```
In [1]: import pandas as pd
         import warnings
         warnings.filterwarnings("ignore")
         data = pd.read csv("/home/placement/Downloads/fiat500.csv")
In [3]: data.describe()
                         ID engine power age in days
                                                                   km previous owners
                                                                                                                          price
                                                                                                 lat
Out[3]:
                                                                                                              lon
         count 1538.000000
                                                           1538.000000
                                                                                        1538.000000
                               1538.000000
                                            1538.000000
                                                                            1538.000000
                                                                                                     1538.000000
                                                                                                                   1538.000000
                 769.500000
                                           1650.980494
                                                          53396.011704
                                                                               1.123537
                                                                                           43.541361
                                                                                                       11.563428
                                                                                                                   8576.003901
                                 51.904421
          mean
            std
                 444.126671
                                  3.988023
                                           1289.522278
                                                          40046.830723
                                                                               0.416423
                                                                                            2.133518
                                                                                                         2.328190
                                                                                                                   1939.958641
                   1.000000
                                 51.000000
                                             366.000000
                                                           1232.000000
                                                                               1.000000
                                                                                           36.855839
                                                                                                         7.245400
           min
                                                                                                                   2500.000000
                 385.250000
           25%
                                 51.000000
                                             670.000000
                                                          20006.250000
                                                                               1.000000
                                                                                           41.802990
                                                                                                         9.505090
                                                                                                                   7122.500000
                 769.500000
                                                                               1.000000
           50%
                                 51.000000
                                            1035.000000
                                                          39031.000000
                                                                                           44.394096
                                                                                                        11.869260
                                                                                                                   9000.000000
               1153.750000
                                                                               1.000000
                                                                                           45.467960
           75%
                                 51.000000
                                            2616.000000
                                                          79667.750000
                                                                                                        12.769040
                                                                                                                  10000.000000
           max 1538.000000
                                 77.000000
                                           4658.000000
                                                        235000.000000
                                                                               4.000000
                                                                                           46.795612
                                                                                                        18.365520 11100.000000
         data.head()
In [4]:
            ID model engine power age in days
                                                                                   lat
                                                                                             Ion price
Out[4]:
                                                       km previous owners
                lounge
                                  51
                                                    25000
                                                                            44.907242
                                                                                        8.611560
         0
             1
                                              882
                                                                                                  8900
             2
         1
                                  51
                                             1186
                                                    32500
                                                                            45.666359 12.241890
                                                                                                  8800
                   pop
         2
             3
                                  74
                                             4658
                                                  142228
                                                                            45.503300 11.417840
                                                                                                  4200
                 sport
         3
                lounge
                                  51
                                             2739
                                                   160000
                                                                            40.633171 17.634609
                                                                                                  6000
         4
             5
                                  73
                                             3074 106880
                                                                          1 41.903221 12.495650
                                                                                                  5700
                   pop
```

```
In [5]: data1=data.drop(['lat','lon','ID'],axis=1)
In [6]: data1
              model engine_power age_in_days
                                                  km previous_owners price
Out[6]:
           0 lounge
                               51
                                          882
                                                25000
                                                                       8900
                pop
                               51
                                         1186
                                                32500
                                                                       8800
                                         4658 142228
                                                                       4200
                               74
                sport
                                         2739 160000
                                                                       6000
              lounge
                               51
                               73
                                                                       5700
                pop
                                         3074 106880
         1533
                               51
                                         3712 115280
                                                                       5200
                sport
         1534 lounge
                               74
                                         3835 112000
                                                                       4600
        1535
                               51
                                         2223
                                                60457
                                                                       7500
                pop
         1536 lounge
                               51
                                         2557
                                                80750
                                                                       5990
                                                                    1 7900
         1537
                pop
                               51
                                         1766
                                                54276
        1538 rows × 6 columns
In [7]: data3=data.loc[(data.model=='lounge')]
        data3
```

Out[7]:		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 [8]: data=pd.get_dummies(data1)
data

	engine_power	age_in_days	km	previous_owners	price	model_lounge	model_pop	model_sport
0	51	882	25000	1	8900	True	False	False
1	51	1186	32500	1	8800	False	True	False
2	74	4658	142228	1	4200	False	False	True
3	51	2739	160000	1	6000	True	False	False
4	73	3074	106880	1	5700	False	True	False
		***				•••		
1533	51	3712	115280	1	5200	False	False	True
1534	74	3835	112000	1	4600	True	False	False
1535	51	2223	60457	1	7500	False	True	False
1536	51	2557	80750	1	5990	True	False	False
1537	51	1766	54276	1	7900	False	True	False
	1 2 3 4 1533 1534 1535 1536	0 51 1 51 2 74 3 51 4 73 1533 51 1534 74 1535 51 1536 51	0 51 882 1 51 1186 2 74 4658 3 51 2739 4 73 3074 1533 51 3712 1534 74 3835 1535 51 2223 1536 51 2557	0 51 882 25000 1 51 1186 32500 2 74 4658 142228 3 51 2739 160000 4 73 3074 106880 1533 51 3712 115280 1534 74 3835 112000 1535 51 2223 60457 1536 51 2557 80750	0 51 882 25000 1 1 51 1186 32500 1 2 74 4658 142228 1 3 51 2739 160000 1 4 73 3074 106880 1 1533 51 3712 115280 1 1534 74 3835 112000 1 1535 51 2223 60457 1 1536 51 2557 80750 1	0 51 882 25000 1 8900 1 51 1186 32500 1 8800 2 74 4658 142228 1 4200 3 51 2739 160000 1 6000 4 73 3074 106880 1 5700 1533 51 3712 115280 1 5200 1534 74 3835 112000 1 4600 1535 51 2223 60457 1 7500 1536 51 2557 80750 1 5990	0 51 882 25000 1 8900 True 1 51 1186 32500 1 8800 False 2 74 4658 142228 1 4200 False 3 51 2739 160000 1 6000 True 4 73 3074 106880 1 5700 False 1533 51 3712 115280 1 5200 False 1534 74 3835 112000 1 4600 True 1535 51 2223 60457 1 7500 False 1536 51 2557 80750 1 5990 True	0 51 882 25000 1 8900 True False 1 51 1186 32500 1 8800 False True 2 74 4658 142228 1 4200 False False 3 51 2739 160000 1 6000 True False 4 73 3074 106880 1 5700 False True 1533 51 3712 115280 1 5200 False False 1534 74 3835 112000 1 4600 True False 1535 51 2223 60457 1 7500 False True 1536 51 2557 80750 1 5990 True False

1538 rows × 8 columns

data.shape

In [9]: data2=pd.get_dummies(data3)
 data2

Out[9]:		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	True
	3	4	51	2739	160000	1	40.633171	17.634609	6000	True
	6	7	51	731	11600	1	44.907242	8.611560	10750	True
	7	8	51	1521	49076	1	41.903221	12.495650	9190	True
	11	12	51	366	17500	1	45.069679	7.704920	10990	True
				•••						
	1528	1529	51	2861	126000	1	43.841980	10.515310	5500	True
	1529	1530	51	731	22551	1	38.122070	13.361120	9900	True
	1530	1531	51	670	29000	1	45.764648	8.994500	10800	True
	1534	1535	74	3835	112000	1	45.845692	8.666870	4600	True
	1536	1537	51	2557	80750	1	45.000702	7.682270	5990	True

1094 rows × 9 columns

```
In [10]: data2.shape
Out[10]: (1094, 9)
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]:		ID	engine_power	age in days	km	previous_owners	lat	lon	model_lounge
	676	677	51		18609		41.572239		True
	215	216	51	701	25000	1	44.988739	9.01050	True
	146	147	51	4018	152900	1	43.067532	12.55155	True
	1319	1320	51	731	20025	1	41.689281	13.25494	True
	1041	1042	51	640	38231	1	41.107880	14.20881	True
In [14]:	x_tra	in.sh	ape						
Out[14]:	(732,	8)							
In [15]:	y tra	in							
Out[15]:	701 695 1415 404 459 654 189 1455 1218	10 5 10 9 10 5 10 9	980 300 880 490 499 850 900 000 400 900 e, Length: 7	32, dtype: i	nt64				
In [16]:	y_tes	t.hea	d()						
Out[16]:	215 146 1319 1041 Name:	9 5 9 8 pric	250 790 500 900 900 e, dtype: in	t64					
In [17]:	y_tra	in.sh	ape						

```
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)
         GridSearchCV
Out[18]:
          ► estimator: Ridge
                ► Ridge
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]: 529111.0455362239
In [22]: from sklearn.metrics import r2 score
         r2 score(y test,y pred ridge)
Out[22]: 0.8343797517106646
```

about:srcdoc

```
In [23]: import seaborn as sns
         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)
                            Predicted ID
            index Actual
Out[23]:
         0
             676
                  10250
                          9912.601754 0
                   9790 10141.748493 1
              215
                          4775.235521 2
         2
              146
                   5500
            1319
                   9900
                          9870.926966 3
            1041
                   8900
                          9630.417885 4
                   9500
                          8697.092014 5
            1425
             409
                  10450 10265.822884 6
         7
             617
                   9790 10293.851867 7
            1526
                   9300
                          8614.349738 8
            1010
                   4600
                          5749.673567 9
In [24]: 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()
```

ridge

Out[24]: []

