# Amazon Sales Data Analysis Project



# **Objective**

- Development of a predictive model for predicting sales.
- Perform ETL (Extract-Transform-Load) on dataset.
- Develop dashboard by using tableau.

# **Benefits**

- Better understand and optimise revenue generation in future
- Maximize forecasting accuracy
- Make current sales experience our top priority

# Architechture

Project is done with 4 steps



## **Data Collection**

The Dataset provided by ineuron. Collected from their website in csy fromat.





# **Data Analysis**

This steps is a combination of Exploratory data analysis,
Statics and Label Encoding





# **Business Analysis**

This Step made with Machine Learning algorithms. Used Pycaret library and Data visualization.





## **Business Intelligence**

Used Tableu for creating
Dashboard and combined Data
Analytics with Business KPI's.



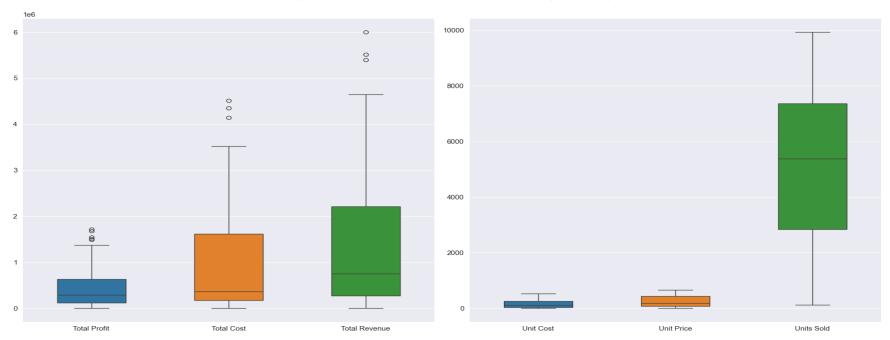


# **Data Preprocessing**

- Importing necessary libraries for data analysis such as: Pandas, Numpy, Matplotlib & Seaborn etc.
- Using pd.read\_csv() function stores the data in pandas dataframe named data.
- Using data.column showing columns present in dataframe.
- info() function show basic information of dataframe like null value count of each column and their data type
- Changing the data type of different column for model training and analysis
- Using describe function on dataframe for getting basic stats of numerical dataset
- Adding extra column to dataframe which contain only month, year and month with year
- Using isnull().sum() checking out total null value in all the column of dataframe

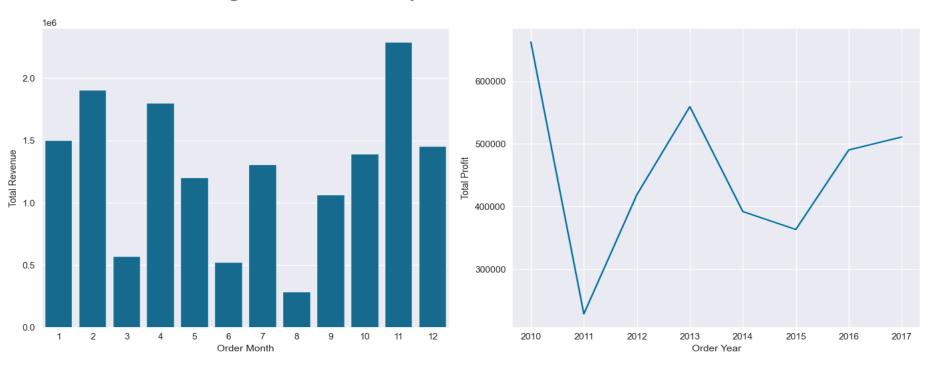
# **Exploratory Data Analysis**

**Checking Outliers in the dataframe by using Box Plot** 



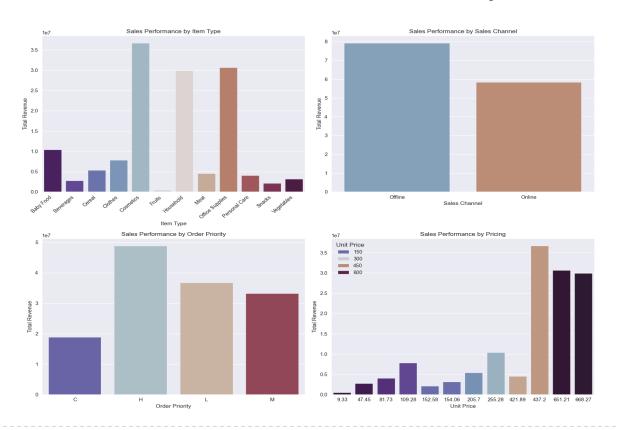
There are some outliers in Total Profit, Total Cost and Total Revenue colums. No outliers in Unit Cost, Unit Price and Unit Sold Columns.

## Understanding the Total Revenue per Month with bar chart and a line chart for trend



The total revenue shows variations from year to year. For instance, there was a significant increase in revenue from 2011 to 2013, followed by a decline in 2015. Each month's revenue varies across different years, indicating potential seasonal trends or fluctuations in consumer behavior. August has lowest rrevenue and November has highest revenue.

## **Factor Analysis**



- The total revenue and units sold vary significantly across different regions and countries.
- The item types 'Cosmetics', 'Household', and 'Office Supplies' contribute significantly to total revenue.
- Offline sales generate higher total revenue and units sold compared to online sales.
- Orders with order priority
   'H' (high) contribute the most to total revenue and units sold.
- There is a variation in total revenue and units sold based on the unit cost of products.



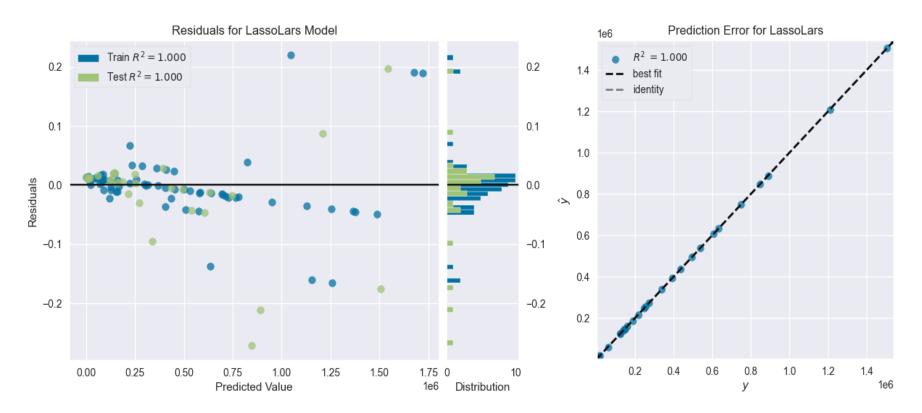
# **Predictive Analysis**

- ILabel Encoding of Item Type, Sales Channel and Order Priority for modeltraining.
- Dropping columns Region, Country, Order Date MonthYear, Order ID and ShipDate.

# **Pycaret Library**

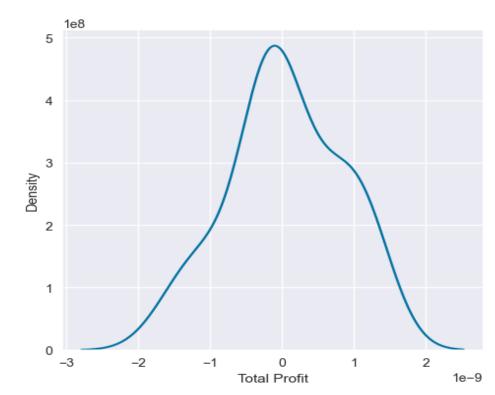
- PyCaret is an open-source, low-code machine learning library in Python.
- Allows users to quickly and easily build, compare, and deploy machine
- learning models on structured and tabular data.
- Reduce the amount of code needed to build a model.
- It provides preprocessing and feature engineering functions.
- Automatic model selection and hyperparameter tuning.
- Support for a wide range of machine learning algorithms

# Plotting residuals and prediction error for Lasso Least Angle Regression based trained model

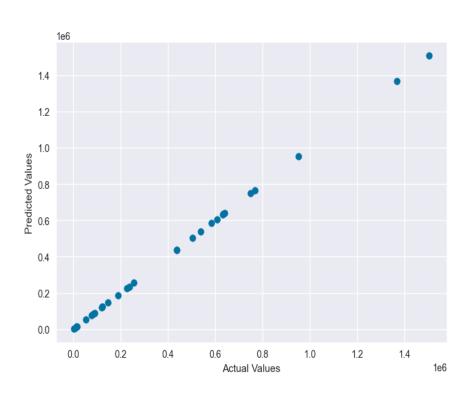


# Implementation of linear regression

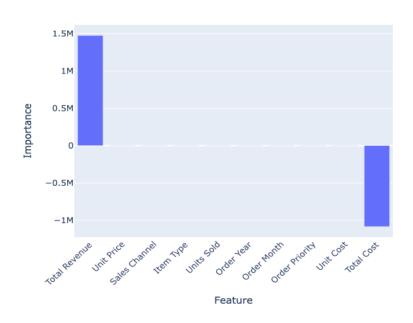
- Selecting the independent variables and target variable.
- Splitting the data into training and testing datasets.
- Standardizing the dataset.
- Performing fit transform on X\_train dataframe.
- Performing fit transform on X\_test dataframe.
- Applying Linear Regression on X\_train and y\_train.
- Calculating mean squared error.
- Creating kernel density estimate plot



# Plotting the predicted values against the actual values to visualize how well the model is fitting the data.



#### Feature Importance (R-squared = 1.00)



# Key Findings and Recommendation



### **Sales Performance**

Certain product categories, such as cosmetics and office supplies, generated higher total revenue and units sold compared to others.

#### **Recommendation:**

Invest in popular product categories to capitalize on existing demand. Consider expanding product offerings within high-performing categories to further boost sales.





## **Time Trends:**

Monthly and yearly sales trends indicated seasonal fluctuations and long-term growth patterns.

#### **Recommendation:**

Anticipate seasonal demand variations and adjust inventory levels and marketing efforts accordingly. Implement strategies to sustain long-term growth and capitalize on emerging market trends.





## **Relationship Analysis:**

Correlation analysis identified strong positive correlations between variables such as total revenue, units sold, and total profit.

#### **Recommendation:**

Continuously monitor and analyze key metrics to identify potential areas for improvement. Implement data-driven strategies to enhance sales performance and profitability.

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