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

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

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
11	51	366	17500	1	10990	1
1528	51	2861	126000	1	5500	1
1529	51	731	22551	1	9900	1
1530	51	670	29000	1	10800	1
1534	74	3835	112000	1	4600	1
1536	51	2557	80750	1	5990	1

1094 rows × 6 columns

```
In [6]: y=data1['price']
X=data1.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)

jun20ridge2 - Jupyter Notebook

```
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
                ► Ridge
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
In [12]: from sklearn.metrics import r2 score
         r2 score(y test,y pred ridge)
Out[12]: 0.8373030813683994
```

```
In [13]: Results=pd.DataFrame(columns=['price','predicted'])
    Results['price']=y_test
    Results['predicted']=y_pred_ridge
    Results=Results.reset_index()
    Results['ID']=Results.index
    Results
```

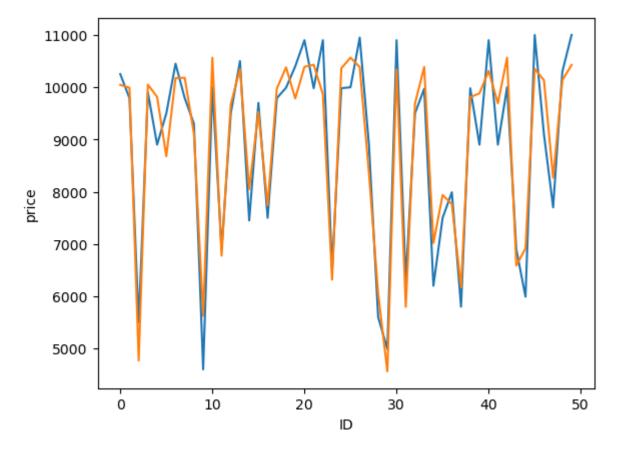
Out[13]:

	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 [16]: 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[16]: []



In []: