

Exploratory Data Analysis BANK MARKETING CAMPAIGN

Background and Agenda

ABC Bank wants to sell it's 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).

The data is related with direct marketing campaigns of a Portuguese banking institution. The marketing campaigns were based on phone calls. Often, more than one contact to the same client was required, in order to access if the product (bank term deposit) would be ('yes') or not ('no') subscribed.

There are four datasets:

- 1) bank-additional-full.csv with all examples (41188) and 20 inputs, ordered by date (from May 2008 to November 2010), very close to the data analyzed in [Moro et al., 2014]
- 2) bank-additional.csv with 10% of the examples (4119), randomly selected from 1), and 20 inputs.
- 3) bank-full.csv with all examples and 17 inputs, ordered by date (older version of this dataset with less inputs).
- 4) bank.csv with 10% of the examples and 17 inputs, randomly selected from 3 (older version of this dataset with less inputs).

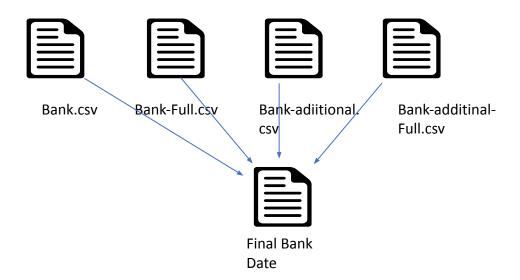
The smallest datasets are provided to test more computationally demanding machine learning algorithms (e.g., SVM).

The classification goal is to predict if the client will subscribe (yes/no) a term deposit (variable y).



Data Exploration

- 1) Bank-additional-full.csv with all examples (41188) and 20 inputs, ordered by date (from May 2008 to November 2010), very close to the data analyzed in [Moro et al., 2014]
- 2) Bank-additional.csv with 10% of the examples (4119), randomly selected from 1, and 20 inputs.
- 3) Bank-full.csv with all examples and 17 inputs, ordered by date (older version of this dataset with less inputs).
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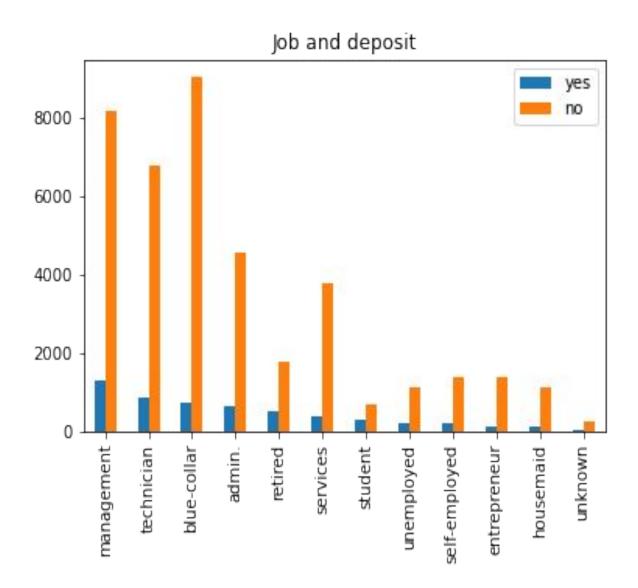
The Bank Data contained four(4) files, namely; Bank, Bank-Full, Bank-additional and the Bank-additional-full documents and all were in the >.csv format.



ANALYSIS OF JOBS AND IMPACT ON DEPOSITS

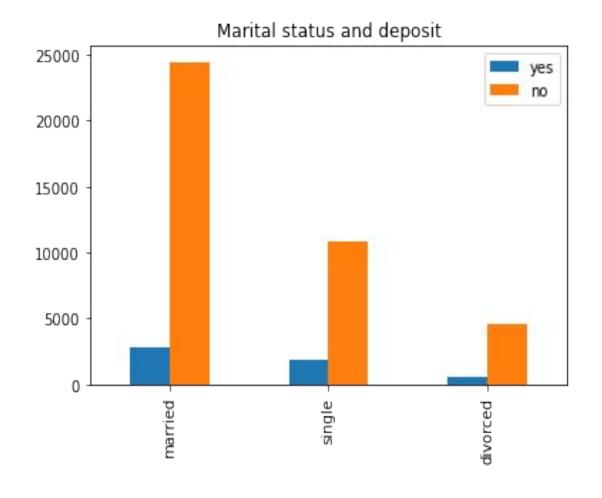


JOBS AND DEPOSITS ANALYSIS



ANALYSIS OF MARITAL STATUS AND IMPACT ON **DEPOSITS**

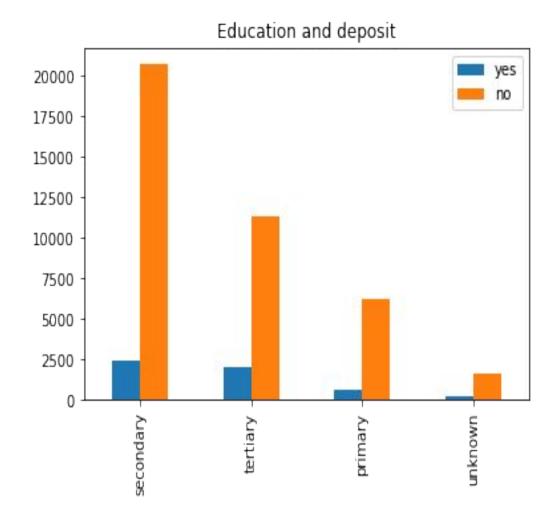
MARITAL STATUS AND DEPOSITS ANALYSIS





ANALYSIS OF EDUCATION AND IMPACT ON DEPOSITS

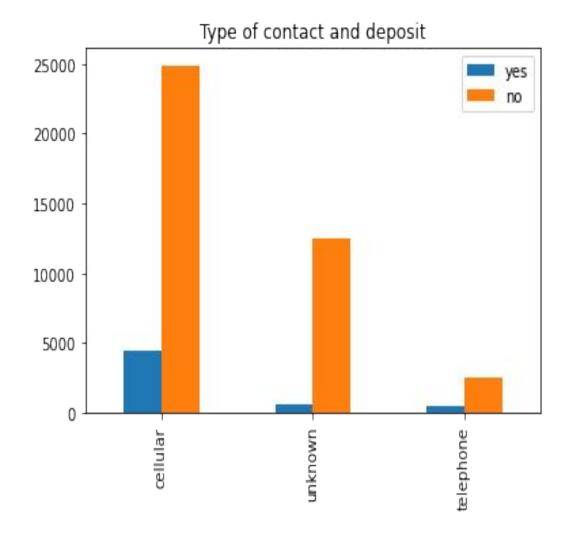
EDUCATION AND DEPOSITS ANALYSIS





ANALYSIS OF THE METHOD OF CONTACT WITH **CUSTOMER AND** IMPACT ON **DEPOSITS**

METHOD OF CONTACT AND DEPOSITS ANALYSIS



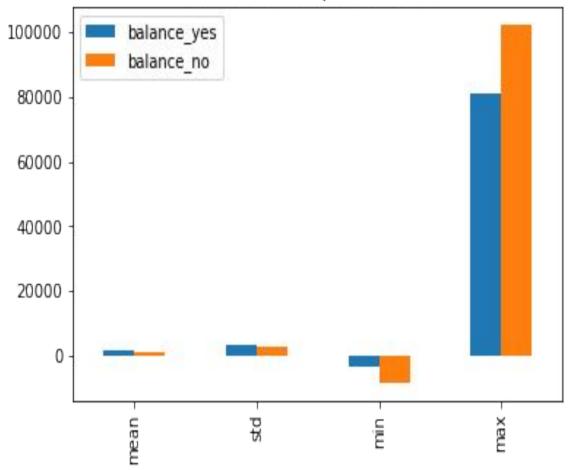


ANALYSIS

OF CUSTOMER'S BANK BALANCE AND IMPACT ON DEPOSITS

BANK BALANCE AND DEPOSITS ANALYSIS



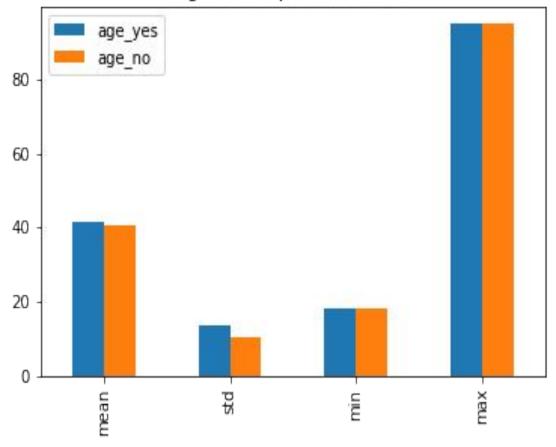




ANALYSIS OF **CUSTOMER'S AGE AND** IMPACT ON **DEPOSITS**

CUSTOMER'S AGE AND DEPOSITS ANALYSIS

Age and deposit statistics

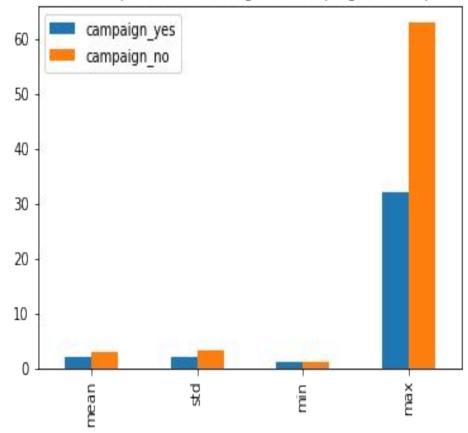




ANALYSIS OF THE **NUMBER OF** CONTACTED ON DEPOSITS

NUMBER OF CONTACT AND DEPOSITS ANALYSIS

Number of contacts performed during this campaign and deposit statistics





FINAL RECOMMENDATION AND FINDINGS

From Analysis we can tell according to our dataset:

- 1. Customers with 'blue-collar' and 'services' jobs are less likely to subscribe for term deposit.
- 2. Married customers are less likely to subscribe for term deposit.
- 3. Customers with 'cellular' type of contact are less likely to subscribe for term deposit.
- 4. People who subscribed for term deposit tend to have greater balance and age values.
- 5. People who subscribed for term deposit tend to have fewer number of contacts during this campaign.



PROPOSED MODELLING TECHNIQUE

THE TEAM IS GOING TO TEST OF THE FOLLOWING MODELS AND PICK WHICH HAS THE BEST ACCURACY SCORE;

- 1. LOGISTIC REGRESSION
- 2. RANDOM FOREST CLASSIFIER
- 3. SUPPORT VECTOR CLASSIFIER (SVC)
- DECISION TREE CLASSIFIER

