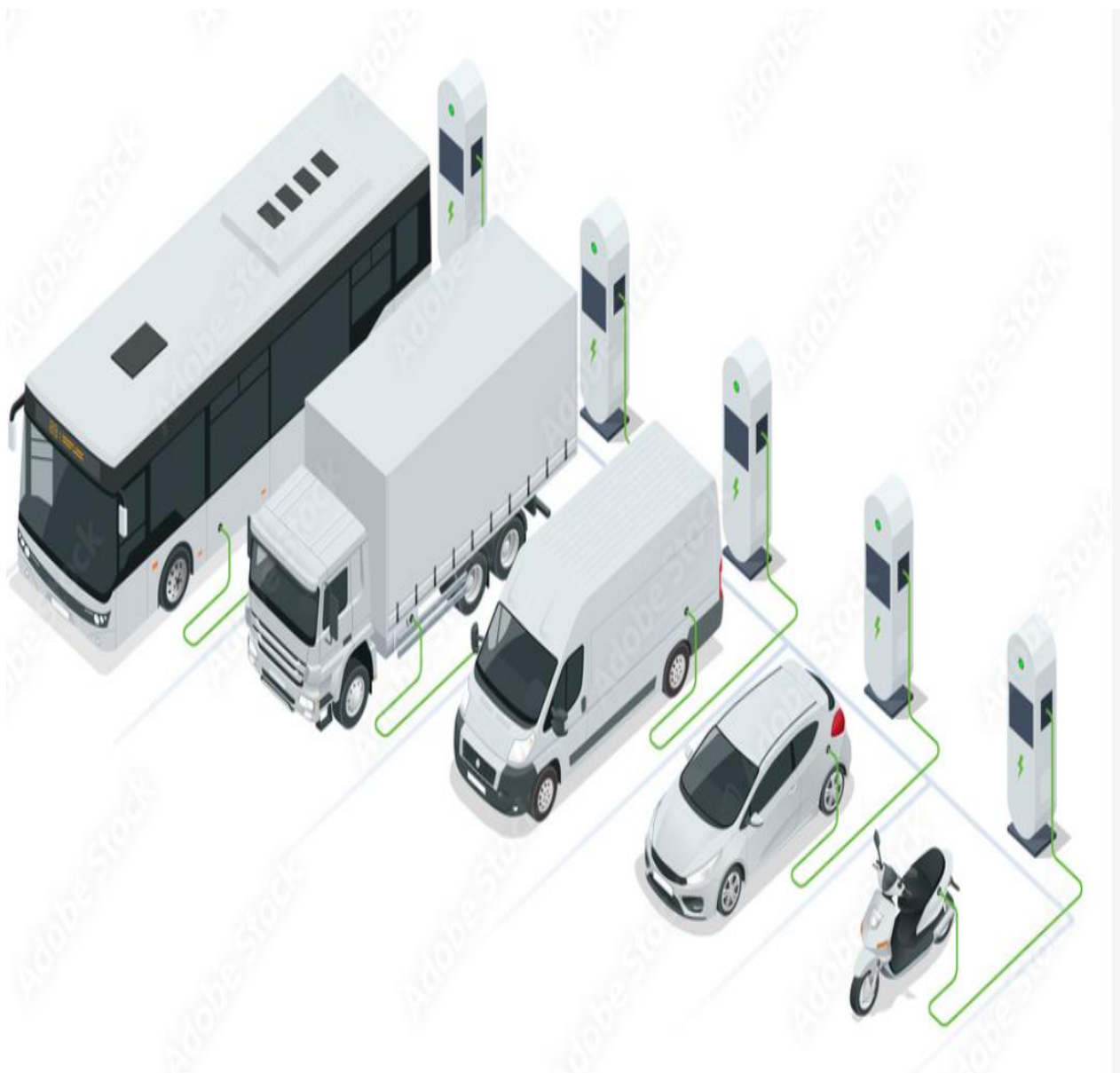


Feynn Labs
Internship Project 2

EV Market Segmentation Analysis for Indian Market



Contributor: Vinay K
Date: 30-08-2024

Introduction

The global electric vehicle (EV) market witnessed substantial growth in 2023, with sales reaching nearly 14 million, a 35% increase from the previous year. This growth expanded the global EV fleet to 40 million vehicles. Projections for 2024 suggest sales could reach 17 million, potentially accounting for 20% of total car sales. While China, Europe, and the USA lead the market, e-mobility is gradually spreading to other regions. Europe saw a notable increase in sales due to stricter CO2 standards and government incentives, despite a decline in Germany caused by subsidy reductions. The number of available EV models is also rising, from 590 in 2023 to a projected 1,000 by 2028.

Indian Context

India's EV market is rapidly expanding, valued at USD 8.03 billion in 2023 and expected to reach USD 117.78 billion by 2032, with a CAGR of 22.4%. The push towards EVs is driven by rising fuel prices, government incentives (such as FAME India), and the need to reduce dependence on imported crude oil. States like Uttar Pradesh, Karnataka, and Tamil Nadu are leading in EV adoption, particularly in two- and three-wheelers due to their affordability and suitability for India's traffic conditions. Major players like Tata Motors and Mahindra & Mahindra are also increasing their EV offerings, further boosting market growth.

Problem Statement

The primary objective of this project is to collect and analyze data related to vehicles, markets, customers, demographics, and more, including electric vehicles. The goal is to extract meaningful insights from these datasets and conduct segmentation of vehicles, markets, and customers. This will provide companies wishing to enter the Indian EV market with valuable insights, helping them decide on strategies such as vehicle segment, customer segment, technical aspects, and pricing.

Data Collection

Data has been collected from the following sources:

1. [Data Source 1](#)
2. [Data Source 2](#)

Finalized Datasets

After considering various parameters such as geography, cities, vehicle types, and customers, the following datasets were finalized for extracting meaningful insights:

1. Indian Automobile Buying Behavior Study 1.0
2. EV Stats-1
3. Indian-EV-Data

Dataset Columns Explanation

1. Indian Automobile Buying Behavior Study 1.0

- Age: Age of the customer purchasing a car
- Profession: Profession of the customer (Business/Employed)

- Marital Status: Married or Unmarried
- Education: Graduate or Postgraduate
- Number of Dependents: Number of dependents
- Personal Loan: Whether the person has an existing personal loan
- House Loan: Whether the person has a house loan
- Wife's Occupation: Whether the wife has an occupation
- Wife's Salary: Salary of the wife (if working)
- Salary: Salary of the customer
- Total Salary: Combined salary of the customer and wife
- Car Make: Make of the car purchased
- Car Price: Price of the car

2. EV Stats-1

- Serial Number: Unique identifier for records
- State: Indian state
- Two-Wheelers (Category L1 & L2): Specific two-wheeler categories as per Central Motor Vehicles Rules (CMVR)
- Three-Wheelers (Category L5): Specific three-wheeler categories as per CMVR
- Passenger Cars (Category M1): Passenger car category as per CMVR
- Buses: Number of buses in the state
- Total Vehicles in State: Total count of vehicles in the state

Note: Two-wheelers and three-wheelers data are grouped under 'Two-Wheeler' and 'Three-Wheeler' categories to simplify classification.

3. Indian-EV-Data

- Model: Vehicle model
- Manufacturer: Vehicle make
- Vehicle Type: Type of two-wheeler (Scooter/Bike)
- Battery Capacity (kWh): Battery capacity of the two-wheeler
- Range per Charge (km): Distance covered on a single charge
- Charging Time: Time taken to fully charge the vehicle
- Price: Price of the two-wheeler

- Power (HP or kW): Power output of the two-wheeler
- Top Speed (km/h): Maximum speed of the vehicle
- Year of Manufacture: Year the vehicle was manufactured
-

Steps Involved in Market Segmentation Project

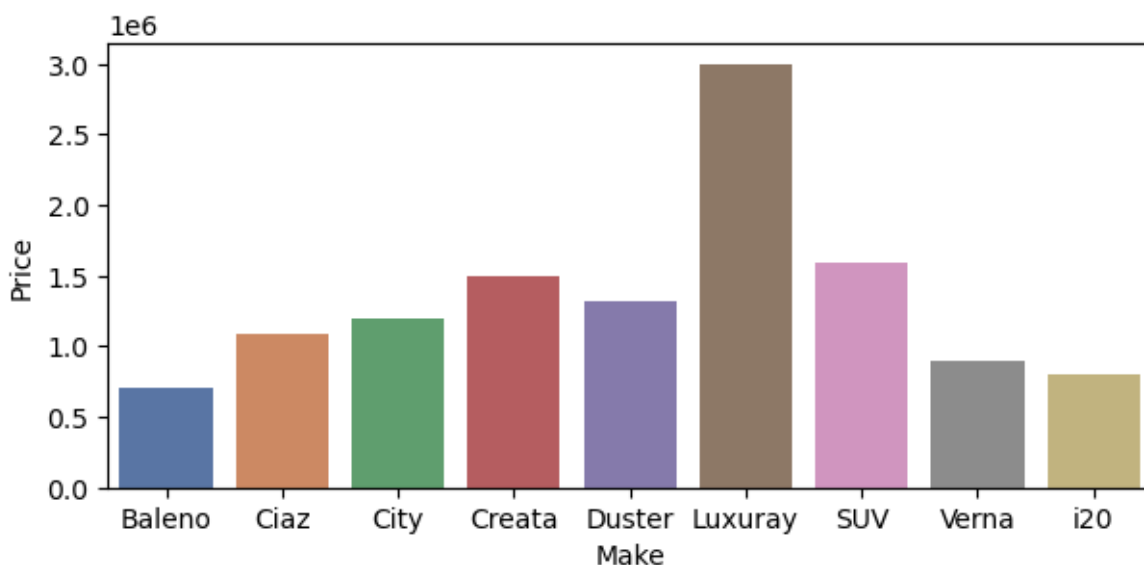
1. Data Import: Import data into Jupyter Notebook using the Pandas library.
2. Data Cleaning: Check for null values, duplicates, missing values, and errors in the datasets.
3. Exploratory Data Analysis (EDA): Perform EDA to gain meaningful insights and visualize them through charts, graphs, etc.
4. Feature Selection: Select features with positive correlations and eliminate those that do not significantly contribute to the analysis.
5. Label Encoding & One-Hot Encoding: Prepare categorical data for analysis.
6. Principal Component Analysis (PCA): Reduce dimensionality for better clustering.
7. Clustering Algorithm: Apply clustering algorithms to segment the data.
8. Plot Clusters: Visualize the clusters.
9. Extract Insights: Derive meaningful insights from the generated clusters.

Exploratory Data Analysis (EDA)

EDA involves analyzing the data to uncover hidden trends and patterns. It includes visualizing the data through graphs such as line graphs, bar graphs, scatter plots, histograms, and maps. EDA helps in understanding the data better without assumptions and can be univariate, bivariate, or multivariate.

EDA for Indian Automobile Buying Behavior Study 1.0

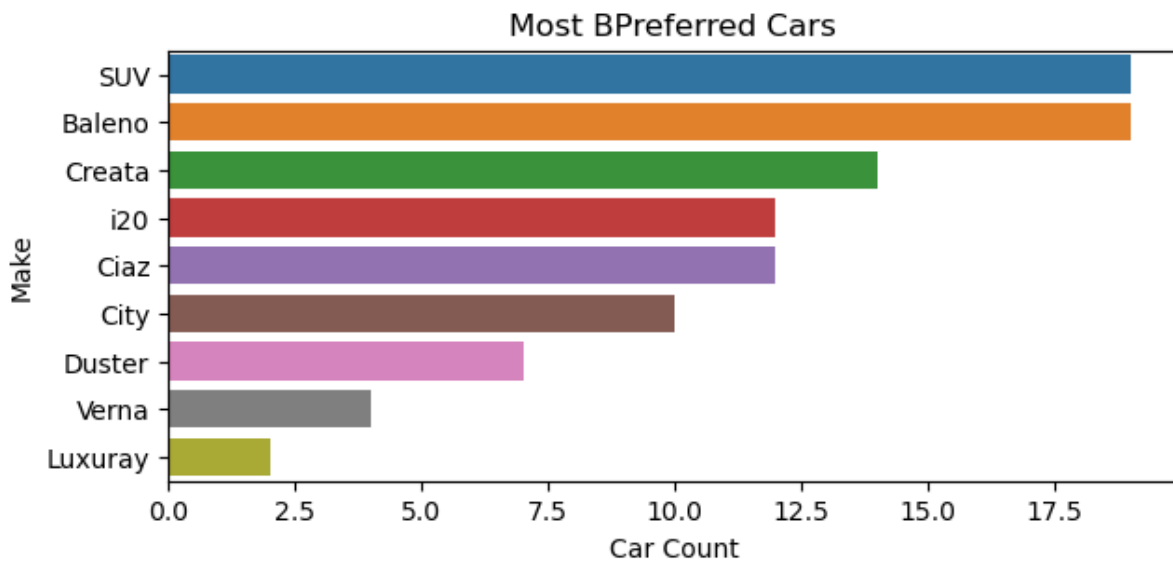
This Data set focuses on Customer, Age, Salary, Car Bought, No of Dependents, spouse salary, Brand Purchased etc .Lets get insights to the data set



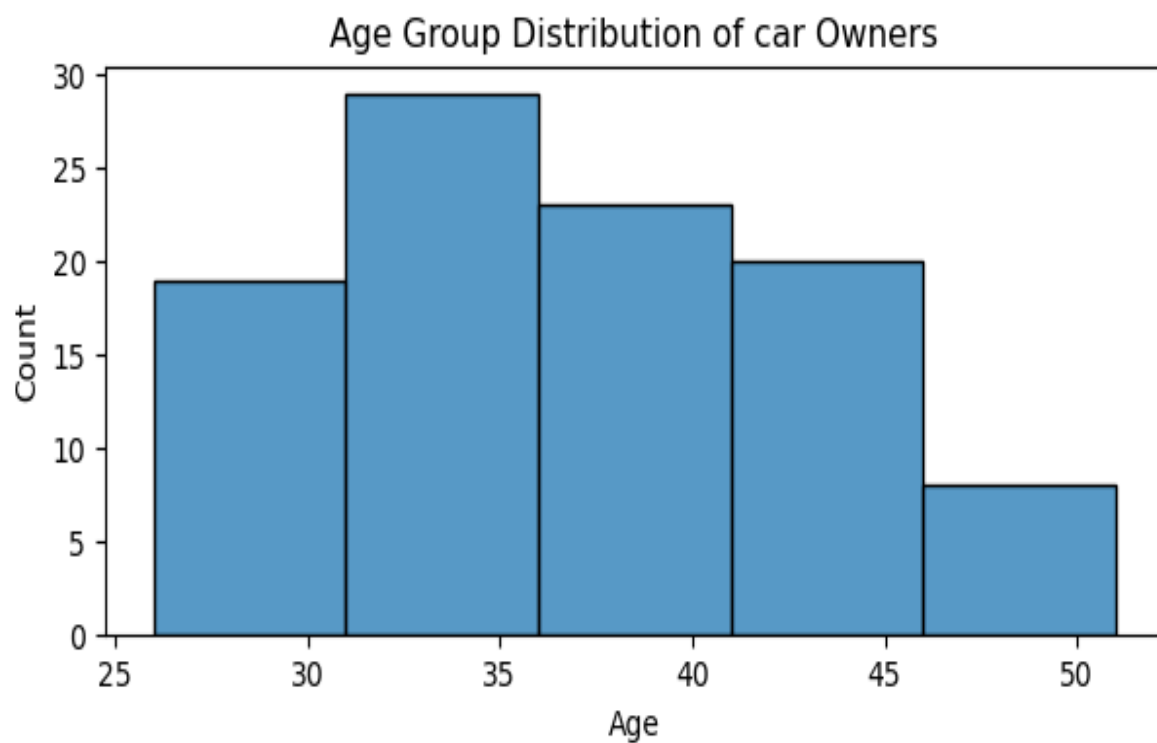
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Above

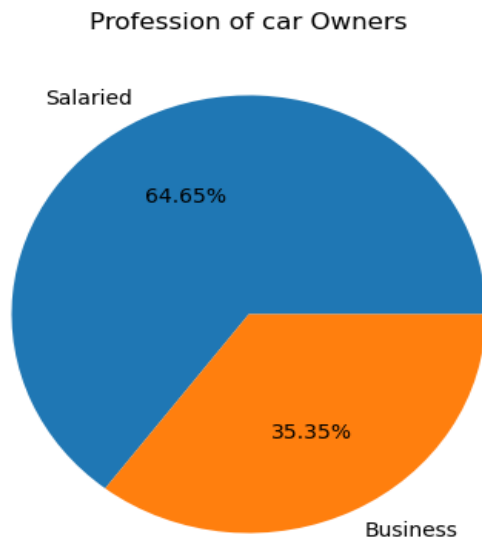
Bar Graph Illustrates the Average Price of a Car available in Indian Market.



2.Above Bar Chart Indicates the Most Preferred Car Brands in India



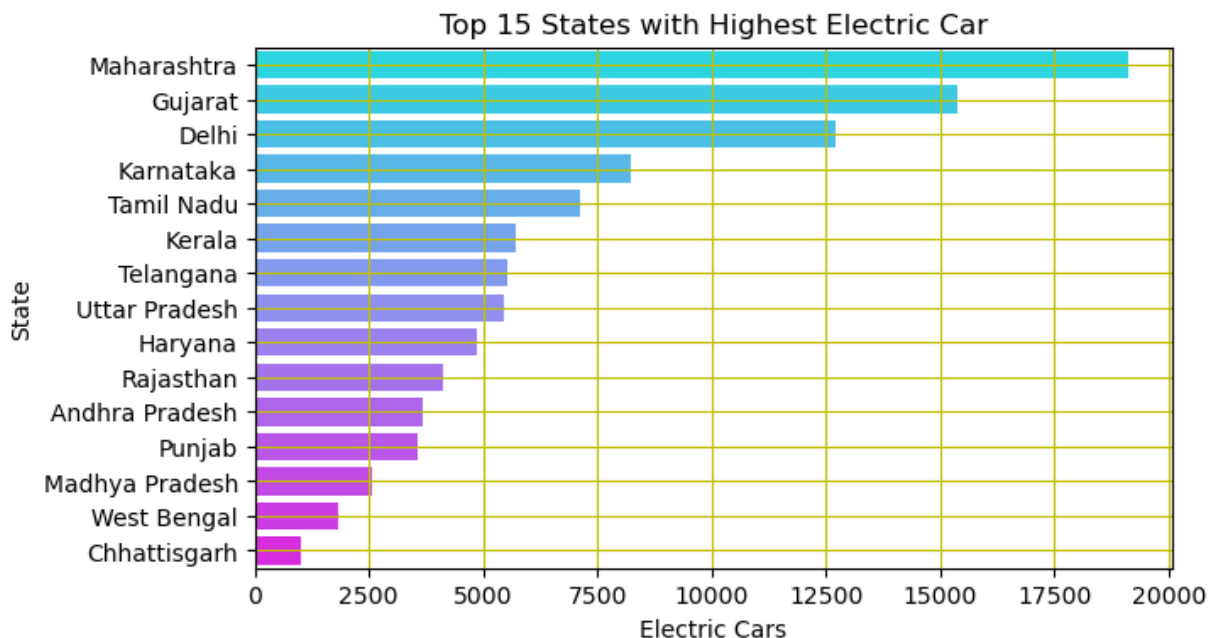
3.Above Histogram Summarizes the Frequency Distribution of Car Owners. Largest Number of Customers are spread in age group 35-40



4)Above Pie Chart Indicated Profession of Car Owners as Percentage of Total

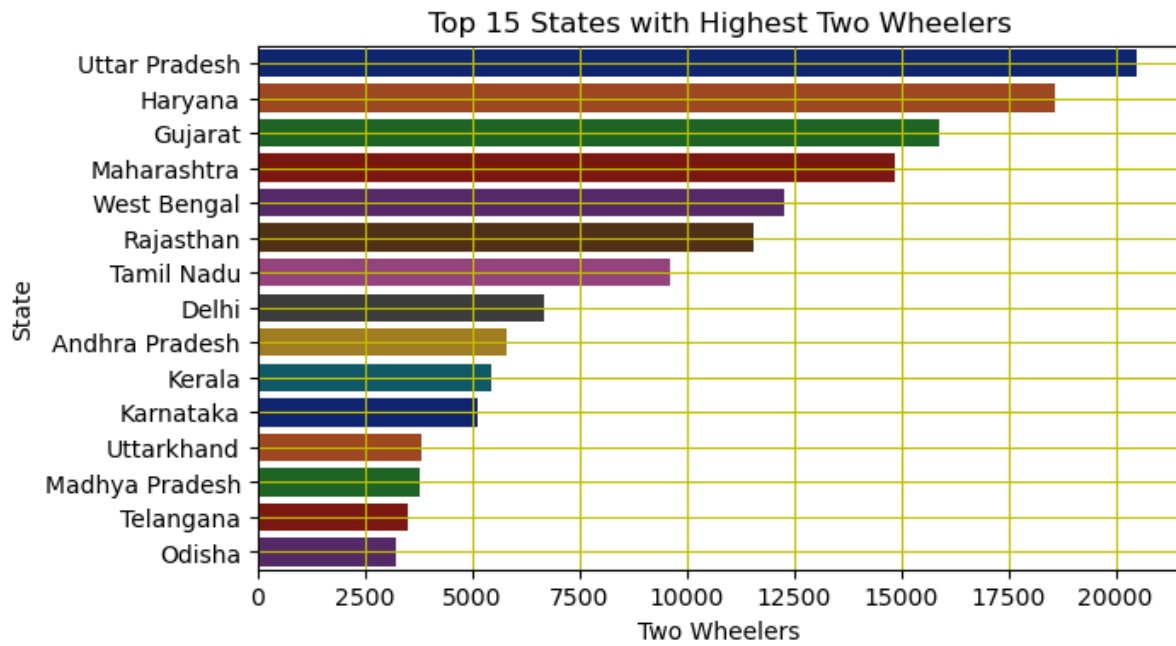
2. EDA for EV Stats-1 Data Set

This Data set Gives an idea of EV'S of Distinct Segments operating in India and their Count. Analysis should focus on states in which specific type of 2 wheeler is operating and Numbers, Further Analysis should be made on observations.



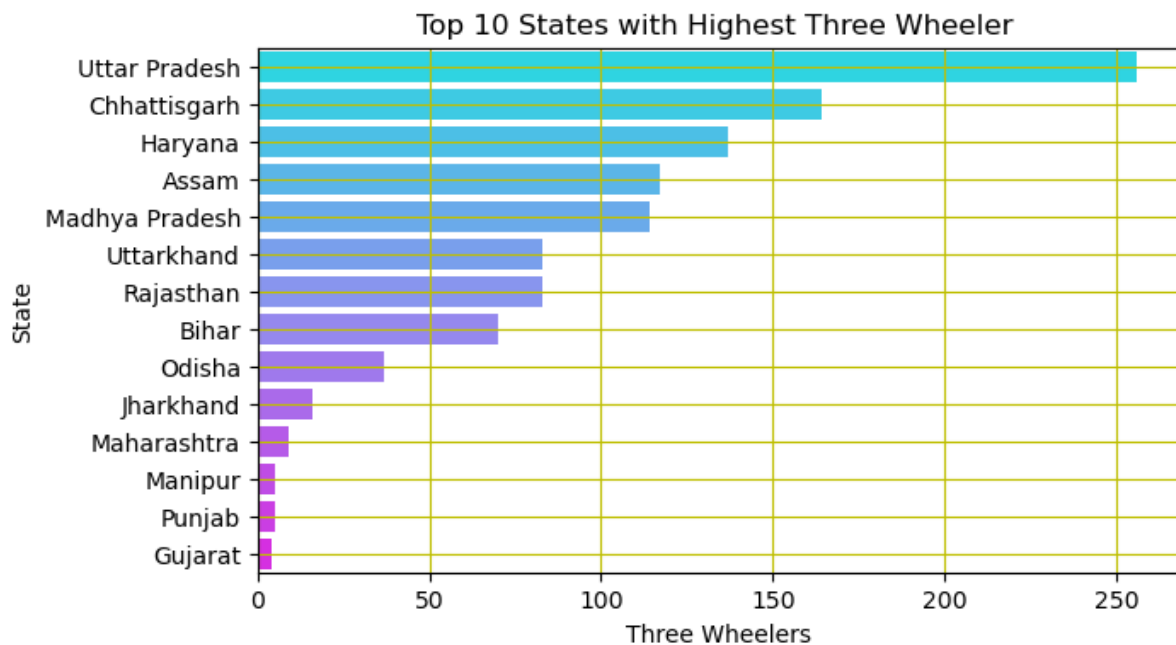
1. A.

Above Bar Chart Shows Top 15 States with Highest Number of Electric Cars. Electric Cars are operating across india with negligible Numbers in North East. Hence further Analysis is needed to expand the scope in North East.

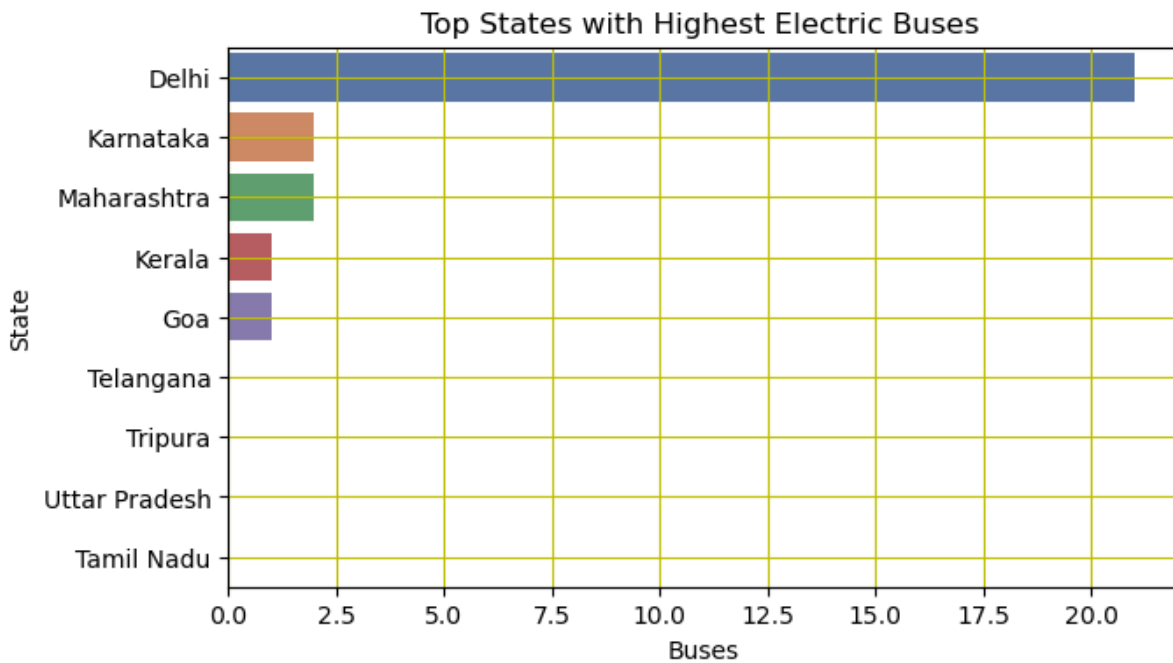


B.

Above Bar Chart Shows Top 15 States with Highest Number of Electric Two Wheelers. Again Two Wheelers operate across india except NorthEast.No conclusion can be made for poor development pf E lectric 2 Wheelers in North East

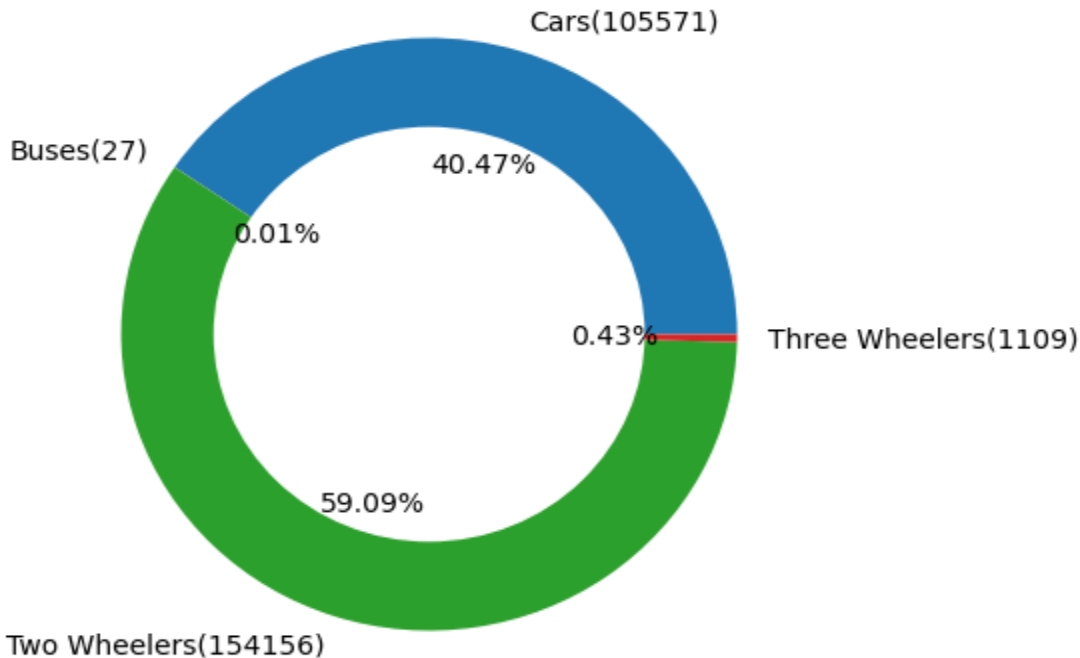


3.Above Bar Chart Shows Top 15 States with Highest Number of Electric Three Wheelers. Highest Number of 3 Wheelers are operating in Nort india and North East as a means of Public Transport and Neglegible in South india and West india as per Data in Data set.



4. Above Bar Chart Shows Top 15 States with Highest Number of Buses. The Above Graph is a clear indication that Public Transport segment is not well developed and there is a huge scope for vehicles in this segment in all states across india.Total only 5 States has Electric Buses operating.

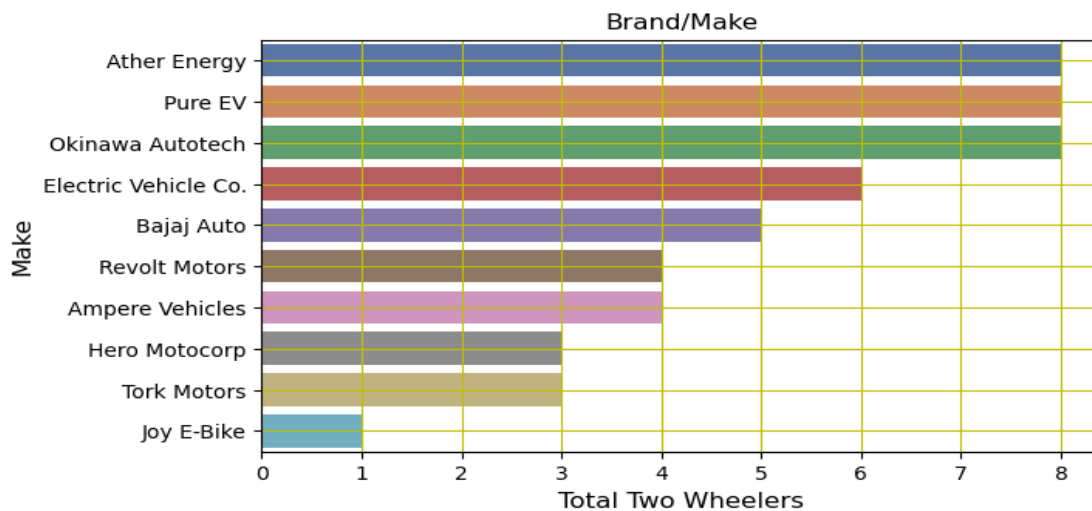
Market Segment for EV in India by Vehicle Type



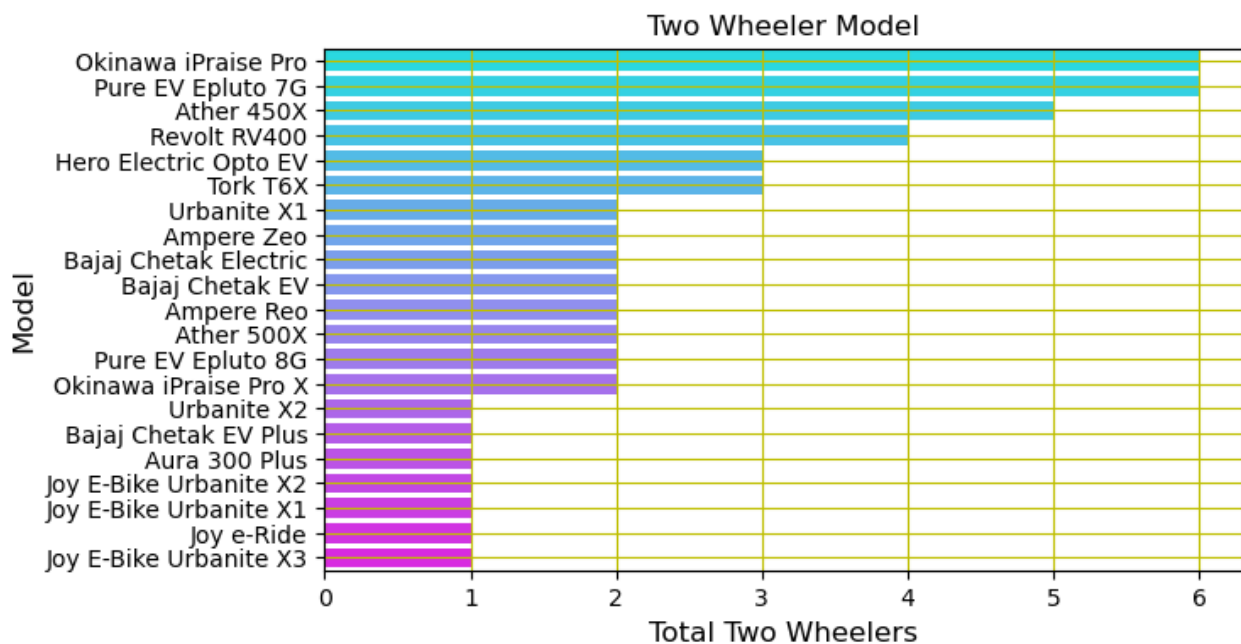
Above Donut Chart indicates the Market Share of each segment of each Vehicle Category in India. As Indicated Maximum Market of EV'S is Dominated by Two Wheelers(59%),followed by Cars Wheelers(40%).Less than 1 percent is contribution of Three Wheelers and Two Wheelers. Further Data Analysis can potentially open a Market for 3 Wheelers, Buses and Trucks in india.

3.Indian-EV-Data

This Data set Has Lists all the Electric Two Wheelers available in India and Price for the same with other parameters like Make, Model, Power Torque etc, Lets Analyse this Data Set using EDA.

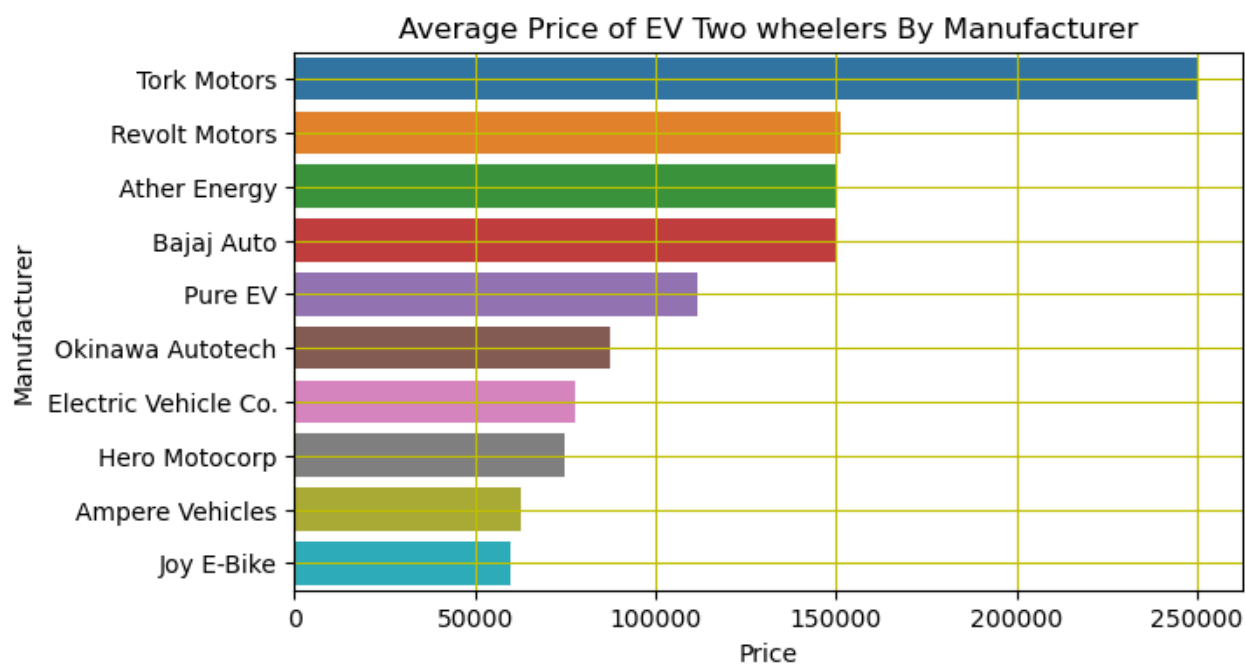


Top Brands in India Manufacturing EV

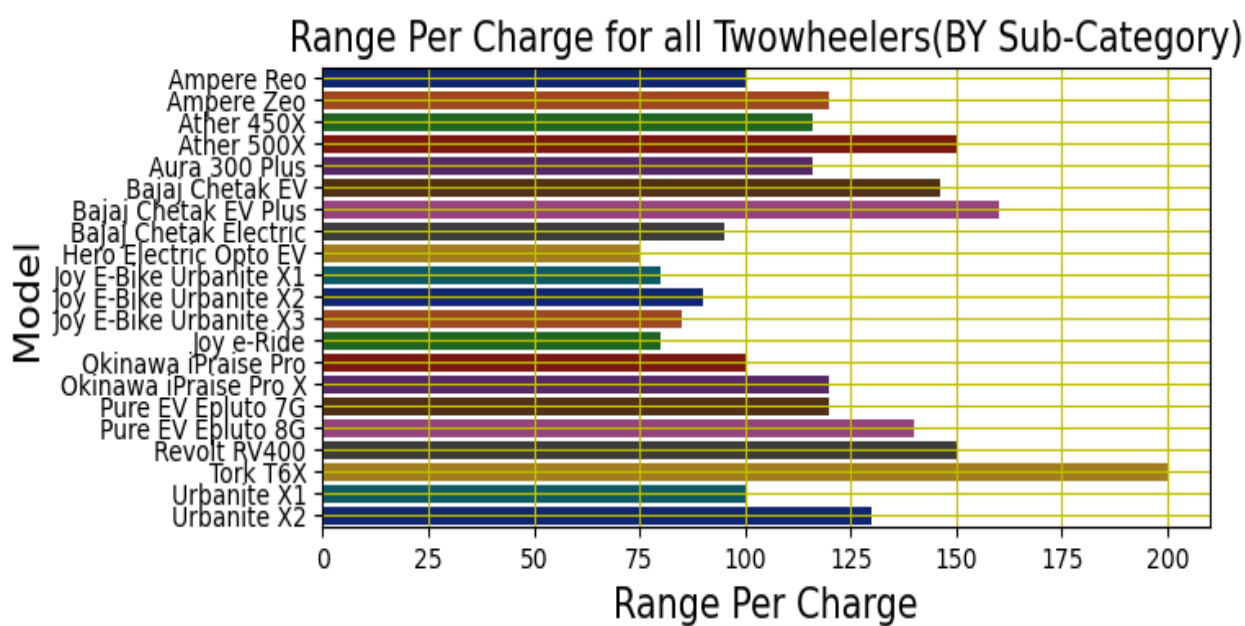


EV Two Wheelers Models in India

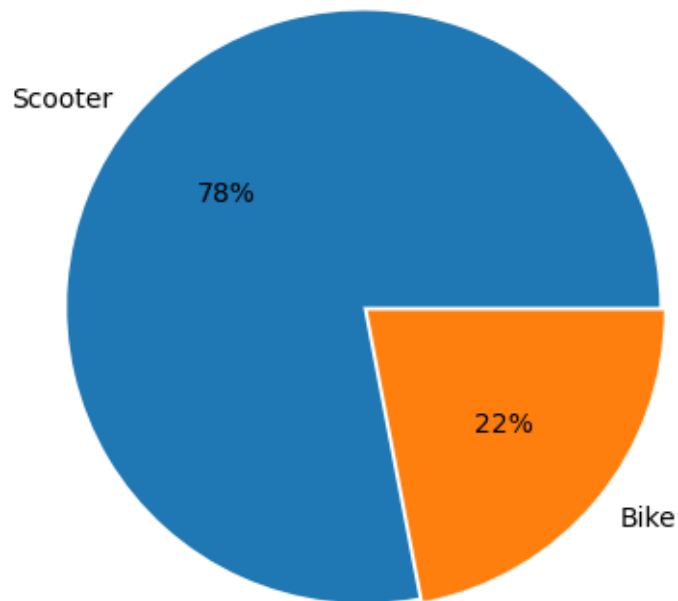
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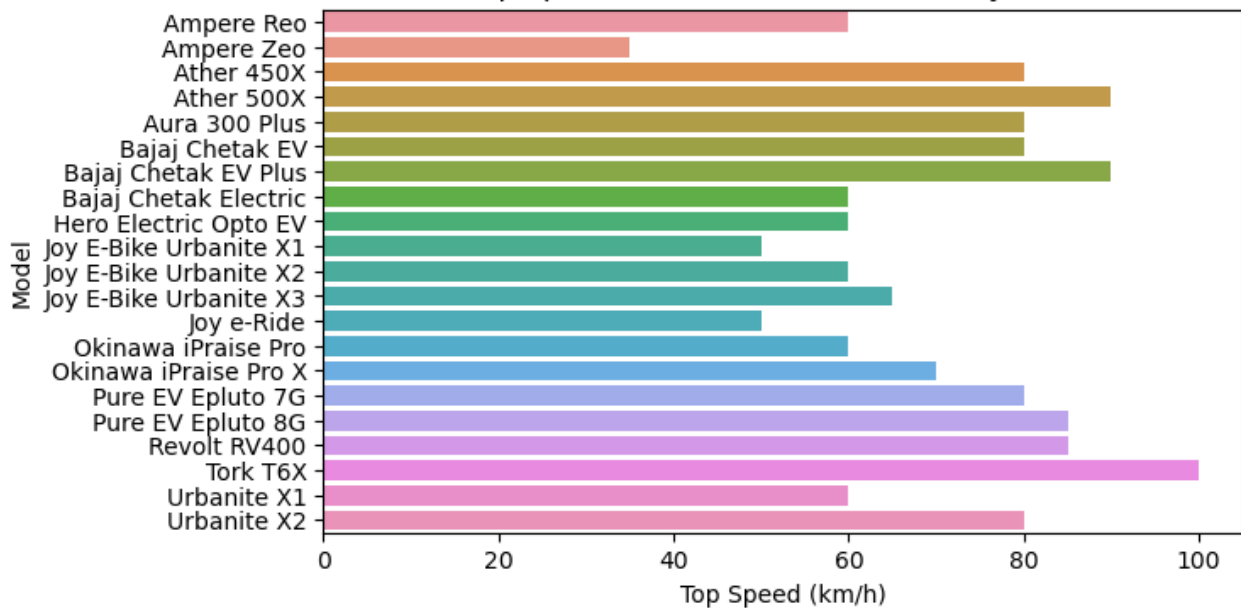
Average Price of Electric Two Wheelers By Manufacturer

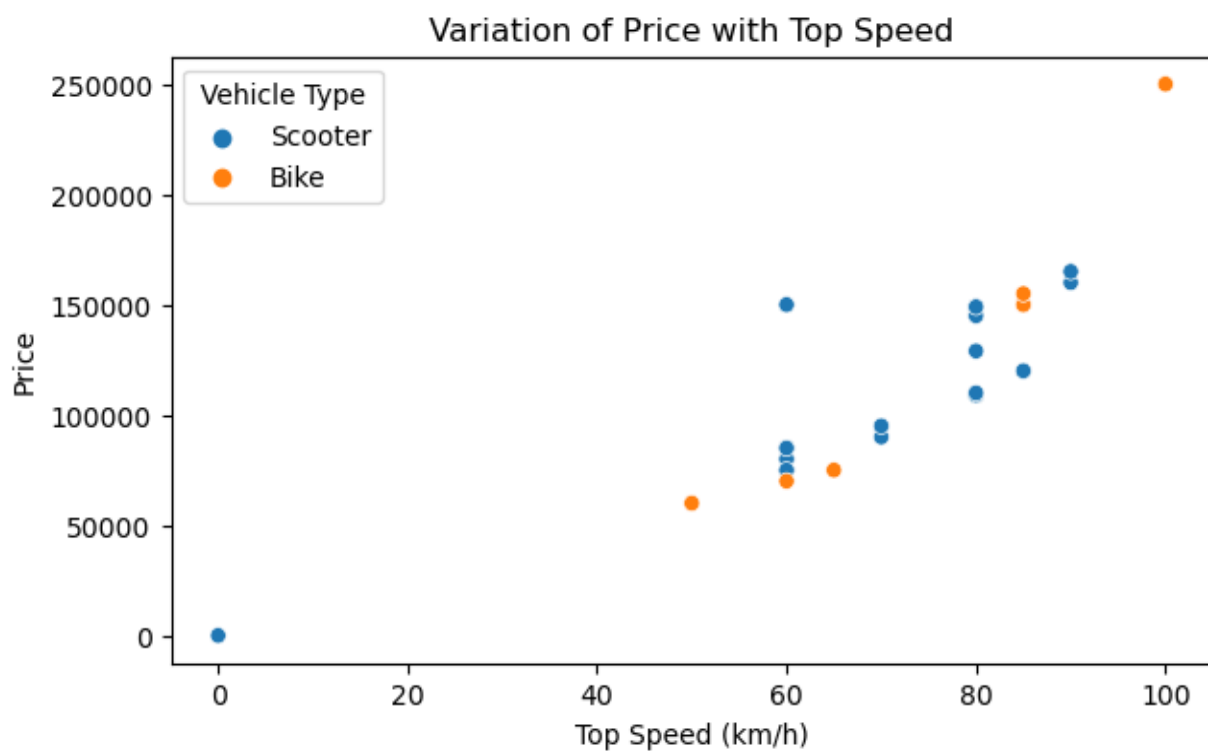
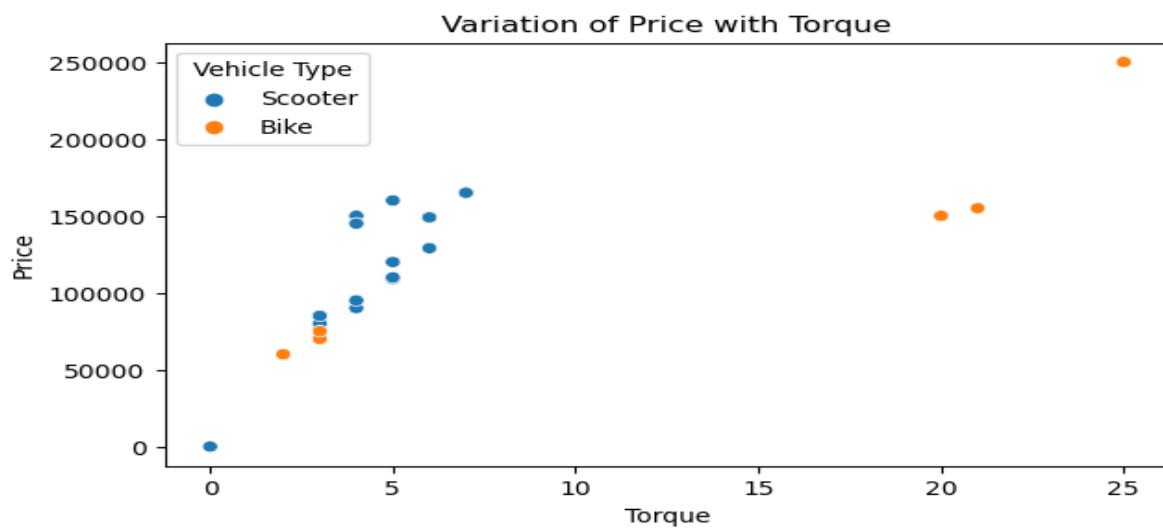


Two Wheeler Category Type(Bike/Scooter)

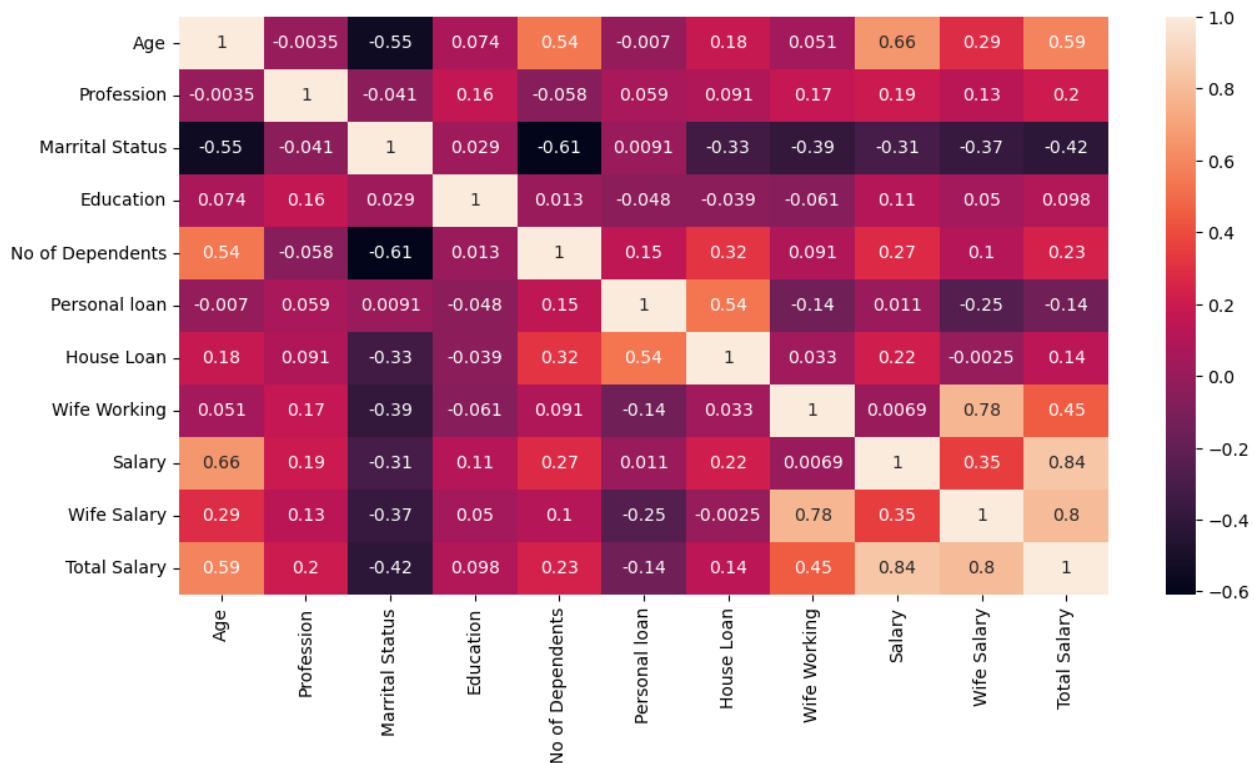


Top Speed of Electric Two wheelers by Model





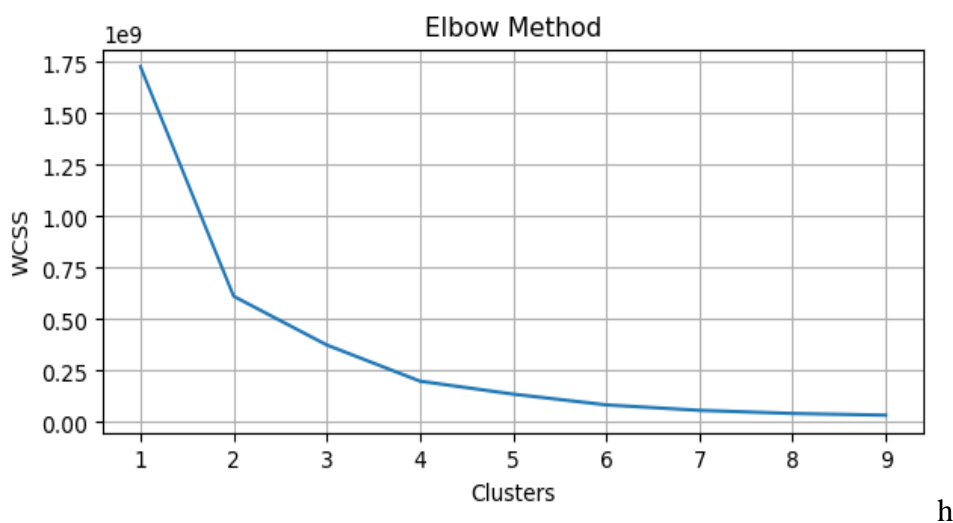
Corelation for one of the Data sets Used



Corelation Matrix for Customer Behaviour Data set. Please Note that Higher the Value higher is correlation. Those with a negative correlation can be excluded and this way we can eliminate few Features.

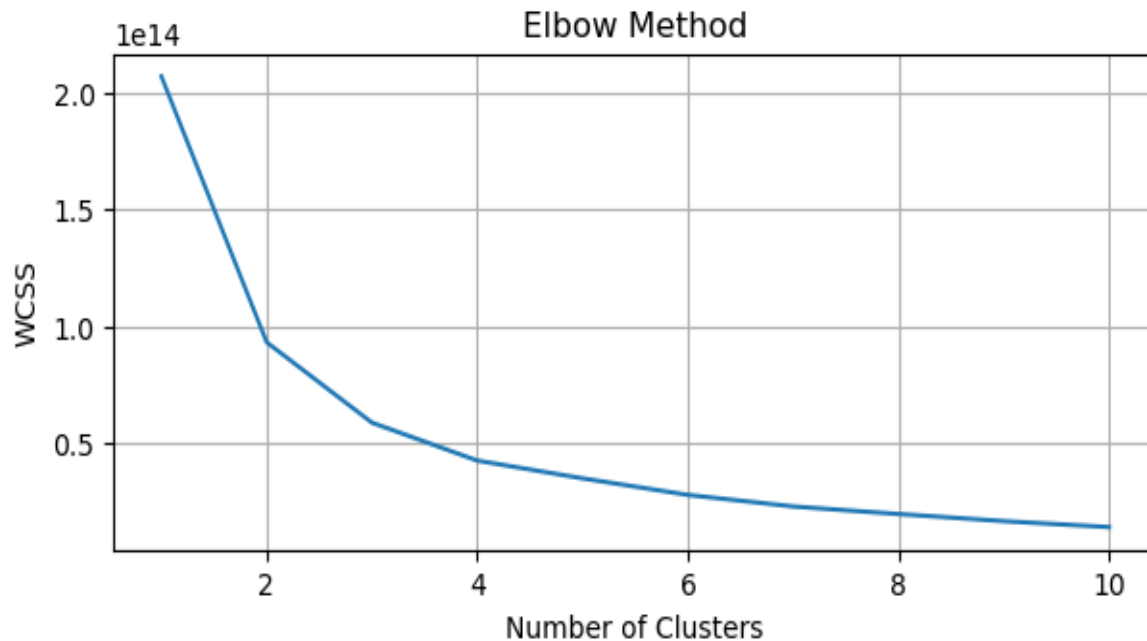
K-Means Clustering

K-Means clustering is an Unsupervised learning technique that's used for clustering the data set into clusters based on patterns in the data we are calculating WCSS (Within-Cluster Sum of Square). WCSS is the sum of squared distance between each point and the centroid in a cluster. When we plot the WCSS with the K value, the plot looks like an Elbow. WCSS is highest when Number of clusters is 1 and keeps on increasing. We Iterate from 1 to 11 and plot the graph and point where elbow is formed indicates number of segments.



A. Market Segmentation for Indian Automobile Buying Behavior Study 1.0

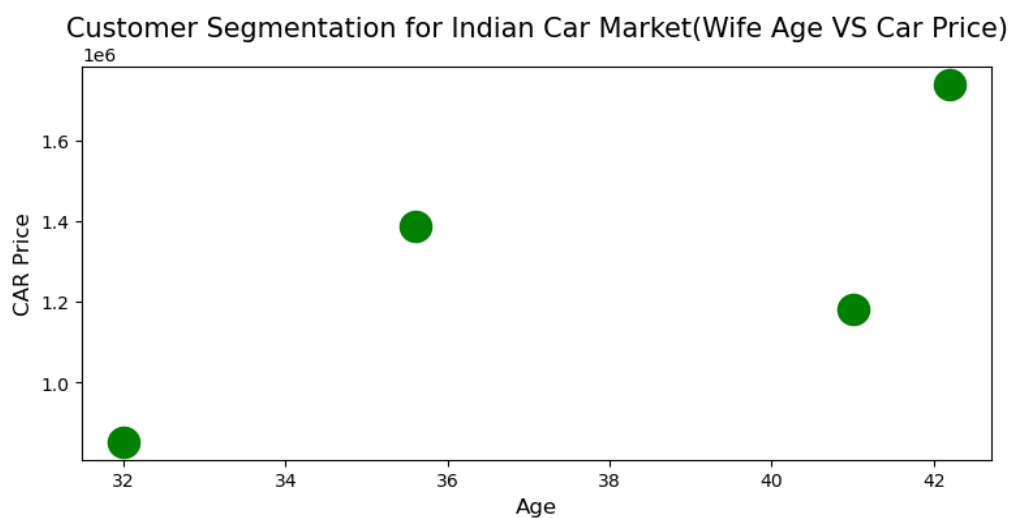
Attached is the Elbow Chart for the above Data set.

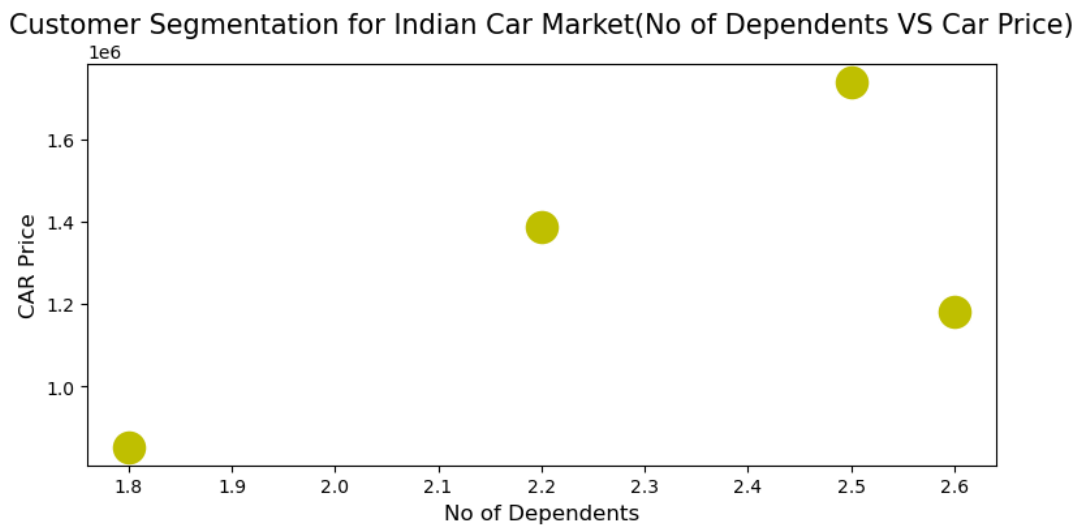
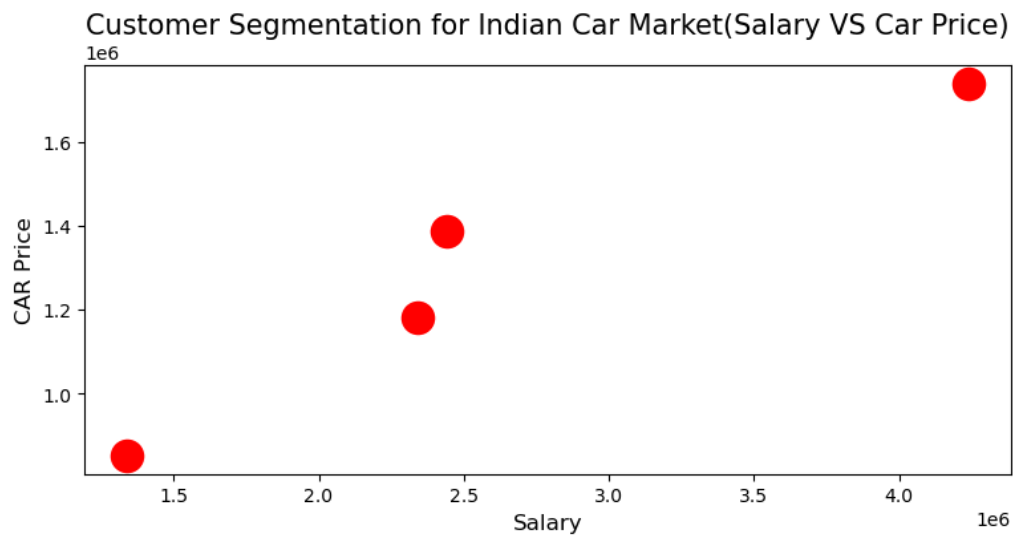
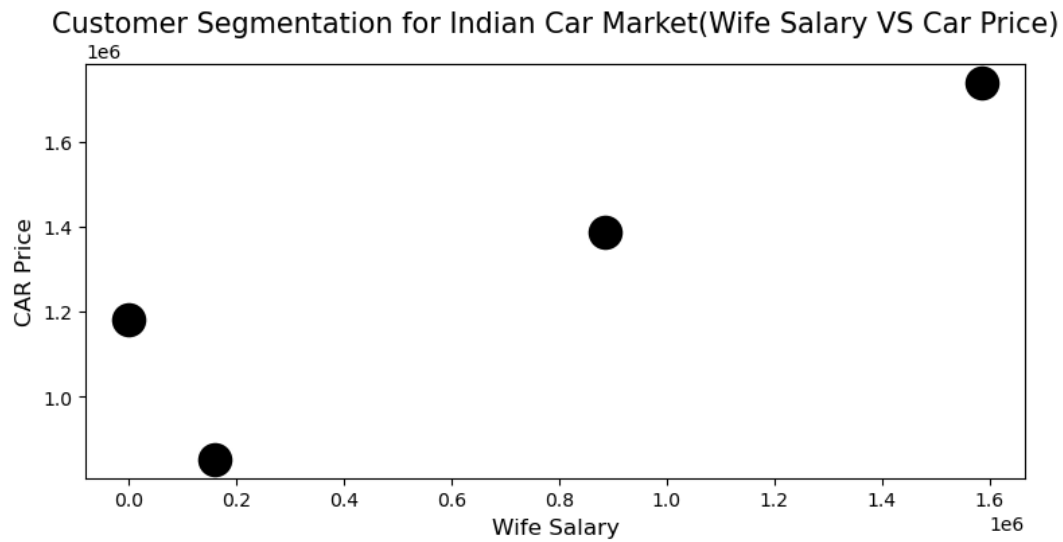


From the above we can observe that Number of clusters=3 and entire Data set will be divided to 3 Clusters and below is the count of each segment.

| Cluster_No | count |
|------------|-------|
| 1 | 38 |
| 3 | 26 |
| 0 | 20 |
| 2 | 15 |

Split Based on Distinct Parameters is as follows.



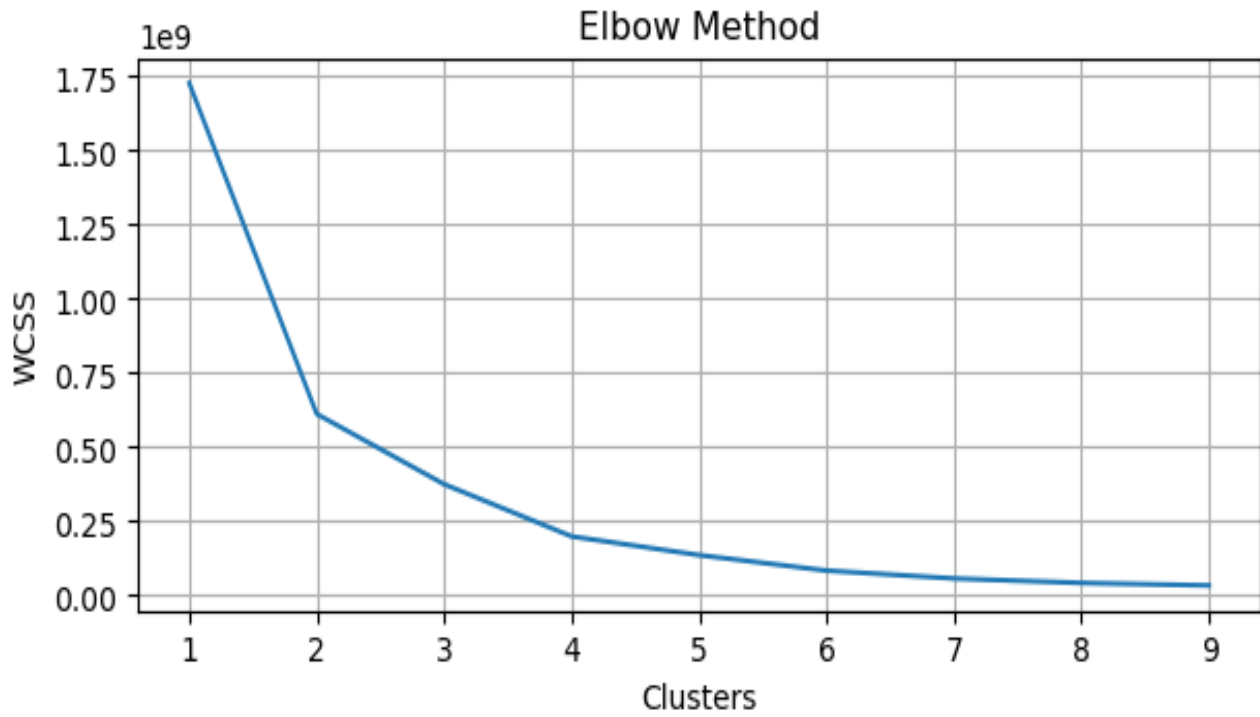


Clustered Data Set for Customers

| ID | Age | Profession | Marital Status | Education | No of Dependents | Personal loan | House Loan | Wife Working | Salary | Wife Salary | Total Salary | Make | Price | Segment |
|----|-----|------------|----------------|-----------|------------------|---------------|------------|--------------|---------|-------------|--------------|---------|---------|---------|
| 2 | 45 | 0 | 0 | 0 | 4 | 1 | 1 | 0 | 1800000 | 0 | 1800000 | Duster | 1200000 | 3 |
| 8 | 34 | 0 | 0 | 1 | 4 | 0 | 0 | 0 | 2000000 | 0 | 2000000 | Verna | 1100000 | 3 |
| 34 | 42 | 1 | 0 | 0 | 4 | 1 | 1 | 0 | 2100000 | 0 | 2100000 | Ciaz | 1100000 | 3 |
| 50 | 49 | 0 | 0 | 1 | 4 | 0 | 0 | 0 | 2000000 | 0 | 2000000 | Duster | 1300000 | 3 |
| 52 | 44 | 1 | 0 | 1 | 4 | 0 | 0 | 0 | 2700000 | 0 | 2700000 | SUV | 1600000 | 3 |
| 55 | 41 | 1 | 0 | 1 | 3 | 1 | 1 | 0 | 3100000 | 0 | 3100000 | Creata | 1500000 | 3 |
| 56 | 41 | 1 | 0 | 0 | 3 | 0 | 0 | 0 | 2600000 | 0 | 2600000 | Ciaz | 1100000 | 3 |
| 62 | 31 | 0 | 1 | 0 | 0 | 0 | 0 | 0 | 2000000 | 0 | 2000000 | SUV | 1500000 | 3 |
| 67 | 39 | 1 | 0 | 1 | 2 | 0 | 0 | 0 | 1900000 | 0 | 1900000 | City | 1200000 | 3 |
| 76 | 36 | 0 | 0 | 1 | 2 | 1 | 1 | 0 | 1900000 | 0 | 1900000 | i20 | 800000 | 3 |
| 78 | 44 | 1 | 0 | 1 | 3 | 0 | 0 | 0 | 3100000 | 0 | 3100000 | Creata | 1500000 | 3 |
| 80 | 46 | 1 | 0 | 0 | 3 | 1 | 1 | 0 | 2100000 | 0 | 2100000 | Baleno | 800000 | 3 |
| 82 | 43 | 1 | 0 | 1 | 3 | 1 | 1 | 0 | 2400000 | 0 | 2400000 | Ciaz | 1100000 | 3 |
| 83 | 42 | 1 | 0 | 1 | 2 | 0 | 1 | 0 | 2900000 | 0 | 2900000 | Duster | 1300000 | 3 |
| 84 | 42 | 1 | 0 | 0 | 3 | 1 | 1 | 0 | 2700000 | 0 | 2700000 | i20 | 800000 | 3 |
| 89 | 34 | 0 | 0 | 1 | 3 | 1 | 1 | 0 | 1900000 | 0 | 1900000 | Baleno | 700000 | 3 |
| 92 | 37 | 1 | 0 | 0 | 2 | 0 | 1 | 0 | 2800000 | 0 | 2800000 | City | 1200000 | 3 |
| 94 | 27 | 0 | 1 | 0 | 0 | 0 | 0 | 0 | 2400000 | 0 | 2400000 | SUV | 1600000 | 3 |
| 96 | 51 | 0 | 0 | 0 | 2 | 1 | 1 | 0 | 2200000 | 0 | 2200000 | Ciaz | 1100000 | 3 |
| 98 | 51 | 1 | 0 | 1 | 2 | 1 | 1 | 0 | 2200000 | 0 | 2200000 | Ciaz | 1100000 | 3 |
| 1 | 35 | 1 | 0 | 1 | 2 | 1 | 1 | 1 | 1400000 | 600000 | 2000000 | Ciaz | 1000000 | 2 |
| 3 | 41 | 0 | 0 | 1 | 3 | 0 | 0 | 0 | 1600000 | 600000 | 2200000 | City | 1200000 | 2 |
| 4 | 31 | 1 | 0 | 1 | 2 | 1 | 0 | 0 | 1800000 | 800000 | 2600000 | SUV | 1600000 | 2 |
| 10 | 35 | 1 | 0 | 1 | 4 | 0 | 0 | 0 | 1300000 | 700000 | 2000000 | SUV | 1600000 | 2 |
| 20 | 37 | 1 | 0 | 1 | 3 | 0 | 1 | 0 | 1700000 | 800000 | 2500000 | City | 1200000 | 2 |
| 22 | 36 | 0 | 0 | 0 | 3 | 0 | 0 | 1 | 1400000 | 1000000 | 2400000 | SUV | 1600000 | 2 |
| 24 | 35 | 1 | 0 | 0 | 3 | 1 | 0 | 0 | 1400000 | 600000 | 2000000 | Ciaz | 1100000 | 2 |
| 25 | 35 | 1 | 0 | 0 | 2 | 0 | 0 | 0 | 1800000 | 1100000 | 2900000 | SUV | 1600000 | 2 |
| 28 | 36 | 1 | 0 | 1 | 3 | 1 | 1 | 1 | 1700000 | 900000 | 2600000 | Verna | 1200000 | 2 |
| 29 | 36 | 1 | 0 | 1 | 2 | 0 | 0 | 0 | 1800000 | 900000 | 2700000 | SUV | 1600000 | 2 |
| 31 | 41 | 1 | 0 | 0 | 3 | 0 | 0 | 1 | 1400000 | 700000 | 2100000 | Duster | 1300000 | 2 |
| 32 | 41 | 1 | 0 | 1 | 4 | 0 | 1 | 1 | 1900000 | 1100000 | 3000000 | Creata | 1500000 | 2 |
| 33 | 43 | 0 | 0 | 0 | 3 | 0 | 0 | 1 | 1300000 | 900000 | 2200000 | Creata | 1500000 | 2 |
| 37 | 30 | 1 | 0 | 1 | 0 | 0 | 0 | 1 | 1300000 | 800000 | 2100000 | Creata | 1500000 | 2 |
| 39 | 31 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 1400000 | 800000 | 2200000 | Ciaz | 1100000 | 2 |
| 42 | 32 | 1 | 0 | 0 | 0 | 0 | 0 | 1 | 1600000 | 800000 | 2400000 | SUV | 1600000 | 2 |
| 43 | 34 | 1 | 0 | 1 | 3 | 0 | 0 | 0 | 1400000 | 900000 | 2300000 | Duster | 1300000 | 2 |
| 47 | 36 | 0 | 0 | 0 | 2 | 0 | 1 | 1 | 1500000 | 1100000 | 2600000 | Duster | 1300000 | 2 |
| 49 | 36 | 1 | 0 | 0 | 3 | 1 | 0 | 1 | 2200000 | 900000 | 3100000 | Creata | 1500000 | 2 |
| 64 | 31 | 1 | 0 | 0 | 2 | 0 | 1 | 1 | 1400000 | 700000 | 2100000 | Ciaz | 1100000 | 2 |
| 70 | 37 | 0 | 0 | 1 | 2 | 0 | 0 | 0 | 1400000 | 1300000 | 2700000 | Duster | 1500000 | 2 |
| 71 | 37 | 1 | 0 | 1 | 2 | 0 | 1 | 0 | 1900000 | 1300000 | 3200000 | SUV | 1600000 | 2 |
| 73 | 35 | 1 | 0 | 1 | 0 | 0 | 0 | 0 | 1400000 | 1100000 | 2500000 | SUV | 1600000 | 2 |
| 74 | 36 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1300000 | 700000 | 2000000 | City | 1200000 | 2 |
| 77 | 36 | 0 | 0 | 0 | 3 | 0 | 1 | 0 | 1800000 | 1100000 | 2900000 | SUV | 1600000 | 2 |
| 87 | 33 | 1 | 0 | 1 | 2 | 0 | 0 | 0 | 1400000 | 800000 | 2200000 | City | 1200000 | 2 |
| 15 | 49 | 1 | 0 | 1 | 4 | 0 | 0 | 1 | 2500000 | 2000000 | 4500000 | Luxuray | 3000000 | 1 |
| 35 | 42 | 1 | 0 | 1 | 3 | 0 | 0 | 0 | 2400000 | 1300000 | 3700000 | SUV | 1600000 | 1 |
| 45 | 34 | 1 | 0 | 0 | 3 | 0 | 1 | 1 | 2200000 | 1400000 | 3600000 | Creata | 1500000 | 1 |
| 51 | 49 | 1 | 0 | 1 | 3 | 0 | 0 | 0 | 2500000 | 1800000 | 4300000 | SUV | 1600000 | 1 |
| 53 | 44 | 0 | 0 | 1 | 3 | 0 | 0 | 0 | 2000000 | 1800000 | 3800000 | Creata | 1500000 | 1 |
| 54 | 41 | 1 | 0 | 0 | 2 | 0 | 0 | 1 | 2900000 | 1800000 | 4700000 | Luxuray | 3000000 | 1 |
| 66 | 39 | 0 | 0 | 0 | 2 | 1 | 1 | 1 | 2200000 | 1400000 | 3600000 | Creata | 1500000 | 1 |
| 68 | 39 | 1 | 0 | 1 | 2 | 1 | 1 | 1 | 2700000 | 1300000 | 4000000 | Creata | 1500000 | 1 |
| 75 | 36 | 1 | 0 | 1 | 2 | 0 | 0 | 1 | 2300000 | 1300000 | 3600000 | Creata | 1500000 | 1 |
| 79 | 45 | 1 | 0 | 1 | 2 | 0 | 1 | 1 | 2700000 | 1800000 | 4500000 | SUV | 1600000 | 1 |
| 81 | 44 | 0 | 0 | 0 | 2 | 0 | 0 | 0 | 3100000 | 2100000 | 5200000 | SUV | 1600000 | 1 |
| 88 | 34 | 1 | 0 | 0 | 2 | 0 | 1 | 1 | 2700000 | 1400000 | 4100000 | Creata | 1500000 | 1 |
| 91 | 36 | 1 | 0 | 1 | 3 | 1 | 1 | 1 | 3100000 | 1800000 | 4900000 | SUV | 1600000 | 1 |
| 95 | 50 | 1 | 0 | 1 | 3 | 0 | 0 | 1 | 3800000 | 1300000 | 5100000 | SUV | 1600000 | 1 |
| 97 | 51 | 1 | 0 | 1 | 2 | 0 | 0 | 1 | 2700000 | 1300000 | 4000000 | Creata | 1500000 | 1 |
| 0 | 27 | 1 | 1 | 1 | 0 | 1 | 1 | 0 | 800000 | 0 | 800000 | i20 | 800000 | 0 |
| 5 | 28 | 1 | 0 | 0 | 3 | 1 | 1 | 0 | 900000 | 0 | 900000 | Baleno | 700000 | 0 |
| 6 | 31 | 1 | 0 | 0 | 4 | 0 | 0 | 0 | 1200000 | 600000 | 1800000 | City | 1200000 | 0 |
| 7 | 33 | 0 | 0 | 1 | 4 | 0 | 0 | 0 | 1400000 | 0 | 1400000 | Baleno | 700000 | 0 |
| 9 | 34 | 1 | 0 | 0 | 3 | 1 | 1 | 1 | 1200000 | 700000 | 1900000 | i20 | 800000 | 0 |
| 11 | 35 | 1 | 0 | 0 | 4 | 1 | 1 | 2 | 1400000 | 0 | 1400000 | Baleno | 700000 | 0 |
| 12 | 29 | 1 | 0 | 1 | 0 | 0 | 0 | 0 | 900000 | 800000 | 1700000 | Verna | 1100000 | 0 |
| 13 | 30 | 0 | 1 | 1 | 2 | 1 | 0 | 0 | 1400000 | 0 | 1400000 | i20 | 800000 | 0 |
| 14 | 31 | 0 | 0 | 0 | 3 | 1 | 1 | 1 | 900000 | 400000 | 1300000 | Baleno | 700000 | 0 |
| 16 | 26 | 1 | 1 | 1 | 0 | 0 | 0 | 0 | 800000 | 0 | 800000 | i20 | 800000 | 0 |
| 17 | 27 | 1 | 1 | 0 | 0 | 0 | 1 | 0 | 800000 | 0 | 800000 | Baleno | 700000 | 0 |
| 18 | 29 | 1 | 1 | 1 | 2 | 0 | 0 | 0 | 900000 | 0 | 900000 | City | 1200000 | 0 |
| 19 | 30 | 1 | 0 | 0 | 0 | 0 | 0 | 1 | 800000 | 500000 | 1300000 | Baleno | 700000 | 0 |
| 21 | 35 | 0 | 0 | 0 | 3 | 0 | 1 | 1 | 1100000 | 800000 | 1900000 | i20 | 800000 | 0 |
| 23 | 35 | 0 | 0 | 1 | 4 | 0 | 0 | 1 | 900000 | 500000 | 1400000 | Baleno | 700000 | 0 |
| 26 | 35 | 1 | 0 | 1 | 4 | 1 | 1 | 0 | 1300000 | 0 | 1300000 | Baleno | 700000 | 0 |
| 27 | 36 | 0 | 0 | 0 | 3 | 0 | 0 | 0 | 1600000 | 0 | 1600000 | i20 | 800000 | 0 |
| 30 | 41 | 0 | 0 | 0 | 3 | 0 | 0 | 0 | 1100000 | 0 | 1100000 | Baleno | 700000 | 0 |
| 36 | 29 | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 900000 | 700000 | 1600000 | City | 1200000 | 0 |
| 38 | 30 | 0 | 1 | 1 | 0 | 0 | 0 | 0 | 1400000 | 0 | 1400000 | SUV | 1600000 | 0 |
| 40 | 31 | 0 | 0 | 0 | 3 | 1 | 1 | 1 | 1100000 | 600000 | 1700000 | i20 | 800000 | 0 |
| 41 | 32 | 0 | 0 | 1 | 0 | 0 | 0 | 0 | 1400000 | 0 | 1400000 | Ciaz | 1100000 | 0 |
| 44 | 34 | 1 | 0 | 1 | 2 | 1 | 1 | 0 | 1600000 | 0 | 1600000 | i20 | 800000 | 0 |
| 46 | 36 | 0 | 0 | 0 | 4 | 0 | 0 | 0 | 200000 | 0 | 200000 | Ciaz | 1100000 | 0 |
| 48 | 36 | 1 | 0 | 1 | 2 | 0 | 1 | 1 | 1100000 | 500000 | 1600000 | i20 | 800000 | 0 |
| 57 | 27 | 1 | 1 | 1 | 0 | 0 | 0 | 0 | 1100000 | 0 | 1100000 | Baleno | 700000 | 0 |
| 58 | 27 | 0 | 1 | 0 | 0 | 0 | 0 | 0 | 1200000 | 0 | 1200000 | Ciaz | 1100000 | 0 |
| 59 | 28 | 1 | 1 | 0 | 0 | 0 | 1 | 0 | 900000 | 0 | 900000 | Baleno | 700000 | 0 |
| 60 | 28 | 1 | 1 | 1 | 0 | 0 | 0 | 0 | 1400000 | 0 | 1400000 | City | 1200000 | 0 |
| 61 | 30 | 1 | 1 | 1 | 0 | 0 | 0 | 0 | 1500000 | 0 | 1500000 | Creata | 1500000 | 0 |
| 63 | 31 | 1 | 0 | 1 | 3 | 1 | 1 | 0 | 1800000 | 0 | 1800000 | Baleno | 700000 | 0 |
| 65 | 39 | 0 | 0 | 1 | 3 | 0 | 0 | 0 | 1600000 | 0 | 1600000 | Baleno | 700000 | 0 |
| 69 | 37 | 0 | 0 | 0 | 3 | 0 | 0 | 0 | 1700000 | 0 | 1700000 | Verna | 1200000 | 0 |
| 72 | 37 | 1 | 0 | 0 | 2 | 0 | 0 | 0 | 1300000 | 0 | 1300000 | Baleno | 700000 | 0 |
| 85 | 42 | 0 | 0 | 1 | 3 | 1 | 1 | 0 | 1800000 | 0 | 1800000 | Baleno | 700000 | 0 |
| 86 | 29 | 0 | 1 | 0 | 0 | 0 | 1 | 0 | 1100000 | 0 | 1100000 | Baleno | 700000 | 0 |
| 90 | 35 | 1 | 0 | 0 | 2 | 0 | 0 | 0 | 1400000 | 0 | 1400000 | i20 | 800000 | 0 |
| 93 | 27 | 1 | 1 | 1 | 0 | 0 | 0 | 0 | 900000 | 0 | 900000 | Baleno | 700000 | 0 |

B. Market Segmentation for EV Stats-1

This Data Set has Data of Distinct EV S with Number of vehicles in each segment and respective states,so clustering will be purely based on segmenting States to distinct segments based on vehicles and Vehicle Count. Segmentation is as follows.



From the above Number of Clusters=4 and Number of units in each Cluster is as follows.

| Cluster | count |
|---------|-------|
| 0 | 18 |
| 3 | 6 |
| 1 | 4 |
| 2 | 2 |

Assigning Clusters to Respective States

| | State | cars | Buses | Two Wheelers | Three Wheelers | Total | Cluster_Number |
|----|-----------------------|--------|-------|--------------|----------------|--------|----------------|
| 24 | Uttar Pradesh | 5445 | 0 | 20508 | 256 | 26209 | 3 |
| 7 | Haryana | 4878 | 0 | 18574 | 137 | 23589 | 3 |
| 20 | Rajasthan | 4116 | 0 | 11564 | 83 | 15763 | 3 |
| 26 | West Bengal | 1840 | 0 | 12297 | 3 | 14140 | 3 |
| 14 | Maharashtra | 19129 | 2 | 14873 | 9 | 34013 | 2 |
| 6 | Gujarat | 15388 | 0 | 15875 | 4 | 31267 | 2 |
| 19 | Punjab | 3567 | 0 | 2966 | 5 | 6538 | 1 |
| 13 | Madhya Pradesh | 2562 | 0 | 3785 | 114 | 6461 | 1 |
| 3 | Chhattisgarh | 997 | 0 | 3073 | 164 | 4234 | 1 |
| 25 | Uttarkhand | 265 | 0 | 3830 | 83 | 4178 | 1 |
| 18 | Odisha | 594 | 0 | 3232 | 37 | 3863 | 1 |
| 2 | Bihar | 271 | 0 | 2830 | 70 | 3171 | 1 |
| 28 | Chandigarh | 974 | 0 | 1526 | 0 | 2500 | 1 |
| 1 | Assam | 151 | 0 | 1607 | 117 | 1875 | 1 |
| 10 | Jharkhand | 655 | 0 | 1039 | 16 | 1710 | 1 |
| 29 | Dadra and Nagar Havel | 803 | 0 | 13 | 0 | 816 | 1 |
| 5 | Goa | 513 | 1 | 0 | 0 | 514 | 1 |
| 9 | Jammu & Kashmir | 208 | 0 | 230 | 0 | 438 | 1 |
| 8 | Himachal Pradesh | 98 | 0 | 0 | 0 | 98 | 1 |
| 27 | Andaman & Nicobar isl | 82 | 0 | 0 | 0 | 82 | 1 |
| 23 | Tripura | 8 | 0 | 73 | 0 | 81 | 1 |
| 15 | Manipur | 12 | 0 | 35 | 5 | 52 | 1 |
| 17 | Nagaland | 1 | 0 | 23 | 0 | 24 | 1 |
| 16 | Meghalaya | 6 | 0 | 0 | 0 | 6 | 1 |
| 4 | Delhi | 12695 | 21 | 6664 | 1 | 19381 | 0 |
| 21 | Tamil Nadu | 7132 | 0 | 9614 | 0 | 16746 | 0 |
| 11 | Karnataka | 8242 | 2 | 5140 | 2 | 13386 | 0 |
| 12 | Kerala | 5729 | 1 | 5471 | 1 | 11202 | 0 |
| 0 | Andhra Pradesh | 3680 | 0 | 5812 | 0 | 9492 | 0 |
| 22 | Telangana | 5530 | 0 | 3502 | 2 | 9034 | 0 |
| 30 | Total | 105571 | 27 | 154156 | 1109 | 260863 | |

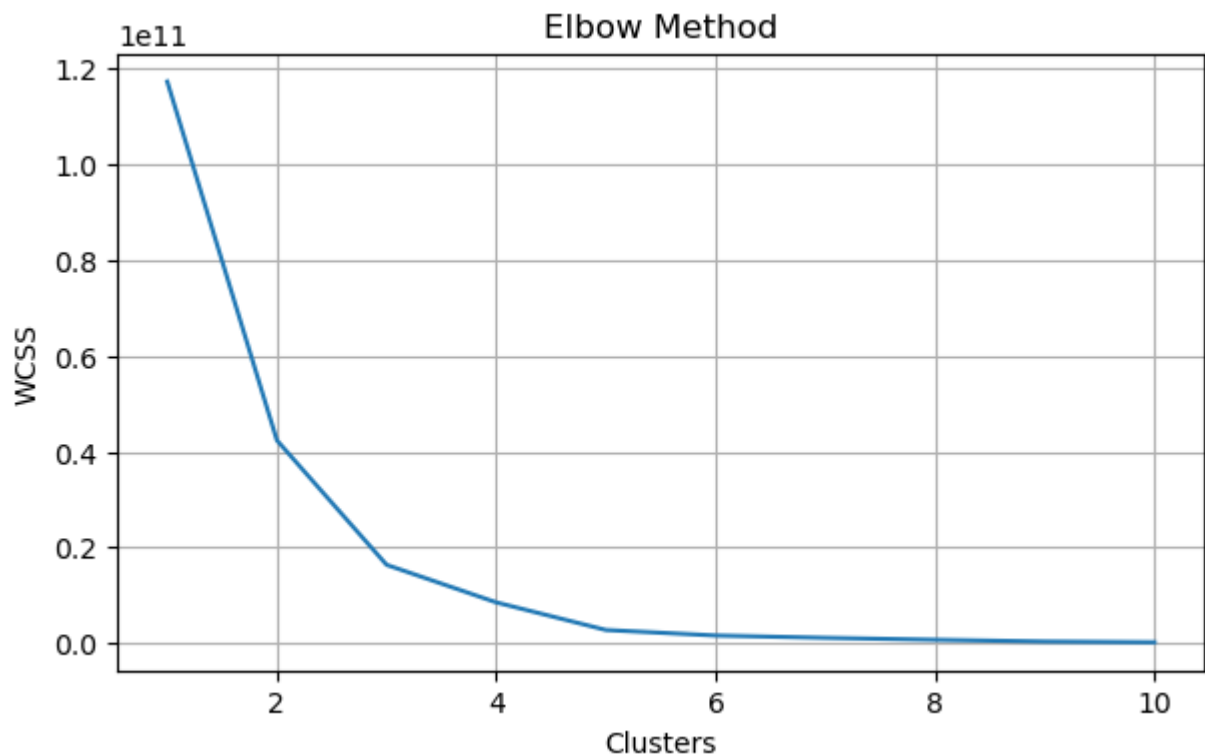
C. Market Segmentation for Electric Two Wheelers.

In this Case Segmentation will be Based on lot of parameters like Power,Torque, Speed ,Vehicle Category(Bike/Scooter),Range per Charge etc.

Feature Selection: Below is the list of features Selected and their Correlationship.

| | |
|------------------------|----------|
| Top Speed (km/h) | 0.838972 |
| Charging Time | 0.826697 |
| Range per Charge (km) | 0.823016 |
| Battery Capacity (kWh) | 0.811086 |
| Power (HP or kW) | 0.784330 |
| Year of Manufacture | 0.348236 |

K-Means Cluster for Electric Two Wheeler.

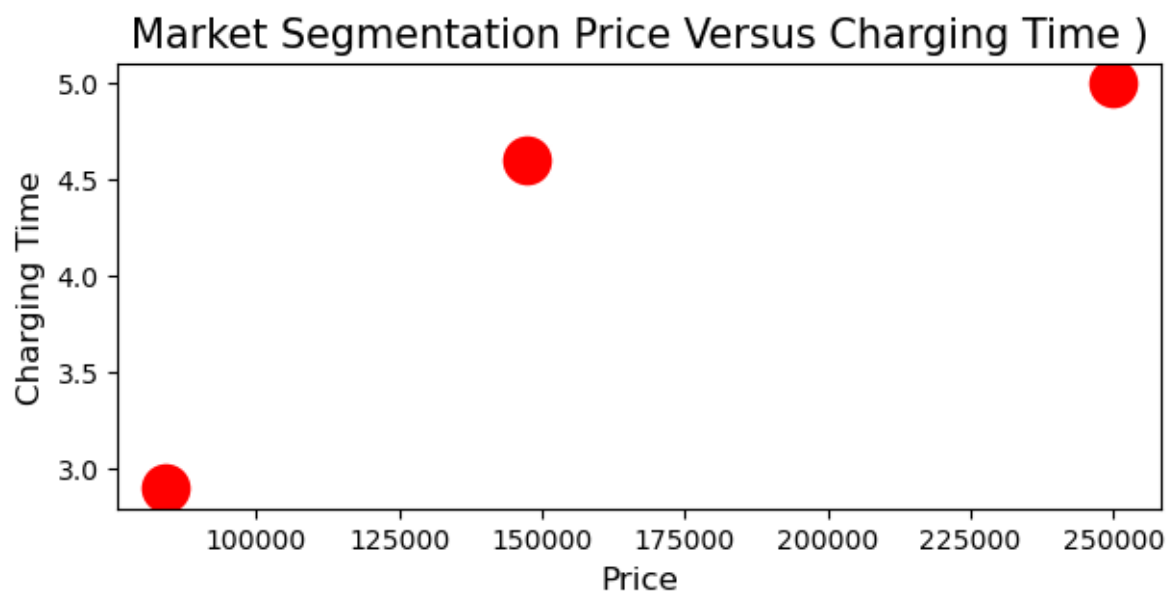


\ From above diagram Number of clusters=3

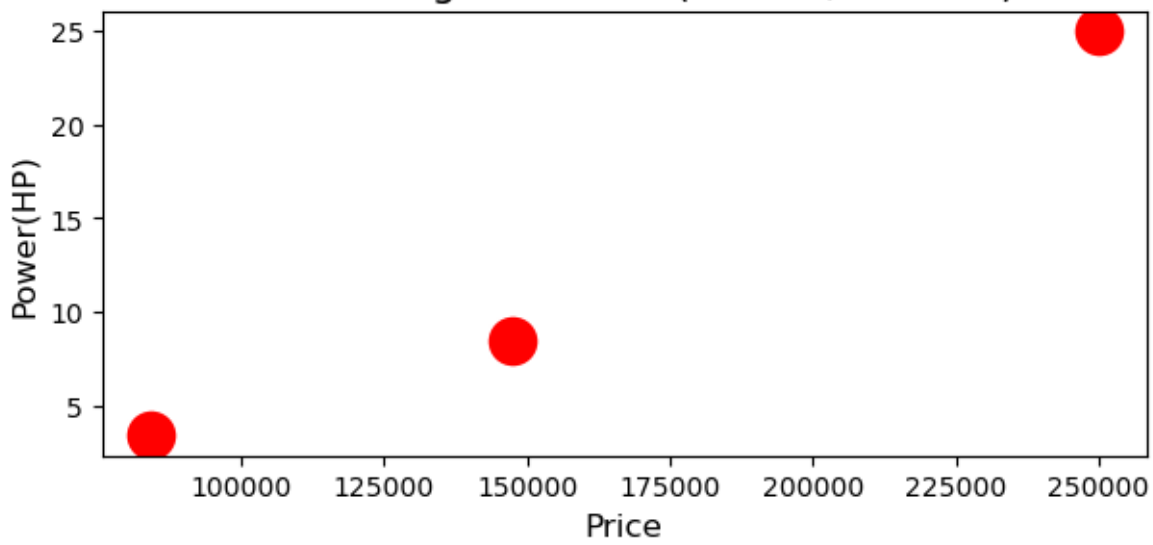
Attached is the Value Count for Each Segment.

| Cluster | count |
|---------|-------|
| 1 | 28 |
| 0 | 19 |
| 2 | 3 |

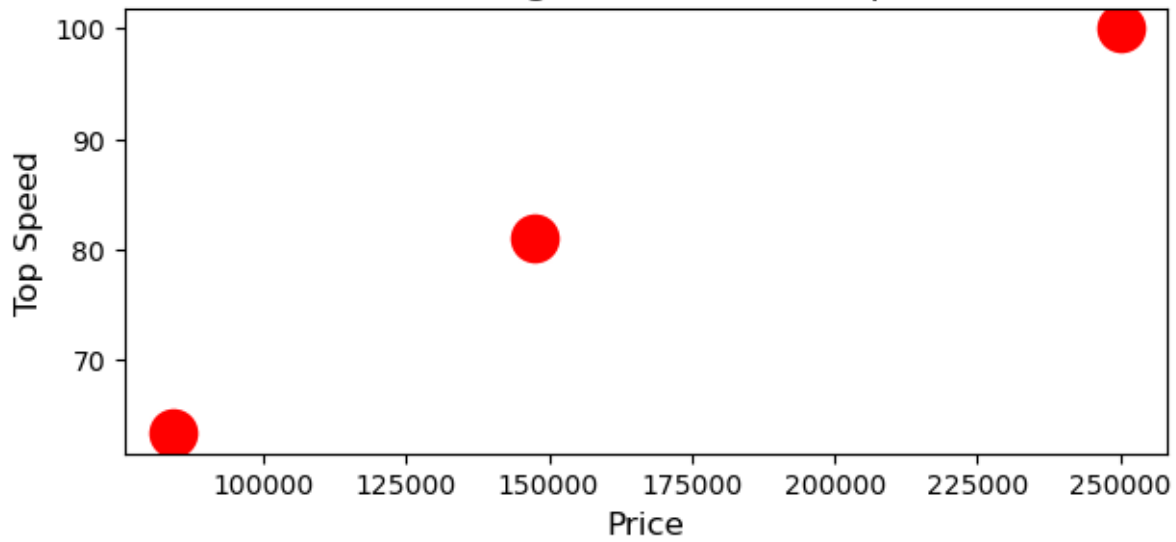
Below are Segments Extracted Based on Features.



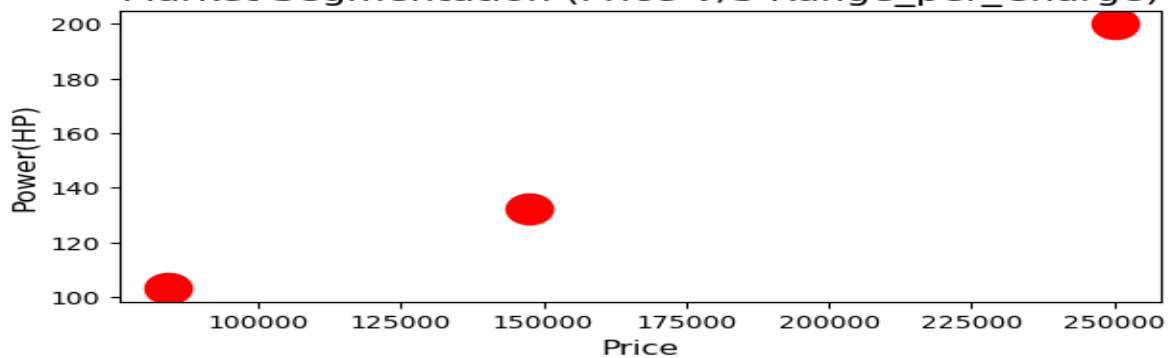
Market Segmentation (Price V/S Power)



Market Segment-Price VS Speed



Market Segmentation (Price V/S Range_per_Charge)



| Column | id | Model | Manufacturer | Vehicle Type | Battery Capacity (kWh) | Range per Charge (km) | Charging Time | Price | Power (HP or kW) | Top Speed (km/h) | Year of Manufacture | Cluster |
|--------|----|-----------------------|----------------------|--------------|------------------------|-----------------------|---------------|--------|------------------|------------------|---------------------|---------|
| 5 | 5 | Tork T6X | Tork Motors | Bike | 6.2 | 200 | 5 | 250000 | 25 | 100 | 2021 | 2 |
| 12 | 12 | Tork T6X | Tork Motors | Bike | 6.2 | 200 | 5 | 250000 | 25 | 100 | 2021 | 2 |
| 18 | 18 | Tork T6X | Tork Motors | Bike | 6.2 | 200 | 5 | 250000 | 25 | 100 | 2021 | 2 |
| 0 | 0 | Aura 300 Plus | Ather Energy | Scooter | 2.9 | 116 | 4.5 | 129000 | 6 | 80 | 2021 | 1 |
| 2 | 2 | Bajaj Chetak Electric | Bajaj Auto | Scooter | 4 | 95 | 5 | 150000 | 4 | 60 | 2020 | 1 |
| 6 | 6 | Revolt RV400 | Revolt Motors | Bike | 3.2 | 150 | 4 | 150000 | 20 | 85 | 2021 | 1 |
| 7 | 7 | Bajaj Chetak Electric | Bajaj Auto | Scooter | 4 | 95 | 5 | 150000 | 4 | 60 | 2020 | 1 |
| 8 | 8 | Ather 450X | Ather Energy | Scooter | 2.9 | 116 | 4.5 | 149000 | 6 | 80 | 2021 | 1 |
| 13 | 13 | Revolt RV400 | Revolt Motors | Bike | 3.2 | 150 | 4 | 150000 | 20 | 85 | 2021 | 1 |
| 14 | 14 | Ather 450X | Ather Energy | Scooter | 2.9 | 116 | 4.5 | 149000 | 6 | 80 | 2021 | 1 |
| 19 | 19 | Bajaj Chetak EV | Bajaj Auto | Scooter | 4 | 146 | 5.5 | 145000 | 4 | 80 | 2021 | 1 |
| 23 | 23 | Ather 450X | Ather Energy | Scooter | 2.9 | 116 | 4.5 | 149000 | 6 | 80 | 2021 | 1 |
| 26 | 26 | Revolt RV400 | Revolt Motors | Bike | 3.2 | 150 | 4 | 150000 | 20 | 85 | 2021 | 1 |
| 27 | 27 | Ather 450X | Ather Energy | Scooter | 2.9 | 116 | 4.5 | 149000 | 6 | 80 | 2021 | 1 |
| 32 | 32 | Bajaj Chetak EV | Bajaj Auto | Scooter | 4 | 146 | 5.5 | 145000 | 4 | 80 | 2021 | 1 |
| 34 | 34 | Ather 450X | Ather Energy | Scooter | 2.9 | 116 | 4.5 | 149000 | 6 | 80 | 2021 | 1 |
| 37 | 37 | Revolt RV400 | Revolt Motors | Bike | 3.2 | 150 | 4 | 155000 | 21 | 85 | 2022 | 1 |
| 38 | 38 | Ather 500X | Ather Energy | Scooter | 4 | 150 | 5 | 165000 | 7 | 90 | 2022 | 1 |
| 39 | 39 | Pure EV Epluto 8G | Pure EV | Scooter | 3.5 | 140 | 3.5 | 120000 | 5 | 85 | 2022 | 1 |
| 43 | 43 | Bajaj Chetak EV Plus | Bajaj Auto | Scooter | 4.5 | 160 | 6 | 160000 | 5 | 90 | 2022 | 1 |
| 45 | 45 | Ather 500X | Ather Energy | Scooter | 4 | 150 | 5 | 165000 | 7 | 90 | 2022 | 1 |
| 46 | 46 | Pure EV Epluto 8G | Pure EV | Scooter | 3.5 | 140 | 3.5 | 120000 | 5 | 85 | 2022 | 1 |
| 1 | 1 | Pure EV Epluto 7G | Pure EV | Scooter | 2.7 | 120 | 3 | 109000 | 5 | 80 | 2021 | 0 |
| 3 | 3 | Okinawa iPraise Pro | Okinawa Autotech | Scooter | 2.5 | 100 | 3 | 85000 | 3 | 60 | 2021 | 0 |
| 4 | 4 | Hero Electric Opto | Hero Motocorp | Scooter | 2.2 | 75 | 3 | 75000 | 3 | 60 | 2021 | 0 |
| 9 | 9 | Pure EV Epluto 7G | Pure EV | Scooter | 2.7 | 120 | 3 | 109000 | 5 | 80 | 2021 | 0 |
| 10 | 10 | Okinawa iPraise Pro | Okinawa Autotech | Scooter | 2.5 | 100 | 3 | 85000 | 3 | 60 | 2021 | 0 |
| 11 | 11 | Hero Electric Opto | Hero Motocorp | Scooter | 2.2 | 75 | 3 | 75000 | 3 | 60 | 2021 | 0 |
| 15 | 15 | Pure EV Epluto 7G | Pure EV | Scooter | 2.7 | 120 | 3 | 109000 | 5 | 80 | 2021 | 0 |
| 16 | 16 | Okinawa iPraise Pro | Okinawa Autotech | Scooter | 2.5 | 100 | 3 | 85000 | 3 | 60 | 2021 | 0 |
| 17 | 17 | Hero Electric Opto | Hero Motocorp | Scooter | 2.2 | 75 | 3 | 75000 | 3 | 60 | 2021 | 0 |
| 20 | 20 | Ampere Reo | Ampere Vehicles | Scooter | 3 | 100 | 3.5 | 80000 | 3 | 60 | 2021 | 0 |
| 21 | 21 | Urbanite X1 | Electric Vehicle Co. | Scooter | 2.5 | 100 | 3 | 75000 | 3 | 60 | 2021 | 0 |
| 22 | 22 | Joy e-Ride | Joy E-Bike | Bike | 2.2 | 80 | 2.5 | 60000 | 2 | 50 | 2021 | 0 |
| 24 | 24 | Pure EV Epluto 7G | Pure EV | Scooter | 2.7 | 120 | 3 | 109000 | 5 | 80 | 2021 | 0 |
| 25 | 25 | Okinawa iPraise Pro | Okinawa Autotech | Scooter | 2.5 | 100 | 3 | 85000 | 3 | 60 | 2021 | 0 |
| 28 | 28 | Pure EV Epluto 7G | Pure EV | Scooter | 2.7 | 120 | 3 | 109000 | 5 | 80 | 2021 | 0 |
| 29 | 29 | Okinawa iPraise Pro | Okinawa Autotech | Scooter | 2.5 | 100 | 3 | 85000 | 3 | 60 | 2021 | 0 |
| 30 | 30 | Joy E-Bike Urbanite | Electric Vehicle Co. | Bike | 2.2 | 80 | 2.5 | 60000 | 2 | 50 | 2021 | 0 |
| 31 | 31 | Ampere Reo | Ampere Vehicles | Scooter | 3 | 100 | 3.5 | 80000 | 3 | 60 | 2021 | 0 |
| 33 | 33 | Urbanite X1 | Electric Vehicle Co. | Scooter | 2.5 | 100 | 3 | 75000 | 3 | 60 | 2021 | 0 |
| 35 | 35 | Pure EV Epluto 7G | Pure EV | Scooter | 2.7 | 120 | 3 | 109000 | 5 | 80 | 2021 | 0 |
| 36 | 36 | Okinawa iPraise Pro | Okinawa Autotech | Scooter | 2.5 | 100 | 3 | 85000 | 3 | 60 | 2021 | 0 |
| 40 | 40 | Okinawa iPraise Pro | Okinawa Autotech | Scooter | 3 | 120 | 3 | 95000 | 4 | 70 | 2022 | 0 |
| 41 | 41 | Joy E-Bike Urbanite | Electric Vehicle Co. | Bike | 2.8 | 90 | 2.5 | 70000 | 3 | 60 | 2022 | 0 |
| 42 | 42 | Ampere Zeo | Ampere Vehicles | Scooter | 3.5 | 120 | 4 | 90000 | 4 | 70 | 2022 | 0 |
| 44 | 44 | Urbanite X2 | Electric Vehicle Co. | Scooter | 3.2 | 130 | 4 | 110000 | 5 | 80 | 2022 | 0 |
| 47 | 47 | Okinawa iPraise Pro | Okinawa Autotech | Scooter | 3 | 120 | 3 | 95000 | 4 | 70 | 2022 | 0 |
| 48 | 48 | Joy E-Bike Urbanite | Electric Vehicle Co. | Bike | 2.9 | 85 | 2.5 | 75000 | 3 | 65 | 2022 | 0 |
| 49 | 49 | Ampere Zeo | Ampere Vehicles | Scooter | 3.5 | 120 | 0 | 0 | 0 | 0 | 0 | 0 |

Cluster for segmentation of Two Wheeler Data Set

Conclusion.

Following Conclusions are obtained from the above Analysis.

State Wise EV Distribution Data Set

A)Two Wheelers and Three Wheelers contribute to 99 percent of Market segment and there is a Huge scope for expansion in all other segments.

B)There is a Huge Potential for Development of EV Market in Northeast India as North east has poor Contribution to EV'S from Above Data Set

3)3 Wheelers operate in few Regions North East India and North India.

4)Huge Potential for Heavy Vehicles and Buses provided Conditions are studied carefully and designed for Indian conditions and weather

5)EV Market Segmentation Based on geography effectively segments market and suitable strategies for each segment can be taken based on the Segment observations.

Customer Behaviour Data Set.

A)Most of the Customers are in the Range of 30 to 40 years and Significantly contribute for Driving 4 Wheeler Market(Car Sales)

B)65 Percent of customers are of salaried. Hence Corporate Employees needs to be targeted having a fixed salary and stable Income

C)There is a relation Between Wife Salary and Price of Car Purchased. Hence luxury Brands can target working couples

D)Further referring to Clusters help significantly in identifying segments. And trends of each segment.

E)Based on Data in each segment specific groups can be targeted.

Electric Two Wheeler Data Set

A) There is scope for Bike Segment in india as Majority of Two Wheelers are Scooters.

B) From Market Segment Analysis for Vehicle segment we observed that 59% of all EV'S were Two Wheelers. Hence Demand for Two wheelers seemed high. Further Analysis of current Market demand, customer preferences etc can be further analyzed to check feasibility to capture this Market.

c)Lot of technicalities need to be analysed to meet or exceed current market demands for technicality. As observed as technical features improve cost also increases. Cost reduction for same features may increase demand for vehicle in specific segment

Scope for Further Improvements.

A)It is not clearly understood as to why in few states EV Market is not well developed or not developed in specific segment. Analysis needs to be carried out to check constrains for the same.

B)Public Transport seems less developed in EV Segment. Further analysis needs to be made on the same with additional data and analysis.

C)Two wheeler Data set is limited in terms of number of vehicles. Need larger Data set for analysis

Libraries and Algorithms Used in Market Segmentation.

A)Pandas for Data loading, cleaning and Transformation

B)Matplotlib for Data Visualisation

c)Pandas for EDA

D)K-Means Clustering for Segmentation.