In [1]: import numpy as np
import pandas as pd

In [2]: df=pd.read_csv(r"C:\Users\Sushma sree\Downloads\used_cars_data.csv")
 df

Out[2]:

| | S.No. | Name | Location | Year | Kilometers_Driven | Fuel_Type | Transmission | Owner_ |
|------|-------|---|------------|------|-------------------|-----------|--------------|--------|
| 0 | 0 | Maruti Wagon R LXI CNG | Mumbai | 2010 | 72000 | CNG | Manual | |
| 1 | 1 | Hyundai Creta 1.6 CRDi SX Option | Pune | 2015 | 41000 | Diesel | Manual | |
| 2 | 2 | Honda Jazz V | Chennai | 2011 | 46000 | Petrol | Manual | |
| 3 | 3 | Maruti Ertiga VDI | Chennai | 2012 | 87000 | Diesel | Manual | |
| 4 | 4 | Audi A4 New 2.0 TDI Multitronic | Coimbatore | 2013 | 40670 | Diesel | Automatic | Se |
| | | | | | | | | |
| 7248 | 7248 | Volkswagen Vento Diesel Trendline | Hyderabad | 2011 | 89411 | Diesel | Manual | |
| 7249 | 7249 | Volkswagen Polo GT TSI | Mumbai | 2015 | 59000 | Petrol | Automatic | |
| 7250 | 7250 | Nissan Micra Diesel XV | Kolkata | 2012 | 28000 | Diesel | Manual | |
| 7251 | 7251 | Volkswagen Polo GT TSI | Pune | 2013 | 52262 | Petrol | Automatic | |
| 7252 | 7252 | Mercedes- Benz E- Class 2009- 2013 E 220 CDI Avan | Kochi | 2014 | 72443 | Diesel | Automatic | |

7253 rows × 14 columns

In [3]: df.head()

Out[3]:

| | S.No. | Name | Location | Year | Kilometers_Driven | Fuel_Type | Transmission | Owner_Type |
|---|-------|---|------------|------|-------------------|-----------|--------------|------------|
| 0 | 0 | Maruti Wagon R LXI CNG | Mumbai | 2010 | 72000 | CNG | Manual | First |
| 1 | 1 | Hyundai Creta 1.6 CRDi SX Option | Pune | 2015 | 41000 | Diesel | Manual | First |
| 2 | 2 | Honda Jazz V | Chennai | 2011 | 46000 | Petrol | Manual | First |
| 3 | 3 | Maruti Ertiga VDI | Chennai | 2012 | 87000 | Diesel | Manual | First |
| 4 | 4 | Audi A4 New 2.0 TDI Multitronic | Coimbatore | 2013 | 40670 | Diesel | Automatic | Second |
| 4 | | | | | | | | • |

In [4]: df.tail()

Out[4]:

| | S.No. | Name | Location | Year | Kilometers_Driven | Fuel_Type | Transmission | Owner_1 |
|------|-------|---|-----------|------|-------------------|-----------|--------------|---------|
| 7248 | 7248 | Volkswagen Vento Diesel Trendline | Hyderabad | 2011 | 89411 | Diesel | Manual | |
| 7249 | 7249 | Volkswagen Polo GT TSI | Mumbai | 2015 | 59000 | Petrol | Automatic | |
| 7250 | 7250 | Nissan Micra Diesel XV | Kolkata | 2012 | 28000 | Diesel | Manual | |
| 7251 | 7251 | Volkswagen Polo GT TSI | Pune | 2013 | 52262 | Petrol | Automatic | 1 |
| 7252 | 7252 | Mercedes- Benz E- Class 2009- 2013 E 220 CDI Avan | Kochi | 2014 | 72443 | Diesel | Automatic | |
| 4 | | | | | | | | |

In [5]: df.shape

Out[5]: (7253, 14)

In [6]: df.describe

| Out[6]: | <box< th=""><th>d method Locati</th><th>NDFrame.des</th><th>cribe o</th><th>f</th><th>S.N</th><th>No.</th><th></th><th></th><th></th><th></th></box<> | d method Locati | NDFrame.des | cribe o | f | S.N | No. | | | | | |
|---------|---|--------------------|-----------------------|------------|------------|---------|---------------|------------|---------|---------|---|--|
| | 0 | 0 | | | | Mar | ruti Wagon | R LXI CNG | Mum | bai | ١ | |
| | 1 | 1 | | Hv | undai (| | 1.6 CRDi | | | une | | |
| | 2 | 2 Honda Jazz V | | | | | | | | Chennai | | |
| | 3 | 3 | | | | | | rtiga VDI | Chen | | | |
| | 4 | 4 | | Α | udi A4 | New | 2.0 TDI Mu | • | Coimbat | | | |
| | | | | | | | | • • • | | | | |
| | 7248 | 7248 | | Vol | .kswager | า Ver | nto Diesel | Trendline | Hydera | bad | | |
| | 7249 | 7249 | | | | Vo] | lkswagen Po | olo GT TSI | Mum | bai | | |
| | 7250 | 7250 | | | | Nis | ssan Micra | Diesel XV | Kolk | ata | | |
| | 7251 | 7251 | | | | Vo] | lkswagen Po | olo GT TSI | Р | une | | |
| | 7252 | 7252 M | lercedes-Ben | z E-Cla | ss 2009 | 9-201 | L3 E 220 CD | OI Avan | Ko | chi | | |
| | | Year Ki | lometers_Dr | ivon Eu | ol Type | Tna | nemiccion | Owner Type | м; 1 | eage | | |
| | 0 | 2010 | - | 2000 | CNC CNC | | Manual | First | 26.6 k | _ | | |
| | \ | 2010 | , | 2000 | CIVIC | , | Manual | 11130 | 20.0 K | III/ Ng | | |
| | 1 | 2015 | Δ | 1000 | Diese] | l | Manual | First | 19.67 | kmn1 | | |
| | 2 | 2013 | | 6000 | Petro] | | Manual | First | 18.2 | - | | |
| | 3 | 2012 | | 7000 | Diese | | Manual | First | 20.77 | | | |
| | 4 | 2013 | | 9670 | Diesel | | Automatic | Second | 15.2 | | | |
| | • • • | ••• | · | ••• | ••• | | • • • | ••• | | • • • | | |
| | 7248 | 2011 | 8 | 9411 | Diese] | | Manual | First | 20.54 | kmp1 | | |
| | 7249 | 2015 | | 9000 | Petro] | | Automatic | First | 17.21 | | | |
| | 7250 | 2012 | | 8000 | Diese] | | Manual | First | 23.08 | | | |
| | 7251 | 2013 | 5 | 2262 | Petro] | L | Automatic | Third | 17.2 | | | |
| | 7252 | 2014 | 7 | 2443 | Diese] | L | Automatic | First | 10.0 | kmp1 | | |
| | | | Davies | C+- | Nav. D | | Duice | | | | | |
| | 0 | Engine | Power | Seats | New_Pr | | Price | | | | | |
| | 0 | 998 CC | 58.16 bhp | 5.0 | | NaN | 1.75 | | | | | |
| | 1 2 | 1582 CC 1199 CC | 126.2 bhp 88.7 bhp | 5.0 | 8.61 l | NaN | 12.50 4.50 | | | | | |
| | 3 | 1199 CC 1248 CC | 88.76 bhp | 5.0 7.0 | 8.01 | NaN | 6.00 | | | | | |
| | 4 | 1968 CC | 140.8 bhp | 5.0 | | | 17.74 | | | | | |
| | | | - | | | NaN | | | | | | |
| | 7248 | 1598 CC | 103.6 bhp | 5.0 | | NaN | NaN | | | | | |
| | 7249 | 1197 CC | 103.6 bhp | 5.0 | | NaN | NaN | | | | | |
| | 7250 | 1461 CC | 63.1 bhp | 5.0 | | NaN | NaN | | | | | |
| | 7251 | 1197 CC | 103.6 bhp | 5.0 | | NaN | NaN | | | | | |
| | 7252 | 2148 CC | 170 bhp | 5.0 | | NaN | NaN | | | | | |
| | 1232 | 2170 00 | 1/0 onb | ٥.٠ | | Naiv | IVAIN | | | | | |

[7253 rows x 14 columns]>

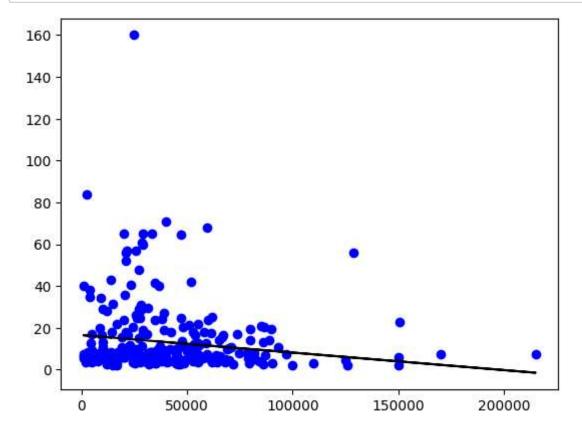
```
In [7]: df.isnull().sum()
Out[7]: S.No.
                                  0
                                  0
         Name
                                  0
         Location
         Year
                                  0
         Kilometers_Driven
                                  0
         Fuel_Type
                                  0
         Transmission
                                  0
                                  0
         Owner_Type
                                  2
         Mileage
                                 46
         Engine
         Power
                                 46
         Seats
                                 53
         New Price
                               6247
         Price
                               1234
         dtype: int64
In [8]: df.dropna(inplace=True)
In [9]: df.isnull().sum()
Out[9]: S.No.
                               0
                               0
         Name
                               0
         Location
                               0
         Year
                               0
         Kilometers_Driven
         Fuel_Type
                               0
                               0
         Transmission
                               0
         Owner_Type
                               0
         Mileage
                               0
         Engine
         Power
                               0
                               0
         Seats
                               0
         New_Price
                               0
         Price
         dtype: int64
In [10]: from sklearn.model_selection import train_test_split
         from matplotlib import pyplot as plt
```

```
In [11]: plt.scatter(df['Kilometers_Driven'],df['Price'])
Out[11]: <matplotlib.collections.PathCollection at 0x23948799810>
           160
           140
           120
           100
            80
            60
            40
            20
             0
                             50000
                                                                      200000
                  0
                                          100000
                                                        150000
In [12]: from sklearn.linear_model import LinearRegression
         lr=LinearRegression()
In [13]: x=df[['Kilometers_Driven']]
         y=df['Price']
In [14]: | x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
In [15]: lr.fit(x_train,y_train)
Out[15]:
          ▼ LinearRegression
          LinearRegression()
```

In [16]: lr.score(x_test,y_test)

Out[16]: 0.008270420401903733

```
In [17]: y_pred=lr.predict(x_test)
    plt.scatter(x_test,y_test,color='b')
    plt.plot(x_test,y_pred,color='k')
    plt.show()
```



In [30]: df1000=df[:1000] df1000

Out[30]:

| | S.No. | Name | Location | Year | Kilometers_Driven | Fuel_Type | Transmission | Owner_Ty |
|------|-------|--|----------|------|-------------------|-----------|--------------|----------|
| 2 | 2 | Honda Jazz V | Chennai | 2011 | 46000 | Petrol | Manual | Fi |
| 7 | 7 | Toyota Innova Crysta 2.8 GX AT 8S | Mumbai | 2016 | 36000 | Diesel | Automatic | Fi |
| 10 | 10 | Maruti Ciaz Zeta | Kochi | 2018 | 25692 | Petrol | Manual | Fi |
| 15 | 15 | Mitsubishi Pajero Sport 4X4 | Delhi | 2014 | 110000 | Diesel | Manual | Fi |
| 20 | 20 | BMW 3 Series 320d | Kochi | 2014 | 32982 | Diesel | Automatic | Fi |
| | | | | | | | | |
| 5999 | 5999 | Tata Bolt Revotron XT | Chennai | 2016 | 10000 | Petrol | Manual | Fi |
| 6002 | 6002 | Volkswagen Vento 1.6 Highline | Mumbai | 2011 | 38000 | Petrol | Manual | Fi |
| 6005 | 6005 | Maruti Vitara Brezza VDi | Pune | 2016 | 37208 | Diesel | Manual | Fi |
| 6010 | 6010 | Honda Brio 1.2 VX MT | Delhi | 2013 | 33746 | Petrol | Manual | Fi |
| 6014 | 6014 | Maruti Swift VDI | Delhi | 2014 | 27365 | Diesel | Manual | Fi |
| | | | | | | | | |

823 rows × 14 columns

In [31]: x=df1000[['Kilometers_Driven']]
y=df1000['Price']

In [32]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2)

In [33]: lr.fit(x_train,y_train)

Out[33]: v LinearRegression LinearRegression()

```
In [34]: |lr.score(x_test,y_test)
Out[34]: 0.023020345210620463
In [35]: y_pred=lr.predict(x_test)
         plt.scatter(x_test,y_test,color='b')
         plt.plot(x_test,y_pred,color='k')
         plt.show()
          160
          140
          120
          100
           80
           60
            40
           20
             0
                 0
                       25000
                                50000
                                        75000
                                                100000 125000 150000 175000
```

In []: