```
In [1]:
          import numpy as np
          import pandas as pd
          import matplotlib.pyplot as py
          import seaborn as sns
In [2]:
          d=pd.read csv(r"C:\Users\user\Downloads\19_nuclear_explosions - 19_nuclear_explosions.c
                WEAPON
                             WEAPON
Out[2]:
                 SOURCE
                         DEPLOYMENT
                                       Data.Source Location.Cordinates.Latitude Location.Cordinates.Longitude
               COUNTRY
                            LOCATION
            0
                    USA
                           Alamogordo
                                              DOE
                                                                        32.54
                                                                                                  -105.57
                    USA
            1
                             Hiroshima
                                              DOE
                                                                        34.23
                                                                                                   132.27
            2
                    USA
                              Nagasaki
                                              DOE
                                                                        32.45
                                                                                                   129.52
            3
                    USA
                                 Bikini
                                                                                                   165.20
                                              DOE
                                                                        11.35
                    USA
                                              DOE
                                                                                                   165.20
            4
                                 Bikini
                                                                        11.35
            •••
                                                ...
         2041
                  CHINA
                               Lop Nor
                                              HFS
                                                                        41.69
                                                                                                    88.35
         2042
                   INDIA
                               Pokhran
                                              HFS
                                                                        27.07
                                                                                                    71.70
         2043
                   INDIA
                               Pokhran
                                              NRD
                                                                        27.07
                                                                                                    71.70
         2044
                  PAKIST
                                Chagai
                                              HFS
                                                                        28.90
                                                                                                    64.89
         2045
                  PAKIST
                                Kharan
                                              HFS
                                                                        28.49
                                                                                                    63.78
        2046 rows × 16 columns
In [3]:
          d.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 2046 entries, 0 to 2045
         Data columns (total 16 columns):
          #
              Column
                                                Non-Null Count
                                                                 Dtype
          0
              WEAPON SOURCE COUNTRY
                                                2046 non-null
                                                                 object
          1
              WEAPON DEPLOYMENT LOCATION
                                                2046 non-null
                                                                 object
          2
              Data.Source
                                                2046 non-null
                                                                 object
                                                                 float64
          3
              Location.Cordinates.Latitude
                                                2046 non-null
          4
                                                                 float64
              Location.Cordinates.Longitude
                                                2046 non-null
          5
              Data.Magnitude.Body
                                                2046 non-null
                                                                 float64
              Data.Magnitude.Surface
                                                                 float64
          6
                                                2046 non-null
                                                                 float64
          7
              Location.Cordinates.Depth
                                                2046 non-null
                                                                 float64
          8
              Data.Yeild.Lower
                                                2046 non-null
          9
              Data.Yeild.Upper
                                                2046 non-null
                                                                 float64
          10
              Data.Purpose
                                                2046 non-null
                                                                 object
          11
              Data.Name
                                                2046 non-null
                                                                 object
          12
                                                                 object
              Data.Type
                                                2046 non-null
          13
              Date.Day
                                                2046 non-null
                                                                 int64
```

2046 non-null

int64

Date.Month

15 Date.Year 2046 non-null int64 dtypes: float64(7), int64(3), object(6) memory usage: 255.9+ KB In [4]: d.columns Out[4]: Index(['WEAPON SOURCE COUNTRY', 'WEAPON DEPLOYMENT LOCATION', 'Data.Source', 'Location.Cordinates.Latitude', 'Location.Cordinates.Longitude', 'Data.Magnitude.Body', 'Data.Magnitude.Surface',
'Location.Cordinates.Depth', 'Data.Yeild.Lower', 'Data.Yeild.Upper',
'Data.Purpose', 'Data.Name', 'Data.Type', 'Date.Day', 'Date.Month', 'Date.Year'], dtype='object') In [5]: d1=d.head(100) d1 Out[5]: **WEAPON WEAPON** SOURCE DEPLOYMENT Data.Source Location.Cordinates.Latitude Location.Cordinates.Longitude I **COUNTRY** LOCATION 0 USA Alamogordo DOE 32.54 -105.57 1 USA Hiroshima DOE 34.23 132.27 2 USA DOE 32.45 129.52 Nagasaki 3 USA Bikini DOE 11.35 165.20 4 USA Bikini DOE 11.35 165.20 95 USSR Semi Kazakh DOE 50.00 78.00 96 USA Nts DOE 37.00 -116.00

46.00

50.00

50.00

float64

61.00

78.00

78.00

100 rows × 16 columns

USSR

USSR

USSR

In [6]:

d1.info()

97

98

99

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99

Kazakh

Semi Kazakh

Semi Kazakh

MTM

DOE

DOE

Data columns (total 16 columns): Non-Null Count Dtype # Column ---0 WEAPON SOURCE COUNTRY 100 non-null object 1 WEAPON DEPLOYMENT LOCATION 100 non-null object 2 Data.Source 100 non-null object 3 Location.Cordinates.Latitude 100 non-null float64 4 Location.Cordinates.Longitude 100 non-null float64 5 Data.Magnitude.Body 100 non-null float64 6 Data.Magnitude.Surface 100 non-null float64

Location.Cordinates.Depth 100 non-null

```
Data.Yeild.Lower
                                              100 non-null
                                                               float64
          8
          9
                                              100 non-null
                                                               float64
              Data.Yeild.Upper
          10 Data.Purpose
                                                               object
                                              100 non-null
          11 Data.Name
                                              100 non-null
                                                               object
          12 Data. Type
                                              100 non-null
                                                               object
                                              100 non-null
                                                               int64
          13 Date.Day
          14 Date.Month
                                              100 non-null
                                                               int64
          15 Date.Year
                                              100 non-null
                                                               int64
         dtypes: float64(7), int64(3), object(6)
         memory usage: 12.6+ KB
 In [7]:
                  'Location.Cordinates.Latitude', 'Location.Cordinates.Longitude',
          x=d1[[
                  'Data.Magnitude.Body', 'Data.Magnitude.Surface',
                  'Location.Cordinates.Depth', 'Data.Yeild.Lower', 'Data.Yeild.Upper', 'Date.Day',
                  'Date.Year']]
          y=d1['Location.Cordinates.Longitude']
 In [8]:
          from sklearn.model_selection import train_test_split
          x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
 In [9]:
          from sklearn.linear model import LinearRegression
In [10]:
          lr=LinearRegression()
          lr.fit(x train,y train)
Out[10]: LinearRegression()
In [13]:
          prediction =lr.predict(x_test)
          py.scatter(y_test,prediction)
Out[13]: <matplotlib.collections.PathCollection at 0x22d1dffb100>
           150
           100
            50
             0
           -50
          -100
                          -50
                                          50
                                                 100
                 -100
                                   0
                                                         150
In [14]:
          print(lr.score(x_test,y_test))
         1.0
In [15]:
          print(lr.score(x_train,y_train))
```

```
1.0
In [16]:
          from sklearn.linear_model import Ridge,Lasso
In [17]:
          rr=Ridge(alpha=10)
          rr.fit(x_train,y_train)
Out[17]: Ridge(alpha=10)
In [18]:
          rr.score(x_test,y_test)
         0.999999998608537
Out[18]:
In [19]:
          la=Lasso(alpha=10)
          la.fit(x_train,y_train)
Out[19]: Lasso(alpha=10)
In [20]:
          la.score(x test,y test)
         0.9999993187710697
Out[20]:
In [21]:
          from sklearn.linear model import ElasticNet
          en=ElasticNet()
          en.fit(x_train,y_train)
Out[21]: ElasticNet()
In [22]:
          print(en.coef_)
         [ 0.00000000e+00 9.99926983e-01 0.00000000e+00 0.00000000e+00
           0.00000000e+00 1.49611139e-06 -1.80911530e-07 -0.00000000e+00
           0.00000000e+00 -0.00000000e+00]
In [23]:
          print(en.intercept_)
          -0.0019103221962488703
In [24]:
          print(en.predict(x_test))
         [-115.99343908 -115.99344013
                                       129.50866018 -115.99341146 165.18605495
           162.13654595
                          77.99240083
                                        77.99272318 -115.99343119
                                                                    77.99240753
           162.13847272
                          77.99247592 165.20575533 -115.99343908
                                                                   132.33843645
            60.99363606
                          77.99243121 -115.99342593 -115.99343908 -115.99338384
          -115.99343513 -115.99342987 -115.99344039 162.13629869
                                                                  165.19510221
            54.99412637 -115.99341146
                                       77.99239964
                                                     77.99240201 165.18605495]
In [25]:
          print(en.score(x_test,y_test))
```

0.9999999947312163

```
In [26]:
          from sklearn import metrics
In [27]:
          print("Mean Absolute Error:", metrics.mean_absolute_error(y_test, prediction))
         Mean Absolute Error: 2.9250675955457456e-13
In [28]:
          print("Mean Squared Error:",metrics.mean_squared_error(y_test,prediction))
         Mean Squared Error: 1.4304846038334128e-25
In [29]:
          print("Root Mean Squared Error:",np.sqrt(metrics.mean_squared_error(y_test,prediction))
         Root Mean Squared Error: 3.782174776280722e-13
In [30]:
          import pickle
In [32]:
          filename="nuclear"
          pickle.dump(lr,open(filename,'wb'))
 In [ ]:
```