# EV MARKET ANALYSIS-Copy1

July 4, 2022

[1]:
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
```

### 0.1 Import the dataset

[2]:
```
#data = pd.read_csv('C:/Users/DELL/Desktop/Feynn Lab/EV details/archive/
↪EV_CARS_INDIA.csv')
df = pd.read_csv('C:\\Users\\DELL\\Desktop\\python project\\ev_cars.csv')
```

[3]:
```
df
```

[3]:

| | Brand Name | Battery Capacity(kWh) | Acceleration(sec) | \ |
|---|---|---|---|---|
| 0 | Audi RS e-tron GT | 93.4 | 3.3 | |
| 1 | Audi e-tron GT | 93.4 | 4.1 | |
| 2 | Audi e-tron | 95.0 | 5.7 | |
| 3 | Tata Nexon EV | 30.2 | 9.9 | |
| 4 | Tata Tigor EV | 26.0 | 5.7 | |
| 5 | Hyudai Kona Electric | 39.2 | 9.7 | |
| 6 | Jaguar I-Pace | 90.0 | 4.8 | |
| 7 | Mahindra eVerito | 21.2 | 11.2 | |
| 8 | MG ZS EV | 44.5 | 8.5 | |

| | TopSpeed(km/h) | Range(km) | Max Power(kW) | Max Torque(Nm) | Transmission | \ |
|---|---|---|---|---|---|---|
| 0 | 250 | 480 | 500 | 830 | Automatic | |
| 1 | 245 | 500 | 523 | 630 | Automatic | |
| 2 | 200 | 484 | 300 | 664 | Automatic | |
| 3 | 180 | 312 | 96 | 245 | Automatic | |
| 4 | 120 | 306 | 55 | 170 | Automatic | |
| 5 | 155 | 452 | 103 | 395 | Automatic | |
| 6 | 200 | 470 | 294 | 696 | Automatic | |
| 7 | 86 | 140 | 33 | 91 | Automatic | |
| 8 | 120 | 340 | 107 | 353 | Automatic | |

| | No. of Seats | Charging T(h) | No. of Airbags | Drive Type | Price(Lakhs) |
|---|---|---|---|---|---|

```
0              5              9           Yes        AWD           204
1              5              9           Yes        AWD           179
2              5              9           Yes        AWD           123
3              5              9           Yes        FWD            17
4              5              9           Yes        FWD            14
5              5              7           Yes        FWD            24
6              5             13           Yes        AWD           112
7              5             12           Yes        FWD            10
8              5              8           Yes        FWD            25
```

### 0.2 Exploratory data analysis

[4]:
```
df.isna().sum()
```

[4]:
```
Brand Name              0
Battery Capacity(kWh)   0
Acceleration(sec)       0
TopSpeed(km/h)          0
Range(km)               0
Max Power(kW)           0
Max Torque(Nm)          0
Transmission            0
No. of Seats            0
Charging T(h)           0
No. of Airbags          0
Drive Type              0
Price(Lakhs)            0
dtype: int64
```

[5]:
```
df.columns
```

[5]:
```
Index(['Brand Name', 'Battery Capacity(kWh)', 'Acceleration(sec)',
       'TopSpeed(km/h)', 'Range(km)', 'Max Power(kW)', 'Max Torque(Nm)',
       'Transmission', 'No. of Seats', 'Charging T(h)', 'No. of Airbags',
       'Drive Type', 'Price(Lakhs)'],
      dtype='object')
```

[6]:
```
df.shape
```

[6]:
```
(9, 13)
```

[7]:
```
df.describe()
```

[7]:
```
       Battery Capacity(kWh)  Acceleration(sec)  TopSpeed(km/h)   Range(km)  \
count               9.000000           9.000000        9.000000    9.000000
mean               59.211111           6.988889      172.888889  387.111111
std                32.735472           2.870298       57.152088  121.111978
```

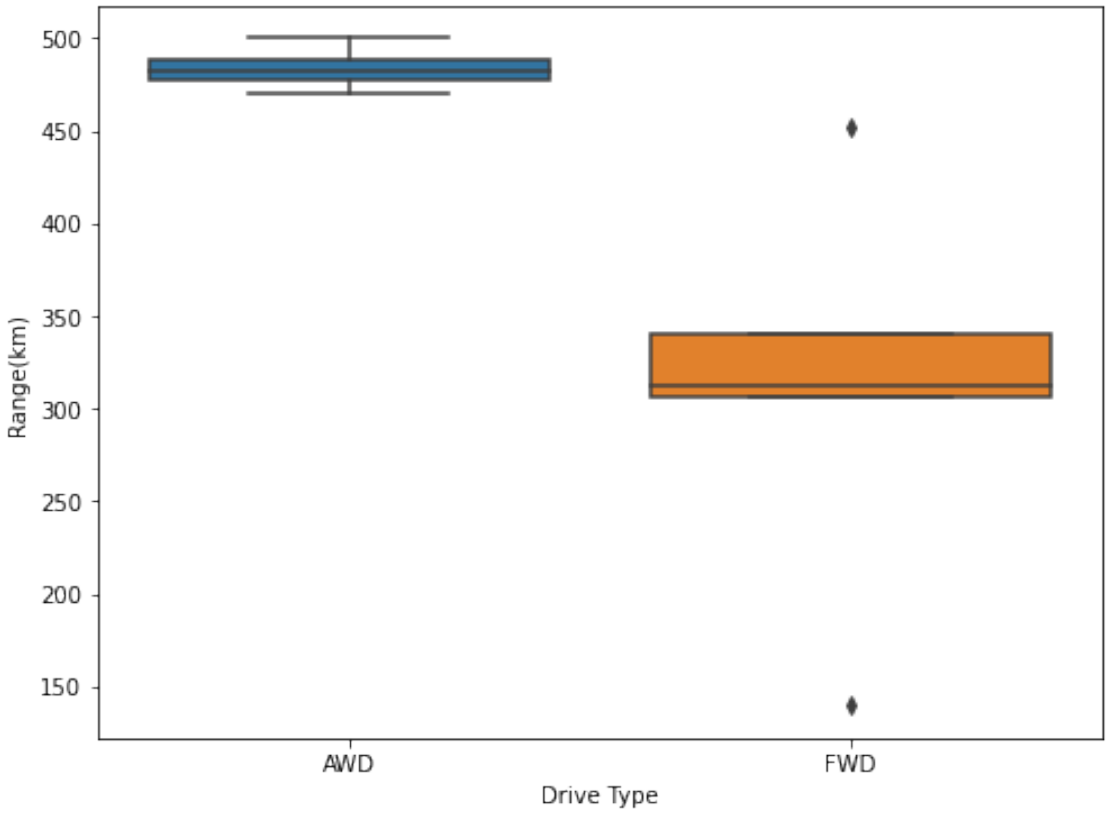
```
min                21.200000           3.300000       86.000000  140.000000
25%                30.200000           4.800000      120.000000  312.000000
50%                44.500000           5.700000      180.000000  452.000000
75%                93.400000           9.700000      200.000000  480.000000
max                95.000000          11.200000      250.000000  500.000000

       Max Power(kW)  Max Torque(Nm)  No. of Seats  Charging T(h)  \
count       9.000000        9.000000           9.0       9.000000
mean      223.444444      452.666667           5.0       9.444444
std       189.078761      260.956893           0.0       1.878238
min        33.000000       91.000000           5.0       7.000000
25%        96.000000      245.000000           5.0       9.000000
50%       107.000000      395.000000           5.0       9.000000
75%       300.000000      664.000000           5.0       9.000000
max       523.000000      830.000000           5.0      13.000000

       Price(Lakhs)
count      9.000000
mean      78.666667
std       76.990259
min       10.000000
25%       17.000000
50%       25.000000
75%      123.000000
max      204.000000
```

[8]:
```
df.groupby('Drive Type')['Brand Name'].agg('count').reset_index()
```

[8]:
```
  Drive Type  Brand Name
0        AWD           4
1        FWD           5
```

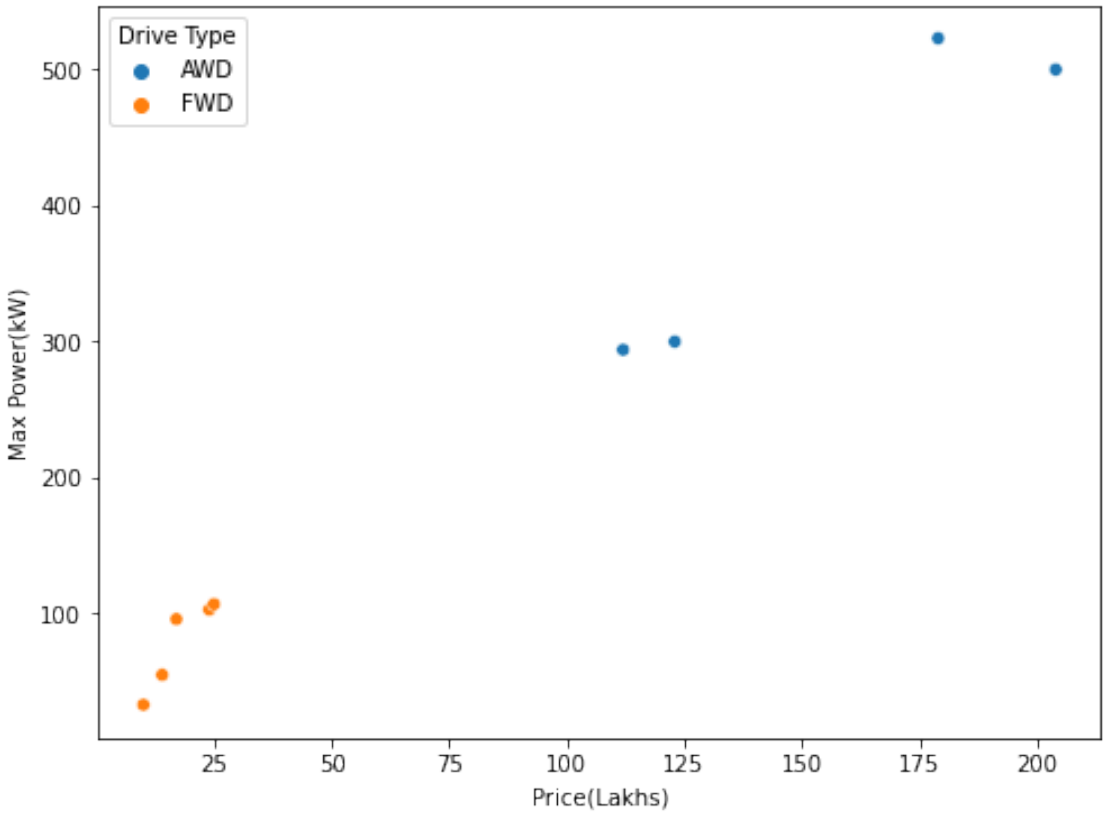
[9]:
```
fig=plt.figure(figsize=(8,6))
sns.boxplot(data=df,x='Drive Type',y="Range(km)")
```

[9]:
```
<AxesSubplot:xlabel='Drive Type', ylabel='Range(km)'>
```




[10]:
```
fig=plt.figure(figsize=(8,6))
sns.scatterplot(data=df,hue='Drive Type',y='Max Power(kW)',x='Price(Lakhs)')
```

[10]: <AxesSubplot:xlabel='Price(Lakhs)', ylabel='Max Power(kW)'>




### 0.3 Converting all catagorical value's into integer

[11]:
```
df['Transmission'] = pd.get_dummies(df['Transmission'])
df['No. of Airbags'] = pd.get_dummies(df['No. of Airbags'])
```

[12]:
```
df['Drive Type'] = df['Drive Type'].map({'AWD':0, 'FWD': 1})
```

[13]:
```
df = df.drop(['Brand Name'], axis=1)
```

[14]:
```
df.head()
```

[14]:

| | Battery Capacity(kWh) | Acceleration(sec) | TopSpeed(km/h) | Range(km) | \ |
|---|---|---|---|---|---|
| 0 | 93.4 | 3.3 | 250 | 480 | |
| 1 | 93.4 | 4.1 | 245 | 500 | |
| 2 | 95.0 | 5.7 | 200 | 484 | |
| 3 | 30.2 | 9.9 | 180 | 312 | |
| 4 | 26.0 | 5.7 | 120 | 306 | |

| | Max Power(kW) | Max Torque(Nm) | Transmission | No. of Seats | Charging T(h) | \ |
|---|---|---|---|---|---|---|
| 0 | 500 | 830 | 1 | 5 | 9 | |

```
1            523             630             1             5             9
2            300             664             1             5             9
3             96             245             1             5             9
4             55             170             1             5             9

   No. of Airbags  Drive Type  Price(Lakhs)
0               1           0           204
1               1           0           179
2               1           0           123
3               1           1            17
4               1           1            14
```

[15]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9 entries, 0 to 8
Data columns (total 12 columns):
 #   Column                 Non-Null Count  Dtype
---  ------                 --------------  -----
 0   Battery Capacity(kWh)  9 non-null      float64
 1   Acceleration(sec)      9 non-null      float64
 2   TopSpeed(km/h)         9 non-null      int64
 3   Range(km)              9 non-null      int64
 4   Max Power(kW)          9 non-null      int64
 5   Max Torque(Nm)         9 non-null      int64
 6   Transmission           9 non-null      uint8
 7   No. of Seats           9 non-null      int64
 8   Charging T(h)          9 non-null      int64
 9   No. of Airbags         9 non-null      uint8
 10  Drive Type             9 non-null      int64
 11  Price(Lakhs)           9 non-null      int64
dtypes: float64(2), int64(8), uint8(2)
memory usage: 866.0 bytes
```

[16]:
```
df['Battery Capacity(kWh)'] = df['Battery Capacity(kWh)'].astype(int)
df['Acceleration(sec)'] = df['Acceleration(sec)'].astype(int)
```

[17]:
```
X = df.iloc[:, :-1]
y = df.iloc[:, -1]
```

[18]:
```
X
```

[18]:
```
   Battery Capacity(kWh)  Acceleration(sec)  TopSpeed(km/h)  Range(km)  \
0                     93                  3             250        480
1                     93                  4             245        500
2                     95                  5             200        484
3                     30                  9             180        312
```

```
4                    26                   5              120         306
5                    39                   9              155         452
6                    90                   4              200         470
7                    21                  11               86         140
8                    44                   8              120         340

   Max Power(kW)  Max Torque(Nm)  Transmission  No. of Seats  Charging T(h)  \
0            500             830             1             5              9
1            523             630             1             5              9
2            300             664             1             5              9
3             96             245             1             5              9
4             55             170             1             5              9
5            103             395             1             5              7
6            294             696             1             5             13
7             33              91             1             5             12
8            107             353             1             5              8

   No. of Airbags  Drive Type
0               1           0
1               1           0
2               1           0
3               1           1
4               1           1
5               1           1
6               1           0
7               1           1
8               1           1
```

[19]:
```
y
```

[19]:
```
0    204
1    179
2    123
3     17
4     14
5     24
6    112
7     10
8     25
Name: Price(Lakhs), dtype: int64
```

Splitting the data set

[20]:
```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=.2,␣
 ↪random_state=1)
```

##Feature Scaling

[21]:
```
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
```

[22]:
```
X_train, X_test
```

[22]:
```
(array([[ 1.07640529, -0.83149718,  0.41320705,  0.73350688,  0.33248592,
          0.98160034,  0.        ,  0.        ,  1.72392288,  0.        ,
         -1.15470054],
        [-1.10806426,  1.56517116, -1.597398  , -1.95601834, -1.00551359,
         -1.30880045,  0.        ,  0.        ,  1.1992507 ,  0.        ,
          0.8660254 ],
        [ 1.17138222, -0.83149718,  1.20686694,  0.97800917,  1.50643952,
          0.73173844,  0.        ,  0.        , -0.37476584,  0.        ,
         -1.15470054],
        [ 1.17138222, -1.17387837,  1.29505137,  0.81500764,  1.38853152,
          1.48889572,  0.        ,  0.        , -0.37476584,  0.        ,
         -1.15470054],
        [-0.94976937, -0.48911599, -0.99774386, -0.60310565, -0.89273202,
         -1.00972333,  0.        ,  0.        , -0.37476584,  0.        ,
          0.8660254 ],
        [-0.82313345,  0.88040878,  0.06046932, -0.5542052 , -0.68254819,
         -0.72578934,  0.        ,  0.        , -0.37476584,  0.        ,
          0.8660254 ],
        [-0.53820264,  0.88040878, -0.38045284,  0.5868055 , -0.64666315,
         -0.15792138,  0.        ,  0.        , -1.42411021,  0.        ,
          0.8660254 ]]),
 array([[-0.37990775,  0.53802759, -0.99774386, -0.32600306, -0.62615741,
         -0.31692441,  0.        ,  0.        , -0.89943803,  0.        ,
          0.8660254 ],
        [ 1.23470018, -0.48911599,  0.41320705,  0.84760795,  0.36324453,
          0.86045517,  0.        ,  0.        , -0.37476584,  0.        ,
         -1.15470054]]))
```

[26]:
```
from sklearn.linear_model import LogisticRegression
log_classifier = LogisticRegression(random_state = 0)
log_classifier.fit(X_train, y_train)
```

[26]:
```
LogisticRegression(random_state=0)
```

[27]:
```
y_pred = log_classifier.predict(X_test)
```

[28]:
```
y_pred
```

[28]:
```
array([ 24, 179], dtype=int64)
```

[ ]: