

# Study of French labour market and inequalities

L. Insolia, J. Kim and Y. Yeghikyan

SNS

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— *Midterm results* —

# Objectives

- Structure of French labour market
- Inequalities (in terms of salary):
  - ages
  - gender
  - job categories
  - spatial distribution
- Firms' distribution
- Exploratory analyses

# Methodology

## INSEE data

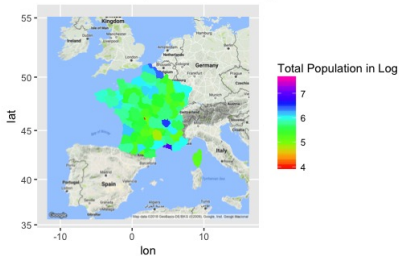
- Population: age, sex and cohabitation mode
- Salary: job categories, age and sex (mean net salary per hour in €)
- Firms: number of firms for each size
- Geography: GPS location

for different geographical levels (communes, departments, towns) in 2014

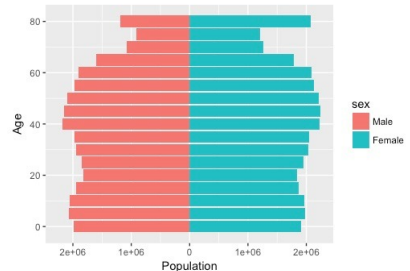
# What has been done so far . . .

# Demographic profiles

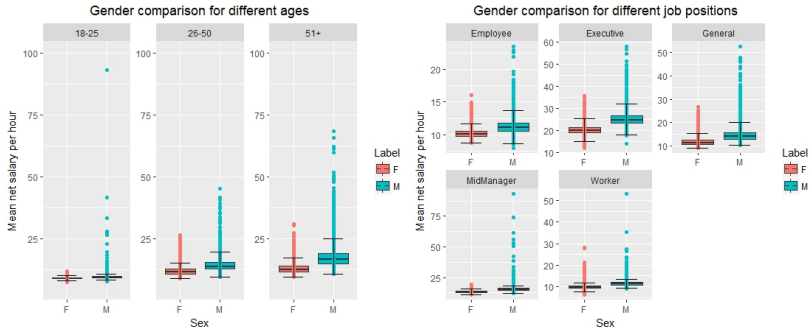
Distribution of Population for each department



Pyramid of Population

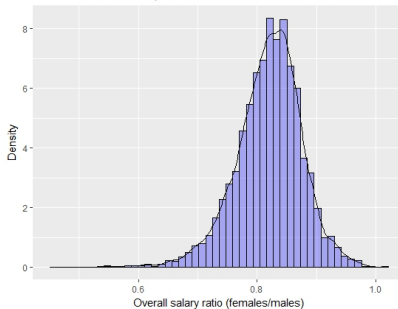


# Inequality of salary

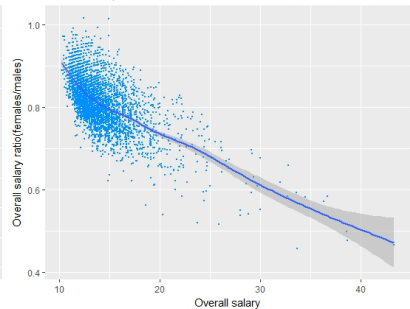


# Inequality of salary

Overall salary ratio between females and males

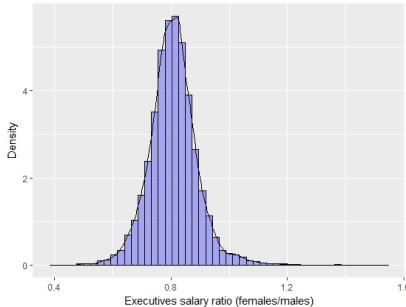


Overall salary ratio between females and males vs. overall salary



# Inequality of salary

Executives salary ratio between females and males

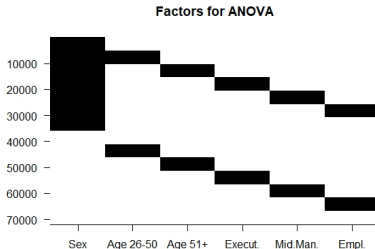


Executives salary ratio between females and males  
vs. overall executives salary





# ANOVA using sex, job, age and interaction effects



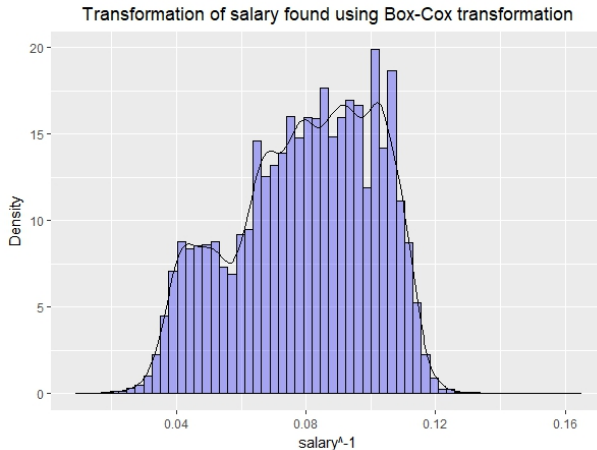
```
Call:
lm(formula = sal_y ~ sal_sex + sal_age + sal_job + sal_sex:sal_age +
    sal_sex:sal_job)

Residuals:
    Min       1q   Median       3Q      Max
-0.084405 -0.004353  0.000683  0.005477  0.057842

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  1.061e-01  8.471e-05 1252.443  < 2e-16 ***
sal_sex      -1.097e-02  1.198e-04  -91.569  < 2e-16 ***
sal_age1     -2.160e-02  1.467e-04 -147.227  < 2e-16 ***
sal_age2     -2.838e-02  1.467e-04 -193.440  < 2e-16 ***
sal_job1     -5.601e-02  1.467e-04 -381.776  < 2e-16 ***
sal_job2     -3.036e-02  1.467e-04 -206.917  < 2e-16 ***
sal_job3     -8.621e-03  1.467e-04  -58.758  < 2e-16 ***
sal_sex:sal_age1 -2.502e-03  2.075e-04  -12.057  < 2e-16 ***
sal_sex:sal_age2 -7.572e-03  2.075e-04  -36.491  < 2e-16 ***
sal_sex:sal_job1 1.197e-03  2.075e-04   5.770 7.94e-09 ***
sal_sex:sal_job2 4.873e-04  2.075e-04   2.349  0.0188 *
sal_sex:sal_job3 3.059e-03  2.075e-04  14.742  < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

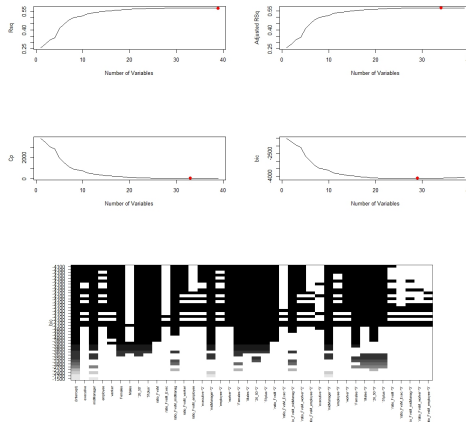
Residual standard error: 0.008585 on 71892 degrees of freedom
Multiple R-squared:  0.841,    Adjusted R-squared:  0.841
F-statistic: 3.458e+04 on 11 and 71892 DF,  p-value: < 2.2e-16
```

# ANOVA using sex, job, age and interaction effects

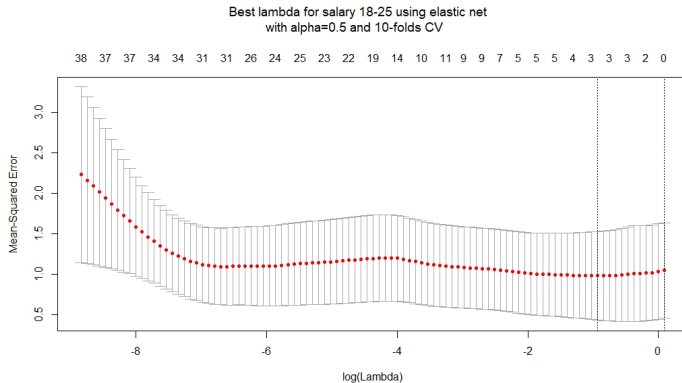


# Prediction for young people using BSS

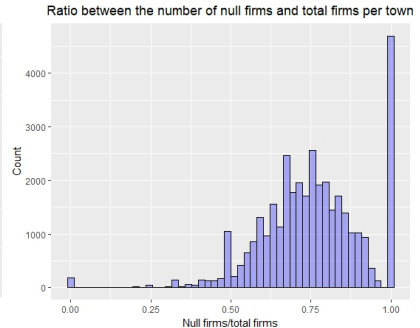
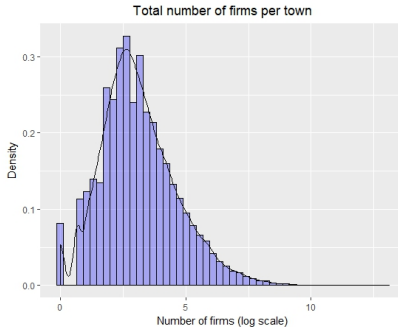
Best subset selection for salary 18-25



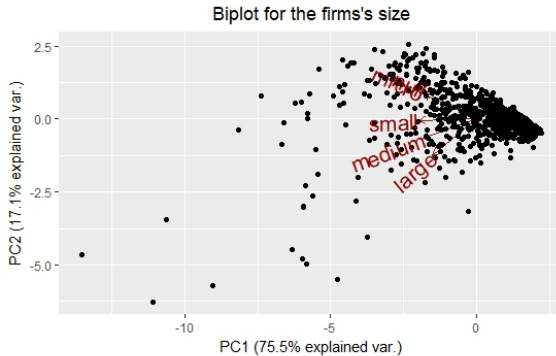
# Prediction for young people using elastic net and 10-folds CV



# Distribution of firms per town



# PCA



# Issues

- A lot of NA in geo locations (retrieved from Google API)
- Unique code for salary data 1/7 of the total
- Missing additional information
- French DOM-TOM regions
- Outliers and spatial correlation

# Future works

- Combine the separated datasets
- Create meaningful indicators
- Take correlation into account (especially spatial)
- Perform clustering techniques to identify geographical clusters
- Perform groupwise lasso to predict salary data
- Verification/improvement of the obtained results
- Compare the methodologies used with robust ones
- Find complementary datasets



– *Thank you* –