```
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 data=pd.read_csv(r"C:\Users\anu\Downloads\fiat500_VehicleSelection_Dataset.csv") data

Out[2]:

	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
1	2	рор	51	1186	32500	1	45.666359	12.241890	8800
2	3	sport	74	4658	142228	1	45.503300	11.417840	4200
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
4	5	рор	73	3074	106880	1	41.903221	12.495650	5700
1533	1534	sport	51	3712	115280	1	45.069679	7.704920	5200
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1535	1536	рор	51	2223	60457	1	45.481541	9.413480	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990
1537	1538	pop	51	1766	54276	1	40.323410	17.568270	7900

1538 rows × 9 columns

```
In [6]: data = data[['engine_power','price']]
data.columns=['Eng','pri']
```

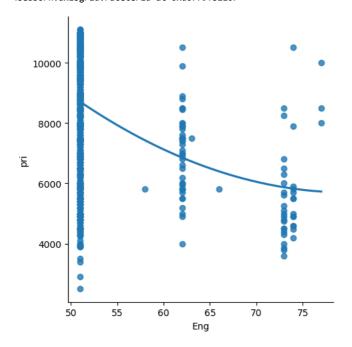
In [7]: data.head()

Out[7]:

	Eng	pri
0	51	8900
1	51	8800
2	74	4200
3	51	6000
4	73	5700

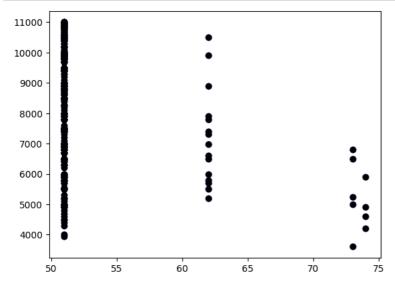
In [8]: sns.lmplot(x='Eng',y='pri',data=data,order=2,ci=None)

Out[8]: <seaborn.axisgrid.FacetGrid at 0xd0779fb210>



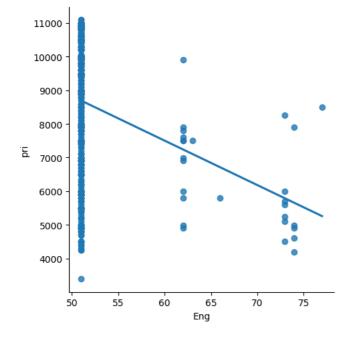
```
In [9]: data.tail()
 Out[9]:
                Eng
                      pri
                 51
                    5200
          1533
          1534
                 74 4600
           1535
                 51 7500
          1536
                 51 5990
           1537
In [10]: data.info()
          <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 [11]: data.describe()
Out[11]:
                       Eng
                                    pri
                1538.000000
                             1538.000000
          count
           mean
                   51 904421
                             8576.003901
             std
                   3.988023
                             1939.958641
            min
                   51.000000
                             2500.000000
                            7122.500000
           25%
                  51.000000
            50%
                  51.000000
                             9000.000000
           75%
                  51.000000 10000.000000
                  77.000000 11100.000000
           max
In [12]: data.fillna(method='ffill')
Out[12]:
                Eng
                      pri
                 51
                    8900
             0
                 51
                    8800
             2
                 74 4200
                 73 5700
           1533
                 51 5200
           1534
                 74 4600
          1535
                 51 7500
           1536
                 51 5990
          1537
                 51 7900
          1538 rows × 2 columns
In [13]: x=np.array(data['Eng']).reshape(-1,1)
         y=np.array(data['pri']).reshape(-1,1)
In [14]: data.dropna(inplace=True)
          C:\Users\anu\AppData\Local\Temp\ipykernel_10712\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/indexing.html#returning-a-view-v
          ersus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)
           data.dropna(inplace=True)
In [15]: 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))
          0.09139659405015021
```

```
In [18]: y_pred = regr.predict(X_test)
    plt.scatter(X_test, y_test, color = 'b')
    plt.scatter(X_test, y_test, color = 'k')
    plt.show()
```



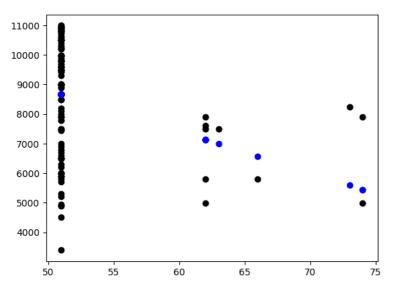
```
In [19]: df500 = data[:][:500]
# Selecting the 1st 500 rows of teh data
sns.lmplot(x = "Eng", y = "pri", data = df500, order = 1, ci = None)
```

Out[19]: <seaborn.axisgrid.FacetGrid at 0xd079a6e690>



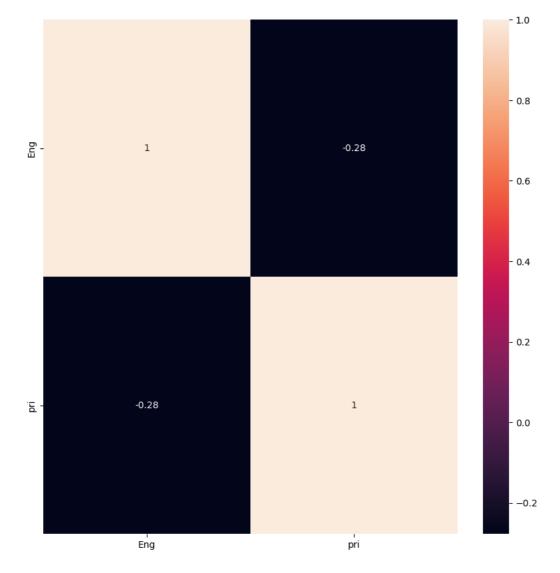
```
In [20]: df500.fillna(method = 'ffill', inplace = True)
    x = np.array(df500['Eng']).reshape(-1, 1)
    y = np.array(df500['pri']).reshape(-1, 1)
    df500.dropna(inplace = True)
    X_train, X_test, y_train, y_test = train_test_split(x, y, test_size = 0.25)
    regr = LinearRegression()
    regr.fit(X_train, 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: 0.05889693200959778



```
In [21]: plt.figure(figsize = (10, 10))
sns.heatmap(data.corr(), annot = True)
```

Out[21]: <Axes: >



```
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: 0.05889693200959778
In [23]: |#Model
                  lr = LinearRegression()
                   #Fit model
                  lr.fit(X_train, y_train)
                   #predict
                   #prediction = lr.predict(X_test)
                   #actual
                   actual = y_test
                   train_score_lr = lr.score(X_train, y_train)
                   test_score_lr = lr.score(X_test, y_test)
                   print("\nLinear Regression Model:\n")
                   print("The train score for lr model is {}".format(train_score_lr))
                  print("The test score for lr model is {}".format(test_score_lr))
                   Linear Regression Model:
                   The train score for lr model is 0.07173702583107355
                   The test score for 1r model is 0.05889693200959778
In [24]: |#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 0.07173681092195672
                  The test score for ridge model is 0.05896933621054745
In [25]: #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 1s model is 0.0717354853716482
                   The test score for 1s model is 0.05908941499687659
In [26]: #Using the linear CV model
                   from sklearn.linear_model import RidgeCV
                   #Ridge Cross validation
                   \label{eq:ridge_cv} {\tt ridge_cv} = {\tt RidgeCV(alphas} = [0.0001, \, 0.001, \, 0.01, \, \, 0.1, \, \, 1, \, \, 10]). \\ {\tt fit(X\_train, \, y\_train)} \\ {\tt ridge\_cv} = {\tt RidgeCV(alphas} = [0.0001, \, 0.001, \, 0.01, \, \, 0.1, \, \, 1, \, \, 10]). \\ {\tt fit(X\_train, \, y\_train)} \\ {\tt ridge\_cv} = {\tt RidgeCV(alphas} = [0.0001, \, 0.001, \, 0.01, \, \, 0.1, \, \, 1, \, \, 10]). \\ {\tt fit(X\_train, \, y\_train)} \\ {\tt ridge\_cv} = {\tt RidgeCV(alphas} = [0.0001, \, 0.001, \, 0.01, \, \, 0.1, \, \, 1, \, \, 10]). \\ {\tt fit(X\_train, \, y\_train)} \\ {\tt ridge\_cv} = {\tt RidgeCV(alphas} = [0.0001, \, 0.001, \, 0.01, \, \, 0.1, \, \, 1, \, \, 10]). \\ {\tt fit(X\_train, \, y\_train)} \\ {\tt ridge\_cv} = {\tt RidgeCV(alphas, \, y\_train)} \\ {\tt ridge\_cv} = {\tt RidgeCV(alphas
                   #score
                  print("The train score for ridge model is {}".format(ridge_cv.score(X_train, y_train)))
                  print("The train score for ridge model is {}".format(ridge_cv.score(X_test, y_test)))
                   The train score for ridge model is 0.07173681092195705
                   The train score for ridge model is 0.05896933621050371
                   ELASTICNET REGRESSION
In [27]: from sklearn.linear_model import ElasticNet
                  regr=ElasticNet()
                   regr.fit(x,y)
                   print(regr.coef_)
                   print(regr.intercept_)
                   [-128.05913739]
                   [15219.18170389]
```

In [28]: y_predict_elastic = regr.predict(X_train)

In [22]: from sklearn.linear_model import LinearRegression

In [29]: mean_squared_error=np.mean((y_predict_elastic-y_train)**2)
print("mean_squared_error on test_set",mean_squared_error)

mean squared error on test set 4473533.8762751445

In []: