```
In [151]: import pandas as pd
In [152]: data=pd.read csv("/home/palcement/Downloads/fiat500.csv")
In [153]: data.head(10)
Out[153]:
                ID model engine_power age_in_days
                                                        km previous_owners
                                                                                                 price
                                                                                  lat
                                                                                           lon
                1 lounge
                                    51
                                                     25000
                                                                                       8.611560
             0
                                                882
                                                                         1 44.907242
                                                                                                 8900
                2
                                    51
                                               1186
                                                     32500
                                                                         1 45.666359 12.241890
                                                                                                 8800
             1
                      pop
                                               4658
                                                    142228
                                                                         1 45.503300 11.417840
             2
                 3
                     sport
                                    74
                                                                                                 4200
                                                    160000
                                               2739
                                                                         1 40.633171 17.634609
                    lounge
                                    51
                                                                                                 6000
                                    73
                                               3074
                                                    106880
                                                                         1 41.903221 12.495650
                                                                                                 5700
                 5
                      pop
                                                     70225
                                                                         1 45.000702
                                                                                      7.682270
                      pop
                                    74
                                               3623
                                                                                                 7900
                                                731
                                                     11600
                    lounge
                                    51
                                                                         1 44.907242
                                                                                       8.611560 10750
                                               1521
                                                     49076
                                                                         1 41.903221 12.495650
                    lounge
                                    51
                                                                                                 9190
                                               4049
                                                      76000
                                                                         1 45.548000 11.549470
                     sport
                                    73
                                                                                                 5600
                                    51
                                               3653
                                                     89000
                                                                         1 45.438301 10.991700
             9 10
                     sport
                                                                                                 6000
```

datal=data.drop(['ID','lat','lon'],axis=1) #To drop the colums in a data frame

In [154]:

In [155]: data1

Out[155]:

	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 [156]: data1=pd.get_dummies(data1) ##This command is used to encode the string in to the numbers.
                                ## Here we observe that the lounge is converted to the 1 0 0
                                ## Here we observe that the pop is converted to the 0 1
                                ##Here we observe that the sport is converted to the 0 0
                                                                                        1
```

In [157]: data1

Λ.,	4.1	115	71
υu	u	TO	/ I

	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 [158]: datal.shape #It will show the no of rows and the columns
# After removing the columns of ID, lat, lon
```

Out[158]: (1538, 8)

In [159]: data.shape

Out[159]: (1538, 9)

In [160]: #asdf;lkj asdf ;lkj asdf ;lkj

```
In [161]: y=data1['price'] #in the dataset named as fiat500, we simply only take the price as seperate and store the
          X=datal.drop('price',axis=1) # in the data frame we removed the price column and remaining data stored in the
In [162]: y
Out[162]: 0
                  8900
                  8800
                  4200
          2
          3
                  6000
                  5700
          4
          1533
                  5200
          1534
                  4600
          1535
                  7500
          1536
                  5990
          1537
                  7900
```

Name: price, Length: 1538, dtype: int64

In [163]: X

Out[163]:

	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 [164]: #!pip3 install scikit-learn to install skleran package run this command
```

```
In [165]: 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 [166]: X_test.head(5)

Out[166]:

	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 [167]: X_train.head()

Out[167]:

	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 [168]: y_test.head(5)

Out[168]: 481

 481
 7900

 76
 7900

 1502
 9400

 669
 8500

1409 9700

Name: price, dtype: int64

```
In [169]: y train.head(8)
Out[169]: 527
                     9990
           129
                     9500
           602
                     7590
           331
                     8750
           323
                     9100
           1358
                   10900
           522
                   10800
           584
                     9999
           Name: price, dtype: int64
In [170]: from sklearn.linear model import LinearRegression
           reg=LinearRegression() #creating of Linear Regression
           reg.fit(X train,y train) #training and fitting LR object using training data
Out[170]: 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 [171]: ypred=reg.predict(X test)
```

```
In [172]: ypred
                                  9910.2/303/91, 1020/.43003992,
                                                                  9904.3213209 .
                  0040.0039/200,
                                                                  9743.68712672.
                  8403.51255128.
                                  9345.81907605.
                                                  8521.46225147.
                  9791.34520178,
                                 9779.16293972,
                                                  6753.27416058, 7354.16762745,
                  8760.24542762,
                                  9923.66596418,
                                                  9812.92276721, 10466.90125415,
                                  6659.46839415.
                                                  9987.65677522.
                                                                  8866.7826029 .
                  8163.46726237,
                  9952.37340054, 10187.72427693, 10231.39378767, 10091.11325493,
                  9365.98570732, 10009.10088406,
                                                  9141.00566394, 10099.11667176,
                  7803.77049829,
                                  6009.84398185,
                                                  8800.33824151, 10237.60733785,
                  5609.98366311, 10097.61555355,
                                                  9684.99946572, 7644.67379732,
                  9276.37891542,
                                 7371.5492091 , 10287.98873148 , 10067.26428381 ,
                 10552.64805598,
                                  9966.72383894, 10068.46126756,
                                                                  6232.53552963,
                 10584.55044373.
                                  9965.98687522. 10529.44404458.
                                                                  9602.67646085.
                  9665.77720284,
                                  6186.06948587,
                                                  8073.87436253, 10345.58323918,
                  6344.74803956,
                                  7361.62678204, 10058.57116223,
                                                                  6792.219309
                                                                  8709.36468047,
                  7897.72464823,
                                  5261.45936067,
                                                  4540.24137423,
                  6882.0117409 ,
                                  7406.73353952,
                                                  6795.61189392,
                                                                  7047.27998963,
                                  8856.93910595,
                                                  9378.02074127, 10389.561154
                  9945.33400083,
                 10092.46332921, 10381.52000388,
                                                  9723.92466625,
                                                                  5996.3331428 ,
                  9786.14866981, 7708.49649098,
                                                                  4932.92788329,
                                                  5583.48163469,
                  9856.66053994,
                                  9236.22981005, 10092.64052142,
                                                                  6256.43516278,
In [174]: from sklearn.metrics import r2 score #model efficinecy step
                                                                          #v test is the actual price
                                                                                                         #v predit is
          r2_score(y test,ypred)
Out[174]: 0.8415526986865394
In [176]: from sklearn.metrics import mean squared error ##Calulating mean square error
          mean squared error(y test,ypred)
Out[176]: 581887.727391353
In [181]: import math
          a=581887.727391353
          math.sqrt(a)
Out[181]: 762.8156575420782
```

localhost:8888/notebooks/19june.ipynb#

```
In [183]:
          y test.head()
Out[183]:
          481
                   7900
          76
                   7900
          1502
                   9400
          669
                   8500
          1409
                   9700
          Name: price, dtype: int64
          ypred
In [186]:
                                   9916.27565791, 10287.45603992,
                                                                    9964.3213269 ,
                   8840.08397206,
                  8403.51255128,
                                   9345.81907605,
                                                    8521.46225147,
                                                                    9743.68712672,
                                   9779.16293972,
                                                    6753.27416058,
                   9791.34520178,
                                                                    7354.16762745,
                   8760.24542762,
                                   9923.66596418,
                                                    9812.92276721, 10466.90125415,
                   8163.46726237,
                                   6659.46839415,
                                                    9987.65677522,
                                                                    8866.7826029 ,
                   9952.37340054, 10187.72427693, 10231.39378767, 10091.11325493,
                  9365.98570732, 10009.10088406,
                                                    9141.00566394, 10099.11667176,
                   7803.77049829,
                                   6009.84398185,
                                                    8800.33824151, 10237.60733785,
                   5609.98366311, 10097.61555355,
                                                    9684.99946572,
                                                                    7644.67379732,
                   9276.37891542,
                                   7371.5492091 ,
                                                   10287.98873148, 10067.26428381,
                  10552.64805598,
                                   9966.72383894, 10068.46126756,
                                                                     6232.53552963,
                 10584.55044373,
                                   9965.98687522, 10529.44404458,
                                                                    9602.67646085,
                                                    8073.87436253, 10345.58323918,
                   9665.77720284,
                                   6186.06948587,
                   6344.74803956,
                                   7361.62678204, 10058.57116223,
                                                                     6792.219309
                   7897.72464823,
                                   5261.45936067,
                                                    4540.24137423,
                                                                    8709.36468047,
                                   7406.73353952,
                                                                    7047.27998963,
                   6882.0117409 ,
                                                    6795.61189392,
                   9945.33400083,
                                   8856.93910595,
                                                    9378.02074127, 10389.561154
                  10092.46332921, 10381.52000388,
                                                                    5996.3331428 ,
                                                    9723.92466625,
                   9786.14866981,
                                   7708.49649098,
                                                    5583.48163469,
                                                                     4932.92788329,
                   0256 66053001
                                   0736 77001005
                                                   10007 6/0571/7
                                                                     6756 1251677Q
```

Out[201]:

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

```
In [202]: Results['diff_price']=Results.apply(lambda row:row.Price-row.Predicted,axis=1)
```

In [204]: Results.head(15)

Out[204]:

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

In []:

In []: