In [86]: import pandas as pd
import seaborn as sns

In [87]: data=pd.read_csv("/home/placement/Desktop/usha g1/fiat500.csv")

In [88]: data.describe()

Out[88]:

	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 [89]: data.head()

Out[89]:

	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 [90]: data

Out[90]:

	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	рор	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 [91]: data1=data.loc[(data.previous_owners==1)]
 data1

Out[91]:

	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	рор	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	рор	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

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```
In [92]: data2=data1.drop(['lat','lon','ID'],axis=1)
data2
```

Out[92]:

	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

1389 rows × 6 columns

```
In [93]: data2=pd.get_dummies(data2)
```

In [94]: data2.shape

Out[94]: (1389, 8)

In [95]: data2

Out[95]:

engine_power	age_in_days	km	previous_owners	price	model_lounge	model_pop	model_sport
51	882	25000	1	8900	1	0	0
51	1186	32500	1	8800	0	1	0
74	4658	142228	1	4200	0	0	1
51	2739	160000	1	6000	1	0	0
73	3074	106880	1	5700	0	1	0
51	3712	115280	1	5200	0	0	1
74	3835	112000	1	4600	1	0	0
51	2223	60457	1	7500	0	1	0
51	2557	80750	1	5990	1	0	0
51	1766	54276	1	7900	0	1	0
	51 51 74 51 73 51 74 51	51 1186 74 4658 51 2739 73 3074 51 3712 74 3835 51 2223 51 2557	51 882 25000 51 1186 32500 74 4658 142228 51 2739 160000 73 3074 106880 51 3712 115280 74 3835 112000 51 2223 60457 51 2557 80750	51 882 25000 1 51 1186 32500 1 74 4658 142228 1 51 2739 160000 1 73 3074 106880 1 51 3712 115280 1 74 3835 112000 1 51 2223 60457 1 51 2557 80750 1	51 882 25000 1 8900 51 1186 32500 1 8800 74 4658 142228 1 4200 51 2739 160000 1 6000 73 3074 106880 1 5700 51 3712 115280 1 5200 74 3835 112000 1 4600 51 2223 60457 1 7500 51 2557 80750 1 5990	51 882 25000 1 8900 1 51 1186 32500 1 8800 0 74 4658 142228 1 4200 0 51 2739 160000 1 6000 1 73 3074 106880 1 5700 0 51 3712 115280 1 5200 0 74 3835 112000 1 4600 1 51 2223 60457 1 7500 0 51 2557 80750 1 5990 1	51 882 25000 1 8900 1 0 51 1186 32500 1 8800 0 1 74 4658 142228 1 4200 0 0 51 2739 160000 1 6000 1 0 73 3074 106880 1 5700 0 1 51 3712 115280 1 5200 0 0 74 3835 112000 1 4600 1 0 51 2223 60457 1 7500 0 1 51 2557 80750 1 5990 1 0

1389 rows × 8 columns

```
In [96]: y=data2['price']
x=data2.drop('price',axis=1)
```

```
In [97]: y
Out[97]: 0
                   8900
                   8800
           2
                   4200
           3
                   6000
                   5700
           4
           1533
                   5200
           1534
                   4600
           1535
                   7500
           1536
                   5990
           1537
                   7900
           Name: price, Length: 1389, dtype: int64
In [98]: 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 [99]: x_test.head(5)
Out[99]:
                                          km previous_owners model_lounge model_pop model_sport
                engine_power age_in_days
            625
                         51
                                  3347 148000
                                                         1
                                                                      1
                                                                               0
                                                                                           0
            187
                         51
                                       117000
                                                         1
                                                                      1
                                                                               0
                                                                                           0
                                  4322
            279
                                  4322 120000
                                                         1
                                                                      0
                         51
                                                                               1
                                       12500
            734
                         51
                                   974
                                                         1
                                                                      0
                                                                               1
            315
                         51
                                  1096
                                        37000
                                                         1
                                                                     1
                                                                               0
                                                                                          0
In [100]: import warnings
           warnings.filterwarnings("ignore")
```

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```
In [101]:
          from sklearn.linear model import ElasticNet
          from sklearn.model selection import GridSearchCV
          elastic = ElasticNet()
          parameters = { 'alpha': [1e-15, 1e-10, 1e-8, 1e-4, 1e-3,1e-2, 1, 5, 10, 20]}
          elastic regressor = GridSearchCV(elastic, parameters)
          elastic_regressor.fit(x_train, y_train)
Out[101]:
                                              GridSearchCV
           GridSearchCV(estimator=ElasticNet(),
                        param grid={'alpha': [1e-15, 1e-10, 1e-08, 0.0001, 0.001, 0.01, 1,
                                               5, 10, 20]})
                                        ▶ estimator: ElasticNet
                                              ▶ ElasticNet
In [102]: elastic_regressor.best_params_
Out[102]: {'alpha': 0.01}
In [103]: elastic=ElasticNet(alpha=0.1)
          elastic.fit(x train,y train)
          y pred elastic=elastic.predict(x test)
In [104]: from sklearn.metrics import r2 score
          r2 score(y test,y pred elastic)
Out[104]: 0.8601270407940889
```

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```
In [105]: from sklearn.metrics import mean_squared_error
    elastic_Error=mean_squared_error(y_pred_elastic,y_test)
    elastic_Error

Out[105]: 515678.8171884504

In [106]: Results= pd.DataFrame(columns=['Actual','Predicted'])
    Results['Actual']=y_test
    Results['Predicted']=y_pred_elastic
    Results=Results.reset_index()
    Results['Id']=Results.index
    Results.head(10)
```

Out[106]:

	index	Actual	Predicted	ld
0	625	5400	5478.361166	0
1	187	5399	5124.950418	1
2	279	4900	4833.208393	2
3	734	10500	9688.909121	3
4	315	9300	9402.252771	4
5	652	10850	10343.695328	5
6	1472	9500	9831.910050	6
7	619	7999	8380.454867	7
8	992	6300	5910.641522	8
9	1154	10000	10142.252255	9

```
In [107]: 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[107]: []

