```
In [1]: import pandas as pd
data=pd.read_csv("/home/placement/Downloads/fiat500.csv")
```

In [2]: data.describe()

Out[2]:

	ID	engine_power	age_in_days	km	previous_owners	lat	lon	price
cou	nt 1538.000000	1538.000000	1538.000000	1538.000000	1538.000000	1538.000000	1538.000000	1538.000000
mea	n 769.500000	51.904421	1650.980494	53396.011704	1.123537	43.541361	11.563428	8576.003901
s	td 444.126671	3.988023	1289.522278	40046.830723	0.416423	2.133518	2.328190	1939.958641
m	in 1.000000	51.000000	366.000000	1232.000000	1.000000	36.855839	7.245400	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
m	ax 1538.000000	77.000000	4658.000000	235000.000000	4.000000	46.795612	18.365520	11100.000000

```
In [3]: datal=data.drop(['lat','lon','ID'],axis=1)
```

In [4]: data1

Out[4]:

model	engine_power	age_in_days	km	previous_owners	price
lounge	51	882	25000	1	8900
pop	51	1186	32500	1	8800
sport	74	4658	142228	1	4200
lounge	51	2739	160000	1	6000
pop	73	3074	106880	1	5700
sport	51	3712	115280	1	5200
lounge	74	3835	112000	1	4600
pop	51	2223	60457	1	7500
lounge	51	2557	80750	1	5990
pop	51	1766	54276	1	7900
	lounge pop sport lounge pop sport lounge pop lounge	lounge 51 pop 51 sport 74 lounge 51 pop 73 sport 51 lounge 74 pop 51 lounge 51	lounge 51 882 pop 51 1186 sport 74 4658 lounge 51 2739 pop 73 3074 sport 51 3712 lounge 74 3835 pop 51 2223 lounge 51 2557	lounge 51 882 25000 pop 51 1186 32500 sport 74 4658 142228 lounge 51 2739 160000 pop 73 3074 106880 sport 51 3712 115280 lounge 74 3835 112000 pop 51 2223 60457 lounge 51 2557 80750	lounge 51 882 25000 1 pop 51 1186 32500 1 sport 74 4658 142228 1 lounge 51 2739 160000 1 pop 73 3074 106880 1 sport 51 3712 115280 1 lounge 74 3835 112000 1 pop 51 2223 60457 1 lounge 51 2557 80750 1

1538 rows × 6 columns

In [5]: data1.shape

Out[5]: (1538, 6)

In [6]: data1=pd.get_dummies(data1)#converting the string column model into 3 individual columns as digits

In [7]: data1

Out[7]:

	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 [8]: data1.shape
```

Out[8]: (1538, 8)

```
In [9]: y=datal['price'] #adding to seperate dataframe the value, we want to predict x=datal.drop('price',axis=1)#removing the value we want to predict from the original dataframe
```

```
In [10]: y
Out[10]: 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 [11]: #divide data into training and testing
         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 [12]: x_test

Out[12]:

	engine_power	age_in_days	km	previous_owners	model_lounge	model_pop	model_sport
481	51	3197	120000	2	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
291	51	701	22000	1	1	0	0
596	51	3347	85500	1	0	1	0
1489	51	366	22148	1	0	1	0
1436	51	1797	61000	1	1	0	0
575	51	366	19112	1	1	0	0

508 rows × 7 columns

In [13]: x_train

Out[13]:

	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
1130	51	1127	24000	1	1	0	0
1294	51	852	30000	1	1	0	0
860	51	3409	118000	1	0	1	0
1459	51	762	16700	1	1	0	0
1126	51	701	39207	1	1	0	0

1030 rows × 7 columns

In [14]: x_test.head(5)

Out[14]:

_		engine_power	age_in_days	km	previous_owners	model_lounge	model_pop	model_sport
-	481	51	3197	120000	2	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 [15]: y train.head()
Out[15]: 527
                9990
         129
                9500
         602
                7590
         331
                8750
         323
                9100
         Name: price, dtype: int64
In [16]: y_train
Out[16]: 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 [17]: y_test
Out[17]: 481
                   7900
         76
                   7900
         1502
                   9400
         669
                   8500
         1409
                   9700
                  . . .
         291
                 10900
         596
                   5699
         1489
                   9500
         1436
                   6990
         575
                 10900
         Name: price, Length: 508, dtype: int64
```

```
In [18]: from sklearn.linear model import LinearRegression
         reg=LinearRegression()#creating object of Linear Regression
         reg.fit(x train, v train)#training and fitting LR object using training data
Out[18]:
          ▼ LinearRegression
          LinearRedression()
         vpred=req.predict(x test)
In [19]:
In [20]: vpred
                                                                   9952.37340054,
                 10370.51555682, 10391.60424404,
                                                   7529.06622456.
                  7006.13845729,
                                  9000.1780961 ,
                                                   4798.36770637,
                                                                   6953.10376491,
                  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,
                                                   9069.05761443,
                                                                   8866.7826029 ,
                  7058.39787506,
                                  9073.33877162,
                                                   9412.68162121, 10293.69451263,
                                  6748.5794244 ,
                                                                   9354.09969973,
                 10072.49011135,
                                                   9785.95841801,
                  9507.9444386 , 10443.01608254,
                                                   9795.31884316,
                                                                   7197.84932877,
                 10108.31707235,
                                  7009.6597206 ,
                                                   9853.90699412,
                                                                   7146.87414965,
                  6417.69133992,
                                  9996.97382441,
                                                   9781.18795953,
                                                                   8515.83255277,
                 8456.30006203,
                                  6499.76668237,
                                                                   6832.86406122,
                                                   7768.57829985,
                  8347.96113362, 10439.02404036,
                                                   7356.43463051,
                                                                   8562.56562053,
                  9820.78555199, 10035.83571539,
                                                   7370.77198022,
                                                                   9411.45894006,
                 10352.85155564,
                                  8045.21588007, 10446.80664758,
                                                                   3736.20118868,
                 10348.63930496, 10435.96627494,
                                                   6167.80169017, 10390.11317804,
                                  9116.4755691 ,
                                                 10484.52829
                                                                   9335.69889855,
                  6527.69471073,
                  6709.57413543.
                                  3390.72353093. 10106.33753331.
                                                                   9792.46732008
                                                                   0060 00050440
In [21]: from sklearn.metrics import r2 score
         r2 score(y test,ypred)
Out[21]: 0.8415526986865394
```

```
In [22]: from sklearn.metrics import mean_squared_error #calculating MSE
mean_squared_error(ypred,y_test)

Out[22]: 581887.727391353

In [23]: import math
print(math.sqrt(581887.727391353))
```

762.8156575420782

```
In [24]: 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[24]:

	index	Price	Predicted	Id	
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 [25]: Results['diff']=Results.apply(lambda row: row.Price - row.Predicted,axis=1)
```

In [26]: Results.head(10)

Out[26]:

	index	Price	Predicted	ld	diff
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

In []: