In [1]: import pandas as pd
import warnings
warnings.filterwarnings("ignore")

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

Out[2]:

	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
1533	1534	sport	51	3712	115280	1	45.069679	7.704920	5200
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1535	1536	pop	51	2223	60457	1	45.481541	9.413480	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990
1537	1538	pop	51	1766	54276	1	40.323410	17.568270	7900

1538 rows × 9 columns

Out[3]:

	model	engine_power	age_in_days	km	previous_owners	price
0	lounge	51	882	25000	1	8900
1	рор	51	1186	32500	1	8800
2	sport	74	4658	142228	1	4200
3	lounge	51	2739	160000	1	6000
4	рор	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 [35]: data1.head(10)

Out[35]:

	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
5	74	3623	70225	1	7900	0	1	0
6	51	731	11600	1	10750	1	0	0
7	51	1521	49076	1	9190	1	0	0
8	73	4049	76000	1	5600	0	0	1
9	51	3653	89000	1	6000	0	0	1

In [5]: data1=pd.get_dummies(data1)
 data1

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

```
In [7]: y
Out[7]: 0
                8900
                8800
                4200
        2
        3
                6000
                5700
        4
        1533
                5200
        1534
                4600
        1535
                7500
        1536
                5990
        1537
                7900
        Name: price, Length: 1538, dtype: int64
```

In [8]: X

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	engine_power	age_in_days	km	previous_owners	model_lounge	model_pop	model_sport
0	51	882	25000	1	1	0	0
1	51	1186	32500	1	0	1	0
2	74	4658	142228	1	0	0	1
3	51	2739	160000	1	1	0	0
4	73	3074	106880	1	0	1	0
1533	51	3712	115280	1	0	0	1
1534	74	3835	112000	1	1	0	0
1535	51	2223	60457	1	0	1	0
1536	51	2557	80750	1	1	0	0
1537	51	1766	54276	1	0	1	0

1538 rows × 7 columns

```
In [9]: 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 [10]: from sklearn.linear model import LinearRegression
          reg=LinearRegression() #creating of Linear Regression
         reg.fit(X train,y train) #training and fitting LR object using training data
Out[10]: 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 nbyiewer.org.
In [11]: ypred=reg.predict(X test)
In [12]: from sklearn.metrics import r2 score #model efficinecy step
                                                                             #y test is the actual price
                                                                                                              #v predit is
          r2 score(y test,ypred)
Out[12]: 0.8415526986865394
In [13]: from sklearn.metrics import mean squared error ##Calulating mean square error
         mean squared error(y test,ypred)
Out[13]: 581887.727391353
```

Out[14]:

	index	Price	Predicted
0	481	7900	5867.650338
1	76	7900	7133.701423
2	1502	9400	9866.357762
3	669	8500	9723.288745
4	1409	9700	10039.591012
5	1414	9900	9654.075826
6	1089	9900	9673.145630
7	1507	9950	10118.707281
8	970	10700	9903.859527
9	1198	8999	9351.558284
10	1088	9890	10434.349636
11	576	7990	7732.262557
12	965	7380	7698.672401
13	1488	6800	6565.952404
14	1432	8900	9662.901035

```
In [15]: Results['diff price']=Results.apply(lambda row:row.Price-row.Predicted,axis=1)
```

In [16]: Results

Out[16]:

	index	Price	Predicted	diff_price
0	481	7900	5867.650338	2032.349662
1	. 76	7900	7133.701423	766.298577
2	1502	9400	9866.357762	-466.357762
3	669	8500	9723.288745	-1223.288745
4	1409	9700	10039.591012	-339.591012
•••				
503	291	10900	10032.665135	867.334865
504	596	5699	6281.536277	-582.536277
505	1489	9500	9986.327508	-486.327508
506	1436	6990	8381.517020	-1391.517020
507	575	10900	10371.142553	528.857447

508 rows × 4 columns

```
In [17]: from sklearn.metrics import mean_squared_error ##Calulating mean square error
mean_squared_error(y_test,ypred)
```

Out[17]: 581887.727391353

Ridge Regression

In [19]: from sklearn.model_selection import GridSearchCV
from sklearn.linear model import Ridge

```
alpha = [1e-15, 1e-10, 1e-8, 1e-4, 1e-3,1e-2, 1, 5, 10, 20,30]
          ridge = Ridge()
         parameters = {'alpha': alpha}
          ridge regressor = GridSearchCV(ridge, parameters)
          ridge regressor.fit(X train, y train)
Out[19]: GridSearchCV(estimator=Ridge(),
                       param grid={'alpha': [1e-15, 1e-10, 1e-08, 0.0001, 0.001, 0.01, 1,
                                               5, 10, 20, 30]})
         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 [20]: ridge regressor.best params
Out[20]: {'alpha': 30}
In [21]: ridge=Ridge(alpha=30)
          ridge.fit(X train,y train)
         y pred ridge=ridge.predict(X test)
In [22]: from sklearn.metrics import mean squared error
         Ridge Error=mean squared error(y pred ridge, y test)
         Ridge Error
Out[22]: 579521.7970897449
In [23]: from sklearn.metrics import r2 score
         r2 score(y test,y pred ridge)
Out[23]: 0.8421969385523054
```

Elastic

In [24]: **from** sklearn.linear model **import** ElasticNet

from sklearn.model selection import GridSearchCV

```
elastic = ElasticNet()
         parameters = { 'alpha': [1e-15, 1e-10, 1e-8, 1e-4, 1e-3,1e-2, 1, 5, 10, 20]}
         elastic regressor = GridSearchCV(elastic, parameters)
         elastic regressor.fit(X train, y train)
Out[24]: GridSearchCV(estimator=ElasticNet(),
                       param grid={'alpha': [1e-15, 1e-10, 1e-08, 0.0001, 0.001, 0.01, 1,
                                               5, 10, 201})
         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 [25]: elastic regressor.best params
Out[25]: {'alpha': 0.01}
In [26]: elastic=ElasticNet(alpha=0.1)
         elastic.fit(X train,y train)
         y pred elastic=elastic.predict(X test)
In [27]: from sklearn.metrics import mean squared error
         elastic Error=mean squared error(y pred elastic,y test)
         elastic Error
Out[27]: 578326.9853103004
In [28]: from sklearn.metrics import r2_score
         r2 score(y test,y pred elastic)
Out[28]: 0.8425222843073693
```

```
In [ ]:
```

FOR ALL MEAN SQUARE ERROR

In []:		
In [29]:	<pre>from sklearn.metrics import mean_squared_error ##Calulating mean square error mean_squared_error(y_test,ypred)</pre>	##LINEAR REGRESION
Out[29]:	581887.727391353	
In [30]:	<pre>from sklearn.metrics import mean_squared_error Ridge_Error=mean_squared_error(y_pred_ridge,y_test) Ridge_Error</pre>	##RIDGE REGRSSION
Out[30]:	579521.7970897449	
In [31]:	<pre>from sklearn.metrics import mean_squared_error elastic_Error=mean_squared_error(y_pred_elastic,y_test) elastic_Error</pre>	##ELASTIC REGRESSION
Out[31]:	578326.9853103004	
In []:		
	FOR ALL R2_SCORE	

```
In [ ]:
```