#### R instructions for the 7th seminar

In a data set *Holiday.RData* there are information about 50 families and their attitude to holiday destinations:

- ID | yes/no indicator showing whether a family visited particular holiday destination during last 2 years
- $X_1$  states an annual family income in thousands of dollars
- $X_2$  states an attitude to travelling (on a scale 1 to 9, where 1 = absolutely negative, 9 = absolutely pozitive)
- $X_3$  states an importance attributed to a family holiday (on a scale 1 to 9, where 1 = the lowest, 9 = the highest)
- $X_4$  states the number of family members
- $X_5$  states an age of the oldest member of a family
- V states planed holiday expenditures of a family: little (1), medium (2), much (3)

All following tasks do first when:

a) The ID is a grouping variable (this must be of class factor). Explanatory variables are  $X_1, \ldots, X_5$ . (Data are considered to be training data set.)

At home when:

b) The V is a grouping variable. Explanatory variables are  $X_1, \ldots, X_5$  the same.

## R Instructions for the problem 1:

Get familiar with data

•Plot 2D graphs with pairs of explanatory variables and distinguish the points with regard to grouping variable.

```
plot(Holiday[,2:6],col=Holiday[,1])
```

•Calculate means of all explanatory variables separated by grouping variable; what do the results suggest?

```
for(i in 2:6)cat("X",i-1,": ",tapply(Holiday[,i],Holiday[,1],mean), ' n')
```

CANONICAL DISCRIMINANT ANALYSIS:

### R Instructions for the problem 2:

•How many canonical discriminant variables are in the model? (Answer: for grouping variable = ID it is  $l = \min\{p = 5, k - 1 = 1\} = 1$ . Thus we are looking for just one discriminant variable.)

# R Instructions for the problem 3:

• Create the object in R bearing all essential results of Canonical discriminant analysis.

```
!library(candisc)
```

```
LinModel <- lm(cbind(X1, X2, X3, X4, X5) ID, data=Holiday)
```

CanResults1<-candisc(LinModel,term="ID")</pre>

or for groupin variable = V:

```
LinModel2<-lm(cbind(X1,X2,X3,X4,X5) V,data=Holiday)</pre>
```

CanResults2<-candisc(LinModel2,term="V")</pre>

#### R Instructions for the problem 4:

•Find out raw and standardized discriminant coefficients. (Standardized allows better interpretation.) Which variable bears the largest peace of information for discrimination?

```
CanResults1$coeffs.raw CanResults1$coeffs.std
```

(for standardized:  $Y = -0.848X_1 + ... + -0.469X_5$  Thus the direction of discriminant variable Y is mostly influenced by  $X_1$  which is an income...)

## R Instructions for the problem 5:

•Find out correlations between original variables and canonical variable(s). Compare its signs with signs of discriminant coefficients. Interprete it.

```
CanResults1$structure (e.g.: R(X_1, Y) = -0.929).
```

# R Instructions for the problem 6:

•Find out canonical scores. In other words express all cases by the mean of the new canonical variable(s).

CanResults1\$scores

## R Instructions for the problem 7:

•Find out means of the new canonical variable(s) for both groups.

```
CanResults1$means
```

```
("No" group mean of Y: 1.059855; "Yes" group mean of Y: -1.463609)
```

# R Instructions for the problem 8:

• Express graphically the results of canonical discriminant analysis.

```
plot(CanResults1, fill=TRUE)
NewData<-data.frame(CanResults1$scores,rep(0,times=50))
head(NewData)
plot(NewData[,2:3],col=NewData[,1])</pre>
```

for voluntiers TODO: LINEAR DISCRIMINANT ANALYSIS LDA: !library(MASS)

#### R Instructions for the problem 9:

- •Check assumptions (multivariate normality in all groups (we wish not to reject), equality of Variance matrices (we wish not to reject), equality of mean vectors (we wish to reject)
- •Create objects for LDA:

```
\label{eq:loss} LdaResults1 <- lda(ID ~ X1+X2+X3+X4+X5, data=Holiday, prior=c(0.5,0.5), CV=F) \\ LdaResults1
```

LdaPredict1<-predict(LdaResults1, newdata=Holiday[,2:6])\$class

To get the classification matrix:

```
LdaPredict1 table(LdaPredict1,Holiday[,1])  
LdaResults1.1<-lda(ID \sim X1+X2+X3+X4+X5,data=Holiday,prior=c(0.5,0.5),CV=T)  
LdaResults1.1
```

Nefachci pro jine prior, koreny i tabulka uspesnosti stejne, jako pro  $0,\!5$ 

```
 LdaResults 2 < -lda(ID \sim X1 + X2 + X3 + X4 + X5, data = Holiday, prior = c(0.58, 0.42), CV = F) \\ LdaResults 2
```

### Homework:

File *Iris.RData* contains data about length and width of sepals and petals of three species of *Iris* plant. Use DA to distunguish the three classes based on their petal and sepal dimensions.