Q.1) Download the car dataset from the following link-

https://www.kaggle.com/CooperUnion/cardataset

Import required libraries, read data, display car companies with car numbers, clean data and display sample data for 2 variables.

Solution:

```
#Importing Libraries
import pandas as pd
#Reading data
df = pd.read csv('C:/TYBSC/car data.csv') # Importing the data set
df.sample(5)
print(df.sample(5))
print(df.shape)
#Print car companies with car numbers
print(df['Make'].value counts())
new df = df[df]'Make'] == 'Volkswagen']
print(new df.shape) # Viewing the new dataset shape
#Data Cleaning
print(new df.isnull().sum())
new df = new df.dropna()
print(new df.isnull().sum())
new_df.shape
new_df.isnull().sum()
print(new_df.sample(2))
#Sample data for 2 variables..
new df = new df[['Engine HP','MSRP']]
print(new df.sample(5))
```

Output:

runfile('C:/TYBSC/LinearReg1.py', wdir='C:/TYBSC')

Make	Mode	l Year city m	pg P	opulari	ty MSRP
10785	Chevrolet	Trax 2017		24	1385 22500
1991	Pontiac	Bonneville 2005	;	15	210 35585
1772	Subaru	B9 Tribeca 2007	7	16	640 34495
9387	GMC	Sierra 1500 201	7	16	549 52455

[5 rows x 16 columns] (11914, 16)

Chevrolet 1123 Ford 881 809 Volkswagen 746 Toyota Dodge 626 Nissan 558 GMC515 Honda 449 Mazda 423 397 Cadillac Mercedes-Benz 353 351 Suzuki BMW334 Infiniti 330 Audi 328 Hyundai 303 Volvo 281 Subaru 256 252 Acura Kia 231 Mitsubishi 213 Lexus 202 196 Buick Chrysler 187 Pontiac 186 Lincoln 164 Oldsmobile 150 Land Rover 143 Porsche 136 111 Saab Aston Martin 93 82 Plymouth Bentley 74 Ferrari 69 **FIAT** 62 60 Scion Maserati 58 52 Lamborghini 31 Rolls-Royce Lotus 29 18 Tesla **HUMMER** 17 Maybach 16 5 Alfa Romeo 5 McLaren Spyker 3 Genesis 3 3

Bugatti

Name: Make, dtype: int64 (809, 16)0 Make 0 Model 0 Year Engine Fuel Type 0 0 Engine HP **Engine Cylinders** 4 Transmission Type 0 Driven_Wheels 0 Number of Doors 0 224 Market Category Vehicle Size 0 0 Vehicle Style 0 highway MPG 0 city mpg **Popularity** 0 **MSRP** 0 dtype: int64

0 Make 0 Model Year Engine Fuel Type 0 Engine HP 0 **Engine Cylinders** 0 Transmission Type 0 Driven_Wheels 0 Number of Doors 0 0 Market Category 0 Vehicle Size Vehicle Style 0 highway MPG 0 0 city mpg **Popularity** 0 0 **MSRP** dtype: int64

Make Model Year ... city mpg Popularity MSRP

1864 Volkswagen Beetle Convertible 2016 ... 873 32670 23 2959 Volkswagen Corrado 1992 ... 16 873 2000

[2 rows x 16 columns]

Engine HP MSRP

6079 140.0 25795 1915 170.0 21795 5332 200.0 24770 5459 170.0 19595 6031 210.0 30875

Q.2)) Download the car dataset from the following link-

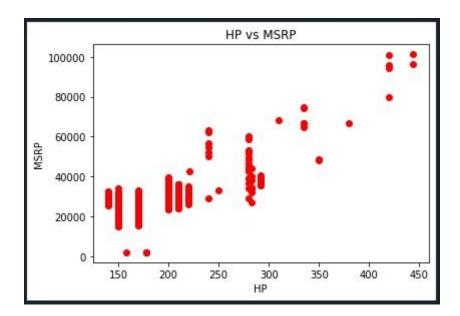
https://www.kaggle.com/CooperUnion/cardataset

Import required libraries, read data and split test and train data.

Solution:

```
#Importing libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model selection import train test split
from sklearn.linear model import LinearRegression
#Read Data
df = pd.read csv('C:/TYBSC/car data.csv') # Importing the data set
df.sample(5) #previewing dataset randomly
print(df.sample(5))
new df = df[df]'Make']=='Volkswagen'] # in this new data set we only take 'Volkswagen' Cars
#print(new df.isnull().sum()) # Is there any Null or Empty cell presents
new df = new df.dropna() # Deleting the rows which have Empty cells
#print(new df.isnull().sum()) # Is there any Null or Empty cell presents
new df.isnull().sum() #Is there any Null or Empty cell presents
print(new df.sample(2)) # Checking the random dataset sample
new df = new df[['Engine HP', 'MSRP']] # We only take the 'Engine HP' and 'MSRP' columns
print(new df.sample(5))
#Split Train and Test dataset
X = \text{np.array}(\text{new df}[[\text{Engine HP'}]]) \# \text{Storing into } X \text{ the 'Engine HP' as np.array}
y = np.array(new df[['MSRP']]) # Storing into y the 'MSRP' as np.array
plt.scatter(X,y,color="red") # Plot a graph X vs y
plt.title('HP vs MSRP')
plt.xlabel('HP')
plt.ylabel('MSRP')
plt.show()
```

```
X train, X test, y train, y test = train test split(X, y, test size = 0.25, random state=15)
       regressor = LinearRegression()
       regressor.fit(X train,y train)
       plt.scatter(X test,y test,color="green") # Plot a graph with X test vs y test
       plt.plot(X train,regressor.predict(X train),color="red",linewidth=3)
       plt.title('Regression(Test Set)')
       plt.xlabel('HP')
       plt.ylabel('MSRP')
       plt.show()
       plt.scatter(X train,y train,color="blue") # Plot a graph with X train vs y train
       plt.plot(X train,regressor.predict(X train),color="red",linewidth=3)
       plt.title('Regression(training Set)')
       plt.xlabel('HP')
       plt.ylabel('MSRP')
       plt.show()
Output:
       runfile('C:/TYBSC/LinearReg1.py', wdir='C:/TYBSC')
              Make Model Year ... city mpg Popularity MSRP
                                            25
                                                    640 25295
       7341
               Subaru Outback 2016 ...
       44
               BMW 2 Series 2016 ...
                                           23
                                                  3916 34850
       4080 Hyundai
                         Equus 2015 ...
                                           15
                                                   1439 61500
       10350
                 GMC Terrain 2016 ...
                                            22
                                                   549 23975
       9977 Cadillac
                          SRX 2016 ...
                                           17
                                                  1624 48920
       [5 rows x 16 columns]
              Make Model Year ... city mpg Popularity MSRP
       5466 Volkswagen Golf 2017 ...
                                           25
                                                   873 20995
       5675 Volkswagen GTI 2013 ...
                                           21
                                                   873 28795
       [2 rows x 16 columns]
           Engine HP MSRP
       1841
                210.0 29895
       10530
                 280.0 39300
       5325
                200.0 25375
       1835
                210.0 30995
       6059
                140.0 28390
```



OBJ

