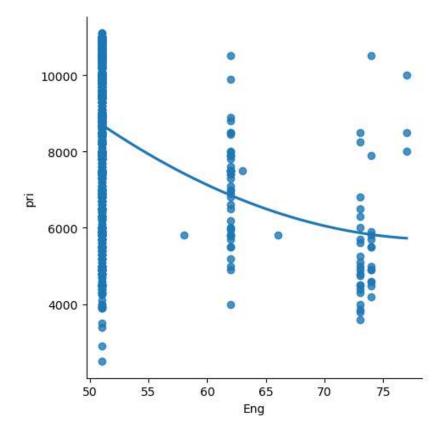
# **ELASTIC NET REGRESSION**

```
In [1]: import pandas as pd
         import numpy as np
         import seaborn as sns
         import matplotlib.pyplot as plt
         from sklearn.model selection import train test split
         from sklearn.linear_model import LinearRegression
         from sklearn.linear_model import Ridge, RidgeCV, Lasso
         from sklearn.preprocessing import StandardScaler
In [2]: data=pd.read_csv(r"C:\Users\Dell\Downloads\fiat500_VehicleSelection_Dataset.csv")
         data
Out[2]:
                 ID model engine_power age_in_days
                                                         km previous_owners
                                                                                   lat
                                                                                             Ion price
                                                      25000
             0
                  1
                     lounge
                                      51
                                                 882
                                                                             44.907242
                                                                                        8.611560
                                                                                                 8900
             1
                   2
                                      51
                                                1186
                                                      32500
                                                                             45.666359
                                                                                       12.241890
                                                                                                 8800
                       pop
             2
                   3
                      sport
                                      74
                                                4658
                                                     142228
                                                                             45.503300
                                                                                       11.417840
                                                                                                 4200
                                                     160000
             3
                                      51
                                                                             40.633171
                                                                                       17.634609
                                                                                                 6000
                   4
                     lounge
                                                2739
                   5
                                                     106880
             4
                       pop
                                      73
                                                3074
                                                                             41.903221
                                                                                       12.495650
                                                                                                 5700
            ...
                                      ...
                                                                             45.069679
                                                                                        7.704920
                                                                                                 5200
          1533 1534
                                      51
                                                3712 115280
                      sport
          1534 1535 lounge
                                                3835 112000
                                                                                        8.666870
                                      74
                                                                            45.845692
                                                                                                 4600
                                      51
                                                2223
                                                      60457
                                                                             45.481541
                                                                                        9.413480
                                                                                                 7500
          1535 1536
          1536 1537
                                      51
                                                2557
                                                      80750
                                                                             45.000702
                                                                                        7.682270
                                                                                                 5990
                     lounge
          1537 1538
                                      51
                                                1766
                                                      54276
                                                                            40.323410 17.568270
                                                                                                 7900
                       pop
         1538 rows × 9 columns
In [3]: data = data[['engine_power','price']]
         data.columns=['Eng','pri']
In [4]: data.head()
Out[4]:
                   pri
             Eng
          0
              51
                 8900
          1
              51
                 8800
          2
              74 4200
          3
              51
                 6000
```

73 5700

```
In [5]: sns.lmplot(x='Eng',y='pri',data=data,order=2,ci=None)
```

Out[5]: <seaborn.axisgrid.FacetGrid at 0x1ef51dbe0b0>



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

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1538 entries, 0 to 1537
Data columns (total 2 columns):
# Column Non-Null Count Dtype
--- 0 Eng 1538 non-null int64
1 pri 1538 non-null int64
dtypes: int64(2)
memory usage: 24.2 KB

In [7]: data.describe()

### Out[7]:

	Eng	pri
count	1538.000000	1538.000000
mean	51.904421	8576.003901
std	3.988023	1939.958641
min	51.000000	2500.000000
25%	51.000000	7122.500000
50%	51.000000	9000.000000
75%	51.000000	10000.000000
max	77.000000	11100.000000

```
In [8]: data.fillna(method='ffill')
```

