```
In [1]: import numpy as np
         import pandas as pd
         import seaborn as sns
         import matplotlib.pyplot as plt
         from sklearn import preprocessing,svm
         from sklearn.model_selection import train_test_split
         from sklearn.linear_model import LinearRegression
In [2]: | df=pd.read_csv(r"C:\Users\91720\Downloads\fiat500_VehicleSelection_Dataset (1).csv")
Out[2]:
                  ID model engine_power age_in_days
                                                         km previous_owners
                                                                                    lat
                                                                                             Ion price
                   1 lounge
                                                                           1 44.907242
                                                                                         8.611560
             0
                                      51
                                                 882
                                                       25000
                                                                                                  8900
                   2
                                      51
                                                       32500
                                                                           1 45.666359 12.241890
                                                                                                  8800
             1
                        pop
                                                 1186
             2
                                      74
                                                      142228
                                                                           1 45.503300 11.417840
                   3
                                                 4658
                                                                                                  4200
                       sport
             3
                   4
                     lounge
                                      51
                                                 2739
                                                      160000
                                                                             40.633171 17.634609
                                                                                                  6000
                   5
                                      73
                                                 3074
                                                      106880
                                                                           1 41.903221 12.495650
                                                                                                  5700
                        pop
          1533 1534
                                                3712
                                      51
                                                      115280
                                                                           1 45.069679
                                                                                         7.704920
                                                                                                  5200
                       sport
          1534 1535
                     lounge
                                      74
                                                 3835
                                                      112000
                                                                           1 45.845692
                                                                                         8.666870
                                                                                                  4600
          1535 1536
                                      51
                                                 2223
                                                       60457
                                                                           1 45.481541
                                                                                         9.413480
                                                                                                  7500
                        pop
          1536 1537 lounge
                                      51
                                                 2557
                                                       80750
                                                                           1 45.000702
                                                                                         7.682270
                                                                                                  5990
                                                                           1 40.323410 17.568270 7900
          1537 1538
                                      51
                                                 1766
                                                       54276
                        pop
         1538 rows × 9 columns
In [3]: |df=df[['lon','price']]
         df.columns=['l','Pri']
In [4]: | df.head(10)
Out[4]:
                         Pri
                    ı
              8.611560
                        8900
          0
            12.241890
                        8800
            11.417840
                        4200
             17.634609
                        6000
            12.495650
                        5700
              7.682270
                        7900
              8.611560
                       10750
            12.495650
                        9190
            11.549470
                        5600
            10.991700
                        6000
In [5]:
         df.describe()
Out[5]:
                          ı
                                     Pri
          count 1538.000000
                             1538.000000
                   11.563428
          mean
                             8576.003901
            std
                   2.328190
                             1939.958641
```

7.245400

9.505090

11.869260

min 25%

50%

75%

max

2500.000000

7122.500000

9000.000000

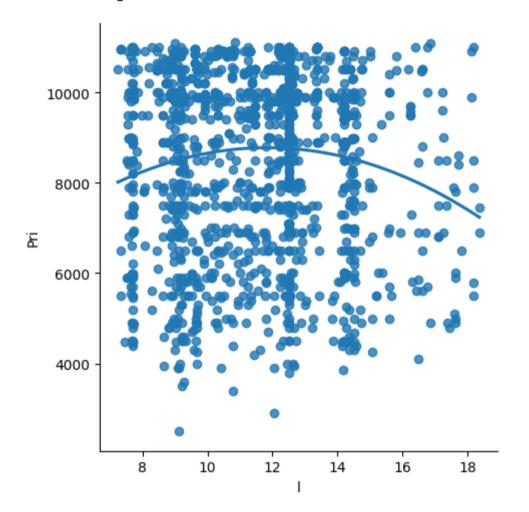
12.769040 10000.00000018.365520 11100.000000

```
In [6]: df.info()
```

dtypes: float64(1), int64(1)
memory usage: 24.2 KB

```
In [7]: sns.lmplot(x="l",y="Pri",data=df,order=2,ci=None)
```

Out[7]: <seaborn.axisgrid.FacetGrid at 0x205dc53fe50>



```
In [15]: x=np.array(df['l']).reshape(-1,1)
y=np.array(df['Pri']).reshape(-1,1)
df.dropna()
```

Out[15]:

```
        I
        Pri

        0
        8.611560
        8900

        1
        12.241890
        8800

        2
        11.417840
        4200

        3
        17.634609
        6000

        4
        12.495650
        5700

        ...
        ...
        ...

        1533
        7.704920
        5200

        1534
        8.666870
        4600

        1535
        9.413480
        7500

        1536
        7.682270
        5990

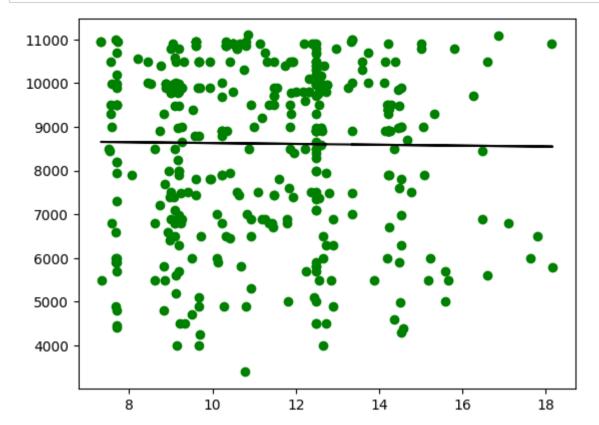
        1537
        17.568270
        7900
```

1538 rows × 2 columns

```
In [9]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
    regr=LinearRegression()
    regr.fit(x_train,y_train)
    print(regr.score(x_test,y_test))
```

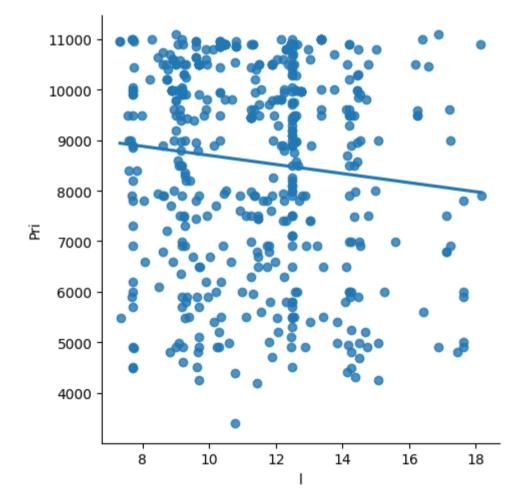
^{-0.006644220228226416}

```
In [10]: y_pred=regr.predict(x_test)
plt.scatter(x_test,y_test,color='g')
plt.plot(x_test,y_pred,color='k')
plt.show()
```



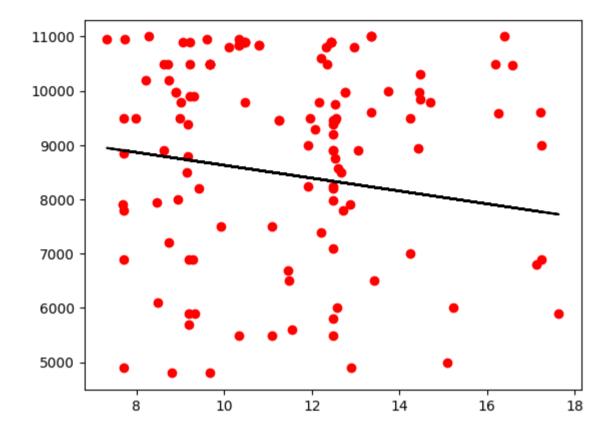
```
In [11]: df500=df[:][:500]
sns.lmplot(x="1",y="Pri",data=df500,order=1,ci=None)
```

Out[11]: <seaborn.axisgrid.FacetGrid at 0x205dc559d50>



```
In [12]: df500.fillna(method='ffill',inplace=True)
    x=np.array(df500['l']).reshape(-1,1)
    y=np.array(df500['Pri']).reshape(-1,1)
    df500.dropna(inplace=True)
    X_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
    regr=LinearRegression()
    regr.fit(X_train,y_train)
    print("Regression:",regr.score(x_test,y_test))
    y_pred=regr.predict(x_test)
    plt.scatter(x_test,y_test,color='r')
    plt.plot(x_test,y_pred,color='k')
    plt.show()
```

Regression: -0.07668022718598899



R2_score: -0.07668022718598899

concluse

Dataset we have taken is poor for linear model but with the smaller data works well with linear mode

1.