



## *Sales Analysis*

By RSA Analyzers

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>Sampada Shyam

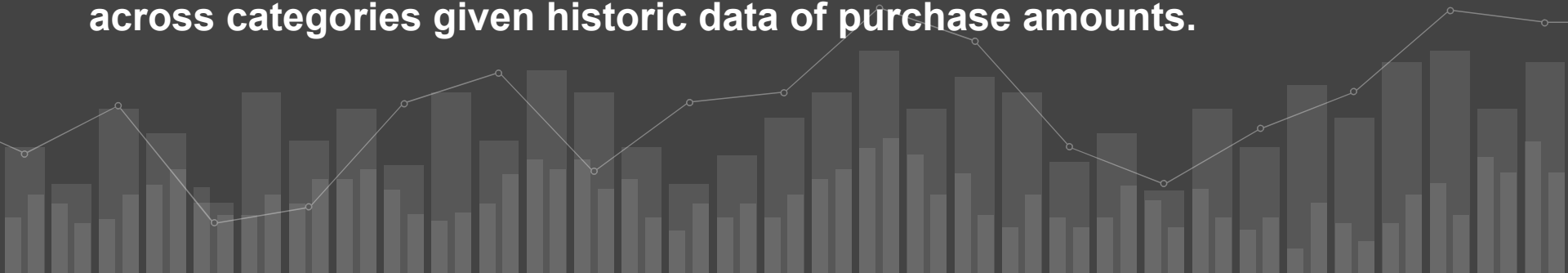
>Rohit Raj Koul

# Motivation

To help the market to focus on the factors that can boost up their sales during Black Friday.

## Problem Statement

The challenge is to predict the purchase of various products by users across categories given historic data of purchase amounts.



# Background and Dataset

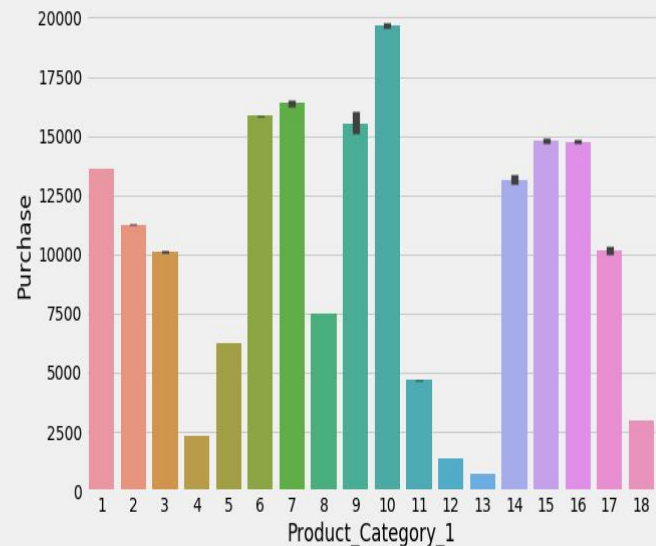
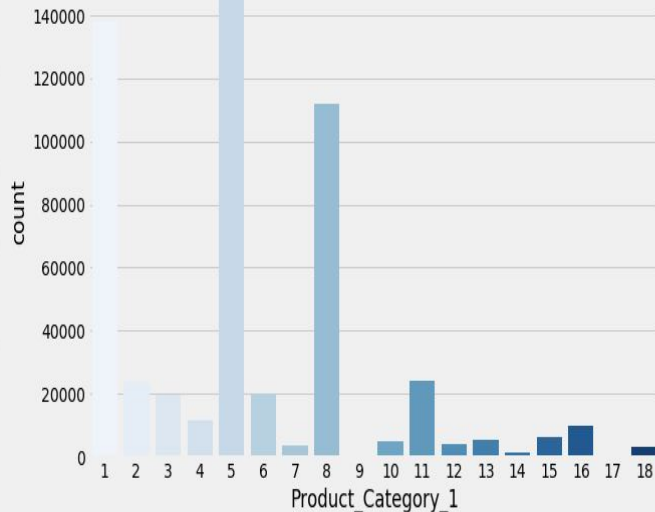
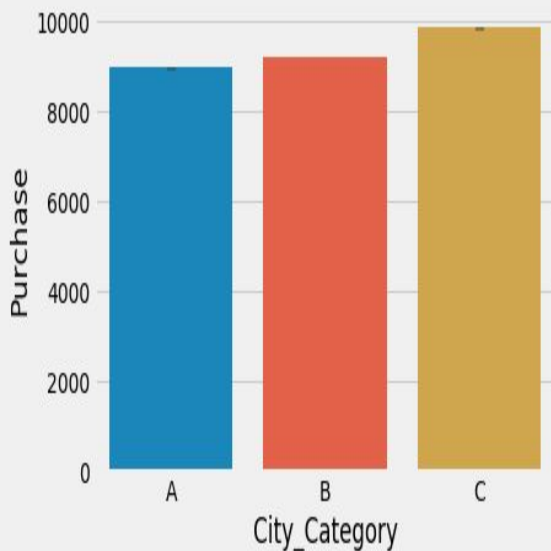
- Black Friday is the most important day for all the companies to maximize their sales and it would be very helpful to predict the sales beforehand.
- Dataset is retrieved from Kaggle.

|        | User_ID | Product_ID | Gender | Age   | Occupation | City_Category | Stay_In_Current_City_Years | Marital_Status | Product_Category_1 | Product_Category_2 | Product_Category_3 | Purchase |
|--------|---------|------------|--------|-------|------------|---------------|----------------------------|----------------|--------------------|--------------------|--------------------|----------|
| 0      | 1000001 | P00069042  | F      | 0-17  | 10         | A             | 2                          | 0              | 3                  | NaN                | NaN                | 8370     |
| 1      | 1000001 | P00248942  | F      | 0-17  | 10         | A             | 2                          | 0              | 1                  | 6.0                | 14.0               | 15200    |
| 2      | 1000001 | P00087842  | F      | 0-17  | 10         | A             | 2                          | 0              | 12                 | NaN                | NaN                | 1422     |
| 3      | 1000001 | P00085442  | F      | 0-17  | 10         | A             | 2                          | 0              | 12                 | 14.0               | NaN                | 1057     |
| 4      | 1000002 | P00285442  | M      | 55+   | 16         | C             | 4+                         | 0              | 8                  | NaN                | NaN                | 7969     |
| ...    | ...     | ...        | ...    | ...   | ...        | ...           | ...                        | ...            | ...                | ...                | ...                | ...      |
| 537572 | 1004737 | P00193542  | M      | 36-45 | 16         | C             | 1                          | 0              | 1                  | 2.0                | NaN                | 11664    |
| 537573 | 1004737 | P00111142  | M      | 36-45 | 16         | C             | 1                          | 0              | 1                  | 15.0               | 16.0               | 19196    |
| 537574 | 1004737 | P00345942  | M      | 36-45 | 16         | C             | 1                          | 0              | 8                  | 15.0               | NaN                | 8043     |
| 537575 | 1004737 | P00285842  | M      | 36-45 | 16         | C             | 1                          | 0              | 5                  | NaN                | NaN                | 7172     |
| 537576 | 1004737 | P00118242  | M      | 36-45 | 16         | C             | 1                          | 0              | 5                  | 8.0                | NaN                | 6875     |

537577 rows x 12 columns

# Exploratory Data Analysis

- Used at an early stage in an Analytical process.
- Explains specific sorts of initial analysis and findings done with data sets.

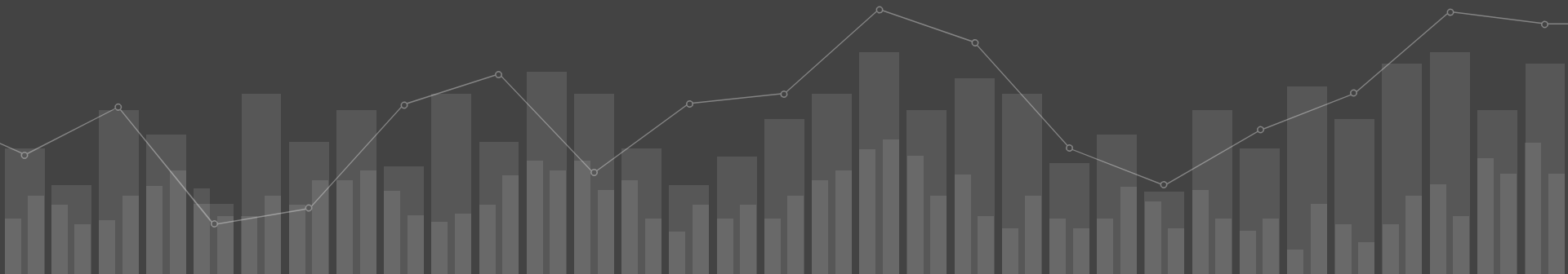


# Data Preprocessing

- Feature Selection
- Handling Missing Values
- Handling Categorical Values
- Splitting Data

Training data = 80%

Testing data = 20%

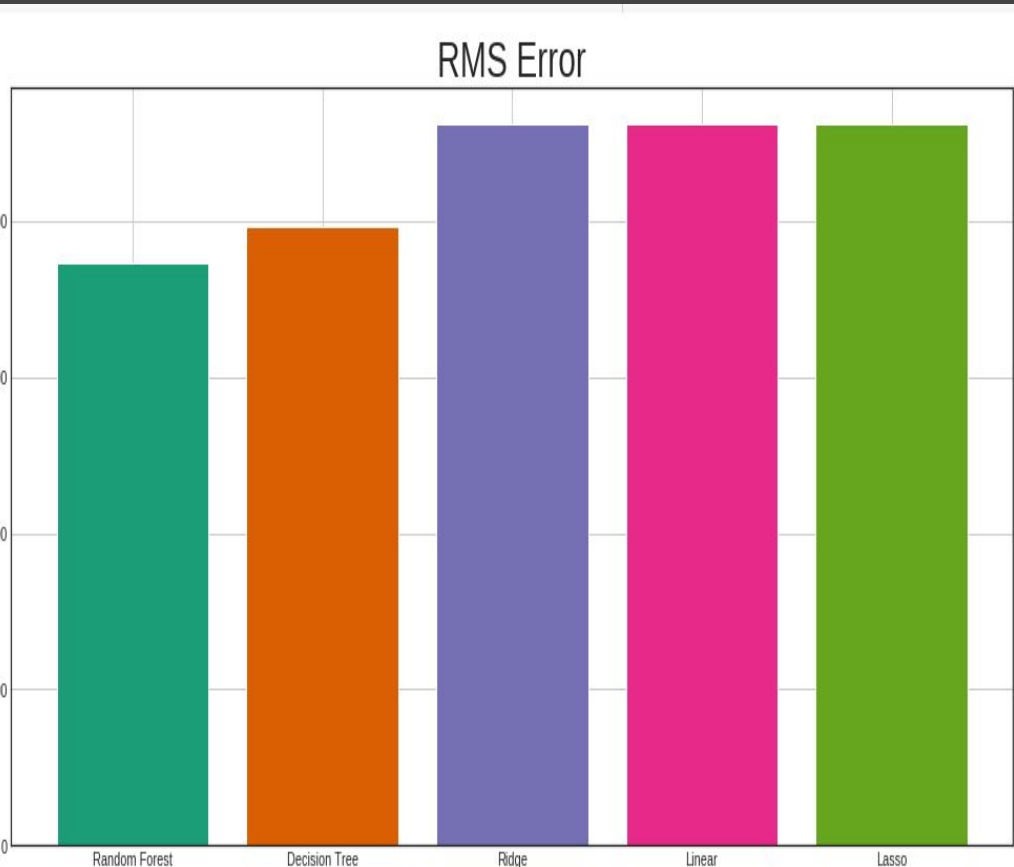


# Regression Model Implementation

- Linear Regression
- Ridge Regression
- Random Forest Regression
- Decision Tree Regression
- Lasso Regression

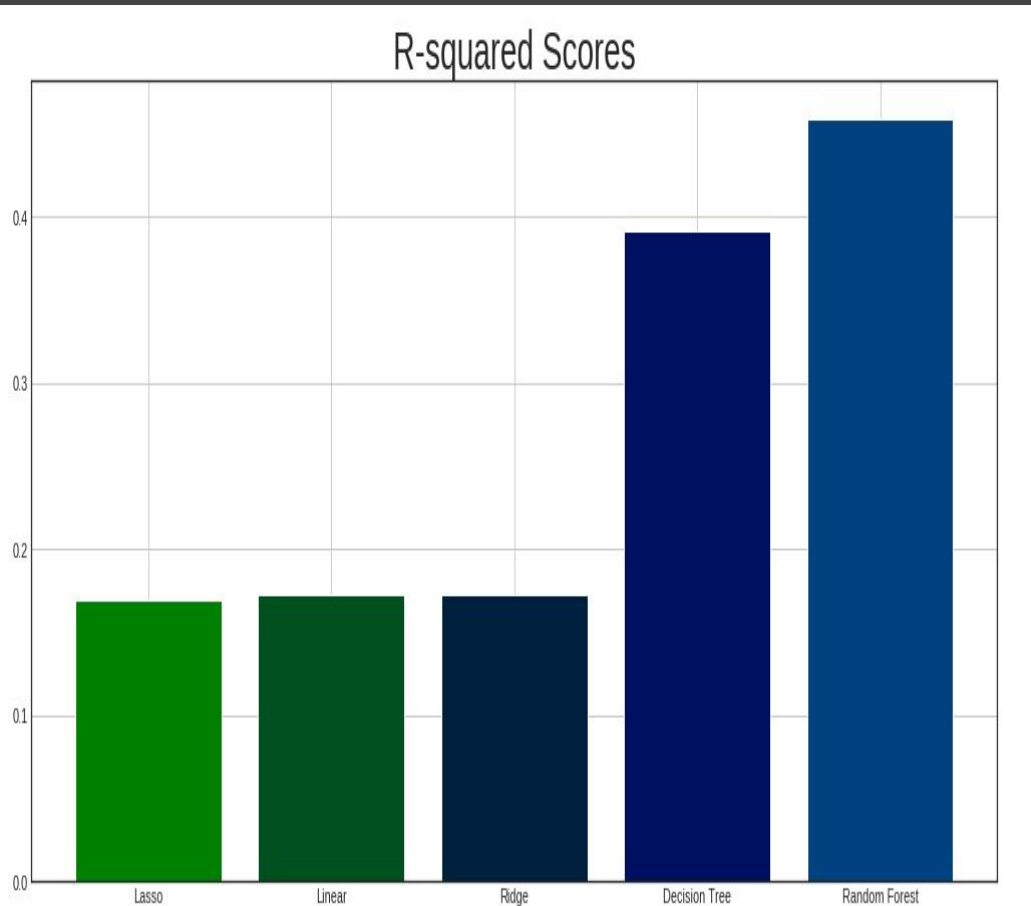
| Regression Models | RMSE     | R-squared scores |
|-------------------|----------|------------------|
| Linear            | 4619.078 | 0.1722           |
| Ridge             | 4619.078 | 0.1722           |
| Random Forest     | 3729.843 | 0.4602           |
| Decision Tree     | 3960.993 | 0.3912           |
| Lasso             | 4627.025 | 0.1693           |

# Comparative Analysis of RMS Errors



- Root Mean Square Error (RMSE) is the standard deviation of the residuals (gauges).
- Residuals are an extent of how far from the backside line data centers are; RMSE is an extent of how spread out these residuals are.
- Root mean square error is consistently used in climatology, deciding, and regression analysis to check preliminary outcomes.
- Comparing the root mean square error of the five algorithms, Random Forest is found to have less root mean square error.

# Comparative Analysis of R-squared scores



- The coefficient of assurance, indicated  $R^2$  or  $r^2$  and articulated "R squared", is the extent of the change in the reliant variable that is unsurprising from the free variable(s).
- It gives a proportion of how all around watched results are repeated by the model, in light of the extent of all out variety of results clarified by the model.
- R-squared score for random forest is observed to be the highest i.e 46% hence providing less RMS errors.
- Decision tree is observed to be the second highest with R-squared score of 39%.
- Linear and ridge regression models are having identical R-squared score of 17% which are the least of all.



# Challenges

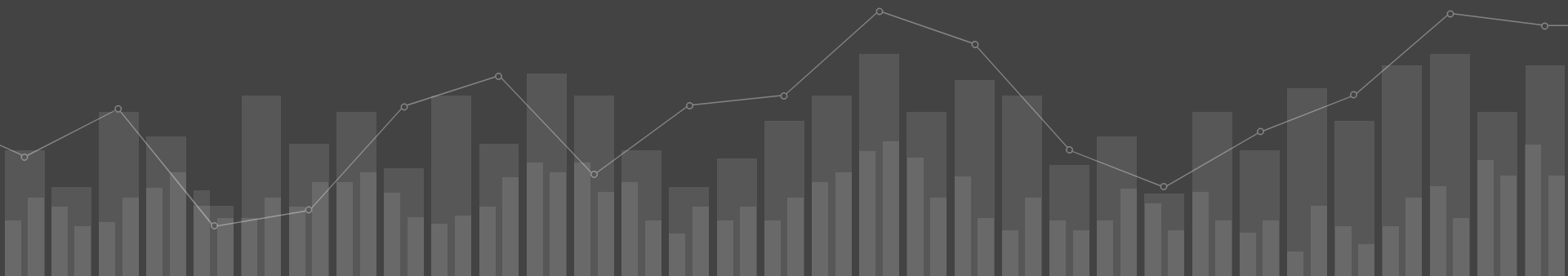
- **Training Large Data Set**
- **Handling Missing Values**
- **Less Number of Features for Analysis**

## Future Work

- **Implementing Logistic Regression, and Classification models**
  - **Recommender System**
  - **Adding more features like states and taxes**
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# References

1. <https://www.google.com/url?q=https://github.com/mathubhalan/Black-Friday-Sales/blob/master/Data/BlackFriday.csv>



# Individual Contribution:

## 1. Sampada Shyam

- Data pre-processing
- Linear regression model implementation

## 2. Aishwarya Bhonde

- Exploratory data analysis
- Ridge regression and random forest model implementation

## 3. Rohit Raj Koul

- Exploratory Data Analysis
- Lasso regression and decision tree classifier model implementation



Thankyou..

Questions?

