

Feynn Labs: Project 2

Market Segment Analysis of EV Market

Abhay Swarnkar

GitHub Link: https://github.com/GENRATECODE/Feynn_lab_real_world_project.git

Electric Vehicle Market in India (Market Segmentation)



Overview:

An **electric vehicle (EV)** is a vehicle that uses one or more electric motors for propulsion. It can be powered by a collector system, with electricity from extravehicular sources, or it can be powered autonomously by a battery (sometimes charged by solar panels, or by converting fuel to electricity using fuel cells or a generator).[1] EVs include, but are not limited to, road and rail vehicles, surface and underwater vessels, electric aircraft and electric spacecraft. EVs first came into existence in the mid-19th century, when electricity was among the preferred methods for motor vehicle propulsion, providing a level of comfort and ease of operation that could not be achieved by the gasoline cars of the time. Internal combustion engines were the dominant propulsion method for cars and trucks for about 100 years, but electric power remained commonplace in other vehicle types, such as trains and smaller vehicles of all types. In the 21st century, EVs have seen a resurgence due to technological developments, and an increased focus on renewable energy and the potential reduction of transportation's impact on climate change and other environmental issues. Project Drawdown describes electric vehicles as one of the 100 best contemporary solutions for addressing climate change. The electric vehicles industry at a nascent stage in India. It is less than 1% of the total vehicle sales however has the potential to grow to more than 5% in a few years. At present there are more than 5 lac electric two-wheelers and few thousand electric cars on Indian roads. The industry volumes have been fluctuating, mostly depending on the incentives offered by the government. Many serious players (Hero Eco, Ather, Avon, Lohia, Ampere, etc) are continuing with the mission and trying to enforce the positive change. More than 90% of electric vehicles on Indian roads are low-speed electric scooters (less than 25km/hr) that do not require registration and licenses. Almost all electric scooters run on lead batteries to keep the prices low, however, battery failures and low life of batteries have become major limiting factors for sales besides government subsidies. Many manufacturers have taken initiatives to install the charging station with limited success. The industry is almost ready for take-off but for the incentives. It is expected that with FAME-2 the industry may witness a quantum leap in volumes and technology. Market segmentation becomes a crucial tool for evolving transportation technology such as electric vehicles (EVs) in emerging markets to explore and implement for extensive adoption. EVs adoption is expected to grow phenomenally in near future as low emission and low operating cost vehicle, and thus it drives. The main aim of this study is to explore and identify distinct sets of potential buyer segments for EVs based on psychographic, Behavioral, and distinct sets of potential buyer segments for EV's based on psychographic Behavioral, and socio-economic characterization by employing an integrated research framework of 'perceived benefits-attitude-intention'. The study applied robust analytical procedures including cluster analysis, multiple discriminant analysis and chi-square test to operationalize and validate segments from the data collected of 563 respondents using a cross-sectional online survey. The finding post the three distinct sets of young consumer groups have been identified and labelled as conservatives', 'indifferent' , and 'enthusiasts' which are deemed to be budding EV buyers. The implications are recommended, which may offer some pertinent guidance for scholars and policy maker to encourage EVs adoption in the backdrop of emerging sustainable transport market.

MARKET SEMENTATION

Target Market:

The target market of Electric Vehicle Market Segmentation can be categorized into Geographic, Sociodemographic, Behavioral, and Psychographic Segmentation.

Behavioral Segmentation: Behavioral Segmentation: searches directly for similarities in Behavioral or reported Behavioral. Example: prior experience with the product, amount spent on the purchase etc



Fig 1. Behavioral Segmentation

Advantage: uses the very Behavioral of interest is used as the basis of segment extraction.

Disadvantage: not always readily available.

Psychographic Segmentation: Grouped based on beliefs, interests, preferences, aspirations, or benefits sought when purchasing a product. Suitable for lifestyle segmentation. Involves many segmentation variables.

Advantage: generally, more reflective of the underlying reasons for differences in consumer Behavioral.

Disadvantage: increased complexity of determining segment memberships for consumers.



Fig 2 Psychographic Segmentation

Socio-Demographic Segmentation: Includes age, gender, income and education.
Useful in industries



Fig.3 Demographic Marketing

Advantage: segment membership can easily be determined for every customer.

Disadvantage: if this criterion is not the cause for customers product preferences then it does not provide sufficient market insight for optimal segmentation decisions.

Data Collection: -

The data has been collected manually, and help of instructor the sources used for this process are listed below:

- <https://www.kaggle.com/datasets/geoffnel/evs-one-electric-vehicle-dataset>
- <https://www.kaggle.com/datasets>
- <https://data.worldbank.org/>
- https://drive.google.com/drive/folders/137KIMhwpB1bx5zx0hTaa486bEKe3kXaB?usp=share_link

Segmenting for Electric Vehicle Market:

The market segmentation approach aims at defining actionable, manageable, homogenous subgroups of individual customers to whom the marketers can target with a similar set of marketing strategies. In practice, there are two ways of segmenting the market-a-priori and post-hoc. An a-priori approach utilizes predefined characteristics such as age, gender, income, education, etc. to predefine the segments followed by profiling based on a host of measured variables (Behavioral, psychographic or benefit). In the post-hoc approach to segmentation on other hand, the segments are identified based on the relationship among the multiple measured variables. The commonality between both approaches lies in the fact that the measured variables determine the ‘segmentation theme’. The present study utilizes an a-priori approach to segmentation so as to divide the potential EV customers into sub-groups.

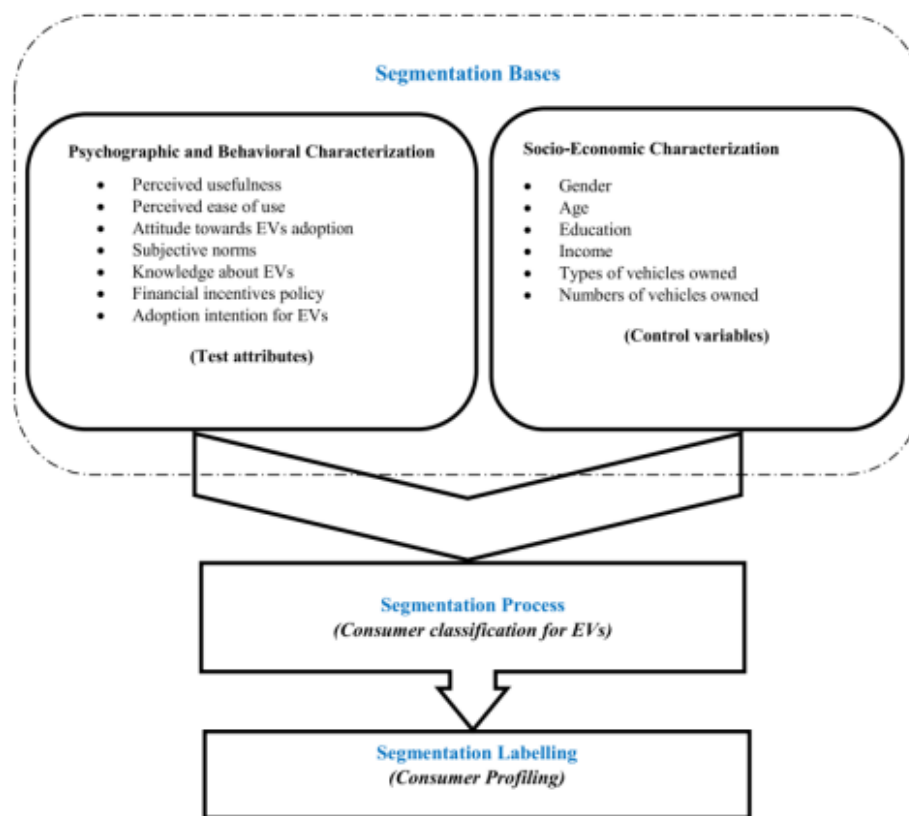


Fig.4 Market Segmentation Electric Vehicles

It is argued that the blended approach of psychographic and socioeconomic attributes for market segmentation enables the formulation of sub-market strategies which in turn satisfy the specific tastes and preferences of the consumer groups. Straughan and Roberts presented a comparison between the usefulness of psychographic, demographic, and economic characteristics based on consumer evaluation for eco-friendly products. Feynn Labs They pinpointed the perceived superiority of the psychographic characteristics over the socio-demographic and economic ones in explaining the environmentally-conscious consumer Behavioral and thus, the study recommended the use of psychographic characteristics in profiling the consumer segments in the market for eco-friendly products. The present study adds perceived-benefit characteristics guided by blended psychographic and socio-economic aspects for segmenting the consumer market.

IMPLEMENTATION

Packages/Tools used:

1. **NumPy** : To Calculate Various calculation related to arrays.
2. **Pandas** : To read or Load and manipulation data datasets.
3. **Matplotlib or Seaborn**: To Plot and all visualization of datasets.
4. **Scikit-learn**: We have used LabelEncoder() to encode our values and used K-mean() unsupervised method.
5. **Python** : Python Language used to write all code in that Language.

```
import numpy as np
import pandas as pd
import matplotlib as plt
import seaborn as sns
%matplotlib inline
```

•[17]:

```
df=pd.read_csv("data/data.csv")
```

[18]:

Fig.5. Library import and Dataset load in Jupyter

Data-Preprocessing:

The Data collected is compact and is partly used for visualization purposed and partly for clustering. The datasets does not have any null value or empty value cell in dataset so we must first pre-processed dataset check dataset does not have empty cell or null value.

```
df.isnull()
It is use for check total number of null value each columns
```

[20]:

```
df.isnull().sum()
```

[20]:

Brand	0
Model	0
AccelSec	0
TopSpeed_KmH	0
Range_Km	0
Efficiency_WhKm	0
FastCharge_KmH	0
RapidCharge	0
PowerTrain	0
PlugType	0
BodyStyle	0
Segment	0
Seats	0
PriceEuro	0
dtype:	int64

Fig. 6. Check total null value

`df.info()` tell about dataset contain null value or not and also tell about data type of each cell

[14]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 103 entries, 0 to 102
Data columns (total 15 columns):
#   Column                Non-Null Count  Dtype  
---  -
0   Unnamed: 0            103 non-null   int64  
1   Brand                  103 non-null   object  
2   Model                  103 non-null   object  
3   AccelSec               103 non-null   float64 
4   TopSpeed_KmH          103 non-null   int64  
5   Range_Km              103 non-null   int64  
6   Efficiency_WhKm        103 non-null   int64  
7   FastCharge_KmH        103 non-null   int64  
8   RapidCharge            103 non-null   object  
9   PowerTrain            103 non-null   object  
10  PlugType               103 non-null   object  
11  BodyStyle              103 non-null   object  
12  Segment                103 non-null   object  
13  Seats                  103 non-null   int64  
14  PriceEuro              103 non-null   int64  
dtypes: float64(1), int64(7), object(7)
memory usage: 12.2+ KB
```

Fig .7. Dataset contain null value or datatype of column

```
df.head()
```

[21]:

	Brand	Model	AccelSec	TopSpeed_KmH	Range_Km	Efficiency_WhKm	FastCharge_KmH	Rapid
0	Tesla	Model 3 Long Range Dual Motor	4.6	233	450	161	940	
1	Volkswagen	ID.3 Pure	10.0	160	270	167	250	
2	Polestar	2	4.7	210	400	181	620	
3	BMW	iX3	6.8	180	360	206	560	
4	Honda	e	9.5	145	170	168	190	

Fig 8. Dataset Print using Head() function

EDA(EXPLORING DATA)

First we explore the key characteristic of the data set by loading the set and inspecting basic feature such as the variables names, the sample size, and the first three rows of the data. We start the Exploratory Data Analysis with some data Analysis drawn from the data without Principal Component analysis and with some Principle Component Analysis in the dataset obtained from the combination of all the data we have. PCA is a statistical process that converts the observations of correlated features into a set of linearly uncorrelated features with the help of orthogonal transformation. These new transformed features are called the Principal Components. The process helps in reducing dimensions of the data to make the process of classification/regression or any form of machine learning, cost-effective.

Comparison of cars in our data

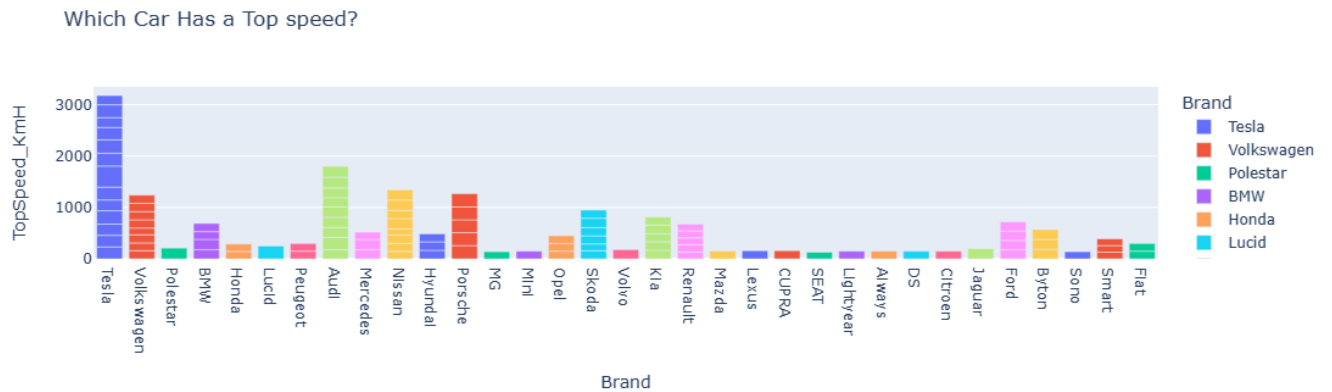


Fig.9. which car has top speed?

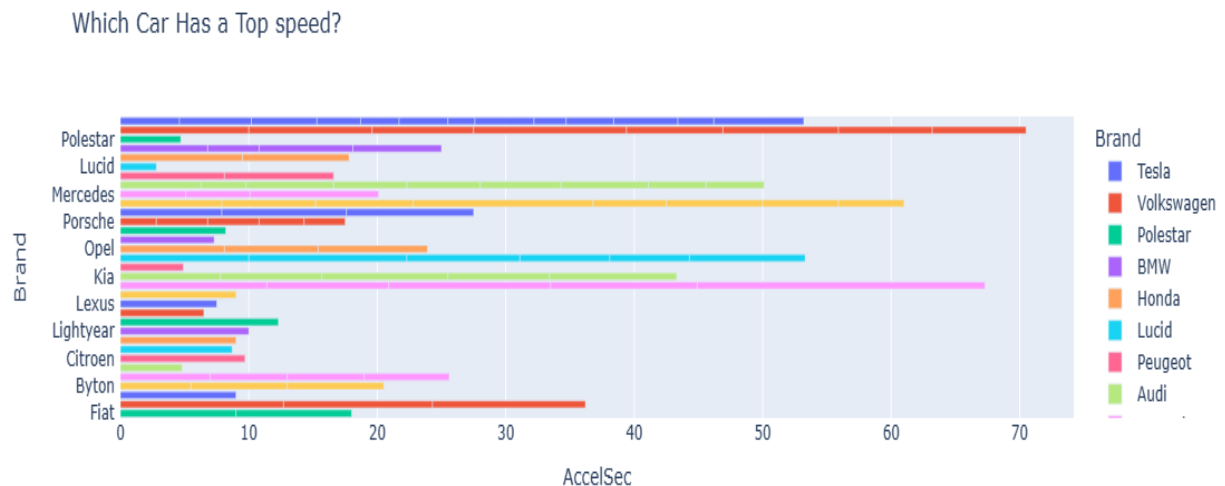
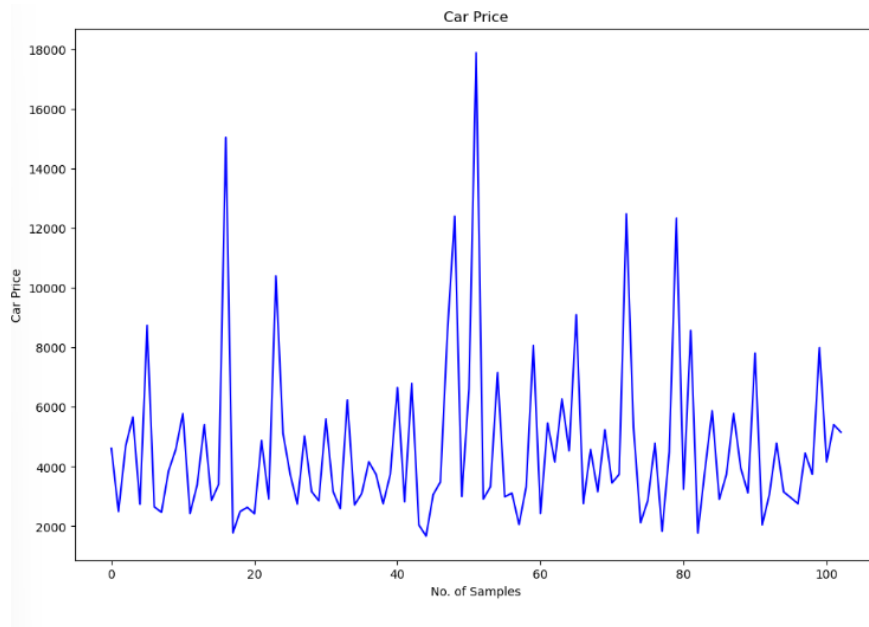
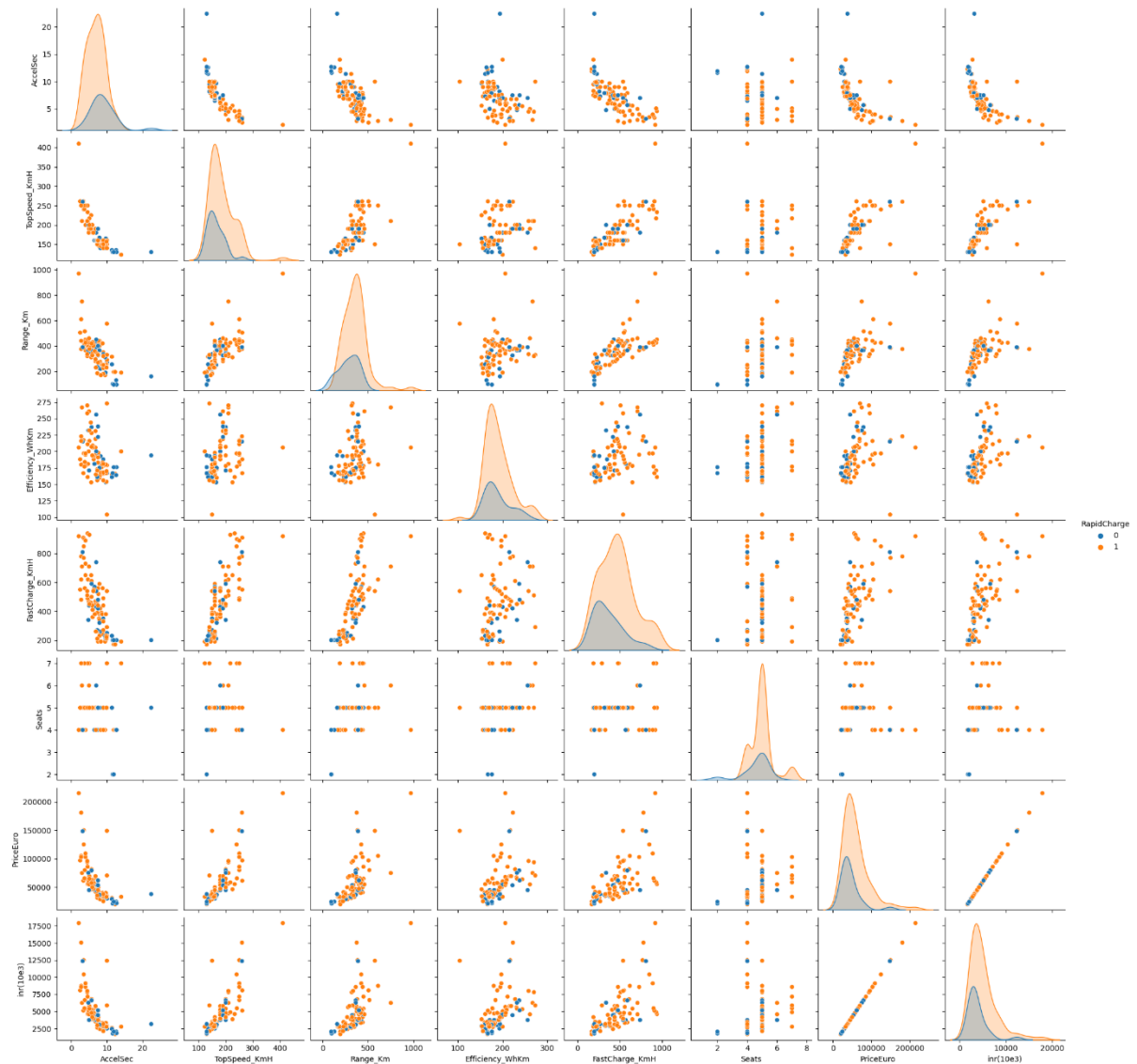


Fig.10. Which car has a Top acceleration



For Electric Vehicle Market one of the most important key is Charging:



Correlation Matrix:

A correlation matrix is simply a table that displays the correlation. It is best used in variables that demonstrate a linear relationship between each other. Coefficients for different variables. The matrix depicts the correlation between all the possible pairs of values through the heatmap in the below figure. The relationship between two variables is usually considered strong when their correlation coefficient value is larger than 0.7

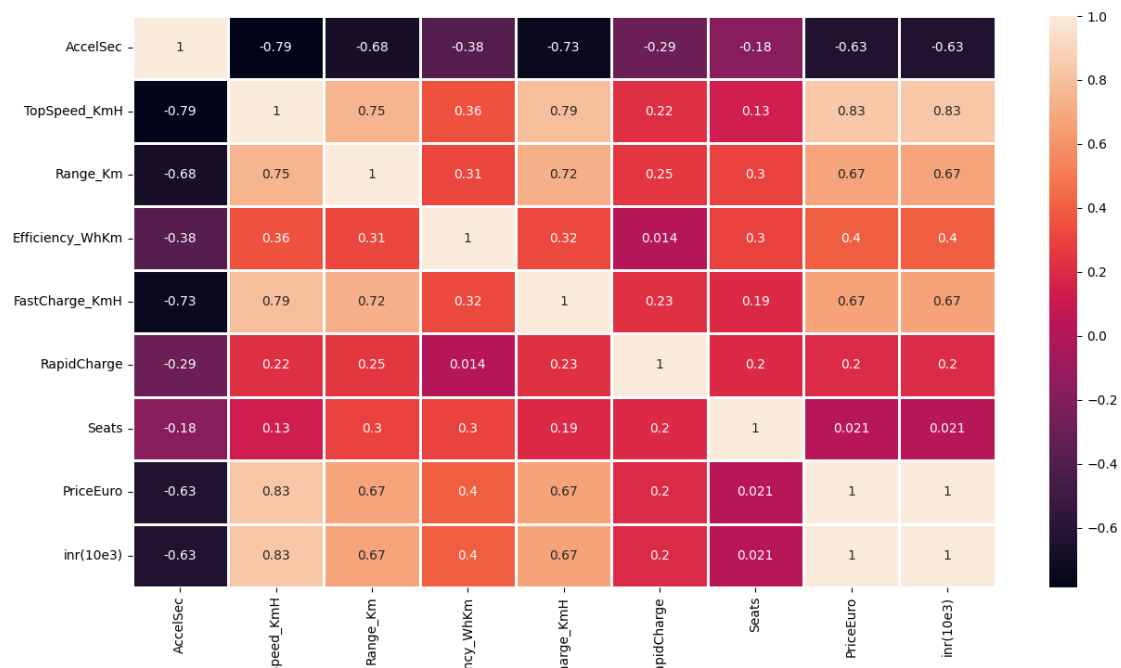


Fig.11. Correlation Matrix

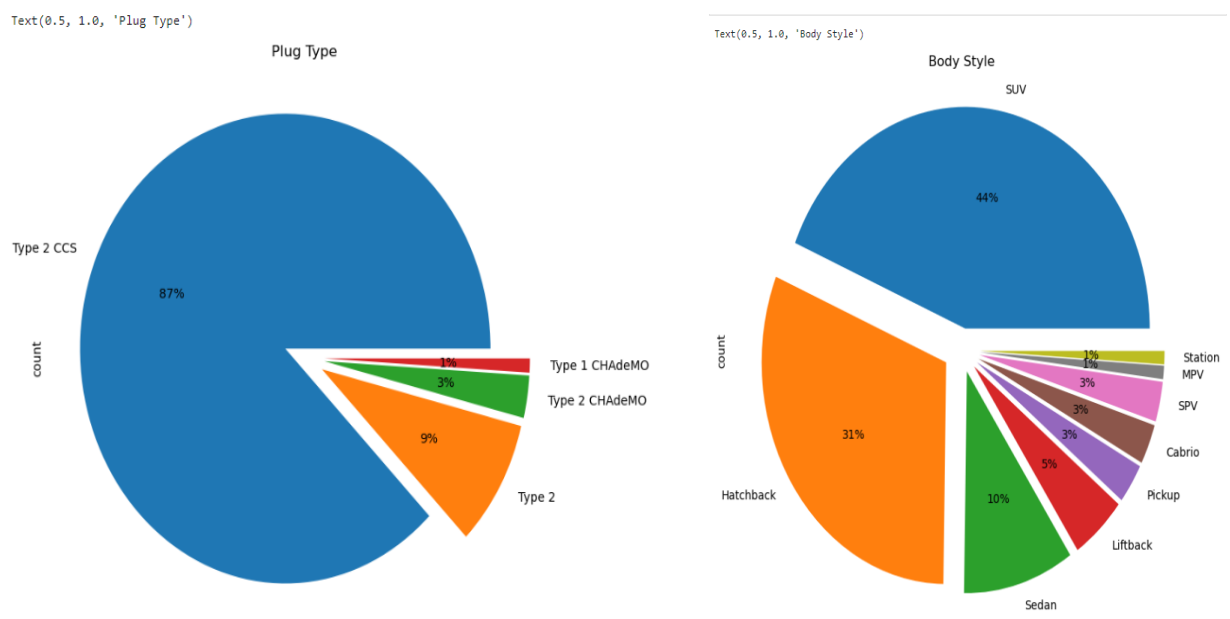


Fig.12. Pie Char According Plug Type and Body Style.

Text(0.5, 1.0, 'Segment')

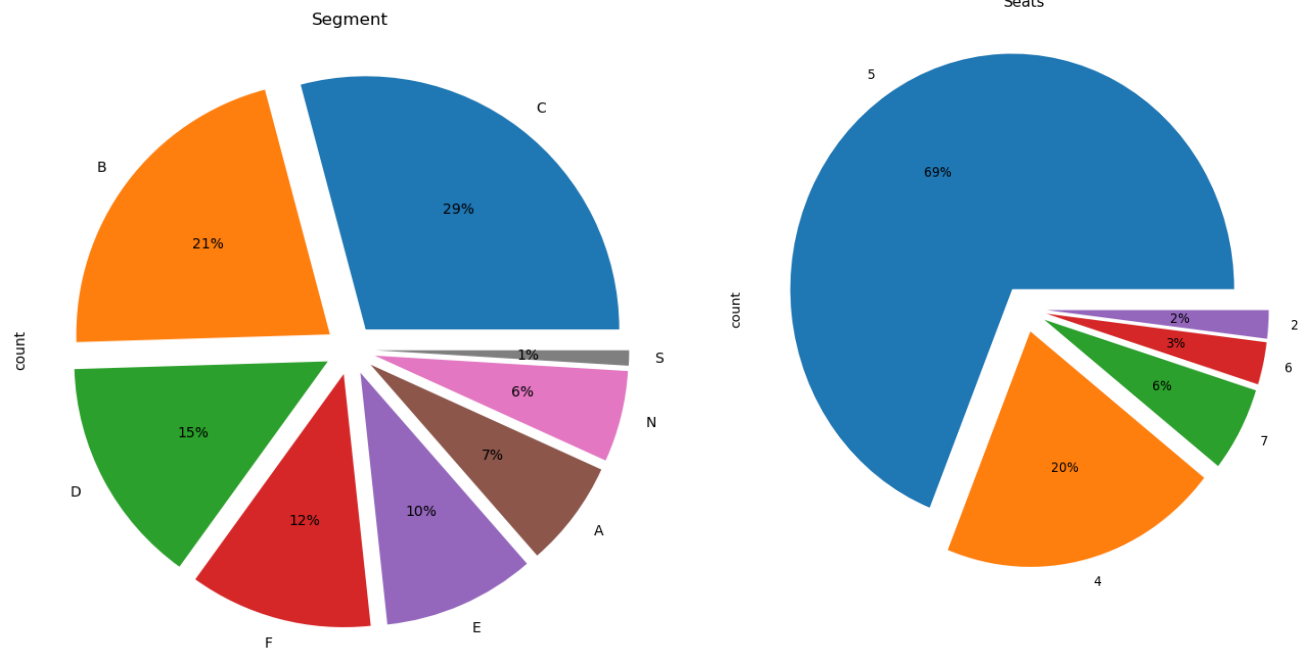


Fig.13. Pie Chart Seat and Segment

Now we can see that the requirements of what type of cars are most need for customer's and from the past 10 year there is a rapid growth of Electric Vehicles usages in India

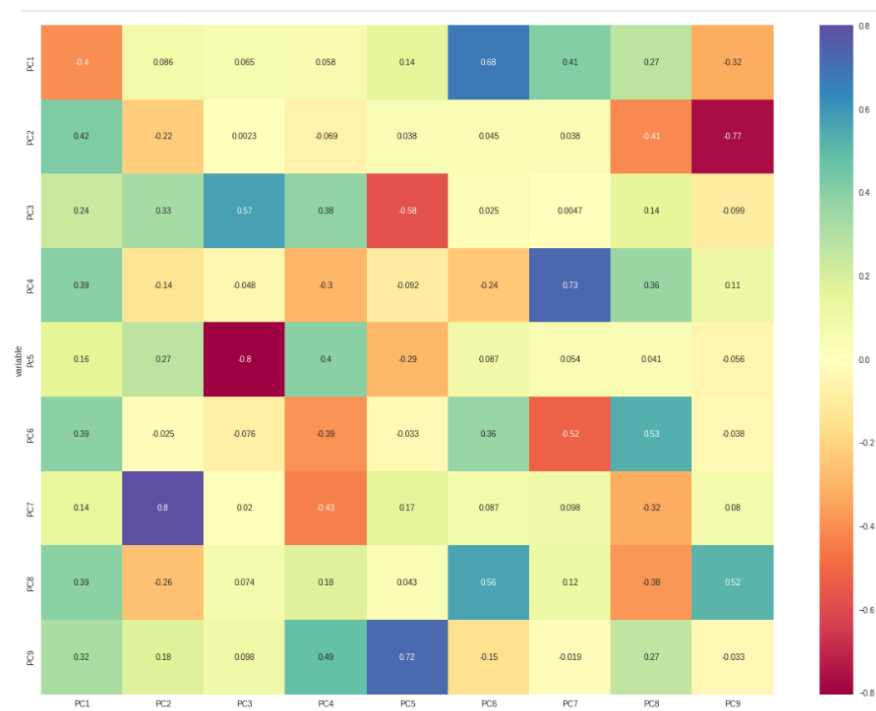


Fig.14. Correlation matrix plot for loading

Scree Plot: is a common method for determining the number of PCs to be retained via graphical representation. It is a simple line segment plot that shows the eigenvalues for each individual PC. It shows the eigenvalues on the y-axis and the number of factors on the x-axis. It always displays a downward curve. Most scree plots look broadly similar in shape, starting high on the left, falling rather quickly, and then flattening out at some point. This is because the first component usually explains much of the variability, the next few components explain a moderate amount, and the latter components only explain a small fraction of the overall variability. The scree plot criterion looks for the “elbow” in the curve and selects all components just before the line flattens out. The proportion of variance plot: The selected PCs should be able to describe at least 80% of the variance.

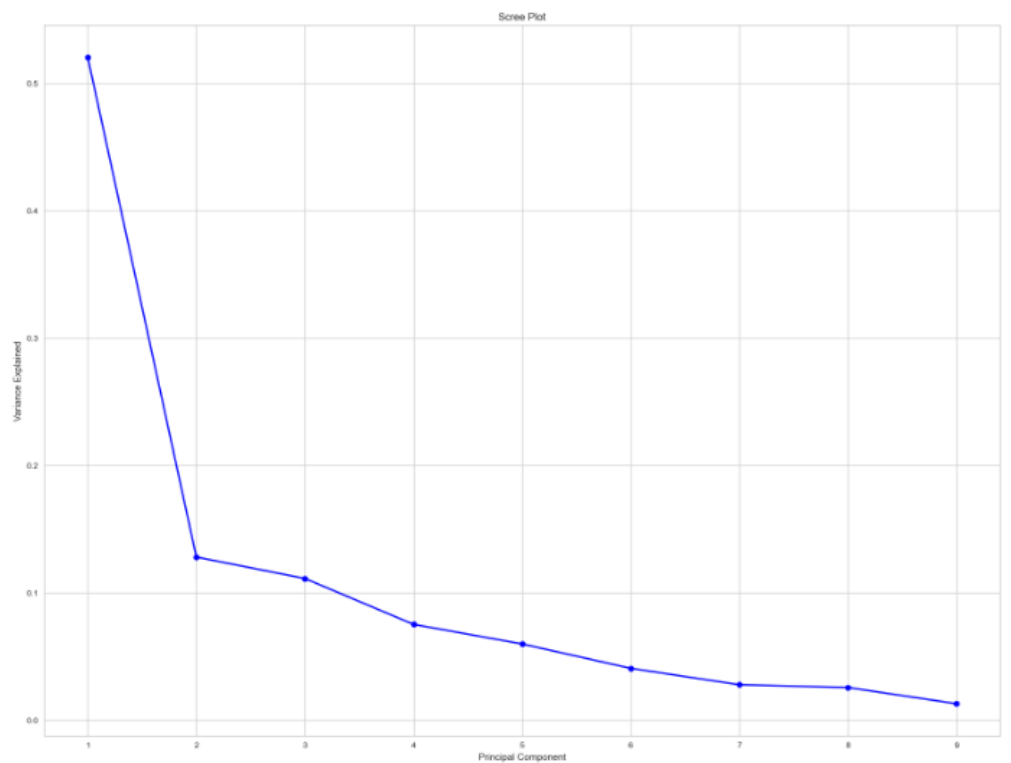


Fig.15. : Screen Plot for our dataset

Extracting Segments

Dendrogram This technique is specific to the agglomerative hierarchical method of clustering. The agglomerative hierarchical method of clustering starts by considering each point as a separate cluster and starts joining points to clusters in a hierarchical fashion based on their distances. To get the optimal number of clusters for hierarchical clustering, we make use of a dendrogram which is a tree-like chart that shows the sequences of merges or splits of clusters. If two clusters are merged, the dendrogram will join them in a graph and the height of the join will be the distance between those clusters. As shown in Figure, we can chose the optimal number of clusters based on hierarchical structure of the dendrogram. As highlighted by other cluster validation metrics, four to five clusters can be considered for the agglomerative hierarchical as well.

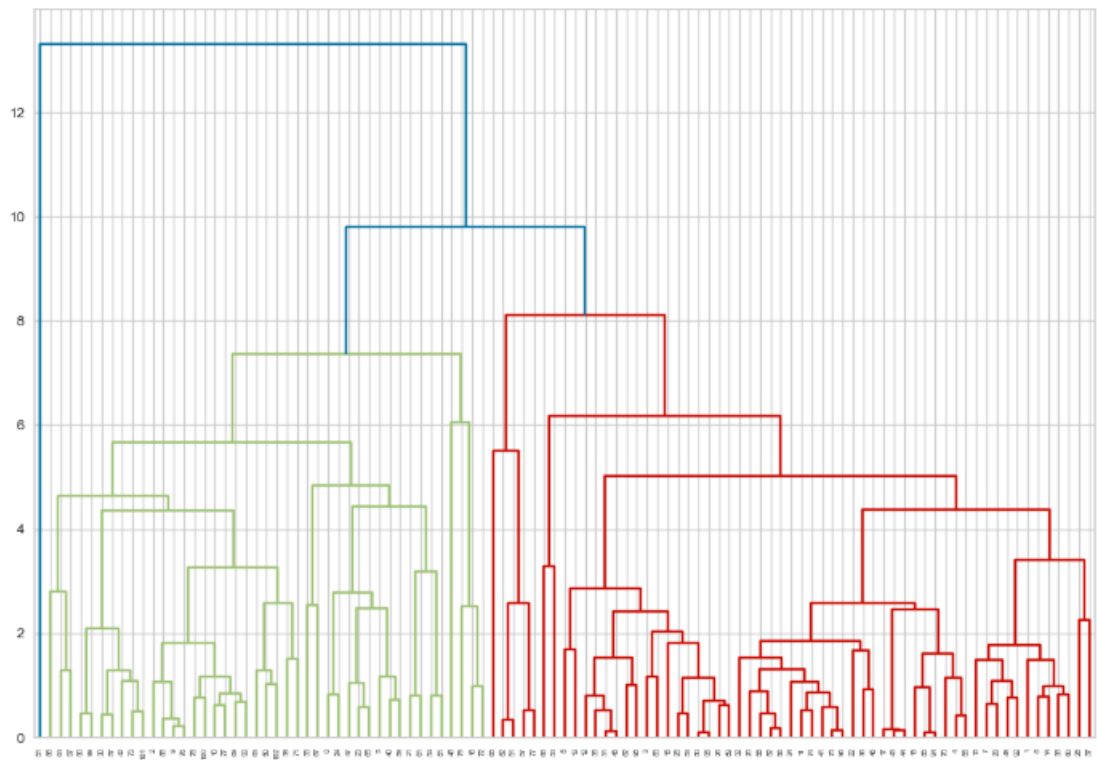


Fig.16. Dendrogram Plot for our Dataset

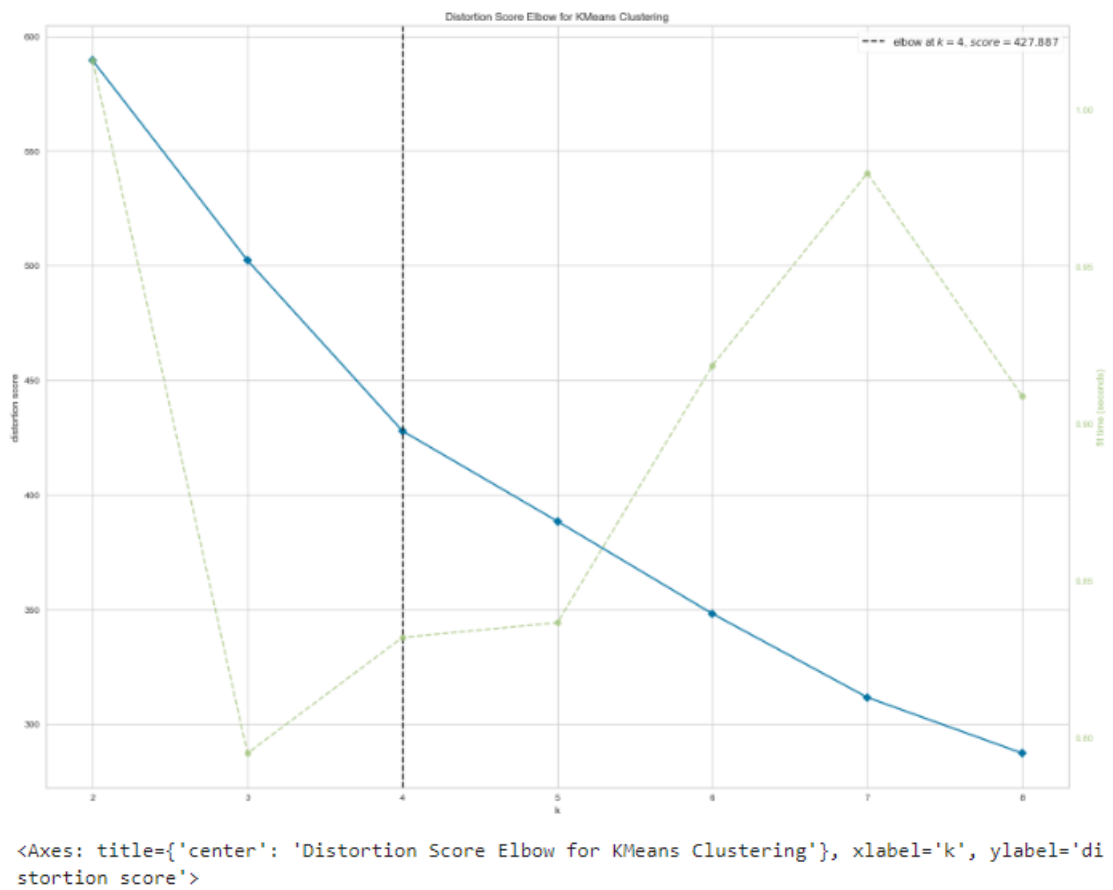


Fig.17. Evaluating The Cluster using Distortion

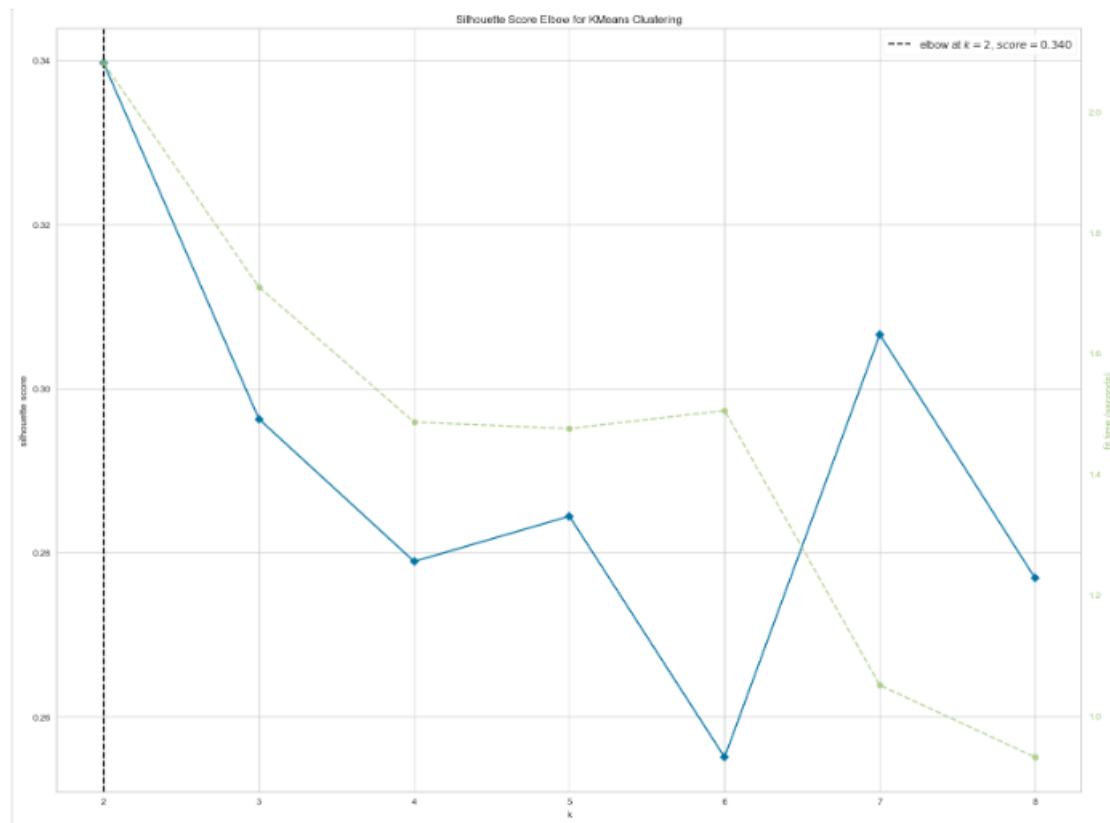


Fig.18. Evaluating the clusters using silhouette

