

```
In [1]: import pandas as pd
```

```
In [2]: data=pd.read_csv("/home/placement/Downloads/fiat500.csv")
```

```
In [3]: data.describe()
```

Out[3]:

	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 [4]: data.head()
```

Out[4]:

	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 [5]: data1=data.drop(['lat','lon','ID'],axis=1)
```

```
In [6]: data1
```

```
Out[6]:
```

	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 [7]: data1.shape
```

```
Out[7]: (1538, 6)
```

```
In [8]: data1=pd.get_dummies(data1)
```

In [9]: data1

Out[9]:

	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 [10]: data1.shape

Out[10]: (1538, 8)

In [11]: y=data1['price']
x=data1.drop('price',axis=1) #

In [12]:

y

Out[12]:

```
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 [13]:

```
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) #split data into testir
```

In [14]:

x_test.head(5)

Out[14]:

	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 [15]:

x_train.shape

Out[15]: (1030, 7)

```
In [16]: y_train
```

```
Out[16]: 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 [17]: y_test.head()
```

```
Out[17]: 481      7900
          76      7900
          1502     9400
          669      8500
          1409     9700
          Name: price, dtype: int64
```

```
In [18]: y_train.shape
```

```
Out[18]: (1030,)
```

```
In [19]: from sklearn.linear_model import LinearRegression
          reg=LinearRegression() #creating object of LinearRegression
          reg.fit(x_train,y_train) #training and fitting LR object using training data
```

```
Out[19]: LinearRegression()
```

```
In [20]: ypred=reg.predict(x_test)
```

In [21]: ypred

Out[21]: 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, 7270.77100000, 8411.45004000])

In [22]: `from sklearn.metrics import r2_score`
`r2_score(y_test,ypred)`

Out[22]: 0.8415526986865394

In [23]: `from sklearn.metrics import mean_squared_error` *#calculating MSE*
`mean_squared_error(ypred,y_test)`

Out[23]: 581887.727391353

In [24]: `n=581887.727391353`
`print(n**(1/2))`

762.8156575420782

In [25]: `y_test.head(10)`

```
Out[25]: 481      7900
          76      7900
          1502    9400
          669     8500
          1409    9700
          1414    9900
          1089    9900
          1507    9950
          970    10700
          1198    8999
          Name: price, dtype: int64
```

In [26]: `ypred`

```
Out[26]: 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,  7370.77108022,  8411.45804006])
```

```
In [27]: Results=pd.DataFrame(columns=['price', 'predicted'])
Results['price']=y_test
Results['predicted']=ypred
Results.head(15)
```

Out[27]:

	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

```
In [28]: Results['diff']=Results.apply(lambda row: row.price - row.predicted,axis=1) #Difference
```


In [29]: Results

Out[29]:

	price	predicted	diff
481	7900	5867.650338	2032.349662
76	7900	7133.701423	766.298577
1502	9400	9866.357762	-466.357762
669	8500	9723.288745	-1223.288745
1409	9700	10039.591012	-339.591012
...
291	10900	10032.665135	867.334865
596	5699	6281.536277	-582.536277
1489	9500	9986.327508	-486.327508
1436	6990	8381.517020	-1391.517020
575	10900	10371.142553	528.857447

508 rows × 3 columns

In []: