

Data Analysis

Discriminant Analysis

Preliminaries

- Under the name of *discriminant analysis* there are reunited various explicative, descriptive and predictive methods designed to study a class or category based population.
- *Discriminant analysis* belongs to the class of supervised learning type of problems, which implies the machine learning task of learning a function that maps an input to an output based on example input-output pairs.
- Each individual observation is characterized by **a set of independent predictor variables and one qualitative variable** whereby the class it belongs to is identified.

Discriminant Analysis

Preliminaries

- The population is divided in 2 subsets:
 - a) the base sample*, for which the qualitative variable value is known, hence the observations are categorized;
 - b) the uninvestigated sample*, case in which the observations are not categorized, and the qualitative variable value is not known.

Discriminant Analysis

Preliminaries

- Discriminant analysis intends to:
 - a) identify the rules based on which the individual observations can be classified, placed in certain classes or categories,
 - b) and, on the other hand, to reduce the number of necessary variables for categorization or for making the discrimination.
- The first aspect highlights the predictive, decisional character of discriminant analysis, while the second one rather reveals the descriptive aspect of the discriminant analysis.

Discriminant Analysis

Preliminaries

- Discriminant analysis is frequently applied in fields and problems, such as:
 - pattern recognition,
 - financial sector, credit institutions, in order to predict the behavior of credit solicitants,
 - medicine, based on laboratory results, there is to be identified a function for estimating the type of symptoms associated to a disease or its probable evolution,
 - meteorology, the prediction of avalanche, based on the weather related variables, snowfalls etc.

Discriminant Analysis

Notations

- Observation matrix:

$$X = \begin{bmatrix} x_{11} & x_{12} & \dots & x_{1m} \\ x_{21} & x_{22} & \dots & x_{2m} \\ \dots & & & \\ x_{n1} & x_{n2} & \dots & x_{nm} \end{bmatrix},$$

where n is the number of observations, and m is the number of predictor variables (independent variables).

Discriminant Analysis

Notations

- Discriminant variable:

$$Y = \begin{bmatrix} y_1 \\ \dots \\ y_n \end{bmatrix} .$$

It is a qualitative variable. A value y_i , $i = 1, n$ represents the class (group or category) the observation i belongs to.

There could be $q \neq n$ number of groups, classes or categories.

Discriminant Analysis

Notations

- Observation vectors:

$w_i, i=1, n$, where w_i is the row i of matrix X .

- Variable vectors:

$x_j, j=1, m$, where x_j is the column j of matrix X .

- Group centers matrix:

$$G = \begin{bmatrix} g_{11} & g_{12} & \cdots & g_{1m} \\ g_{21} & g_{22} & \cdots & g_{2m} \\ \cdots & & & \\ g_{q1} & g_{q2} & \cdots & g_{qm} \end{bmatrix} ,$$

where q is the number of groups, classes or categories.

Discriminant Analysis

Notations

- A value g_{kj} represent the mean of predictor variable j for the k group.
- Group center vectors $G_k, k=1,q, G_k = \begin{bmatrix} g_{k1} \\ \dots \\ g_{km} \end{bmatrix}$.
- The overall mean: $\bar{X} = \begin{bmatrix} \bar{x}_1 \\ \dots \\ \bar{x}_m \end{bmatrix}$.
- The diagonal matrix of group frequencies:

$$D_G = \begin{bmatrix} n_1 & 0 & \dots & 0 \\ 0 & n_2 & \dots & 0 \\ \dots & & & \\ 0 & 0 & \dots & n_q \end{bmatrix}.$$

Discriminant Analysis

Variability indicators and dispersion (scatter)

- Discrimination among groups is achieved with variability and dispersion indicators.
 - A. Scatter matrices (*sum of square and cross product*):
 - reflect the scatter level associated to the whole collectivity (*SST*),
 - within the groups (*SSW*), and
 - the scatter of groups among each other (*SSB*).
- *SST* is the scatter matrix of the whole collectivity, and shows the scatter level around the overall mean.

Discriminant Analysis

Variability indicators and dispersion (scatter)

- The general term of SST matrix is:

$$\begin{aligned}
 SST_{jl} &= \sum_{i=1}^n (x_{ij} - \bar{x}_j)(x_{il} - \bar{x}_l) = \sum_{k=1}^q \sum_{i \in k} (x_{ij} - \bar{x}_j)(x_{il} - \bar{x}_l) = \\
 &= \sum_{k=1}^q \sum_{i \in k} (x_{ij} - g_{kj} + g_{kj} - \bar{x}_j)(x_{il} - g_{kl} + g_{kl} - \bar{x}_l) = \\
 &= \sum_{k=1}^q \sum_{i \in k} (x_{ij} - g_{kj})(x_{il} - g_{kl}) + \sum_{k=1}^q \sum_{i \in k} (g_{kj} - \bar{x}_j)(g_{kl} - \bar{x}_l) + \\
 &+ \sum_{k=1}^q \sum_{i \in k} (x_{ij} - g_{kj})(g_{kl} - \bar{x}_l) + \sum_{k=1}^q \sum_{i \in k} (g_{kj} - \bar{x}_j)(x_{il} - g_{kl}) \\
 &=
 \end{aligned}$$

Discriminant Analysis

Variability indicators and dispersion (scatter)

- The general term of SST matrix is:

$$\begin{aligned} SST_{jl} = & \sum_{k=1}^q \sum_{i \in k} (x_{ij} - g_{kj})(x_{il} - g_{kl}) + \sum_{k=1}^q n_k (g_{kj} - \bar{x}_j)(g_{kl} - \bar{x}_l) + \\ & + \sum_{k=1}^q (g_{kl} - \bar{x}_l) \sum_{i \in k} (x_{ij} - g_{kj}) + \sum_{k=1}^q (x_{il} - g_{kl}) \sum_{i \in k} (g_{kj} - \bar{x}_j) \end{aligned}$$

- a) The first sum, $\sum_{k=1}^q \sum_{i \in k} (x_{ij} - g_{kj})(x_{il} - g_{kl})$, represents the general term of the scatter matrix within the groups, SSW .

Discriminant Analysis

Variability indicators and dispersion (scatter)

b) The second sum, $\sum_{k=1}^q n_k (g_{kj} - \bar{x}_j)(g_{kl} - \bar{x}_l)$, is general term of the scatter matrix among the groups, SSB .

c) The third and the fourth sums have the value 0 (zero), because the simple sums of the deviation from the groups means are 0:

$$\sum_{i \in k} (x_{ij} - g_{kj}) = 0 \quad \text{and} \quad \sum_{i \in k} (x_{il} - g_{kl}) = 0$$

- Therefore: $SST_{jl} = SSW_{jl} + SSB_{jl}$, $j=1, m$, $l=1, m$.
- And in terms of matrices: $SST = SSW + SSB$.

Discriminant Analysis

Variability indicators and dispersion (scatter)

- If there are taken into account the degrees of freedom, then the scatter matrices are computed as follows:

$$MST = \frac{SST}{n-1}, \quad MSW = \frac{SSW}{n-q}, \quad MSB = \frac{SSB}{q-1}.$$

Discriminant Analysis

Variability indicators and dispersion (scatter)

B. Covariance matrices:

– The general terms of the covariance matrices differ from those of the scatter matrices by the fact that they are computed as mean values.

– Overall, total covariance:

$$T_{jl} = \frac{1}{n} \sum_{i=1}^n (x_{ij} - \bar{x}_j)(x_{il} - \bar{x}_l), \quad j=1,m, \quad l=1, m$$

– Intra-group covariance (within the groups)

$$W_{jl} = \sum_{k=1}^q \frac{n_k}{n} \cdot \frac{1}{n_k} \sum_{i \in k} (x_{ij} - g_{kj})(x_{il} - g_{kl}), \quad j=1,m, \quad l=1,m$$

Discriminant Analysis

Variability indicators and dispersion (scatter)

- Inter-group covariance (among the groups):

$$B_{jl} = \sum_{k=1}^q \frac{n_k}{n} (g_{kj} - \bar{x}_j)(g_{kl} - \bar{x}_l), j=1,m, l=1,m$$

- The relation between terms is the same as in the case of scatter matrices:

$$T_{jl} = W_{jl} + B_{jl}, j=1,m, l=1,m.$$

- And in terms of matrices: $T = W + B$.

Discriminant Analysis

Variability indicators and dispersion (scatter)

C. Total variance:

- It is given by the values contained on the principal diagonals of the covariance and scatter matrices:

$VT = \text{Trace}(T)$ is the general (overall) total variance,

$VW = \text{Trace}(W)$ is the total intra-group variance,

$VB = \text{Trace}(B)$ is the total inter-group variance.

- Or, with scatter matrices:

$VT = \text{Trace}(SST)$,

$VW = \text{Trace}(SSW)$,

$VB = \text{Trace}(SSB)$.

Discriminant Analysis

Variability indicators and dispersion (scatter)

D. Generalized variance:

- It is computed as determinant of covariance and scatter matrices:

$$VGT = |T| ,$$

$$VGW = |W| ,$$

$$VGB = |B| .$$

- Or, with scatter matrices:

$$VGT = |SST| ,$$

$$VGW = |SSW| ,$$

$$VGB = |SSB| .$$

Discriminant Analysis

Model significance. Statistical tests.

- Model testing is executed in 2 phases:
 - One Fisher test, based on Wilks statistics, which shows if the set of predictor variables can make the discrimination on the groups of instances as a whole;
 - Individual statistical tests for each predictor variables whereby is decided whether such a variable can be a good predictor.

a) The global F test:

$$H_0: G_1 = G_2 = \dots = G_q$$

$$H_1: \exists \text{ two groups } i, k \text{ such that } G_i \neq G_k$$

Discriminant Analysis

Model significance. Statistical tests.

- It is computed the following lambda indicator:

$$\Lambda = \frac{|SSB|}{|SSB + SSW|}$$

- The greater Λ is, the more likely that the H_0 hypothesis is to be rejected.
- The test statistics is a Fisher value computed as follows:

$$F = \frac{1 - \Lambda^{1/b}}{\Lambda^{1/b}} \frac{ab - c}{m(q - 1)}$$

where a , b and c are computed as:

Discriminant Analysis

Model significance. Statistical tests.

where a , b and c are computed as:

$$a = n - q - \frac{m - q + 2}{2}$$

$$b = \begin{cases} \sqrt{\frac{m^2(q-1) - 4}{m^2 + (q-1)^2 - 5}} & , \text{if } m^2 + (q-1)^2 - 5 > 0 \\ 1 & , \text{if } m^2 + (q-1)^2 - 5 \leq 0 \end{cases}$$

$$c = \frac{m(q-1) - 2}{2}$$

Discriminant Analysis

Model significance. Statistical tests.

- If $F > F_{m(q-1), ab-c; \alpha}^{Critic}$,

then the null hypothesis (H0) is rejected with a degree of credence $1-\alpha$.

b) Individual statistical tests, for each predictor variable:

- A predictor variable is considered a good predictor if is able to separate the groups as clear as possible.
- Therefore, the ratio between the inter-group variance and the intra-group variance is as great as possible.

Discriminant Analysis

Model significance. Statistical tests.

- Having a ration between variances, the test F is to be applied.
- Therefore, for a given predictor variable j , the null and the alternative hypothesis are:

$$H_0: g_{1j} = g_{2j} = \dots = g_{qj}$$

$$H_1: \exists k, i \text{ two groups such that } g_{kj} \neq g_{ij}$$

- The statistics of the test is: $F_j = \frac{SSB_{jj}}{SSW_{jj}}$.
- The critical value for $q-1$ and $n-q$ degrees of freedom and a significance threshold α is:

$$F_{q-1; n-q; \alpha}^{Critic} .$$

Discriminant Analysis

Model significance. Statistical tests.

- If $F_j > F_{q-1;n-q;\alpha}^{Critic}$,

then the null hypothesis (H0) is rejected with level of trust $1-\alpha$.