

Exploratory Data Analysis Bank Marketing (Campaign)

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Agenda

Problem Statement
Data Analysis
Data Cleaning
EDA
Final Recommendations
GitHub Repo Link



Problem Statement

Overview

ABC Bank wants to sell its term deposit product to customers and before launching the product they want to develop a model which help them in understanding whether a particular customer will buy their product or not (based on customer's past interaction with bank or other Financial Institution).

Solution

By analysing the dataset we will be able to convert this problem into a machine learning classification and build a model to predict whether a client will subscribe a term deposit or not.



Data Analysis

- 21 Features
- 41188 rows

Assumptions:

- The data seems to be cleaned and a little bit skewed, however you can see that the variables have outliers that need to be cleaned with data cleaning process.
- There are no null values, but there are some "unknown" values.
- We've got 41188 rows, some of the columns has 85% of the values repeated.
- The value types seem to be correct but as said there are some "unknown" values.

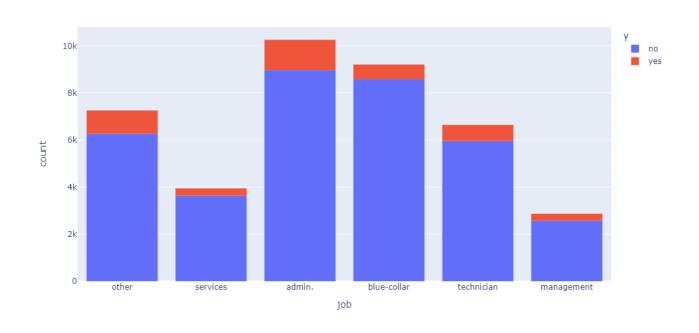
Data Cleaning

The most used techniques to treat the data frame were:

- Dropping columns and rows
- Replacing missing values with the mode
- Filling outliers with ffill/bfill based on interquantile range
- Grouping text features that doesn't appear many time

Cleaned dataset reduced from (41188, 21) to (40165, 18)

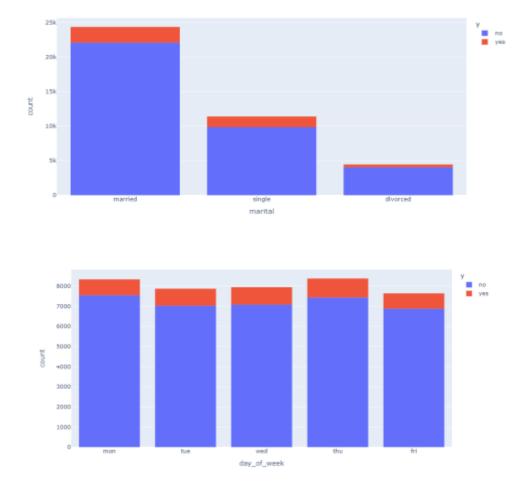
EDA – Categorical Attributes



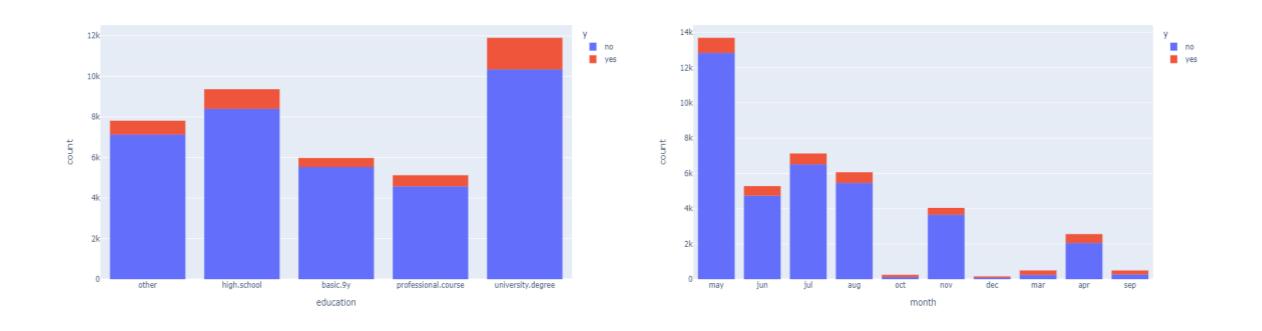
Job: Highest number of subscriptions to a term deposit work on admin.

Marital: Most of the clients approached were married.

Day of Week: There's no significant difference in the numbers of clients approached and people subscribed.

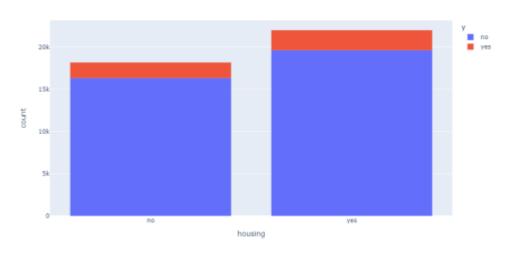


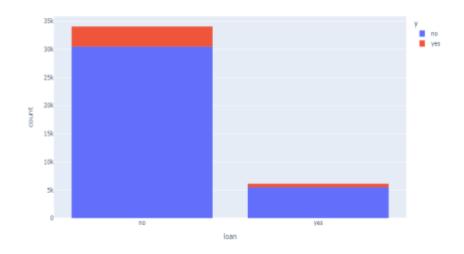
EDA – Categorical Attributes

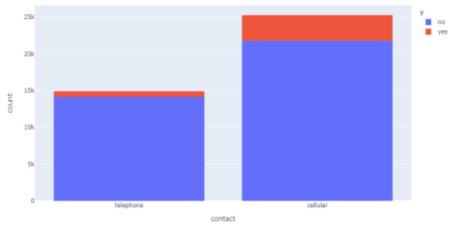


Education: In the education column we can see that most people who has subscribed have a university degree.

Month: The last contact month of year was way bigger in may.



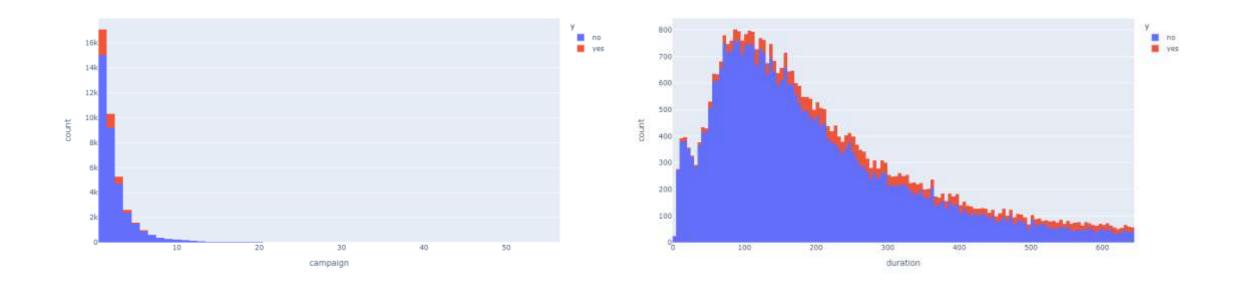




Housing: A housing loan does not have much of effect on the number of term deposits purchased.

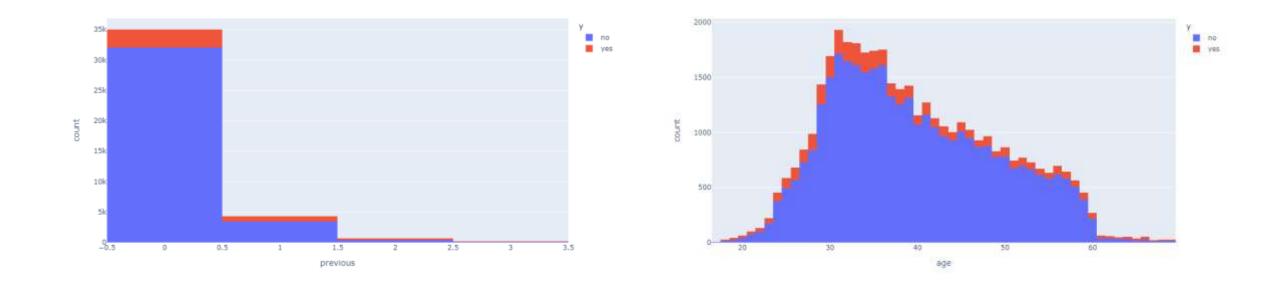
Loan: Most of the clients approached did not have a personal loan.

Contact: The contact by phone seems to be more effective.



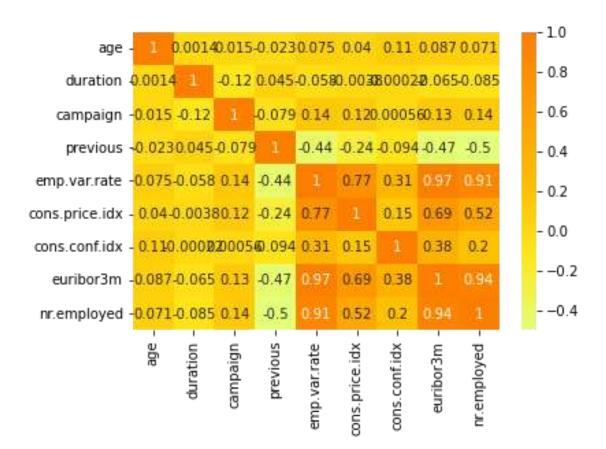
Campaign: Most of the numbers of contacts performed during the campaign and for the client were only once or twice.

Duration: The duration of the last contact is highly variated.



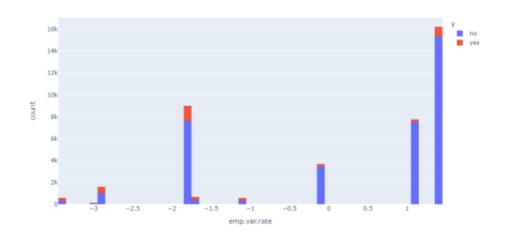
Previous: The number of contacts performed before this campaign and for a specific client was 0 in more than 80% of the cases

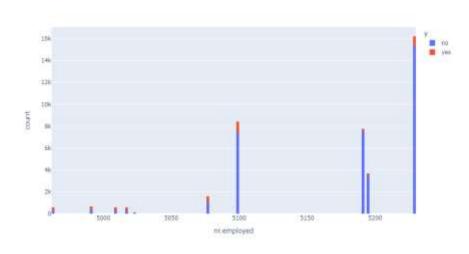
Age: Most of the people contacted was between 30 and 40 years old.

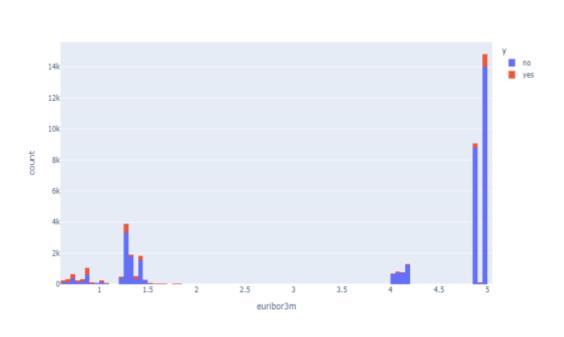


As we can see in this heatmap, the columns 'euribor3m', 'nr.employed' and 'emp.var.rate' are highly correlated.

You can see in the next page their graphics when comparing who did and who did not subscribe to a term deposit.







Final Recommendations

Now the project is almost finished and there are, basically, 2 important things left to do:

- Perform features encode:
- Turn words / yes or no / true or false into 0, 1, etc.
- Get dummy variables
- WoE encode
- Explore classification models to predict if the clients have the tendence to subscribe to a term deposit:
- Logistic Regression
- Decision Tree Classifier
- Random Forest Classifier
- Gaussian NB (Naive Bayes)
- SVC

GitHub Repo Link

https://github.com/sharuhinda/bank marketing campaign/tree/review https://github.com/vgiih/EvolveData

Thank You!

