In [147]: import pandas as pd In [148]: data=pd.read csv("/home/placement/Downloads/fiat500.csv") In [149]: data.describe() Out[149]: 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 1650.980494 769.500000 1.123537 51.904421 53396.011704 43.541361 11.563428 8576.003901 mean 444.126671 3.988023 1289.522278 40046.830723 0.416423 2.133518 2.328190 1939.958641 std 1.000000 51.000000 366.000000 1232.000000 1.000000 36.855839 7.245400 2500.000000 min 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 12.769040 10000.000000 75% 1153.750000 51.000000 2616.000000 79667.750000 1.000000 45.467960 max 1538.000000 77.000000 4658.000000 235000.000000 4.000000 46.795612 18.365520 11100.000000 data.head() In [150]: Out[150]: model engine power age in days km previous owners lon price lat lounge 0 1 51 882 25000 1 44.907242 8.611560 8900 2 pop 51 1186 32500 45.666359 12.241890 8800

45.503300 11.417840

40.633171 17.634609

1 41.903221 12.495650

4200

6000

5700

sport

lounge

pop

5

74

51

73

4658

2739

142228

160000

3074 106880

/ \ I I I +	1 15 1	
17111		
O G C		

	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
153	3 sport	51	3712	115280	1	5200
153	4 lounge	74	3835	112000	1	4600
153	5 pop	51	2223	60457	1	7500
153	6 lounge	51	2557	80750	1	5990
153	7 pop	51	1766	54276	1	7900

1538 rows × 6 columns

In [152]: data=pd.get_dummies(data)

In [153]: data

\sim		F 1 F 2 I
11	11#	1 15 7 1
w	u L	1 1
_		

	ID	engine_power	age_in_days	km	previous_owners	lat	lon	price	model_lounge	model_pop	model_sport
0	1	51	882	25000	1	44.907242	8.611560	8900	1	0	0
1	2	51	1186	32500	1	45.666359	12.241890	8800	0	1	0
2	3	74	4658	142228	1	45.503300	11.417840	4200	0	0	1
3	4	51	2739	160000	1	40.633171	17.634609	6000	1	0	0
4	5	73	3074	106880	1	41.903221	12.495650	5700	0	1	0
1533	1534	51	3712	115280	1	45.069679	7.704920	5200	0	0	1
1534	1535	74	3835	112000	1	45.845692	8.666870	4600	1	0	0
1535	1536	51	2223	60457	1	45.481541	9.413480	7500	0	1	0
1536	1537	51	2557	80750	1	45.000702	7.682270	5990	1	0	0
1537	1538	51	1766	54276	1	40.323410	17.568270	7900	0	1	0

1538 rows × 11 columns

```
In [154]:
```

data.shape

Out[154]: (1538, 11)

In [155]: data2=pd.get_dummies(data1)

In [156]: data2

	- 1	- 1	-	_	
1111		- 1	- 11		

	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 [157]: #for checking rows & columns
data2.shape
```

Out[157]: (1538, 8)

```
In [158]: #predicted value we removed from data frame
y=data2['price']
x=data2.drop('price',axis=1)
```

```
In [159]: y
Out[159]: 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 [160]: #!pip install scikit-learn
           #to install sklearn
In [161]: #divide the data into testing & training
           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 [162]: x_test.head(5)
Out[162]:
                 engine_power age_in_days
                                          km previous_owners model_lounge model_pop model_sport
                                  3197 120000
                                                          2
             481
                         51
                                                                      0
                                                                                1
                                                                                           0
                                  2101 103000
             76
                                                          1
                                                                      0
                                                                                           0
                         62
                                                                                1
            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 [163]: x_test.head(5)

	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 [164]: x_train.head(5)

Out[164]:

	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 [165]: y_test.head(5)

Out[165]: 481

7900 7900 76 1502 9400 669 8500 1409 9700

Name: price, dtype: int64

```
In [166]: y train.head(5)
Out[166]: 527
                  9990
           129
                  9500
           602
                  7590
           331
                  8750
           323
                  9100
           Name: price, dtype: int64
In [167]: from sklearn.linear_model import LinearRegression
In [168]: reg=LinearRegression()#creating object of LinearRegression
In [169]: reg.fit(x train,y train)#training and fitting LR object using training data and the is created by training d
Out[169]: 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 [170]: #prediction price
           y_pred=reg.predict(x_test)
```

```
In [171]: y pred
Out[171]: array([ 5867.6503378 ,
                                  7133.70142341.
                                                   9866.35776216.
                                                                   9723.28874535.
                                  9654.07582608,
                                                   9673.14563045, 10118.70728123,
                 10039.59101162,
                  9903.85952664,
                                  9351.55828437, 10434.34963575, 7732.26255693,
                                                   9662.90103518, 10373.20344286,
                  7698.67240131,
                                  6565.95240435,
                                                   4941.33017994, 10455.2719478,
                  9599.94844451,
                                  7699.34400418,
                 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,
                  9822.1231461 ,
                                                                   8457.38443276,
                                  8258.46551788,
                                                   6279.2040404 ,
                  9773.86444066,
                                  6767.04074749,
                                                   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,
                                                   9795.31884316,
                                                                   7197.84932877,
                 10108.31707235, 7009.6597206,
                                                   9853.90699412, 7146.87414965,
                  6417.69133992,
                                  9996.97382441,
                                                   9781.18795953,
                                                                   8515.83255277,
                  8456.30006203,
                                  6499.76668237,
                                                   7768.57829985,
                                                                   6832.86406122,
                  8347.96113362, 10439.02404036,
                                                   7356.43463051,
                                                                   8562.56562053,
In [172]: from sklearn.metrics import r2 score
          r2 score(y test,y pred)#y pred=actual price,y pred=predict price
Out[172]: 0.8415526986865394
In [177]: | from sklearn.metrics import mean squared_error#calculating MSE
          mean squared error(y pred,y test)
Out[177]: 581887.727391353
In [178]:
          import math
          print(math.sqrt(581887.727391353))
```

762.8156575420782

```
In [179]: y pred
Out[179]: array([ 5867.6503378 ,
                                   7133.70142341,
                                                    9866.35776216,
                                                                     9723.28874535.
                                   9654.07582608,
                  10039.59101162,
                                                    9673.14563045, 10118.70728123,
                   9903.85952664,
                                   9351.55828437,
                                                   10434.34963575,
                                                                     7732.26255693,
                                   6565.95240435,
                                                    9662.90103518, 10373.20344286,
                   7698.67240131,
                   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,
                                                                     5666.62365159,
                                   5157.65652129,
                                                    8948.63632836,
                   9822.1231461 ,
                                   8258.46551788,
                                                    6279.2040404 ,
                                                                     8457.38443276,
                   9773.86444066,
                                   6767.04074749,
                                                    9182.99904787, 10210.05195479,
                   8694.90545226, 10328.43369248,
                                                    9069.05761443,
                                                                     8866.7826029 ,
                   7058.39787506,
                                   9073.33877162,
                                                    9412.68162121, 10293.69451263,
                  10072.49011135,
                                   6748.5794244 ,
                                                    9785.95841801,
                                                                     9354.09969973,
                   9507.9444386 , 10443.01608254,
                                                                     7197.84932877,
                                                    9795.31884316,
                                   7009.6597206 ,
                  10108.31707235,
                                                    9853.90699412,
                                                                     7146.87414965,
                   6417.69133992,
                                   9996.97382441,
                                                    9781.18795953,
                                                                     8515.83255277,
                   8456.30006203,
                                   6499.76668237,
                                                    7768.57829985,
                                                                     6832.86406122,
                                                                     8562.56562053,
                   8347.96113362, 10439.02404036,
                                                    7356.43463051,
```

```
In [181]: #Results=pd.DataFrame(columns=['Actual', 'predicted'])
    #Results['Actual']=y_test
    Results['price']=y_test
    Results['predicted']=y_pred
    Results=Results.reset_index()
    Results['ID']=Results.index
    Results.head(5)
```

Out[181]: predicted ID index price 5867.650338 0 481 7900 7133.701423 1 76 7900 1502 9400 9866.357762 2 669 8500 9723.288745 3 9700 10039.591012 4 1409

```
In [182]: Results['diff']=Results.apply(lambda row:row.price-row.predicted,axis=1)
```

In [183]: Results

Out[183]:

	index	price	predicted	ID	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
503	291	10900	10032.665135	503	867.334865
504	596	5699	6281.536277	504	-582.536277
505	1489	9500	9986.327508	505	-486.327508
506	1436	6990	8381.517020	506	-1391.517020
507	575	10900	10371.142553	507	528.857447

508 rows × 5 columns

In []: