```
In [126]: import pandas as pd
In [127]: | data=pd.read_csv("fiat500.csv")
In [128]: data
Out[128]:
                     ID model engine_power age_in_days
                                                            km previous_owners
                                                                                       lat
                                                                                                lon price
                0
                      1 lounge
                                         51
                                                    882
                                                          25000
                                                                              1 44.907242
                                                                                            8.611560
                                                                                                     8900
                                                   1186
                                         51
                                                          32500
                                                                              1 45.666359 12.241890
                                                                                                     8800
                1
                           pop
                                                         142228
                                                                              1 45.503300 11.417840
                2
                                         74
                                                   4658
                                                                                                     4200
                          sport
                                                         160000
                3
                                         51
                                                   2739
                                                                              1 40.633171 17.634609
                                                                                                     6000
                      4 lounge
                                         73
                                                   3074 106880
                                                                              1 41.903221 12.495650
                                                                                                     5700
                           pop
             1533
                   1534
                                                   3712
                                                         115280
                                                                              1 45.069679
                                                                                           7.704920
                                                                                                     5200
                                         51
                          sport
                  1535
                                                         112000
                                                                              1 45.845692
                                                                                           8.666870
                                                   3835
                                                                                                     4600
             1534
                        lounge
                                         74
                                                   2223
                                                          60457
                  1536
             1535
                           pop
                                         51
                                                                              1 45.481541
                                                                                           9.413480
                                                                                                     7500
             1536
                  1537
                        lounge
                                         51
                                                   2557
                                                          80750
                                                                              1 45.000702
                                                                                           7.682270
                                                                                                     5990
             1537 1538
                                         51
                                                          54276
                                                   1766
                                                                              1 40.323410 17.568270
                                                                                                     7900
                           pop
            1538 rows × 9 columns
In [129]: data=data.drop(['lat','lon','ID'],axis=1)
```

In [130]: data

Out[130]:

	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 [131]: data2=pd.get_dummies(data)

In [132]: data

Out[132]:

	model	engine_power	age_in_days	km	previous_owners	price
0	lounge	51	882	25000	1	8900
1	рор	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 [133]: data

Out[133]:

	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 [134]: data2.shape
```

#data['model']=data['model']

Out[134]: (1538, 8)

In [135]: data2

Out I	[125]
out	[TOO]

	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 [136]: y=data2['price']
x=data2.drop('price',axis=1)
```

```
In [137]: y
Out[137]: 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 [138]: 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 [139]: x test.head(5)
Out[139]:
                 engine_power age_in_days
                                           km previous_owners model_lounge model_pop model_sport
                         51
                                   3197 120000
                                                           2
                                                                       0
             481
                                                                                 1
                                                                                            0
                                   2101 103000
             76
                                                           1
                         62
                                                                       0
                                                                                 1
                                                                                            0
            1502
                         51
                                    670
                                         32473
                                                           1
                                                                       1
                                                                                 0
                                                                                            0
             669
                         51
                                         29000
                                                           1
                                                                       1
                                                                                 0
                                                                                            0
                                    913
                                         18800
            1409
                         51
                                    762
                                                           1
                                                                       1
                                                                                 0
                                                                                            0
In [140]: x_train.shape
```

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Out[140]: (1030, 7)

```
In [141]: y_train
Out[141]: 527
                     9990
           129
                     9500
           602
                     7590
           331
                     8750
           323
                     9100
           1130
                    10990
           1294
                     9800
           860
                     5500
           1459
                     9990
           1126
                     8900
           Name: price, Length: 1030, dtype: int64
In [142]: 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
Out[142]: 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 [143]: |ypred=reg.predict(x test)
```

```
In [144]: ypred
                                   7568.91955482,
                                                   6804.97705225,
                                                                    8065.01292384,
                 10582.50828537.
                 10310.29143419.
                                   8836.34894739.
                                                   8390.05091229.
                                                                    9582.13932508.
                  9745.34784981. 10045.45021387. 10294.09872915.
                                                                    7145.15315349.
                  9727.85493167,
                                   6281.78952194,
                                                   7901.36245623,
                                                                    9387.9203723 ,
                                   9351.49777725,
                                                   9980.70844784, 10094.79341516,
                  5039.55649797,
                  6359.24321991,
                                   9856.10227211,
                                                                    5234.05388382.
                                                   9099.07023804.
                  5534.45288323,
                                   4495.02309231, 10199.78432943, 10024.87037067,
                  5465.58034188,
                                   8520.72057674,
                                                   7034.71038647, 10054.65061446,
                 10191.12067767.
                                   6008.34860428,
                                                   9748.18097947,
                                                                    9669.4333196 .
                  9145.3756075 ,
                                   9175.66562699, 10087.86753845,
                                                                    9825.02990067,
                  7340.29803785,
                                   5083.8487301 ,
                                                    9441.50914802, 10243.05490667,
                  5556.42300245, 10676.01945733,
                                                   6126.99295838,
                                                                    9845.16661356,
                                                                    9938.82104889,
                  9850.77978959,
                                   7840.83596305,
                                                   6552.05146566,
                  8327.79232274,
                                   9119.62204137,
                                                   6111.83787367, 10410.00504522,
                  6360.97695249,
                                   8601.59209793,
                                                   8377.80258216,
                                                                    9803.81343895,
                  8285.09831762, 10091.75635129, 10003.86694939, 10028.60283146,
                                   8552.21002673,
                                                                    9381.22662706,
                 10354.61956534,
                                                   6726.65446676,
                  6520.9999373 , 10352.85155564,
                                                    9063.7534579 , 10456.89121831,
                  9127.72470241.
                                   9952.37340054.
                                                   8376.6975881 .
                                                                    9220.36267675,
In [145]: #savedmodel=pickle.load(open(filename, 'rb'))
          \#x \ test=[[1,75,1062,8000,1]]
          #savedmodel.predict(x test)
In [146]: from sklearn.metrics import r2 score
          r2 score(v test, vpred)
Out[146]: 0.8415526986865394
 In [ ]:
In [147]: from sklearn.metrics import mean squared error
          mean squared error(ypred,y test)
Out[147]: 581887.727391353
```

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```
In [148]: #from sklear.metrics import accuracy score
          #accuracy score(y test,ypred)
In [151]: #results=pd.dataframe(coloumns=['actual', 'predicted'])
          #Results['Acutal']=y test
          Results=pd.DataFrame(columns=['Price', 'Predicted'])
          Results['Price']=y test
          Results['Predicted']=ypred
          #result['km']=x test['km']
          Results=Results.reset index()
          Results['Id']=Results.index
          Results.head(15)
Out[151]:
```

	index	Price	Predicted	ld
0	481	7900	5867.650338	0
1	76	7900	7133.701423	1
2	1502	9400	9866.357762	2
3	669	8500	9723.288745	3
4	1409	9700	10039.591012	4
5	1414	9900	9654.075826	5
6	1089	9900	9673.145630	6
7	1507	9950	10118.707281	7
8	970	10700	9903.859527	8
9	1198	8999	9351.558284	9
10	1088	9890	10434.349636	10
11	576	7990	7732.262557	11
12	965	7380	7698.672401	12
13	1488	6800	6565.952404	13
14	1432	8900	9662.901035	14

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