034843	4.88839721	0.05238560	0.16995348	0.01360772	1.89015563
	X7	X8				
	0.17412892	-0.02078089				

X5	-0.11707317	-1.23646341	0.00586423	0.01360772	0.21623374	0.13920325
	X7	X8				
	0.12768293	0.03783943				

X6	5.89790941	61.01010453	0.57159698	1.89015563	0.13920325	22.43545897
	X7	X8				
	0.90034843	-0.21113008				

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS WITHIN COVARIANCE MATRICES

DF = 41

GROUP = 1

VARIABLE	X1	X2	X3	X4	X5	X6
X7	-0.50522648	-3.90766551	0.80055749	0.17412892	0.12768293	0.90034843
	X7	X8				
	9.09233449	-0.04158537				

	X1	X2	X3	X4	X5	X6
X8	-0.15073171	-1.15817073	0.00698943	-0.02078089	0.03783943	-0.21113008
	X7	X8				
	-0.04158537	0.06712154				

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS		WITHIN COVARIANCE MATRICES					
DF = 11		GROUP = 2					
VARIABLE	X1	X2	X3	X4	X5	X6	
X1	36.87878780	616.25000000	-0.12969697	1.79818182	-0.75727273	26.70757576	
	X7	X8					
	-0.46369397	-0.37575758					

X2	616.25000000	12511.36931818	-1.98340909	24.09227273	-9.20318182	331.70795455	
	X7	X8					
	-0.14772727	-8.12113636					

X3	-0.12969697	-1.98340909	0.09885152	-0.00150000	-0.07840909	0.72919697	
	X7	X8					
	0.03924242	-0.04802424					

X4	1.79818182	24.09227273	-0.00150000	0.18825455	0.13748182	2.22627273	
	X7	X8					
	-0.06272727	0.03071818					

X5	-0.75727273	-9.20318182	-0.07840909	0.13748182	1.16789091	0.32736364	
	X7	X8					
	-0.10363636	0.24259091					

X6	26.70757576	331.70795455	0.72919697	2.22627273	0.32736364	43.94628785	
	X7	X8					
	-1.77343485	-0.46365152					

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS		WITHIN COVARIANCE MATRICES				
DF = 11		GROUP = 2				
VARIABLE	X1	X2	X3	X4	X5	X6
X7	-0.46969697	-0.14772727	0.03924242	-0.06272727	-0.10362636	-1.77346465
	X7	X8				
	0.44696970	-0.01651515				

	X1	X2	X3	X4	X5	X6
X8	-0.27575758	-8.12113636	-0.04802424	0.03071518	0.24259091	-0.46365151
	X7	X8				
	-0.01651515	0.11372424				

DISCRIMINANT ANALYSIS

VARIABLE	DISCRIMINANT ANALYSIS							
	WITHIN CORRELATION COEFFICIENTS / PROBABILITY > R							
	GROUP = 1							
	X1	X2	X3	X4	X5	X6	X7	X8
X1	1.000000 0.0000	0.902576 0.0001	0.046047 0.7692	0.476031 0.0017	-0.099163 0.5390	0.490396 0.0013	-0.065994 0.6811	-0.229155 0.1407
X2	0.902576 0.0001	1.000000 0.0000	0.098757 0.5407	0.501989 0.0010	-0.112569 0.5154	0.545248 0.0004	-0.054363 0.7299	-0.189252 0.2280
X3	0.046047 0.7692	0.098757 0.5407	1.000000 0.0000	0.357681 0.0190	0.035498 0.8180	0.339655 0.0262	0.747326 0.0001	0.075939 0.6379
X4	0.476031 0.0017	0.501989 0.0010	0.357681 0.0190	1.000000 0.0000	0.070933 0.6594	0.967876 0.0001	0.140075 0.6202	-0.194564 0.2146
X5	-0.099163 0.5390	-0.112569 0.5154	0.035498 0.8180	0.070933 0.6594	1.000000 0.0000	0.063195 0.6933	0.091061 0.5730	0.314088 0.0404
X6	0.490396 0.0013	0.545248 0.0004	0.339655 0.0262	0.967876 0.0001	0.063195 0.6933	1.000000 0.0000	0.063033 0.6940	-0.172033 0.2755
X7	-0.065994 0.6811	-0.054363 0.7299	0.747326 0.0001	0.140075 0.6202	0.091061 0.5730	0.063033 0.6940	1.000000 0.0000	-0.053232 0.7371
X8	-0.229155 0.1407	-0.189252 0.2280	0.075939 0.6379	-0.194564 0.2146	0.314088 0.0404	-0.172033 0.2755	-0.053232 0.7371	1.000000 0.0000

DISCRIMINANT ANALYSIS

VARIABLE	DISCRIMINANT ANALYSIS							
	WITHIN CORRELATION COEFFICIENTS / PROBABILITY > R							
	GROUP = 2							
	X1	X2	X3	X4	X5	X6	X7	X8
X1	1.000000 0.0000	0.907227 0.0001	-0.067928 0.8277	0.682453 0.0140	-0.115389 0.7205	0.663415 0.0180	-0.115689 0.7198	-0.183482 0.5736
X2	0.907227 0.0001	1.000000 0.0000	-0.056399 0.8557	0.496424 0.0980	-0.076135 0.8084	0.447345 0.1422	-0.001975 0.9910	-0.215297 0.5073
X3	-0.067928 0.8277	-0.056399 0.8557	1.000000 0.0000	-0.010996 0.9717	-0.230767 0.5238	0.349858 0.2644	0.186692 0.5668	-0.452942 0.1366
X4	0.682453 0.0140	0.496424 0.0980	-0.010996 0.9717	1.000000 0.0000	0.293205 0.6429	0.774006 0.0034	-0.216244 0.5054	0.200941 0.5182
X5	-0.115389 0.7205	-0.076135 0.8084	-0.230767 0.5238	0.293205 0.6429	1.000000 0.0000	0.045695 0.8825	-0.143441 0.6595	0.665651 0.0175
X6	0.663415 0.0180	0.447345 0.1422	0.349858 0.2644	0.774006 0.0034	0.045695 0.8825	1.000000 0.0000	-0.400154 0.1955	-0.207393 0.5235
X7	-0.115689 0.7198	-0.001975 0.9910	0.186692 0.5668	-0.216244 0.5054	-0.143441 0.6595	-0.400154 0.1955	1.000000 0.0000	-0.073252 0.8152
X8	-0.183482 0.5736	-0.215297 0.5073	-0.452942 0.1366	0.200941 0.5182	0.665651 0.0175	-0.207398 0.5235	-0.073252 0.8152	1.000000 0.0000

DISCRIMINANT ANALYSIS

DF = 52

VARIABLE	DISCRIMINANT ANALYSIS			POOLED COVARIANCE MATRIX		
	X1	X2	X3	X4	X5	X6
X1	12.98369963	173.03914835	0.00531136	0.77489011	-0.25250000	10.29995421
	X7	X8				
	-0.49771062	-0.19333333				
X2	173.03914835	3086.56438874	0.23385302	8.95075549	-2.92173077	118.27311126
	X7	X8				
	-3.11229396	-2.63110577				
X3	0.00531136	0.23385302	0.12042170	0.04098672	-0.01196282	0.60493544
	X7	X8				
	0.63351007	-0.00464808				
X4	0.77489011	8.95075549	0.04098672	0.17382880	0.03981186	1.96125733
	X7	X8				
	0.12402473	-0.00988636				
X5	-0.25250000	-2.92173077	-0.01196282	0.03981186	0.41754583	0.17900641
	X7	X8				
	0.07875000	0.08115224				
X6	10.29995421	118.27311126	0.60493544	1.96125733	0.17900641	26.98500412
	X7	X8				
	0.33472965	-0.25454808				

DISCRIMINANT ANALYSIS

DF = 52

	DISCRIMINANT ANALYSIS			POOLED COVARIANCE MATRIX		
VARIABLE	X1	X2	X3	X4	X5	X6
X7	-0.49771062	-3.11229396	0.63951007	0.12402473	0.07875000	0.33472925
	X7	X8				
	7.26350733	-0.03628205				

	X1	X2	X3	X4	X5	X6
X8	-0.19833333	-2.63110577	-0.00464308	-0.00988636	0.03115224	-0.26454008
	X7	X8				
	-0.03628205	0.07697981				

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS PARTIAL CORRELATION COEFFICIENTS COMPUTED FROM POOLED COVARIANCE MATRIX / PROB > |R|

VARIABLE	X1	X2	X3	X4	X5	X6	X7	X8
X1	1.000000 0.0000	0.867734 0.0001	0.004264 0.9744	0.517796 0.0002	-0.108865 0.5562	0.552359 0.0001	-0.051450 0.7154	-0.199153 0.1492
X2	0.867734 0.0001	1.000000 0.0000	0.012130 0.9288	0.386421 0.0045	-0.081386 0.5692	0.409784 0.0027	-0.020786 0.8773	-0.170692 0.2195
X3	0.004264 0.9744	0.012130 0.9288	1.000000 0.0000	0.283289 0.0375	-0.053349 0.7060	0.335555 0.0134	0.683769 0.0001	-0.048276 0.7312
X4	0.517796 0.0002	0.386421 0.0045	0.283289 0.0375	1.000000 0.0000	0.147775 0.2911	0.905483 0.0001	0.110376 0.5628	-0.085469 0.5498
X5	-0.108865 0.5562	-0.081386 0.5692	-0.053349 0.7060	0.147775 0.2911	1.000000 0.0000	0.053324 0.7061	0.045219 0.7465	0.452647 0.0010
X6	0.552359 0.0001	0.409784 0.0027	0.335555 0.0134	0.905483 0.0001	0.053324 0.7061	1.000000 0.0000	0.023907 0.8593	-0.183537 0.1653
X7	-0.051450 0.7154	-0.020786 0.8773	0.683769 0.0001	0.110376 0.5628	0.045219 0.7465	0.023907 0.8593	1.000000 0.0000	-0.048276 0.7300
X8	-0.199153 0.1492	-0.170692 0.2195	-0.048276 0.7312	-0.085469 0.5498	0.452647 0.0010	-0.183537 0.1653	-0.048276 0.7300	1.000000 0.0000

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS		WITHIN COVARIANCE MATRIX INFORMATION	
GROUP	RANK OF COVARIANCE MATRIX	NATURAL LOG OF THE DETERMINANT OF THE COVARIANCE MATRIX	
1	8	-0.99468787	
2	8	3.16945640	
POOLED	8	3.53281769	

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS TEST OF HOMOGENIETY OF WITHIN COVARIANCE MATRICES

NOTATION: K = NUMBER OF GROUPS
 P = NUMBER OF VARIABLES
 N = TOTAL NUMBER OF OBSERVATIONS
 N(I) = NUMBER OF OBSERVATIONS IN THE I' TH GROUP

$$V = \frac{\prod \text{[WITHIN SS MATRIX(I)]}^{N(I)/2}}{\text{[POOLED SS MATRIX]}^{N/2}}$$

$$\text{RHO} = 1.0 - \left[\text{SUM} \frac{1}{N(I)-1} - \frac{1}{N-K} \right] \frac{2P + 3P - 1}{6(P+1)(K-1)}$$

$$\text{DF} = .5(K-1)P(P+1)$$

UNDER NULL HYPOTHESIS: $-2 \text{ RHO LN} \left[\frac{\prod N^{PN/2} V}{\prod N(I)^{PN(I)/2}} \right]$ IS DISTRIBUTED APPROXIMATELY AS CHI-SQUARE(DF)

TEST CHI-SQUARE VALUE = 142.36846474 WITH 36 DF PROB > CHI-SQ = 0.0001

SINCE THE CHI-SQUARE VALUE IS SIGNIFICANT AT THE 0.0500 LEVEL, THE WITHIN COVARIANCE MATRICES WILL BE USED IN THE DISCRIMINANT FUNCTION.

REFERENCE: KENDALL, M.G. AND STUART, A. THE ADVANCED THEORY OF STATISTICS VOL.3, P266 & 282.

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS PAIRWISE SQUARED GENERALIZED DISTANCES BETWEEN GROUPS

$$\text{WHERE: } D^2(I|J) = (\bar{X}_I - \bar{X}_J)' \text{COV}_J^{-1} (\bar{X}_I - \bar{X}_J) + \text{LN} |\text{COV}_J| - 2 \text{LN PRIOR}_J$$

GENERALIZED SQUARED DISTANCE TO GROUP

FROM GROUP	1	2
1	-0.49205901	71.70813752
2	14.87855452	6.17761119

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS CLASSIFICATION RESULTS FOR EACH OBSERVATION
 GIVING GENERALIZED SQUARED DISTANCE TO EACH GROUP / POSTERIOR PROBABILITY OF MEMBERSHIP IN EACH GROUP

WHERE: $D_J^2(X) = (X - \bar{X}_J)' COV_J^{-1} (X - \bar{X}_J) + LN |COV_J| - 2 LN PRICR_J$ AND $PR(J|X) = \frac{EXP[-.5 D_J^2(X)]}{\sum_K EXP[-.5 D_K^2(X)]}$

GENERALIZED SQUARED DISTANCE TO GROUP

OBS	FRJM GROUP	CLASSIFIED INTO GROUP	1	2
1	1	1	12.935399 0.9973	24.777567 0.0027
2	1	1	10.326322 0.9817	18.789670 0.0183
3	1	1	3.847199 0.9922	13.528990 0.0078
4	1	1	4.923170 0.9503	14.178049 0.0097
5	1	1	2.744549 0.9999	20.418493 0.0001
6	1	1	12.977099 1.0000	124.916269 0.0000
7	1	1	2.474798 0.9999	21.783300 0.0001
8	1	1	1.688878 0.9998	19.036582 0.0002
9	1	1	1.576131 0.9999	12.572584 0.0041
10	1	1	1.893001 0.9995	17.275311 0.0005
11	1	1	6.430258 0.9985	19.489257 0.0015
12	1	1	3.147143 0.9991	17.272594 0.0009
13	1	1	33.002755 1.0000	56.074765 0.0000
14	1	1	1.940459 0.9904	11.223535 0.0096
15	1	1	1.966196 0.9951	12.587829 0.0049

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS CLASSIFICATION RESULTS FOR EACH OBSERVATION
 GIVING GENERALIZED SQUARED DISTANCE TO EACH GROUP / POSTERIOR PROBABILITY OF MEMBERSHIP IN EACH GROUP

OBS	GENERALIZED SQUARED DISTANCE TO GROUP			
	FROM GROUP	CLASSIFIED INTO GROUP	1	2
16	1	1	1.781334 0.9902	11.013286 0.0098
17	1	1	1.901937 0.9956	12.724216 0.0044
18	1	1	19.217417 1.0000	62.641169 0.0000
19	1	1	33.389658 1.0000	1189.856481 0.0000
20	1	2	14.838607 0.4545	14.714894 0.5155
21	1	1	6.642013 0.9962	17.768379 0.0038
22	1	1	0.762468 0.9993	15.163327 0.0007
23	1	1	6.633738 0.9926	16.435963 0.0074
24	1	1	3.192285 0.9995	18.396065 0.0005
25	1	1	2.166141 0.9878	10.956954 0.0122
26	1	1	2.222423 0.9741	9.476114 0.0259
27	1	1	0.711625 0.9997	16.716412 0.0003
28	1	1	2.277366 0.9964	13.544857 0.0036
29	1	1	1.174411 0.9999	19.921607 0.0001
30	1	1	4.842068 1.0000	78.429224 0.0000
31	1	1	4.011712 0.9985	16.977548 0.0015

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS CLASSIFICATION RESULTS FOR EACH OBSERVATION
 GIVING GENERALIZED SQUARED DISTANCE TO EACH GROUP / POSTERIOR PROBABILITY OF MEMBERSHIP IN EACH GROUP

OBS	GENERALIZED SQUARED DISTANCE TO GROUP		POSTERIOR PROBABILITY OF MEMBERSHIP IN EACH GROUP	
	FROM GROUP	CLASSIFIED INTO GROUP	1	2
32	1	1	11.000187 0.9045	15.490706 0.0955
33	1	1	3.580966 0.9928	13.435340 0.0072
34	1	1	4.026944 1.0000	25.931109 0.0000
35	1	1	1.979179 1.0000	30.769296 0.0000
36	1	1	14.300782 0.9746	21.603303 0.0254
37	1	1	32.741203 1.0000	120.013361 0.0000
38	1	1	8.836561 0.9144	13.573141 0.0356
39	1	1	9.692490 0.9980	22.120466 0.0020
40	1	1	3.148359 0.9933	13.147575 0.0067
41	1	1	1.990927 0.9972	13.735411 0.0028
42	1	1	7.591054 1.0000	145.745810 0.0000
43	2	2	187.193111 0.0000	13.943135 1.0000
44	2	2	128.032699 0.0000	16.047209 1.0000
45	2	2	139.659960 0.0000	13.824966 1.0000
46	2	2	38.509279 0.0000	15.616264 1.0000
47	2	2	241.766596 0.0000	15.530083 1.0000

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS CLASSIFICATION RESULTS FOR EACH OBSERVATION
 GIVING GENERALIZED SQUARED DISTANCE TO EACH GROUP / POSTERIOR PROBABILITY OF MEMBERSHIP IN EACH GROUP

OBS	FROM GROUP	CLASSIFIED INTO GROUP	GENERALIZED SQUARED DISTANCE TO GROUP	
			1	2
48	2	1	8.533729 0.8589	12.145670 0.1411
49	2	2	36.219628 0.0000	13.025212 1.0000
50	2	2	49.373878 0.0000	14.334518 1.0000
51	2	1	5.074779 0.9781	12.676181 0.0219
52	2	2	262.942251 0.0000	12.757794 1.0000
53	2	2	559.466029 0.0000	13.263007 1.0000
54	2	2	168.871437 0.0000	8.947232 1.0000

DISCRIMINANT ANALYSIS

DISCRIMINANT ANALYSIS SUMMARY OF CLASSIFICATION PERFORMANCE USING GENERALIZED SQUARED DISTANCE

$$\text{WHERE: } D_J^2(X) = (X - \bar{X}_J)' \text{COV}_J^{-1} (X - \bar{X}_J) + \text{LN } |\text{COV}_J| - 2 \text{ LN PRIOR}_J$$

NUMBER OF OBSERVATIONS CLASSIFIED INTO GROUP

FROM GROUP	1	2
1	41	1
2	2	10

APPENDIX E

BIOMEDICAL COMPUTER PROGRAMS-STEPWISE
DISCRIMINANT ANALYSIS OUTPUT

BMD07M - STEPWISE DISCRIMINANT ANALYSIS - REVISED SEPT 14, 1970
 UNIVERSITY OF IDAHO
 COMPUTER SERVICES

PROBLEM CODE FRNGRD
 NUMBER OF VARIABLES 8
 NUMBER OF GROUPS 2
 NUMBER OF CASES IN EACH GROUP 42 12
 PRIOR PROBABILITIES 0.5000 0.5000
 VARIABLE FORMAT (1X,F5.0,7F6.0)

DATA INPUT FROM CARDS

MEANS (THE LAST COLUMN CONTAINS THE GRAND MEANS OVER THE GROUPS USED IN THE ANALYSIS)

VARIABLE	GROUP		
	A	B	
1	3.57143	10.83333	5.18518
2	17.57143	143.12500	45.47221
3	0.22905	0.23167	0.22963
4	0.25024	0.63000	0.33463
5	0.47833	1.00000	0.59426
6	2.36190	5.54166	3.06851
7	1.07143	0.41667	0.92593
8	0.15167	0.53833	0.23759

STANDARD DEVIATIONS

VARIABLE	GROUP	
	A	B
1	2.53889	6.07278
2	23.62111	111.85413
3	0.35526	0.31441
4	0.41226	0.43388
5	0.46501	1.08069
6	4.73702	6.62919
7	3.01534	0.66856
8	0.25908	0.33723

WITHIN GROUPS COVARIANCE MATRIX

VARIABLE	VARIABLES							
	1	2	3	4	5	6	7	8
1	12.88366							
2	173.03860	3086.55762						
3	0.00531	0.23386	0.12042					
4	0.77489	8.95073	0.04099	0.17383				
5	-0.25250	-2.92173	-0.01196	0.03981	0.41754			
6	10.29994	118.27290	0.60493	1.96125	0.17901	26.98891		
7	-0.49771	-3.11226	0.63951	0.12402	0.07875	0.33473	7.26343	
8	-0.19833	-2.63110	-0.00465	-0.00989	0.08115	-0.26455	-0.03628	0.07698

WITHIN GROUPS CORRELATION MATRIX

VARIABLE	VARIABLES							
	1	2	3	4	5	6	7	8
1	1.00000							
2	0.86773	1.00000						
3	0.00426	0.01213	1.00000					
4	0.51780	0.38642	0.28329	1.00000				
5	-0.10887	-0.08139	-0.05335	0.14777	1.00000			
6	0.55236	0.40978	0.33556	0.90548	0.05332	1.00000		
7	-0.05145	-0.02079	0.68379	0.11038	0.04522	0.02391	1.00000	
8	-0.19915	-0.17069	-0.04828	-0.08547	0.45265	-0.18354	-0.04852	1.00000

SUBPROBLEM 1
 F-LEVEL FOR INCLUSION 0.0100
 F-LEVEL FOR DELETION 0.0050
 TOLERANCE LEVEL 0.0001
 CONTROL VALUES 1111111

STEP NUMBER 0
 VARIABLE ENTERED

VARIABLES NOT INCLUDED AND F TO ENTER - DEGREES OF FREEDOM 1 52

1	38.2031	3	0.0005	5	6.0830	7	0.5509
2	47.6673	4	7.7435	6	3.4966	8	18.1273

STEP NUMBER 1
 VARIABLE ENTERED 2

VARIABLES INCLUDED AND F TO REMOVE - DEGREES OF FREEDOM 1 52

2 47.6673

VARIABLES NOT INCLUDED AND F TO ENTER - DEGREES OF FREEDOM 1 51

1	0.0751	3	0.0019	4	0.0080	5	4.7238	6	0.5660	7	0.1835	8	15.5752
---	--------	---	--------	---	--------	---	--------	---	--------	---	--------	---	---------

U-STATISTIC	0.52174	DEGREES OF FREEDOM	1	1	52
APPROXIMATE F	47.66727	DEGREES OF FREEDOM	1	52.00	

F MATRIX - DEGREES OF FREEDOM 1 52

GROUP
 A
 GROUP B 47.66727

STEP NUMBER 2
 VARIABLE ENTERED 8

VARIABLES INCLUDED AND F TO REMOVE - DEGREES OF FREEDOM 1 51

2 43.6188 8 15.5752

VARIABLES NOT INCLUDED AND F TO ENTER - DEGREES OF FREEDOM 1 50

1	0.3563	3	0.0151	4	0.0227	5	0.1579	6	0.0491	7	0.0364
---	--------	---	--------	---	--------	---	--------	---	--------	---	--------

U-STATISTIC	0.39968	DEGREES OF FREEDOM	2	1	52
APPROXIMATE F	38.30165	DEGREES OF FREEDOM	2	51.00	

F MATRIX - DEGREES OF FREEDOM 2 51

GROUP
 A
 GROUP B 38.30168

STEP NUMBER 3
VARIABLE ENTERED 1

VARIABLES INCLUDED AND F TO REMOVE - DEGREES OF FREEDOM 1 50

1 0.3563 2 3.8910 8 15.6382

VARIABLES NOT INCLUDED AND F TO ENTER - DEGREES OF FREEDOM 1 49

3 0.0173 4 0.0088 5 0.1701 6 0.2696 7 0.0211

U-STATISTIC 0.39685 DEGREES OF FREEDOM 3 1 52
APPROXIMATE F 25.33092 DEGREES OF FREEDOM 3 50.00

F MATRIX - DEGREES OF FREEDOM 3 50

GROUP
A
GROUP
B 25.33057

STEP NUMBER 4
VARIABLE ENTERED 6

VARIABLES INCLUDED AND F TO REMOVE - DEGREES OF FREEDOM 1 49

1 0.5720 2 3.3151 6 0.2696 8 14.5170

VARIABLES NOT INCLUDED AND F TO ENTER - DEGREES OF FREEDOM 1 48

3 0.1368 4 0.5211 5 0.2741 7 0.0120

U-STATISTIC 0.39468 DEGREES OF FREEDOM 4 1 52
APPROXIMATE F 18.78806 DEGREES OF FREEDOM 4 49.00

F MATRIX - DEGREES OF FREEDOM 4 49

GROUP
A
GROUP
B 18.78770

STEP NUMBER 5
VARIABLE ENTERED 4

VARIABLES INCLUDED AND F TO REMOVE - DEGREES OF FREEDOM 1 48

1 0.5081 2 3.2166 4 0.5211 6 0.7793 8 11.8491

VARIABLES NOT INCLUDED AND F TO ENTER - DEGREES OF FREEDOM 1 47

3 0.1538 5 0.1563 7 0.0778

U-STATISTIC 0.39044 DEGREES OF FREEDOM 5 1 52
APPROXIMATE F 14.98779 DEGREES OF FREEDOM 5 48.00

F MATRIX - DEGREES OF FREEDOM 5 48

GROUP
A
GROUP

B 14.98733

STEP NUMBER 6
VARIABLE ENTERED 5

VARIABLES INCLUDED AND F TO REMOVE - DEGREES OF FREEDOM 1 47

1 0.5629 2 3.0270 4 0.3976 5 0.1563 6 0.7219 8 7.6949

VARIABLES NOT INCLUDED AND F TO ENTER - DEGREES OF FREEDOM 1 46

3 0.1947 7 0.0804

U-STATISTIC 0.38914 DEGREES OF FREEDOM 6 1 52
APPROXIMATE F 12.29634 DEGREES OF FREEDOM 6 47.00

F MATRIX - DEGREES OF FREEDOM 6 47

GROUP
A

GROUP
B 12.29597

STEP NUMBER 7
VARIABLE ENTERED 3

VARIABLES INCLUDED AND F TO REMOVE - DEGREES OF FREEDOM 1 46

1 0.6726 2 2.7522 3 0.1947 4 0.3961 5 0.1972 6 0.8504 8 7.3142

VARIABLES NOT INCLUDED AND F TO ENTER - DEGREES OF FREEDOM 1 45

7 0.8811

U-STATISTIC 0.38750 DEGREES OF FREEDOM 7 1 52
APPROXIMATE F 10.38695 DEGREES OF FREEDOM 7 46.00

F MATRIX - DEGREES OF FREEDOM 7 46

GROUP
A

GROUP
B 10.38663

STEP NUMBER 8
VARIABLE ENTERED 7

VARIABLES INCLUDED AND F TO REMOVE - DEGREES OF FREEDOM 1 45

1 0.7941 3 0.9949 5 0.3547 7 0.8811
2 2.5477 4 0.8228 6 1.5602 8 5.4723

U-STATISTIC 0.38006 DEGREES OF FREEDOM 8 1 52
APPROXIMATE F 9.17524 DEGREES OF FREEDOM 8 45.00

F MATRIX - DEGREES OF FREEDOM 8 45

GROUP
A

GROUP

F LEVEL INSUFFICIENT FOR FURTHER COMPUTATION

VARIABLE	FUNCTION	
	A	B
1	1.12544	1.45173
2	-0.04698	-0.01417
3	4.49524	7.25309
4	1.67609	4.88811
5	1.49667	2.12245
6	-0.35767	-0.75064
7	-0.21544	-0.52850
8	0.84231	6.24854

CONSTANT	A	B
	-2.89865	-10.47528

GROUP WITH LARGEST PROB. SQUARE OF DISTANCE FROM AND POSTERIOR PROBABILITY FOR GROUP -

GROUP CASE	A	A				B	
1	A	11.285	0.796,	14.006	0.204,		
2	A	9.759	0.941,	15.281	0.059,		
3	A	3.079	0.994,	13.325	0.006,		
4	A	1.659	0.996,	12.518	0.004,		
5	A	2.545	0.999,	16.772	0.001,		
6	A	11.018	0.998,	23.790	0.002,		
7	A	2.408	0.997,	13.743	0.003,		
8	A	1.916	0.999,	15.114	0.001,		
9	A	1.310	0.992,	11.044	0.000,		
10	A	1.184	0.994,	11.513	0.006,		
11	A	3.956	0.976,	11.353	0.024,		
12	A	2.504	0.986,	11.079	0.014,		
13	B	34.635	0.034,	27.926	0.966,		
14	A	0.954	0.993,	10.953	0.007,		
15	A	2.438	0.970,	9.363	0.030,		
16	A	1.485	0.984,	9.676	0.016,		
17	A	1.679	0.995,	12.264	0.005,		
18	A	16.383	0.998,	28.747	0.002,		
19	A	40.648	0.998,	52.940	0.002,		
20	A	5.833	0.738,	7.904	0.262,		
21	A	5.720	0.913,	10.425	0.087,		
22	A	1.233	0.994,	11.367	0.006,		
23	A	3.266	0.994,	13.385	0.006,		
24	A	2.679	0.994,	12.859	0.006,		
25	A	1.730	0.967,	8.481	0.033,		
26	A	2.381	0.914,	7.098	0.086,		
27	A	0.796	0.998,	13.286	0.002,		
28	A	1.478	0.998,	13.469	0.002,		
29	A	0.938	0.998,	13.124	0.002,		
30	A	4.399	0.998,	16.711	0.002,		
31	A	2.350	0.995,	12.802	0.005,		
32	A	4.280	0.811,	7.195	0.189,		
33	A	1.554	0.995,	12.192	0.005,		
34	A	4.669	0.998,	16.889	0.002,		
35	A	1.718	0.998,	14.551	0.002,		
36	A	7.954	0.998,	20.150	0.002,		
37	A	29.507	0.848,	32.943	0.152,		
38	A	2.794	0.974,	10.024	0.026,		
39	A	3.957	0.996,	15.225	0.004,		
40	A	1.333	0.982,	9.291	0.018,		
41	A	2.019	0.995,	12.565	0.005,		

42	A	6.302 0.997,	18.068 0.003,
GROUP		A	B
CASE	B		
1	B	21.885 0.000,	3.846 1.000,
2	B	19.465 0.003,	7.598 0.997,
3	A	15.960 0.956,	22.099 0.044,
4	B	42.375 0.000,	25.459 1.000,
5	A	7.152 0.824,	10.243 0.176,
6	B	7.903 0.172,	4.758 0.828,
7	B	15.054 0.197,	12.240 0.803,
8	A	3.695 0.925,	8.708 0.075,
9	B	21.033 0.007,	11.050 0.993,
10	B	45.081 0.000,	23.453 1.000,
11	B	35.437 0.000,	11.099 1.000,
12	B	40.275 0.001,	25.710 0.999,

NUMBER OF CASES CLASSIFIED INTO GROUP -

	A	B
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GROUP	A	41	1
	B	3	9

SUMMARY TABLE

STEP NUMBER	VARIABLE ENTERED	VARIABLE REMOVED	F VALUE TO ENTER OR REMOVE	NUMBER OF VARIABLES INCLUDED	U-STATISTIC
1	2		47.6673	1	0.5217
2	8		15.5752	2	0.3997
3	1		0.3563	3	0.3968
4	6		0.2696	4	0.3947
5	4		0.5211	5	0.3904
6	5		0.1563	6	0.3891
7	3		0.1947	7	0.3875
8	7		0.8811	8	0.3801

EIGENVALUES

1.63109	0.00001	0.00000	0.00000	-0.00000	-0.00000	-0.00000	-0.00000
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CUMULATIVE PROPORTION OF TOTAL DISPERSION

1.00000	1.00000	1.00000	1.00000	1.00000	1.00000	1.00000	1.00000
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CANONICAL CORRELATIONS

0.78736	0.00291	0.00157	0.00092	0.00018	0.00064	0.00064	0.00104
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COEFFICIENTS FOR CANONICAL VARIABLE -

ORIGINAL VARIABLE	1	2	3	4	5	6	7	8
1	0.10824	-0.56604	0.21136	-0.13818	0.11312	-0.02377	0.06072	0.07840
2	0.01088	0.03338	-0.01126	0.00127	0.00261	-0.00014	0.00069	-0.00150
3	0.91484	-1.40911	-2.36101	-3.20436	-0.86386	1.95822	0.54830	-0.66621
4	1.06548	-2.16088	-5.02430	2.68509	-0.09756	0.14274	0.86542	-0.42151
5	0.20758	-0.20230	0.34646	-0.42690	0.24151	-0.10850	0.50168	-1.64539
6	-0.13036	0.26984	0.44760	-0.01320	-0.13428	0.04328	-0.01402	0.06364
7	-0.10385	0.12899	0.31281	0.31666	0.28742	0.10729	-0.15775	0.06026
8	1.79337	0.19745	1.01881	0.78773	-1.60443	0.30986	-3.04712	1.23312

GROUP CANONICAL VARIABLES EVALUATED AT GROUP MEANS

1	-0.66990	-0.00000	-0.00000	-0.00000	0.00000	0.00000	0.00000	-0.00000
2	2.34465	-0.00000	0.00000	0.00000	-0.00000	0.00000	0.00000	-0.00000

CHECK ON FINAL U-STATISTIC 0.38007

POINTS PLOTTED ON THE FOLLOWING GRAPH

X = FIRST CANONICAL VARIABLE

Y = SECOND CANONICAL VARIABLE

CASE NUMBER FOLLOWED BY * INDICATES THE POINT IS OFF THE GRAPH

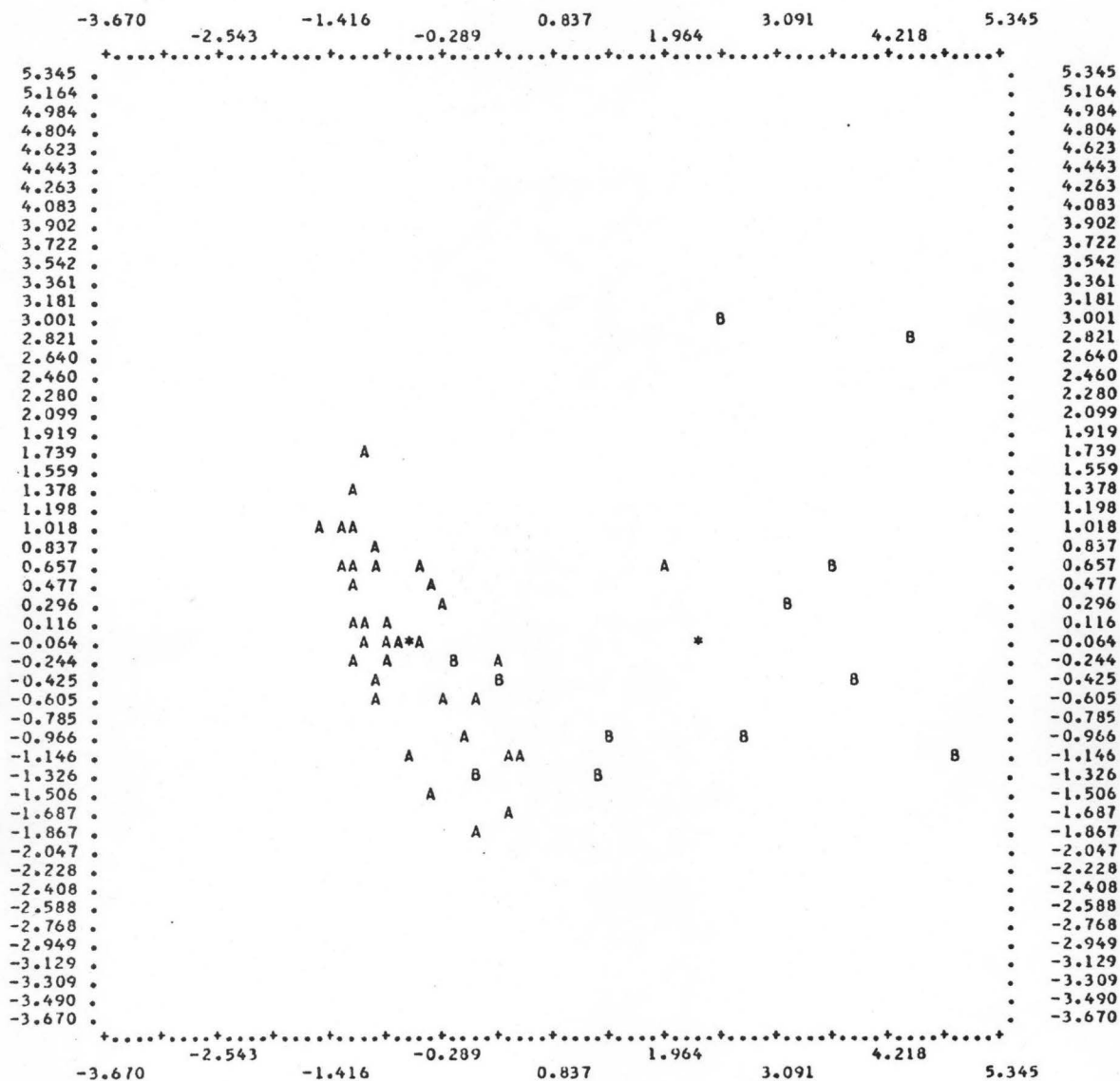
GROUP A MEAN COORDINATES -0.670 -0.000

CASE	X	Y	CASE	X	Y	CASE	X	Y	CASE	X	Y	CASE	X	Y
1	0.386	-1.716	11	-0.389	-1.545	2	-0.078	-1.002	12	-0.585	-1.058	3	-0.862	0.136
4	-0.964	-0.513	14	-0.821	-0.201	5	-1.522	1.060	15	-0.311	-0.642	6	-1.281	0.620
7	-1.043	0.133	17	-0.918	0.601	8	-1.352	0.974	18	-1.213	0.637	9	-0.777	-0.151
10	-0.876	-0.144	20	0.494	-1.108							13	1.950	0.725
21	0.057	-1.852	31	-0.896	0.178	22	-0.843	-0.212	32	0.354	-1.089	23	-0.841	-0.177
24	-0.851	-0.103	34	-1.189	1.446	25	-0.283	0.231	35	-1.291	1.029	26	0.055	-0.550
27	-1.234	0.422	37	0.268	-0.248	28	-1.152	0.147	38	-0.362	0.412	29	-1.184	0.072
30	-1.205	-0.261	40	-0.483	0.721							33	-0.927	-0.521
41	-0.912	0.836										36	-1.185	0.972
42	-1.114	-0.124										39	-1.032	1.730

GROUP B MEAN COORDINATES 2.345 -0.000

CASE	X	Y	CASE	X	Y
1	3.829	-0.435	11	4.874	-1.169
2	2.806	-0.986	12	3.253	0.302
3	-0.181	-0.166			
4	3.643	0.686			
5	0.325	-0.499			
6	1.359	-0.942			
7	1.304	-1.258			
8	0.006	-1.346			
9	2.493	2.919			
10	4.425	2.894			

OVERLAP IS INDICATED BY \$, GROUP MEANS BY *.



FINISH CARD ENCOUNTERED, JOB TERMINATED