Out[8]:

```
Eng
            pri
   0
       51
          8900
   1
       51 8800
       74
          4200
   3
          6000
       51
       73 5700
  ...
1533
       51 5200
       74 4600
1534
1535
       51 7500
1536
       51 5990
1537
       51 7900
```

1538 rows × 2 columns

```
In [9]: x=np.array(data['Eng']).reshape(-1,1)
y=np.array(data['pri']).reshape(-1,1)
```

```
In [10]: data.dropna(inplace=True)
```

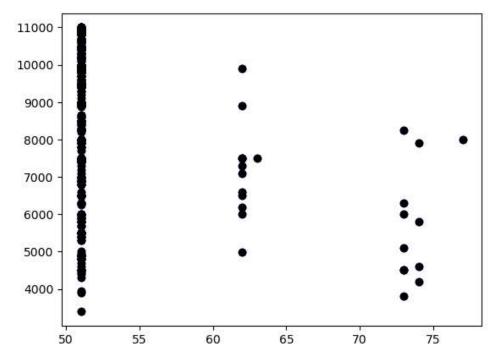
C:\Users\Dell\AppData\Local\Temp\ipykernel\_6612\1368182302.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/ind exing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy) data.dropna(inplace=True)

```
In [11]: X_train,X_test,y_train,y_test = train_test_split(x, y, test_size = 0.25)
# Splitting the data into training data and test data
regr = LinearRegression()
regr.fit(X_train, y_train)
print(regr.score(X_test, y_test))
```

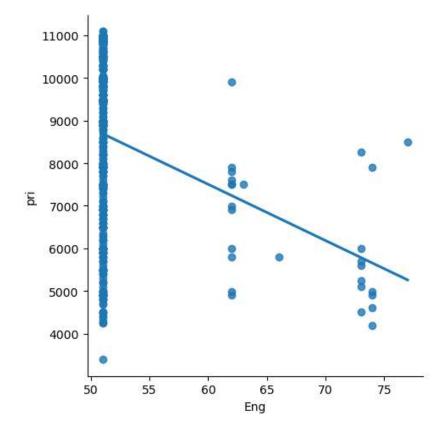
0.08037534242081401

```
In [12]: y_pred = regr.predict(X_test)
    plt.scatter(X_test, y_test, color = 'b')
    plt.scatter(X_test, y_test, color = 'k')
    plt.show()
```



```
In [13]: df500 = data[:][:500]
# Selecting the 1st 500 rows of teh data
sns.lmplot(x = "Eng", y = "pri", data = df500, order = 1, ci = None)
```

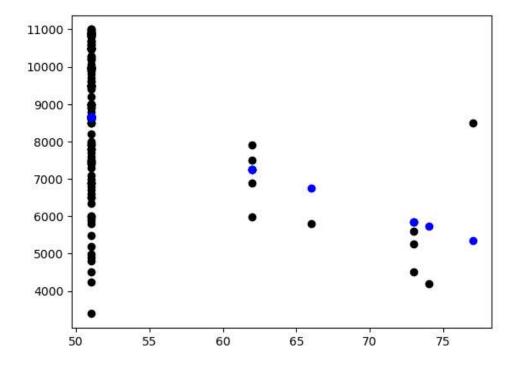
Out[13]: <seaborn.axisgrid.FacetGrid at 0x1ef587f99c0>



```
In [14]:

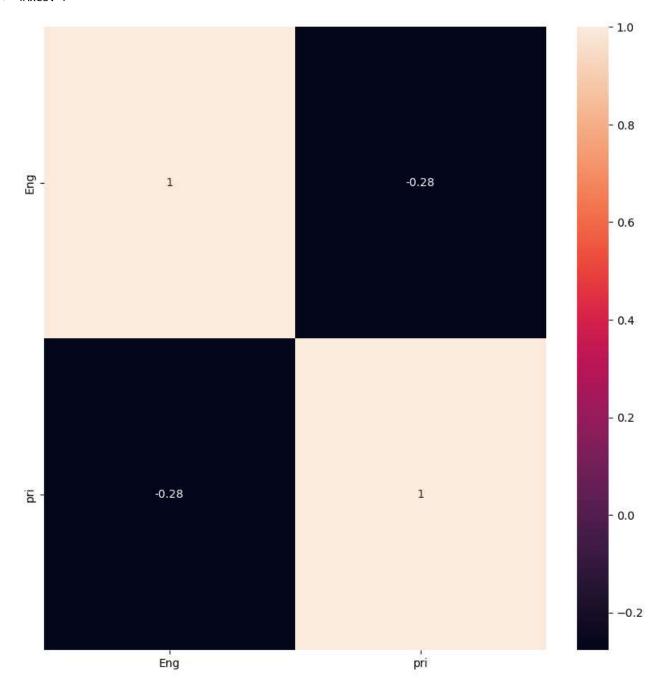
df500.fillna(method = 'ffill', inplace = True)
x = np.array(df500['Eng']).reshape(-1, 1)
y = np.array(df500['pri']).reshape(-1, 1)
df500.dropna(inplace = True)
X_train, X_test, y_train, y_test = train_test_split(x, y, test_size = 0.25)
regr = LinearRegression()
regr.fit(X_train, y_train)
print("Regression:",regr.score(X_test,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

Regression: 0.12622221859731864



```
In [15]: plt.figure(figsize = (10, 10))
sns.heatmap(data.corr(), annot = True)
```

Out[15]: <Axes: >



```
In [16]: from sklearn.linear_model import LinearRegression
    from sklearn.metrics import r2_score
    #Train the model
    model = LinearRegression()
    model.fit(X_train, y_train)
    #Evaluating the model on the test set
    y_pred = model.predict(X_test)
    r2 = r2_score(y_test, y_pred)
    print("R2 score:",r2)
```

R2 score: 0.12622221859731864

#### Linear Regression Model:

The train score for lr model is 0.05153214007654372 The test score for lr model is 0.12622221859731864

```
In [18]: #Ridge Regression Model
    ridgeReg = Ridge(alpha=10)
    ridgeReg.fit(X_train,y_train)
    #train and test scorefor ridge regression
    train_score_ridge = ridgeReg.score(X_train, y_train)
    test_score_ridge = ridgeReg.score(X_test, y_test)
    print("\nRidge Model:\n")
    print("The train score for ridge model is {}".format(train_score_ridge))
    print("The test score for ridge model is {}".format(test_score_ridge))
```

#### Ridge Model:

The train score for ridge model is 0.05153193102598341 The test score for ridge model is 0.1261842603069645

```
In [19]: #Lasso regression model
print("\nLasso Model: \n")
lasso = Lasso(alpha = 10)
lasso.fit(X_train,y_train)
train_score_ls =lasso.score(X_train,y_train)
test_score_ls =lasso.score(X_test,y_test)
print("The train score for ls model is {}".format(train_score_ls))
print("The test score for ls model is {}".format(test_score_ls))
```

### Lasso Model:

The train score for ls model is 0.051530322720581334 The test score for ls model is 0.12610778604113482

```
In [21]: #Using the Linear CV model
from sklearn.linear_model import LassoCV
#Lasso Cross validation
lasso_cv = LassoCV(alphas = [0.0001, 0.001, 0.01, 1, 10], random_state=0).fit(X_train, y_train)
#score
print(lasso_cv.score(X_train, y_train))
print(lasso_cv.score(X_test, y_test))
```

0.051530322720581334
0.12610778604113482

C:\ProgramData\anaconda3\lib\site-packages\sklearn\linear\_model\\_coordinate\_descent.py:1568: Data
ConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the s
hape of y to (n\_samples, ), for example using ravel().
y = column\_or\_1d(y, warn=True)

# **Elestic net regression**