Capstone Project (Week 2) Bank Marketing dataset

Introduction

- In banks it is important to meet the goals that are set
- For meeting those goals banks market their product through telemarketing
- For effective marketing, banks have to identify the customers who are very like to take
- By targeting only those customers banks can get huge margin of profit

Data

- Dataset used here is of a portuguese bank
- They recorded the outcome of direct marketing campaign
- Dataset contains 45211 records and 17 attributes
- Publicly available dataset found at UCI Machine Learning Repository

Variables

```
<u>Age</u> – Numeric attribute
<u>job</u> – categorical attribute having type of job they have
<u>marital</u> – Marital status of customer (Married, single, divorced)
<u>education</u> – categorical attribute (Primary, secondary, tertiary, unknown)
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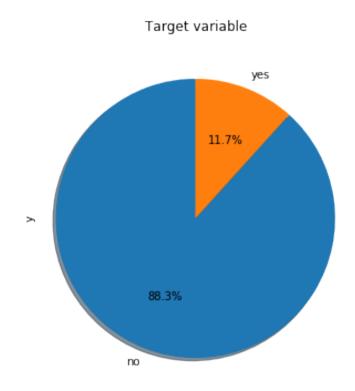
<u>default</u> – categorical attribute (has credit in default or not)
<u>balance</u> – Numeric attribute (average yearly balance in euros)
<u>housing</u> – categorical attribute (has housing loan or not)

<u>loan</u> – categorical attribute (has personal loan or not)

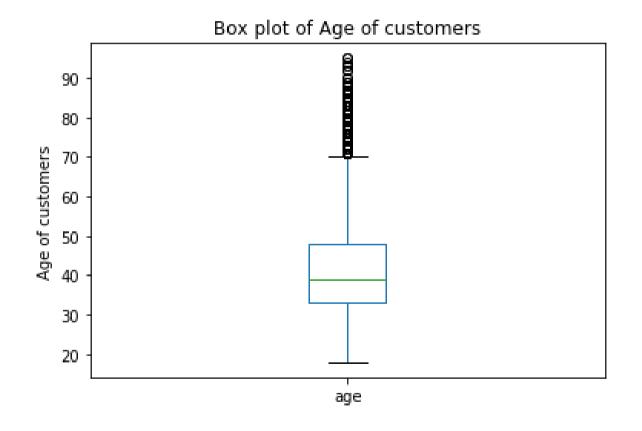
Variables cont..

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contact – mode of contact (cellular or telephone)
day – numeric (last contact day of the month)
month – categorical (last contact month of the year)
duration – Numeric (contact duration)
campaign – numeric (number of times contacted during the campaign)
pdays – numeric (number of days passed by after contacting)
previous – numeric (number of contacts performed before the campaign)
poutcome – categorical (out come of previous campaign)
y – categorical - customer subscribed to product or not (target variable)
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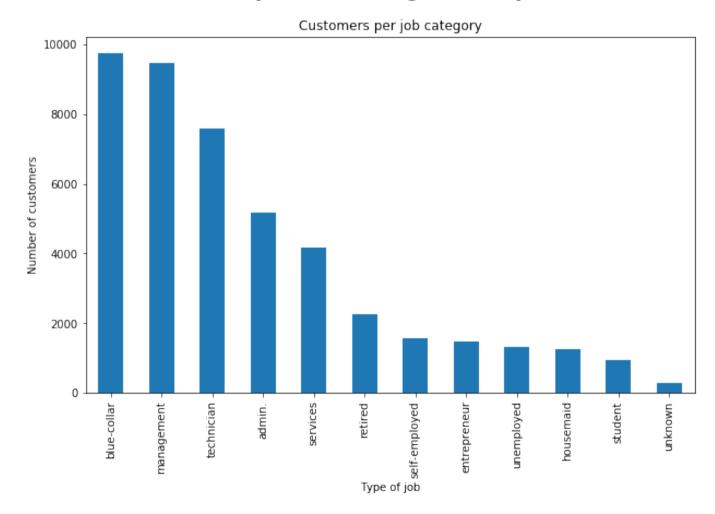
Target variable exploration



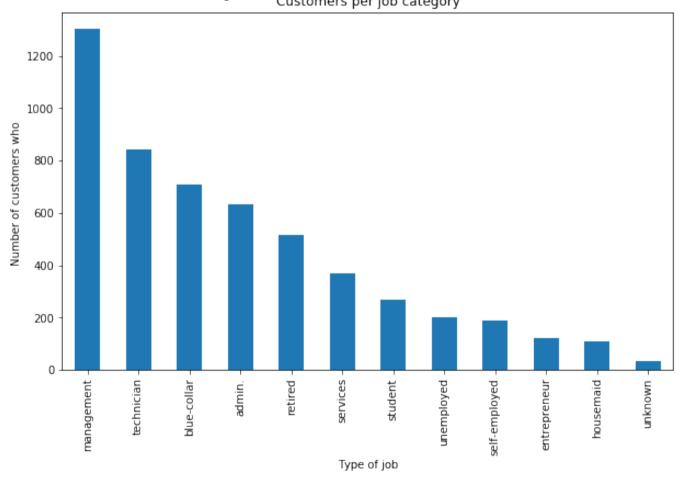
Age Attribute (Min: 18, Max: 94)



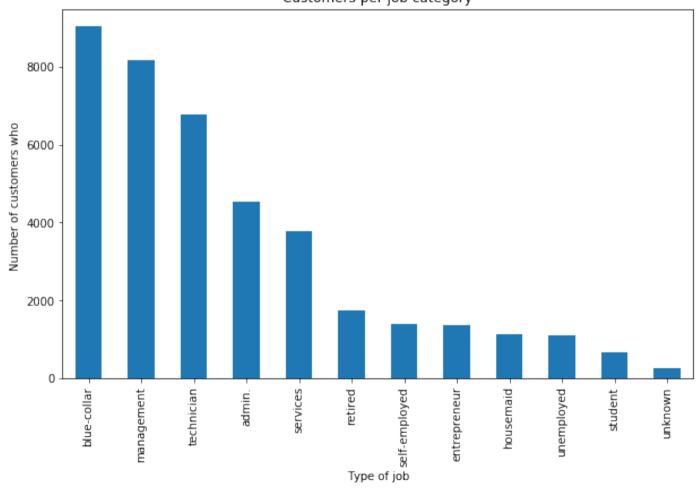
Job Attribute (12 categories)



• Job Attribute (Customers who subscribed)

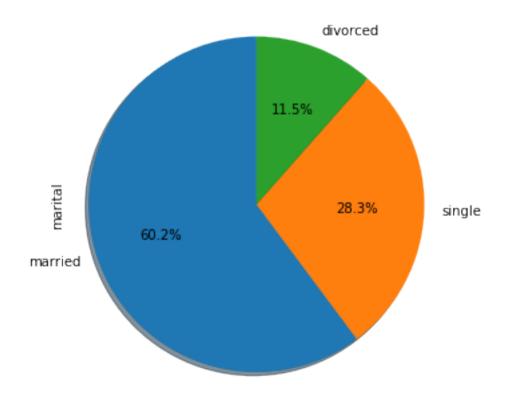


Job Attribute (Customers who not subscribed)



Marital status distribution

Pie chart of marital status



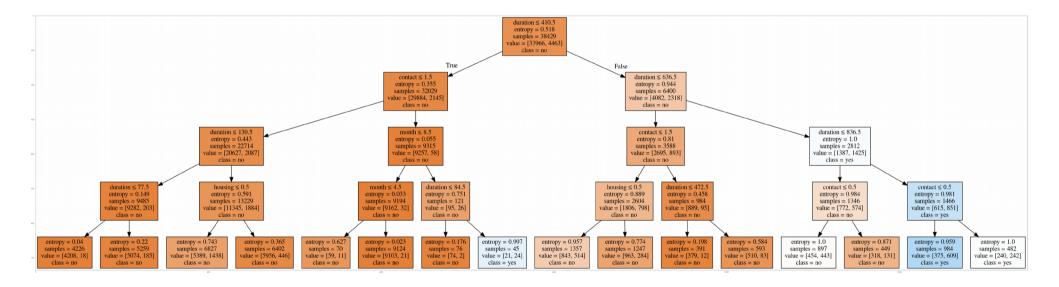
Pre-processing

- Dataset is very well formated
- It has no missing values
- Data explored for outliers
- Whole dataset considered for analysis resulting in very good accuracy

Model Building

- 3 classifiers are tested
- Logistic regression gave best results
- Decision tree can also be considered eventhough it gave less accuracy as it gives ifthen rules
- If-then rules are easy to interpret and implement

Decision tree



Results and comparision

Algorithm	Accuracy	Jaccard similarity	F1 score	Log loss
Decision Tree	0.89	0.89	0.86	NA
Logistic Regression	1	1	1	0.01
KNN	0.99	0.99	1	NA

Thank you