```
import pandas as pd
In [191]:
In [192]: data=pd.read csv("/home/placement/Downloads/fiat500.csv") #reads the file
In [193]: data.describe()
Out[193]:
                             ID engine_power age_in_days
                                                                    km previous_owners
                                                                                                 lat
                                                                                                             lon
                                                                                                                         price
              count 1538.000000
                                  1538.000000
                                              1538.000000
                                                            1538.000000
                                                                             1538.000000
                                                                                         1538.000000
                                                                                                     1538.000000
                                                                                                                  1538.000000
                     769.500000
                                              1650.980494
                                                            53396.011704
                                                                                           43.541361
                                                                                                                  8576.003901
                                    51.904421
                                                                                1.123537
                                                                                                       11.563428
              mean
                                                                                                        2.328190
                std
                     444.126671
                                     3.988023
                                              1289.522278
                                                            40046.830723
                                                                                0.416423
                                                                                            2.133518
                                                                                                                  1939.958641
               min
                       1.000000
                                    51.000000
                                               366.000000
                                                            1232.000000
                                                                                1.000000
                                                                                           36.855839
                                                                                                        7.245400
                                                                                                                  2500.000000
                     385.250000
                                    51.000000
                                                            20006.250000
                                                                                1.000000
                                                                                           41.802990
                                                                                                        9.505090
                                                                                                                  7122.500000
               25%
                                               670.000000
               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 [194]: data1=data.drop(['ID','lat','lon'],axis=1) #removes the specific columns from dataframe
```

In [195]: data1

Out[195]:

_		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 [196]: data2=pd.get_dummies(data1) #encodes the string into bits

In [197]: data2

Out[197]:

	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 [198]: data2.shape #shows num of rows and columns
Out[198]: (1538, 8)
```

```
In [199]: y=data2['price']#predicted value removed feom data frame
x=data2.drop(['price'],axis=1)
```

```
In [200]: y #prices only will display
Out[200]: 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
```

In [201]: x #no prices

Out[201]:		engine_power	age_in_days	km	previous_owners	model_lounge	model_pop	model_sport
	0	51	882	25000	1	1	0	0
	1	51	1186	32500	1	0	1	0
	2	74	4658	142228	1	0	0	1
	3	51	2739	160000	1	1	0	0
	4	73	3074	106880	1	0	1	0
	1533	51	3712	115280	1	0	0	1
	1534	74	3835	112000	1	1	0	0
	1535	51	2223	60457	1	0	1	0
	1536	51	2557	80750	1	1	0	0
	1537	51	1766	54276	1	0	1	0

1538 rows × 7 columns

In [202]: !pip3 install scikit-learn #to install sklearn library

Requirement already satisfied: scikit-learn in ./anaconda3/lib/python3.10/site-packages (1.2.1)

Requirement already satisfied: joblib>=1.1.1 in ./anaconda3/lib/python3.10/site-packages (from scikit-lear n) (1.1.1)

Requirement already satisfied: scipy>=1.3.2 in ./anaconda3/lib/python3.10/site-packages (from scikit-learn) (1.10.0)

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)

In [203]: pm sklearn.model_selection import train_test_split
train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.33,random_state=42)#splitting into training and

In [204]: x test.head(5)#shows the testing column values

Out[204]: km previous owners model lounge model pop model sport engine power age in days 481 51 3197 120000 2 0 1 0 76 2101 103000 1 0 62 1 0 1502 51 670 32473 1 1 0 0 669 51 29000 1 1 0 0 913 1409 51 762 18800 1 1 0 0

In [205]: y_test.head(5)

Out[205]: 481 7900 76 7900 1502 9400 669 8500 1409 9700

Name: price, dtype: int64

In [206]: x_train.head(5)

Out[206]:

	engine_power	age_in_days	km	previous_owners	model_lounge	model_pop	model_sport
527	51	425	13111	1	1	0	0
129	51	1127	21400	1	1	0	0
602	51	2039	57039	1	0	1	0
331	51	1155	40700	1	1	0	0
323	51	425	16783	1	1	0	0

```
In [207]: y_train.head(5)
```

Out[207]: 527

527 9990

129 9500

602 7590 331 8750

323 9100

Name: price, dtype: int64

In [208]: 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[208]: 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 [209]: ypred=reg.predict(x_test) #prediction of values(x_tesy*reg)

```
In [210]: ypred
                  8398.59735084.
                                  9680.77538859.
                                                   4334.81943405. 10015.00600846.
                  9850.72458719,
                                  7864.73798641, 10072.71245374, 10552.64805598,
                                  6861.80736606,
                 10253.47474908,
                                                   6484.22649656, 10374.62123623,
                  8426.37409382.
                                  5447.47569851,
                                                   9914.20077691. 4687.39013431.
                  7885.32100747,
                                  5431.00822998,
                                                   9911.86294348, 10390.16991322,
                  9680.84745901,
                                  8844.57815539,
                                                   7764.08471024, 4257.54640953,
                  9882.76503303, 10341.35258769,
                                                   5736.4484335 , 10179.87154436,
                                  7997.3181334 ,
                                                                   9894.57834738,
                  9501.423448
                                                   5532.33458288,
                 10437.97459358,
                                  6381.35845844,
                                                   9591.23555726,
                                                                   9574.27908517,
                 10322.30715736,
                                  9501.22785499,
                                                   9789.955758
                                                                   9593.26549752,
                  6775.82788536,
                                  7915.34831306, 10389.98590521, 10351.58343315,
                  7381.32686464,
                                  9966.53983093, 10430.87188433, 10554.43156462,
                 10285.85574963, 10035.88086558,
                                                   9526.63034431,
                                                                   7742.78157141,
                  9297.64938364, 10051.42272678, 10004.81256571,
                                                                   9985.84167026,
                  9374.6573594 ,
                                  9561.57499854,
                                                   9754.94184269,
                                                                   9819.85893758,
                  8780.31447831,
                                  6255.99008069,
                                                   6281.53627686,
                                                                   8190.88781577,
                                                   6850.70237466,
                                                                   5511.29438169,
                  8588.91394592,
                                  6566.97963218,
                  8119.97866315,
                                  9847.74830838,
                                                   7775.93862032,
                                                                   9875.05509733,
                 10121.29366536,
                                  5791.92464084,
                                                   9835.42728501, 10043.91426822,
                  8027 28015259
                                   4527 22080416
                                                  10609 02444098
                                                                    3808 29240951
In [211]: from sklearn.metrics import r2 score#efficiency
          r2 score(y test, ypred) #ypred is predicted value
Out[211]: 0.8415526986865394
In [212]: from sklearn.metrics import mean squared error#to calculate rms value
In [213]: mean squared error(ypred,y test)
Out[213]: 581887.727391353
In [214]: import math
```

```
In [215]: n=581887.727391353 #to find square root
math.sqrt(n)

Out[215]: 762.8156575420782

In [216]: #from sklearn.metrics import accuracy_score
#accuracy_score(y_test,ypred)

In [224]: results=pd.DataFrame(columns=['Price','Predicted'])
results['Price']=y_test
results['Predicted']=ypred
results['Predicted']=ypred
results['Id']=results.index
results['Id']=results.index
results.head(15)
Out[224]: index Price Predicted Id
```

		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 [226]: results['final']=results.apply(lambda row:row.Price - row.Predicted,axis=1)#difference between prices and pr

In [227]: results.head(15)

Out[227]:

	index	Price	Predicted	ld	final
0	481	7900	5867.650338	0	2032.349662
1	76	7900	7133.701423	1	766.298577
2	1502	9400	9866.357762	2	-466.357762
3	669	8500	9723.288745	3	-1223.288745
4	1409	9700	10039.591012	4	-339.591012
5	1414	9900	9654.075826	5	245.924174
6	1089	9900	9673.145630	6	226.854370
7	1507	9950	10118.707281	7	-168.707281
8	970	10700	9903.859527	8	796.140473
9	1198	8999	9351.558284	9	-352.558284
10	1088	9890	10434.349636	10	-544.349636
11	576	7990	7732.262557	11	257.737443
12	965	7380	7698.672401	12	-318.672401
13	1488	6800	6565.952404	13	234.047596
14	1432	8900	9662.901035	14	-762.901035

In []: