

# Banking predictions

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# Data presentation

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Marketing campaign of a Portuguese banking institution

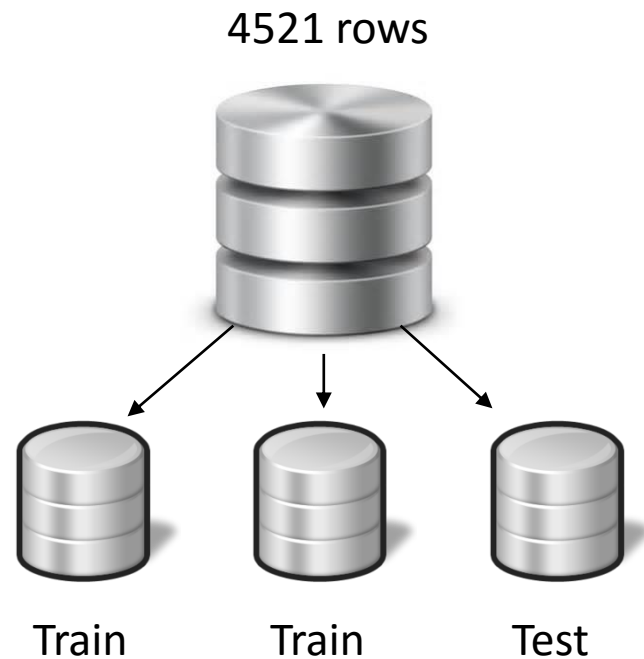
Phone calls

| Age  | Job      | Marital status    | Education           | Default           | Loan           | Contact |
|------|----------|-------------------|---------------------|-------------------|----------------|---------|
| Date | Duration | Number of contact | Days last contacted | Previous campaign | Has subscribed |         |

=> Tasks repartition

# Preparing to modeling

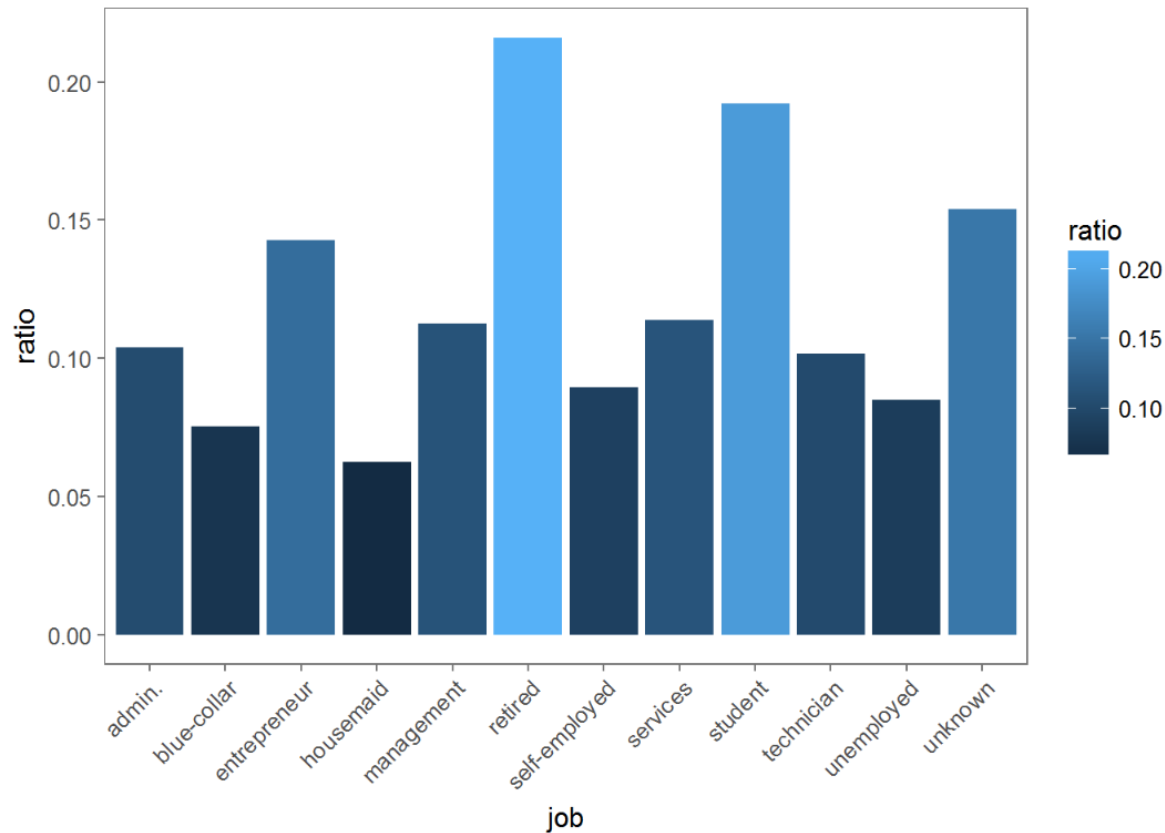
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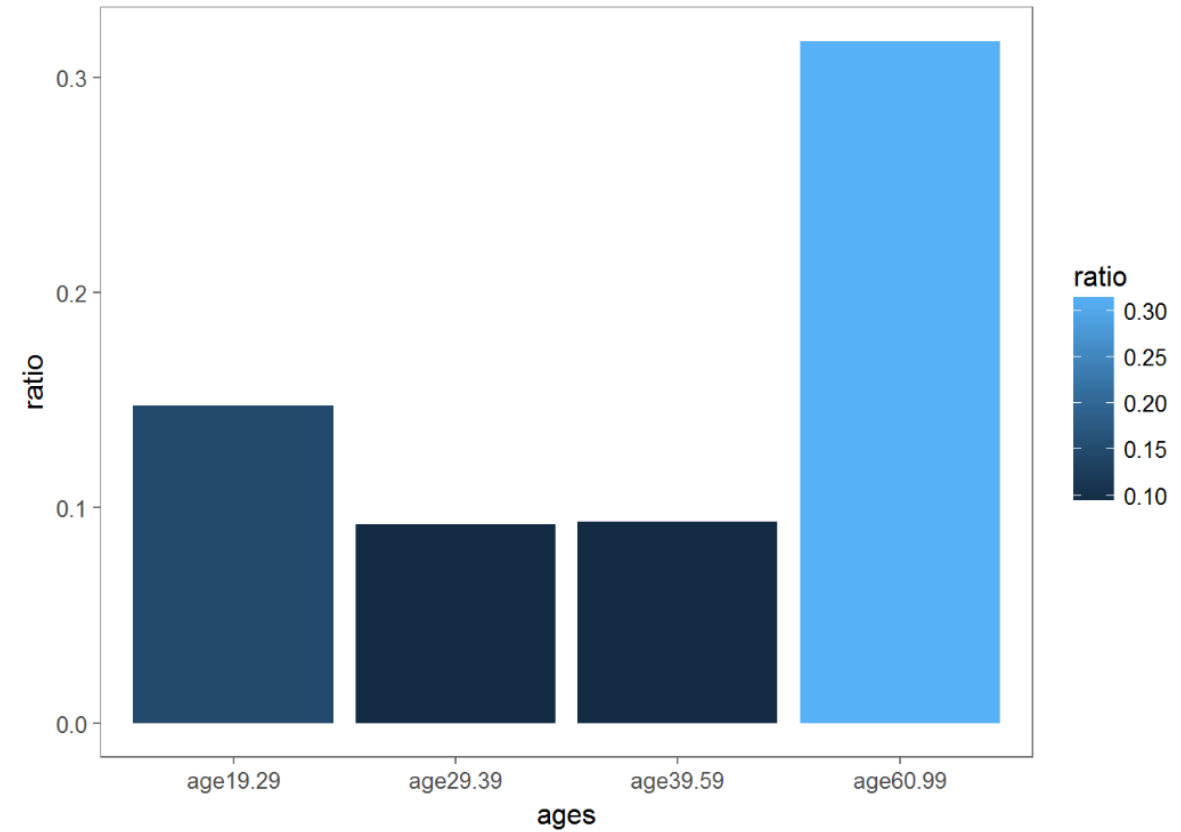
## Overview

```
$ age      : int   30 33 35 30 59 35 36 39 41 43 ...
$ job      : Factor w/ 12 levels "admin.", "blue-collar",...: 11 8 5 5 2 5 7 10 3 8 ...
$ marital  : Factor w/ 3 levels "divorced", "married",...: 2 2 3 2 2 3 2 2 2 2 ...
$ education: Factor w/ 4 levels "primary", "secondary",...: 1 2 3 3 2 3 3 2 3 1 ...
$ default  : Factor w/ 2 levels "no", "yes": 1 1 1 1 1 1 1 1 1 1 ...
$ balance  : int   1787 4789 1350 1476 0 747 307 147 221 -88 ...
$ housing  : Factor w/ 2 levels "no", "yes": 1 2 2 2 2 1 2 2 2 2 ...
$ loan     : Factor w/ 2 levels "no", "yes": 1 2 1 2 1 1 1 1 1 2 ...
$ day      : int   19 11 16 3 5 23 14 6 14 17 ...
$ month    : Factor w/ 12 levels "apr", "aug", "dec",...: 11 9 1 7 9 4 9 9 9 1 ...
$ duration : int   79 220 185 199 226 141 341 151 57 313 ...
$ campaign : int    1 1 1 4 1 2 1 2 2 1 ...
$ pdays    : int   -1 339 330 -1 -1 176 330 -1 -1 147 ...
$ previous : int    0 4 1 0 0 3 2 0 0 2 ...
$ poutcome : Factor w/ 4 levels "failure", "other",...: 4 1 1 4 4 1 2 4 4 1 ...
$ y        : Factor w/ 2 levels "no", "yes": 1 1 1 1 1 1 1 1 1 1 ...
```

# Data inferences



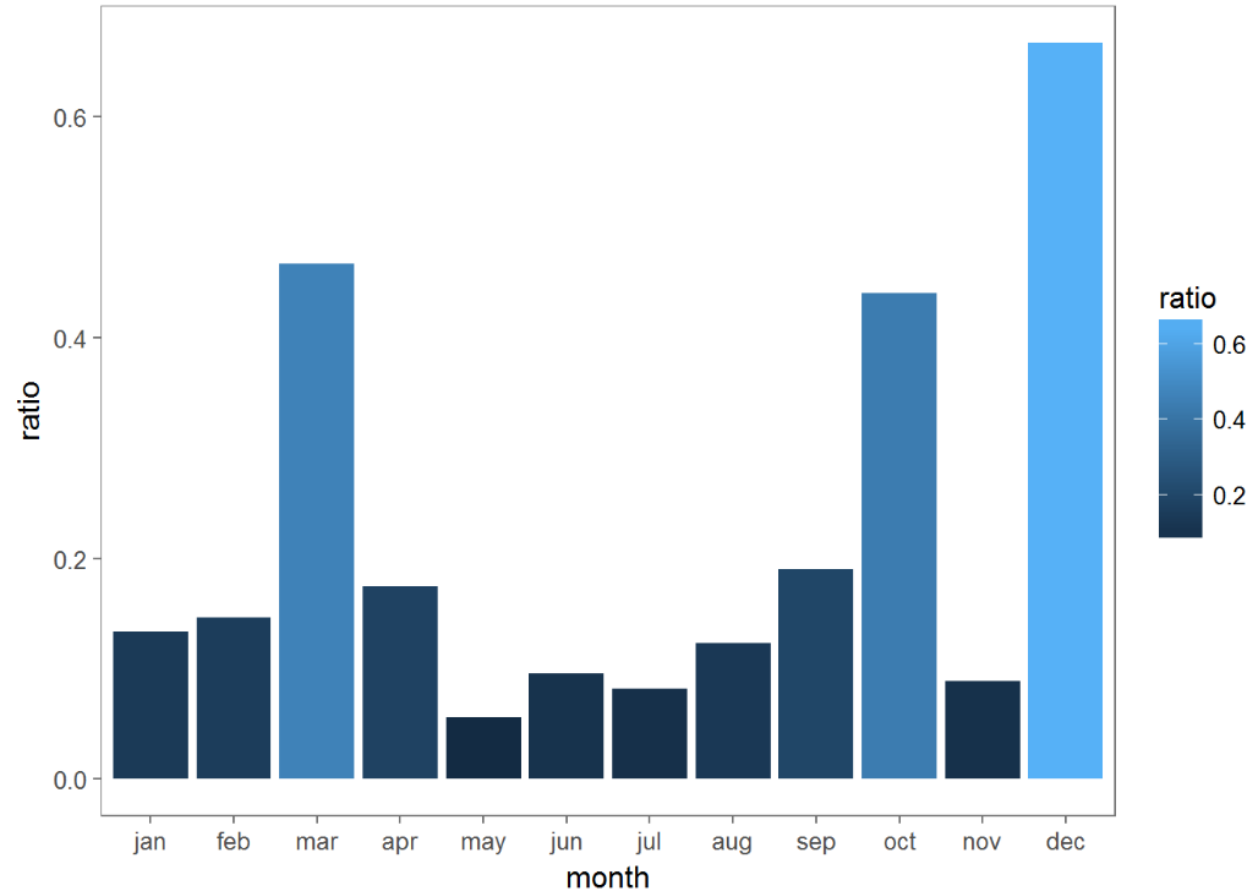
*Ratio of product contractors per job*



*Ratio of product contractors per age range*

# Data inferences

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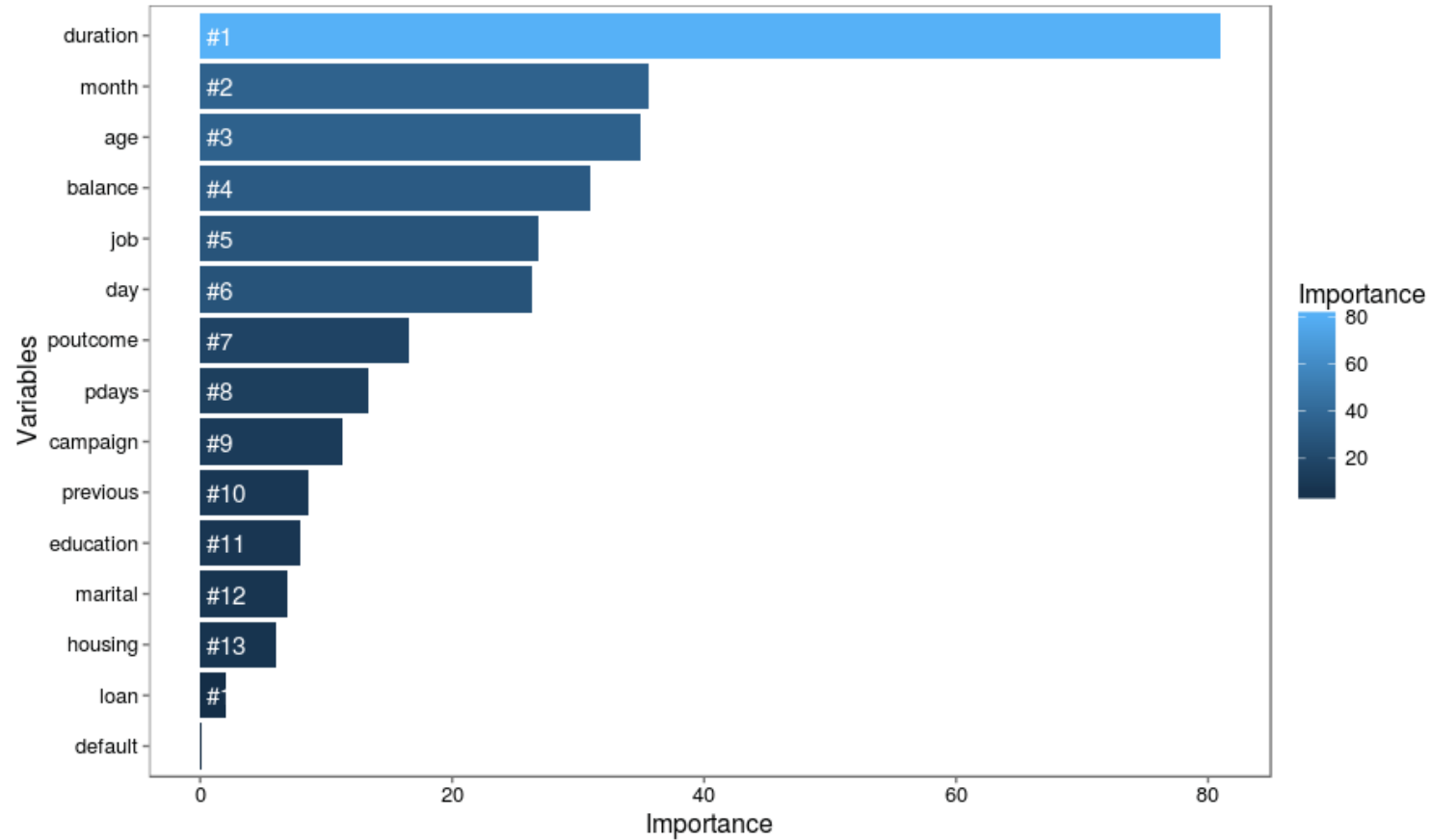
*Ratio of product contractors per month*

# Choosing the best model

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| Model type                     | Accurcay (%) | AUC   | Std. Dev |
|--------------------------------|--------------|-------|----------|
| Logistic regression            | 91           | 0,945 | 0.0059   |
| Classification tree            | 89           | NA    | 0.0061   |
| Naive Bayes                    | 86           | 0,821 | 0.0090   |
| Random Forest                  | 90           | 0,664 | 0.0060   |
| Stacking (Logistic Regression) | 89           | 0,928 | NA       |

# Variable importance with RF



*Variable importance ranking*

# Marketing campaign

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Goal : see if a targeted advertising campaign could be usefull for the bank

| –                      | Original dataset | Modified dataset |
|------------------------|------------------|------------------|
| Number of subscription | 159              | 619              |
| Ratio of subscribers   | 0.1065684        | 0.4148794        |

*Simulated campaign results*



# Conclusion

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## ➤ Our analysis

- month, balance, age, duration, job are the most relevant predictors

## ➤ Future work

- Improve parameters of baseline classifiers
- Use voting stacking system
- Deeper analysis of each dataset parameter to highlight other axes of study