Correspondence Analysis

In practice

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Data

We consider the relationship between hair and eyes colors measured on 592 students in Delaware University.

```
mydata <- read.table("http://pbil.univ-lyon1.fr/R/donnees/snee74.txt",
    header = TRUE, stringsAsFactors = TRUE)
names(mydata)
```

```
## [1] "cheveux" "yeux"    "sexe"
```

```
head(mydata)
```

```
##    cheveux   yeux    sexe
## 1     Noir Marron    Male
## 2    Blond   Bleu Femelle
## 3     Noir   Bleu    Male
## 4   Marron Marron Femelle
## 5     Roux Marron    Male
## 6   Marron   Bleu    Male
```

Factors in R

A qualitative variables is stored as `factor` with different categories (`levels`)

```
is.factor(mydata$yeux)
```

```
## [1] TRUE
```

```
levels(mydata$yeux)
```

```
## [1] "Bleu"     "Marron"    "Noisette" "Vert"
```
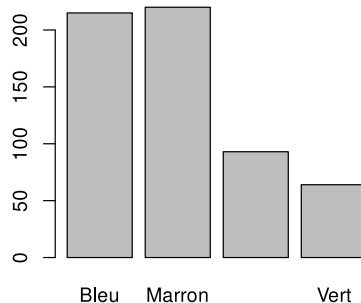
Univariate analysis

- Compute the vector with the number of individuals for each eye color category and display the results on a plot.

```
summary(mydata$yeux)
```

```
##     Bleu   Marron Noisette     Vert
##      215      220       93       64
```

```
plot(mydata$yeux)
```

Contingency table

- Build the contingency table crossing hair and eyes colors. See ?table

```
mytab <- table(mydata$cheveux, mydata$yeux)
```

- Compute row and columns marginal relative frequencies

```
rowSums(mytab)/sum(mytab)
```

```
##     Blond    Marron      Noir      Roux
## 0.2145270 0.4831081 0.1824324 0.1199324
```

```
colSums(mytab)/sum(mytab)
```

```
##      Bleu    Marron  Noisette      Vert
## 0.3631757 0.3716216 0.1570946 0.1081081
```

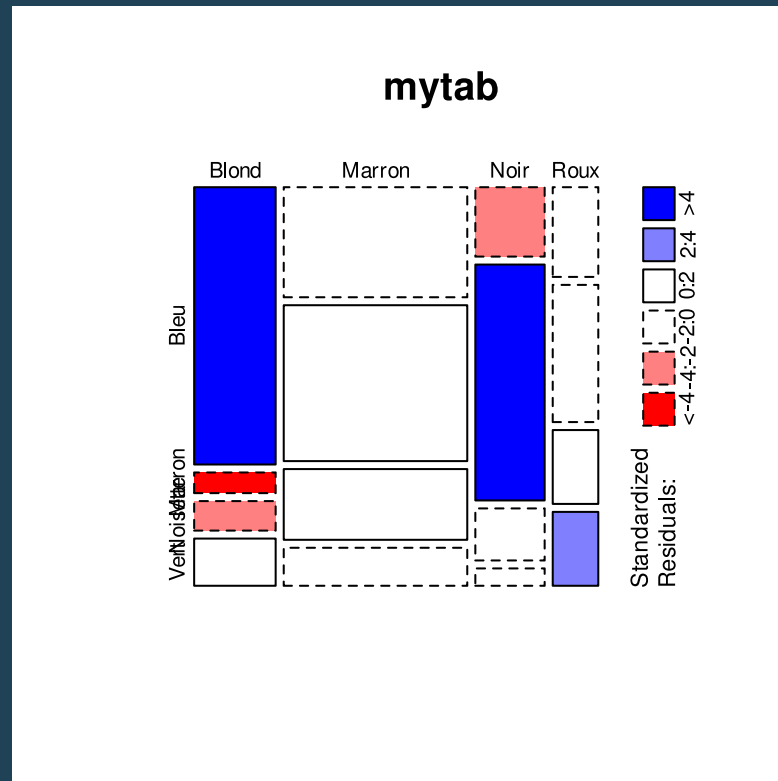# Bivariate analysis

- Compute and interpret the results of a χ^2 test

```
chisq.test(mytab)
```

```
##
##      Pearson's Chi-squared test
##
## data:  mytab
## X-squared = 138.29, df = 9, p-value < 2.2e-16
```
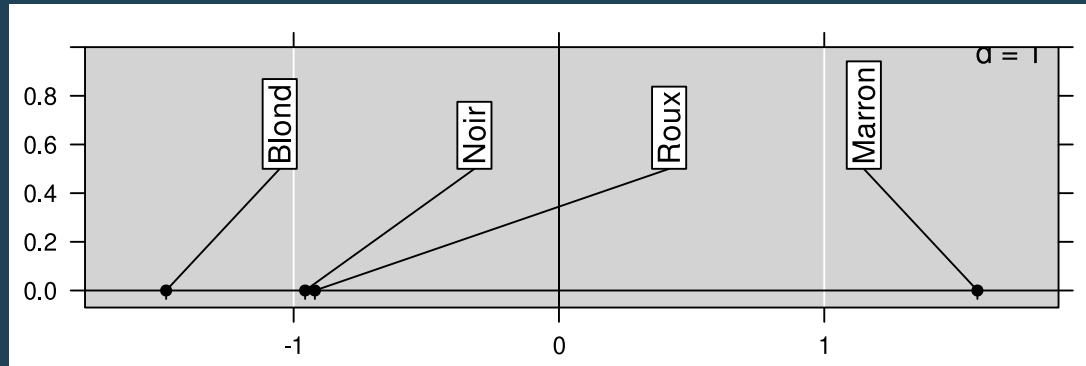
Bivariate analysis

- Display and interpret the associations between categories with `mosaicplot`

```
mosaicplot(mytab, shade = TRUE)
```
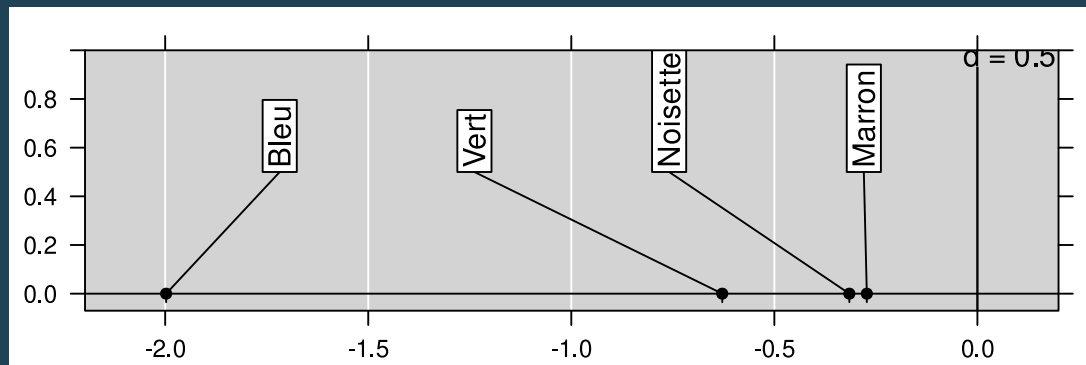
Scoring

Suppose that we can display hair colors by a score

what would be the best way to display eyes colors? (given the distribution of students)

Weighted averaging

- Assign a random score to hair colors

```
set.seed(12)
score.h <- rnorm(4)
```

- Compute the position for the eye color 'Vert'

```
sum(score.h * mytab[, 4]/sum(mytab[, 4]))
```

```
## [1] 0.06851621
```

```
## or

weighted.mean(score.h, w = mytab[, 4])
```

```
## [1] 0.06851621
```

Compute scores for all eye colors

By repeating the same formulas

```
score.e <- sapply(1:4, function(x) sum(score.h * mytab[,
    x]/sum(mytab[, x])))
```
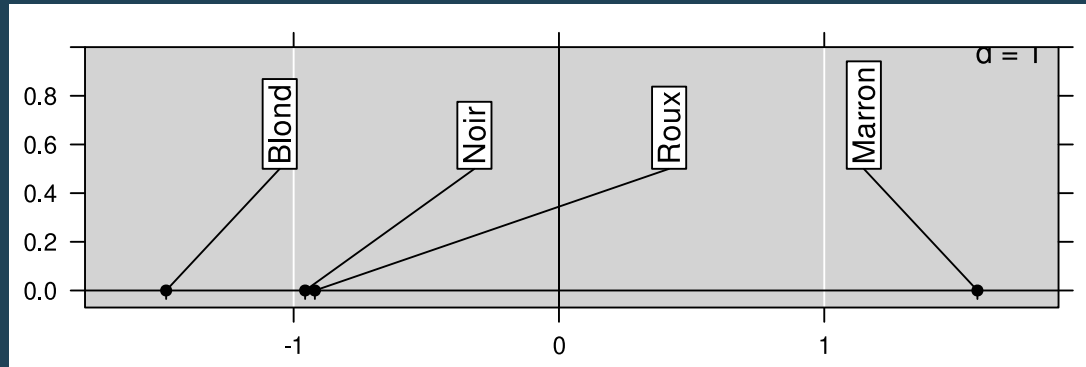
or using matrix algebra

```
t(prop.table(mytab, 2)) %*% score.h
```
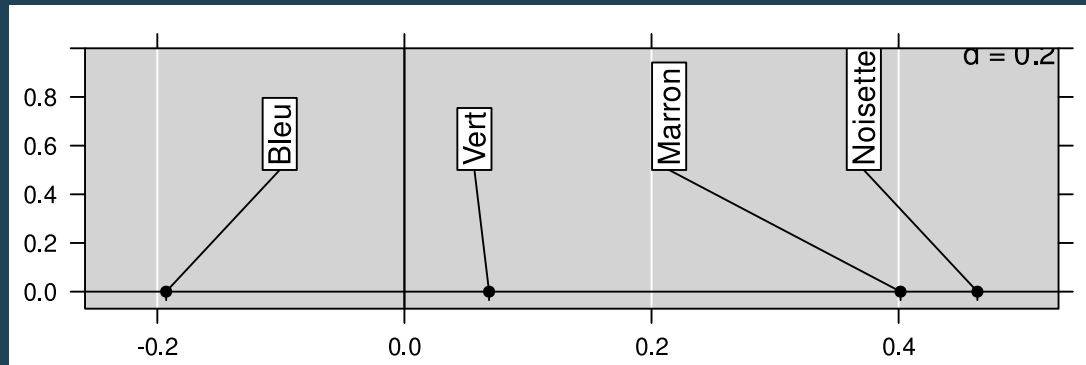
```
##
##                   [,1]
##   Bleu      -0.19286557
##   Marron     0.40154742
##   Noisette   0.46376597
##   Vert       0.06851621
```

Best representation

if

then

Reciprocal representation

- Start from a random score for eyes colors to position hair colors by averaging

```
set.seed(13)
score.e <- rnorm(4)
prop.table(mytab, 1) %*% score.e
```

```
##
##             [,1]
##   Blond  0.5582172
##   Marron 0.4003574
##   Noir   0.1814083
##   Roux   0.4170599
```

Reciprocal averaging

1. Set a random score for columns
2. Use the column score to compute a score for rows by weighted averaging
3. Compute a new score for columns by weighted averaging of row score
4. Center and scale the two scores using row and column weights
5. Repeat steps 2-4 until convergence

Correspondence analysis

The solution of the iterative algorithm can also be obtained by the diagonalization procedure provided by the `dudi.coa` function

```r
library(ade4)
coa1 <- dudi.coa(unclass(mytab), scannf = FALSE)
```

- Check the link between χ^2 statistic and total inertia

```r
sum(coa1$eig)
```

```
## [1] 0.2335977
```

```r
chisq.test(mytab)$statistic/sum(mytab)
```

```
## X-squared
## 0.2335977
```

Interpretation

- Compute a row score with unit norm ($l1) to obtain a column score by weighted averaging ($co) with maximal variance ($eig)

- Compute a row score with unit norm ($c1) to obtain a column score by weighted averaging ($li) with maximal variance ($eig)

- Check both results

```
prop.table(mytab, 1) %*% coa1$c1[, 1]
```

```
##
##                [,1]
##   Blond    0.8353478
##   Marron  -0.1482527
##   Noir    -0.5045624
##   Roux    -0.1295233
```

```
coa1$li[, 1]
```

```
## [1]   0.8353478 -0.1482527 -0.5045624 -0.1295233
```

Graphical representation

- Represent and interpret the results using the `scatter` function (check the `method` argument of `scatter.coa`)

```
scatter(coa1, method = 2)
```

Inertia statistics

- Use the `inertia.dudi` function to compute inertia statistics for rows and columns

```
ic = inertia.dudi(coa1, row.inertia = TRUE, col.inertia = TRUE)
```

- Represent absolute contributions using the `plot` function (see `?plot.inertia`)

```
plot(ic, contrib = "abs")
```