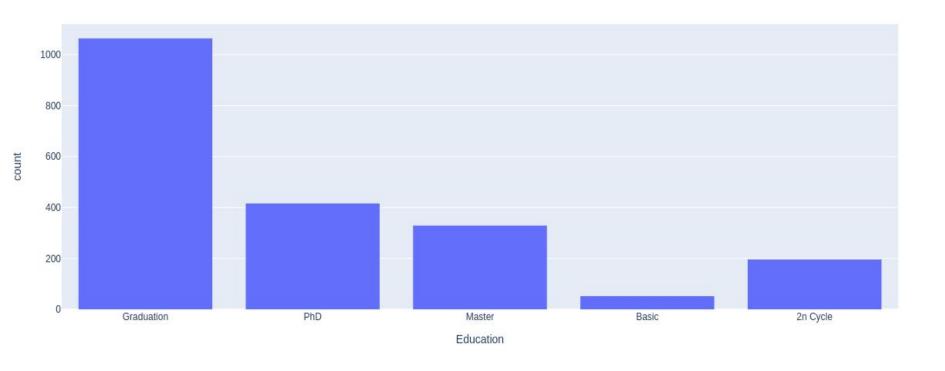
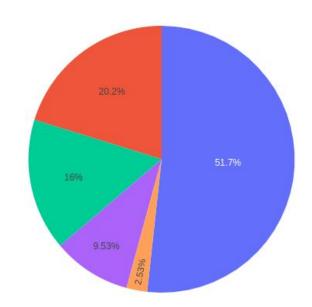
Data Advanced Analytics Test

Total customers by graduation



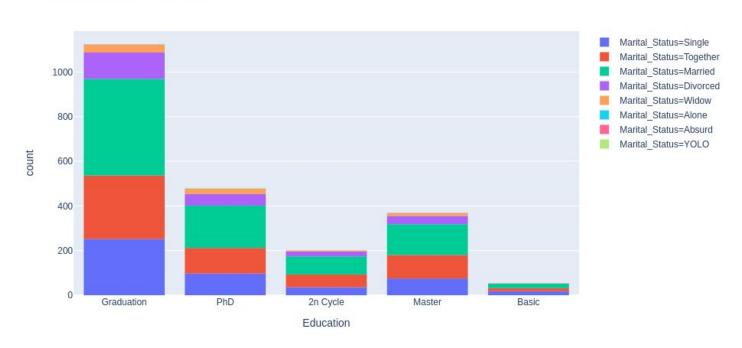
Total customers% by graduation



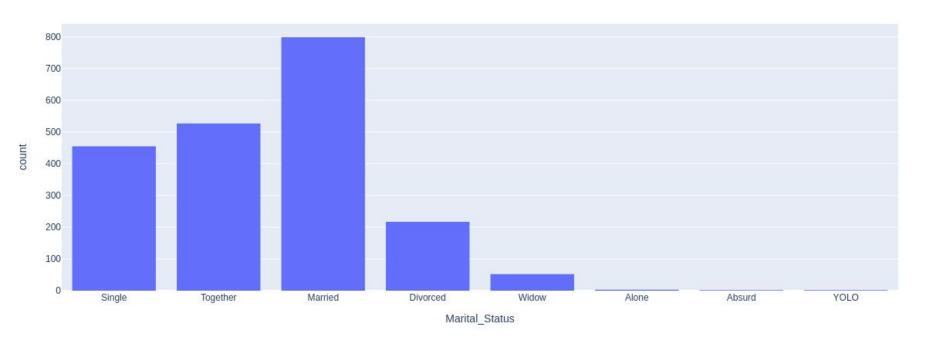
Graduation

Master 2n Cycle Basic

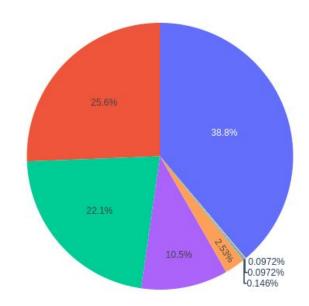
Education per Marital Status

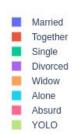


Marital Status

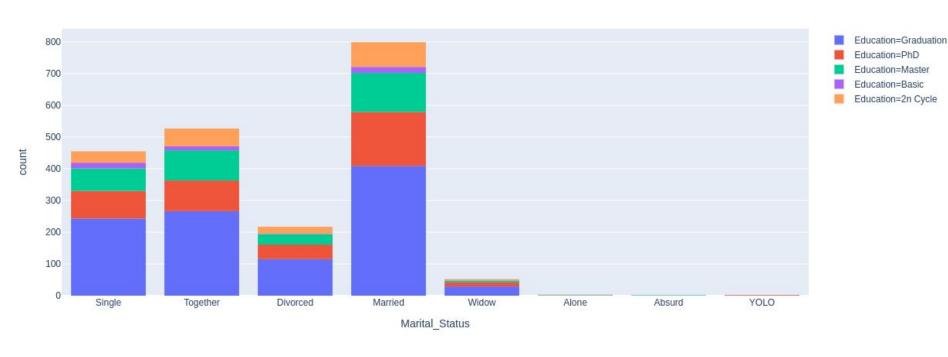


Marital Status%

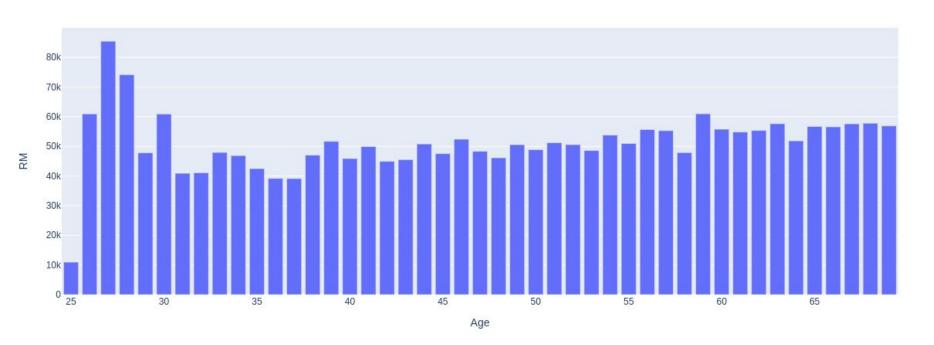




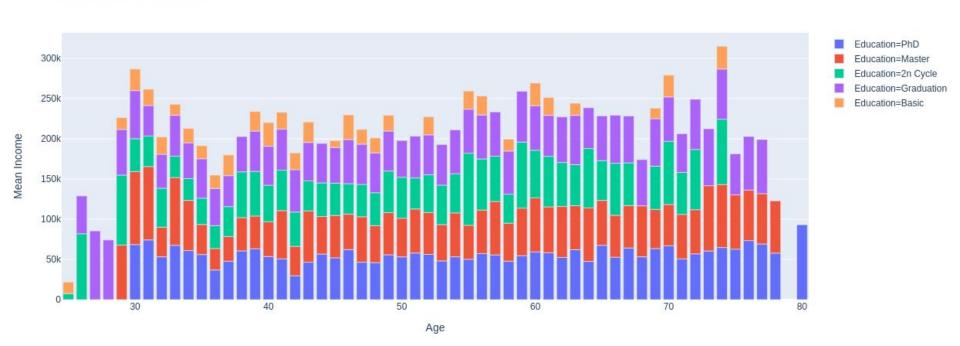
Marital Status per Education

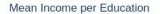


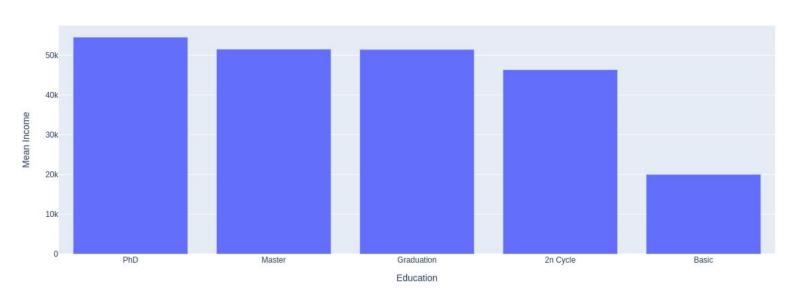
Mean Income per Age



Mean Income per Education

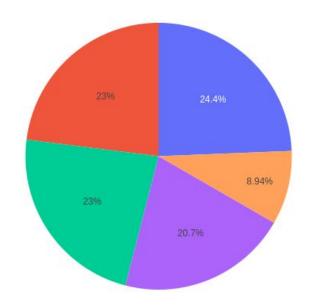






The average income does not have much variation, only the basic one (2.5% of the dataset)

Mean Income per Education%



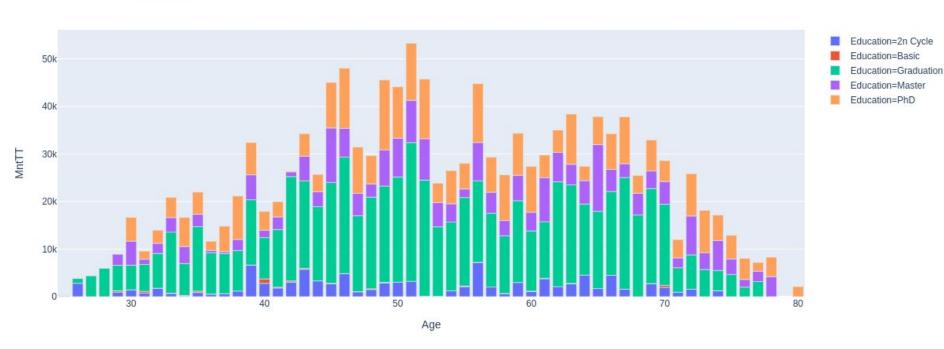
PhD

Graduation 2n Cycle Basic

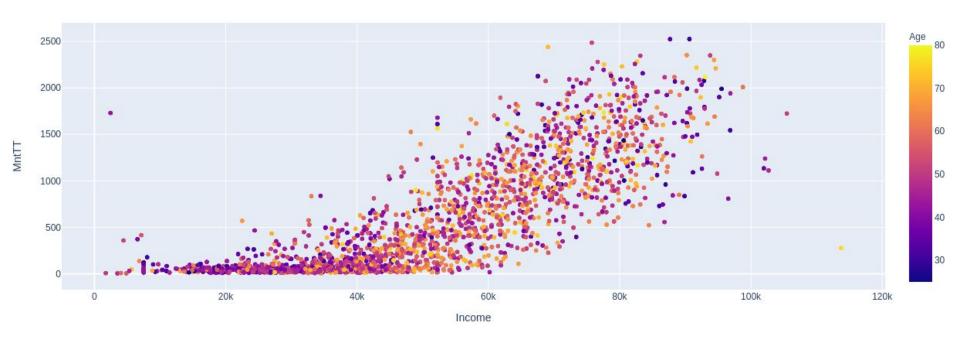
Income per Education



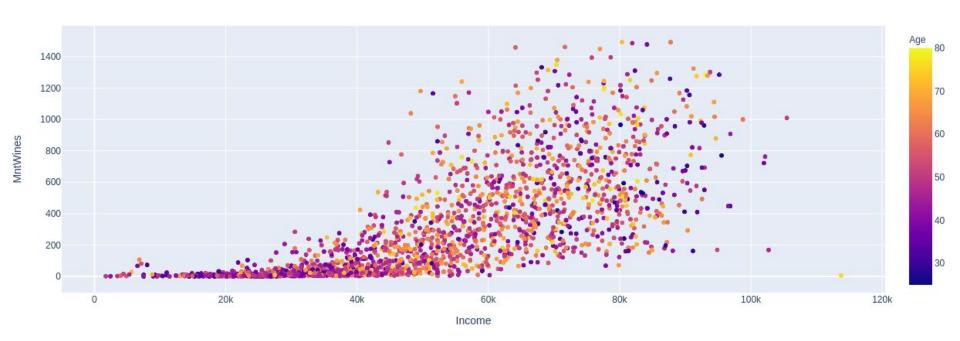
Total Spent per Education



Total Spent per Income



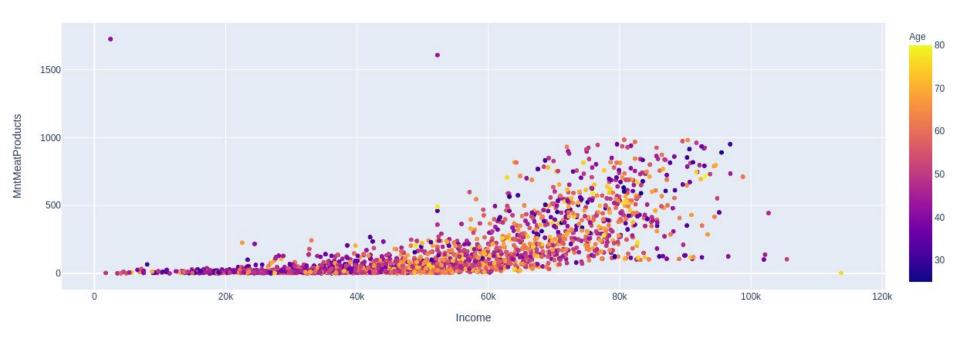
Total Spent with Wine per Income



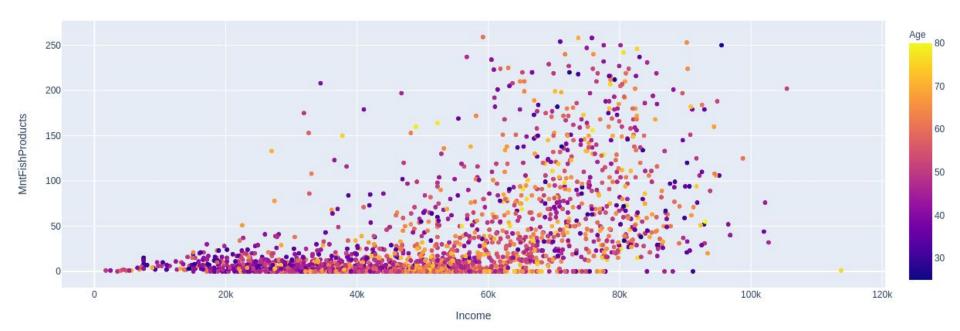
Total Spent with Fruits per Income



Total Spent with Meat Products per Income



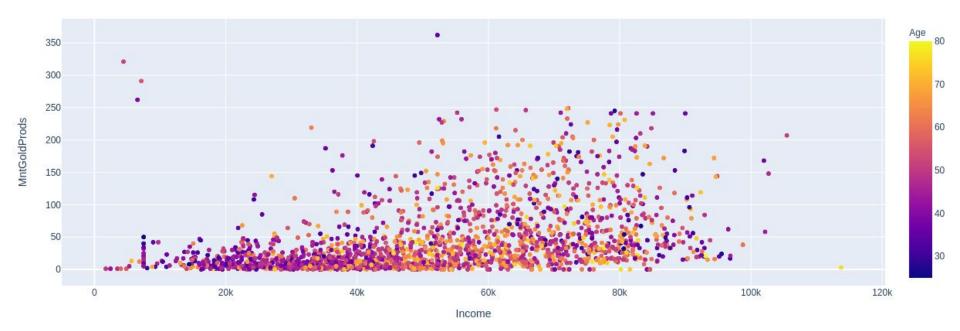
Total Spent with Fish Products



Total Spent with Sweet Products per Income



Total Spent with Gold Products per Income



Pearson Correlation of Income per Category of Products

MntWines

Income

MntFruits



MntMeatProducts

MntFishProducts

MntSweetProducts

MntGoldProds

MntTT

- 0.9

- 0.8

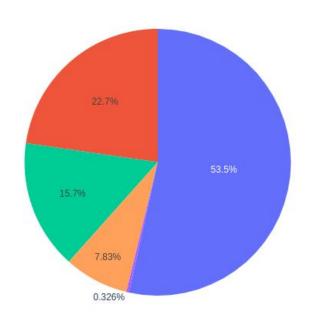
- 0.7

- 0.6

- 0.5

- 0.4

Total Spend% per Education



Graduation

Master 2n Cycle Basic

Total Spend per Education



Total Spend with Wine per Education



Total Spend with Fruits per Education



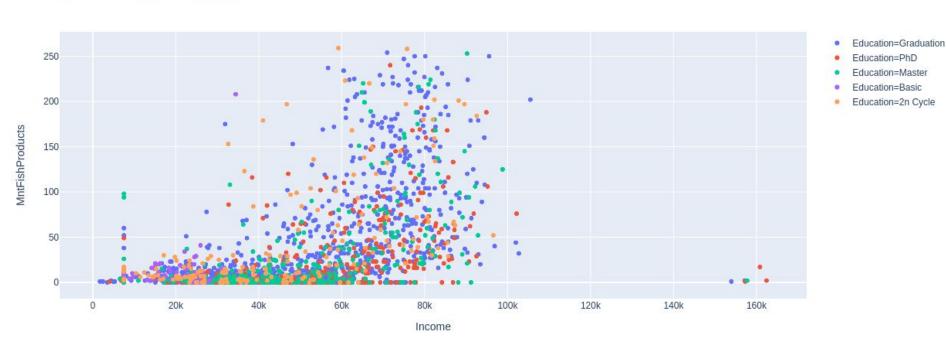
Total Spend with Meat per Education



Education=Graduation Education=PhD Education=Master

Education=Basic Education=2n Cycle

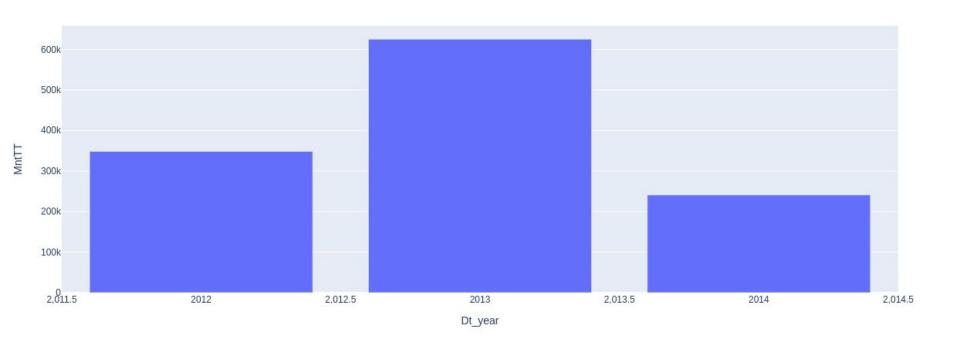
Total Spend with Fish per Education



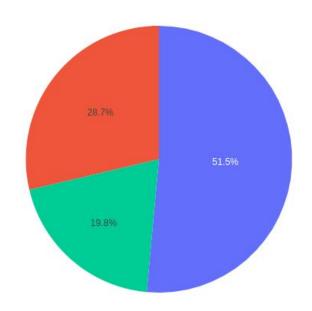
Total Spend with Sweet Products per Education



Total Spend per Year

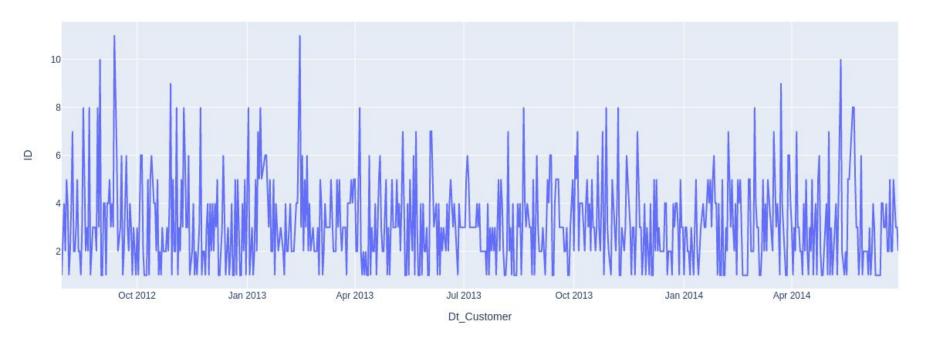


Total Spend% per Year

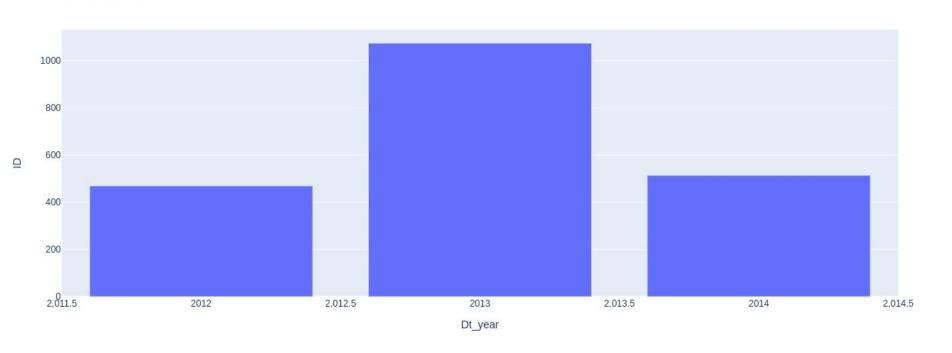




Total of Clients per date

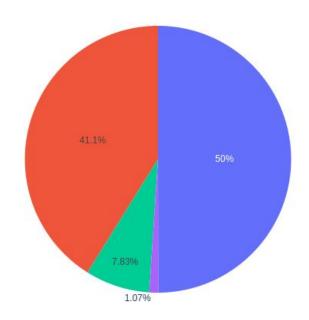


Total of Customers per Year



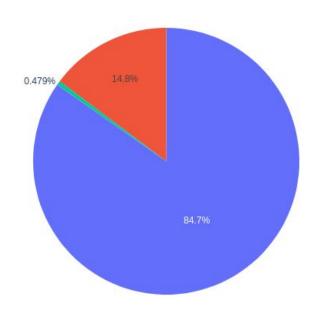
Dependents

Total Spend of Customers with Dependents%



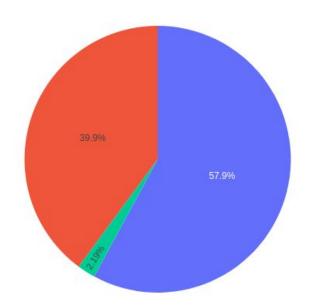
Dependents

Total Spend of Customers with Kids%



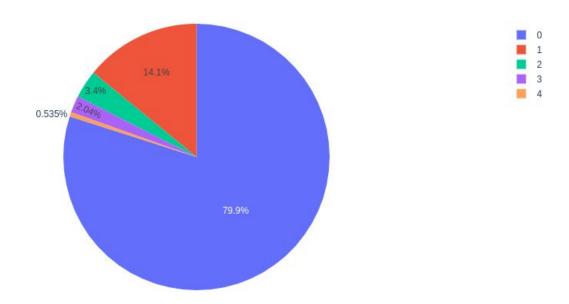
Dependents

Total Spend of Customers with Teens%



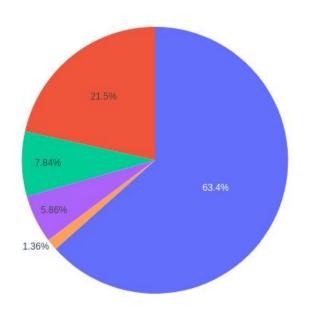
Campaigns

Total of Customers per Total of Campaigns Accepted



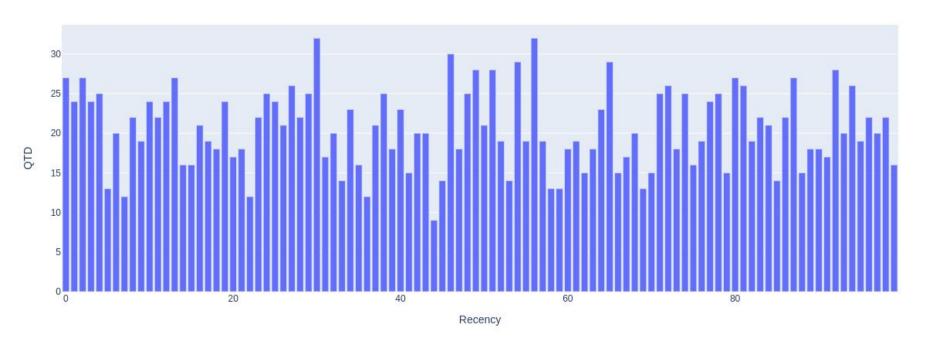
Campaigns

Total Spend per Number of Campaigns Accepted



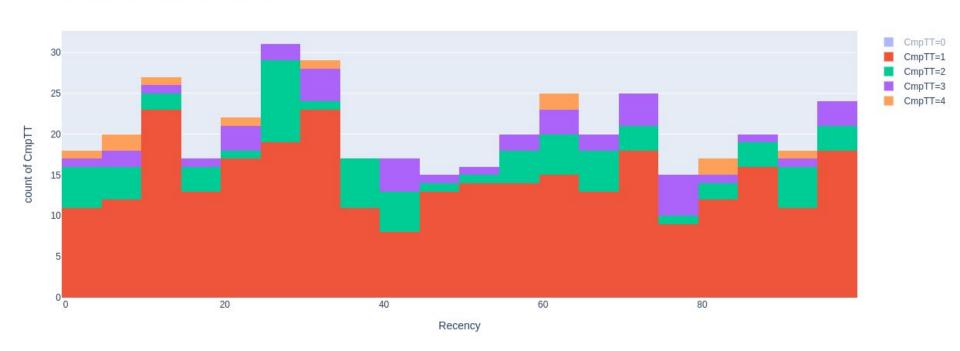
Recency

Total of Customers per Recency

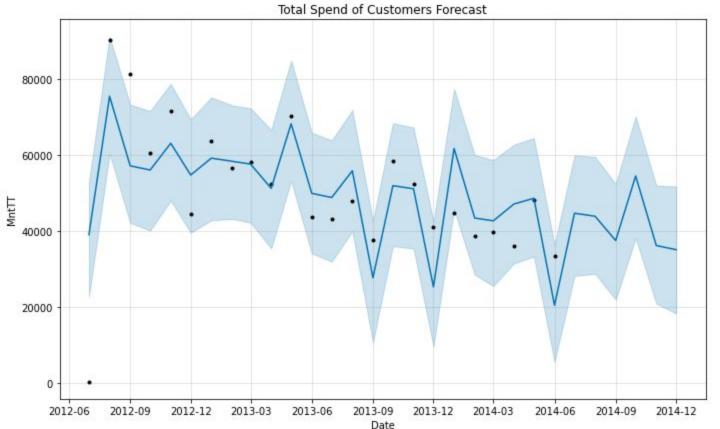


Recency

Total of Campaigns Accepted per Recency



Time Series Forecast



Six months forecast

	ds	yhat
24	2014-07-01	44685.808586
25	2014-08-01	43874.656448
26	2014-09-01	37494.388518
27	2014-10-01	54475.231201
28	2014-11-01	36183.497223
29	2014-12-01	35083.189117

Classification

3 models were chosen for training, testing and validation: Random Forest Classifier: It creates decision trees at random, where each tree will be used in the choice of the final result.

Logistic Regression: aims to produce, from a set of observations, a model that allows the prediction (probability) of values taken by a categorical variable.

XGBClassifier: It is an algorithm based on Decision Trees (Gradient Boosting). means that the algorithm uses the Gradient Descent algorithm to minimize the loss (loss) while new models are being added

Classification

The models had the following results:

Metric	Random Forest	Logistic Regression	XGBClassifier
Recall	73,33%	61,66%	36,66%
Precision	37,60%	25,69%	62,85%
Accuracy	80,04%	70,85%	88,56%

Linkedin

Presentation by Alysson Guimarães

Contact me

Github