```
import pandas as pd
In [1]:
         import warnings
         warnings.filterwarnings("ignore")
In [2]: data=pd.read csv("/home/placement/Downloads/fiat500.csv")
In [3]: #we are doing ridge
         data1=data.loc[(data.model=='lounge')]
          data1
Out[31:
                  ID model engine_power age_in_days
                                                         km previous owners
                                                                                  lat
                                                                                           lon
                                                                                                price
                   1 lounge
                                                      25000
             0
                                      51
                                                 882
                                                                          1 44.907242
                                                                                       8.611560
                                                                                                 8900
                                                     160000
             3
                   4 lounge
                                      51
                                                2739
                                                                          1 40.633171 17.634609
                                                                                                 6000
             6
                   7 lounge
                                      51
                                                 731
                                                      11600
                                                                          1 44.907242
                                                                                       8.611560
                                                                                               10750
             7
                                                1521
                                                      49076
                                                                          1 41.903221 12.495650
                   8 lounge
                                      51
                                                                                                 9190
                                                 366
                                                                                       7.704920
             11
                  12 lounge
                                      51
                                                      17500
                                                                          1 45.069679
                                                                                               10990
           1528
                                                2861 126000
                                                                          1 43.841980 10.515310
                                                                                                 5500
                1529
                     lounge
                                      51
           1529
                1530
                     lounge
                                      51
                                                 731
                                                      22551
                                                                          1 38.122070 13.361120
                                                                                                 9900
           1530
                1531 lounge
                                      51
                                                 670
                                                      29000
                                                                          1 45.764648
                                                                                       8.994500
                                                                                               10800
                1535
                                                     112000
           1534
                     lounge
                                      74
                                                3835
                                                                          1 45.845692
                                                                                       8.666870
                                                                                                 4600
                                                                                                 5990
           1536 1537 lounge
                                                2557
                                                      80750
                                                                          1 45.000702
                                                                                       7.682270
                                      51
         1094 rows × 9 columns
         data1=data1.drop(['ID','lat','lon'],axis=1)
```

Out[5]:		engine_power	age_in_days	km	previous_owners	price	model_lounge
	0	51	882	25000	1	8900	1
	3	51	2739	160000	1	6000	1
	6	51	731	11600	1	10750	1
	7	51	1521	49076	1	9190	1
1	1	51	366	17500	1	10990	1
152	8	51	2861	126000	1	5500	1
152	9	51	731	22551	1	9900	1
153	0	51	670	29000	1	10800	1
153	4	74	3835	112000	1	4600	1
153	6	51	2557	80750	1	5990	1

1094 rows × 6 columns

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

```
In [7]: 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 [8]: 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[8]: GridSearchCV(estimator=Ridge(),
                       param grid={'alpha': [1e-15, 1e-10, 1e-08, 0.0001, 0.001, 0.01, 1,
                                               5, 10, 20, 301})
         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 [9]: ridge regressor.best params
 Out[9]: {'alpha': 30}
In [10]: ridge=Ridge(alpha=30)
          ridge.fit(X train,y train)
         y pred ridge=ridge.predict(X test)
In [11]: from sklearn.metrics import mean squared error
         Ridge Error=mean squared error(y pred ridge,y test)
         Ridge Error
Out[11]: 519771.8129989745
```

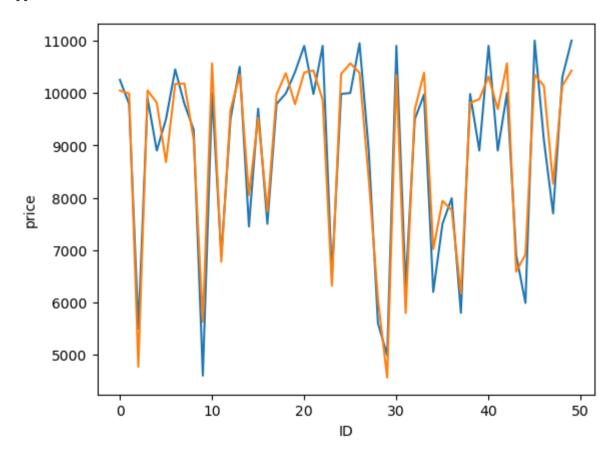
Out[13]:

	index	price	predicted	ID
0	676	10250	10045.347779	0
1	215	9790	9989.171535	1
2	146	5500	4769.099603	2
3	1319	9900	10048.683238	3
4	1041	8900	9813.944798	4
357	757	6000	5640.378648	357
358	167	10950	10431.681162	358
359	156	8000	8765.506865	359
360	1145	10700	10384.884273	360
361	1393	9400	9929.721685	361

362 rows × 4 columns

```
In [14]: import seaborn as sns
import matplotlib.pyplot as plt
sns.lineplot(x='ID',y='price',data=Results.head(50))
sns.lineplot(x='ID',y='predicted',data=Results.head(50))
plt.plot()
```

Out[14]: []



In []: