

# Software Development for 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 forth sums have the value 0 (zero), because the simple sums of the deviation from the groups means are 0:

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

- 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:

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

# 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:

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

- Intra-group covariance (within the groups)

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

# 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  $\Lambda$  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  $F^{Computed} > F^{Critic}_{m(q-1), ab-c; \alpha}$  ,

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 the ratio 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:  $F_j = \frac{SSB_j}{SSW_j}$  .
- The critical value for  $q-1$  and  $n-q$  degrees of freedom and a significance threshold  $\alpha$  is:

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

# Discriminant Analysis

**Model significance. Statistical tests.**

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

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