```
In [4]: import pandas as pd
In [5]: data=pd.read csv("/home/placement/Desktop/naren/fiat500.csv")
         #data['model']=data['model'].map({'lounge':1,'pop':2,'sport':3})
In [6]: data1=data.drop(['ID','lat','lon'],axis=1)
In [7]: data1
Out[7]:
                model engine_power age_in_days
                                                      previous_owners price
             0 lounge
                                51
                                          882
                                                25000
                                                                   1 8900
                                51
                                          1186
                                                32500
                                                                      8800
             1
                  pop
             2
                                          4658
                                               142228
                                                                     4200
                 sport
                                74
                                51
                                                                     6000
                lounge
                                          2739
                                               160000
                                73
                                          3074
                                               106880
                                                                     5700
                  pop
          1533
                 sport
                                51
                                          3712 115280
                                                                      5200
          1534
                lounge
                                74
                                          3835
                                               112000
                                                                     4600
          1535
                                51
                                          2223
                                                60457
                                                                     7500
                  pop
          1536
                lounge
                                51
                                          2557
                                                80750
                                                                      5990
          1537
                                51
                                         1766
                                                54276
                                                                   1
                                                                     7900
                  pop
```

1538 rows × 6 columns

Type *Markdown* and LaTeX: α^2

In [8]: #data1=pd.get dummies(data1)

```
In [9]: #data1
In [10]: data1['model']=data1['model'].map({'lounge':1,'pop':2,'sport':3})
In [11]: data1
Out[11]:
                       engine_power age_in_days
                 model
                                                  km previous_owners price
              0
                    1
                                51
                                           882
                                                25000
                                                                  1 8900
                    2
                                51
              1
                                          1186
                                                32500
                                                                     8800
                                                                   1
              2
                    3
                                74
                                          4658 142228
                                                                  1 4200
                                51
                                          2739
                                               160000
                                                                     6000
              3
                    1
                                73
                                                                     5700
                    2
                                          3074
                                              106880
                                                                   1
                                          3712 115280
                                                                     5200
           1533
                    3
                                51
                                                                   1
           1534
                    1
                                74
                                          3835
                                               112000
                                                                     4600
                                51
                                          2223
                                                60457
                                                                     7500
           1535
                    2
           1536
                                51
                                          2557
                                                80750
                                                                     5990
                    1
           1537
                    2
                                51
                                          1766
                                                54276
                                                                   1
                                                                     7900
          1538 rows × 6 columns
In [12]: data1=pd.get_dummies(data1)
```

In [13]: data1

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(1)	+ 1		
υu	L	 	

	model	engine_power	age_in_days	km	previous_owners	price
0	1	51	882	25000	1	8900
1	2	51	1186	32500	1	8800
2	3	74	4658	142228	1	4200
3	1	51	2739	160000	1	6000
4	2	73	3074	106880	1	5700
1533	3	51	3712	115280	1	5200
1534	1	74	3835	112000	1	4600
1535	2	51	2223	60457	1	7500
1536	1	51	2557	80750	1	5990
1537	2	51	1766	54276	1	7900

1538 rows × 6 columns

```
In [14]: y=data1['price']
x=data1.drop('price',axis=1)
```

```
In [15]: y
Out[15]: 0
                  8900
                  8800
                  4200
         2
                  6000
          3
          4
                  5700
         1533
                  5200
         1534
                  4600
         1535
                  7500
         1536
                  5990
         1537
                  7900
         Name: price, Length: 1538, dtype: int64
In [16]: x
Out[16]:
               model engine_power age_in_days
                                              km previous_owners
```

0	1	51	882	25000	1	_
1	2	51	1186	32500	1	
2	3	74	4658	142228	1	
3	1	51	2739	160000	1	
4	2	73	3074	106880	1	
		•••		***		
1533	3	51	3712	115280	1	
1534	1	74	3835	112000	1	
1535	2	51	2223	60457	1	
1536	1	51	2557	80750	1	
1537	2	51	1766	54276	1	

1538 rows × 5 columns

In [17]: #!pip3 install scikit-learn

```
In [18]: #data1=pd.get dummies(data1)
In [19]: #data1
 In [ ]:
In [20]: 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 [21]: x_test.head(5)
Out[21]:
                model engine_power age_in_days
                                                km previous_owners
                   2
                                        3197 120000
            481
                               51
                                                                2
            76
                   2
                               62
                                        2101 103000
                                                                1
           1502
                                                                1
                   1
                               51
                                         670
                                              32473
                               51
                                                                1
            669
                   1
                                         913
                                              29000
           1409
                   1
                               51
                                         762
                                              18800
                                                                1
```

```
In [22]: x_train.shape
Out[22]: (1030, 5)
In [23]: y_train.shape
Out[23]: (1030,)
```

In [24]: x_train.head()

ο.		$\Gamma \cap A \perp 1$	
υι	ıτ	I Z4 I	

	model	engine_power	age_in_days	km	previous_owners
527	1	51	425	13111	1
129	1	51	1127	21400	1
602	2	51	2039	57039	1
331	1	51	1155	40700	1
323	1	51	425	16783	1

In [25]: y_train.head()

Out[25]: 527

9990

129 9500

602 7590

331 8750

323 9100

Name: price, dtype: int64

In [26]: x_test.head()

Out[26]:

	model	engine_power	age_in_days	km	previous_owners
481	2	51	3197	120000	2
76	2	62	2101	103000	1
1502	1	51	670	32473	1
669	1	51	913	29000	1
1409	1	51	762	18800	1

In [27]: y test.head()

```
Out[27]: 481
                   7900
          76
                   7900
          1502
                   9400
          669
                   8500
          1409
                   9700
          Name: price, dtype: int64
In [29]: from sklearn.linear model import LinearRegression
          reg=LinearRegression()
          reg.fit(x_train,y_train)
Out[29]: 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 [ ]: #!pip3 install scikit-learn
In [31]: |ypred=reg.predict(x test)
```

```
In [38]: | ypred
Out[38]: arrav([ 5994.51703157.
                                  7263.58726658,
                                                   9841.90754881.
                                                                   9699.31627673.
                 10014.19892635,
                                  9630.58715835,
                                                   9649.4499026 . 10092.9819664 .
                                                                   7716.91706011,
                  9879.19498711,
                                  9329.19347948, 10407.2964056,
                                  6673.95810983,
                  7682.89152522,
                                                   9639.42618839, 10346.53679153,
                                                  4727.33552438, 10428.17092937,
                 9366.53363673,
                                  7707.90063494,
                 10359.87663878, 10364.84674179,
                                                   7680.16157493,
                                                                   9927.58506055,
                                                   4929.31229715.
                                                                   6940.60225317.
                  7127.7284177 ,
                                  9097.51161986,
                 7794.35120591,
                                  9600.43942019,
                                                  7319.85877519,
                                                                   5224.05298205,
                 5559.52039134,
                                  5201.35403287,
                                                   8960.11762682,
                                                                   5659.72968338,
                                                                   8556.73835062,
                  9915.79926869,
                                  8255.93615893,
                                                   6270.40332834,
                 9749.72882426,
                                  6873.76758364,
                                                   8951.72659758, 10301.95669828,
                  8674.89268564, 10301.93257222,
                                                  9165.73586068,
                                                                   8846.92420399,
                  7044.68964545,
                                  9052.4031418 ,
                                                   9390.75738772, 10267.3912561,
                 10046.90924744,
                                  6855.71260655,
                                                   9761.93338967,
                                                                   9450.05744337,
                  9274.98388541, 10416.00474283,
                                                  9771.10646661,
                                                                   7302.96566423,
                 10082.61483093,
                                  6996.96553454,
                                                   9829.40534825,
                                                                   7134.21944391,
                                                                   8614.84049875,
                  6407.26222178,
                                  9971.82132188,
                                                   9757.01618446,
                  8437.92452169,
                                  6489.24658616,
                                                   7752.65456507,
                                                                   6626.60510856,
                  8329.88998217, 10412.00324329,
                                                   7342.77348105,
                                                                   8543.63624413,
In [34]: from sklearn.metrics import r2 score
         r2_score(y test,ypred)
Out[34]: 0.8383895235218546
In [35]: from sklearn.metrics import mean squared error as ns
         o=ns(y test,ypred)
         0
Out[35]: 593504.2888137395
         import math
In [36]:
         math.sgrt(o)
Out[36]: 770.3922954013361
```

```
In [39]: | ypred
Out[39]: array([ 5994.51703157,
                                   7263.58726658,
                                                    9841.90754881.
                                                                    9699.31627673.
                 10014.19892635,
                                   9630.58715835,
                                                    9649.4499026 . 10092.9819664 .
                  9879.19498711,
                                   9329.19347948,
                                                  10407.2964056 ,
                                                                    7716.91706011,
                  7682.89152522,
                                   6673.95810983,
                                                    9639.42618839, 10346.53679153,
                  9366.53363673,
                                   7707.90063494,
                                                   4727.33552438, 10428.17092937,
                 10359.87663878, 10364.84674179,
                                                    7680.16157493,
                                                                    9927.58506055,
                  7127.7284177 ,
                                   9097.51161986,
                                                    4929.31229715,
                                                                    6940.60225317,
                  7794.35120591,
                                   9600.43942019,
                                                    7319.85877519,
                                                                    5224.05298205,
                  5559.52039134,
                                   5201.35403287,
                                                    8960.11762682,
                                                                    5659.72968338,
                                                                    8556.73835062,
                  9915.79926869,
                                   8255.93615893,
                                                    6270.40332834,
                  9749.72882426,
                                   6873.76758364,
                                                    8951.72659758, 10301.95669828,
                  8674.89268564, 10301.93257222,
                                                                    8846.92420399,
                                                    9165.73586068,
                  7044.68964545,
                                   9052.4031418 ,
                                                   9390.75738772, 10267.3912561 ,
                 10046.90924744,
                                   6855.71260655,
                                                    9761.93338967,
                                                                    9450.05744337,
                  9274.98388541, 10416.00474283,
                                                                    7302.96566423,
                                                    9771.10646661,
                 10082.61483093,
                                   6996.96553454,
                                                    9829.40534825,
                                                                    7134.21944391,
                  6407.26222178,
                                   9971.82132188,
                                                    9757.01618446,
                                                                    8614.84049875,
                  8437.92452169,
                                   6489.24658616,
                                                    7752.65456507,
                                                                    6626.60510856,
                  8329.88998217, 10412.00324329,
                                                    7342.77348105,
                                                                    8543.63624413,
```

```
In [42]: 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[42]:		index	price	predicted	ID	
	0	481	7900	5994.517032	0	
	1	76	7900	7263.587267	1	
	2	1502	9400	9841.907549	2	
	3	669	8500	9699.316277	3	
	4	1409	9700	10014.198926	4	
	5	1414	9900	9630.587158	5	
	6	1089	9900	9649.449903	6	
	7	1507	9950	10092.981966	7	
	8	970	10700	9879.194987	8	
	9	1198	8999	9329.193479	9	
	10	1088	9890	10407.296406	10	
	11	576	7990	7716.917060	11	
	12	965	7380	7682.891525	12	
	13	1488	6800	6673.958110	13	
	14	1432	8900	9639.426188	14	

```
In [43]: Results['price_diff']=Results.apply(lambda row: row.price - row.predicted,axis=1)
```

In [44]: Results

Out[44]:

	index	price	predicted	ID	price_diff
0	481	7900	5994.517032	0	1905.482968
1	76	7900	7263.587267	1	636.412733
2	1502	9400	9841.907549	2	-441.907549
3	669	8500	9699.316277	3	-1199.316277
4	1409	9700	10014.198926	4	-314.198926
503	291	10900	10007.364639	503	892.635361
504	596	5699	6390.174715	504	-691.174715
505	1489	9500	10079.478928	505	-579.478928
506	1436	6990	8363.337585	506	-1373.337585
507	575	10900	10344.486077	507	555.513923

508 rows × 5 columns

In []: