

## Module 19 Multivariate Analysis for Genetic data

### Session 07: Canonical Correlation Analysis

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# Contents

1 Introduction

2 Maximization problem

3 Biplots

4 Examples

# Some History



- Hotelling, H. (1935) The most predictable criterion, *Journal of Educational Psychology* 26 pp. 139-142.
- Hotelling, H. (1936) Relations between two sets of variates. *Biometrika*, 28 pp. 321-377.

## Objective

$n \times p$

$n \times q$

- Study the relationship between two sets of variables **X** and **Y**
  - Find linear combinations of **X** and **Y** that have maximal correlation

# Notes

Canonical correlation analysis is important from a theoretical point of view, since it:

- underlies multiple linear regression ( $q = 1$ )
- underlies discriminant analysis (**Y** categorical)
- underlies correspondence analysis (**X** and **Y** categorical)
- underlies .... many multivariate methods!

# Some notation

- $\mathbf{x}$  a p-variate random vector
- $\mathbf{y}$  a q-variate random vector
- $\Sigma_{xx}$  within set correlation matrix of  $x$  variates
- $\Sigma_{yy}$  within set correlation matrix of  $y$  variates
- $\Sigma_{xy}$  between set correlation matrix

$$\Sigma = \begin{bmatrix} \Sigma_{xx} & \Sigma_{xy} \\ \Sigma_{yx} & \Sigma_{yy} \end{bmatrix},$$

- $\mathbf{u} = \mathbf{a}'\mathbf{x}$  first canonical  $x$  variate
- $\mathbf{v} = \mathbf{b}'\mathbf{y}$  first canonical  $y$  variate

# Maximization problem

We have:

$$V(\mathbf{u}) = \mathbf{a}'\boldsymbol{\Sigma}_{xx}\mathbf{a} \quad V(\mathbf{v}) = \mathbf{b}'\boldsymbol{\Sigma}_{yy}\mathbf{b} \quad Cov(\mathbf{u}, \mathbf{v}) = \mathbf{a}'\boldsymbol{\Sigma}_{xy}\mathbf{b}$$

Maximize

$$\rho(\mathbf{u}, \mathbf{v}) = \frac{\mathbf{a}'\boldsymbol{\Sigma}_{xy}\mathbf{b}}{\sqrt{\mathbf{a}'\boldsymbol{\Sigma}_{xx}\mathbf{a}}\sqrt{\mathbf{b}'\boldsymbol{\Sigma}_{yy}\mathbf{b}}}$$

Equivalently

$$\max_{\mathbf{a}, \mathbf{b}} \mathbf{a}'\boldsymbol{\Sigma}_{xy}\mathbf{b} \text{ subject to } \mathbf{a}'\boldsymbol{\Sigma}_{xx}\mathbf{a} = 1 \quad \mathbf{b}'\boldsymbol{\Sigma}_{yy}\mathbf{b} = 1$$

# Solution as eigenvalue problem

The coefficients **a** and **b** are obtained as eigenvectors from the equations

$$\boldsymbol{\Sigma}_{xx}^{-1} \boldsymbol{\Sigma}_{xy} \boldsymbol{\Sigma}_{yy}^{-1} \boldsymbol{\Sigma}_{yx} \mathbf{a} = \lambda_1 \mathbf{a}$$

$$\boldsymbol{\Sigma}_{yy}^{-1} \boldsymbol{\Sigma}_{yx} \boldsymbol{\Sigma}_{xx}^{-1} \boldsymbol{\Sigma}_{xy} \mathbf{b} = \lambda_1 \mathbf{b}$$

- Eigenvalues are squared canonical correlations
- $\min(p, q)$  successive uncorrelated canonical variates can be extracted
- The covariance matrices  $\boldsymbol{\Sigma}_{xx}$ ,  $\boldsymbol{\Sigma}_{xy}$  and  $\boldsymbol{\Sigma}_{yy}$  are estimated by the respective sample covariance or correlation matrices.

# Solution as a singular value decomposition

$$\mathbf{K} = \mathbf{R}_{xx}^{-1/2} \mathbf{R}_{xy} \mathbf{R}_{yy}^{-1/2} = \tilde{\mathbf{A}} \mathbf{D} \tilde{\mathbf{B}}'$$

- $\mathbf{D}$  diagonal matrix with canonical correlations
- $\mathbf{A} = \mathbf{R}_{xx}^{-1/2} \tilde{\mathbf{A}}$  matrix of canonical weights for  $x$  variables,  
 $\mathbf{A}' \mathbf{R}_{xx} \mathbf{A} = \mathbf{I}$ .
- $\mathbf{B} = \mathbf{R}_{yy}^{-1/2} \tilde{\mathbf{B}}$  matrix of canonical weights for  $y$  variables,  
 $\mathbf{B}' \mathbf{R}_{yy} \mathbf{B} = \mathbf{I}$ .
- Canonical  $x$  variates  $\mathbf{U} = \mathbf{X}_s \mathbf{A}$
- Canonical  $y$  variates  $\mathbf{V} = \mathbf{Y}_s \mathbf{B}$

# Biplots in Canonical correlation analysis

$$\mathbf{K} = \mathbf{R}_{xx}^{-1/2} \mathbf{R}_{xy} \mathbf{R}_{yy}^{-1/2} = \tilde{\mathbf{A}} \mathbf{D} \tilde{\mathbf{B}}'$$

$$\mathbf{F} = \mathbf{R}_{xx} \mathbf{AD} \quad \mathbf{G} = \mathbf{R}_{yy} \mathbf{B}$$

Joint plot of first two columns of  $\mathbf{F}$  and  $\mathbf{G}$  is a biplot of the between set correlation matrix.

$$\begin{aligned}\mathbf{FG}' &= \mathbf{R}_{xx} \mathbf{AD} (\mathbf{R}_{yy} \mathbf{B})' = \mathbf{R}_{xx} \mathbf{AD} \mathbf{B}' \mathbf{R}_{yy} = \mathbf{R}_{xx}^{1/2} \tilde{\mathbf{A}} \mathbf{D} \tilde{\mathbf{B}}' \mathbf{R}_{yy}^{1/2} \\ &= \mathbf{R}_{xx}^{1/2} \mathbf{R}_{xx}^{-1/2} \mathbf{R}_{xy} \mathbf{R}_{yy}^{-1/2} \mathbf{R}_{yy}^{1/2} = \mathbf{R}_{xy}\end{aligned}$$

Like in PCA, there are different scalings

$$\begin{aligned}\mathbf{F} &= \mathbf{R}_{xx} \mathbf{AD} & \mathbf{G} &= \mathbf{R}_{yy} \mathbf{B} \\ \mathbf{F} &= \mathbf{R}_{xx} \mathbf{A} & \mathbf{G} &= \mathbf{R}_{yy} \mathbf{BD} \\ \mathbf{F} &= \mathbf{R}_{xx} \mathbf{AD}^{1/2} & \mathbf{G} &= \mathbf{R}_{yy} \mathbf{BD}^{1/2}\end{aligned}$$

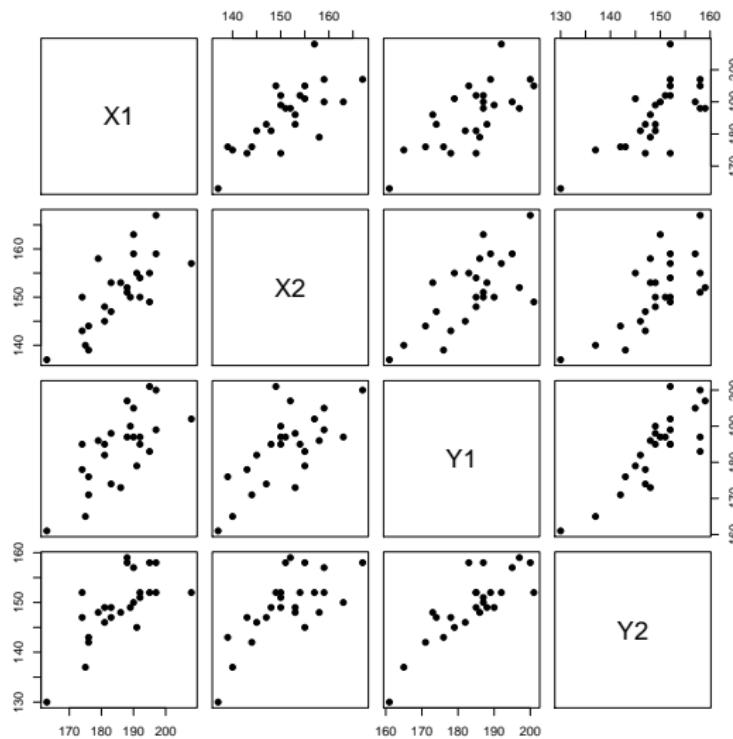
# Textbook example: Fret's (1921) data

- 25 families
- $x_1$  head length of first son
- $y_1$  head length of second son
- $x_2$  head breadth of first son
- $y_2$  head breadth of second son

Family	$x_1$	$x_2$	$y_1$	$y_2$
1	191	155	179	145
2	195	149	201	152
3	181	148	185	149
4	183	153	188	149
5	176	144	171	142
6	208	157	192	152
7	189	150	190	149
8	197	159	189	152
9	188	152	197	159
10	192	150	187	151
11	179	158	186	148
12	183	147	174	147
13	174	150	185	152
14	190	159	195	157
15	188	151	187	158
16	163	137	161	130
17	195	155	183	158
18	186	153	173	148
19	181	145	182	146
20	175	140	165	137
21	192	154	185	152
22	174	143	178	147
23	176	139	176	143
24	197	167	200	158
25	190	163	187	150

```
library(calibrate); data(heads);
```

# Scatterplot matrix



# Correlation matrix

	$X_1$	$X_2$	$Y_1$	$Y_2$
$X_1$	1.00	0.73	0.71	0.70
$X_2$	0.73	1.00	0.69	0.71
$Y_1$	0.71	0.69	1.00	0.84
$Y_2$	0.70	0.71	0.84	1.00

# Canonical variates

$$U_1 = 0.5522x_1 + 0.5215x_2$$

$$V_1 = 0.5044y_1 + 0.5383y_2$$

$$U_2 = -1.3664x_1 + 1.3784x_2$$

$$V_2 = -1.7686y_1 + 1.7586y_2$$

# Standard numerical output

		$r_1 = 0.7885$	$r_2 = 0.0537$	
		$U_1$	$U_2$	
$x_1$		0.5522	(0.9353)	-1.3664 (-0.3539)
$x_2$		0.5215	(0.9272)	1.3784 ( 0.3747)
Var. Expl.		0.8672	0.1328	
		$V_1$	$V_2$	
$y_1$		0.5044	(0.9562)	-1.7686 (-0.2927)
$y_2$		0.5383	(0.9616)	1.7586 ( 0.2743)
Var. Expl.		0.9195	0.0805	

# Goodness-of-fit of between-set correlation matrix

	1	2
eigenvalue	0.6217	0.0029
fraction	0.9954	0.0046
cumulative	0.9954	1.0000

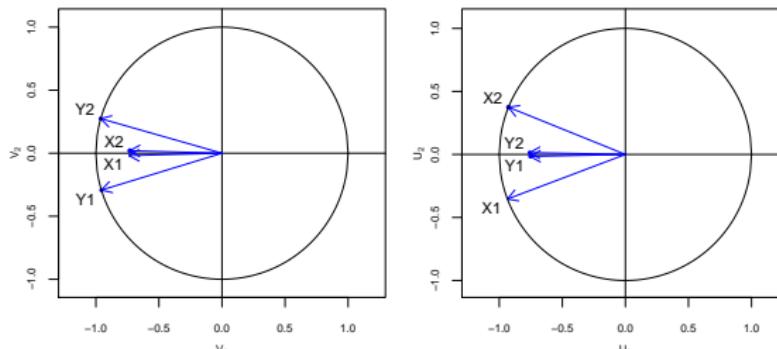
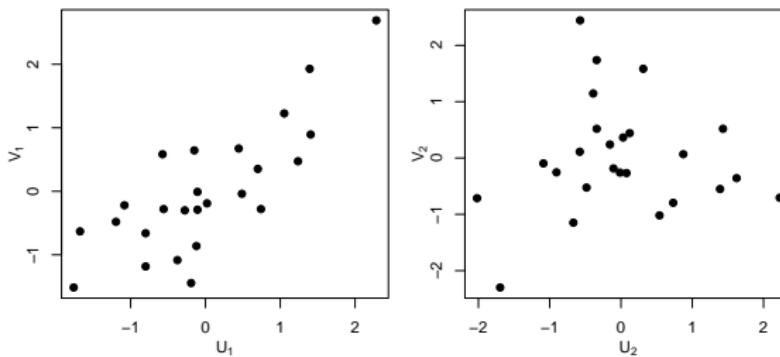
# Correlation matrix of the canonical variates

	$U_1$	$U_2$	$V_1$	$V_2$
$U_1$	1.00	0.00	0.79	0.00
$U_2$	0.00	1.00	0.00	0.05
$V_1$	0.79	0.00	1.00	0.00
$V_2$	0.00	0.05	0.00	1.00

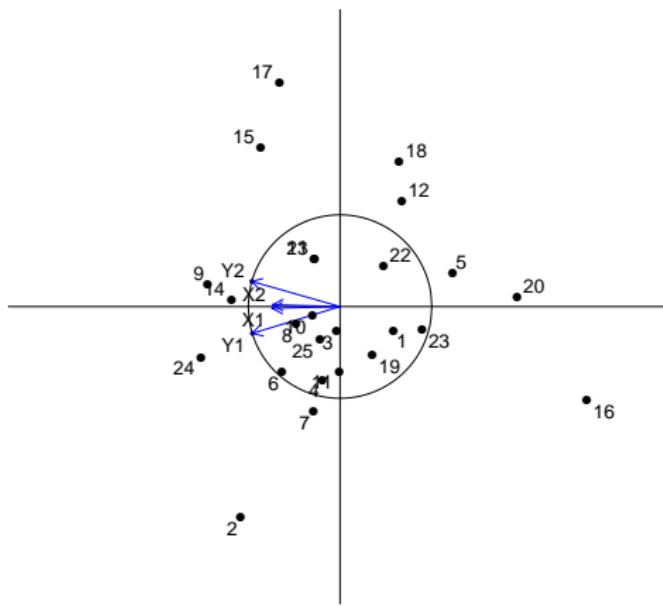
# Shared and explained variance

- Shared variance between X and Y variables: squared canonical correlations
- Adequacy coefficients
  - Variance in X set explained by canonical x-variates
  - Variance in Y set explained by canonical y-variates
- Redundancy coefficients
  - Variance in X set explained by canonical y-variates
  - Variance in Y set explained by canonical x-variates

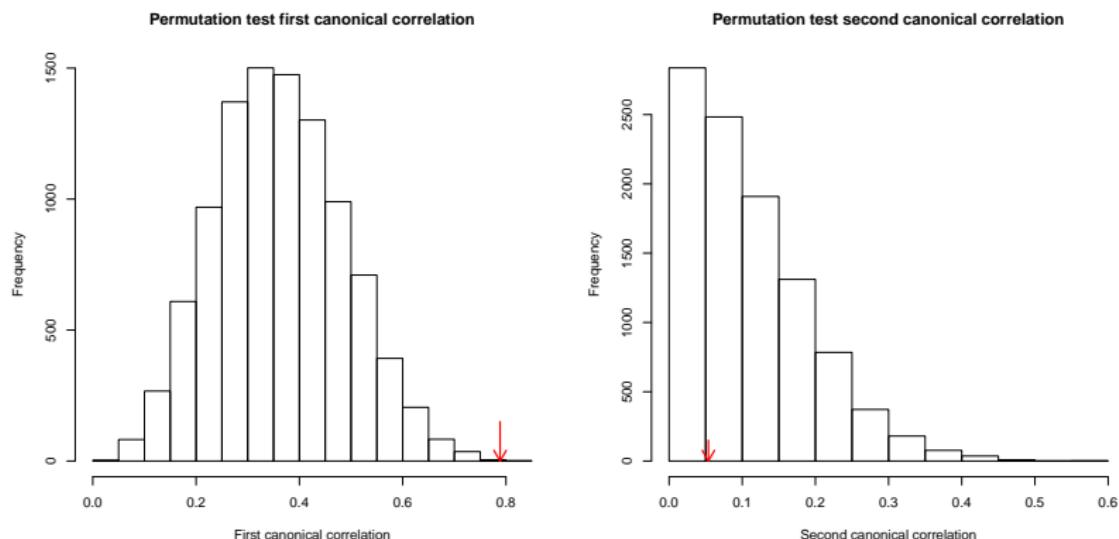
# Classical plots and Biplots



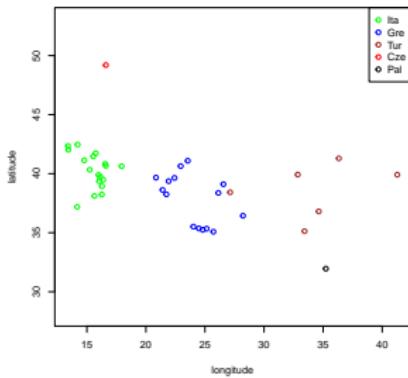
# Representing cases



# Significance of the canonical correlations



# Genetic example



- 41 samples, 16 STRs
- Do allele frequencies depend on geographic coordinates?
- We use D10S1248

Messina F, et al. (2016) Spatially Explicit Models to Investigate Geographic Patterns in the Distribution of Forensic STRs: Application to the North-Eastern Mediterranean. *PLoS ONE* 11(11): e0167065.

# The data

	Sampling location	Country/Island	Long.E	Lat.N	N	A9	A11	A12	A13	A14	A15	A16	A17	A18
1	Brno	Czech Rep. (Moravia)	16.61	49.20	49	0.00	0.00	0.04	0.24	0.31	0.24	0.10	0.05	0.01
2	L'Aquila	Italy (Abruzzo)	13.40	42.35	32	0.00	0.02	0.03	0.25	0.33	0.09	0.17	0.11	0.00
3	Avezzano	Italy (Abruzzo)	13.43	42.03	28	0.00	0.00	0.02	0.25	0.32	0.23	0.14	0.04	0.00
4	Pescara	Italy (Abruzzo)	14.22	42.46	18	0.00	0.06	0.08	0.33	0.28	0.14	0.11	0.00	0.00
5	Benevento	Italy (Campania)	14.78	41.13	45	0.00	0.01	0.06	0.22	0.34	0.20	0.12	0.04	0.00
6	Foggia	Italy (Apulia)	15.54	41.46	26	0.00	0.00	0.04	0.31	0.31	0.12	0.14	0.10	0.00
7	Gargano promontory	Italy (Apulia)	15.75	41.73	30	0.00	0.00	0.03	0.22	0.28	0.27	0.17	0.03	0.00
8	Cilento promontory	Italy (Campania)	15.25	40.33	44	0.00	0.00	0.03	0.26	0.29	0.24	0.12	0.04	0.00
9	Matera	Italy (Basilicata)	16.60	40.67	34	0.00	0.00	0.04	0.26	0.29	0.19	0.16	0.04	0.00
10	Altamura	Italy (Apulia)	16.55	40.83	20	0.00	0.00	0.00	0.25	0.32	0.25	0.17	0.00	0.00
11	Brindisi	Italy (Apulia)	17.94	40.63	93	0.00	0.00	0.02	0.28	0.34	0.18	0.10	0.06	0.01
12	Lungro	Italy (Calabria)	16.12	39.74	24	0.00	0.02	0.00	0.25	0.23	0.27	0.21	0.02	0.00
13	Acri	Italy (Calabria)	16.38	39.49	30	0.00	0.00	0.03	0.20	0.43	0.22	0.05	0.07	0.00
14	Mormanno	Italy (Calabria)	15.99	39.89	36	0.00	0.00	0.00	0.24	0.29	0.26	0.12	0.08	0.00
15	Paola	Italy (Calabria)	16.04	39.36	38	0.00	0.00	0.01	0.34	0.35	0.13	0.09	0.05	0.01
16	Lamezia	Italy (Calabria)	16.28	38.93	17	0.00	0.00	0.00	0.38	0.12	0.38	0.06	0.06	0.00
17	Loci	Italy (Calabria)	16.26	38.23	38	0.00	0.01	0.01	0.14	0.38	0.22	0.17	0.04	0.01
18	Reggio Calabria	Italy (Calabria)	15.65	38.11	41	0.00	0.04	0.02	0.26	0.29	0.17	0.17	0.05	0.00
19	Butera	Italy (Sicily)	14.18	37.19	24	0.00	0.00	0.04	0.25	0.40	0.21	0.08	0.02	0.00
20	Ioannina	Greece (Epirus)	20.85	39.66	25	0.00	0.00	0.00	0.18	0.26	0.30	0.14	0.10	0.02
21	Agrinion	Greece (Aetolia-Acarania)	21.41	38.62	22	0.00	0.00	0.00	0.32	0.29	0.20	0.14	0.04	0.00
22	Patrai	Greece (Akhaia)	21.73	38.25	62	0.00	0.01	0.02	0.18	0.33	0.22	0.23	0.02	0.00
23	Kardhitsa	Greece (Thessaly)	21.92	39.36	17	0.00	0.00	0.03	0.21	0.38	0.09	0.23	0.03	0.03
24	Larisa	Greece (Thessaly)	22.42	39.64	17	0.00	0.00	0.03	0.21	0.29	0.26	0.18	0.00	0.03
25	Thessaloniki	Greece (Central Macedonia)	22.94	40.64	15	0.00	0.00	0.03	0.17	0.43	0.30	0.07	0.00	0.00
26	Serrai	Greece (Central Macedonia)	23.54	41.09	27	0.00	0.00	0.04	0.17	0.43	0.24	0.09	0.04	0.00
27	Mitilini	Greece (Lesvos)	26.56	39.11	18	0.00	0.00	0.03	0.17	0.44	0.22	0.06	0.06	0.03
28	Khios	Greece (Khios)	26.14	38.37	30	0.00	0.00	0.00	0.35	0.37	0.15	0.10	0.03	0.00
29	Rhodos	Greece (Rhodos)	28.22	36.44	50	0.00	0.00	0.01	0.27	0.33	0.17	0.14	0.07	0.01
30	Crete (unspecified)	Greece (Crete)	24.81	35.24	65	0.00	0.01	0.04	0.20	0.33	0.20	0.16	0.06	0.00
31	Khania	Greece (Crete)	24.02	35.51	45	0.00	0.03	0.06	0.17	0.37	0.18	0.14	0.06	0.00
32	Rethymnon	Greece (Crete)	24.48	35.36	67	0.00	0.00	0.04	0.20	0.42	0.13	0.16	0.04	0.00
33	Iraklion	Greece (Crete)	25.14	35.34	48	0.00	0.02	0.03	0.18	0.41	0.16	0.17	0.03	0.01
34	Lasithi plateau	Greece (Crete)	25.71	35.08	38	0.00	0.00	0.07	0.14	0.43	0.18	0.10	0.07	0.00
35	Western Mediterranean coast	Turkey	27.13	38.42	54	0.00	0.00	0.03	0.20	0.34	0.30	0.09	0.04	0.00
36	Central Anatolia	Turkey	32.85	39.92	94	0.00	0.00	0.04	0.27	0.25	0.24	0.17	0.03	0.00
37	Black Sea coast	Turkey	36.33	41.29	42	0.00	0.00	0.06	0.23	0.29	0.23	0.14	0.06	0.00
38	Eastern Anatolia	Turkey	41.26	39.91	33	0.00	0.00	0.01	0.20	0.27	0.20	0.27	0.04	0.00
39	Eastern Mediterranean coast	Turkey	34.63	36.80	27	0.00	0.00	0.06	0.26	0.26	0.30	0.09	0.04	0.00
40	Cyprus	Turkey (Cyprus)	33.43	35.13	46	0.00	0.01	0.02	0.26	0.35	0.24	0.08	0.04	0.00
41	Palestine (CEPH)	Palestine	35.23	31.95	50	0.01	0.01	0.05	0.19	0.29	0.25	0.13	0.07	0.00

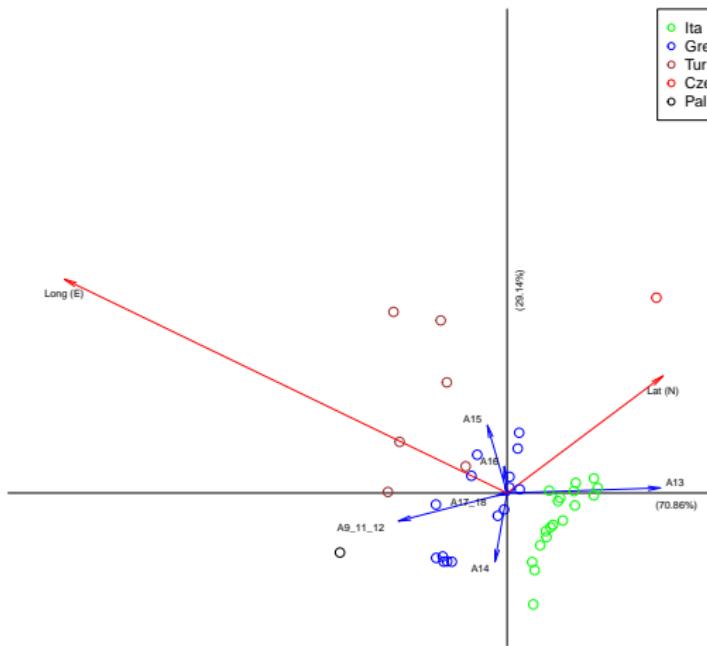
# The analysis

- Canonical analysis with  $\mathbf{X}$  = geographical coordinates,  $\mathbf{Y}$  = allele frequencies.
  - We treat the allele frequencies as compositional, and do the clr transformation.
  - We need to map samples to the biplot.
  - We need to accomodate for singularity of the covariance matrix of the clr transformed data.
- 
- Graffelman, J. (2005) Enriched biplots for canonical correlation analysis. *Journal of Applied Statistics* 32(2) pp. 173–188.
  - Graffelman, J., Pawlowsky-Glahn, V., Egozcue, J.J. Buccianti, A. (2018) Exploration of geochemical data with compositional canonical biplots. *Journal of Geochemical Exploration* 194 pp. 120–133.

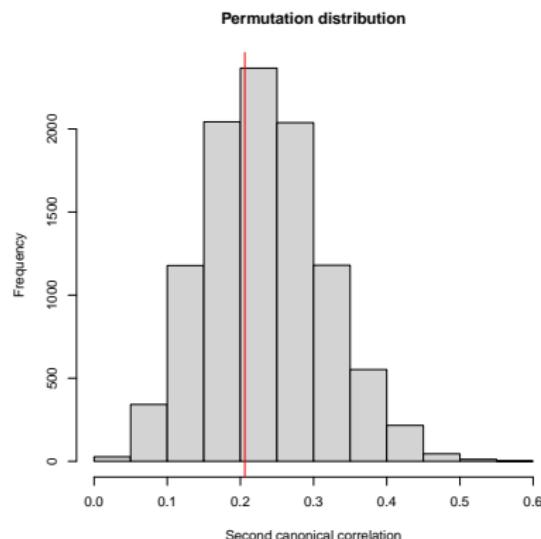
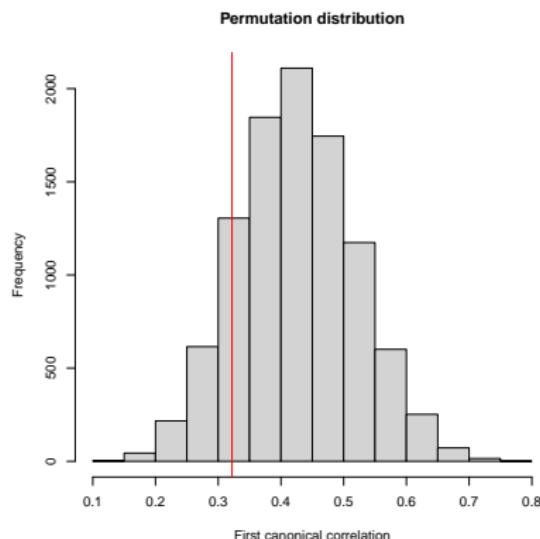
# Goodness-of-fit

	1	2
correlation	0.322	0.206
eigenvalue	0.103	0.043
fraction	0.709	0.291
cumulative	0.709	1.000

# The canonical biplot



# The permutation test



permutation p-values 0.8631 and 0.6116

# References

- Johnson & Wichern, (2002) Applied Multivariate Statistical Analysis, 5th edition, Prentice Hall, Chapter 10.
- Mardia, K.V. et al. (1979) Multivariate Analysis. Academic press. Chapter 10.