```
In [2]: import pandas as pd
In [3]: data=pd.read csv("/home/placement/Desktop/naren/fiat500.csv")
In [5]: data1=data.drop(['ID','lat','lon'],axis=1)
In [6]: data1
Out[6]:
                model engine_power age_in_days
                                                  km previous_owners price
                                51
                                                                   1 8900
             0 lounge
                                           882
                                                25000
                                51
                                          1186
                                                                      8800
             1
                  pop
                                                32500
                                74
                                               142228
                                                                      4200
                 sport
                                          4658
                                               160000
                                                                      6000
                lounge
                                51
                                          2739
                                73
                                          3074
                                               106880
                                                                      5700
                  pop
          1533
                 sport
                                51
                                          3712 115280
                                                                      5200
          1534
                                74
                                               112000
                                                                      4600
                lounge
                                          3835
          1535
                                51
                                                                      7500
                                          2223
                                                60457
                  pop
                                          2557
                                                80750
                                                                      5990
          1536
                                51
                lounge
          1537
                                51
                                          1766
                                                54276
                                                                   1 7900
                  pop
         1538 rows × 6 columns
         Type Markdown and LaTeX: \alpha^2
In [7]: data1['model']=data1['model'].map({'lounge':1,'pop':2,'sport':3})
```

In [8]: data1

Out[8]:

	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 [9]: y=datal['price']
x=datal.drop('price',axis=1)
```

```
In [10]: y
Out[10]: 0
                 8900
                 8800
                 4200
         2
                 6000
         3
                 5700
         4
         1533
                 5200
         1534
                 4600
         1535
                 7500
         1536
                 5990
         1537
                 7900
         Name: price, Length: 1538, dtype: int64
```

In [11]: x

## Out[11]:

	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

	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 [15]: x_train.shape
Out[15]: (1030, 5)
In [16]: y_train.shape
Out[16]: (1030,)
```

In [17]: x\_train.head()

Out[17]:

		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 [18]: y_train.head()
```

Out[18]: 527

9990

129 9500

602 7590

331 8750

323 9100

Name: price, dtype: int64

In [19]: x\_test.head()

Out[19]:

	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 [20]: y test.head()
Out[20]: 481
                 7900
         76
                 7900
         1502
                 9400
         669
                 8500
         1409
                 9700
         Name: price, dtype: int64
In [21]: from sklearn.linear_model import LinearRegression
         reg=LinearRegression()
         reg.fit(x_train,y_train)
Out[21]:
          ▼ LinearRegression
          LinearRegression()
In [22]: ypred=reg.predict(x_test)
```

```
In [23]: ypred
                                 9755.03033027,
                 9767.12327701,
                                                 6859.84033207. 7339.68592914.
                 8740.34003982.
                                 9898.84623968.
                                                 9788.7072129 . 10439.74281794.
                 8145.90808395,
                                 6767.15633519,
                                                 9962.57850061, 8846.92420399,
                 9927.58506055, 10279.88133318, 10205.11210182, 10065.46678709,
                 9343.97683092,
                                 9983.85933876,
                                                 9237.93178546, 10073.45985579,
                 7906.63849672,
                                 6017.75726035,
                                                 8780.77873324, 10211.55465771,
                 5737.35007744, 10190.21750673,
                                                 9661.444679
                                                             , 7747.41088806,
                 9396.65945773,
                                 7357.03908605, 10261.68730153, 10041.70922157,
                10525.09542651.
                                 9941.6915233 , 10042.87112799,
                                                                 6342.10368715.
                10588.92756092,
                                 9940.98736563, 10501.95046891,
                                                                 9697.00608104,
                 9642.20441674,
                                 6177.49903451,
                                                 8056.81304643, 10318.99744586,
                                 7347.76781534, 10049.18638926,
                                                                 6780.85650138,
                 6334.90676093,
                 7897.31981053,
                                 5062.64376289,
                                                 4656.55980585,
                                                                 8690.25433913,
                 6988.39956167,
                                 7416.44791638,
                                                 6784.57575877,
                                                                 7034.60046808,
                 9920.49385534, 8837.10014026,
                                                 9473.86037871, 10362.758227
                                                                 6004.14282877,
                10185.09072066, 10354.75677873,
                                                 9700.03561924,
                                 7693.58259057,
                                                 5694.99266265,
                                                                 5047.53800681,
                 9762.0944756 ,
                 9832.25820502, 9332.68577808, 10066.98651296,
                                                                 6061.66190757,
                 8691.8702504 , 10314.99712678.
                                                  5011.82707762. 10007.36463896.
                 6300 17/71/70 10070 /7802753
                                                  2262 22752526 102AA A260771 1\
In [24]: from sklearn.metrics import r2 score
         r2_score(y test,ypred)
Out[24]: 0.8383895235218546
In [25]: from sklearn.metrics import mean squared error as ns
         o=ns(y test,ypred)
         0
Out[25]: 593504.2888137395
In [26]: import math
         math.sgrt(o)
Out[26]: 770.3922954013361
```

```
In [27]: ypred
                  9915.79926869.
                                  8255.93615893.
                                                   6270.40332834.
                                                                    8556.73835062.
                  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,
                  6407.26222178,
                                  9971.82132188,
                                                                    8614.84049875,
                                                   9757.01618446,
                  8437.92452169,
                                                   7752.65456507,
                                                                    6626.60510856,
                                  6489.24658616,
                  8329.88998217, 10412.00324329,
                                                   7342.77348105,
                                                                    8543.63624413,
                  9796.44743777, 10010.43582651,
                                                   7356.86786062,
                                                                    9523.1488851 ,
                                  8076.40060487, 10419.77659259,
                                                                    3528.47185472,
                 10326.22987622,
                                                   6183.26049817, 10363.3075255 ,
                 10322.06476759, 10425.03337221,
                                                                    9431.7474931 ,
                  6517.32275335,
                                  9095.0117622 , 10457.31199063,
                                  3459.16038547, 10198.89642316,
                                                                    9768.32453463,
                  6698.67676865,
                  6349.19977644,
                                  4992.62377497,
                                                   9157.57378424,
                                                                    9843.66900765,
                  5300.60403791,
                                  5481.29193836, 10061.18107826,
                                                                    8216.74437963,
                                                                    5747.75873375,
                 10365.58751936,
                                  6942.04538818,
                                                   6646.91921519,
```

9396.86589768,

9927.58506055, 10363.24676269,

9076.99796116. 10215.3021386 . 10383.15669206. 10126.03200513.

8993.90183905,

```
In [28]: 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[28]:

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

In [30]: Results

## Out[30]:

	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 [ ]: