

## **AGENDA:**

- Introduction: About me
- Retail: The Big picture
- Context and Problem Statement
- Objectives
- Process and Approach
- Analysis and Recommendations
- Machine Learning Models and Evaluations
- Adaptability and Future Scope

## **RETAIL: THE BIG PICTURE**



### Context/Problem statement: Marketing team needs to:

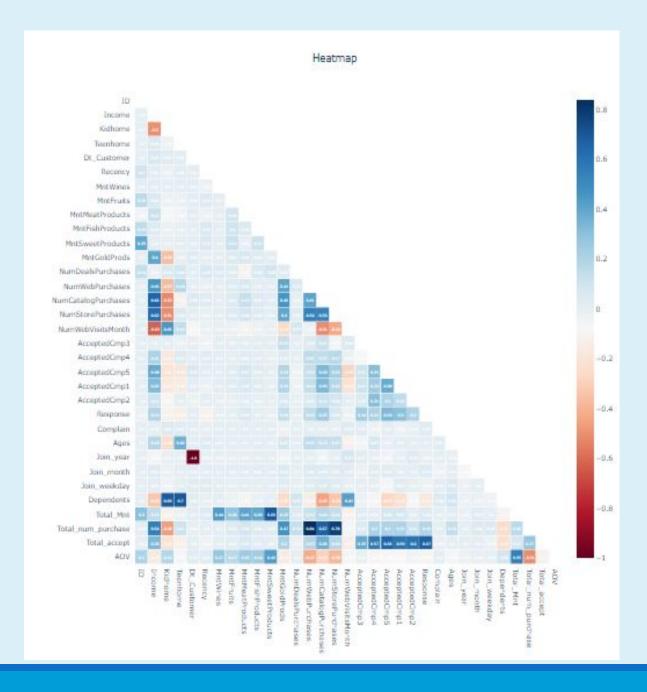
- ✓ Understand the problems or challenges with their earlier campaigns.
- ✔ Predict the customer response to upcoming campaigns.
- Ensure the effectiveness of the campaign.
- Increase the number of new customers/retain existing customers.

### **Objective:**

- Build the machine learning model to predict the customer response to the campaigns based on customer demographics and last purchases.
- ✓ Increase the responses through personalized promotions by using this model and boost campaign efficacy.

**About Data:** The data set of 39983 customers of the company with data on:

- Customer profiles
- Product preferences
- ✓ Campaign successes/failures
- Channel performance



### PROCESS AND APPROACH

#### **Model Performance:**

Evaluating the performance of model on training and test dataset.



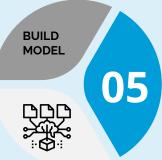


## **Understanding Requirement:**

Understand the problem. Gather the data.

#### **Building a Productive model:**

Algorithm Selection from Knn classifier, Random Forest classifier, Boosting. Split the dataset and Train the model in training and test data set.



DATA SCIENCE

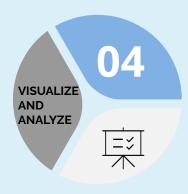


### **Exploratory Data Analysis:**

Analysing data sets to summarize their main characteristics, with visual methods.

## Data Visualization and Further Statistical Analysis:

Feature Identification, analysing the correlation between different variables. Preprocessing the data.





**Data Cleaning:** Remove Irrelevant Values, Filling the Missing Values.

### **ANALYSIS AND RECOMMENDATIONS**

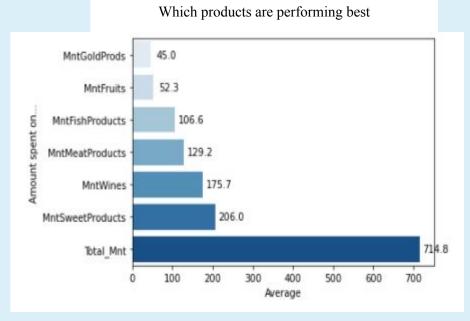
## **Exploratory Data Analysis for the last campaign:**

- Performed nearly twice better from the previous campaigns.
- Attracted more customers on various factors.
- The customers purchased more evenly through catalogs, websites and stores.
- The customers earned 25% more income than the customers in the previous campaigns.

### **Data-Driven Recommendations:**

- Use the same marketing techniques as in the last campaign.
- Focus on promoting certain products.
- Have marketing campaigns to convert customers who shop online only to in-store purchases.
- People who spent an above average amount on certain products would have more in store purchases, so there should be focussed in store campaigns for them.





## **Machine Learning Models and Evaluations**

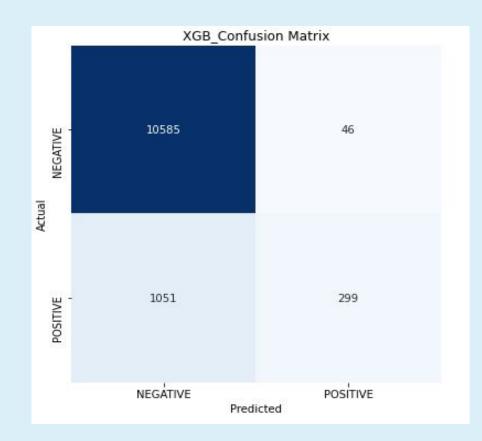
	method	mean score	std score	precision score	f1 score
0	Logistic Regression	0.000000	0.000000	0.000000	0.000000
1	Decision Tree Classifier	0.956315	0.013380	0.968727	0.966209
2	KNN Classifier	0.972901	0.006715	0.984906	0.975701
3	Random Forest Classifier	0.982673	0.007802	0.986228	0.970267
0	Ada Boost Classifier	0.972921	0.008368	0.981777	0.969629
1	Gradient Boost Classifier	0.876815	0.026270	0.899749	0.410520
2	XGB Classifier	0.884991	0.020243	0.866667	0.410520



Comparison of different algorithms used in the machine learning models.

# Best F1 score(97.57%) is of KNN Classifier model.

	precision	recall	f1-score	support
NEGATIVE	1.00	1.00	1.00	10631
POSITIVE	0.98	0.97	0.98	1350
accuracy			0.99	11981
macro avg	0.99	0.98	0.99	11981
weighted avg	0.99	0.99	0.99	11981



	precision	recall	f1-score	support
NEGATIVE	0.91	1.00	0.95	10631
POSITIVE	0.87	0.22	0.35	1350
accuracy			0.91	11981
macro avg	0.89	0.61	0.65	11981
veighted avg	0.90	0.91	0.88	11981

## Adaptability of this model for other domains:

Banking - Mortgage approval predictions Healthcare - Disease prediction Real estate - Pricing predictions.

## **Future scope:**

Sales forecasting More customized campaigns for the customers. Promotional event based campaigns.

