```
In [13]:
          import pandas as pd
           import warnings
           warnings.filterwarnings("ignore")
In [14]: data=pd.read csv("/home/placement/Downloads/fiat500.csv")
In [24]:
          data.head(10)
Out[24]:
                  model engine_power age_in_days
                                                      km previous owners
                                                                               lat
                                                                                        lon
                                                                                              price
                  lounge
            0
               1
                                   51
                                              882
                                                   25000
                                                                       1 44.907242
                                                                                    8.611560
                                                                                              8900
               2
                                   51
                                             1186
                                                   32500
                                                                       1 45.666359 12.241890
                                                                                              8800
                    pop
                    sport
                                   74
                                             4658
                                                  142228
                                                                       1 45.503300 11.417840
                                                                                              4200
                                   51
                                             2739
                                                  160000
                                                                       1 40.633171 17.634609
                                                                                              6000
                  lounge
                                                  106880
                                   73
                                             3074
                                                                       1 41.903221 12.495650
                                                                                              5700
                     pop
                                             3623
                                                   70225
                                                                       1 45.000702
                                                                                    7.682270
                                                                                              7900
                                   74
                     pop
                                              731
                                                   11600
                                                                                    8.611560
                  lounge
                                   51
                                                                       1 44.907242
                                                                                            10750
               8
                  lounge
                                   51
                                             1521
                                                   49076
                                                                       1 41.903221 12.495650
                                                                                              9190
                                   73
                                             4049
                                                   76000
                                                                         45.548000 11.549470
                                                                                              5600
                    sport
            9 10
                                   51
                                             3653
                                                   89000
                                                                       1 45.438301 10.991700
                                                                                              6000
                    sport
In [16]:
           data1=data.loc[(data.previous owners==1)]
```

localhost:8888/notebooks/Elastic 22.ipynb

In [17]: data1

Out[17]:

	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	рор	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

1389 rows × 9 columns

```
In [18]: data1=data.drop(['lat','lon','lon'],axis=1)
```

In [19]: datal=pd.get_dummies(datal)

In [20]: data1

Out[20]:

	ID	engine_power	age_in_days	km	previous_owners	price	model_lounge	model_pop	model_sport
0	1	51	882	25000	1	8900	1	0	0
1	2	51	1186	32500	1	8800	0	1	0
2	3	74	4658	142228	1	4200	0	0	1
3	4	51	2739	160000	1	6000	1	0	0
4	5	73	3074	106880	1	5700	0	1	0
1533	1534	51	3712	115280	1	5200	0	0	1
1534	1535	74	3835	112000	1	4600	1	0	0
1535	1536	51	2223	60457	1	7500	0	1	0
1536	1537	51	2557	80750	1	5990	1	0	0
1537	1538	51	1766	54276	1	7900	0	1	0

1538 rows × 9 columns

```
In [21]: y=data1['price']
x=data1.drop('price',axis=1)
```

```
In [22]: 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 [26]: from sklearn.model selection import GridSearchCV
         from sklearn.linear model import ElasticNet
         elastic = ElasticNet()
         parameters={'alpha':[1e-15,1e-10,1e-8,1e-4,1e-3,1e-2,1,5,10,20,30]}
         elastic regressor=GridSearchCV(elastic,parameters)
         elastic_regressor.fit(X_train,Y_train)
Out[26]:
                GridSearchCV
          ▶ estimator: ElasticNet
                ▶ ElasticNet
In [27]: elastic regressor.best params
Out[27]: {'alpha': 0.01}
In [28]: elastic=ElasticNet(alpha=0.1)
         elastic.fit(X train,Y train)
         y_pred_elastic=elastic.predict(X_test)
In [29]: from sklearn.metrics import r2_score
         r2_score(Y_test,y_pred_elastic)
Out[29]: 0.8414565299012147
```

```
In [30]: from sklearn.metrics import mean_squared_error
    elastic_Error=mean_squared_error(y_pred_elastic,Y_test)
    elastic_Error

Out[30]: 582240.9011940917

In [31]: Results=pd.DataFrame(columns=['price','predicted'])
    Results['price']=Y_test
    Results['predicted']=y_pred_elastic
    Results=Results.reset_index()
    Results['ID']=Results.index
    Results
```

Out[31]:

	index	price	predicted	ID
0	481	7900	5903.123297	0
1	76	7900	7223.403259	1
2	1502	9400	9794.808007	2
3	669	8500	9724.783858	3
4	1409	9700	9974.200594	4
503	291	10900	10069.090270	503
504	596	5699	6320.490299	504
505	1489	9500	9959.608673	505
506	1436	6990	8310.780259	506
507	575	10900	10384.079146	507

508 rows × 4 columns

```
In [33]: 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[33]: []

