```
In [115]: import pandas as pd
In [116]: data=pd.read_csv("/home/placement/Downloads/fiat500 (another copy).csv")
In [117]: data.describe()
Out[117]:
                             ID engine power
                                                                      km previous owners
                                               age in days
                                                                                                    lat
                                                                                                               lon
                                                                                                                           price
              count 1538.000000
                                   1538.000000
                                               1538.000000
                                                              1538.000000
                                                                                           1538.000000
                                                                                                       1538.000000
                                                                                                                     1538.000000
                                                                               1538.000000
                      769.500000
                                               1650.980494
                                                                                  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
               75%
                    1153.750000
                                     51.000000
                                               2616.000000
                                                             79667.750000
                                                                                  1.000000
                                                                                             45.467960
                                                                                                                   10000.000000
               max 1538.000000
                                     77.000000
                                                                                  4.000000
                                                                                             46.795612
                                               4658.000000 235000.000000
                                                                                                          18.365520 11100.000000
            data2=data.drop(['ID','lat','lon'],axis=1)
```

In [119]: data2

Out[119]:

	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 [120]: data2=pd.get_dummies(data2)

In [121]: data2

\sim			г -	-	-	-
- ()		-				- 1
u	u	L	IJ	LZ	. т	

	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 [122]: data2.shape
Out[122]: (1538, 8)
In [123]: y=data2['price']
x=data2.drop('price',axis=1)
```

```
In [124]: y
Out[124]: 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 [125]: 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 [126]: x_test.head(5)
Out[126]:
                 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
                         62
                                                           1
                                                                       0
                                                                                 1
                                                                                            0
            1502
                         51
                                    670
                                         32473
                                                           1
                                                                       1
                                                                                 0
                                                                                            0
             669
                          51
                                    913
                                         29000
                                                           1
                                                                       1
                                                                                 0
                                                                                            0
                                         18800
                                                                                            0
            1409
                         51
                                    762
                                                           1
                                                                       1
                                                                                 0
```

```
In [127]: x train.head(5)
Out[127]:
                                         km previous owners model lounge model pop model sport
                engine power age in days
                                   425 13111
            527
                         51
                                                         1
                                                                     1
                                                                               0
                                                                                          0
            129
                         51
                                  1127
                                       21400
                                                         1
                                                                     1
                                                                                          0
                                  2039 57039
            602
                         51
                                                         1
                                                                     0
                                                                                          0
                                                                               1
                                  1155 40700
            331
                         51
                                                         1
                                                                     1
                                                                                          0
            323
                         51
                                   425 16783
                                                         1
                                                                     1
                                                                                          0
In [128]: y_test.head(5)
Out[128]: 481
                   7900
                   7900
           76
           1502
                   9400
           669
                   8500
           1409
                   9700
           Name: price, dtype: int64
In [129]: y_train.head(5)
Out[129]: 527
                  9990
           129
                  9500
           602
                  7590
           331
                  8750
           323
                  9100
           Name: price, dtype: int64
In [130]: x_train.shape
Out[130]: (1030, 7)
```

```
In [131]: y train
Out[131]: 527
                     9990
           129
                     9500
           602
                     7590
           331
                     8750
           323
                     9100
           1130
                    10990
           1294
                     9800
           860
                     5500
                     9990
           1459
           1126
                     8900
           Name: price, Length: 1030, dtype: int64
In [132]: from sklearn.linear model import LinearRegression
           reg=LinearRegression()
           reg.fit(x_train,y_train)
Out[132]: 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 [133]: ypred=reg.predict(x_test)
```

```
In [144]: | ypred
                  9727.85493167,
                                   6281.78952194.
                                                                    9387.9203723 .
                                                   7901.36245623.
                  5039.55649797.
                                   9351.49777725.
                                                   9980.70844784. 10094.79341516.
                  6359.24321991,
                                   9856.10227211,
                                                   9099.07023804,
                                                                    5234.05388382,
                  5534.45288323,
                                   4495.02309231, 10199.78432943, 10024.87037067,
                  5465.58034188,
                                   8520.72057674,
                                                   7034.71038647, 10054.65061446,
                 10191.12067767.
                                   6008.34860428.
                                                                    9669.4333196 .
                                                   9748.18097947.
                  9145.3756075 ,
                                   9175.66562699, 10087.86753845,
                                                                    9825.02990067,
                  7340.29803785,
                                   5083.8487301 ,
                                                   9441.50914802, 10243.05490667,
                  5556.42300245, 10676.01945733,
                                                   6126.99295838,
                                                                    9845.16661356,
                  9850.77978959,
                                   7840.83596305,
                                                   6552.05146566,
                                                                    9938.82104889,
                  8327.79232274,
                                   9119.62204137,
                                                   6111.83787367, 10410.00504522,
                  6360.97695249,
                                   8601.59209793,
                                                   8377.80258216,
                                                                    9803.81343895,
                  8285.09831762, 10091.75635129, 10003.86694939, 10028.60283146,
                 10354.61956534,
                                   8552.21002673,
                                                   6726.65446676,
                                                                    9381.22662706,
                  6520.9999373 , 10352.85155564,
                                                   9063.7534579 , 10456.89121831,
                                   9952.37340054,
                                                                    9220.36267675,
                  9127.72470241,
                                                   8376.6975881 ,
                 10036.24981328,
                                   8418.65456209,
                                                   4717.7579531 , 10076.86950203,
                 10017.8490121 , 10590.33289679,
                                                  10161.75393066,
                                                                    4927.49556508
                                   9678.26477249.
                                                   9764.65653403.
                  7276.18410037.
                                                                    5643.53722047
In [135]: from sklearn.metrics import r2 score
          r2 score(y test,ypred)
Out[135]: 0.8415526986865394
In [137]: from sklearn.metrics import mean squared error
          mean squared error(ypred,y test)
Out[137]: 581887.727391353
In [140]:
          import math
          y=math.sgrt(581887)
In [141]: y
Out[141]: 762.815180761369
```

```
(581887)**(1/2)
In [143]:
Out[143]: 762.815180761369
In [146]: 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[146]:
                      price
                               predicted ID
                index
             0
                 481
                       7900
                             5867.650338
                                         0
                  76
                       7900
                             7133.701423
                                        1
             1
                1502
                       9400
                             9866.357762 2
                 669
                       8500
                             9723.288745
                                         3
                1409
                       9700
                            10039.591012 4
                1414
                       9900
                             9654.075826
                                        5
             5
                1089
                       9900
                             9673.145630
                                        6
             7
                1507
                       9950
                            10118.707281
                                        7
                      10700
                             9903.859527
                                         8
                 970
                1198
                       8999
                             9351.558284
                                         9
                1088
            10
                       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 [147]: Results['diff']=Results.apply(lambda row: row.price-row.predicted,axis=1)
```

In [149]: Results

Out[149]:

	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 []: