```
In [12]:
           import pandas as pd
           import warnings
           warnings.filterwarnings
           warnings.filterwarnings("ignore")
In [13]: data=pd.read csv("/home/placement/Downloads/fiat500.csv")
           data.describe()
In [14]:
Out[14]:
                            ID engine_power
                                                                    km previous_owners
                                              age_in_days
                                                                                                  lat
                                                                                                              lon
                                                                                                                          price
                   1538.000000
                                 1538.000000
                                              1538.000000
                                                            1538.000000
                                                                             1538.000000
                                                                                         1538.000000
                                                                                                      1538.000000
                                                                                                                   1538.000000
             count
                    769.500000
                                   51.904421
                                              1650.980494
                                                            53396.011704
                                                                                1.123537
                                                                                            43.541361
                                                                                                        11.563428
                                                                                                                   8576.003901
             mean
                                                                                                         2.328190
               std
                    444.126671
                                    3.988023
                                              1289.522278
                                                            40046.830723
                                                                                0.416423
                                                                                             2.133518
                                                                                                                   1939.958641
                      1.000000
                                                                                                         7.245400
              min
                                   51.000000
                                               366.000000
                                                            1232.000000
                                                                                1.000000
                                                                                            36.855839
                                                                                                                   2500.000000
              25%
                    385.250000
                                   51.000000
                                               670.000000
                                                            20006.250000
                                                                                1.000000
                                                                                            41.802990
                                                                                                         9.505090
                                                                                                                   7122.500000
              50%
                    769.500000
                                   51.000000
                                              1035.000000
                                                            39031.000000
                                                                                1.000000
                                                                                            44.394096
                                                                                                        11.869260
                                                                                                                   9000.000000
              75%
                   1153.750000
                                   51.000000
                                              2616.000000
                                                            79667.750000
                                                                                1.000000
                                                                                            45.467960
                                                                                                        12.769040
                                                                                                                  10000.000000
              max 1538.000000
                                   77.000000
                                              4658.000000
                                                           235000.000000
                                                                                4.000000
                                                                                            46.795612
                                                                                                        18.365520
                                                                                                                  11100.000000
In [15]:
           data.head()
Out[15]:
                   model engine power age in days
                                                         km previous owners
                                                                                    lat
                                                                                              Ion price
                1
                   lounge
                                    51
                                                      25000
                                                                                                  8900
                                                882
                                                                           1 44.907242
                                                                                         8.611560
                2
                      pop
                                     51
                                               1186
                                                      32500
                                                                              45.666359
                                                                                       12.241890
                                                                                                  8800
                                     74
                                               4658
                                                     142228
                                                                              45.503300 11.417840
                                                                                                  4200
                    sport
                                     51
                                                     160000
                                                                              40.633171 17.634609
                                                                                                  6000
                4
                   lounge
                                               2739
                5
                                     73
                                               3074 106880
                                                                           1 41.903221 12.495650
                                                                                                  5700
                      pop
In [16]: data1=data.drop(['lat','lon','ID'],axis=1)
```

In [17]: data1

Out[17]:

	model	engine_power	age_in_days	km	previous_owners	price
0	lounge	51	882	25000	1	8900
1	pop	51	1186	32500	1	8800
2	sport	74	4658	142228	1	4200
3	lounge	51	2739	160000	1	6000
4	pop	73	3074	106880	1	5700
1533	sport	51	3712	115280	1	5200
1534	lounge	74	3835	112000	1	4600
1535	pop	51	2223	60457	1	7500
1536	lounge	51	2557	80750	1	5990
1537	pop	51	1766	54276	1	7900

1538 rows × 6 columns

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	engine_power	age_in_days	km	previous_owners	price	model_lounge	model_pop	model_sport
0	51	882	25000	1	8900	1	0	0
1	51	1186	32500	1	8800	0	1	0
2	74	4658	142228	1	4200	0	0	1
3	51	2739	160000	1	6000	1	0	0
4	73	3074	106880	1	5700	0	1	0
1533	51	3712	115280	1	5200	0	0	1
1534	74	3835	112000	1	4600	1	0	0
1535	51	2223	60457	1	7500	0	1	0
1536	51	2557	80750	1	5990	1	0	0
1537	51	1766	54276	1	7900	0	1	0

1538 rows × 8 columns

```
In [19]: data2.shape
Out[19]: (1538, 8)
In [20]: #pridected value we removed from data frame
    y=data2['price']
    x=data2.drop('price',axis=1)
```

```
In [21]: y
Out[21]: 0
                  8900
                  8800
                  4200
          2
          3
                  6000
          4
                  5700
                   . . .
          1533
                  5200
          1534
                  4600
          1535
                  7500
          1536
                  5990
          1537
                  7900
          Name: price, Length: 1538, dtype: int64
In [23]: #divide the data into testing & training
          from sklearn.model selection import train test split
          x train,x test,y train,y test=train test split(x,y,test size=0.33,random state=42)
In [24]: #t0 show starting rows
         x test.head(5)
Out[24]:
               engine_power age_in_days
                                         km previous_owners model_lounge model_pop model_sport
           481
                        51
                                 3197 120000
                                                         2
                                                                     0
                                                                              1
                                                                                         0
                                 2101 103000
            76
                                                         1
                                                                     0
                                                                                         0
                        62
                                                                              1
           1502
                        51
                                  670
                                       32473
                                                         1
                                                                    1
                                                                              0
                                                                                         0
           669
                        51
                                  913
                                       29000
                                                         1
                                                                     1
                                                                              0
                                                                                         0
```

```
In [25]: |y_train.head(5)
Out[25]: 527
                 9990
                 9500
          129
          602
                 7590
          331
                 8750
          323
                 9100
          Name: price, dtype: int64
In [26]: y test.head(5)
Out[26]: 481
                  7900
          76
                  7900
          1502
                  9400
          669
                  8500
          1409
                  9700
          Name: price, dtype: int64
In [27]: y train.head(5)
Out[27]: 527
                 9990
          129
                 9500
          602
                 7590
          331
                 8750
          323
                 9100
          Name: price, dtype: int64
In [28]: #linear regression
          from sklearn.linear model import LinearRegression
          reg=LinearRegression()#creating object of LinearRegression
          reg.fit(x train,y train)#training and fitting LR object using training data and the model is created by trai
Out[28]: LinearRegression()
          In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
```

```
In [29]:
         #prediction price
         y pred=req.predict(x test)
         y pred
Out[29]: array([ 5867.6503378 ,
                                 7133.70142341.
                                                  9866.35776216.
                                                                  9723.28874535.
                10039.59101162,
                                 9654.07582608,
                                                  9673.14563045, 10118.70728123,
                                 9351.55828437, 10434.34963575, 7732.26255693,
                 9903.85952664,
                                                  9662.90103518, 10373.20344286.
                                 6565.95240435.
                 7698.67240131.
                 9599.94844451,
                                 7699.34400418,
                                                  4941.33017994, 10455.2719478 ,
                10370.51555682, 10391.60424404,
                                                  7529.06622456,
                                                                  9952.37340054,
                                                                  6953.10376491,
                 7006.13845729.
                                 9000.1780961 .
                                                  4798.36770637.
                 7810.39767825,
                                 9623.80497535,
                                                  7333.52158317,
                                                                  5229.18705519,
                 5398.21541073,
                                 5157.65652129,
                                                  8948.63632836,
                                                                  5666.62365159,
                 9822.1231461 ,
                                                  6279.2040404 ,
                                                                  8457.38443276,
                                 8258.46551788,
                 9773.86444066,
                                 6767.04074749,
                                                  9182.99904787, 10210.05195479,
                 8694.90545226, 10328.43369248,
                                                  9069.05761443,
                                                                  8866.7826029 ,
                 7058.39787506,
                                 9073.33877162,
                                                  9412.68162121, 10293.69451263,
                10072.49011135,
                                 6748.5794244 .
                                                  9785.95841801,
                                                                  9354.09969973.
                 9507.9444386 , 10443.01608254,
                                                  9795.31884316,
                                                                  7197.84932877,
                10108.31707235, 7009.6597206,
                                                  9853.90699412,
                                                                  7146.87414965,
                 6417.69133992.
                                 9996.97382441.
                                                  9781.18795953,
                                                                  8515.83255277,
                                 6499.76668237,
                                                  7768.57829985,
                                                                  6832.86406122,
                 8456.30006203,
                 8347.96113362. 10439.02404036.
                                                  7356.43463051.
                                                                  8562.56562053.
In [30]: from sklearn.metrics import r2 score
         r2 score(y test,y pred)#y test=actual price,y pred=predicted price
Out[30]: 0.8415526986865394
In [31]: from sklearn.metrics import mean squared error#calculating MSE
         mean squared error(v pred, v test)
Out[31]: 581887.727391353
 In [ ]:
```