In [1]:
```
import pandas as pd
```

In [2]:
```
import warnings
warnings.filterwarnings("ignore")
```

In [3]:
```
data=pd.read_csv("/home/placement/Desktop/EEE(238)/fiat500.csv")
```

In [4]:
```
data.describe()
```

Out[4]:

	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 [5]:
```
data
```

Out[5]:

	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

In [6]:
```
data1=data.drop(['ID','lat','lon'],axis=1)
```

In [7]:
```
data1
```

Out[7]:

	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 [8]:
```
data1.shape
```

Out[8]:
```
(1538, 6)
```

In [9]:
```
data2=data1.loc[(data1.model=='lounge')]
```

In [10]:
```
data2=pd.get_dummies(data2)
```

In [11]:
```
data2
```

Out[11]:

	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 [12]:
```
y=data2['price']
x=data2.drop('price',axis=1)
```

In [13]:
```
y
```

Out[13]:
```
0        8900
3        6000
6       10750
7        9190
11      10990
        ...
1528     5500
1529     9900
1530    10800
1534     4600
1536     5990
Name: price, Length: 1094, dtype: int64
```

In [14]:
```
x
```

Out[14]:

	engine_power	age_in_days	km	previous_owners	model_lounge
0	51	882	25000	1	1
3	51	2739	160000	1	1
6	51	731	11600	1	1
7	51	1521	49076	1	1
11	51	366	17500	1	1
...	...	...	...	...	...
1528	51	2861	126000	1	1
1529	51	731	22551	1	1
1530	51	670	29000	1	1
1534	74	3835	112000	1	1
1536	51	2557	80750	1	1

1094 rows × 5 columns

In [15]:
```
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 [16]:
```
x_test.head()
```

Out[16]:

	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 [17]:
```
y_test.head()
```

Out[17]:
```
676     10250
215      9790
146      5500
1319     9900
1041     8900
Name: price, dtype: int64
```

In [18]:
```
x_train.head()
```

Out[18]:

	engine_power	age_in_days	km	previous_owners	model_lounge
441	51	762	36448	1	1
701	51	701	27100	1	1
695	51	3197	51083	1	1
1415	51	670	33000	1	1
404	51	456	14000	1	1

In [19]:
```
y_train.head()
```

Out[19]:
```
441      8980
701     10300
695      5880
1415    10490
404      9499
Name: price, dtype: int64
```

In [20]:
```
x_train
```

Out[20]:

	engine_power	age_in_days	km	previous_owners	model_lounge
441	51	762	36448	1	1
701	51	701	27100	1	1
695	51	3197	51083	1	1
1415	51	670	33000	1	1
404	51	456	14000	1	1
...	...	...	...	...	...
459	51	397	15628	1	1
654	51	3227	95554	1	1
189	51	1431	81900	1	1
1455	51	701	33942	1	1
1218	51	882	25000	1	1

732 rows × 5 columns

In [21]:
```
y_train
```

Out[21]:
```
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 [22]:
```
x_test
```

Out[22]:

	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
...	...	...	...	...	...
757	51	4018	102841	1	1
167	51	397	15341	1	1
156	51	1858	35304	1	1
1145	51	456	14970	1	1
1393	51	609	32665	2	1

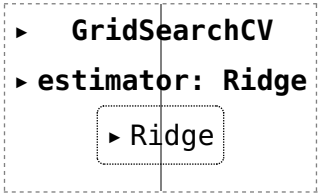
362 rows × 5 columns

In [23]:
```
y_test
```

Out[23]:
```
676     10250
215      9790
146      5500
1319     9900
1041     8900
        ...
757      6000
167     10950
156      8000
1145    10700
1393     9400
Name: price, Length: 362, dtype: int64
```

In [24]:
```
from sklearn.model_selection import GridSearchCV#ridge regression
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[24]:



In [25]:
```
ridge_regressor.best_params_
```

Out[25]:
```
{'alpha': 30}
```

In [26]:
```
ridge=Ridge(alpha=30)
ridge.fit(x_train,y_train)
y_pred_ridge=ridge.predict(x_test)
```

In [27]:
```
y_pred_ridge
```

Out[27]:
```
array([10045.34777889,  9989.17153543,  4769.09960336, 10048.68323752,
        9813.94479825,  8678.14356117, 10173.79792135, 10180.6270078 ,
        9107.31525896,  5625.00740732, 10565.71108835,  6776.12815534,
        9677.36019112, 10348.97135978,  8049.20104733,  9526.33575316,
        7738.85607226,  9973.09944563, 10379.76191917,  9784.95620261,
       10390.79428386, 10429.52293694,  9867.32992522,  6316.76795239,
       10363.01826786, 10565.71108835, 10385.15644406,  8356.2693706 ,
        6052.94959183,  4562.66804027, 10340.47145405,  5796.55307957,
        9687.69883182, 10386.93279686,  7018.31868443,  7936.55917599,
        7765.92126381,  6169.45640953,  9811.27845178,  9882.52937837,
       10312.76262569,  9691.63232633, 10565.71108835,  6585.82855773,
        6916.6311432 , 10347.90965216, 10136.14357831,  8266.05175267,
       10133.53282186, 10426.05302378, 10264.14549009,  9629.21583316,
        9977.36553225,  9716.74149368,  9353.11972737,  9573.46229983,
        6761.6689103 ,  9804.79795157,  9932.37164515,  5625.07326046,
       10146.17519266, 10332.82578954,  9734.59814219,  6678.28902489,
       10293.21149128, 10312.42707921,  9427.86530055,  9815.46093328,
       10394.02774477, 10436.31090369,  7098.97365343,  9677.370361  ,
        9828.47077394,  7021.16294159,  9930.12732016, 10196.92829788,
        8386.74648114,  9540.25435824,  9765.3639485 , 10368.1517171 ,
       10082.88941809,  6357.42683864, 10430.78719546, 10093.60179309,
        9516.80780142,  7122.64740135,  7719.55327264,  9832.05621279,
        9757.42338865, 10436.88527011,  6004.71929229,  9921.3339132 ,
        8877.95292358, 10041.38666426, 10462.68481494,  7733.76691433,
        8850.71117831, 10421.05777646,  6942.97382254,  6912.87326813,
        6066.80915414,  6315.26734883,  9774.79076844,  6398.14508382,
        9806.68987526,  9801.49099341, 10501.27761938,  5604.99606646,
        9794.78184597,  6999.20995646,  9467.39307862,  9706.5227695 ,
       10378.72288778, 10455.23018537,  6012.0099509 ,  5118.30421508,
        9752.19360218,  9931.77067455,  5449.00003099, 10334.91439469,
        9977.36553225,  9855.40550683, 10381.30992956,  4889.95648612,
        5226.68791203,  9813.4154118 , 10324.79575513, 10315.57646644,
        7935.99268535,  9857.80550646,  9808.2552455 ,  9853.29949663,
        9977.12724572,  9697.81111657, 10462.92321435,  9239.75790604,
        8056.55468638, 10437.56406678,  8673.59658098,  7098.99055677,
        7816.00486189,  8942.22604215, 10187.4782851 , 10424.94541065,
        8172.19745151, 10070.42740734,  5450.70371602,  9057.23399734,
        9940.60129574, 10141.83209263, 10029.2031342 ,  5882.88399169,
```

```
 6988.85761289,  9792.97361779,  9740.70731227,  5189.58908619,
10448.51710998, 10350.05678205,  5930.14249834,  7973.31320996,
10408.72918181, 10398.45493529,  9622.06348934,  7981.39552103,
 9831.98659262, 10339.22453573, 10450.74495548,  9894.02540083,
 8829.9851558 ,  9632.58136126, 10108.59863049,  7997.00644995,
10378.37478692, 10250.24717957,  9665.62091858,  9985.97917741,
10382.21603023, 10409.01979636, 10371.00614414,  8581.84728281,
10394.31253507,  9657.1968326 ,  9854.59232047, 10053.40822971,
 9611.32203526,  8756.17468877, 10261.99836019,  6636.54831182,
 9320.38823484,  8937.41241054,  4584.0579125 ,  6496.04532397,
 6874.60996914, 10445.08831652, 10417.86155817, 10465.59728135,
 5908.85959612, 10364.13115218,  7440.14661653, 10462.92321435,
 7364.84878898, 10073.35249298,  9996.90323477, 10382.04539989,
 8468.286263  ,  7128.4688464 , 10157.57549578,  5690.55405714,
10473.87098636,  9394.7588997 ,  7965.99904423,  8857.22351744,
 6567.36918077, 10264.14549009,  9723.18859473,  9650.11302532,
 7558.9528078 ,  9092.04408323, 10195.32072431,  9417.92066621,
 9023.56448358, 10203.89031505,  9857.70928211,  7869.82213501,
10094.23854452, 10016.87839982, 10393.56846387,  9939.33366233,
 7473.59251736,  9548.45848828,  6931.35602476, 10128.41187897,
 5476.07841754,  9482.94932674, 10004.71809905,  8840.35803082,
 9183.58600823, 10193.44728929,  6693.42378212,  9891.64090994,
 7941.18281108,  8893.89594292,  9977.36553225,  9646.02284023,
10115.36826665,  9977.36553225, 10416.41693961,  9866.37068388,
 9612.95325864,  6763.86620095,  7758.56279091, 10511.3377342 ,
10110.03601389,  5985.71210505,  5379.55474479, 10086.77270664,
 8590.88930427, 10392.53722069,  6098.57359345,  9709.1703    ,
10325.18497569,  9715.54054572,  7572.96558285,  9765.3639485 ,
 9813.51896171,  7228.99093841,  8007.83781593,  9904.86874258,
10012.0567824 ,  7912.56700171, 10375.39325339,  9673.69685851,
10398.45493529,  5460.03755965,  7280.5349007 ,  9810.03252385,
 5418.53480582, 10380.92701862,  8877.95292358,  8041.02830728,
 9790.60757038, 10459.71922749,  6263.41225143,  9554.10680845,
10436.61206061,  9968.01613467,  9853.07133478,  9880.78402341,
10046.86686828,  9961.05411749,  9977.36553225, 10370.13693042,
10161.11057059,  9192.46590982,  9830.4375438 ,  8012.10493814,
 9453.45887438,  7877.55383435,  5587.00369094, 10386.10262599,
 9878.41601203,  7003.92764864,  7173.72195805, 10229.36112804,
10403.26253937,  9734.78855924, 10021.13736416,  9349.69093391,
10387.84313028,  7733.76691433, 10056.64892295,  7656.3382783 ,
 5702.23720624,  9811.21606677, 10415.02980736,  9872.39093245,
```

```
 3901.76632267,  9789.01354159, 10420.22864401,  7511.00890116,
 7204.44567391, 10074.56980756,  9103.99857109,  9688.23834295,
 5656.79259391,  9811.27845178, 10076.19675014, 10339.10901004,
10125.897234  ,  5479.82528759,  5904.27386392,  7576.88220289,
 9765.75023434,  6890.55397887,  6691.60769781, 10461.98334203,
 6319.35371886,  8877.95292358, 10196.29154645, 10317.76515621,
 9935.62007856, 10042.02341569, 10431.48866837, 10403.11317719,
 5859.48252489,  5133.29631187, 10447.51975135, 10307.64700202,
 5794.71820485,  5855.33690786,  8722.08988368, 10059.34866858,
10732.79990752,  8834.7001814 , 10565.71108835, 10324.31472354,
 6791.95158544,  5640.37864803, 10431.68116227,  8765.50686495,
10384.88427298,  9929.72168494])
```

In [28]:
```
y_test
```

Out[28]:
```
676     10250
215      9790
146      5500
1319     9900
1041     8900
        ...
757      6000
167     10950
156      8000
1145    10700
1393     9400
Name: price, Length: 362, dtype: int64
```

In [29]:
```
from sklearn.metrics import mean_squared_error
Ridge_Error=mean_squared_error(y_pred_ridge,y_test)
Ridge_Error
```

Out[29]:
```
519771.8129989745
```

In [30]:
```
from sklearn.metrics import r2_score
r2_score(y_test,y_pred_ridge)
```

Out[30]:
```
0.8373030813683994
```

In [31]:
```
Results=pd.DataFrame(columns=['Actual','Predicted'])
Results['Actual']=y_test
Results['Predicted']=y_pred_ridge
#Results['Km']=x_test['Km']
Results=Results.reset_index()
Results['ID']=Results.index
Results.head(10)
```

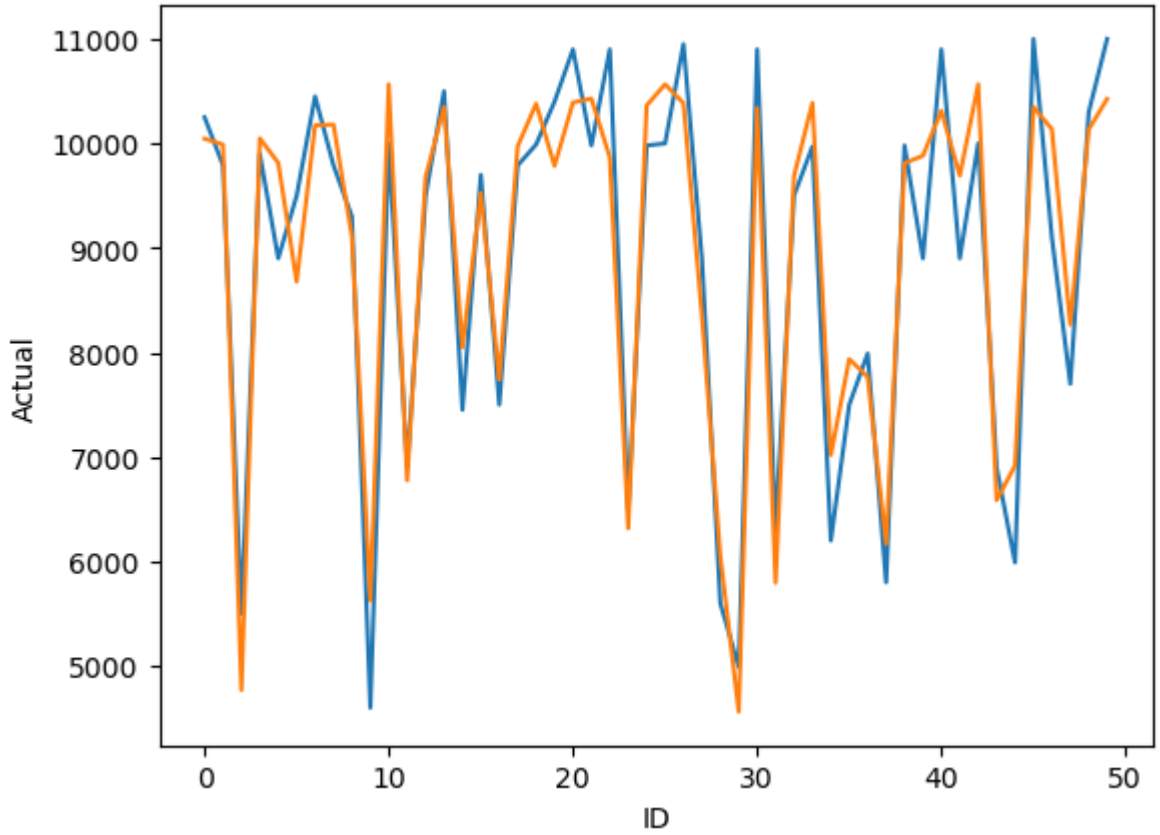
Out[31]:

	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 [35]:
```
import seaborn as sns
import matplotlib.pyplot as plt
```

In [40]:
```
sns.lineplot(x='ID',y='Actual',data=Results.head(50))
sns.lineplot(x='ID',y='Predicted',data=Results.head(50))
plt.plot()
```

Out[40]: []



In [ ]:

In [ ]: