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

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)
data1
```

Out[5]:

	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 [6]:

```
data2=data1.loc[(data.model=='lounge')]
data2
```

Out[6]:

```
      model  engine_power  age_in_days      km  previous_owners  price
   0  lounge           51          882   25000                1   8900
   3  lounge           51         2739  160000                1   6000
   6  lounge           51          731   11600                1  10750
   7  lounge           51         1521   49076                1   9190
  11  lounge           51          366   17500                1  10990
 ...     ...          ...          ...     ...              ...    ...
1528  lounge           51         2861  126000                1   5500
1529  lounge           51          731   22551                1   9900
1530  lounge           51          670   29000                1  10800
1534  lounge           74         3835  112000                1   4600
1536  lounge           51         2557   80750                1   5990

1094 rows × 6 columns
```

In [7]:
```
data=pd.get_dummies(data1)
data
```

Out[7]:

	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 [8]:
```
data.shape
```

Out[8]:
```
(1538, 8)
```

In [9]:

```
data2=pd.get_dummies(data2)
data2
```

Out[9]:

```
      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 [10]:

```
data2.shape
```

Out[10]:

```
(1094, 6)
```

In [11]:

```
y=data2['price']
x=data2.drop('price',axis=1)
```

In [12]:

```
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 [13]:
```
x_test.head(5)
```

Out[13]:

	engine_power	age_in_days	km	previous_owners	model_lounge
676	51	762	18609	1	1
215	51	701	25000	1	1
146	51	4018	152900	1	1
1319	51	731	20025	1	1
1041	51	640	38231	1	1

In [14]:
```
x_train.shape
```

Out[14]:
```
(732, 5)
```

In [15]:
```
y_train
```

Out[15]:
```
441      8980
701     10300
695      5880
1415    10490
404      9499
        ...
459     10850
654      5900
189     10000
1455     9400
1218     8900
Name: price, Length: 732, dtype: int64
```

In [16]:

```
y_test.head()
```

Out[16]:

```
676     10250
215      9790
146      5500
1319     9900
1041     8900
Name: price, dtype: int64
```

In [17]:

```
y_train.shape
```

Out[17]:

```
(732,)
```

In [18]:

```
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[18]:

```
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 [19]:

```
ridge_regressor.best_params_
```

Out[19]:

```
{'alpha': 30}
```

In [20]:

```
ridge=Ridge(alpha=30)
ridge.fit(x_train,y_train)
y_pred_ridge=ridge.predict(x_test)
```

In [21]:

```
from sklearn.metrics import mean_squared_error
Ridge_Error=mean_squared_error(y_pred_ridge,y_test)
Ridge_Error
```

Out[21]:

```
519771.8129989745
```

In [22]:

```
from sklearn.metrics import r2_score
r2_score(y_test,y_pred_ridge)
```

Out[22]:

```
0.8373030813683994
```

In [24]:

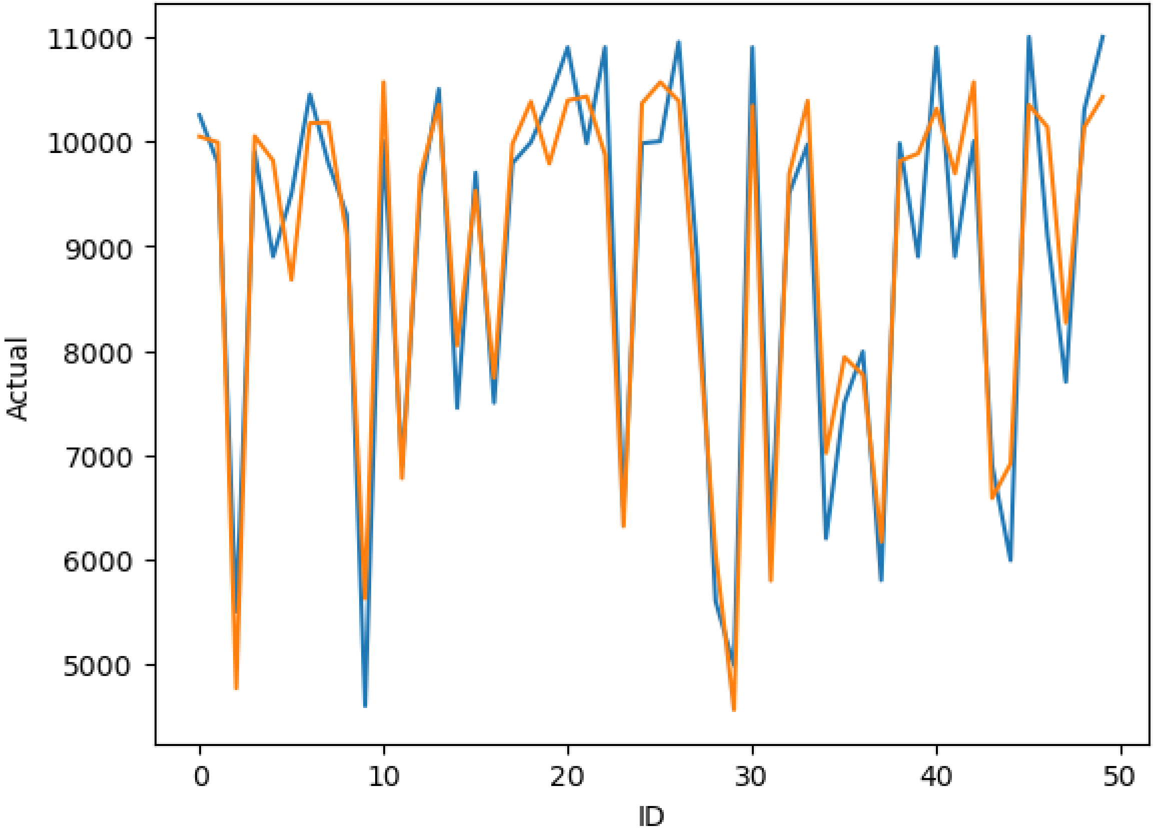
```
Results=pd.DataFrame(columns=['Actual','predicted'])
Results['Actual']=y_test
Results['predicted']=y_pred_ridge
Results=Results.reset_index()
Results['ID']=Results.index
Results.head(10)
```

Out[24]:

	index	Actual	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
5	1425	9500	8678.143561	5
6	409	10450	10173.797921	6
7	617	9790	10180.627008	7
8	1526	9300	9107.315259	8
9	1010	4600	5625.007407	9

```
In [25]: import seaborn as sns
         import matplotlib.pyplot as plt
         sns.lineplot(x='ID',y='Actual',data=Results.head(50))
         sns.lineplot(x='ID',y='predicted',data=Results.head(50))
         plt.plot()

Out[25]: []
```



```
In [ ]:
```