Assignment 4 - Canonical Correlation Analysis

GROUP 03

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Canonical correlation analysis by utilizing suit able software

Look at the data described in Exercise 10.16 of Johnson, Wichern. You may find it in the file P10-16.DAT. The data for 46 patients are summarized in a covariance matrix, which will be analyzed in R. Read through the description of the different R packages and functions so you may chose the must suitable one for the analysis. Supplement with own code where necessary.

The given matrix is the following:

V1	V2	V3	V4	V5
1106.000	396.700	108.400	0.787	26.230
396.700	2382.000	1143.000	-0.214	-23.960
108.400	1143.000	2136.000	2.189	-20.840
0.787	-0.214	2.189	0.016	0.216
26.230	-23.960	-20.840	0.216	70.560

Thus, separating the variance-covariance matrix it is obtained that

$$\Sigma_{11} =$$

V1	V2	V3
1106.0	396.7	108.4
396.7	2382.0	1143.0
108.4	1143.0	2136.0

$$\Sigma_{22} =$$

	V4	V5
$\overline{4}$	0.016	0.216
5	0.216	70.560

$$\Sigma_{21} =$$

	V1	V2	V3
4	0.787	-0.214	2.189
5	26.230	-23.960	-20.840

$$\Sigma_{12} =$$

V4	V5
0.787	26.23
-0.214	-23.96
2.189	-20.84

It can be computed that

$$A = S_{11}^{-1/2} S_{12} S_{22}^{-1} S_{21} S_{11}^{-1/2} =$$

0.0468068	-0.0374259	0.0813854
-0.0374259	0.0306007	-0.0716743
0.0813854	-0.0716743	0.2059907

Its eigen-decomposition is given by

Eigenvalues $\overrightarrow{\alpha}$:

X
0.2676458
0.0157523
0.0000000

Eigenvectors \overrightarrow{a} :

0.3749438	-0.7634104	0.5259484
-0.3220478	0.4247428	0.8460962
0.8693114	0.4866191	0.0866000

Also,

$$B = S_{22}^{-1/2} S_{21} S_{11}^{-1} S_{12} S_{22}^{-1/2} =$$

0.2671702	0.0109346
0.0109346	0.0162279

Its eigen-decomposition is given by

Eigenvalues $\overrightarrow{\beta}$:

 $\frac{x}{0.2676458}\\0.0157523$

Eigenvectors \overrightarrow{b} :

-0.9990556 0.0434508 -0.0434508 -0.9990556

Then the canonical correlations are

$$\rho_1 = \sqrt{\overrightarrow{\alpha}}$$

$$\frac{\mathbf{x}}{0.5172440}$$

 $\begin{array}{c} 0.5173449 \\ 0.1255082 \end{array}$

$$\rho_2 = \sqrt{\overrightarrow{\beta}}$$

 $\frac{x}{0.5173449}$ 0.1255082

a) Test at the 5% level if there is any association between the groups of variables.

```
alpha <- 0.05
#critical value
crit <- qchisq(p = (1-alpha), df = p*q)
#test statistic</pre>
```

- b) How many pairs of canonical variates are significant?
- c) Interpret the "significant" squared canonical correlations. Tip: Read section "Canonical Correlations as Generalizations of Other Correlation Coefficients".
- d) Interpret the canonical variates by using the coefficients and suitable correlations.
- e) Are the "significant" canonical variates good summary measures of the respective data sets? Tip: Read section "Proportions of Explained Sample Variance".
- f) Give your opinion on the success of this canonical correlation analysis.

Appendix A - Code

```
RNGversion('3.5.1')
knitr::opts_chunk$set(echo = TRUE)
library(expm)
library(knitr)
library(CCP)
data <- read.table("./Data/P10-16.DAT")</pre>
#number of observations (patients)
n <- 46
#number of primary variables
p <- 3
#number of secondary variables
q <- 2
kable(data)
#separating the variance-covariance matrix
sigma11 <- as.matrix(data[1:3,1:3])</pre>
sigma22 <- as.matrix(data[4:5, 4:5])</pre>
sigma21 <- as.matrix(data[4:5, 1:3])</pre>
sigma12 <- as.matrix(data[1:3, 4:5])</pre>
A <- solve(sqrtm(sigma11)) %*% sigma12 %*% solve(sigma22) %*% sigma21 %*% solve(sqrtm(sigma11))
B <- solve(sqrtm(sigma22)) %*% sigma21 %*% solve(sigma11) %*% sigma12 %*% solve(sqrtm(sigma22))
kable(sigma11)
kable(sigma22)
kable(sigma21)
kable(sigma12)
kable(A)
aEigenDecom <- eigen(A)
aEigVect <- aEigenDecom$vectors</pre>
aEigVals <- aEigenDecom$values
kable(aEigVals)
aEigVals <- aEigVals[1:2]
kable(aEigVect)
kable(B)
bEigenDecom <- eigen(B)
bEigVect <- bEigenDecom$vectors</pre>
bEigVals <- bEigenDecom$values
kable(bEigVals)
kable(bEigVect)
rho1 <- sqrt(aEigVals)</pre>
kable(rho1)
rho2 <- sqrt(bEigVals)</pre>
kable(rho2)
alpha \leftarrow 0.05
#critical value
crit <- qchisq(p = (1-alpha), df = p*q)</pre>
#test statistic
```