```
In [1]: import pandas as pd
In [2]: data = pd.read csv("/home/placement/Desktop/yamuna/fiat500.csv")
In [3]: data.describe()
Out[3]:
                                                                   km previous_owners
                           ID engine power
                                            age in days
                                                                                                 lat
                                                                                                            lon
                                                                                                                         price
           count 1538.000000
                                1538.000000
                                                                            1538.000000
                                                                                        1538.000000
                                                                                                     1538.000000
                                                                                                                  1538.000000
                                             1538.000000
                                                           1538.000000
                   769.500000
                                                                                                       11.563428
                                  51.904421
                                             1650.980494
                                                           53396.011704
                                                                               1.123537
                                                                                          43.541361
                                                                                                                  8576.003901
           mean
                   444.126671
                                   3.988023
                                             1289.522278
                                                           40046.830723
                                                                               0.416423
                                                                                           2.133518
                                                                                                        2.328190
                                                                                                                  1939.958641
              std
                     1.000000
                                  51.000000
                                              366.000000
                                                           1232.000000
                                                                               1.000000
                                                                                          36.855839
                                                                                                        7.245400
             min
                                                                                                                  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 [4]:
          data.head()
Out[4]:
                  model engine power age in days
                                                       km previous owners
                                                                                             Ion price
                                                                                   lat
              1 lounge
           0
                                   51
                                               882
                                                     25000
                                                                          1 44.907242
                                                                                        8.611560
                                                                                                 8900
               2
                                   51
                                              1186
                                                     32500
                                                                            45.666359
                                                                                      12.241890
                                                                                                 8800
                    pop
                   sport
                                   74
                                              4658
                                                    142228
                                                                            45.503300 11.417840
                                                                                                 4200
                                                    160000
                                                                            40.633171 17.634609
                                   51
                                              2739
                                                                                                 6000
                  lounge
                                                                          1 41.903221 12.495650 5700
               5
                    qoq
                                   73
                                              3074 106880
In [5]: data1 = data.drop(['lat','lon','ID'],axis=1) # umwanted columns removed
```

In [6]: data1

Out[6]:

	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

```
In [7]: data1 = pd.get_dummies(data1)
In [8]: data1.shape
    #data1['model'] = data['model'].map({'lounge':1,'pop':2,'sport':3})
Out[8]: (1538, 8)
```

In [9]: data1

Out[9]:		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 [10]: y = data1['price']
x = data1.drop('price',axis= 1)
```

```
In [11]: y
Out[11]: 0
                  8900
                  8800
          2
                  4200
          3
                  6000
                  5700
          4
          1533
                  5200
          1534
                  4600
          1535
                  7500
          1536
                  5990
          1537
                  7900
          Name: price, Length: 1538, dtype: int64
In [12]: #!pip3 install scikit-learn
In [13]: 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 [14]: x_test.head(5)
Out[14]:
                engine_power age_in_days
                                         km previous_owners model_lounge model_pop model_sport
                        51
                                  3197 120000
                                                         2
                                                                     0
                                                                                          0
            481
                                                                               1
            76
                        62
                                  2101
                                      103000
                                                         1
                                                                     0
                                                                               1
                                                                                          0
           1502
                        51
                                  670
                                       32473
                                                         1
                                                                     1
                                                                               0
                                                                                          0
           669
                        51
                                                         1
                                                                     1
                                                                               0
                                                                                          0
                                  913
                                       29000
           1409
                        51
                                  762
                                       18800
                                                         1
                                                                     1
                                                                               0
                                                                                          0
In [15]: x_train.shape
Out[15]: (1030, 7)
```

```
In [16]: y test.head()
Out[16]: 481
                   7900
          76
                   7900
          1502
                   9400
          669
                   8500
          1409
                   9700
          Name: price, dtype: int64
In [17]: |y_train.shape
Out[17]: (1030,)
In [18]: from sklearn.linear model import LinearRegression
          reg = LinearRegression()
          reg.fit(x_train,y_train)
Out[18]: 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 [19]: ypred=reg.predict(x test)
```

```
In [20]: | ypred
Out[20]: arrav([ 5867,6503378 ,
                                  7133.70142341.
                                                  9866.35776216.
                                                                   9723.28874535.
                                  9654.07582608,
                                                  9673.14563045, 10118.70728123,
                 10039.59101162,
                 9903.85952664,
                                  9351.55828437, 10434.34963575, 7732.26255693,
                                  6565.95240435,
                                                  9662.90103518, 10373.20344286,
                 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 ,
                                                                   8457.38443276,
                                  8258.46551788,
                                                  6279.2040404 ,
                 9773.86444066,
                                  6767.04074749.
                                                  9182.99904787, 10210.05195479,
                 8694.90545226, 10328.43369248,
                                                                   8866.7826029 ,
                                                  9069.05761443,
                 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,
                                                                  8515.83255277,
                 6417.69133992,
                                  9996.97382441,
                                                  9781.18795953,
                 8456.30006203,
                                  6499.76668237,
                                                  7768.57829985,
                                                                  6832.86406122,
                 8347.96113362, 10439.02404036,
                                                  7356.43463051,
                                                                   8562.56562053,
In [21]: from sklearn.metrics import r2 score
         r2 score(v test, vpred)
Out[21]: 0.8415526986865394
In [22]: from sklearn.metrics import mean squared error
         mean squared error(ypred,y test)
Out[22]: 581887.727391353
In [23]: n = 581887.727391353 ** (1/2)
         print(n)
```

762.8156575420782

```
In [24]: y test.head()
Out[24]: 481
                  7900
          76
                  7900
         1502
                  9400
          669
                  8500
          1409
                  9700
         Name: price, dtype: int64
In [26]:
         ypred
Out[26]: array([ 5867.6503378 ,
                                   7133.70142341,
                                                   9866.35776216,
                                                                    9723.28874535,
                 10039.59101162,
                                   9654.07582608,
                                                    9673.14563045, 10118.70728123,
                  9903.85952664,
                                   9351.55828437, 10434.34963575,
                                                                    7732.26255693,
                                   6565.95240435,
                                                    9662.90103518, 10373.20344286,
                  7698.67240131,
                  9599.94844451,
                                   7699.34400418,
                                                   4941.33017994, 10455.2719478,
                 10370.51555682, 10391.60424404,
                                                   7529.06622456,
                                                                    9952.37340054,
                                                    4798.36770637,
                                                                    6953.10376491,
                  7006.13845729,
                                   9000.1780961 ,
                  7810.39767825,
                                   9623.80497535,
                                                   7333.52158317,
                                                                    5229.18705519,
                                   5157.65652129,
                  5398.21541073,
                                                    8948.63632836,
                                                                    5666.62365159,
                  9822.1231461 ,
                                   8258.46551788,
                                                   6279.2040404 ,
                                                                    8457.38443276,
                  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,
                                                                    8515.83255277,
                                   9996.97382441,
                                                    9781.18795953,
                  8456.30006203,
                                   6499.76668237,
                                                    7768.57829985,
                                                                    6832.86406122,
                                                                    8562.56562053,
                  8347.96113362, 10439.02404036,
                                                    7356.43463051,
                  0020 70555100
                                  10025 02571520
                                                    7270 77100022
                                                                    0411 45004006
```

```
In [35]: Results=pd.DataFrame(columns=['Price','Predicted'])
    Results['Price']=y_test
    Results['Predicted']=ypred
    Results.head(20)
```

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	Price	Predicted
481	7900	5867.650338
76	7900	7133.701423
1502	9400	9866.357762
669	8500	9723.288745
1409	9700	10039.591012
1414	9900	9654.075826
1089	9900	9673.145630
1507	9950	10118.707281
970	10700	9903.859527
1198	8999	9351.558284
1088	9890	10434.349636
576	7990	7732.262557
965	7380	7698.672401
1488	6800	6565.952404
1432	8900	9662.901035
380	10500	10373.203443
754	10690	9599.948445
30	6990	7699.344004
49	4300	4941.330180
240	10500	10455.271948

In []: