

02/24/21

## Introduction

#### Data

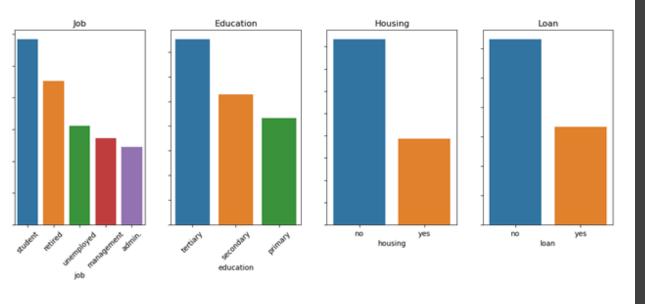
➤ Portuguese bank (2008 – 2010) throughout 17 past campaigns

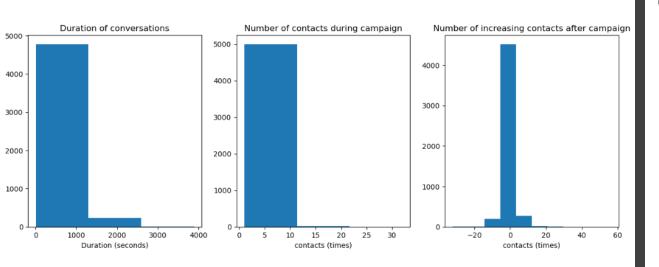
#### Reason

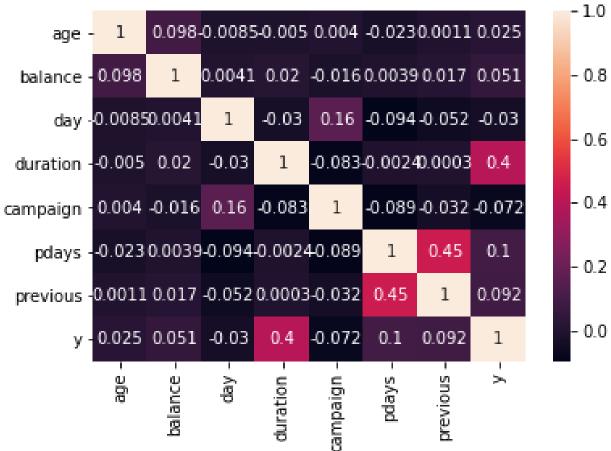
➤ Large number of records from customers

#### Goal

- **≻**EDA
- > Feature Engineering
- ➤ Model Selection
- **→** Optimization
- > Feature Importance
- ➤ Recommendations







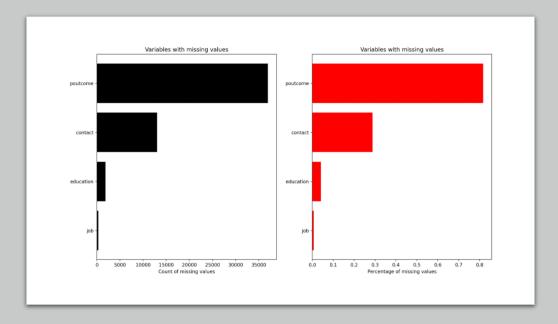
# **Data Cleaning**

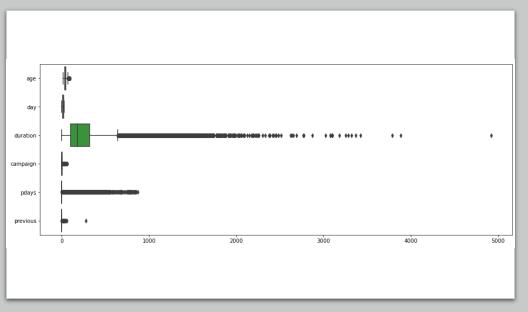
#### **Continuous variables**

Keep them for analysis of extreme values

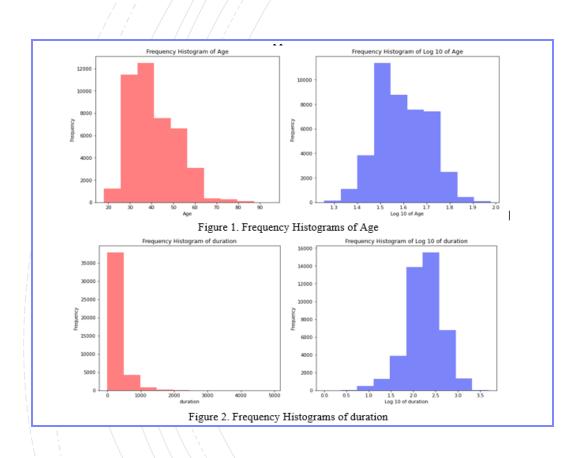
### **Categorical variables**

- Remove variable Outcome
- Remove missing rows from education and job
- Impute Contact





## Feature Engineering & Selection



### Feature Engineering

- Dummy variable encoding
- Log transformation

#### Feature Selection

- Feature importance selection
- Dimension Reduction
- Split Data Set(80/20)

## **Model Selection**



## **Logistic Regression**

Simple and easy to interpret



### **Random Forest**

Decision Trees

Good at predicting noise data set

Reduces overfitting



## **Gradient Boosting**

Learning rate
Real-time solution

### Model Accuracy & Optimization

**Logistic Regression** 

87%

LOG

**TRANSFORMATION** 

**GRIDSEARCHCV** 



**Logistic Regression** 

89%

Random Forest & Gradient Boosting

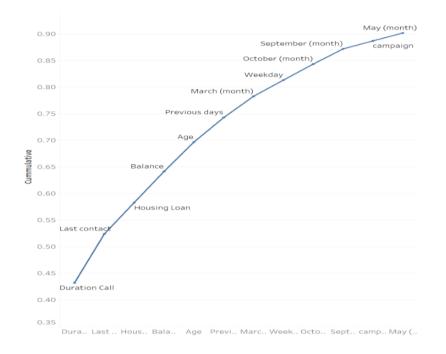
90%



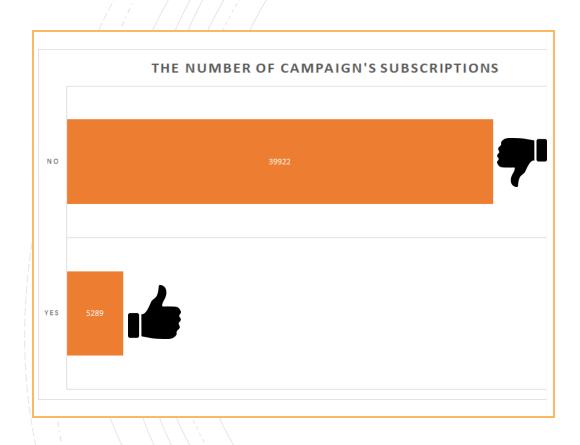
**Random Forest** 

#### Feature Importance





### Recommendation



### **Duration of Call**

- Create products that allows customers to be more interested.
- Improve customer service.

### **Housing Loan**

- Separate customer with housing loan and without housing loan.
  - Create a new product with lower balance for customer with housing loan.

## **Conclusion**





BUILD A MODEL TO SAVE TIME AND INCREASE PROFIT FOR BANK.

RECOMMENDED AREAS FOR THE BANK TO IMPROVE TO MAKE THE NEXT CAMPAIGN MORE SUCCESSFUL.

## References

 Moro, Cortez, & Rita. A Data-Driven Approach to Predict the Success of Bank Telemarketing. Decision Support Systems, Elsevier, 62:22-31, June 2014