```
In [70]: import pandas as pd
In [71]: |data=pd.read_csv("fiat500.csv")
In [72]: data
Out[72]:
                   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
                                        51
                                                  1186
                                                         32500
                                                                             1 45.666359 12.241890
                                                                                                    8800
               1
                         pop
                                                       142228
                                                                            1 45.503300 11.417840
               2
                                        74
                                                  4658
                                                                                                    4200
                         sport
               3
                                        51
                                                  2739
                                                       160000
                                                                             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
                                                                                          8.666870
                                                  3835
                                                                             1 45.845692
                                                                                                    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 [73]: data=data.drop(['lat','lon','ID',],axis=1)
```

In [74]: data

Out[74]:

	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 [75]: data=pd.get_dummies(data)
In [76]: data.shape
    #data['model'] =data['model'].map({'lounge':1,'pop':2,'sport':3})
Out[76]: (1538, 8)
```

In [77]: data

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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 [78]: y=data['price']
```

```
In [80]: y
Out[80]: 0
                 8900
                 8800
         2
                 4200
         3
                 6000
         4
                 5700
         1533
                 5200
         1534
                 4600
         1535
                 7500
         1536
                 5990
         1537
                 7900
         Name: price, Length: 1538, dtype: int64
         !pip install scikit-learn
In [81]:
         Requirement already satisfied: scikit-learn in ./anaconda3/lib/python3.10/site-packages (1.2.1)
         Requirement already satisfied: threadpoolctl>=2.0.0 in ./anaconda3/lib/python3.10/site-packages (from sciki
         t-learn) (2.2.0)
         Requirement already satisfied: numpy>=1.17.3 in ./anaconda3/lib/python3.10/site-packages (from scikit-lear
         n) (1.23.5)
         Requirement already satisfied: scipy>=1.3.2 in ./anaconda3/lib/python3.10/site-packages (from scikit-learn)
         (1.10.0)
         Requirement already satisfied: joblib>=1.1.1 in ./anaconda3/lib/python3.10/site-packages (from scikit-lear
         n) (1.1.1)
In [82]: 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 [83]: x test.head(5)
Out[83]:
                engine power age in days
                                              previous owners model lounge model pop model sport
                                                          2
            481
                         51
                                  3197 120000
                                                                      0
                                                                                1
                                                                                           0
            76
                         62
                                  2101 103000
                                                          1
                                                                      0
                                                                                1
                                                                                           0
           1502
                         51
                                   670
                                        32473
                                                          1
                                                                      1
                                                                                0
                                                                                           0
            669
                         51
                                   913
                                        29000
                                                          1
                                                                      1
                                                                                0
                                                                                           0
           1409
                         51
                                   762
                                        18800
                                                          1
                                                                      1
                                                                                0
                                                                                           0
In [84]: x train.shape
Out[84]: (1030, 7)
In [85]: y_train
Out[85]: 527
                    9990
          129
                    9500
          602
                    7590
                    8750
          331
          323
                    9100
                   . . .
          1130
                   10990
          1294
                    9800
          860
                    5500
          1459
                    9990
          1126
                    8900
          Name: price, Length: 1030, dtype: int64
In [86]: 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[86]: 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.

```
ypred=reg.predict(x test)
In [871:
In [88]:
         ypred
Out[88]: array([ 5867.6503378 ,
                                 7133.70142341,
                                                                  9723.28874535,
                                                  9866.35776216,
                10039.59101162,
                                 9654.07582608,
                                                  9673.14563045, 10118.70728123,
                                 9351.55828437, 10434.34963575, 7732.26255693,
                 9903.85952664,
                                                  9662.90103518, 10373.20344286,
                 7698.67240131,
                                 6565.95240435,
                 9599.94844451,
                                 7699.34400418,
                                                  4941.33017994, 10455.2719478 ,
                10370.51555682, 10391.60424404,
                                                  7529.06622456,
                                                                  9952.37340054,
                 7006.13845729,
                                 9000.1780961 ,
                                                  4798.36770637,
                                                                  6953.10376491,
                 7810.39767825,
                                 9623.80497535,
                                                  7333.52158317,
                                                                  5229.18705519,
                 5398.21541073,
                                 5157.65652129,
                                                  8948.63632836,
                                                                  5666.62365159,
                                                                  8457.38443276,
                 9822.1231461 ,
                                 8258.46551788,
                                                  6279.2040404 ,
                                 6767.04074749,
                 9773.86444066,
                                                  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.
                                                                  7197.84932877,
                                                  9795.31884316,
                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 [89]: from sklearn .metrics import r2 score
         r2 score(y test,ypred)
Out[89]: 0.8415526986865394
In [90]: from sklearn.metrics import mean squared error
         mean squared error(ypred,y test)
Out[90]: 581887.727391353
 In [ ]: #from sklearn.metrics import accuracy score
         #accuracy score(y test)
```

```
In [92]:
    Results= pd.DataFrame(columns=['price','predicted'])
    Results['price']=y_test
    Results['predicted']=ypred
    Results=Results.reset_index()
    Results['Id']=Results.index
    Results.head(15)
```

Out[92]:

	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

```
In [ ]:
In [ ]:
```

In	[]:	
In	[]:	
In	[]:	
In	[]:	
In	[1:	
In	[]:	
In	[]:	
In	[]:	
In	[1:	
In	[1:	