

```
In [55]: import pandas as pd
```

```
In [56]: data=pd.read_csv("/home/placement/Desktop/usha g1/fiat500.csv")
```

```
In [57]: data.describe()
```

```
Out[57]:
```

	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
mean	769.500000	51.904421	1650.980494	53396.011704	1.123537	43.541361	11.563428	8576.003901
std	444.126671	3.988023	1289.522278	40046.830723	0.416423	2.133518	2.328190	1939.958641
min	1.000000	51.000000	366.000000	1232.000000	1.000000	36.855839	7.245400	2500.000000
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
75%	1153.750000	51.000000	2616.000000	79667.750000	1.000000	45.467960	12.769040	10000.000000
max	1538.000000	77.000000	4658.000000	235000.000000	4.000000	46.795612	18.365520	11100.000000

```
In [58]: data.head()
```

```
Out[58]:
```

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.611560	8900
1	2	pop	51	1186	32500	1	45.666359	12.241890	8800
2	3	sport	74	4658	142228	1	45.503300	11.417840	4200
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
4	5	pop	73	3074	106880	1	41.903221	12.495650	5700

```
In [59]: data1=data.drop(['lat','lon','ID'],axis=1) #unwanted columns removed
```

```
In [60]: data1
```

```
Out[60]:
```

	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 [61]: data1=pd.get_dummies(data1)
```

```
In [62]: data1.shape  
#data1['model']=data['model'].map({'lounge':1,'pop':2,'sport':3})
```

```
Out[62]: (1538, 8)
```

```
In [63]: data1
```

```
Out[63]:
```

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

In [65]:

y

Out[65]:

0	8900
1	8800
2	4200
3	6000
4	5700
	...
1533	5200
1534	4600
1535	7500
1536	5990
1537	7900

Name: price, Length: 1538, dtype: int64

In [66]:

```
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 [67]:

x_test.head(5)

Out[67]:

	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 [68]:

x_train.shape

Out[68]: (1030, 7)

```
In [69]: y_train
```

```
Out[69]: 527      9990
         129      9500
         602      7590
         331      8750
         323      9100
         ...
        1130     10990
        1294      9800
         860      5500
        1459      9990
        1126      8900
        Name: price, Length: 1030, dtype: int64
```

```
In [70]: y_test.head()
```

```
Out[70]: 481      7900
         76      7900
        1502      9400
         669      8500
        1409      9700
        Name: price, dtype: int64
```

```
In [71]: y_train.shape
```

```
Out[71]: (1030,)
```

```
In [72]: from sklearn.linear_model import LinearRegression
         reg = LinearRegression()
         reg.fit(x_train,y_train)
```

```
Out[72]: 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 [73]: ypred=reg.predict(x_test)
```

In [74]: ypred

```
Out[74]: array([ 5867.6503378 ,  7133.70142341,  9866.35776216,  9723.28874535,
                10039.59101162,  9654.07582608,  9673.14563045, 10118.70728123,
                9903.85952664,  9351.55828437, 10434.34963575,  7732.26255693,
                7698.67240131,  6565.95240435,  9662.90103518, 10373.20344286,
                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,  5157.65652129,  8948.63632836,  5666.62365159,
                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,  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,
                8820.78555100, 10025.82571520,  7270.77108022,  8411.45804006]
```

```
In [75]: from sklearn.metrics import r2_score
         r2_score(y_test,ypred)
         # efficiency
```

Out[75]: 0.8415526986865394

```
In [76]: from sklearn.metrics import mean_squared_error
         mean_squared_error(ypred,y_test)
         #calculating MSE
```

Out[76]: 581887.727391353

```
In [77]: n=581887.727391353 ** (1/2)
         print(n)
```

762.8156575420782

In [78]:

```
y_test.head()
```

```
Out[78]: 481      7900
          76      7900
          1502    9400
          669     8500
          1409    9700
          Name: price, dtype: int64
```

In [80]:

```
ypred
```

```
Out[80]: array([ 5867.6503378 ,  7133.70142341,  9866.35776216,  9723.28874535,
        10039.59101162,  9654.07582608,  9673.14563045, 10118.70728123,
        9903.85952664,  9351.55828437, 10434.34963575,  7732.26255693,
        7698.67240131,  6565.95240435,  9662.90103518, 10373.20344286,
        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,  5157.65652129,  8948.63632836,  5666.62365159,
        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,  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,
        8820.78555100, 10025.82571520,  7270.77108022,  8411.45804006])
```

```
In [86]: Results= pd.DataFrame(columns=['Price', 'Predicted'])
Results['Price']=y_test
Results['Predicted']=ypred
Results.head(20)
```

Out[86]:

	Price	Predicted
481	7900	5867.650338
76	7900	7133.701423
1502	9400	9866.357762
669	8500	9723.288745
1409	9700	10039.591012
1414	9900	9654.075826
1089	9900	9673.145630
1507	9950	10118.707281
970	10700	9903.859527
1198	8999	9351.558284
1088	9890	10434.349636
576	7990	7732.262557
965	7380	7698.672401
1488	6800	6565.952404
1432	8900	9662.901035
380	10500	10373.203443
754	10690	9599.948445
30	6990	7699.344004
49	4300	4941.330180
240	10500	10455.271948

In []: