```
In [2]: import pandas as pd
In [3]: data=pd.read csv("/home/placement/Downloads/fiat500.csv")
In [4]: data.head()
Out[4]:
                model engine_power age_in_days
                                                  km previous owners
                                                                            lat
                                                                                     Ion price
                lounge
                                                                                8.611560
             1
                                51
                                           882
                                                25000
                                                                   1 44.907242
                                                                                         8900
             2
                                51
                                          1186
                                                32500
                                                                   1 45.666359 12.241890
                                                                                         8800
                   pop
                                               142228
                                                                   1 45.503300 11.417840
             3
                  sport
                                74
                                          4658
                                                                                         4200
                                                                   1 40.633171 17.634609
                lounge
                                51
                                          2739
                                               160000
                                                                                         6000
                                73
                                          3074 106880
                                                                   1 41.903221 12.495650 5700
             5
                   pop
In [5]: data1=data.drop(['lat','lon','ID'],axis=1)
In [6]: data1.head()
Out[6]:
             model engine_power age_in_days
                                               km previous_owners price
          0 lounge
                                             25000
                                                                1 8900
                             51
                                        882
                             51
                                       1186
                                             32500
                                                                   8800
               pop
                             74
                                      4658 142228
                                                                1 4200
              sport
                                       2739 160000
          3 lounge
                             51
                                                                1 6000
               pop
                             73
                                       3074 106880
                                                                1 5700
In [7]: data=pd.get dummies(data)
```

```
In [8]: data.shape
 Out[8]: (1538, 11)
In [9]: y=data['price']
In [10]: x=data.drop('price',axis=1)
In [11]: y
Out[11]: 0
                 8900
                 8800
                 4200
         2
                 6000
         4
                 5700
         1533
                 5200
         1534
                 4600
         1535
                 7500
         1536
                 5990
         1537
                 7900
         Name: price, Length: 1538, dtype: int64
In [12]: 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 [13]: x\_test.head(5)

Out[13]:

	ID	engine_power	age_in_days	km	previous_owners	lat	lon	model_lounge	model_pop	model_sport
481	482	51	3197	120000	2	40.174702	18.167629	0	1	0
76	77	62	2101	103000	1	45.797859	8.644440	0	1	0
1502	1503	51	670	32473	1	41.107880	14.208810	1	0	0
669	670	51	913	29000	1	45.778591	8.946250	1	0	0
1409	1410	51	762	18800	1	45.538689	9.928310	1	0	0

In [14]: x\_train.shape

Out[14]: (1030, 10)

In [15]: x\_train.head()

Out[15]:

	ID	engine_power	age_in_days	km	previous_owners	lat	lon	model_lounge	model_pop	model_sport
527	528	51	425	13111	1	45.022388	7.58602	1	0	0
129	130	51	1127	21400	1	44.332531	7.54592	1	0	0
602	603	51	2039	57039	1	40.748241	14.52835	0	1	0
331	332	51	1155	40700	1	42.143860	12.54016	1	0	0
323	324	51	425	16783	1	41.903221	12.49565	1	0	0

In [16]: y\_train.shape

Out[16]: (1030,)

```
In [17]: | y_train.head()
Out[17]: 527
                  9990
          129
                  9500
          602
                  7590
          331
                  8750
          323
                  9100
          Name: price, dtype: int64
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
                10066.02424638. 10430.97776811. 10050.79995384.
                                                                  7801.53792597.
                 8738.32379912,
                                 9963.07184541, 10250.69391036,
                                                                  9856.67153089.
                 8383.84152492,
                                 9307.84587539,
                                                                  9859.23075392,
                                                 8530.90168144.
                                 9744.86150125,
                 9733.54483496,
                                                  6741.410463
                                                                  7342.18893371,
                 8772.20704958,
                                 9959.77345301,
                                                  9692.26944677, 10524.54487623,
                 8221.41396472,
                                 6722.97284178,
                                                  9894.93188478,
                                                                  8849.71168914,
                 9786.53980838, 10262.59139607, 10382.67498044,
                                                                  9988.41681508.
                 9336.80741819,
                                 9902.52039123,
                                                  9109.63147621, 10147.01866123,
                                                  8827.96184211, 10302.33416028,
                 7831.00036415,
                                 6059.56493387,
                                                                  7698.86996869,
                 5660.1705204 , 10068.83508852,
                                                  9595.70115109,
                 9319.54039166.
                                 7421.93077111, 10397.65812756, 10008.49656229,
                10572.26845119,
                                 9890.79746015,
                                                  9995.86970892,
                                                                  6328.88724858,
                10434.22517244,
                                 9981.92833783, 10478.31842709,
                                                                  9584.67757276,
                 9795.59966427,
                                 6215.62308925,
                                                  8012.67431998, 10289.49085168,
                 6351.65397303,
                                 7447.35295678,
                                                                  6753.92994153,
                                                  9954.0491226 ,
                 7806.68212311, 5292.72896136,
                                                  4479.07164048,
                                                                  8743.32482334,
                 6930.07078154, 7474.31727616,
                                                  6868.13323766,
                                                                  7152.35036884,
                 9982.54626745, 8788.00494177,
                                                  9330.50348958, 10377.44826079,
                                                                  5814.72404507,
                10022.1505514 , 10337.05151745,
                                                  9800.2030809 ,
                 9709 16172189
                                 7742 98208443
                                                  5572.51518045.
                                                                  4925,50951785
In [21]: from sklearn.metrics import r2 score
         r2 score(y test,ypred)
Out[21]: 0.8428319728488683
In [22]: from sklearn.metrics import mean squared error
         mean squared error(ypred,y test)
Out[22]: 577189.6736608233
In [23]: import math
         a=577189.6736608233
         print(math.sqrt(a))
```

759.7300005007195

```
In [24]: Results=pd.DataFrame(columns=['price','predicate'])
    Results['price']=y_test
    Results['predicate']=ypred
    Results=Results.reset_index()
    Results['Id']=Results.index
    Results.head(15)
```

## Out[24]:

	index	price	predicate	Id
0	481	7900	5819.193088	0
1	76	7900	7248.829142	1
2	1502	9400	9741.893697	2
3	669	8500	9798.980331	3
4	1409	9700	10055.006246	4
5	1414	9900	9551.495568	5
6	1089	9900	9758.017439	6
7	1507	9950	10122.977837	7
8	970	10700	9654.966181	8
9	1198	8999	9251.140326	9
10	1088	9890	10478.095123	10
11	576	7990	7807.300526	11
12	965	7380	7705.158738	12
13	1488	6800	6295.632449	13
14	1432	8900	9545.404863	14

```
In [26]: Results['diff_price']=Results.apply(lambda row:row.price-row.predicate,axis=1)
```

## In [27]: Results.head(15)

## Out[27]:

	index	price	predicate	ld	diff_price
0	481	7900	5819.193088	0	2080.806912
1	76	7900	7248.829142	1	651.170858
2	1502	9400	9741.893697	2	-341.893697
3	669	8500	9798.980331	3	-1298.980331
4	1409	9700	10055.006246	4	-355.006246
5	1414	9900	9551.495568	5	348.504432
6	1089	9900	9758.017439	6	141.982561
7	1507	9950	10122.977837	7	-172.977837
8	970	10700	9654.966181	8	1045.033819
9	1198	8999	9251.140326	9	-252.140326
10	1088	9890	10478.095123	10	-588.095123
11	576	7990	7807.300526	11	182.699474
12	965	7380	7705.158738	12	-325.158738
13	1488	6800	6295.632449	13	504.367551
14	1432	8900	9545.404863	14	-645.404863

## In [ ]: