

# Analysis of a Marketing Campaign: From descriptive analyses to predictive modelling

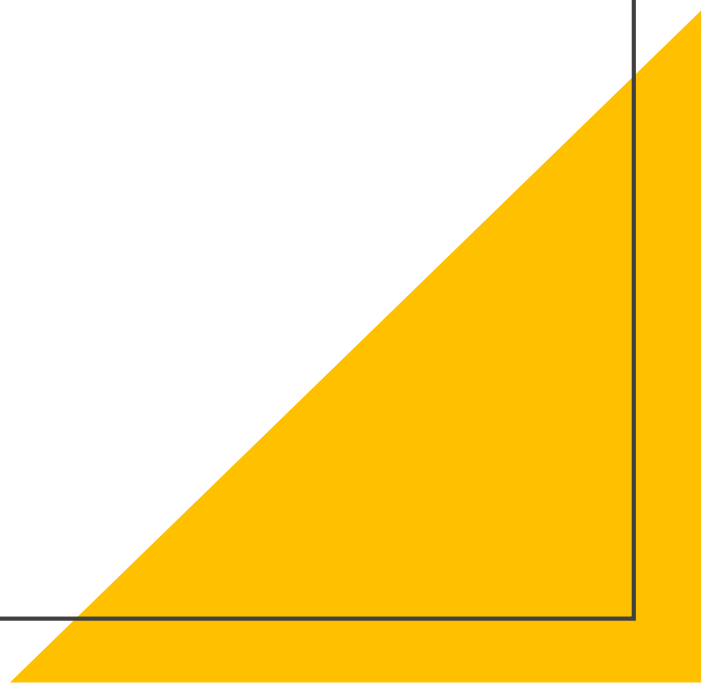
Jacopo De Angelis

PhD Candidate

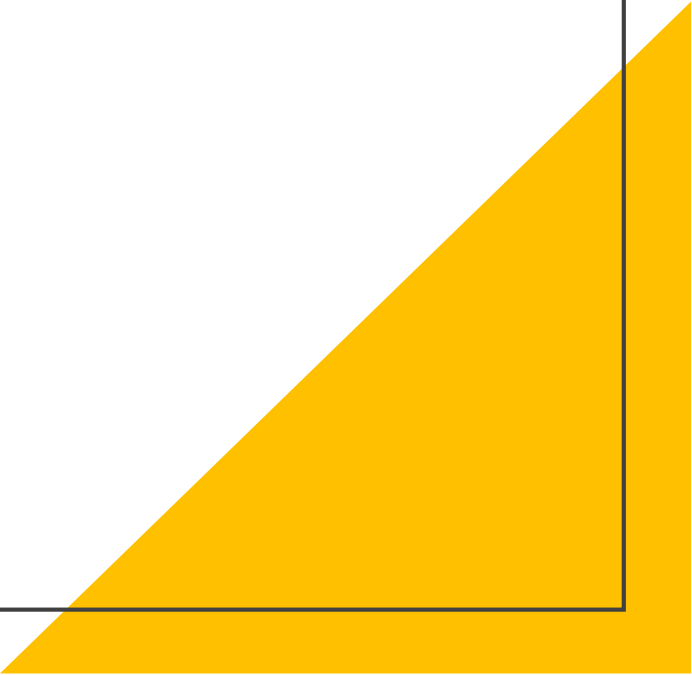
Scientific Consultant

Data Analysis & Data Science

# Topics

- ❖ Aims of the project and dataset description
  - ❖ Data exploration and visualization
  - ❖ Predictive modelling
  - ❖ Model validation
  - ❖ Most relevant predictors of campaigning success
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- A large yellow triangle is positioned in the bottom right corner of the slide, pointing towards the top right.

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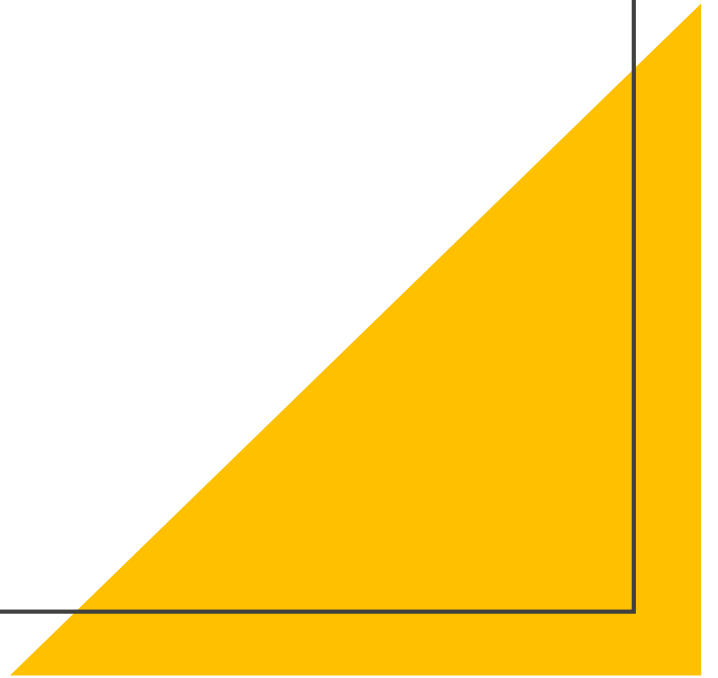
# Aims and dataset description

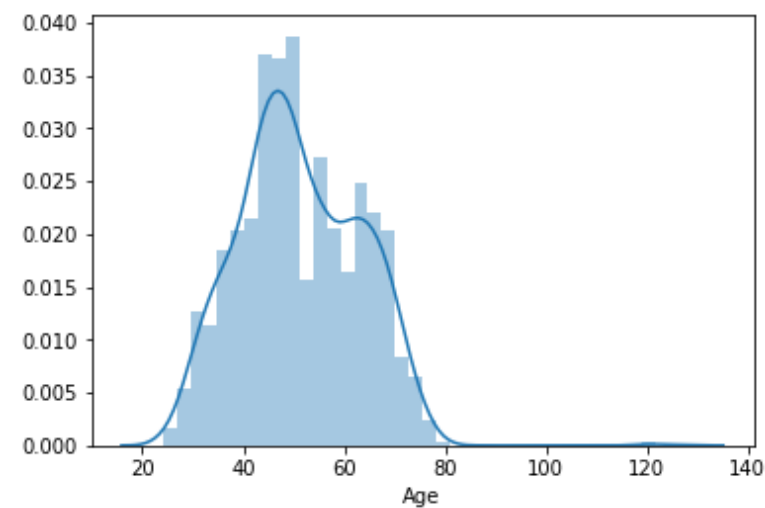
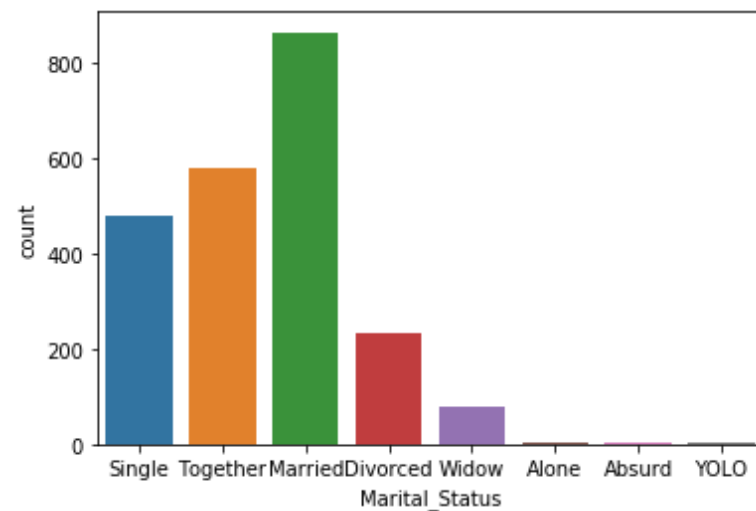


- ❖ The primary aim of this project is to develop models predicting whether a customer will respond, or not, to the marketing campaign;
- ❖ The secondary aim is to analyze which specific factors mostly predict customer responding or not to the campaign;
- ❖ The dataset has been taken from Kraggle (<https://www.kaggle.com/rodsaldanha/marketing-campaign>). It contains profile and online behavior user's data, 2240 rows x 29 columns.

# Topics

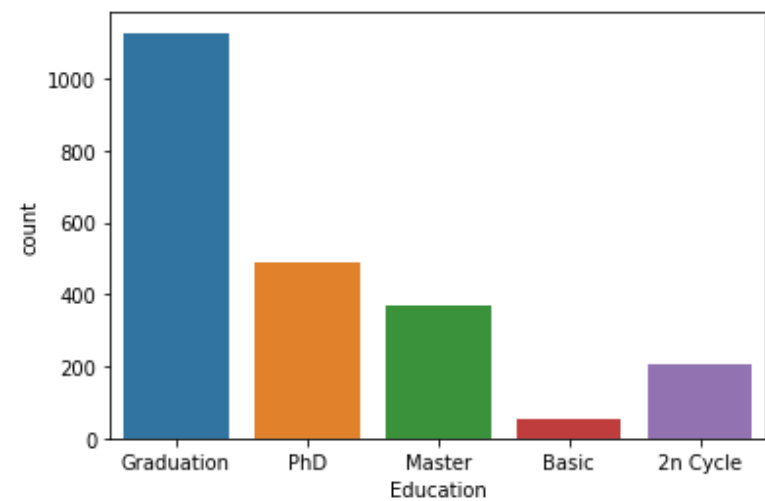
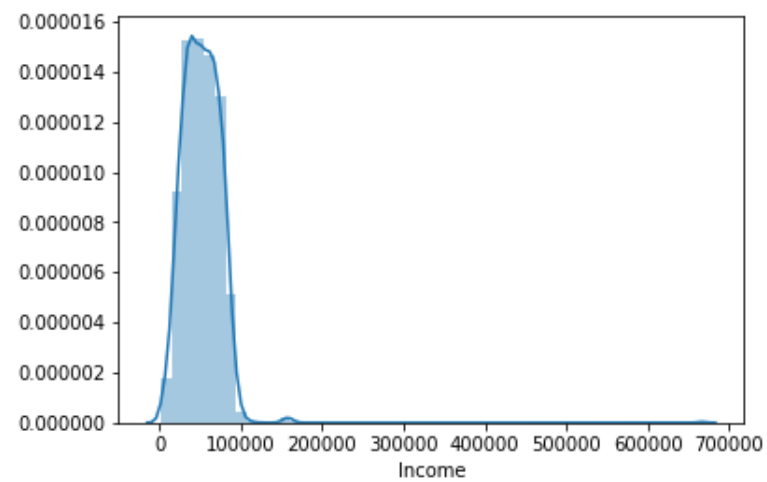
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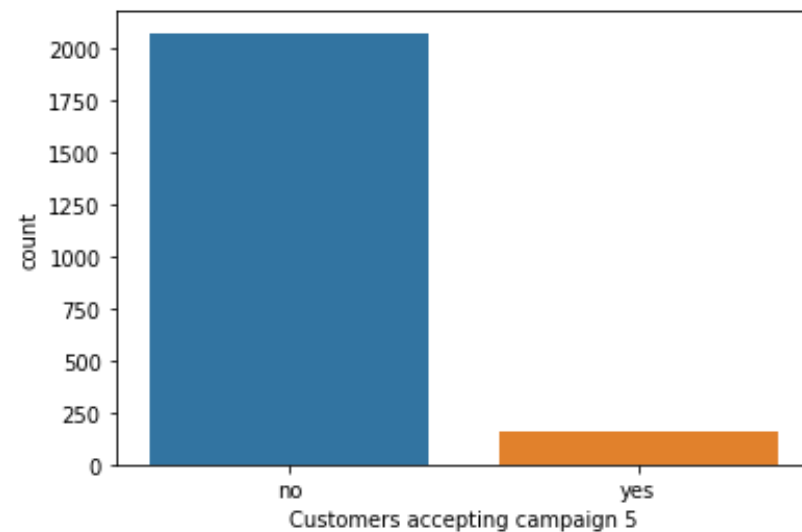
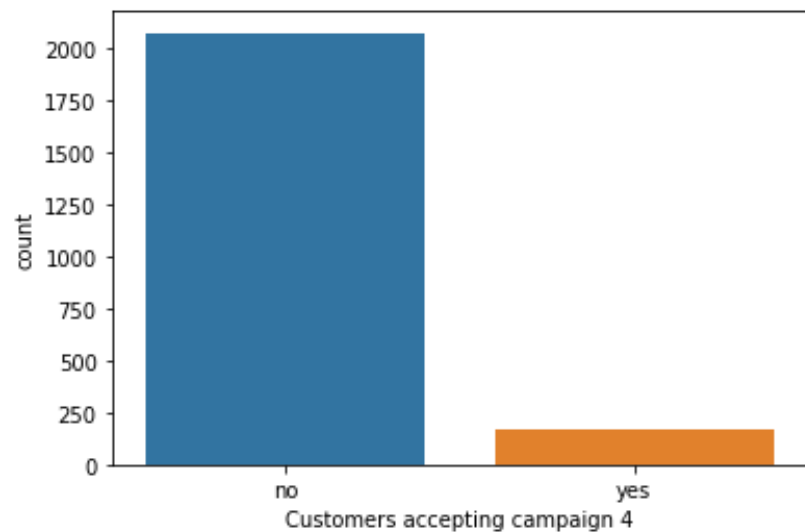
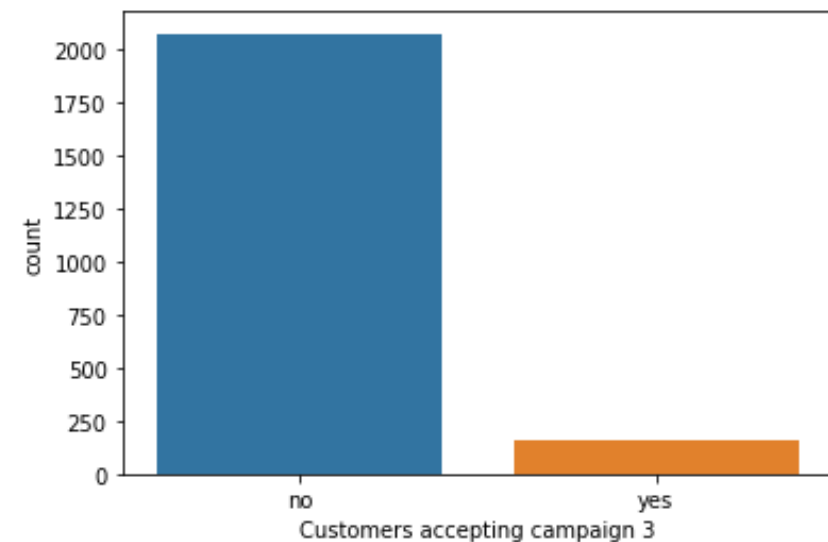
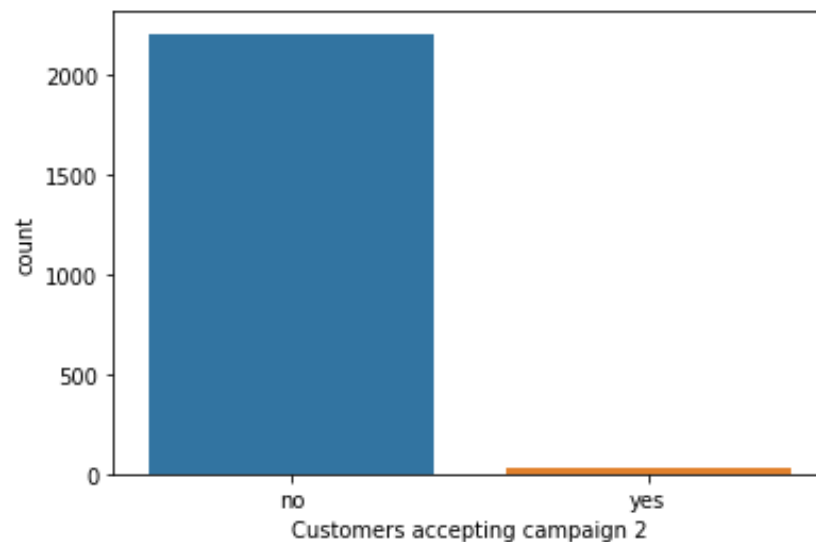
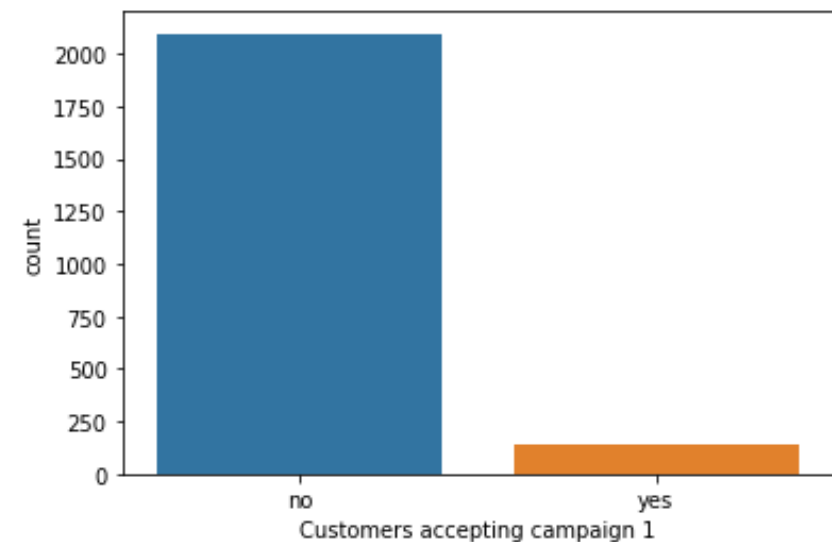


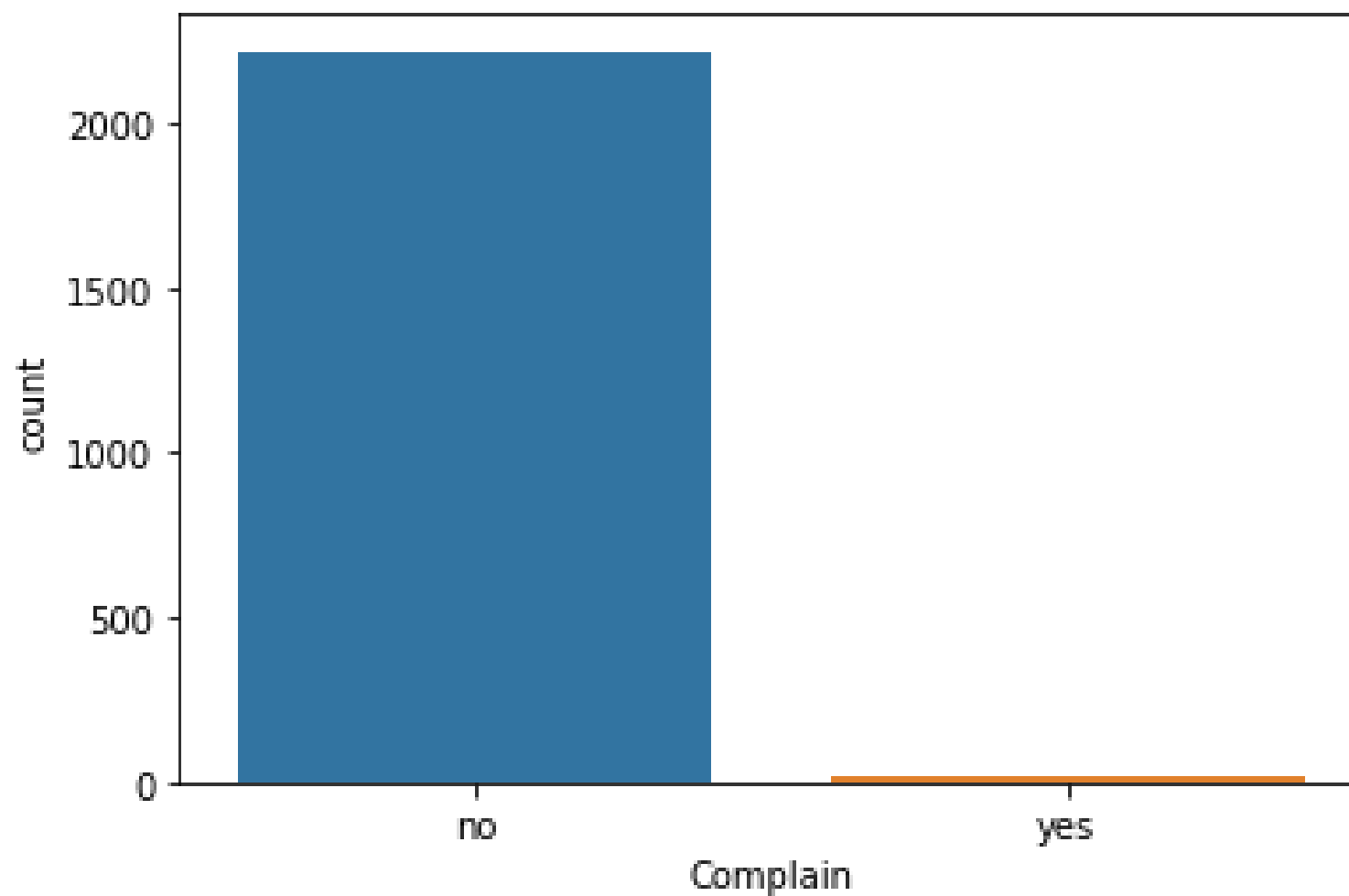
# Personal Info

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# Responses to previous campaigns

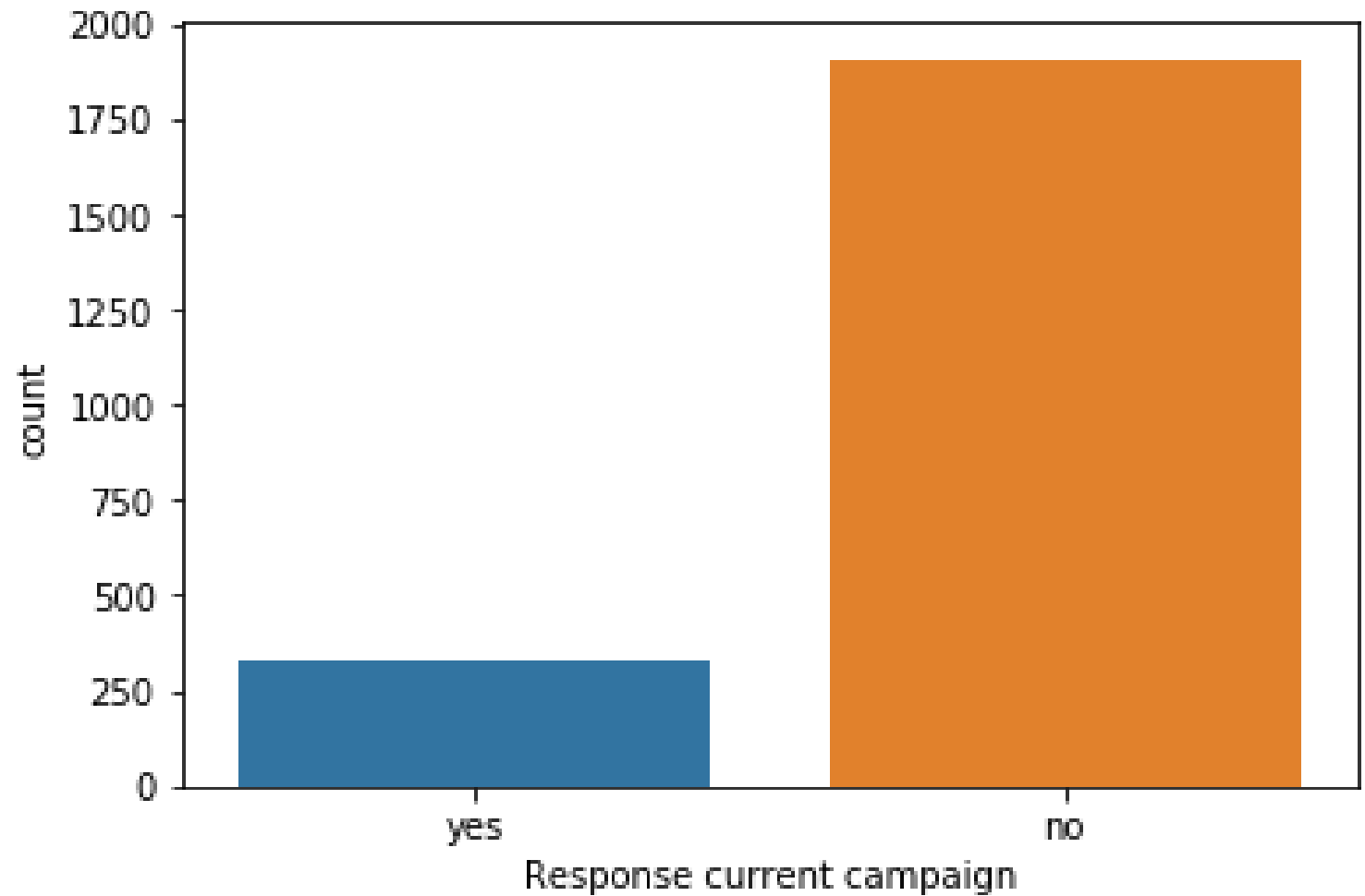




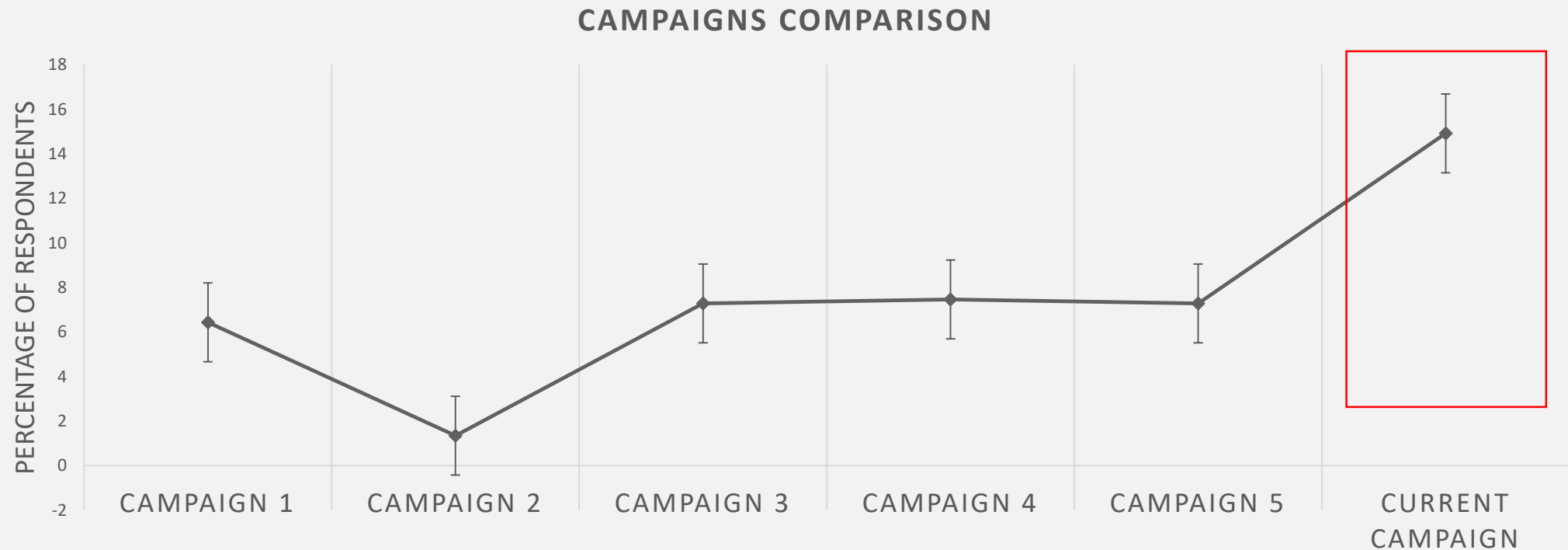
Frequency of  
complains



Number of  
responses to  
the current  
campaign



# Comparing current campaign outcome with previous campaigns outcomes

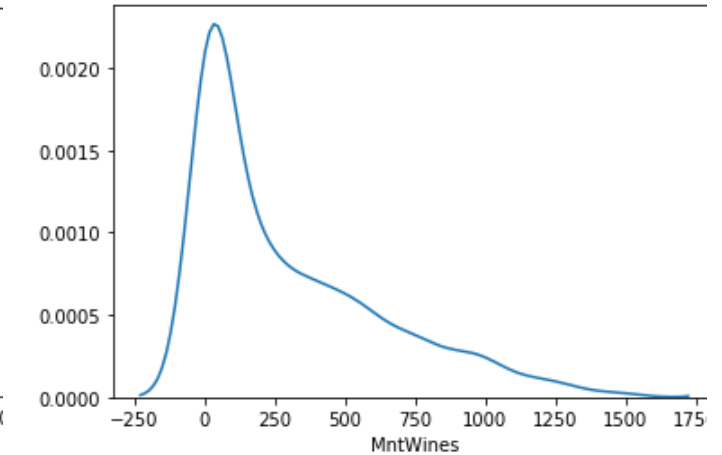
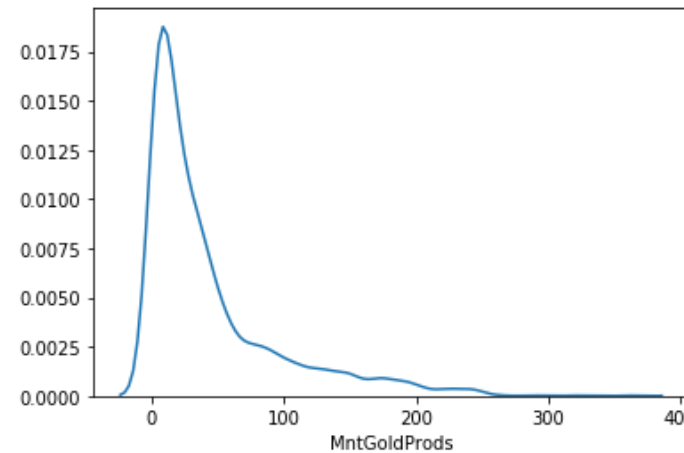
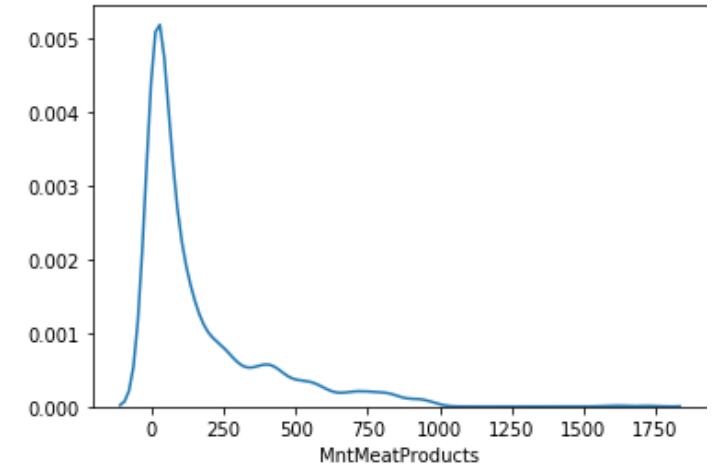
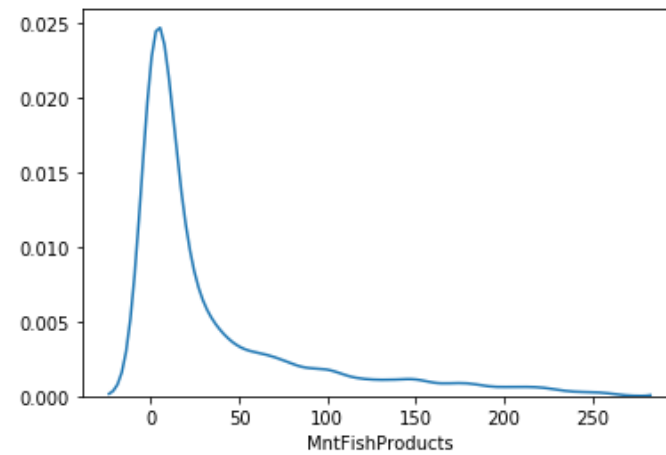


*Binomial Test comparing current campaign proportion with the average of previous campaigns' proportion significant ( $p < .001$ )*

# Amount spent on products in the last 2 years

## Amount spent on:

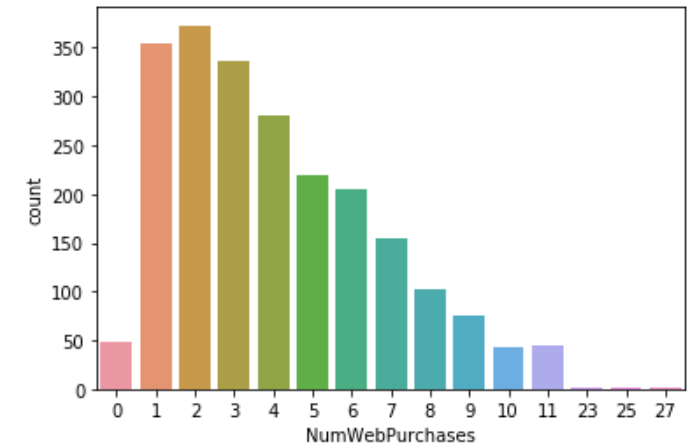
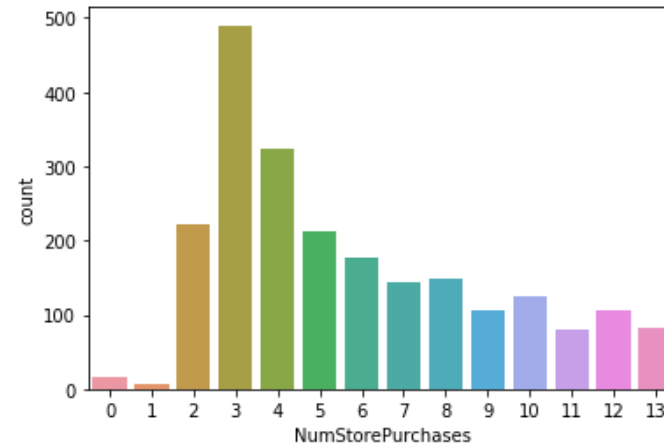
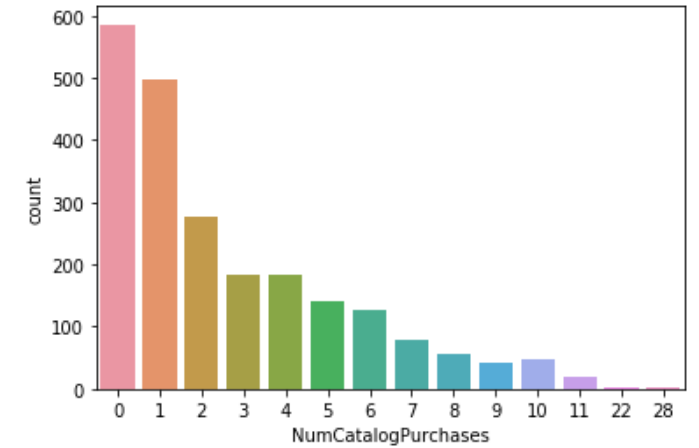
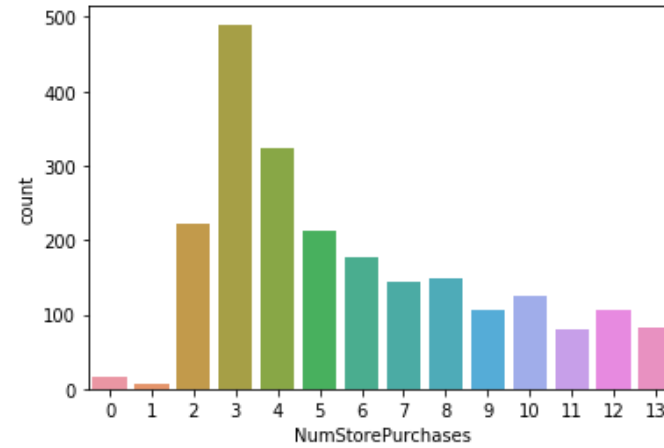
- Fish
- Gold
- Meat
- Wines



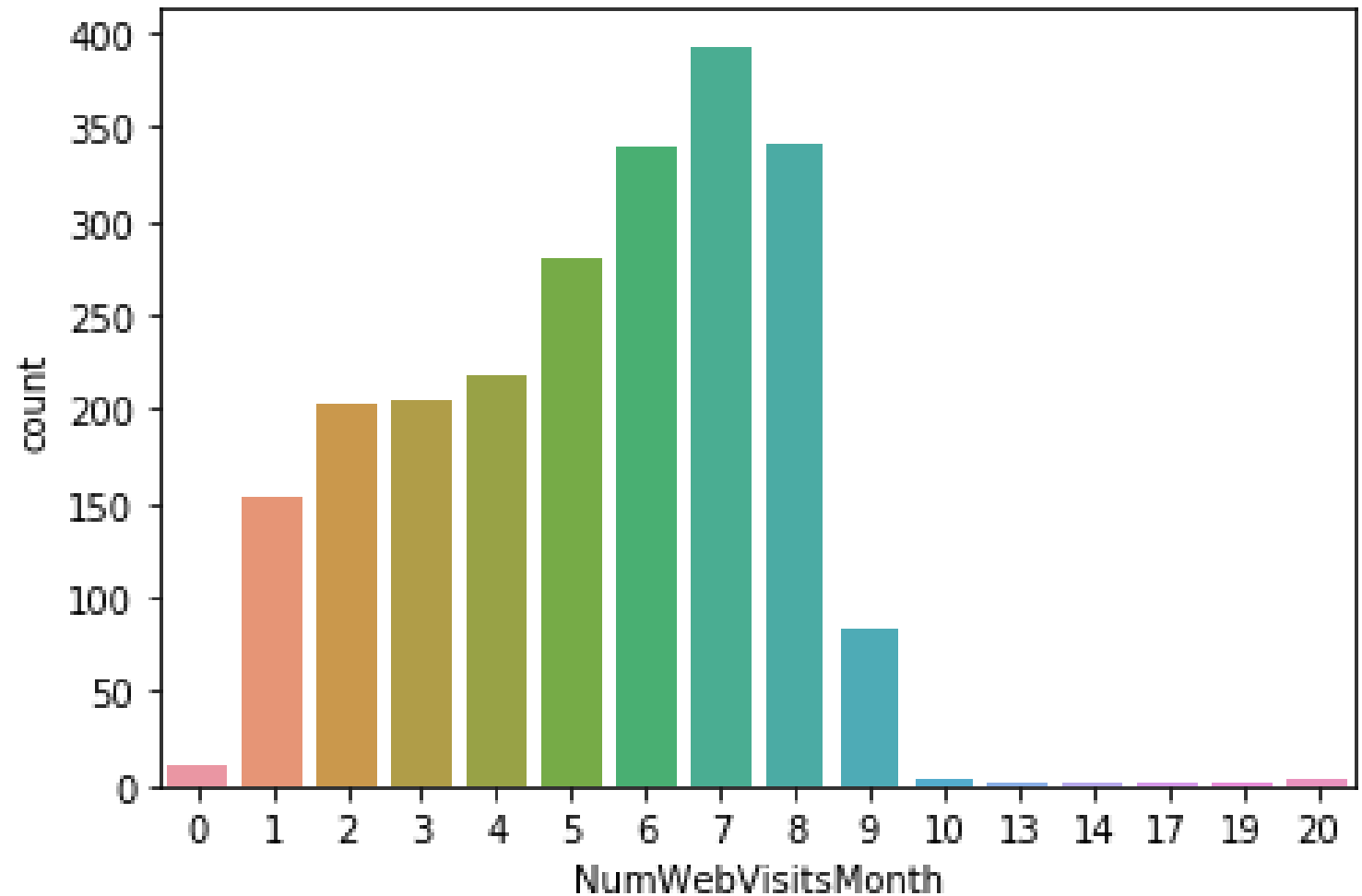
# Number of Purchases

## Options:

- Catalogue
- Discount
- Online
- Store



Number of  
web visits  
per month

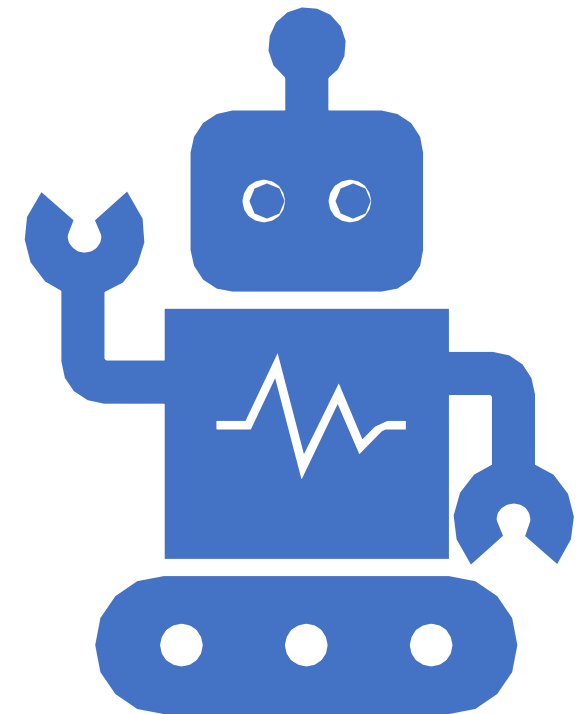


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# Steps

- ❖ Checking and filling missing values (i.e., imputer strategy);
- ❖ Feature engineering (i.e., computing age, transforming education and marital status);
- ❖ Feature scaling (i.e., standardization);
- ❖ Input selection (i.e., selecting most relevant features through recursive feature elimination methodology);
- ❖ Training Test Split (test size = 0.3);
- ❖ SMOTE procedure to balance the dataset;
- ❖ Machine Learning algorithms





# Selected model features/inputs

- ❖ Accepted Cmp 1 (i.e., respondents to campaign 1)
- ❖ Accepted Cmp 2
- ❖ Accepted Cmp 3
- ❖ Accepted Cmp 4
- ❖ Accepted Cmp 5
- ❖ Education
- ❖ Marital Status
- ❖ MntMeatProducts (i.e., Amount spent for meat products)
- ❖ NumStorePurchases (i.e., number of purchases at the store)
- ❖ NumWebVisitsMonth (i.e., number of visits on the website)
- ❖ Recency (i.e., number of days since the last purchase)
- ❖ Teens at home

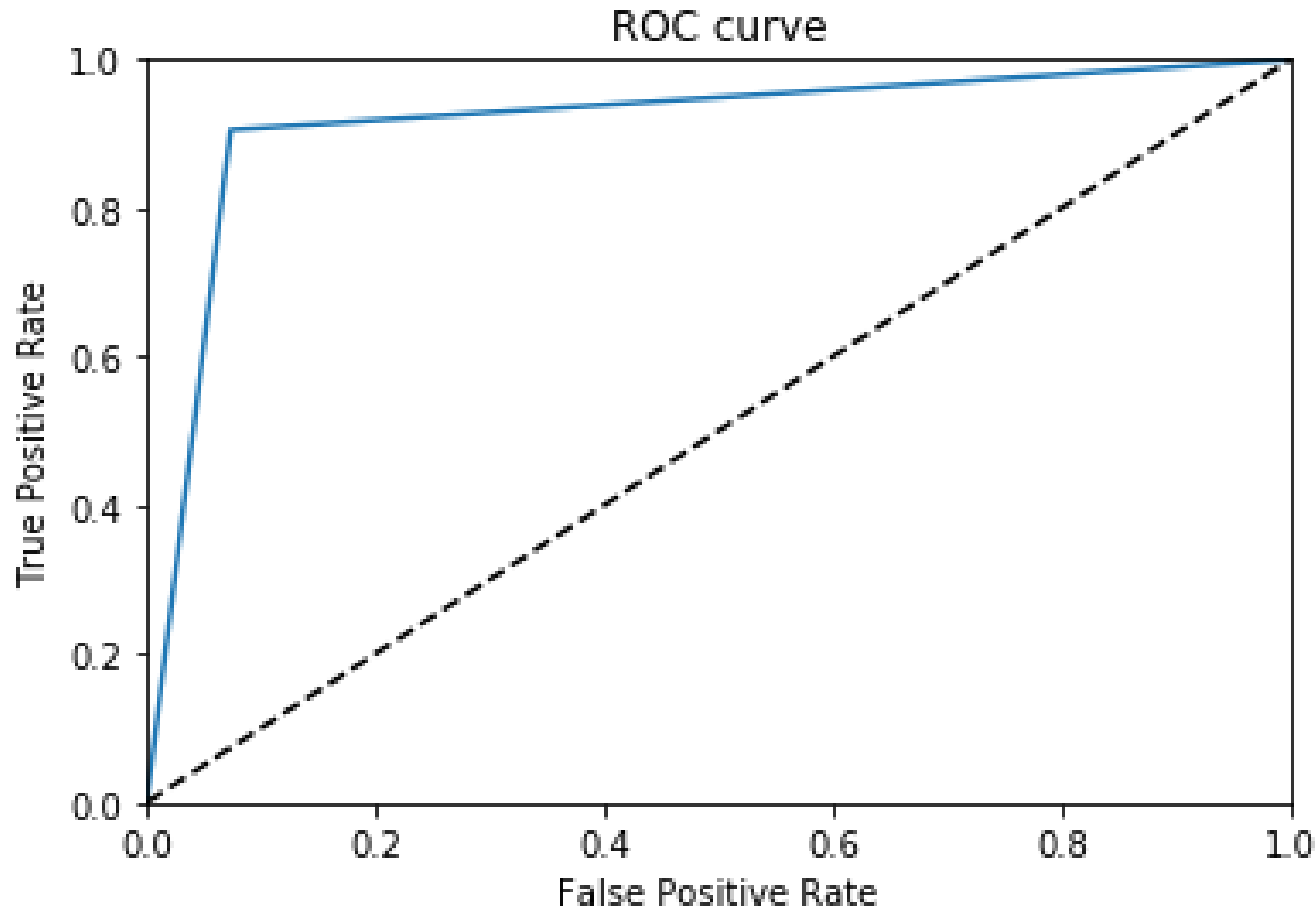


# ML algorithms performances

Model	Precision	Recall	F1-score	Accuracy
SVM*	0.89	0.93	0.90	0.90
Decision Tree	0.88	0.91	0.89	0.90
Random Forest*	0.92	0.91	0.91	0.92

*\*Grid Search applied for hyperparameter optimization*

# ROC Curve Random Forest

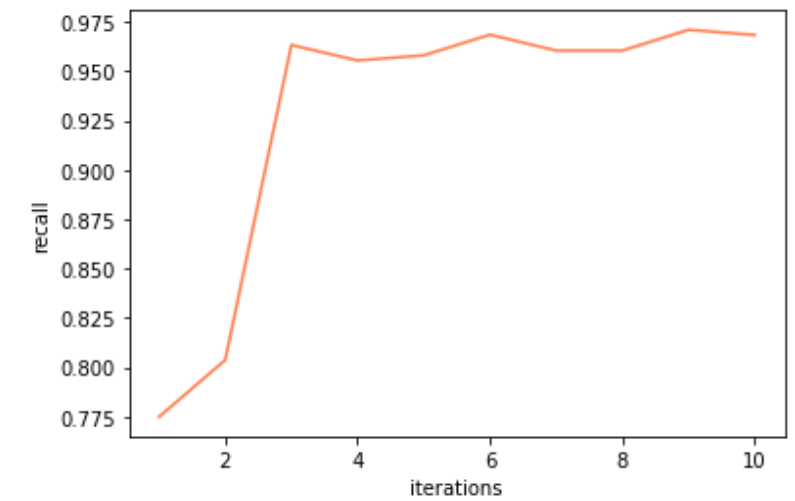
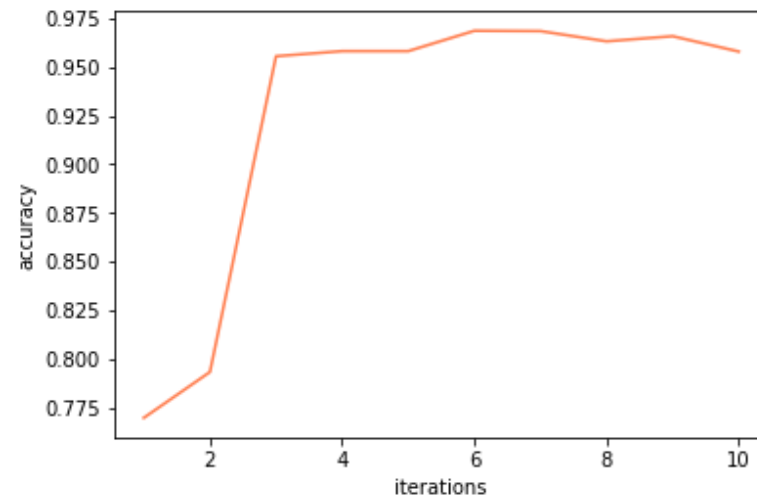
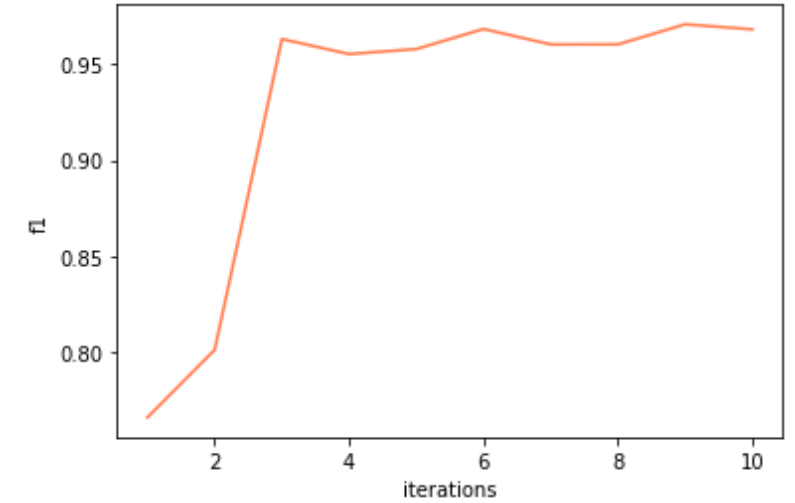
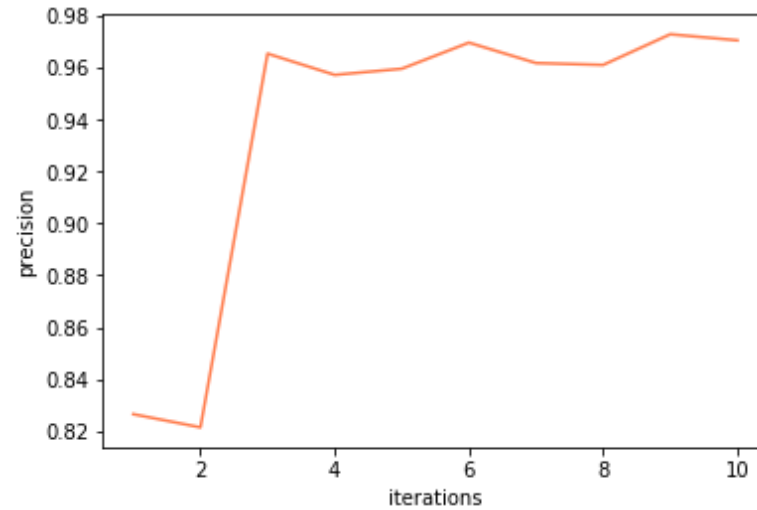


Area Under the Curve  
(AUC) = 0.91

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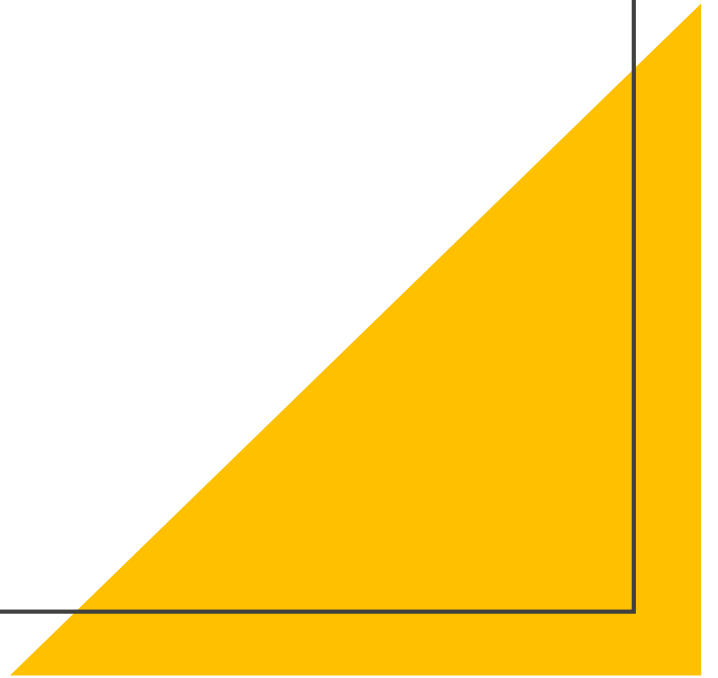
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# 10 – fold Cross- validation

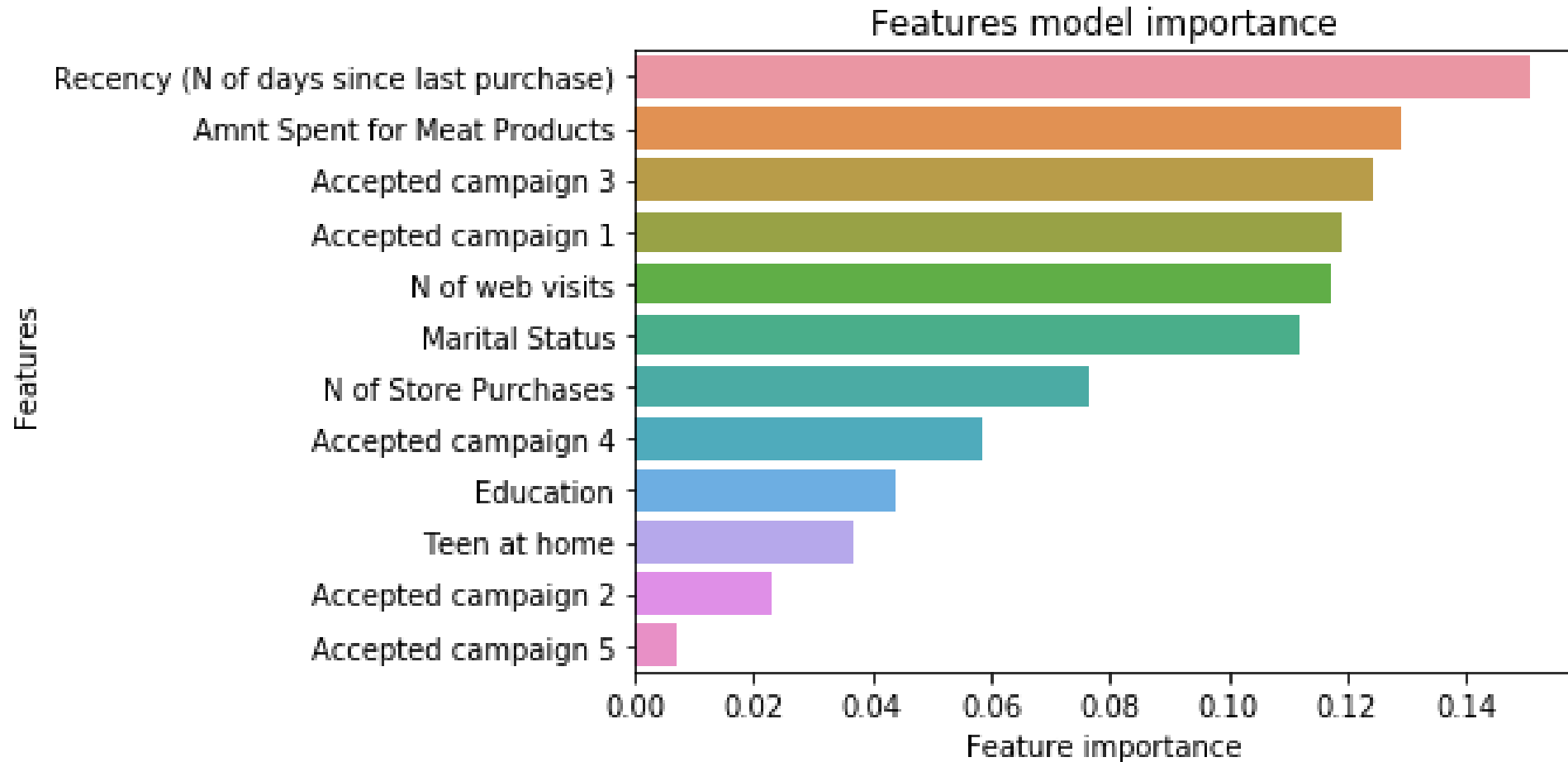


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# Features importance



# Conclusions and future steps

- ❖ The models reached good predictive performance metrics (about 90 % accuracy);
- ❖ Random Forest was the best performing algorithm;
- ❖ Random Forest can be tested for model deployment predicting consumer behavior;
- ❖ The number of days since last purchase, the amount spent for meat products, and behaviour in previous campaigns are the most substantial predictors of consumer response to the current marketing online campaign.