```
In [27]: import pandas as pd
In [28]: data=pd.read csv("/home/placement/Downloads/fiat500.csv")
In [29]:
           data.describe()
Out[29]:
                            ID engine power
                                              age in days
                                                                     km previous owners
                                                                                                   lat
                                                                                                              lon
                                                                                                                          price
                                                                                          1538.000000
                                                                                                      1538.000000
                                                                                                                    1538.000000
             count 1538.000000
                                  1538.000000
                                              1538.000000
                                                             1538.000000
                                                                              1538.000000
                     769.500000
                                    51.904421
                                              1650.980494
                                                            53396.011704
                                                                                 1.123537
                                                                                            43.541361
                                                                                                         11.563428
                                                                                                                    8576.003901
             mean
                                                                                                          2.328190
               std
                    444.126671
                                    3.988023
                                              1289.522278
                                                            40046.830723
                                                                                 0.416423
                                                                                             2.133518
                                                                                                                    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 [30]: | data2=data.drop(['ID','lat','lon'],axis=1)
```

In [31]: data2

Out[31]:

	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 [32]: data2=pd.get_dummies(data2)

In [33]: data2

Out[33]:

	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 [34]: data2.shape
Out[34]: (1538, 8)
```

```
In [35]: y=data2['price']#predicted value removed from data frame
x=data2.drop(['price'],axis=1)
```

```
In [36]: y
Out[36]: 0
                 8900
                 8800
         2
                 4200
         3
                 6000
         4
                 5700
                  . . .
         1533
                 5200
         1534
                 4600
         1535
                 7500
         1536
                 5990
         1537
                 7900
         Name: price, Length: 1538, dtype: int64
```

In [37]: x

Out[37]:

	engine_power	age_in_days	km	previous_owners	model_lounge	model_pop	model_sport
0	51	882	25000	1	1	0	0
1	51	1186	32500	1	0	1	0
2	74	4658	142228	1	0	0	1
3	51	2739	160000	1	1	0	0
4	73	3074	106880	1	0	1	0
			•••				
1533	51	3712	115280	1	0	0	1
1534	74	3835	112000	1	1	0	0
1535	51	2223	60457	1	0	1	0
1536	51	2557	80750	1	1	0	0
1537	51	1766	54276	1	0	1	0

1538 rows × 7 columns

```
In [38]: 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 [39]: data3=data2.loc[(data.model=='lounge')]

In [40]: data3

Out[40]:

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

1094 rows × 8 columns

In [41]: data3=pd.get_dummies(data3)

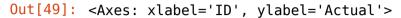
```
In [42]: from sklearn.model selection import GridSearchCV
         from sklearn.linear model import Ridge
         alpha = [1e-15, 1e-\overline{10}, 1e-8, 1e-4, 1e-3, 1e-2, 1, 5, 10, 20, 30]
         ridge = Ridge()
         parameters = {'alpha': alpha}
         ridge regressor = GridSearchCV(ridge, parameters)
         ridge regressor.fit(x train, y train)
Out[42]:
          ▶ GridSearchCV
           ▶ estimator: Ridge
                ▶ Ridge
In [43]: import warnings
         warnings.filterwarnings("ignore")
In [44]: ridge regressor.best params
Out[44]: {'alpha': 30}
In [45]: ridge=Ridge(alpha=30)
         ridge.fit(x train,y train)
         y pred ridge=ridge.predict(x test)
In [46]: from sklearn.metrics import mean squared error
         Ridge Error=mean squared error(y pred ridge, y test)
         Ridge Error
Out[46]: 579521.7970897449
In [47]: from sklearn.metrics import r2 score
         r2 score(y test,y pred ridge)
Out[47]: 0.8421969385523054
```

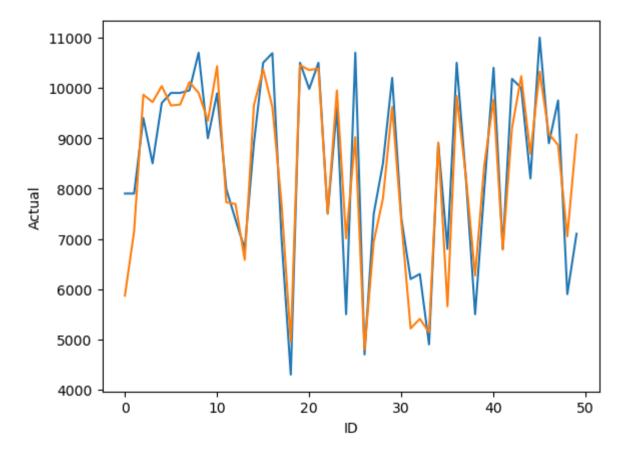
```
In [48]: Results=pd.DataFrame(columns=['Actual','Predicted'])
    Results['Actual']=y_test
    Results['Predicted']=y_pred_ridge
    Results=Results.reset_index()
    Results['ID']=Results.index #replaces id with index number
    Results.head(10)
```

Out[48]:

index		Actual	Predicted	ID
0	481	7900	5869.741155	0
1	76	7900	7149.563327	1
2	1502	9400	9862.785355	2
3	669	8500	9719.283532	3
4	1409	9700	10035.895686	4
5	1414	9900	9650.311090	5
6	1089	9900	9669.183317	6
7	1507	9950	10115.128380	7
8	970	10700	9900.241944	8
9	1198	8999	9347.080772	9

```
In [49]: import seaborn as sns
import matplotlib.pyplot as plt
sns.lineplot(x='ID',y='Actual',data=Results.head(50)) #red is actual
sns.lineplot(x='ID',y='Predicted',data=Results.head(50)) #blue is predicted
```





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