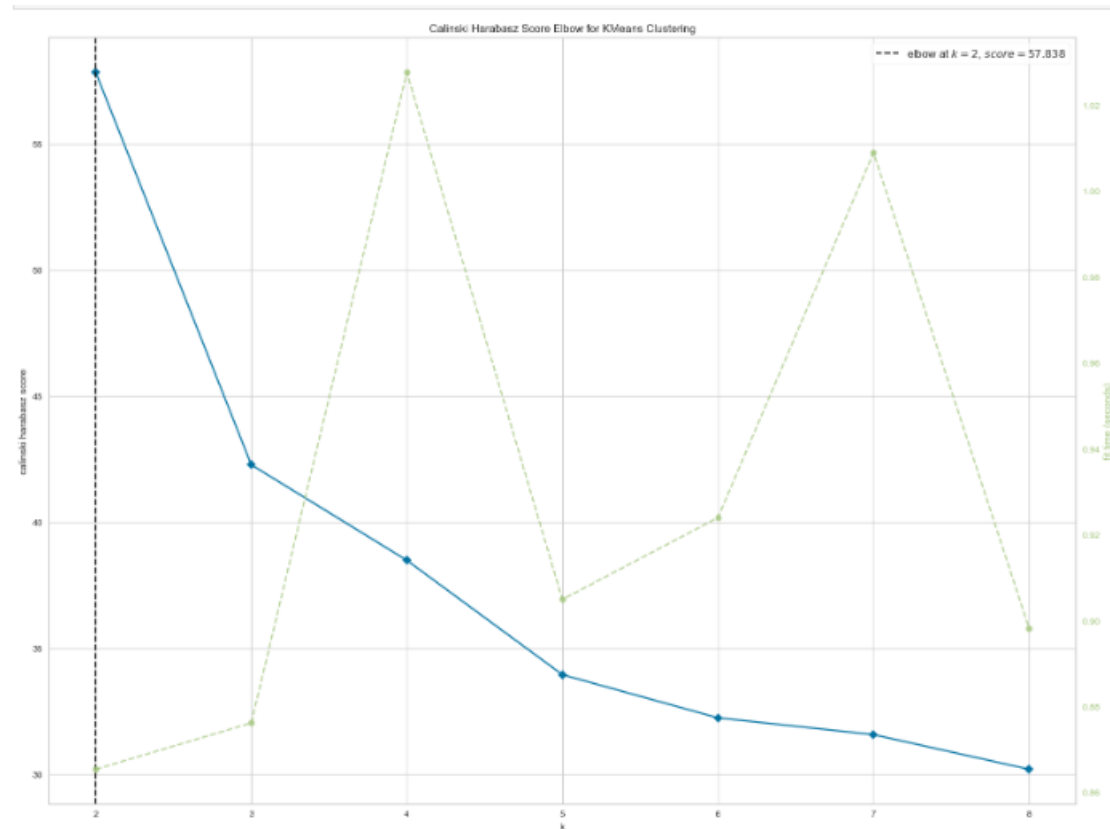


Fig.19. Evaluating the clusters using Calinski Harabasz

ANALYSIS AND APPROACHES USED FOR SEGMENTATION

Clustering

Clustering is one of the most common exploratory data analysis techniques used to get an intuition about the structure of the data. It can be defined as the task of identifying subgroups in the data such that data points in the same subgroup (cluster) are very similar while data points in different clusters are very different. In other words, we try to find homogeneous subgroups within the data such that data points in each cluster are as similar as possible according to a similarity measure such as Euclidean based distance or correlation-based distance. The decision of which similarity measure to use is application-specific. Clustering analysis can be done on the basis of features where we try to find subgroups of samples based on features or on the basis of samples where we try to find subgroups of features based on samples.

K-Means Algorithm

K Means algorithm is an iterative algorithm that tries to partition the dataset into pre-defined distinct non-overlapping subgroups (clusters) where each data point belongs to only one group. It tries to make the intra-cluster data points as similar as possible while also keeping the clusters as different (far) as possible. It assigns data points to a cluster such that the sum of the squared distance between the data points and the cluster's centroid (arithmetic mean of all the data points that belong to that cluster) is at the minimum. The less variation we have within clusters, the more homogeneous (similar) the data points are within the same cluster.

The way k means algorithm works is as follows:

- Specify number of clusters K.
- Initialize centroids by first shuffling the dataset and then randomly selecting K data points for the centroids without replacement.
- Keep iterating until there is no change to the centroids. i.e assignment of data points to clusters isn't changing.

The approach k-means follows to solve the problem is expectation maximization The E-step is assigning the data points to the closest cluster. The M-step is computing the centroid of each cluster. Below is a breakdown of how we can solve it mathematically

The Objective function is

$$J = \sum_{i=0}^m \sum_{k=1}^K W_{ik} \|x^i - \mu_k\|^2$$

And M-Step is

$$\frac{\partial J}{\partial \mu_k} = 2 \sum_{i=1}^m w_{ik} (x^i - \mu_k) = 0$$

$$\Rightarrow \mu_k = \frac{\sum_{i=1}^m W_{ik} x^i}{\sum_{i=1}^m W_{ik}}$$

Applications

K means algorithm is very popular and used in a variety of applications such as market segmentation, document clustering, image segmentation and image compression, etc. The goal usually when we undergo a cluster analysis is either:

1. Get a meaningful intuition of the structure of the data we're dealing with.
2. Cluster-then-predict where different models will be built for different subgroups if we believe there is a wide variation in the Behavioral of different subgroups.

The k-means clustering algorithm performs the following tasks:

- Specify number of clusters K
- Initialize centroids by first shuffling the dataset and then randomly selecting K data points for the centroids without replacement.
- Compute the sum of the squared distance between data points and all centroids.
- Assign each data point to the closest cluster (centroid).
- Compute the centroids for the clusters by taking the average of the all-data points that belong to each cluster.
- Keep iterating until there is no change to the centroids. i.e. assignment of data points to clusters isn't changing. According to the Elbow method, here we take K=4 clusters to train K-Means model. The derived clusters are shown in the following figure

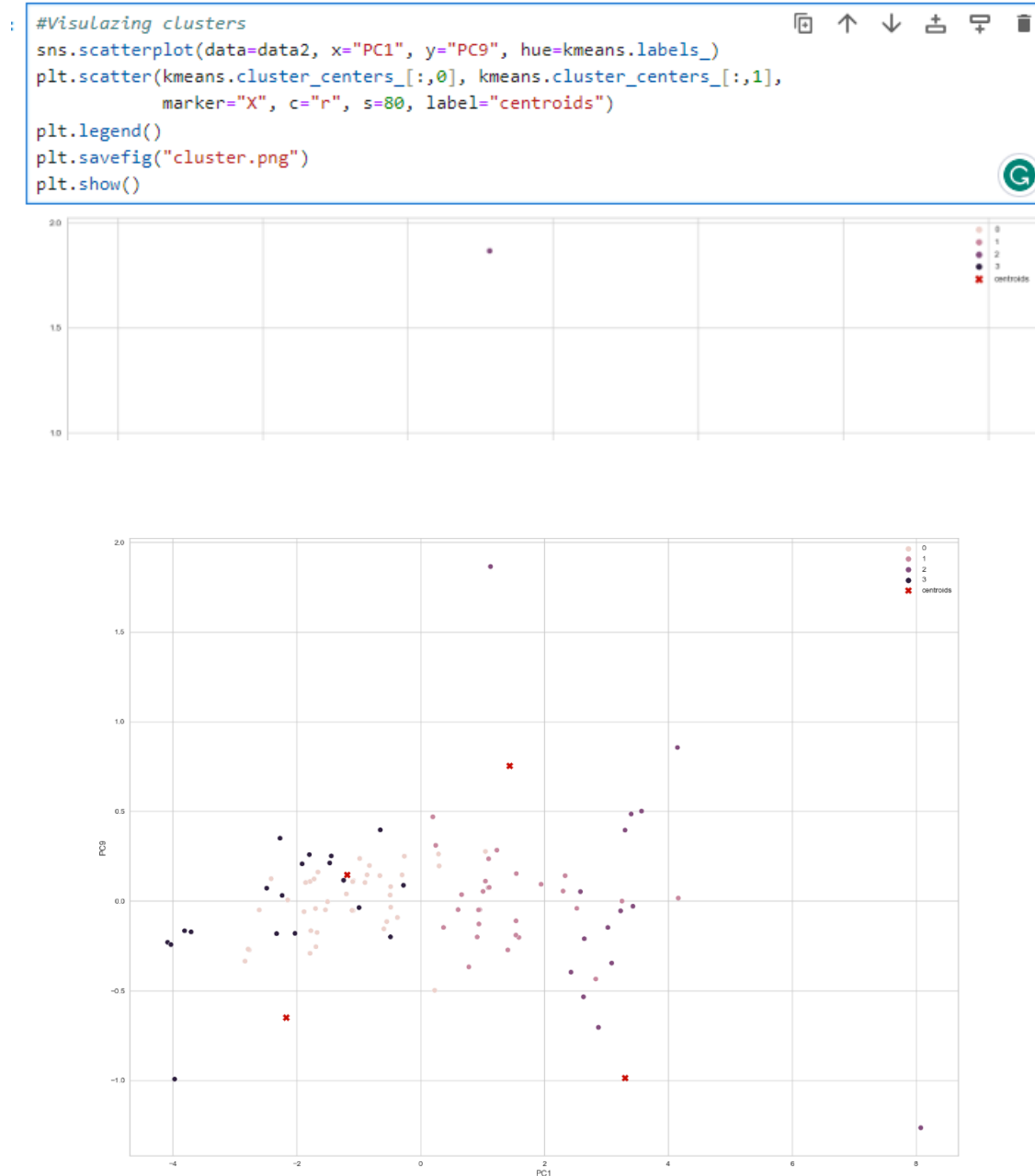


Fig.20. Visualize Cluster using K-mean

Prediction of Prices most used cars

Linear regression is a machine learning algorithm based on supervised learning. It performs a regression task. Regression models target prediction value based on independent variables. It is mostly used for finding out the relationship between variables and forecasting. Here we use a linear regression model to predict the prices of different Electric cars in different companies. X contains the independent variables and y is the dependent Prices that is to be predicted. We train our model with a splitting of data into a 4:6 ratio, i.e. 40% of the data is used to train the model.

LinearRegression().fit(Xtrain,ytrain) command is used to fit the data set into model. The values of intercept, coefficient, and cumulative distribution function (CDF) are described in the figure.

Regression for data2

```
02]: X=data2[['PC1', 'PC2','PC3','PC4','Pc5','PC6', 'PC7','PC8','PC9']]
      y=df['inr(10e3)']
```

```
03]: X_train, X_test, y_train, y_test = train_test_split(X, y,test_size=0.4, random_state=101)
      lm=LinearRegression().fit(X_train,y_train)
```

```
04]: print(lm.intercept_)

4643.522050485438
```

```
05]: lm.coef_
```

```
05]: array([ 1101.5872075 , -741.20904198,   208.53617452,   508.32245827,
            122.35330123,  1579.00685826,   333.61147115, -1079.99511501,
            1461.7226913  ])
```

```
06]: X_train.columns
```

```
06]: Index(['PC1', 'PC2', 'PC3', 'PC4', 'Pc5', 'PC6', 'PC7', 'PC8', 'PC9'], dtype='object')
```

```
07]: cdf=pd.DataFrame(lm.coef_, X.columns, columns=['Coeff'])
      cdf
```

```
07]:
```

	Coeff
PC1	1101.587208
PC2	-741.209042
PC3	208.536175
PC4	508.322458
Pc5	122.353301
PC6	1579.006858
PC7	333.611471
PC8	-1079.995115
PC9	1461.722691

After completion of training the model process, we test the remaining 60% of data on the model. The obtained results are checked using a scatter plot between predicted values and the original test data set for the dependent variable and acquired similar to a straight line as shown in the figure and the density function is also normally distributed.

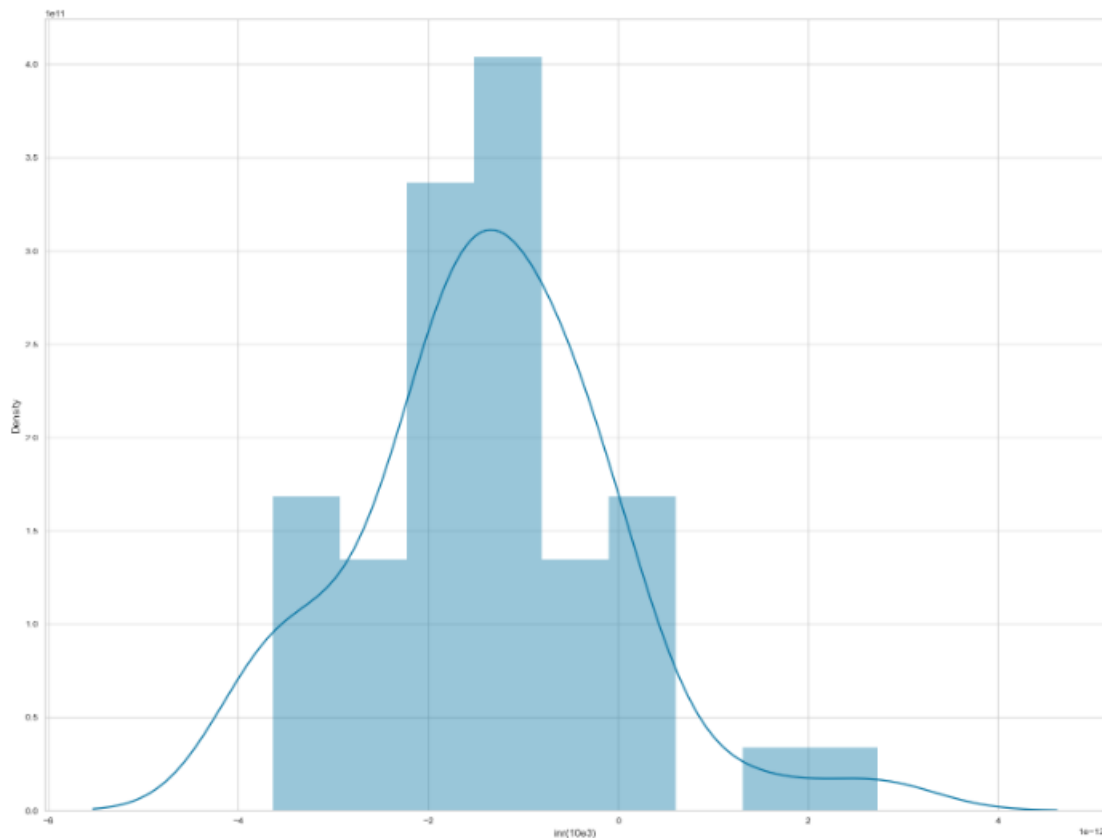


Fig.21. Density Function Is Also Normally Distributed

The metrics of the algorithm, Mean absolute error, Mean Squared error and mean square root error are described in the below figure:

```
8]: print('MAE:',mean_absolute_error(y_test,predictions))
    print('MSE:',mean_squared_error(y_test,predictions))
    print('RMSE:',np.sqrt(mean_squared_error(y_test,predictions)))
```

```
MAE: 1.5699610923461262e-12
MSE: 3.618915179919496e-24
RMSE: 1.902344653294848e-12
```

```
0]: mean_absolute_error(y_test,predictions)
```

```
0]: 1.5699610923461262e-12
```

```
2]: mean_squared_error(y_test,predictions)
```

```
2]: 3.618915179919496e-24
```

```
3]: np.sqrt(mean_squared_error(y_test,predictions))
```

```
3]: 1.902344653294848e-12
```

```
]:
```

PROFILING AND DESCRIBING THE SEGMENTS

Sorting the Top Speeds and Maximum Range in accordance to the Price with head () we can view the Pie Chart.

Pie Chart:

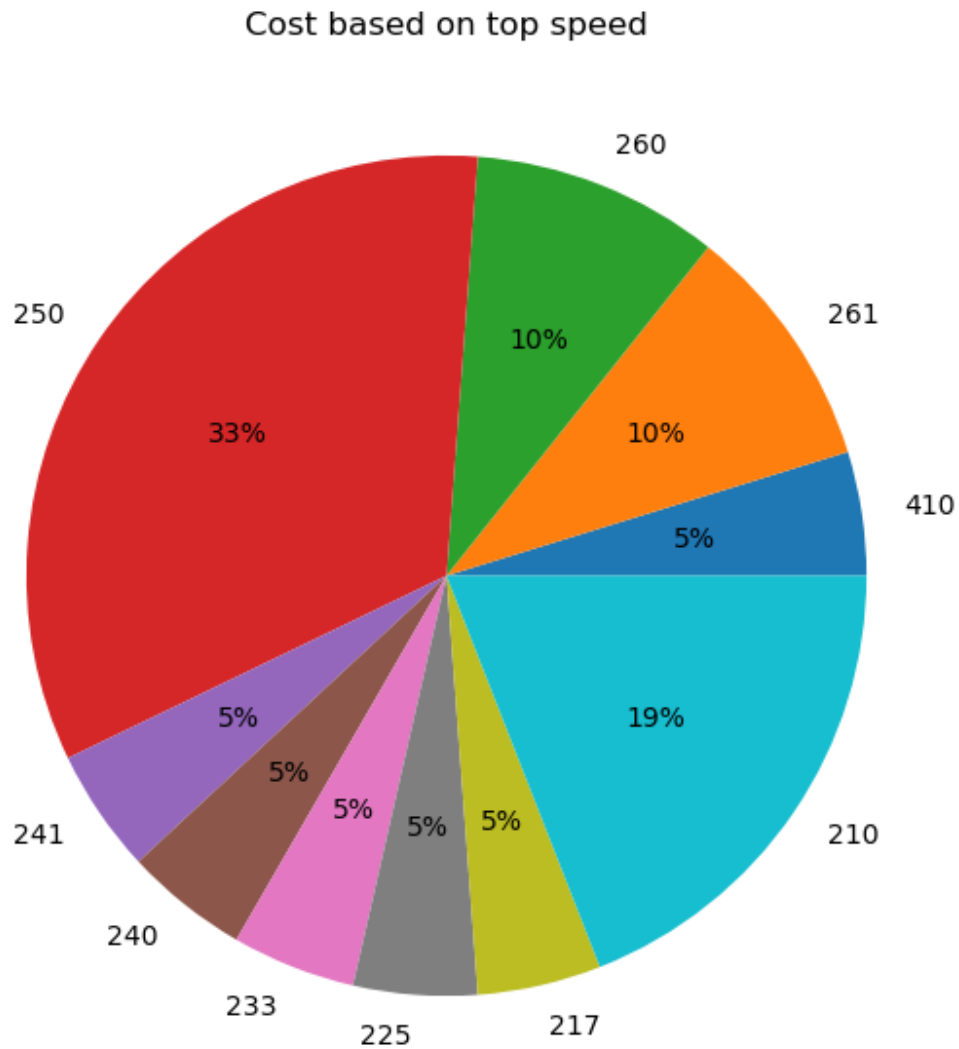


Fig.22. Cost Based on Top Speed

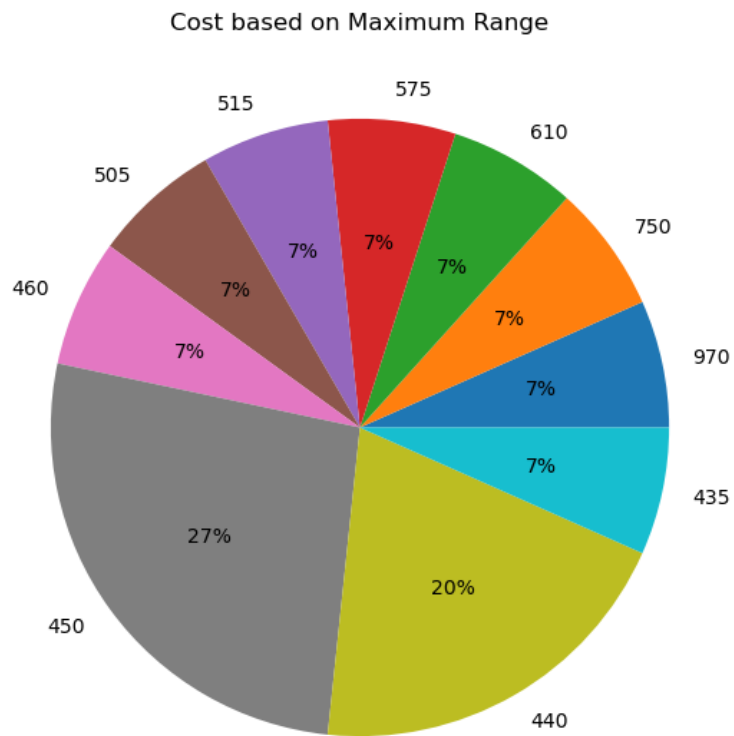


Fig.23. Cost Based on Maximum Range

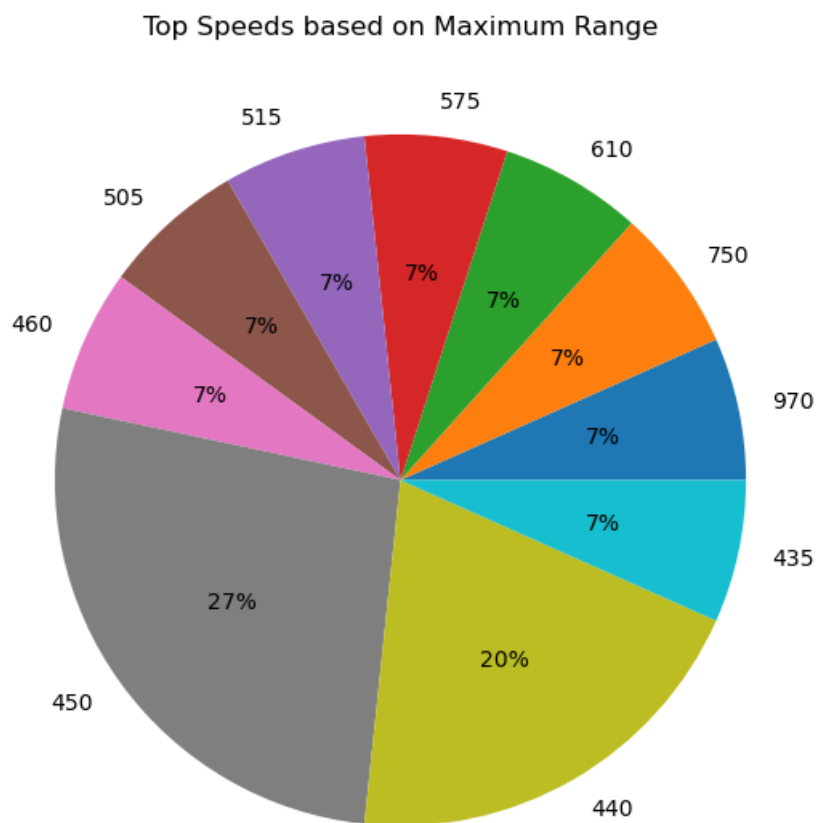


Fig.24. Top Speed Based on Maximum Range

TARGET SEGMENTS:

So, from the analysis we can see that the optimum targeted segment should be belonging to the following categories:

Behavioral:

Mostly from our analysis there are cars with 5 seats.

Demographic:

- Top Speed & Range : With a large area of market the cost is dependent on Top speeds and Maximum range of cars.
- Efficiency : Mostly the segments are with most efficiency.

Psychographic:

- Price : From the above analysis, the price range is between 16,00,000 to 1,80,00,000.

Finally, our target segment should contain cars with most Efficiency, contains Top Speed and price **between 16 to 180 lakhs** with mostly with **5 seats**.

Customizing the Marketing Mix



Reference

- <https://www.kaggle.com/datasets/geoffnel/evs-one-electric-vehicle-dataset>
- <https://c570blog.wordpress.com/2014/05/31/the-four-ps-of-marketing-mix/>
- [https://en.wikipedia.org/wiki/K-means clustering](https://en.wikipedia.org/wiki/K-means_clustering)
- [View: Time for India to move into top gear with an eye on 2030 EV public infrastructure goal - The Economic Times \(indiatimes.com\)](#)