```
In [71]: import pandas as pd
In [72]: data=pd.read csv("/home/placement/Downloads/fiat500.csv")
           import warnings
In [73]:
           warnings.filterwarnings('ignore')
In [74]: data.describe()
Out[74]:
                            ID engine_power
                                             age_in_days
                                                                    km previous_owners
                                                                                                 lat
                                                                                                             lon
                                                                                                                         price
                                              1538.000000
                   1538.000000
                                 1538.000000
                                                            1538.000000
                                                                             1538.000000
                                                                                         1538.000000
                                                                                                     1538.000000
                                                                                                                   1538.000000
             count
                    769.500000
                                   51.904421
                                              1650.980494
                                                           53396.011704
                                                                                1.123537
                                                                                           43.541361
                                                                                                       11.563428
                                                                                                                   8576.003901
             mean
                    444.126671
                                    3.988023
                                              1289.522278
                                                                                0.416423
                                                                                            2.133518
                                                                                                        2.328190
                                                                                                                   1939.958641
               std
                                                           40046.830723
              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
                   1153.750000
                                                                                1.000000
              75%
                                   51.000000
                                              2616.000000
                                                           79667.750000
                                                                                           45.467960
                                                                                                       12.769040
                                                                                                                 10000.000000
              max 1538.000000
                                   77.000000
                                              4658.000000 235000.000000
                                                                                4.000000
                                                                                           46.795612
                                                                                                       18.365520
                                                                                                                 11100.000000
In [75]: | data1=data.drop(['ID','lat','lon'],axis=1)
```

In [76]:	data1							
					<u> </u>			
	0	lounge	51	882	25000	1	9900	
	1	pop	51	1186	32500	1	8800	
	2	sport	74	4658	142228	1	2200	
	3	lounge	51	2739	160000	1	5000	
	4	рор	73	3074	106880	1	5700	
	1533	sport	51	3712	115280	1	3200	
	1534	lounge	74	3835	112000	1	600	
	1535	рор	51	2223	60457	1	7500	
	1536	lounge	51	2557	80750	1	990	
	1537	рор	51	1766	54276	1	7900	
	1538 r	ows × 6 colu	mns					
n [77]:			[(data.mode					

In [78]: data2

Out[78]:

	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
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
6	7	lounge	51	731	11600	1	44.907242	8.611560	10750
7	8	lounge	51	1521	49076	1	41.903221	12.495650	9190
11	12	lounge	51	366	17500	1	45.069679	7.704920	10990
1528	1529	lounge	51	2861	126000	1	43.841980	10.515310	5500
1529	1530	lounge	51	731	22551	1	38.122070	13.361120	9900
1530	1531	lounge	51	670	29000	1	45.764648	8.994500	10800
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990

1094 rows × 9 columns

In [79]: data2=pd.get_dummies(data2)

In [80]: data2

Out[80]:

	ID	engine_power	age_in_days	km	previous_owners	lat	lon	price	model_lounge
0	1	51	882	25000	1	44.907242	8.611560	8900	1
3	4	51	2739	160000	1	40.633171	17.634609	6000	1
6	7	51	731	11600	1	44.907242	8.611560	10750	1
7	8	51	1521	49076	1	41.903221	12.495650	9190	1
11	12	51	366	17500	1	45.069679	7.704920	10990	1
1528	1529	51	2861	126000	1	43.841980	10.515310	5500	1
1529	1530	51	731	22551	1	38.122070	13.361120	9900	1
1530	1531	51	670	29000	1	45.764648	8.994500	10800	1
1534	1535	74	3835	112000	1	45.845692	8.666870	4600	1
1536	1537	51	2557	80750	1	45.000702	7.682270	5990	1

1094 rows × 9 columns

```
In [81]: data2.shape
Out[81]: (1094, 9)
In [82]: y=data2['price']
x=data2.drop('price',axis=1)
```

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```
In [83]: y
Out[83]: 0
                    8900
                    6000
          3
                   10750
          6
          7
                    9190
          11
                   10990
                   . . .
          1528
                    5500
          1529
                    9900
          1530
                   10800
          1534
                    4600
          1536
                    5990
          Name: price, Length: 1094, dtype: int64
In [84]: 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 [85]: x_test.head(5)
Out[85]:
                                                                                Ion model_lounge
                  ID engine_power age_in_days
                                                km previous_owners
                                                                        lat
            676
                 677
                              51
                                         762
                                              18609
                                                                1 41.572239 13.33369
                                                                                              1
            215
                 216
                              51
                                         701
                                              25000
                                                                1 44.988739
                                                                             9.01050
                                                                                              1
                              51
                                             152900
                                                                1 43.067532 12.55155
                 147
                                        4018
                                                                                              1
            146
                              51
                                                                1 41.689281 13.25494
           1319
               1320
                                         731
                                              20025
                                                                                              1
```

1 41.107880 14.20881

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```
In [86]: x_train.head(5)
```

Out[86]:

	ID	engine_power	age_in_days	km	previous_owners	lat	lon	model_lounge
441	442	51	762	36448	1	45.571220	9.15914	1
701	702	51	701	27100	1	41.903221	12.49565	1
695	696	51	3197	51083	1	45.571220	9.15914	1
1415	1416	51	670	33000	1	42.287029	12.40754	1
404	405	51	456	14000	1	40.840141	14.25226	1

```
In [87]: y_test.head(5)
Out[87]: 676
                 10250
                  9790
         215
         146
                  5500
         1319
                  9900
         1041
                  8900
         Name: price, dtype: int64
In [88]: y_train.head(5)
Out[88]: 441
                  8980
         701
                 10300
         695
                  5880
         1415
                 10490
         404
                  9499
         Name: price, dtype: int64
In [89]: x_train.shape
Out[89]: (732, 8)
```

```
In [90]: y_train
Out[90]: 441
                  8980
         701
                 10300
         695
                  5880
         1415
                 10490
                  9499
         404
                 . . .
         459
                 10850
         654
                  5900
         189
                 10000
                  9400
         1455
         1218
                  8900
         Name: price, Length: 732, dtype: int64
```

In [91]: x_train

Out[91]:

	ID	engine_power	age_in_days	km	previous_owners	lat	lon	model_lounge
441	442	51	762	36448	1	45.571220	9.15914	1
701	702	51	701	27100	1	41.903221	12.49565	1
695	696	51	3197	51083	1	45.571220	9.15914	1
1415	1416	51	670	33000	1	42.287029	12.40754	1
404	405	51	456	14000	1	40.840141	14.25226	1
459	460	51	397	15628	1	45.512569	10.32901	1
654	655	51	3227	95554	1	45.069679	7.70492	1
189	190	51	1431	81900	1	44.138371	12.23882	1
1455	1456	51	701	33942	1	41.107880	14.20881	1
1218	1219	51	882	25000	1	44.907242	8.61156	1

732 rows × 8 columns

```
In [92]: #ridge regression
         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[92]:
          ▶ GridSearchCV
          ▶ estimator: Ridge
                ▶ Ridge
In [93]: ridge regressor.best params
Out[93]: {'alpha': 30}
In [94]: ridge=Ridge(alpha=30)
         ridge.fit(x train,y train)
         y pred ridge=ridge.predict(x test)
In [95]: from sklearn.metrics import mean squared error
         Ridge Error=mean squared error(y pred ridge,y test)
         Ridge Error
Out[95]: 529111.0455362241
In [96]: from sklearn.metrics import r2 score
         r2 score(y test,y pred ridge)
Out[96]: 0.8343797517106646
```

```
In [104]: Results=pd.DataFrame(columns=['Actual', 'Pedicted'])
    Results['Actual']=y_test
    Results['Predicted']=y_pred_ridge
    Results=Results.reset_index()
    Results['ID']=Results.index
    Results.head(10)
```

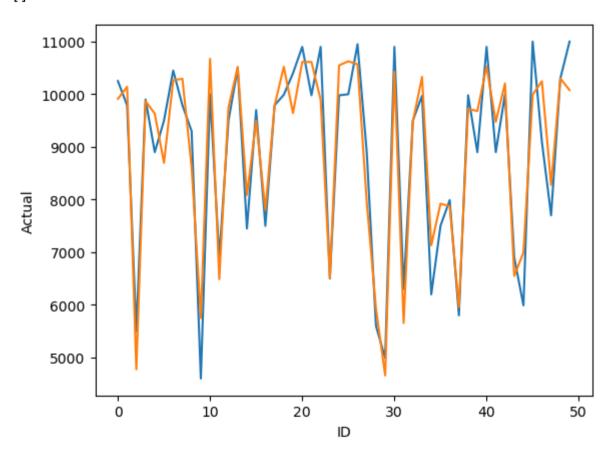
Out[104]:

	index	Actual	Pedicted	Predicted	ID
0	676	10250	NaN	9912.601754	0
1	215	9790	NaN	10141.748493	1
2	146	5500	NaN	4775.235521	2
3	1319	9900	NaN	9870.926966	3
4	1041	8900	NaN	9630.417885	4
5	1425	9500	NaN	8697.092014	5
6	409	10450	NaN	10265.822884	6
7	617	9790	NaN	10293.851867	7
8	1526	9300	NaN	8614.349738	8
9	1010	4600	NaN	5749.673567	9

```
In [105]: import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [107]: sns.lineplot(x='ID',y='Actual',data=Results.head(50))
sns.lineplot(x='ID',y='Predicted',data=Results.head(50))
plt.plot()
```

Out[107]: []



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In []: