

Data Analysis of
Transport
Systems

Block 3. Computational Statistical Inference

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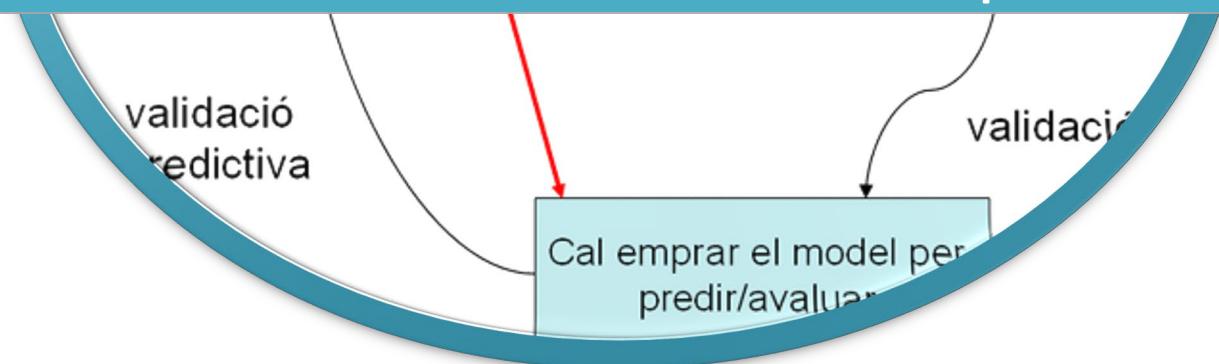


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3-1. STATISTICAL INFERENCE AND COMPUTATIONAL EXPLORATION OF LARGE TABLES (DATABASE)

A single response variable is assumed, quantitative or qualitative, the paradigm is the same, the techniques differ:

- Sort systematically all explanatory variables according to their degree of association with the response variable.
- Tool description: contrast of hypothesis (statistical inference).
- The association can be defined:
 1. A global variable to variable: variables that best explain the response and to what degree.
 2. For groups (= between levels of categorical variables involved, whether response or explanatory).

It will be seen from the computational point of view R:

- Global association between quantitative resp. and a quantitative explanation (covariate).
- Global association between quantitative resp. and an explanatory resp. with I levels (factor).
- Global association between qualitative resp. with J levels and an explanatory resp. with I levels.
- Association between quantitative resp. and every level of an explanatory with I levels.
- Associating each J qualitative resp. level and each I explanatory resp. levels.

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

3-1.1 Contrast of hypothesis

Working Paradigm: Formulation H₀ (null hypothesis) and interpretation associated p value of contrast.

H₀: "The X variable is not associated with Y"

1. Depending on the nature of H₀ (type of the variables involved) will contrast with a statistical method or another.
2. All contrast provides a p value to be interpreted "Probability that the null hypothesis is true, if the conditions of applicability of contrast are given".
3. Calculating the p value according to the probability distribution of the statistical reference.

Interpretation of the p value:

- If p value is small (less than a threshold of 5 or 10%) then H₀ is rejected, it is not credible and therefore evidence that its falsehood and it is claimed that there is evidence to believe that the variable Y is associated with the variable X.
- If p value is large (greater than a threshold of 5 or 10%) then H₀ accepts and therefore there is no evidence that its falsity and hence the assertion that there is no evidence to believe that the variable Y is associated with the variable X.

3-1 ... INTERPRETATION OF TEST: LEVEL OF SIGNIFICANCE

Level of significance α

$$\alpha = P(\text{reject } H_0 \mid H_0 \text{ true})$$

Frequently, α is set to 0.01 or 0.05
(Hypothesis)

	Actually True	Actually False
Accept	$1 - \alpha$	β (Type II error)
Reject	α (Type I error)	$1 - \beta$

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

3-1.2 Global association between quantitative resp. and a quantitative explicative var.

- Measure the linear correlation between the 2 variables. You can test the null hypothesis of linear correlation coefficient equal to zero.
- The continuous explanatory variables would be sorted by their p-value in the contrast of hypothesis, low to high. Low p value indicates greater confidence in rejecting the null hypothesis and therefore greater association between the response and variable.

Parametric tests - Non parametric (method="spearman"): a universal example, weight for height.

```

# Continuous explanatory
attach(davis)
plot(weight ~ height, data=davis)
# Calculate the linear correlation coefficient between 2 quantitative variables involved
cor(davis$weight, davis$height, use="pairwise.complete.obs" )
# Calculate the linear correlation coefficient between all quantitative variables involved
cor(data.frame(weight, height,r_weight, r_height), use="pairwise.complete.obs" )
# cor(x, y, use = "all.obs", method = c("pearson", "kendall", "spearman"))
# cor.test() Methode contrast correlation equal to 0:
# cor.test(x, y,alternative = c("two.sided","less","greater"), exact = TRUE, conf.level = 0.95)
# "spearman"),exact = NULL, conf.level = 0.95
cor.test(weight ~ height, data=davis)

# P.value attribute contains p value
cor.test(weight ~ height, data=davis)

```

```

> cor.test(davis$weight,davis$height)
Pearson's product-moment correlation
data: davis$weight and davis$height
t = 17.0397, df = 198, p-value < 2.2e-16
alternative hypothesis: true correlation is not equal to
0
95 percent confidence interval:
0.7080838 0.8218898

```

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

3-1.3 Global association between quantitative resp. and an explicative variate with I levels (factor)

- Acceptable normality assumption, essential independence and identical variance observations by groups (defined by levels, if not met R resetting G. L. test).
- Automated tools related to models analysis of variance with a factor (one-way). If $I = 2$ then it is a contrast of means from 2 subpopulations in unpaired samples. It is possible to relax the assumption of equal variance in the I subsamples reducing degrees of freedom of reference distributions.

Parametric tests with dichotomous factor: a universal example, weight by gender.

```
# Dichotomous explanatory
options(contrasts=c("contr.treatment","contr.treatment"))
attach(davis)
plot(weight ~ sex, data=davis)
plot.design(davis)
# Contrast equal variances: there are specific contrasts
var.test(weight ~ sex, data=davis)
# t.test() methode bilateral contrast with different variances (default)
t.test(weight ~ sex, data=davis)
```

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

```
# t.test() methode contrast unilateral with different variances (default)
t.test(weight ~ sex, data=davis, alternative="greater")
t.test(weight ~ sex, data=davis, alternative="less")
# Equal variance contrast: generalizing of var.test() I > 2
bartlett.test(weight ~ sex, data=davis)
# One-way
oneway.test(weight ~ sex, data=davis)
```

```
pes <- oneway.test(weight ~ sex, data=davis, var.equal=FALSE)
```

```
# One-way: obtaining p_valor accessing attribute p.value
oneway.test(weight ~ sex, data=davis)$p.value
```

- The explanatory variables available should be related to the p.valor: low to high. A lower p value, the greater degree of association.
- For enthusiasts of nonparametric mean contrasts: Kruskal-Wallis test, kruskal.test() in R.
- For non-parametric tests of equality of variances in subgroups: Fligner test, fligner.test() in R.

- FactoMineR package. Use condes(X, num.var) and check \$quali output. Profiling tool.

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

Test for means ($Y \sim A$)

Normal Y

$A | = 2 \rightarrow t.test(Y \sim A)$

$A | > 2 \rightarrow oneway.test(Y \sim A)$

Non normally distributed Y

$A | = 2 \rightarrow wilcox.test(Y \sim A)$

$A | > 2 \rightarrow kruskal.test(Y \sim A)$

Test for dispersion in groups defined by A ($Y \sim A$)

Normal Y

$A | = 2 \rightarrow var.test(Y \sim A)$

$A | > 2 \rightarrow bartlett.test(Y \sim A)$

Non normally distributed Y

$A | >= 2 \rightarrow fligner.test(Y \sim A)$

Assume H_0 is rejected for a test on means \rightarrow There is at least one group with a mean that is not the common mean.

Pairwise.t.test(Y, A) - Y Normally distributed

Pairwise.wilcox.test(Y, A) - Y any

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

3-1.4 Global association between qualitative resp. with J levels and an explicative var with I levels

- Statistically related to the analysis of contingency tables.
- Habitual Convention: response variable in columns and rows explanatory.
- You can make tables 1, 2 and more dimensions in R, but we focus on the current paragraph and therefore 2 dimensional tables.
- Total observations concept (n), univariate row total ($n_{i\bullet}$), univariate columns total ($n_{\bullet j}$).
- Marginal rows probability ($p_{i\bullet}$), marginal columns ($p_{\bullet j}$), bivariate marginal (p_{ij}).

	1	\dots	J	
1	\vdots	\vdots	\vdots	$n_{1\bullet}$
\vdots	\dots	n_{ij}	\dots	$n_{i\bullet}$
I	\vdots	\vdots	\vdots	$n_{I\bullet}$
	$n_{\bullet 1}$	$n_{\bullet j}$	$n_{\bullet J}$	n

$$p_{ij} = \frac{n_{ij}}{n} \quad p_{i\bullet} = \frac{n_{i\bullet}}{n} \quad p_{\bullet j} = \frac{n_{\bullet j}}{n}$$

B~A

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

Absence of association contrast:

$$H_0 : n_{ij} = n \cdot p_{i\bullet} \cdot p_{\bullet j}$$

$$X^2 = \sum_{i=1}^I \sum_{j=1}^J \frac{(o_{ij} - e_{ij})^2}{e_{ij}} = \sum_{i=1}^I \sum_{j=1}^J \frac{(n_{ij} - n \cdot p_{i\bullet} \cdot p_{\bullet j})^2}{n \cdot p_{i\bullet} \cdot p_{\bullet j}} \approx \chi^2_{(I-1)(J-1)} \rightarrow P(\chi^2_{(I-1)(J-1)} > discrepancy) = pvalue$$

- If the number of observations in each cell is nonzero (and best practice as more than 5), the null hypothesis of no association between the explanatory variable and the response variable, which is equivalent at independence between rows and columns, can be contrasted by **Pearson statistic χ^2** , whose **reference distribution** is a Chi-squared: could ask for the calculation of **p value** of the null hypothesis, if the p value is less than $\alpha = 0.05$ then there is evidence to reject the null hypothesis and if the p-value is above the threshold $\alpha = 0.05$ then there is evidence to accept the null hypothesis.
- **FactoMineR package. Use `catdes(X, num.var)` and check `$test.chi2` output**

```
> chisq.test(table(edad,residencia))
Pearson's Chi-squared test data: table(edad, residencia)
X-squared = 6.4, df = 6, p-value = 0.3799
> attributes(chisq.test(table(edad,residencia)))
$names
[1] "statistic" "parameter" "p.value"      "method"       "data.name"   "observed"    "expected"
"residuals"
```

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

- For each variable with qualitative treatment (well to be a factor initially or by discretization of the original numerical variable) can calculate the p value of the hypothesis of no association between levels of this and the J levels of the response variable.
- Qualitative variables can be listed sorted by p value low to high.
- FactoMineR package. Use `cattes(X, num.var)` and check `$test.chi2` output

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

3-1.5 Association between quantitative resp. and every level of an explanatory with I levels

Given an explanatory with I levels, the issue will be detect the level with the highest association with the response, that is, those whose null hypothesis of no association has a probability (p value) smaller and thus points more significantly to the rejection of the null hypothesis.

- If for each qualitative explanatory variable are created virtually as many levels as auxiliary variables, so that for $k = 1$ to I , the null hypothesis is postulated "no difference in the population mean of response between the groups defined by the level k other levels".
- It would make as many contrasts as detailed in paragraph 3-1.3Error! Reference source not found. as levels I, well now the variable that defines the group is dichotomous *level k against all levels less k* and therefore better to use a specialization of the contrast of the population mean of two subpopulations acceptably normal by t-test for unpaired subsamples. Be:
 - \bar{y}_k the mean response in the subset defined by the category (level k) the treated explanatory variable.
 - \bar{y} mean $\left(\bar{y} = \sum_{l=1}^n \frac{y_l}{n} \right)$ and s the sample standard deviation of the response $\left(s^2 = \sum_{l=1}^n \frac{(y_l - \bar{y})^2}{n-1} \right)$.
 - n_k number of observations belonging to group k and n number of observations.

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

The contrast has described a statistical description:

$$H_0 : \mu_k = \mu \text{ for } k=1 \dots I$$
$$t = \frac{(\bar{y}_k - \bar{y})}{\sqrt{\left(1 - \frac{n_k}{n}\right) \cdot \frac{s^2}{n_k}}} \approx \text{Student } t_{n-1}$$

The computational procedure is:

- For each of the explanatory variables with qualitative treatment and for each level k of the explanatory calculate the p value of the null hypothesis of equal means in response according to the groups defined by level k and all levels.
- The result is a vector of lists, with many lists as explanatory variables within each list and many had p values as explanatory levels. They agree to give a matrix structure that allows sort lists low to high p value. Detects modalities (or levels) discriminant, that is, very informative to characterize the response.
- **FactoMineR package. Use condes(X, num.var) and check \$category output**

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

3-1.6 Associating each J qualitative response level and each I explanatory response levels

It can be statistically quantified using the normal approximation to the contrast of clusters resulting contingency table defining groups based on rows (k vs. rest) and groups of columns (the l vs. rest) of the response. Arguably following a similar reasoning to the previous paragraph but in the context of contingency tables.

For each level $l=1..J$ of the response determine what level of explanatory k is more associated with $k=1..I$, by calculating the p value of:

$$H_0 : p_{l/k} = p_{\bullet l} \quad \forall k, \forall l \quad \text{where } p_{l/k} = \frac{n_{kl}}{n_{k\bullet}} \text{ and } X^2 = \sum_{i=k \text{ vs rest}} \sum_{j=l \text{ rest}} \frac{(o_{ij} - e_{ij})^2}{e_{ij}} \approx \chi^2_{(2-1) \cdot (2-1)} = \chi^2_1,$$

therefore its square root is a standard normal and specializing summation:

$$\frac{n_{kl}}{n_{k\bullet}} \approx N\left(\frac{n_{\bullet l}}{n}, (1-p_{k\bullet}) \frac{p_{\bullet l}(1-p_{\bullet l})}{n_{k\bullet}}\right) \quad \text{and then,} \quad z_{kl} = \frac{\frac{n_{kl}}{n_{k\bullet}} - \frac{n_{\bullet l}}{n}}{\sqrt{(1-p_{k\bullet}) \frac{p_{\bullet l}(1-p_{\bullet l})}{n_{k\bullet}}}} \approx N(0,1)$$

3-1 ... EXPLORATION OF LARGE TABLES (DATABASE)

The result of the calculated p value results (a possibility) a matrix structure with many rows and columns as the explanatory levels (I) and the response (J) and containing the p value of level i of the explanatory ratio does not affect the response of the response level j. If they said matrix sorts by the first dimension discriminates the most explanatory categories for each category of the response appear.

- FactoMineR package. Use `cattes(X, num.var)` and check `$category` output

Now it will proceed to the previous paragraphs illustrate with *CBD Prizing* example.

3-2. CBD PRIZING EXAMPLE: ACCEPTANCE OF TRAFFIC AND MOVEMENT MEASURES AIMED AT REDUCING POLLUTANTS EMISSIONS

Determine a decision rule for accepting traffic and circulation measures aimed at reducing emissions (could come from a survey about the congestion charge applied to central London).

The hypothetical function score should classify individuals into three zones: green (light acceptance), orange (Near doubt) and red (light rejection).

Determine the percentage of acceptances, the percentage of actual rejections who have been labeled as clear acceptance and actual percentage of acceptances who have been labeled a clear rejection.

It features a sample of individuals that has collected information on socioeconomic characteristics and review. It has a data file and the following metadata:

- 99999999 value in the continuous variables indicating missing value.

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

- data dictionary or features of the data matrix:

dictamen (Real review) (response variable)

1 positive / acceptance

2 negative / rejection

anys.feina (seniority) (years)

habitatge (Dwelling)

1 rental

2 writing published

3 private contract

4 ignores contract

5 parents

6 others

edat (Age)

estat.civil (marital status)

1 single

2 married

3 widower

4 separate

5 divorced

tipus.feina (type of work)

1 empleado fijo

2 empleado temporal

3 autonomo

4 otros

despeses (Expenditure) (thousands €)

ingressos (Income) (thousands €)

patrimoni (Heritage) (thousands €)

carregat.patr (Heritage carryforwards) (thousands €)

import.assoc (Loans requested) (thousands €)

Plazo del préstamo más largo (Longer loan term) (months)

preu.final (Value of assets financed) (thousands €)

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

3-2.1 Global debugging more categorization of quantitative variables

```
opina <- read.table("kioto.txt", header=T, sep='\t', na.string='99999999')
```

```
summary(opina)
```

```
> summary(opina)
```

dictamen	anys.feina	habitatge	plas	edat	estat.civil
Min. :0.000	Min. : 0.000	Min. :0.000	Min. : 6.00	Min. :18.00	Min. :0.000
1st Qu.:1.000	1st Qu.: 2.000	1st Qu.:2.000	1st Qu.:36.00	1st Qu.:28.00	1st Qu.:2.000
Median :1.000	Median : 5.000	Median :2.000	Median :48.00	Median :36.00	Median :2.000
Mean :1.281	Mean : 7.987	Mean :2.657	Mean :46.44	Mean :37.08	Mean :1.879
3rd Qu.:2.000	3rd Qu.:12.000	3rd Qu.:4.000	3rd Qu.:60.00	3rd Qu.:45.00	3rd Qu.:2.000
Max. :2.000	Max. :48.000	Max. :6.000	Max. :72.00	Max. :68.00	Max. :5.000
registres	tipus.feina	despeses	ingressos	patrimoni	carrega.patr
Min. :1.000	Min. :0.000	Min. : 35.00	Min. : 0.0	Min. : 0	Min. : 0.0
1st Qu.:1.000	1st Qu.:1.000	1st Qu.: 35.00	1st Qu.: 80.0	1st Qu.: 0	1st Qu.: 0.0
Median :1.000	Median :1.000	Median : 51.00	Median :120.0	Median : 3000	Median : 0.0
Mean :1.174	Mean :1.676	Mean : 55.57	Mean :130.6	Mean : 5403	Mean : 342.9
3rd Qu.:1.000	3rd Qu.:3.000	3rd Qu.: 72.00	3rd Qu.:165.0	3rd Qu.: 6000	3rd Qu.: 0.0
Max. :2.000	Max. :4.000	Max. :180.00	Max. :959.0	Max. :300000	Max. :30000.0
			NA's : 34.0	NA's : 47	NA's : 18.0
import.assoc	preu.final				
Min. : 100	Min. : 105				
1st Qu.: 700	1st Qu.: 1118				
Median :1000	Median : 1400				
Mean :1039	Mean : 1463				
3rd Qu.:1300	3rd Qu.: 1692				
Max. :5000	Max. :11140				

```
save.image("opina_raw.RData")
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

DE between uni and bivariate categorical

```

> # dictamen: obs 3310 is 0, I delete it!
> opina <- opina[opina$dictamen!=0,]
> table(opina$dictamen)
  1   2
3200 1254
> opina$f.dictamen <- factor(opina$dictamen,levels=c("2","1"),labels=c("rebutja","accepta"))
> summary(opina$f.dictamen)
rebutja accepta
  1254   3200

```

Refined and discretization of "anys feina"

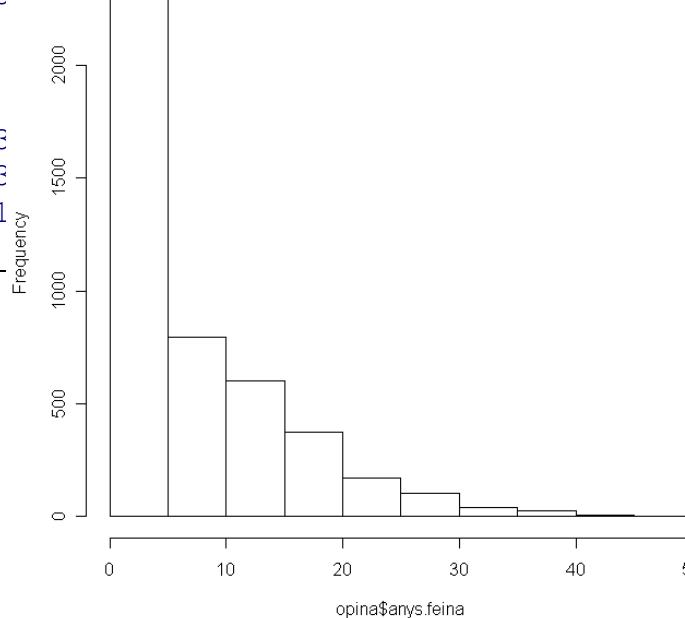
```

> summary(opina$anys.feina)
  Min. 1st Qu. Median     Mea
 0.000  2.000  5.000  7.98
> table(opina$anys.feina)

  0   1   2   3   4   5   6
535 510 454 336 233 266 181 13
  25  26  27  28  29  30  31  3
  62  14  14  14  11  49  10  1
> hist(opina$anys.feina)

```

Histogram of opina\$anys.feina



Bin Range (opina\$anys.feina)	Frequency
17-18	56
18-19	65
19-20	39
20-21	151
21-22	23
22-23	41
23-24	26
24-25	19
25-26	42
26-27	43
27-28	45
28-29	47
29-30	48
30-31	1
31-32	2
32-33	3
33-34	1
34-35	1

Exponential decay of difficult
Discretization i

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

Discretization options

```

> opina$f.afei <- factor(cut(anys.feina, 4)) # In 4 intervals
> table(opina$f.afei)

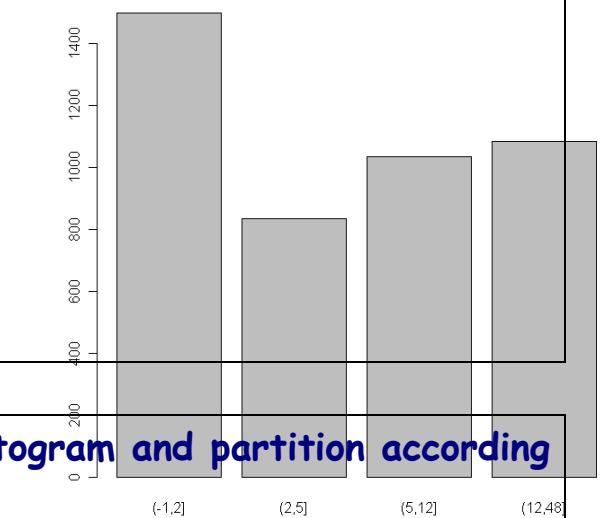
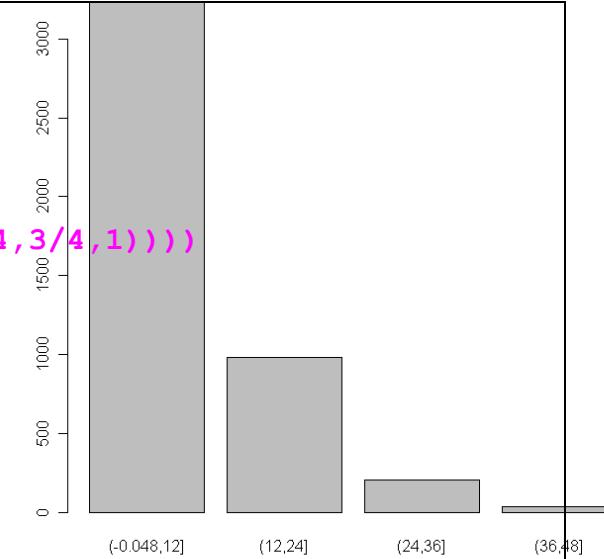
(-0.048,12]      (12,24]      (24,36]      (36,48]
 3235          979          207           33

> # Assistant to find the best grouping (in obs.nb.)
> opina$aux <- factor(cut(opina$anys.feina, quantile(anys.feina,c(0,1/4,2/4,3/4,1))))
> tapply(opina$anys.feina,opina$aux,median)
(0,2]   (2,5]   (5,12]  (12,48]
     1       4       9      18
> table(opina$aux)

(0,2]   (2,5]   (5,12]  (12,48]
 964     835    1034    1086
> sum(table(opina$aux))
[1] 3919
> opina$aux <- factor(cut(opina$anys.feina, breaks=c(-1,2,5,12,48)))
> tapply(opina$anys.feina,opina$aux,median)
(-1,2]   (2,5]   (5,12]  (12,48]
     1       4       9      18
> table(opina$aux)

(-1,2]   (2,5]   (5,12]  (12,48]
1499     835    1034    1086
> sum(table(opina$aux))
[1] 4454
>

```



Tip: Depending on user specified intervals (with meaning), after observing histogram and partition according to quantiles.

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```

opina$f.afei <- factor(cut(opina$anys.feina, breaks=c(-1,1.99,4.99,11.99,48)))
levels(opina$f.afei)<- c('<2','[2-5]', '[5-12]', '>12')
levels(opina$f.afei)<-paste("AFEI",levels(opina$f.afei), sep="")
summary(opina$f.afei)

# Housing: There are 6 missing. Group 4 ignores terms and treated as missing: 20 people
summary(opina$habitatge)
plot(opina$habitatge)

# Term
hist(opina$plas)
summary(opina$plas)
opina$f.plas <- factor(cut(opina$plas, breaks=c(5,35,47,59,72)))
levels(opina$f.plas)<- c('<36','36-47','48-59','>59')
levels(opina$f.plas) <- paste("PLAÇ",levels(opina$f.plas), sep="")

# Age
hist(opina$edat)
summary(opina$edat)
opina$f.edat<- factor(cut(opina$edat, breaks=c(17,28,36,45,68)))
levels(opina$f.edat) <- paste("EDAT",levels(opina$f.edat), sep="")

```

AFEI Category	Frequency
AFEI<2	~1100
AFEI[2-5]	~1000
AFEI[5-12]	~1200
AFEI>12	~1300

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```
# Marital Status: There is a missing (3320). One separated and divorced
in a group

summary(opina$estat.civil)
table(opina$estat.civil)
opina[opina$estat.civil==0,]
opina$f.eciv <- opina$estat.civil
opina$f.eciv[opina$estat.civil==5] <- 4
opina$f.eciv
factor(opina$f.eciv, levels=c("1", "2", "3", "4"), labels=c("ECIV.solter",
,"ECIV.casat", "ECIV.vidu", "ECIV.sepdiv"))
table(opina$f.eciv)

# Records: what is it? Delete a variable undocumented
opina$registres <- NULL

# Type of work
summary(opina$tipus.feina)
table(opina$tipus.feina)
opina$f.tfei
factor(opina$tipus.feina, levels=c("1", "2", "3", "4"), labels=c("TFEI.fi
x", "TFEI.tmp", "TFEI.auto", "TFEI.altr"))
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```
# Now come the quintessential covariates
# Expenditure: without missing. There are more high cost outliers in the group
of rejection
summary(opina$despeses)
table(opina$despeses)
hist(opina$despeses)
boxplot(opina$despeses)
tapply(opina$despeses, opina$f.dictamen, mean)
tapply(opina$despeses, opina$f.dictamen, median)
plot(opina$despeses~opina$f.dictamen)

# To find the best clustering (in obs.nb.)
opina$f.desp<- factor(cut(opina$despeses, breaks=c(34,35,45,60,75,90,180)))
levels(opina$f.desp) <- paste("DESP",levels(opina$f.desp), sep="")
summary(opina$f.desp)

# Income: without missing. There is 1 outlier (807) high income group rejection
(has many assets but little income)

summary(opina$ingressos)
table(opina$ingressos)
par(mfrow=c(1,2))
hist(opina$ingressos)
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```
boxplot(opina$ingressos)
tapply(opina$ingressos,opina$f.dictamen,mean)
tapply(opina$ingressos,opina$f.dictamen,median)
plot(opina$dictamen,opina$ingressos)
text(opina$dictamen,opina$ingressos,labels=row.names(opina))
opina$f.ingr<- factor(cut(opina$ingressos, breaks=c(-1,50,100,150,250,960)))
levels(opina$f.ingr) <- paste("INGR",levels(opina$f.ingr), sep="")
summary(opina$f.ingr)

# Heritage: 47 missing. There is 1 outlier (807) high income group rejection
# (has many assets but little income)

summary(opina$patrimoni)
table(opina$patrimoni)
par(mfrow=c(1,2))
hist(opina$patrimoni)
boxplot(opina$patrimoni)
tapply(opina$patrimoni,opina$f.dictamen,mean)
tapply(opina$patrimoni,opina$f.dictamen,median)
plot(opina$patrimoni~opina$f.dictamen)
opina$f.patr<- factor(cut(opina$patrimoni,
1,500,4000,8000,15000,300000)), breaks=c(-
levels(opina$f.patr) <- c('<0.5','0.5-4','4-8','8-15','>15')
levels(opina$f.patr) <- paste("PATR",levels(opina$f.patr), sep="")
summary(opina$f.patr)
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```
# carrega.patr: 18 missing. There is 1 outlier (807) high income group rejection
# (has many assets but little income)

summary(opina$carrega.patr)
table(opina$carrega.patr)
par(mfrow=c(1,2))
hist(opina$carrega.patr)
boxplot(opina$carrega.patr)
opina$f.carrpatr<- factor(cut(opina$carrega.patr, breaks=c(-1,0,5000,30000)))
levels(opina$f.carrpatr) <- c('Nul','Mig','Alt')
levels(opina$f.carrpatr) <- paste("CARRP",levels(opina$f.carrpatr), sep="")
summary(opina$f.carrpatr)

# import.assoc: 0 missing.
summary(opina$import.assoc)
table(opina$import.assoc)
par(mfrow=c(1,2))
hist(opina$import.assoc)
boxplot(opina$import.assoc)
opina$f.import<-
breaks=c(0,700,1000,1300,5000)))
levels(opina$f.import) <- c('<.7','.7-1','1-1.3','1.3-5')
levels(opina$f.import) <- paste("IMP",levels(opina$f.import), sep="")
summary(opina$f.import)
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```

# preu.final: 0 missing.
summary(opina$preu.final)
table(opina$preu.final)
par(mfrow=c(1,2))
hist(opina$preu.final)
boxplot(opina$preu.final)
opina$f.preu <-
  factor(cut(opina$preu.final,breaks=c(0,1100,1400,1700,12000)))
levels(opina$f.preu) <- c('<1.1','1.1-1.4','1.4-1.7','1.7+')
levels(opina$f.preu) <- paste("PREU",levels(opina$f.preu), sep="")
summary(opina$f.preu)
> opina$registres<-NULL
> summary(opina)
      dictamen      anys.feina      habitatge      plas      edat      estat.civil
Min.   :1.000   Min.   : 0.000   Min.   :0.000   Min.   : 6.00   Min.   :18.00   Min.   :0.000
1st Qu.:1.000   1st Qu.: 2.000   1st Qu.:2.000   1st Qu.:36.00   1st Qu.:28.00   1st Qu.:2.000
Median :1.000   Median : 5.000   Median :2.000   Median :48.00   Median :36.00   Median :2.000
Mean   :1.282   Mean   : 7.987   Mean   :2.657   Mean   :46.44   Mean   :37.08   Mean   :1.879
3rd Qu.:2.000   3rd Qu.:12.000   3rd Qu.:4.000   3rd Qu.:60.00   3rd Qu.:45.00   3rd Qu.:2.000
Max.   :2.000   Max.   :48.000   Max.   :6.000   Max.   :72.00   Max.   :68.00   Max.   :5.000
      tipus.feina      despeses      ingressos      patrimoni      carrega.patr      import.assoc
Min.   :0.000   Min.   : 35.00   Min.   : 0.0   Min.   :     0   Min.   :     0   Min.   : 100
1st Qu.:1.000   1st Qu.: 35.00   1st Qu.: 80.0   1st Qu.:     0   1st Qu.:     0   1st Qu.: 700
Median :1.000   Median : 51.00   Median :120.0   Median : 3000   Median :     0   Median :1000
Mean   :1.676   Mean   : 55.57   Mean   :130.6   Mean   : 5404   Mean   :  343   Mean   :1039
3rd Qu.:3.000   3rd Qu.: 72.00   3rd Qu.:165.0   3rd Qu.: 6000   3rd Qu.:     0   3rd Qu.:1300
Max.   :4.000   Max.   :180.00   Max.   :959.0   Max.   :300000   Max.   :30000   Max.   :5000
      NA's   : 34.0   NA's   : 47   NA's   :  47   NA's   : 18

```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

preu.final	f.dictamen	f.afei	f.habi	f.plas
Min. : 105	rebutja:1254	AFEI<2 :1045	HAB.llloguer: 973	PLAÇ<36 : 673
1st Qu.: 1117	accepta:3200	AFEI[2-5) :1023	HAB.scrpu :2107	PLAÇ36-47: 971
Median : 1400		AFEI[5-12) :1167	HAB.contpri: 246	PLAÇ48-59: 877
Mean : 1463		AFEI>12 :1219	HAB.pares : 783	PLAÇ>59 :1933
3rd Qu.: 1692			HAB.altres : 319	
Max. :11140			NA's : 26	
f.edat	f.eciv	f.tfei	f.desp	f.ingr
EDAT(17,28]:1196	ECIV.solter: 977	TFEI.fix :2805	DESP(34,35] :1211	INGR(-1,50] : 496
EDAT(28,36]:1160	ECIV.casat :3241	TFEI.tmp : 452	DESP(35,45] : 915	INGR(50,100] :1222
EDAT(36,45]:1030	ECIV.vidu : 67	TFEI.auto:1024	DESP(45,60] :1079	INGR(100,150]:1364
EDAT(45,68]:1068	ECIV.sepdiv: 168	TFEI.altr: 171	DESP(60,75] : 743	INGR(150,250]:1051
	NA's : 1	NA's : 2	DESP(75,90] : 349	INGR(250,960]: 287
			DESP(90,180]: 157	NA's : 34
f.patr	f.carppatr	f.import	f.preu	
PATR<0.5 :1636	CARRPNul:3669	IMP<.7 :1165	PREU<1.1 :1065	
PATR0.5-4:1126	CARRPMig: 739	IMP.7-1 :1294	PREU1.1-1.4:1165	
PATR4-8 : 925	CARRPAlt: 28	IMP1-1.3: 948	PREU1.4-1.7:1172	
PATR8-15 : 443	NA's : 18	IMP1.3-5:1047	PREU1.7+ :1052	
PATR>15 : 277				
NA's : 47				

- FactoMineR package. Use catdes(X, num.var) and condes(X, num.var) after data imputation (no NAs).
- Imputation tools: missMDA package. For numeric variables imputePCA(X_num) method and imputeMCA(X_factors) for qualitative variables. Validation of imputation results is mandatory.

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

3-2.2 Global association: quantitative resp. and explicative var with I levels (factor)

Change of roles: quantitative response variables with continuous basis and as explanatory variable the actual review. The addressed question is whether there is a difference according to the actual opinion (opinion of acceptance or rejection) in seniority, age, expenses, income, assets, equity load, amount of credit, longer term loan and the value of items loans.

```

attach(opina)
> attributes(opina)$names
[1] "dictamen"      "anys.feina"     "habitatge"      "plas"          "edat"          "estat.civil"
[7] "tipus.feina"   "despeses"       "ingressos"      "patrimoni"     "carrega.patr"  "import.assoc"
[13] "preu.final"   "f.dictamen"    "f.afei"         "f.habi"        "f.plas"        "f.edat"
[19] "f.eciv"        "f.tfei"        "f.desp"         "f.ingr"        "f.patr"        "f.carrpatr"
[25] "f.import"      "f.preu"

# Variables with digital processing: creating list of variables
attach(opina)

vacs <-
list(opina[,2],opina[,4],opina[,5],opina[,8],opina[,9],opina[,10],opina[,11],opina[,12],
opina[,13])

nomsvacs<-names(opina)[c(2,4,5,8:13)];nomsvacs

pvalvacs <-NULL

# Object initialization loop pvalvacs list: p values contrast null hypothesis
for (i in 1:9) {pvalvacs[ i ] <- oneway.test(vacs[[i]]~f.dictamen,var.equal =
FALSE)$p.value }

pvalvacs

```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```
# Change list formatting to 1-dimensional matrix in columns
pvalvacs <- matrix( pvalvacs )

pvalvacs

# Give name to the rows of the matrix pvalvacs, variables relating to
row.names(pvalvacs) <-
c("anys.feina","plas","edat","despeses","ingressos","patrimoni","carrega.patr","import.assoc","preu.final")

pvalvacs

# Sort first matrix dimension: rows. Output: variables p-value less than major.
sort( pvalvacs[,1] )
> vacs <-
list(opina[,2],opina[,4],opina[,5],opina[,8],opina[,9],opina[,10],opina[,11],opina[,12],opina[,13])
> pvalvacs <-NULL
> for (i in 1:9 ) { pvalvacs[ i ] <- oneway.test(vacs[[i]]~f.dictamen,var.equal = FALSE)$p.value }
> pvalvacs
[1] 3.378876e-89 5.213283e-13 6.032429e-11 7.113231e-02 2.264256e-46 5.429938e-14 5.068654e-01
[8] 2.677710e-21 5.012962e-01
> pvalvacs <- matrix( pvalvacs )
> pvalvacs
      [,1]
[1,] 3.378876e-89
[2,] 5.213283e-13
[3,] 6.032429e-11
[4,] 7.113231e-02
[5,] 2.264256e-46
[6,] 5.429938e-14
[7,] 5.068654e-01
[8,] 2.677710e-21
[9,] 5.012962e-01
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```

> row.names( pvalvacs ) <-  

c("anys.feina","plas","edat","despeses","ingressos","patrimoni","carrega.patr","import.assoc","  

preu.final")  

> pvalvacs  

      [,1]  

anys.feina 3.378876e-89  

plas 5.213283e-13  

edat 6.032429e-11  

despeses 7.113231e-02  

ingressos 2.264256e-46  

patrimoni 5.429938e-14  

carrega.patr 5.068654e-01  

import.assoc 2.677710e-21  

preu.final 5.012962e-01  

> sort( pvalvacs[,1] )  

  anys.feina  ingressos import.assoc  patrimoni          plas          edat       despeses  

3.378876e-89 2.264256e-46 2.677710e-21 5.429938e-14 5.213283e-13 6.032429e-11 7.113231e-02  

  preu.final carrega.patr  

5.012962e-01 5.068654e-01  

>nomsvacs<-names(opina)[c(2,4,5,8:13)];nomsvacs  

>library(FactoMineR)  

> catdes(opina[,c("f.dictamen",nomsvacs)],num.var=1)  

$quanti.var  

          Eta2      P-value  

anys.feina 0.067804443 6.071892e-70  

ingressos 0.043837874 5.628929e-45  

import.assoc 0.023885476 3.261359e-25

```

```

plas          0.010125845 1.686207e-11
patrimoni    0.009420326 1.070367e-10
edat          0.009082235 1.854456e-10
  
```

\$quanti

\$quanti\$rebutja

	v.test	Mean in category	Overall mean	sd in category
import.assoc	10.313197	1156.062998	1038.918276	535.454076
plas	6.714938	48.794258	46.438707	12.859222
edat	-6.359496	35.408293	37.080377	10.421689
patrimoni	-6.501447	3602.786062	5403.979351	8653.748988
ingressos	-14.053672	101.508104	130.564253	80.814576
anys.feina	-17.376225	4.586922	7.986753	6.115582
	Overall	sd	p.value	
import.assoc	474.492724	6.142358e-25		
plas	14.653817	1.881465e-11		
edat	10.983365	2.024164e-10		
patrimoni	11573.118739	7.955097e-11		
ingressos	86.367110	7.314150e-45		
anys.feina	8.173388	1.249107e-67		

\$quanti\$accepta

	v.test	Mean in category	Overall mean	sd in category
anys.feina	17.376225	9.319062	7.986753	8.486593
ingressos	13.890300	141.818267	130.564253	85.820705
patrimoni	6.452196	6104.474945	5403.979351	12455.914562

```
edat      6.359496    37.735625   37.080377  11.127476
plas     -6.714938    45.515625   46.438707  15.200547
import.assoc -10.313197  993.012187 1038.918276 439.922117
                    Overall sd      p.value
anys.feina      8.173388 1.249107e-67
ingressos       86.367110 7.253072e-44
patrimoni      11573.118739 1.102407e-10
edat        10.983365 2.024164e-10
plas         14.653817 1.881465e-11
import.assoc    474.492724 6.142358e-25

attr(,"class")
[1] "catdes" "list "
>
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

- The final review (*f.dictamen*) is associated with all of said quantitative variables except the value of objects loan (*preu.final*) and the financial burden. The costs show the contrast p value of association with the final review of 7% (slightly higher confidence level of 5%), there is to be strict and in my opinion there is evidence to reject the null hypothesis of no association and therefore to say that there are differences in costs between the 2 groups of opinion.
- The **null hypothesis** asserts that the two focus groups (acceptance or rejection) mean values of the variables at work seniority, age, expenses, income, assets, credit amount and term of the loan longer have p values **are equal** lower than the typical threshold of 5% (0.05 in per unit) and therefore **the null hypotheses are rejected** and taken for **valid alternatives differences between the 2 groups**.
- The **null hypothesis** asserts that the two focus groups (acceptance or rejection) mean values of the financial burden and value objects **are equal** loans are given for valid by showing p values above the threshold of 5%.
- Attention: the file is purged and the specific numbers may differ if R script execution with other similar criterion debugging variable is repeated, the more the findings should not differ if debugging is robust and sensible. You can use the FactoMineR R package.

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

3-2.3 Global association between qualitative resp. with J levels and an explicative var with I levels

- The role of response variable in this case is set by f.dictamen (dichotomous, $J = 2$). Categorical explanatory variables are the original.
- Over those created by the process of discretization of numerical variables originally included.

```

> load("E:/LTLIDIA/LIDIA/MLTM-MCAID/raw material/CURS06-07/Tema 2/opina.RData")
> attach(opina)
> names(opina)
[1] "dictamen"      "anys.feina"     "habitatge"      "plas"          "edat"          "estat.civil"
[7] "tipus.feina"   "despeses"       "ingressos"      "patrimoni"     "carrega.patr"  "import.assoc"
[13] "preu.final"   "f.dictamen"    "f.afei"         "f.habi"        "f.plas"        "f.edat"
[19] "f.eciv"        "f.tfei"        "f.desp"         "f.ingr"        "f.patr"        "f.carrpatr"
[25] "f.import"      "f.preu"

```

Manual procedure, untapped computational advantages of R

```

# seniority
summary(opina$f.afei)
# Hypothesis independence dictamen vs. f.afei rejected X-squared= 42.8832, df = 3, p-value = 2.606e-09
chisq.test(opina$f.afei,opina$f.dictamen)
table(opina$f.afei,opina$f.dictamen)

```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```
# Housing: There are 6 missing. Group 4 ignores terms and treated as missing: 20 people
# Hypothesis independence dictamen vs. f.habi rejected (X-squared = 217.8476, df = 4, p-value < 2.2e-16)
summary(opina$f.habi)
chisq.test(opina$f.habi,opina$f.dictamen)
table(opina$f.habi,opina$f.dictamen)
```

... They continue with all factors, analyzing contingency tables

```
> table(opina$f.afei,opina$f.dictamen)
```

	rebutja	accepta
AFEI<2	512	533
AFEI[2-5)	330	693
AFEI[5-12)	263	904
AFEI>12	149	1070

```
> table(opina$f.ingr,opina$f.dictamen)
```

	rebutja	accepta
INGR(-1,50]	279	217
INGR(50,100]	441	781
INGR(100,150]	289	1075
INGR(150,250]	172	879
INGR(250,960]	53	234

```
> table(opina$f.tfei,opina$f.dictamen)
```

	rebutja	accepta
TFEI.fix	580	2225
TFEI.tmp	271	181
TFEI.auto	333	691
TFEI.altr	68	103

```
> table(opina$f.habi,opina$f.dictamen)
```

	rebutja	accepta
HAB.llloguer	388	585
HAB.scrpu	390	1717
HAB.contpri	84	162
HAB.pares	233	550
HAB.altres	146	173

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

Automated procedure using the computational advantages of R

```

> names(opina)
# [1] "dictamen"      "anys.feina"     "habitatge"      "plas"          "edat"          "estat.civil"
# [7] "tipus.feina"    "despeses"       "ingressos"      "patrimoni"     "carrega.patr"  "import.assoc"
#[13] "preu.final"    "f.dictamen"    "f.afei"         "f.habi"        "f.plas"        "f.edat"
#[19] "f.eciv"         "f.tfei"        "f.desp"         "f.ingr"        "f.patr"        "f.carrpatr"
#[25] "f.import"       "f.preu"

vads <-
list(f.afei,f.habi,f.plas,f.edat,f.eciv,f.tfei,f.desp,f.ingr,f.patr,f.carrpatr,f.import,f.preu)
pvalvads <-NULL
for (i in 1:12 ) { pvalvads[ i ] <- chisq.test(vads[[i]],f.dictamen)$p.value }
pvalvads <- matrix( pvalvads )
row.names( pvalvads ) <-
c("f.afei","f.habi","f.plas","f.edat","f.eciv","f.tfei","f.desp","f.ingr","f.patr","f.carrpatr"
,"f.import","f.preu")
sort( pvalvads[,1] )

> sort( pvalvads[,1] )
      f.afei      f.ingr      f.tfei      f.habi      f.patr      f.import      f.plas
2.913760e-87 8.229399e-75 4.303683e-70 5.446068e-46 9.349481e-43 1.291256e-23 1.912465e-12
      f.eciv      f.edat      f.desp      f.preu      f.carrpatr
2.300075e-11 2.605645e-09 7.466957e-08 7.595702e-08 2.534421e-01

```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

- All variables are associated with the review, unless the discretization of the financial burden, which is consistent with the analysis as a covariate in the financial burden.
- The final price as a covariate is not associated with the review: the null hypothesis of equal final price in the 2 opinion groups is accepted, however, when considering the discretization of the final price (f.preu) there are differences. There is an issue of outliers.

```
# Graphs of the four variables associated and discrepant preu.final
par(mfrow=c(1,2))
plot(opina$f.dictamen,opina$f.afei)
plot(opina$f.dictamen,opina$anys.feina)
par(mfrow=c(1,2))
plot(opina$f.dictamen,opina$f.ingr)
plot(opina$f.dictamen,opina$ingressos)
par(mfrow=c(1,2))
plot(opina$f.dictamen,opina$f.tfei)
plot(opina$f.dictamen,opina$f.habi)

par(mfrow=c(1,2))
plot(opina$f.dictamen,opina$f.preu)
plot(opina$f.dictamen,opina$preu.final)
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

Numerical univariate descriptive analysis by focus opinion groups in numeric variables seniority, income and final price of items subject to loans:

```
> tapply(opina$anys.feina, opina$f.dictamen, summary)
```

\$rebutja

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
	0.000	1.000	2.000	4.587	6.000	43.000

\$accepta

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
	0.000	2.000	7.000	9.319	14.000	48.000

```
> tapply(opina$ingressos, opina$f.dictamen, summary)
```

\$rebutja

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
	0.0	58.0	90.0	101.5	136.0	959.0	20.0

\$accepta

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
	0.0	92.0	127.5	141.8	175.0	905.0	14.0

```
> tapply(opina$preu.final, opina$f.dictamen, summary)
```

\$rebutja

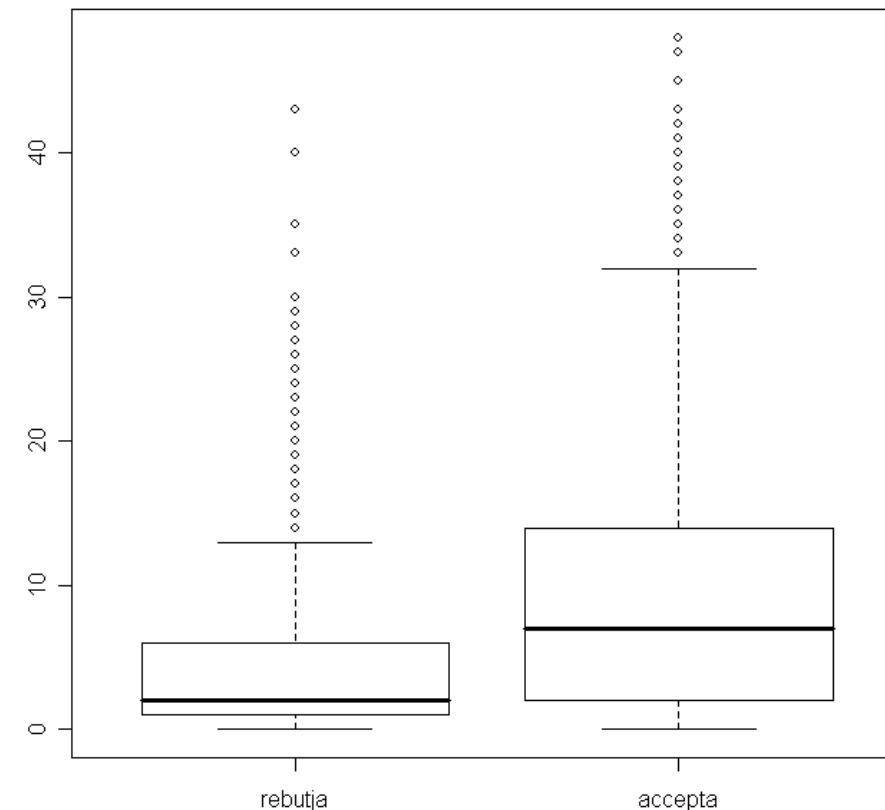
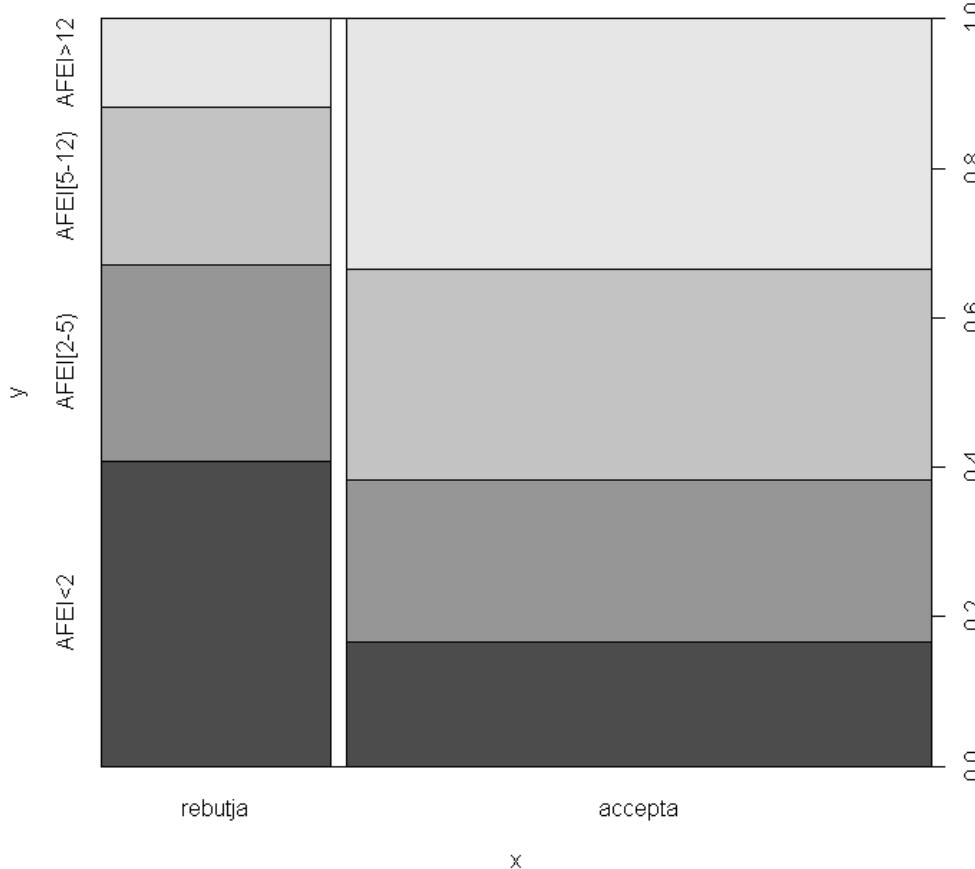
	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
	105	1063	1423	1474	1728	6802

\$accepta

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
	125	1134	1400	1459	1678	11140

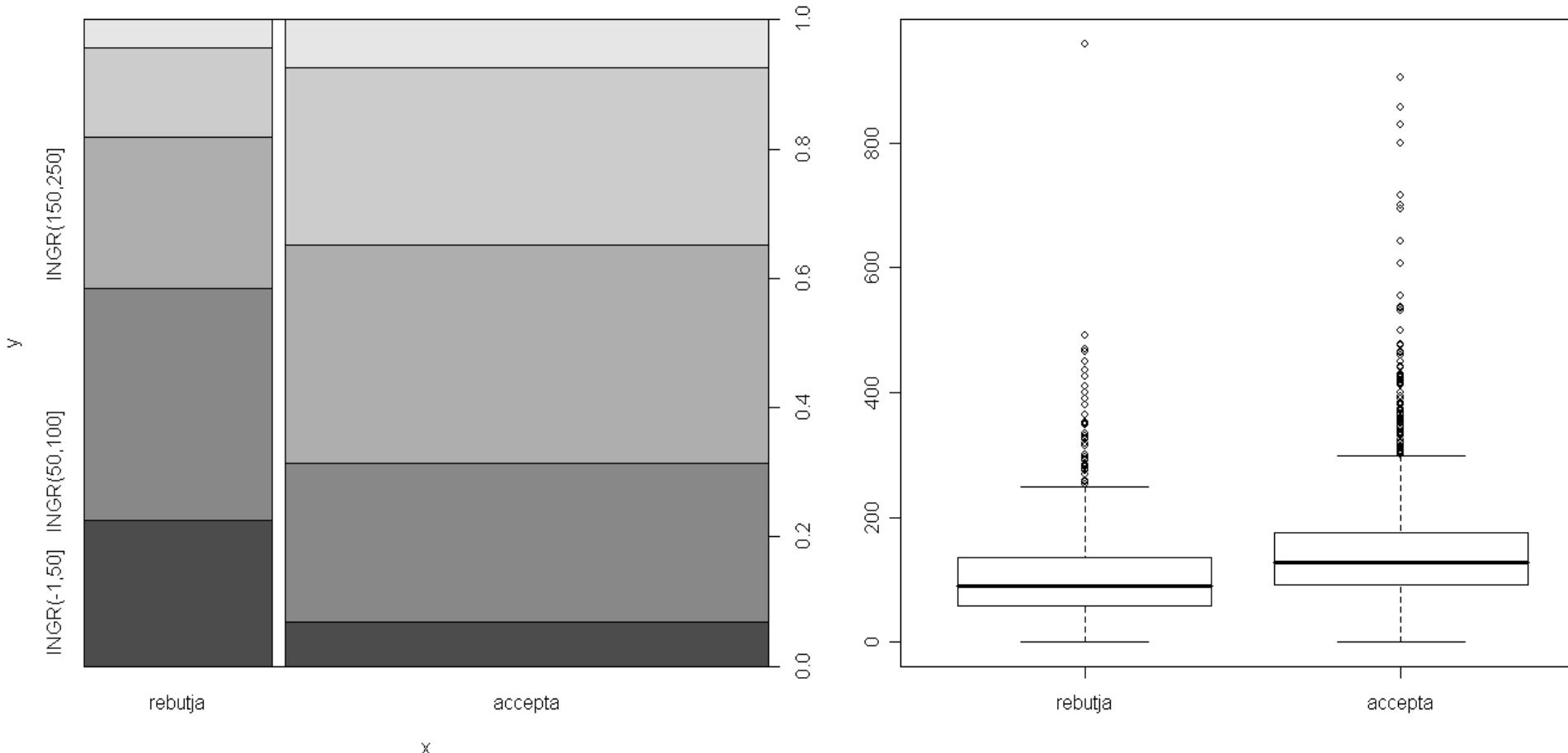
3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

Graphic univariate descriptive analysis by focus opinion groups of varying seniority, income and final price of items subject to loans:



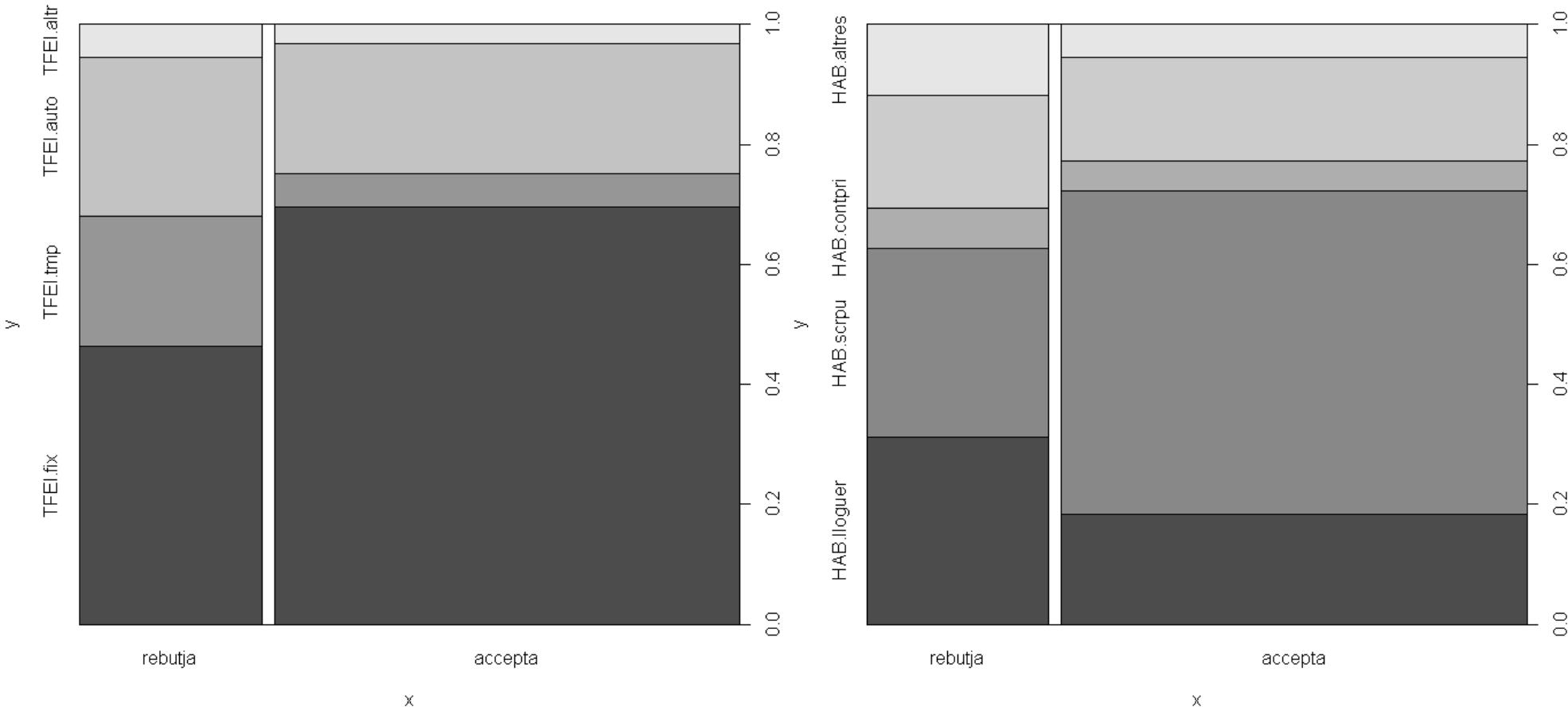
3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

Graphic univariate descriptive analysis by focus groups of variable income:



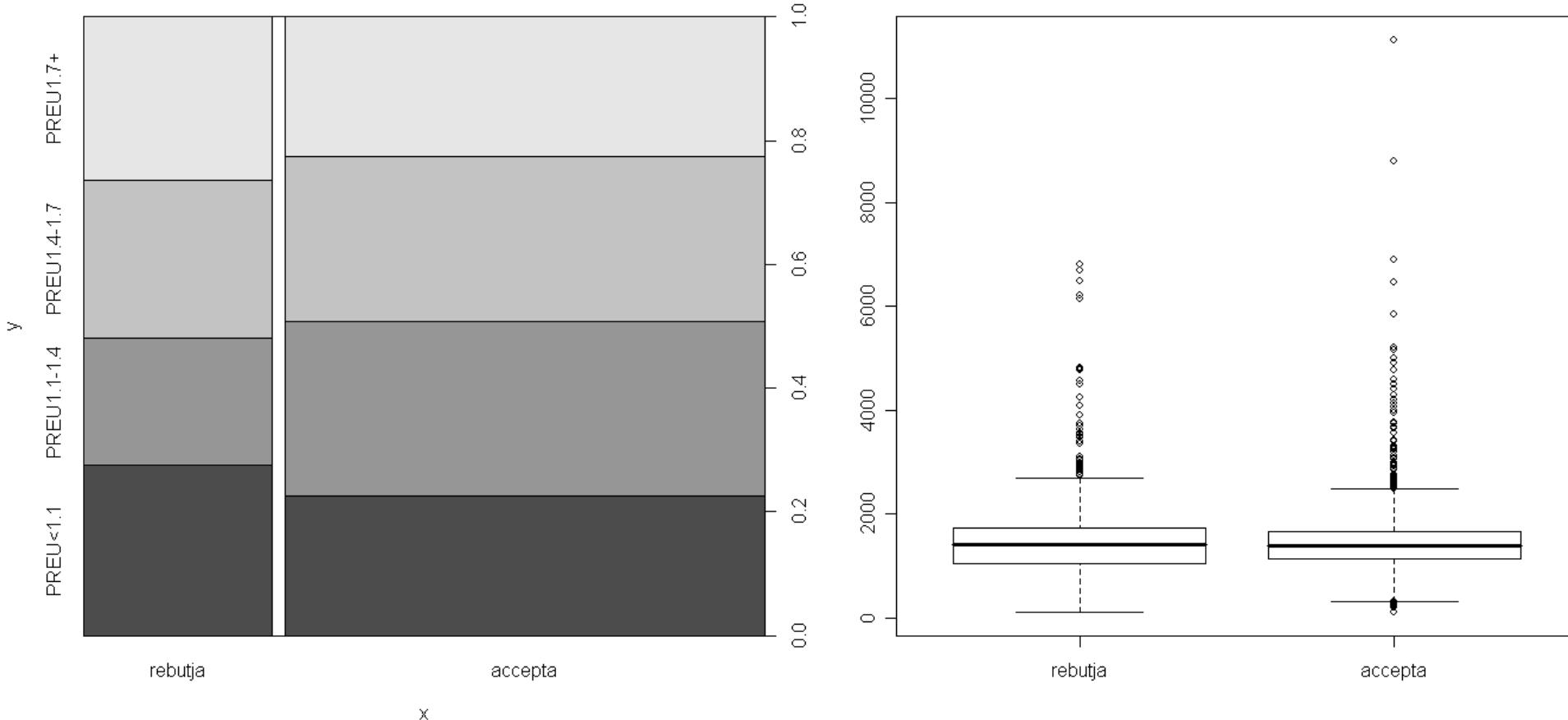
3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

Graphic univariate descriptive analysis by focus opinion groups of the variable type of work and ownership regime residence:



3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

Graphic univariate descriptive analysis by focus opinion groups final price of items subject to loans.:



3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```
>library(FactoMineR)
> names(opina)
[1] "dictamen"      "anys.feina"     "habitatge"      "plas"          "edat"          "estat.civil"
[7] "tipus.feina"   "despeses"       "ingressos"      "patrimoni"    "carrega.patr"  "import.assoc"
[13] "preu.final"   "f.dictamen"   "f.afei"        "f.habi"        "f.plas"        "f.edat"
[19] "f.eciv"        "f.tfei"        "f.desp"        "f.ingr"        "f.patr"        "f.carrpatr"
[25] "f.import"      "f.preu"
> nomsvads<-names(opina)[c(14:25)];nomsvads
[1] "f.dictamen" "f.afei" "f.habi" "f.plas"      "f.edat"      "f.eciv"      "f.tfei"      "f.desp"
[9] "f.ingr"      "f.patr"  "f.carrpatr" "f.import"
> catdes(opina[,c(nomsvads)],num.var=1)
$test.chi2
            p.value df
f.afei    2.913760e-87 3
f.ingr    7.456113e-77 5
f.tfei    4.314139e-70 4
f.habi    2.760016e-46 5
f.patr    9.013802e-43 5
f.import  1.291256e-23 3
f.plas    1.912465e-12 3
f.eciv    8.783901e-11 4
f.edat    2.605645e-09 3
f.desp    7.466957e-08 5
f.carrpatr 1.629441e-04 3

$category .... and much more
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

Library FactoMineR - profiling and feature selection - CSI tools
Y numeric target

condes(df, num.var)

\$quanti - Numeric explanatory vars that are globally associated to Y

\$quali - Qualitative explanatory vars that are globally associated to Y

\$category - Those categories belonging to explanatory factors whose mean is not equal to grand mean

B - Factor target

catdes(df, num.var)

\$ test.chi2 - Qualitative vars that are globally associated to B target

\$quanti.var - Quantitative vars that are globally associated to B target

\$quanti - Means on numeric vars that differ from global mean at each level of B factor

\$category - Specific relation between I factor from a qualitative factor and each level of B factor

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

3-2.4 Association for levels resp. with J levels and qualitative factors

Helper function to calculate the p value of the contrast statistic for each level of an explanatory factor in files (I level) and each of the response levels in columns:

```

# Create a function to calculate z for each cell of the table
# original (for var. Categorically a table): response columns
zij <- function( x, y ){
  taula <- table( x, y );
  # added table rows which is the first
  dimension
  n_ <- sum( taula );
  ni_ <- apply(taula, 1, sum ) ;
  pi_ <- apply(taula, 1, sum ) / n_ ;
  # sum table columns is the second dimension
  n_j <- apply( taula, 2, sum ) ;
  p_j <- apply( taula, 2, sum ) / n_ ;
  # Calculate the row profiles: nij/ni_ (annotate with p(j/i))
  pf <- taula/(ni_) ;
  pc <- taula/(n_j) ;
  # Replicating the marginal univ column many times as rows: n_j/n_
  pcI <- matrix( data=p_j, nrow=dim(pf)[1], ncol=dim(pf)[2], byrow=T ) ;
  # Numerator
  dpf <- pf - pcI;
  # Denominator
  denom <- sqrt( ( (1-pi_)/(ni_)) %*% t( p_j*(1-p_j) ) );
  zij <- dpf / denom;
  pzij <- 1-pnorm(zij);
  list(perfilfila=pf,perfilcolumna=pc,p.filafila=p_j,p.col=pi_, vtest=zij, pval=pzij )
}

```

Output:
 # One list with zij (vtest) and pvalues (pval) plus,
 # p.col is marginal row: pi_=sum_j (nij)/n
 # p.fila is marginal column: p_j=sum_i (nij)/n
 # perfilcolumna are column percentages: nij/n_j
 # perfilfila are row percentages: nij/ni_

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```

> pvalvadij1 <- zиж( f.afei, f.dictamen )$pval
> pvalvadij1
      y
x          rebutja      accepta
AFEI<2    0.000000e+00 1.000000e+00
AFEI[2-5)  4.420680e-04 9.995579e-01
AFEI[5-12) 9.999997e-01 3.394110e-07
AFEI>12   1.000000e+00 0.000000e+00
> pvalvadij2 <- zиж( f.habi, f.dictamen )$pval
> pvalvadij2
      y
x          rebutja      accepta
HAB.lloguer 0.000000e+00 1.000000e+00
HAB.scrpu   1.000000e+00 0.000000e+00
HAB.contpri 1.393059e-02 9.860694e-01
HAB.pares   1.172637e-01 8.827363e-01
HAB.altres  1.201261e-13 1.000000e+00
...
> pvalvadij12 <- zиж( f.preu, f.dictamen )$pval

> pvadij <-
rbind(pvalvadij1,pvalvadij2,pvalvadij3,pvalvadij4,pvalvadij5,pvalvadij6,pvalvadij7,pvalvadij8,pvalvadij
9,pvalvadij10,pvalvadij11,pvalvadij12)
> pvadij
      rebutja      accepta
AFEI<2    0.000000e+00 1.000000e+00
AFEI[2-5)  4.420680e-04 9.995579e-01
AFEI[5-12) 9.999997e-01 3.394110e-07
AFEI>12   1.000000e+00 0.000000e+00
HAB.lloguer 0.000000e+00 1.000000e+00
...
PREU1.7+   3.996315e-03 9.960037e-01
  
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

You could automate more generically using iterative loop in R.

```
names(opina)
# [1] "dictamen"      "anys.feina"     "habitatge"       "plas"          "edat"          "estat.civil"
# [7] "tipus.feina"    "despeses"        "ingressos"       "patrimoni"     "carrega.patr"  "import.assoc"
#[13] "preu.final"     "f.dictamen"     "f.afei"         "f.habi"        "f.plas"        "f.edat"
#[19] "f.eciv"         "f.tfei"         "f.desp"         "f.ingr"        "f.patr"        "f.carrpatr"
#[25] "f.import"       "f.preu"

vads <-
list(f.afei,f.habi,f.plas,f.edat,f.eciv,f.tfei,f.desp,f.ingr,f.patr,f.carrpatr,f.import,f.preu)
pvadkl <-NULL
for (i in 1:12 ) { pvadkl <- rbind(pvadkl, zij(vads[[i]],f.dictamen)$pval) }
pvadkl

#pvadkl <-
#rbind(pvalvads[1],pvalvads[2],pvalvads[3],pvalvads[4],pvalvads[5],pvalvads[6],pvalvads[7],pvalvads[8],pvalvads[9],pvalvads[10],pvalvads[11],pvalvads[12])
pvadkl

sort(pvadkl[,1])
sort(pvadkl[,2])
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```
# Acceptance opinion: factor levels more to less explanatory
> sort(pvadij[,2])
  AFEI>12      HAB.scrpu      TFEI.fix INGR(150,250)      PLAÇ<36 INGR(100,150)      PATR4-8
0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 3.541611e-14 1.329648e-11 2.179690e-11
  ECIV.casat   EDAT(45,68]     IMP.<.7  PREU1.1-1.4      PATR8-15 AFEI[5-12)      PATR>15
1.865655e-10 1.086958e-09 1.680927e-09 5.589849e-08 1.394307e-07 3.394110e-07 8.868735e-07
  IMP.7-1      INGR(250,960]   PATR0.5-4 DESP(35,45]      DESP(45,60] DESP(60,75]      CARRPMig
4.792517e-06 1.115941e-04 2.623134e-03 4.568517e-03 1.027362e-02 5.204700e-02 7.883559e-02
  PREU1.4-1.7  EDAT(28,36]     ECIV.vidu PLAÇ36-47      CARRPAlt HAB.pares      CARRPNul
2.486637e-01 4.220112e-01 5.144387e-01 7.033886e-01 8.199414e-01 8.827363e-01 8.850363e-01
  IMP1-1.3     PLAÇ48-59     EDAT(36,45] DESP(34,35]      HAB.contpri PREU1.7+      AFEI[2-5)
9.296074e-01 9.351384e-01 9.654742e-01 9.821484e-01 9.860694e-01 9.960037e-01 9.995579e-01
  TFEI.altr    PREU<1.1     TFEI.auto DESP(90,180]     PLAÇ>59  DESP(75,90]      EDAT(17,28]
9.997235e-01 9.997898e-01 9.998193e-01 9.998260e-01 9.999113e-01 9.999584e-01 9.999883e-01
  ECIV.solter  ECIV.sepdiv   HAB.altres INGR(50,100]     AFEI<2  HAB.lloguer      TFEI.tmp
9.999896e-01 1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00
  INGR(-1,50]   PATR<0.5     IMP1.3-5
1.000000e+00 1.000000e+00 1.000000e+00

# Rejection opinion: factor levels unless more explanatory
> sort(pvadij[,1])
  AFEI<2      HAB.lloguer      TFEI.tmp INGR(-1,50]      PATR<0.5      IMP1.3-5 INGR(50,100]
0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 3.586020e-14
  HAB.altres  ECIV.sepdiv   ECIV.solter EDAT(17,28]     DESP(75,90]     PLAÇ>59 DESP(90,180]
1.201261e-13 4.014773e-08 1.038939e-05 1.167609e-05 4.158758e-05 8.871978e-05 1.739983e-04
  TFEI.auto    PREU<1.1     TFEI.altr  AFEI[2-5)     PREU1.7+      HAB.contpri DESP(34,35]
1.807043e-04 2.102092e-04 2.765210e-04 4.420680e-04 3.996315e-03 1.393059e-02 1.785162e-02
  EDAT(36,45]   PLAÇ48-59  IMP1-1.3  CARRPNul     HAB.pares     CARRPAlt PLAÇ36-47
3.452581e-02 6.486157e-02 7.039256e-02 1.149637e-01 1.172637e-01 1.800586e-01 2.966114e-01
  ECIV.vidu    EDAT(28,36]   PREU1.4-1.7 CARRPMig     DESP(60,75]   DESP(45,60] DESP(35,45]
4.855613e-01 5.779888e-01 7.513363e-01 9.211644e-01 9.479530e-01 9.897264e-01 9.954315e-01
  PATR0.5-4    INGR(250,960] IMP.7-1    PATR>15     AFEI[5-12)     PATR8-15 PREU1.1-1.4
9.973769e-01 9.998884e-01 9.999952e-01 9.999991e-01 9.999997e-01 9.999999e-01 9.999999e-01
  IMP.<.7      EDAT(45,68]     ECIV.casat PATR4-8     INGR(100,150]   PLAÇ<36 AFEI>12
1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00
  HAB.scrpu    TFEI.fix     INGR(150,250]
1.000000e+00 1.000000e+00 1.000000e+00
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```
# Use FactoMineR package (forget the functions)
> names(opina)
[1] "dictamen"      "anys.feina"     "habitatge"      "plas"          "edat"          "estat.civil"
[7] "tipus.feina"   "despeses"       "ingressos"      "patrimoni"    "carrega.patr" "import.assoc"
[13] "preu.final"   "f.dictamen"    "f.afei"        "f.habi"       "f.plas"       "f.edat"
[19] "f.eciv"        "f.tfei"        "f.desp"        "f.ingr"       "f.patr"       "f.carrpatr"
[25] "f.import"      "f.preu"
> nomsvads<-names(opina)[c(14:25)];nomsvads
[1] "f.dictamen" "f.afei"      "f.habi"      "f.plas"      "f.edat"      "f.eciv"      "f.tfei"      "f.desp"
[9] "f.ingr"      "f.patr"      "f.carrpatr" "f.import"
> catdes(opina[,c(nomsvads)],num.var=1)
$test.chi2
            p.value df
f.afei    2.913760e-87 3
f.ingr    7.456113e-77 5
f.tfei    4.314139e-70 4
f.habi    2.760016e-46 5
f.patr    9.013802e-43 5
f.import  1.291256e-23 3
f.plas    1.912465e-12 3
f.eciv    8.783901e-11 4
f.edat    2.605645e-09 3
f.desp    7.466957e-08 5
f.carrpatr 1.629441e-04 3

$category
$category$rebutja
           Cla/Mod  Mod/Cla  Global  p.value  v.test
f.afei=AFEI<2  48.99522 40.829346 23.4620566 2.066308e-61 16.534634
f.tfei=TFEI.tmp 59.95575 21.610845 10.1481814 2.077713e-50 14.930796
f.ingr=INGR(-1,50] 56.25000 22.248804 11.1360575 3.185727e-44 13.949106
f.patr=PATR<0.5  40.03667 52.232855 36.7310283 3.948997e-40 13.259999
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

f.import=IMP1.3-5	39.63706	33.094099	23.5069600	3.490432e-20	9.202707
f.habi=HAB.lluguer	39.87667	30.940989	21.8455321	3.720151e-19	8.944981
f.ingr=INGR(50,100]	36.08838	35.167464	27.4360126	1.247073e-12	7.100055
f.habi=HAB.altres	45.76803	11.642743	7.1621015	5.145047e-12	6.901516
f.eciv=ECIV.sepdiv	46.42857	6.220096	3.7718904	4.282184e-07	5.055966
f.eciv=ECIV.solter	33.57216	26.156300	21.9353390	3.015307e-05	4.172307
f.edat=EDAT(17,28]	32.85953	31.339713	26.8522676	3.248560e-05	4.155303
f.desp=DESP(75,90]	37.24928	10.366826	7.8356533	1.543626e-04	3.783944
f.plas=PLAÇ>59	31.03983	47.846890	43.3991917	2.080250e-04	3.709066
f.carrrpatr=NA	72.22222	1.036683	0.4041311	2.562996e-04	3.655882
f.ingr=NA	58.82353	1.594896	0.7633588	3.356363e-04	3.586119
f.tfei=TFEI.auto	32.51953	26.555024	22.9905703	5.245459e-04	3.467898
f.desp=DESP(90,180]	40.76433	5.103668	3.5249214	7.273158e-04	3.379070
f.tfei=TFEI.altr	39.76608	5.422648	3.8392456	1.112552e-03	3.260400
f.afei=AFEI[2-5)	32.25806	26.315789	22.9681185	1.123301e-03	3.257672
f.habi=NA	50.00000	1.036683	0.5837449	2.945121e-02	2.177394
f.desp=DESP(34,35]	30.47069	29.425837	27.1890436	3.985629e-02	2.055235
f.habi=HAB.contpri	34.14634	6.698565	5.5231253	4.044316e-02	2.049194
f.patr=NA	42.55319	1.594896	1.0552313	4.688543e-02	1.987334
f.desp=DESP(45,60]	25.39388	21.850080	24.2254154	2.199695e-02	-2.290421
f.desp=DESP(35,45]	24.69945	18.022329	20.5433318	9.678926e-03	-2.587094
f.patr=PATR0.5-4	24.77798	22.248804	25.2806466	3.756515e-03	-2.897916
f.ingr=INGR(250,960]	18.46690	4.226475	6.4436462	1.224361e-04	-3.841196
f.import=IMP.7-1	23.49304	24.242424	29.0525370	9.155198e-06	-4.436221
f.afei=AFEI[5-12)	22.53642	20.972887	26.2011675	5.723537e-07	-5.000320
f.patr=PATR>15	15.52347	3.429027	6.2191289	5.347060e-07	-5.013421
f.patr=PATR8-15	17.60722	6.220096	9.9461159	8.470896e-08	-5.356800
f.import=IMP<.7	21.45923	19.936204	26.1562640	2.228576e-09	-5.980205
f.edat=EDAT(45,68]	20.97378	17.862839	23.9784463	1.302426e-09	-6.067097
f.eciv=ECIV.casat	25.57853	66.108453	72.7660530	8.651867e-10	-6.132482
f.patr=PATR4-8	19.35135	14.274322	20.7678491	8.247049e-12	-6.834189
f.ingr=INGR(100,150]	21.18768	23.046252	30.6241581	3.735941e-12	-6.946827
f.plas=PLAÇ<36	16.19614	8.692185	15.1100135	7.084890e-15	-7.782957

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```
f.ingr=INGR(150,250] 16.36537 13.716108 23.5967670 7.929587e-24 -10.064492
f.habi=HAB.scrpu 18.50973 31.100478 47.3057925 1.491928e-42 -13.672021
f.tfei=TFEI.fix 20.67736 46.251994 62.9770992 2.619632e-46 -14.287468
f.afei=AIFI>12 12.22313 11.881978 27.3686574 8.107785e-53 -15.296180
```

\$category\$accepta	Cla/Mod	Mod/Cla	Global	p.value	v.test
f.afei=AIFI>12	87.77687	33.43750	27.3686574	8.107785e-53	15.296180
f.tfei=TFEI.fix	79.32264	69.53125	62.9770992	2.619632e-46	14.287468
f.habi=HAB.scrpu	81.49027	53.65625	47.3057925	1.491928e-42	13.672021
f.ingr=INGR(150,250]	83.63463	27.46875	23.5967670	7.929587e-24	10.064492
f.plas=PLAC<36	83.80386	17.62500	15.1100135	7.084890e-15	7.782957
f.ingr=INGR(100,150]	78.81232	33.59375	30.6241581	3.735941e-12	6.946827
f.patr=PATR4-8	80.64865	23.31250	20.7678491	8.247049e-12	6.834189
f.eciv=ECIV.casat	74.42147	75.37500	72.7660530	8.651867e-10	6.132482
f.edat=EDAT(45,68]	79.02622	26.37500	23.9784463	1.302426e-09	6.067097
f.import=IMP<.7	78.54077	28.59375	26.1562640	2.228576e-09	5.980205
f.patr=PATR8-15	82.39278	11.40625	9.9461159	8.470896e-08	5.356800
f.patr=PATR>15	84.47653	7.31250	6.2191289	5.347060e-07	5.013421
f.afei=AIFI[5-12]	77.46358	28.25000	26.2011675	5.723537e-07	5.000320
f.import=IMP.7-1	76.50696	30.93750	29.0525370	9.155198e-06	4.436221
f.ingr=INGR(250,960]	81.53310	7.31250	6.4436462	1.224361e-04	3.841196
f.patr=PATR0.5-4	75.22202	26.46875	25.2806466	3.756515e-03	2.897916
f.desp=DESP(35,45]	75.30055	21.53125	20.5433318	9.678926e-03	2.587094
f.desp=DESP(45,60]	74.60612	25.15625	24.2254154	2.199695e-02	2.290421
f.patr=NA	57.44681	0.84375	1.0552313	4.688543e-02	-1.987334
f.habi=HAB.contpri	65.85366	5.06250	5.5231253	4.044316e-02	-2.049194
f.desp=DESP(34,35]	69.52931	26.31250	27.1890436	3.985629e-02	-2.055235
f.habi=NA	50.00000	0.40625	0.5837449	2.945121e-02	-2.177394
f.afei=AIFI[2-5)	67.74194	21.65625	22.9681185	1.123301e-03	-3.257672
f.tfei=TFEI.altr	60.23392	3.21875	3.8392456	1.112552e-03	-3.260400
f.desp=DESP(90,180]	59.23567	2.90625	3.5249214	7.273158e-04	-3.379070
f.tfei=TFEI.auto	67.48047	21.59375	22.9905703	5.245459e-04	-3.467898

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```

f.ingr=NA          41.17647  0.43750  0.7633588 3.356363e-04 -3.586119
f.carpatr=NA       27.77778  0.15625  0.4041311 2.562996e-04 -3.655882
f.plas=PLAÇ>59     68.96017  41.65625 43.3991917 2.080250e-04 -3.709066
f.desp=DESP(75,90] 62.75072  6.84375  7.8356533 1.543626e-04 -3.783944
f.edat=EDAT(17,28] 67.14047  25.09375 26.8522676 3.248560e-05 -4.155303
f.eciv=ECIV.solter 66.42784  20.28125 21.9353390 3.015307e-05 -4.172307
f.eciv=ECIV.sepdiv 53.57143  2.81250  3.7718904 4.282184e-07 -5.055966
f.habi=HAB.altres   54.23197  5.40625  7.1621015 5.145047e-12 -6.901516
f.ingr=INGR(50,100] 63.91162  24.40625 27.4360126 1.247073e-12 -7.100055
f.habi=HAB.lloguer  60.12333  18.28125 21.8455321 3.720151e-19 -8.944981
f.import=IMP1.3-5   60.36294  19.75000 23.5069600 3.490432e-20 -9.202707
f.patr=PATR<0.5     59.96333  30.65625 36.7310283 3.948997e-40 -13.259999
f.ingr=INGR(-1,50]   43.75000  6.78125 11.1360575 3.185727e-44 -13.949106
f.tfei=TFEI.tmp      40.04425  5.65625 10.1481814 2.077713e-50 -14.930796
f.afei=AFEI<2        51.00478  16.65625 23.4620566 2.066308e-61 -16.534634

attr(,"class")
[1] "catdes" "list "
>
  
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

The conclusions drawn from the results of inference are the previous page:

- The most discriminant rules by accepting CBD Prizing measures are AFEI> 12 (years of working more than 12 years), HAB.scrpu (available housing property with formalized writing), TFEI.fix (have a permanent contract) and INGR (150,250] (annual income between 150000 to 250000 €).
- The more discriminating arrangements with rejection of CBD Prizing measures are AFEI<2 (years of working less than 2 years), HAB.lloguer (usual rental housing), TFEI.tmp (have a temporary contract) , INGR (-1.50] (annual income below € 50000), PATR <0.5 (less than 5000 € worth), etc..
- The less discriminating modalities with rejection of CBD Prizing measures are AFEI>12 (years of working more than 12 years), HAB.scrpu (available housing property with formalized writing), TFEI.fix (have a contract permanent employment) and INGR (150,250] (annual income between € 150000 to 250000. Of course, are most associated with the form of acceptance of our response.

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

3-2.5 Association of quantitative target and I level factors

Changing roles: quantitative response variables with continuous character and income as explanatory variables all factors. The addressed question is whether there is a difference in income target depending on job, age, expenses, income, assets, equity load, amount of credit, term loans and longer loan impairment objects.

```

> attributes(opina)$names
[1] "dictamen"      "anys.feina"     "habitatge"      "plas"          "edat"          "estat.civil"
[7] "tipus.feina"   "despeses"       "ingressos"      "patrimoni"     "carrega.patr"  "import.assoc"
[13] "preu.final"   "f.dictamen"    "f.afei"        "f.habi"        "f.plas"        "f.edat"
[19] "f.eciv"        "f.tfei"        "f.desp"        "f.ingr"        "f.patr"        "f.carrpatr"
[25] "f.import"      "f.preu"

# Variables con tratamiento categórico: creación lista de variables
attach(opina)
vads <-
list(f.afei,f.habi,f.plas,f.edat,f.eciv,f.tfei,f.desp,f.ingr,f.patr,f.carrpatr,f.import,f.preu)
pvackl <-NULL
for (i in 1:12) { pvackl <- rbind(pvackl, tijs.test(vads[[i]], ingressos)$pval) }
pvackl
sort(pvackl[,1])

> pvackl
            [,1]
AFEI<2      1.000000e+00
AFEI[2-5)    9.799195e-01
AFEI[5-12)   3.130949e-04
AFEI>12     9.008350e-12
  
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```

HAB.lloguer 7.257076e-01
...
PREU<1.1 1.000000e+00
PREU1.1-1.4 9.591006e-01
PREU1.4-1.7 7.225851e-01
PREU1.7+ 0.000000e+00

> sort(pvackl[,1])
   HAB.scrpu   ECIV.casat     TFEI.fix DESP(90,180] INGR(150,250] INGR(250,960]      PREU1.7+
0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00
    PATR>15    IMP1.3-5     AFEI>12  DESP(60,75]  DESP(75,90]          PATR4-8      CARRPMig
1.065814e-14 8.471002e-14 9.008350e-12 3.800549e-10 4.311581e-10 6.956680e-09 1.069969e-07
    PATR8-15    EDAT(45,68]   CARRPAlt  AFEI[5-12)    PLAÇ<36  EDAT(36,45]  EDAT(28,36]
1.836212e-07 3.321193e-07 4.011699e-05 3.130949e-04 7.755021e-03 2.455788e-02 1.646013e-01
DESP(45,60]    PLAÇ48-59    PLAÇ>59   IMP1-1.3    HAB.contpri  PREU1.4-1.7  HAB.lloguer
2.786199e-01 3.241068e-01 4.018773e-01 5.366704e-01 6.000310e-01 7.225851e-01 7.257076e-01
    ECIV.vidu   IMP.7-1     ECIV.sepdiv  PREU1.1-1.4  AFEI[2-5)  INGR(100,150]  PATR0.5-4
7.745403e-01 8.207252e-01 9.010088e-01 9.591006e-01 9.799195e-01 9.930373e-01 9.950800e-01
    PLAÇ36-47    DESP(35,45]   TFEI.auto   TFEI.altr    HAB.altres  CARRPNul   IMP<.7
9.976938e-01 9.997583e-01 9.999918e-01 9.999995e-01 9.999998e-01 1.000000e+00 1.000000e+00
    HAB.pares   TFEI.tmp    PREU<1.1   EDAT(17,28]  AFEI<2    ECIV.solter  DESP(34,35]
1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00 1.000000e+00
INGR(-1,50]    INGR(50,100]  PATR<0.5
1.000000e+00 1.000000e+00 1.000000e+00

>
> library(FactoMineR)
> nomsvars<-names(opina)[c(14:26)];nomsvars
[1] "f.dictamen" "f.afei" "f.habi" "f.plas" "f.edat" "f.eciv" "f.tfei"      "f.desp"
[9] "f.ingr"      "f.patr"   "f.carrpatr" "f.import" "f.preu"
> # The answer can not be NAs
> df<-opina[!is.na(opina$ingressos),c("ingressos",nomsvars)]
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

```

> dim(df)
[1] 4420   14
> condes(df,num.var=1)
$quali
      R2      p.value
f.ingr 0.848002754 0.000000e+00
f.desp  0.056348645 2.709087e-53
f.dictamen 0.043837874 5.628929e-45
f.patr  0.039470569 1.566999e-36
f.tfei  0.027772478 6.197342e-26
f.preu  0.025942786 5.366424e-25
f.afei  0.023337468 1.855788e-22
f.habi  0.022446905 4.695900e-20
f.eciv  0.017957949 1.722417e-16
f.import 0.016140145 1.701961e-15
f.edat  0.014824500 3.122086e-14
f.carrrpatr 0.012758064 2.965749e-12
f.plas  0.002604839 9.225141e-03

$category
      Estimate      p.value
INGR(250,960] 199.563277 0.000000e+00
INGR(150,250]  38.786475 1.011900e-280
accepta        20.155082 5.628929e-45
PREU1.7+       22.581156 1.772866e-23
DESP(90,180]   52.310244 8.904500e-23
IMP1.3-5       17.050889 6.756882e-14
PATR>15       36.128442 1.936061e-13
  
```

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

AFEI>12	15.211624	4.866698e-12
CARRPAlt	68.932323	4.190370e-07
EDAT (45, 68]	11.089009	8.987020e-07
HAB.scrpu	16.931709	3.202381e-06
AFEI [5-12)	8.353976	1.396559e-04
PATR8-15	15.218194	2.471397e-04
DESP (75, 90]	14.307945	3.751772e-04
PATR4-8	9.603444	4.781514e-03
TFEI.fix	48.227245	4.989573e-03
PLAÇ<36	6.974560	1.122255e-02
NA	-29.467112	2.663771e-03
PLAÇ36-47	-7.506832	1.883317e-03
HAB.pares	-13.459898	1.149712e-03
PATR0.5-4	-10.696171	1.026747e-03
HAB.altres	-18.567615	2.827650e-04
NA	-63.763973	1.744878e-05
DESP (45, 60]	-12.376591	3.185092e-06
IMP<.7	-13.930197	2.819359e-10
PATR<0.5	-20.786798	1.000923e-11
PREU<1.1	-16.983877	5.773270e-14
EDAT (17, 28]	-16.887951	1.246510e-14
DESP (35, 45]	-22.681156	6.839499e-16
AFEI<2	-19.557552	1.262618e-17
DESP (34, 35]	-35.929223	9.766366e-44
rebutja	-20.155082	5.628929e-45
INGR (100, 150]	-26.347475	9.291599e-163
INGR (50, 100]	-71.953740	0.000000e+00
INGR (-1, 50]	-140.048538	0.000000e+00

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

Helper function to calculate the p value of the contrast statistic for each level of an explanatory factor in rows (I levels) and response columns:

```

# Create a function to calculate t for each contrast level of a variable

tijs.test <- function( x, y ){
  pc <- tapply( y, x, mean, na.rm = TRUE );
  # taula amb mitjana pels nivells de x
  mit <- mean(y,na.rm = TRUE);
  std2 <- var(y,use="complete.obs");

  # Replicated as many times as global average levels of x.
  pc <- matrix( data=pc, nrow=1, ncol= length( levels(x) ), byrow=T );
  pci <- matrix( data=mit, nrow=1, ncol= length( levels(x) ), byrow=T );
  taula <- table( x,y );
  tpx <- t(matrix(apply(taula,1,sum)));
  tt <- sum( taula );

  # Numerator
  dpf <- pc - pci;

  # Denominator
  denom <- sqrt( (1-(tpx/tt)) * (std2/tpx) );
  tij <- dpf /denom;
  ptij <- 1-pt(tij,tpx-1);
  tij <- t( tij );
  ptij <- t( ptij );
  row.names(tij) <- levels(x);
  row.names(ptij) <- levels(x);

  list(vtest=tij, pval=ptij )}
```

Output:

One list with tij (vtest) and pvalues (pval)

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

3-2.6 Work Sample and Test

A mandatory standard in any statistical model is its validation and hence the possibility of application to other data and the ability to use their predictions.

A universal technique for analysis and data mining is to divide the data randomly split into two sets: the subset of the test work and the test. The work sample will be used to build and estimate the statistical model (models if different techniques are used), while the test sample will be used to validate the proposed model so that it is possible to construct a validation table for a binary response style:

Test	$y=1$	$y=0$	Total
$\hat{y}_i = 1$	True positives (a)	False positives (b)	$a+b$
$\hat{y}_i = 0$	False negatives (c)	True negatives (d)	$c+d$
	$a+c$	$b+d$	n

- Sensitivity is the proportion of true values 1 that are estimated 1: $S_n = a/(a+c)$.
- Specificity is the proportion of true values 0 that are estimated 0: $S_p = d/(b+d)$.
- The positive and negative predictive values of the model are defined as.: $P_+ = a/(a+b)$ y $P_- = d/(c+d)$.

Predictive capacity : $(a+d)/n$

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

The working subsample generally contain between 50 and 95% of the original data.

If the amount of data is not high then can be applied cross-validation techniques. In this technique data is randomly divided into n groups (equal size, that is $n = 10$.) Reserves as a group subsample of Test and $n-1$ remaining as Working subsample, and this process is repeated n times, leaving a distinct group as Test at each iteration, for each iteration error rate is obtained with the test sample for . Finally, a model is built with all the data and accuracy and error rate is obtained by averaging the n error rates available.

Another technique for estimating the error of a model, when few data are available or when the inference is complicated, lies in the [bootstrapping](#): A first model with the entire sample is constructed and then point estimates are derived, then numerous sample datasets ASCR (over 1000 samples) and (from the point estimates) with these samples confidence intervals or error rates for the whole model is built.

3-2 CBD PRIZING EXAMPLE: EXPLORATION OF LARGE TABLES (DATABASE)

In the R statistical work environment, could be from the starting sample in subsamples of Working and Test (each in a data.frame) with type instructions which are used below:

```
sample(x, size, replace = FALSE, prob = NULL)
```

x: Either a (numeric, complex, character or logical) vector of more than one element from which to choose, or a positive integer, in this case 1:x range is chosen.

size: non-negative integer giving the number of items to choose.

replace: Should sampling be with replacement?

prob: A vector of probability weights for obtaining the elements of the vector being sampled.

```
# Create the WORK and TEST sample: for example 67% and 33% (o 75-25)
```

```
learn <- sample(1:(dim(opina)[1]), round(0.67*dim(opina)[1]))
learn <- sort( learn )
opina$filter <- rep(TRUE,dim(opina)[1])
attach(opina)
opina[-learn,]$filter <- FALSE
opina$filter
```

```
# Create data.frame of Work and for Test, and firstly insightful columns are selected according to the specification.
```

```
df.treball <- data.frame(... )
df.test <- df.treball[ -learn, ]
df.treball <- df.treball[ learn, ]
np <- dim(df.treball)[2]
dim(df.treball)[1] # For validation sample size LEARN
```