

Multifactorial Exploratory Approaches: multiple correspondence analysis

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Multifactorial Exploratory Approaches

multiple correspondence analysis

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outline

1 introduction





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MCA				

Because MCA is an extension of CA, its inner workings are very similar. For this reason, they are not repeated here.

nominal data

As pointed out yesterday, MCA takes as input a table of nominal data.

Table 1: A sample input table for MCA (Desagulier 2017, p. 36)

corpus file	mode	genre	exact match	intensifier	syntax	adjective
KBF.xml	spoken	conv	a quite ferocious mess	quite	preadjectival	ferocious
AT1.xml	written	biography	quite a flirty person	quite	predeterminer	flirty
A7F.xml	written	misc	a rather anonymous name	rather	preadjectival	anonymous
ECD.xml	written	commerce	a rather precarious foothold	rather	preadjectival	precarious
B2E.xml	written	biography	quite a restless night	quite	predeterminer	restless
AM4.xml	written	misc	a rather different turn	rather	preadjectival	different
F85.xml	spoken	unclassified	a rather younger age	rather	preadjectival	younger
J3X.xml	spoken	unclassified	quite a long time	quite	predeterminer	long
KBK.xml	spoken	conv	quite a leading light	quite	predeterminer	leading

beware of inertia

For MCA to yield manageable results, it is best if

- the table is of reasonable size (not too many columns)
- each variable does not break down into too many categories

Otherwise, the contribution of each dimension to ϕ^2 is small, and a large number of dimensions must be inspected(which kind of defeats the purpose)

beware of inertia

There are no hard and fast rules for knowing when there are too many dimensions to inspect.

However, when the eigenvalue that corresponds to a dimension is low, we know that the dimension is of little interest (the chances are that the data points will be close to the intersection of the axes in the summary plot).

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Schmid (2003) provides an analysis of sex differences in the 10M-word spoken section of the British National Corpus (BNC). Schmid shows that women use certain swear-words more than men, although swear-words which tend to have a perceived 'strong' effect are more frequent in male speech. Schmid's study is based on two subcorpora, which are both sampled from the spoken section of the BNC. The subcorpora amount to 8,173,608 words.

The contributions are not equally shared among men and women since for every 100 word spoken by women, 151 are spoken by men. To calculate the distinctive lexical preferences of men and women, while taking the lack of balance in the contributions into account, Schmid's measures rely on the difference coefficient.

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This formula is based on normalized frequencies per million words. Its score ranges from -1 (if a word occurs more frequently in women's utterances) to 1 (if a word occurs more frequently in male speech). Absolute frequencies are used to calculate the significance level of the differences using the hypergeometrical approximation of the binomial distribution. With respect to swear-words, Schmid's conclusion is that both men and women swear, but men tend to use stronger swear-words than women.

Schmid's study is repeated here in order to explore the distribution of swear-words with respect to gender in the BNC-XML. The goal is to see if:

- men swear more than women;
- some swear-words are preferred by men or women;
- the gender-distribution of swear-words is correlated with other variables: age and social class.

The data file for this case study is swearwords_bnc.txt.

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The code for the extraction was partly contributed by Mathilde Léger, a third-year student at Paris 8 University, as part of her end-of-term project. Unlike Schmid, and following Rayson et al. (1997), the data are extracted from the demographic component of the BNC-XML, which consists of spontaneous interactive discourse. The swear-words are: *bloody*, *damn*, *fuck*, *fucked*, *fucker*, *fucking*, *gosh*, and *shit*. Two exploratory variables are included: age and social class.

- > # clear R's memory
- > rm(list=ls(all=TRUE))
- > #load FactoMineR
- > library(FactoMineR)
- > # load the data (choose swearwords_bnc.txt)
- > df <- read.table(file=file.choose(), header=TRUE, sep="\t")</pre>

The data set contains 293,289 swear-words. These words are described by three categorical variables (nominal data):

- gender (2 levels: male and female)
- age (6 levels: Ag0, Ag1, Ag2, Ag3, Ag4, Ag5)
- social class (4 levels: AB, C1, C2, DE)

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Age breaks down into 6 groups:

- Ag0: respondent age between 0 and 14;
- Ag1: respondent age between 15 and 24;
- Ag2: respondent age between 25 and 34;
- Ag3: respondent age between 35 and 44;
- Ag4: respondent age between 45 and 59;
- Ag5: respondent age is 60+.

Social classes are divided into 4 groups:

- AB: higher management: administrative or professional.
- C1: lower management: supervisory or clerical;
- C2: skilled manual;
- DE: semi-skilled or unskilled.

It is advisable to keep an eye on the number of levels for each variable and see if any can be kept to a minimum to guarantee that inertia will not drop.

```
> str(df)
'data.frame': 293289 obs. of 4 variables:
 $ word : Factor w/ 8 levels "bloody", "damn", ...: 2 2 7 7 7 2 7 2 7 7 ...
 $ gender : Factor w/ 2 levels "f", "m": 2 2 2 2 2 2 2 2 2 2 ...
            : Factor w/ 6 levels "Ag0", "Ag1", "Ag2",...: 6 6 6 6 6 6 6 6 6 6 ...
 $ age
 $ soc_class: Factor w/ 4 levels "AB", "C1", "C2",..: 1 1 1 1 1 1 1 1 1 ...
> table(df$word)
 bloody
           damn
                   fuck fucked fucker fucking
                                                            shit
                                                   gosh
 146203
          32294
                   9219
                             11
                                    467
                                           23487
                                                   60678
                                                           20930
```

We can group *fuck*, *fucking*, *fucked*, and *fucker* into a single factor: *f*-words. With gsub(), we replace each word with the single tag f-words.

> df\$word <- gsub("fuck|fucking|fucker|fucked", "f-words", df\$word, ignore.case=TRUE)
> table(df\$word)
bloody damn f-words gosh shit
146203 32294 33184 60678 20930

We convert df\$word back to a factor. The number of levels has been reduced to five.

```
> df$word <- as.factor(df$word)</pre>
```

As in CA, we can declare some variables as active and some other variables as supplementary/illustrative in MCA. We declare the variables corresponding to swear words and gender as active, and the variables age and social class as supplementary/illustrative. Running a MCA involves the following steps:

- determining how many dimensions there are to inspect;
- interpreting the MCA graph.

We run the MCA with the MCA() function. We declare age and soc_class as supplementary (quali.sup=c(3,4)). We do not plot the graph yet (graph=FALSE).

> mca.object <- MCA(df, quali.sup=c(3,4), graph=FALSE)</pre>

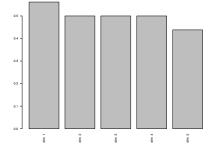
Again, the eig object allows us to see how many dimensions there are to inspect.

> rou	und(mca.obje	ct\$eig, 2)					
	eigenvalue	percentage	of variance	cumulative	percentage	of variance	
dim 1	0.56		22.47			22.47	
dim 2	2 0.50		20.00			42.47	
dim 3	0.50		20.00			62.47	
dim 4	0.50		20.00			82.47	
dim 5	0.44		17.53			100.00	

The number of dimensions is rather large and the first two dimensions account for only 42.47% of ϕ^2 . To inspect a significant share of ϕ^2 , e.g. 80%, we would have to inspect at least 4 dimensions. This issue is common in MCA. The eigenvalues can be vizualized by means of a scree plot. It is obtained as follows.

```
> barplot(mca.object$eig[,1],
+ names.arg=paste("dim ", 1:nrow(mca.object$eig)), las=2)
```

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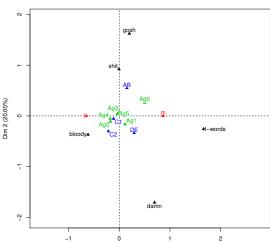


Ideally, we would want to see a sharp decrease after the first few dimensions, and we would want these first few dimensions to account for as much share of ϕ^2 as possible. Here, no sharp decrease is observed.

The MCA map is plotted with the plot.MCA() function. Each category is the color of its variable (habillage="quali"). The title is removed (title="").

```
> plot.MCA(mca.object,
+ invisible="ind",
+ autoLab="yes",
+ shadowtext=TRUE,
+ habillage="quali",
+ title="")
```

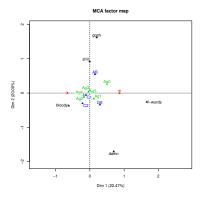
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MCA factor map

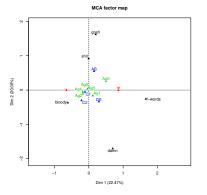
Dim 1 (22.47%)

How men and women swear in the BNC-XML $_{\rm dim \ 1}$



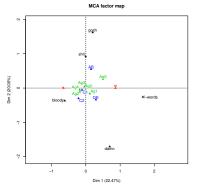
Strikingly, the most explicit swear words (*f*-words) cluster in the right-most part of the plot. These are used mostly by men. Female speakers tend to prefer a softer swear word: *bloody*.

How men and women swear in the BNC-XML $_{\rm dim\ 2}$



Words in the upper part (*gosh* and *shit*) are used primarily by upper-class speakers. *F*-words, *bloody*, and *damn* are used by lower social categories. Age groups are positioned close to the intersection of the axes. This is a sign that the first two dimensions bring little or no information about them.

How men and women swear in the BNC-XML $\dim 1 + \dim 2$



we observe 3 distinct clusters:

cluster 1 (upper-right corner)
gosh and shit, used by male and
female upper class speakers;

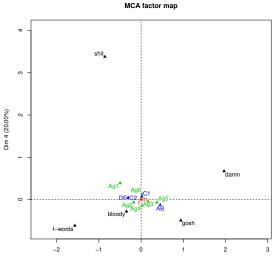
cluster 2 (lower-left corner) bloody, used by female middle-class speakers;

cluster 3 (lower-right corner) *f*-words and *damn*, used by male lower-class speakers.

A divide exists between male (m, right) and female (f, left) speakers. However, as the combined eigenvalues indicate, we should be wary of making final conclusions based on the sole inspection of the first two dimensions. The relevance of age groups becomes more relevant if dimensions 3 and 4 are inspected together

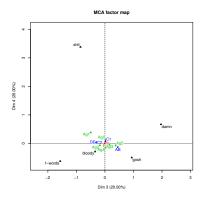
```
> plot.MCA(mca.object,
+ axes=c(3,4),
+ invisible="ind",
+ autoLab="yes",
+ shadowtex=TRUE,
+ habillage="quali",
+ title="")
```

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Dim 3 (20.00%)

How men and women swear in the BNC-XML $\dim 3 + \dim 4$

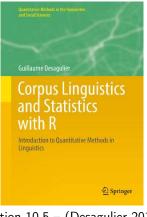


- the male/female distinction disappears
- a divide is observed between f-words and bloody (left), used mostly by younger and middle-aged speakers, and gosh and damn (right), used mostly by upper-class speakers from age groups 3 and 5.
- the most striking feature is the outstanding position of *shit* in the upper-left corner.

Practical Handbook of Corpus Linguistics

Guillaume Desagulier (to appear). "Multifactorial exploratory approaches." In: *Practical Handbook of Corpus Linguistics*. Ed. by Magali Paquot and Stefan Thomas Gries. New York: Springer

Corpus Linguistics and Statistics with R



Section 10.5 - (Desagulier 2017)

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Dibilography I

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