

```
[84] # Importing libraries
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns
```

```
▶ #Importing dataset
carData=pd.read_csv('/content/drive/My Drive/data.csv')
carData.head()
```

```
[86] #Data types
carData.dtypes
```

```
[87] #Statistical summary
carData.describe()
```

```
[88] #Shape
carData.shape
```

```
[89] #NULL values
d=carData.isnull().sum()
d
```




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From the above output we can clearly see that there are maximum number of null values in the 'Market Category' column. Thus, for the ease of things we opt to remove it from the labelled columns. Also, we see that various more parameters do not have much significance in determining the prices of the cars. Thus dropping those columns won't make much of a difference in the processing of the model.

```
[90] #Eliminating the insignificant columns
```

```
carData = carData.drop(['Engine Fuel Type', 'Number of Doors', 'Market Category'], axis = 1)  
carData.head()
```

```
[91] carData.shape
```

```
[92] carData.rename(columns = { "Engine HP": "HP", "Engine Cylinders": "Cylinders",  
                                "Transmission Type": "Transmission", "Driven Wheels": "Drive Mode", "highway MPG": "MPG-H",  
                                "city mpg": "MPG-C", "MSRP": "Price"}, inplace = True)
```

```
[93] carData.drop_duplicates()
```

```
[94] #NULL values
```

```
print(carData.isnull().sum())
```

```
[95] carData=carData.dropna()
```

```
carData.count()
```

```
▶ carData.isnull().sum()
```

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```
[97] #Plotting graphs of Data (Columns)
      sns.boxplot(x=carData['HP'])
```

```
[98] sns.boxplot(x=carData['Cylinders'])
```

```
[99] sns.boxplot(x=carData['MPG-H'])
```

```
[100] sns.boxplot(x=carData['MPG-C'])
```

```
[101] sns.boxplot(x=carData['Popularity'])
```

```
[102] sns.boxplot(x=carData['Price'])
```

```
[103] q1=carData.quantile(0.25)
      q3=carData.quantile(0.75)
      iqr=q3-q1
      iqr
```

```
[104] carData=carData[~((carData<(q1-1.5*iqr))|(carData>(q3+1.5*iqr))).any(axis=1)]
```

```
[105] carData.shape
```

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```
[106] #Percentage of car per brand
```


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```
[108] #Percentage of car per brand
      counts=carData['Make'].value_counts()*100/sum(carData['Make'].value_counts())
      #Top 10 popular brands
      popularCars=counts.index[:10]
      #Plotting the bar plot
      plt.figure(figsize=(10,5))
      plt.bar(popularCars,height=counts[:10])
      plt.title('Top 10 car Brands')
      plt.show()

[109] prices=carData[['Make','Price']].loc[(carData['Make']=='Chevrolet')|
      (carData['Make']=='Volkswagen')|
      (carData['Make']=='Toyota')|
      (carData['Make']=='Nissan')|
      (carData['Make']=='GMC')|
      (carData['Make']=='Dodge')|
      (carData['Make']=='Mazda')|
      (carData['Make']=='Honda')|
      (carData['Make']=='Suzuki')|
      (carData['Make']=='Infiniti')].groupby('Make').mean()

      prices

#Correlation Matrix
corrMatrix=carData.corr()
sns.heatmap(corrMatrix,annot=True)
```




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▼ From the heatmap plotted above, it can be concluded that:

>>> Price is positively dependent on features and Horse Power(HP) and Year

>>> The features HP and Cylinders are positively dependent on each other

i.e. if number of Cylinders are increased, the HP also increases.

>>>MPG-H and MPG-C have strong negative correlation with Cylinders:

i.e. if number of cylinders are increased, MPG-H and MPG-C decreases.

```
[115] sns.barplot(carData['Year'],carData['Price'])
```

▶ `sns.barplot(carData['HP'],carData['Price'])`

▶ `sns.barplot(carData['Cylinders'],carData['Price'])`

▶ `sns.barplot(carData['MPG-H'],carData['Price'])`

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▶ `sns.barplot(carData['Cylinders'],carData['Price'])`

▶ `sns.barplot(carData['MPG-H'],carData['Price'])`

[118] `sns.barplot(carData['MPG-C'],carData['Price'])`

▶ `sns.barplot(carData['Popularity'],carData['Price'])`