```
In [ ]:
```

```
#!/usr/bin/env python
# coding: utf-8
```

COVID - 19 and X Dataset

Assigned dataset - 20 https://github.com/michaelofsbu/CSE-544-Datasets

For the X dataset, we have chosen the Chicago Crime Data

https://data.cityofchicago.org/Public-Safety/Crimes-2001-to-present/ijzp-q8t2

In[6]:

```
In [ ]:
```

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import datetime as dt
import re
import scipy.stats as stats
from scipy.stats import gamma
```

Mandatory tasks

Q 1

Load Data and get statistics

In[7]:

```
In [ ]:
```

```
covid_data = pd.read_csv('20.csv')
print(covid_data.describe())
```

In[10]:

```
In [ ]:
```

```
covid_PA_confirmed = covid_data['PA confirmed']
covid_RI_confirmed = covid_data['RI confirmed']
covid_PA_deaths = covid_data['PA deaths']
covid_RI_deaths = covid_data['RI deaths']
```

Check null values in data set

In[11]:

```
In [ ]:
```

```
print(covid_PA_confirmed.isnull().sum())
print(covid_RI_confirmed.isnull().sum())
print(covid_PA_deaths.isnull().sum())
print(covid_RI_deaths.isnull().sum())
```

No null values observed

getting individual data from cumulative data and assign zero for negative confirmed cases and deaths

In[12]:

```
In [ ]:
```

```
covid_PA_confirmed = covid_PA_confirmed.diff()
covid_PA_confirmed.fillna(0, inplace=True)
covid_PA_confirmed.loc[covid_PA_confirmed<0]=0</pre>
```

In []:

```
covid_RI_confirmed = covid_RI_confirmed.diff()
covid_RI_confirmed.fillna(0, inplace=True)
covid_RI_confirmed.loc[covid_RI_confirmed<0]=0</pre>
```

In []:

```
covid_PA_deaths = covid_PA_deaths.diff().reset_index(drop=True)
covid_PA_deaths.fillna(0, inplace=True)
covid_PA_deaths.loc[covid_PA_deaths<0]=0</pre>
```

In []:

```
covid_RI_deaths = covid_RI_deaths.diff().reset_index(drop=True)
covid_RI_deaths.fillna(0, inplace=True)
covid_RI_deaths.loc[covid_RI_deaths<0]=0</pre>
```

In[13]:

In []:

```
covid_PA_confirmed = pd.concat([covid_data['Date'], covid_PA_confirmed], axis=1)
covid_RI_confirmed = pd.concat([covid_data['Date'], covid_RI_confirmed], axis=1)
covid_PA_deaths = pd.concat([covid_data['Date'], covid_PA_deaths], axis=1)
covid_RI_deaths = pd.concat([covid_data['Date'], covid_RI_deaths], axis=1)
covid_PA_confirmed.columns=['Date', 'Count']
covid_PA_deaths.columns=['Date', 'Count']
covid_RI_confirmed.columns=['Date', 'Count']
covid_RI_deaths.columns=['Date', 'Count']
```

Outlier detection

In[18]:

Tukey's rule to check for outliers for daily Confirmed and death cases

alpha is taken as 1.5

```
In [ ]:
```

```
def outlier_detection(df):
    n = df.size
    df = df.sort_values(ascending=True)
    q1 = df[int(np.ceil(0.25*n))]
    q3 = df[int(np.ceil(0.75*n))]
    iqr = q3 - q1

alpha = 1.5
    upper_limit = q3 + 1.5*iqr
    lower_limit = q1 - 1.5*iqr
    print('upper and lower limits for outliers: ',upper_limit,lower_limit)
```

```
return df[((df < lower_limit) | (df > upper_limit))]
In[19]:
In [ ]:
print(outlier_detection(covid_PA_confirmed['Count']))
print(outlier_detection(covid_RI_confirmed['Count']))
print(outlier detection(covid PA deaths['Count']))
print(outlier detection(covid RI deaths['Count']))
No outliers found
----Q2----
Q2 - iii, iv
In[71]:
In [ ]:
def plot ewma(X, test, predictions):
    plt.plot(X, test, label="original")
    plt.plot(X,predictions,label="Predictions")
    plt.xlabel('Dates - 4th week of August')
    plt.ylabel('Y')
    plt.legend(loc='upper left')
    plt.xticks(rotation=45)
    plt.show()
class EWMA:
    def init (self, alpha):
        self.alpha = alpha
    def predict(self, data, test):
        y t hat = data['Count'][0]
        #calculating y t hat
        for i in range(data.shape[0]):
            y t = data['Count'][i]
            y_t_hat = self.alpha*y_t + (1-self.alpha)*y_t_hat
        mse errors = []
        mape_errors = []
        predictions = np.zeros(len(test))
        #predicting values
        for i in np.arange(len(test)):
            y t = test['Count'][i]
            print("Date: " + str(test['Date'][i]) + " - Test Prediction: " + "{:5.3f}".f
ormat(y t hat) + ", Actual: " + "{:5.3f}".format(y t))
            #Ignoring zero data to avoid infinite error
            if y t!=0:
```

In[72]:

Exponentially Weighted Moving Average (Confirmed COVID Cases in August 2020) redict Confirmed cases for 4th week of August with first three weeks data of August

mape_errors.append(abs(y_t_hat-y_t)*100/y_t)

plot_ewma(np.array(test['Date']), test['Count'], predictions)

print("--------Errors-----print("MAPE:" + "{:5.2f}".format(np.mean(mape_errors)))
print("MSE:" + "{:5.2f}".format(np.mean(mse_errors)))

predictions[i] = y t hat = self.alpha*y t + (1-self.alpha)* y t hat

mse_errors.append((y_t_hat-y_t)**2)

```
train_PA_confirmed = covid_PA_confirmed[(covid PA confirmed['Date']>='2020-08-01') & (cov
id_PA_confirmed['Date'] <= '2020-08-21')].reset_index(drop=True)</pre>
test PA confirmed = covid PA confirmed[(covid PA confirmed['Date']>='2020-08-22') & (covi
d PA confirmed['Date'] <= '2020-08-28')].reset index(drop=True)</pre>
In [ ]:
train RI confirmed = covid RI confirmed[(covid RI confirmed['Date']>='2020-08-01') & (cov
id_RI_confirmed['Date'] <= '2020-08-21')].reset_index(drop=True)</pre>
test RI confirmed = covid RI confirmed[(covid RI confirmed['Date']>='2020-08-22') & (covi
d RI confirmed['Date'] <= '2020-08-28')].reset index(drop=True)
In [ ]:
print("\n----\n")
print("\n----- Predicting Pennsylvania confirmed cases ----\n")
ewma = EWMA(0.5)
ewma.predict(train PA confirmed, test PA confirmed)
In [ ]:
print("\n----- Predicting Rhode Island confirmed cases -----\n")
ewma = EWMA(0.5)
ewma.predict(train RI confirmed, test RI confirmed)
In [ ]:
print("\n----\n")
print("\n----- Predicting Pennsylvania confirmed cases -----\n")
ewma = EWMA(0.8)
ewma.predict(train PA confirmed, test PA confirmed)
In [ ]:
print("\n---- Predicting Rhode Island confirmed cases -----\n")
ewma = EWMA(0.8)
ewma.predict(train RI confirmed, test RI confirmed)
------ EWMA(0.5) ------
----- Predicting Pennsylvania confirmed cases ------
Date: 2020-08-22 - Test Prediction: 697.134, Actual: 796.000
Date: 2020-08-23 - Test Prediction: 746.567, Actual: 619.000
Date: 2020-08-24 - Test Prediction: 682.783, Actual: 426.000
Date: 2020-08-25 - Test Prediction: 554.392, Actual: 561.000
Date: 2020-08-26 - Test Prediction: 557.696, Actual: 501.000
Date: 2020-08-27 - Test Prediction: 529.348, Actual: 620.000
Date: 2020-08-28 - Test Prediction: 574.674, Actual: 835.000
-----Errors-----
MAPE:21.66
MSE:24461.59
----- Predicting Rhode Island confirmed cases ------
```

Date: 2020-08-22 - Test Prediction: 77.676. Actual: 0.000

,
Date: 2020-08-23 - Test Prediction: 38.838, Actual: 0.000
Date: 2020-08-24 - Test Prediction: 19.419, Actual: 0.000
Date: 2020-08-25 - Test Prediction: 9.709, Actual: 0.000
Date: 2020-08-26 - Test Prediction: 4.855, Actual: 700.000
Date: 2020-08-27 - Test Prediction: 352.427, Actual: 0.000
Date: 2020-08-28 - Test Prediction: 176.214, Actual: 0.000
Errors
MAPE:99.31 MSE:483226.95
EWMA(0.8)
Predicting Pennsylvania confirmed cases
Date: 2020-08-22 - Test Prediction: 704.619, Actual: 796.000
Date: 2020-08-23 - Test Prediction: 777.724, Actual: 619.000
Date: 2020-08-24 - Test Prediction: 650.745, Actual: 426.000
Date: 2020-08-25 - Test Prediction: 470.949, Actual: 561.000
Date: 2020-08-26 - Test Prediction: 542.990, Actual: 501.000
Date: 2020-08-27 - Test Prediction: 509.398, Actual: 620.000
Date: 2020-08-28 - Test Prediction: 597.880, Actual: 835.000
Errors
MAPE:22.94 MSE:23197.88
Predicting Rhode Island confirmed cases
Date: 2020-08-22 - Test Prediction: 19.712, Actual: 0.000
Date: 2020-08-23 - Test Prediction: 3.942, Actual: 0.000
Date: 2020-08-24 - Test Prediction: 0.788, Actual: 0.000
Date: 2020-08-25 - Test Prediction: 0.158, Actual: 0.000
Date: 2020-08-26 - Test Prediction: 0.032, Actual: 700.000
Date: 2020-08-27 - Test Prediction: 560.006, Actual: 0.000
Date: 2020-08-28 - Test Prediction: 112.001, Actual: 0.000
Errors

```
MAPE:100.00
MSE:489955.85
```

In[73]:

Exponentially Weighted Moving Average (Confirmed COVID Cases in August 2020) redict deaths for 4th week of August with first three weeks data of August

```
In [ ]:
train PA deaths = covid PA deaths[(covid PA deaths['Date']>='2020-08-01') & (covid PA de
aths['Date'] <= '2020-08-21')].reset index(drop=True)
test PA deaths = covid PA deaths[(covid PA deaths['Date']>='2020-08-22') & (covid PA deat
hs['Date']<='2020-08-28')].reset index(drop=True)
In [ ]:
train RI deaths = covid RI deaths[(covid RI deaths['Date']>='2020-08-01') & (covid RI de
aths['Date'] <= '2020-08-21')].reset index(drop=True)
test RI deaths = covid RI deaths[(covid RI deaths['Date']>='2020-08-22') & (covid RI deat
hs['Date']<='2020-08-28')].reset index(drop=True)
In [ ]:
print("\n-----\n")
print("\n-----\n")
ewma = EWMA(0.5)
ewma.predict(train PA deaths, test PA deaths)
ewma = EWMA(0.5)
In [ ]:
print("\n----- Predicting Rhode Island deaths ----\n")
ewma.predict(train RI deaths, test RI deaths)
In [ ]:
print("\n-----\n")
print("\n----- Predicting Pennsylvania deaths -----\n")
ewma = EWMA(0.8)
ewma.predict(train PA deaths, test PA deaths)
In [ ]:
ewma = EWMA(0.8)
print("\n----\n")
ewma.predict(train RI deaths, test RI deaths)
------ EWMA(0.5) ------
----- Predicting Pennsylvania deaths -----
Date: 2020-08-22 - Test Prediction: 19.131, Actual: 18.000
Date: 2020-08-23 - Test Prediction: 18.565, Actual: 2.000
Date: 2020-08-24 - Test Prediction: 10.283, Actual: 1.000
Date: 2020-08-25 - Test Prediction: 5.641, Actual: 26.000
Date: 2020-08-26 - Test Prediction: 15.821, Actual: 19.000
```

Date: 2020-08-27 - Test Prediction: 17.410. Actual: 11.000

Date: 2020-08-28 - Test Prediction: 14.205, Actual: 20.000
Errors
MAPE:277.87 MSE:123.02
Predicting Rhode Island deaths
Date: 2020-08-22 - Test Prediction: 0.888, Actual: 0.000
Date: 2020-08-23 - Test Prediction: 0.444, Actual: 0.000
Date: 2020-08-24 - Test Prediction: 0.222, Actual: 0.000
Date: 2020-08-25 - Test Prediction: 0.111, Actual: 0.000
Date: 2020-08-26 - Test Prediction: 0.055, Actual: 12.000
Date: 2020-08-27 - Test Prediction: 6.028, Actual: 0.000
Date: 2020-08-28 - Test Prediction: 3.014, Actual: 0.000
Errors
MAPE:99.54 MSE:142.67
EWMA(0.8)
Predicting Pennsylvania deaths
Predicting Pennsylvania deaths Date: 2020-08-22 - Test Prediction: 19.369, Actual: 18.000
Date: 2020-08-22 - Test Prediction: 19.369, Actual: 18.000
Date: 2020-08-22 - Test Prediction: 19.369, Actual: 18.000 Date: 2020-08-23 - Test Prediction: 18.274, Actual: 2.000
Date: 2020-08-22 - Test Prediction: 19.369, Actual: 18.000 Date: 2020-08-23 - Test Prediction: 18.274, Actual: 2.000 Date: 2020-08-24 - Test Prediction: 5.255, Actual: 1.000
Date: 2020-08-22 - Test Prediction: 19.369, Actual: 18.000 Date: 2020-08-23 - Test Prediction: 18.274, Actual: 2.000 Date: 2020-08-24 - Test Prediction: 5.255, Actual: 1.000 Date: 2020-08-25 - Test Prediction: 1.851, Actual: 26.000
Date: 2020-08-22 - Test Prediction: 19.369, Actual: 18.000 Date: 2020-08-23 - Test Prediction: 18.274, Actual: 2.000 Date: 2020-08-24 - Test Prediction: 5.255, Actual: 1.000 Date: 2020-08-25 - Test Prediction: 1.851, Actual: 26.000 Date: 2020-08-26 - Test Prediction: 21.170, Actual: 19.000
Date: 2020-08-22 - Test Prediction: 19.369, Actual: 18.000 Date: 2020-08-23 - Test Prediction: 18.274, Actual: 2.000 Date: 2020-08-24 - Test Prediction: 5.255, Actual: 1.000 Date: 2020-08-25 - Test Prediction: 1.851, Actual: 26.000 Date: 2020-08-26 - Test Prediction: 21.170, Actual: 19.000 Date: 2020-08-27 - Test Prediction: 19.434, Actual: 11.000
Date: 2020-08-22 - Test Prediction: 19.369, Actual: 18.000 Date: 2020-08-23 - Test Prediction: 18.274, Actual: 2.000 Date: 2020-08-24 - Test Prediction: 5.255, Actual: 1.000 Date: 2020-08-25 - Test Prediction: 1.851, Actual: 26.000 Date: 2020-08-26 - Test Prediction: 21.170, Actual: 19.000 Date: 2020-08-27 - Test Prediction: 19.434, Actual: 11.000 Date: 2020-08-28 - Test Prediction: 12.687, Actual: 20.000
Date: 2020-08-22 - Test Prediction: 19.369, Actual: 18.000 Date: 2020-08-23 - Test Prediction: 18.274, Actual: 2.000 Date: 2020-08-24 - Test Prediction: 5.255, Actual: 1.000 Date: 2020-08-25 - Test Prediction: 1.851, Actual: 26.000 Date: 2020-08-26 - Test Prediction: 21.170, Actual: 19.000 Date: 2020-08-27 - Test Prediction: 19.434, Actual: 11.000 Date: 2020-08-28 - Test Prediction: 12.687, Actual: 20.000
Date: 2020-08-22 - Test Prediction: 19.369, Actual: 18.000 Date: 2020-08-23 - Test Prediction: 18.274, Actual: 2.000 Date: 2020-08-24 - Test Prediction: 5.255, Actual: 1.000 Date: 2020-08-25 - Test Prediction: 1.851, Actual: 26.000 Date: 2020-08-26 - Test Prediction: 21.170, Actual: 19.000 Date: 2020-08-27 - Test Prediction: 19.434, Actual: 11.000 Date: 2020-08-28 - Test Prediction: 12.687, Actual: 20.000
Date: 2020-08-22 - Test Prediction: 19.369, Actual: 18.000 Date: 2020-08-23 - Test Prediction: 18.274, Actual: 2.000 Date: 2020-08-24 - Test Prediction: 5.255, Actual: 1.000 Date: 2020-08-25 - Test Prediction: 1.851, Actual: 26.000 Date: 2020-08-26 - Test Prediction: 21.170, Actual: 19.000 Date: 2020-08-27 - Test Prediction: 19.434, Actual: 11.000 Date: 2020-08-28 - Test Prediction: 12.687, Actual: 20.000

Date: 2020-08-25 - Test Prediction: 0.002, Actual: 0.000

Date: 2020-08-26 - Test Prediction: 0.000, Actual: 12.000

Date: 2020-08-27 - Test Prediction: 9.600, Actual: 0.000

Date: 2020-08-28 - Test Prediction: 1.920, Actual: 0.000

-----Errors-----

MAPE:100.00 MSE:143.99

Q2 - i ii

In[88]:

```
In [ ]:
```

```
def plot ar(X, test, predictions):
   plt.plot(X, test, label="original")
   plt.plot(X, predictions, label="Predictions")
   plt.xlabel('Dates - 4th week of August')
   plt.ylabel('Y')
   plt.legend(loc='upper left')
   plt.xticks(rotation=45)
   plt.show()
class AR:
   def init (self, p):
       self.p = p
   def predict(self, train, test):
       test_dates = np.array(test['Date'])
       test counts = np.array(test['Count'])
       data counts = np.hstack([train['Count'], test counts])
       p = self.p
       t = data counts.shape[0] - test counts.shape[0]
       error = []
       predictions = np.zeros(test.shape[0])
       for i in range(t,t+test counts.shape[0]):
            testx = [1]
            testx = np.hstack([[1], data counts[i-p:i]])
            #calculating beta
            X = []
            Y = []
            for j in range(i):
                if (j+p < i):
                    X.append([1])
                    X[j] = X[j]+list(data_counts[j:j+p])
                    Y.append(data_counts[j+p])
                else:
                    break
            beta=np.matmul(np.linalg.inv(np.matmul(np.transpose(X),X)),np.matmul(np.tran
spose(X), Y))
            y t hat = predictions[i-t] = np.dot(testx,beta)
            y t = data counts[i]
            #Ignoring zero data to avoid infinite error
            if y t!=0:
                error.append(abs(y t hat-y t)*100/y t)
                mse.append((y_t_hat - y_t)**2)
            print("Date: " + str(test dates[i-t]) + " - Test prediction: " + "{:5.3f}".f
ormat(predictions[i-t]) + " | Actual: " + str(test counts[i-t]))
```

```
plot ar(test dates, test counts, predictions)
       print("-----Errors-----
       print("MAPE: " + "{:5.3f}".format(np.mean(error)))
       print("MSE : " + "{:5.3f}".format(np.mean(mse)))
       return np.mean(error)
In[89]:
Auto Regression: (Confirmed COVID Cases in August 2020)
redict deaths for 4th week of August with first three weeks data of August
In [ ]:
train PA confirmed = covid PA confirmed[(covid PA confirmed['Date']>='2020-08-01') & (cov
id PA confirmed['Date'] <= '2020-08-21')].reset index(drop=True)</pre>
test PA confirmed = covid PA confirmed[(covid PA confirmed['Date']>='2020-08-22') & (covi
d PA confirmed['Date'] <= '2020-08-28')].reset index(drop=True)</pre>
In [ ]:
train RI confirmed = covid RI confirmed[(covid RI confirmed['Date']>='2020-08-01') & (cov
id RI confirmed['Date'] <= '2020-08-21')].reset_index(drop=True)</pre>
test_RI_confirmed = covid_RI_confirmed[(covid_RI_confirmed['Date']>='2020-08-22') & (covid_RI_confirmed['Date']>='2020-08-22')
d RI confirmed['Date'] <= '2020-08-28')].reset index(drop=True)</pre>
In [ ]:
print("\n----\n")
In [ ]:
print("\n----- Predicting Pennsylvania confirmed cases -----\n")
ar3 = AR(3)
ar3.predict(train PA confirmed, test PA confirmed)
In [ ]:
print("\n----- Predicting Rhode Island confirmed cases -----\n")
ar3 = AR(3)
ar3.predict(train RI confirmed, test RI confirmed)
In [ ]:
print("\n----\n")
In [ ]:
print("\n----- Predicting Pennsylvania confirmed cases -----\n")
ar3 = AR(5)
ar3.predict(train PA confirmed, test PA confirmed)
In [ ]:
print("\n----- Predicting Rhode Island confirmed cases ----\n")
ar3 = AR(5)
ar3.predict(train RI confirmed, test RI confirmed)
----- AR(3) -----
----- Predicting Pennsylvania confirmed cases ------
```

Date: 2020-08-23 - Test prediction: 737.243 | Actual: 619.0

Date: 2020-08-22 - Test prediction: 795.688 | Actual: 796.0

Date: 2020-08-24 - Test prediction: 760.136 | Actual: 426.0 Date: 2020-08-25 - Test prediction: 637.590 | Actual: 561.0 Date: 2020-08-26 - Test prediction: 655.719 | Actual: 501.0 Date: 2020-08-27 - Test prediction: 670.987 | Actual: 620.0 Date: 2020-08-28 - Test prediction: 653.981 | Actual: 835.0 -----Errors------MAPE: 24.573 MSE: 27257.157 ----- Predicting Rhode Island confirmed cases ------Date: 2020-08-22 - Test prediction: 12.529 | Actual: 0.0 Date: 2020-08-23 - Test prediction: 192.698 | Actual: 0.0 Date: 2020-08-24 - Test prediction: 175.213 | Actual: 0.0 Date: 2020-08-25 - Test prediction: 160.637 | Actual: 0.0 Date: 2020-08-26 - Test prediction: 148.300 | Actual: 700.0 Date: 2020-08-27 - Test prediction: -15.950 | Actual: 0.0 Date: 2020-08-28 - Test prediction: -15.944 | Actual: 0.0 ------Errors------MAPE: 78.814 MSE: 304372.448 ------ AR(5) ----------- Predicting Pennsylvania confirmed cases ------Date: 2020-08-22 - Test prediction: 806.785 | Actual: 796.0 Date: 2020-08-23 - Test prediction: 774.646 | Actual: 619.0 Date: 2020-08-24 - Test prediction: 739.230 | Actual: 426.0 Date: 2020-08-25 - Test prediction: 586.330 | Actual: 561.0 Date: 2020-08-26 - Test prediction: 585.629 | Actual: 501.0 Date: 2020-08-27 - Test prediction: 711.559 | Actual: 620.0 Date: 2020-08-28 - Test prediction: 680.020 | Actual: 835.0 -----Errors------

MAPE: 22.109 MSE: 23237.213

```
----- Predicting Rhode Island confirmed cases ------
Date: 2020-08-22 - Test prediction: 21.037 | Actual: 0.0
Date: 2020-08-23 - Test prediction: 21.022 | Actual: 0.0
Date: 2020-08-24 - Test prediction: 21.007 | Actual: 0.0
Date: 2020-08-25 - Test prediction: 322.942 | Actual: 0.0
Date: 2020-08-26 - Test prediction: 258.800 | Actual: 700.0
Date: 2020-08-27 - Test prediction: -28.212 | Actual: 0.0
Date: 2020-08-28 - Test prediction: -28.187 | Actual: 0.0
-----Errors-----
MAPE: 63.029
MSE: 194657.654
In[90]:
Auto Regression: (Confirmed COVID Cases in March 2020)
time series for the month of March (03/01/2020 to 03/31/2020)
we will predict Confirmed Covid Cases
for all Counties
print("Auto Regression: (Confirmed COVID Cases in March 2020)")
In [ ]:
train PA deaths = covid PA deaths[(covid PA deaths['Date']>='2020-08-01') & (covid PA de
aths['Date'] <= '2020-08-21')].reset index(drop=True)
test PA deaths = covid PA deaths[(covid PA deaths['Date']>='2020-08-22') & (covid PA deat
hs['Date'] <= '2020-08-28')].reset index(drop=True)
In [ ]:
train RI deaths = covid RI deaths[(covid RI deaths['Date']>='2020-08-01') & (covid RI de
aths['Date']<='2020-08-21')].reset index(drop=True)
test RI deaths = covid RI deaths[(covid RI deaths['Date']>='2020-08-22') & (covid RI deat
hs['Date'] <= '2020-08-28')].reset index(drop=True)
In [ ]:
print("\n-----\n")
In [ ]:
print("\n----- Predicting Pennsylvania deaths -----\n")
ar3.predict(train PA deaths, test PA deaths)
In [ ]:
print("\n----- Predicting Rhode Island deaths ----\n")
ar3.predict(train RI deaths, test RI deaths)
In [ ]:
print("\n-----\n")
```

```
In [ ]:
print("\n----- Predicting Pennsylvania deaths -----\n")
ar3 = AR(5)
ar3.predict(train PA deaths, test PA deaths)
In [ ]:
print("\n-----\n")
ar3 = AR(5)
ar3.predict(train RI deaths, test RI deaths)
------ AR(3) ------
----- Predicting Pennsylvania deaths ------
Date: 2020-08-22 - Test prediction: 17.888 | Actual: 18.0
Date: 2020-08-23 - Test prediction: 19.717 | Actual: 2.0
Date: 2020-08-24 - Test prediction: 16.444 | Actual: 1.0
Date: 2020-08-25 - Test prediction: 17.587 | Actual: 26.0
Date: 2020-08-26 - Test prediction: 26.527 | Actual: 19.0
Date: 2020-08-27 - Test prediction: 20.524 | Actual: 11.0
Date: 2020-08-28 - Test prediction: 14.622 | Actual: 20.0
-----Errors------
MAPE: 373.762
MSE: 114.213
----- Predicting Rhode Island deaths -----
Date: 2020-08-22 - Test prediction: 0.970 | Actual: 0.0
Date: 2020-08-23 - Test prediction: 2.213 | Actual: 0.0
Date: 2020-08-24 - Test prediction: 2.016 | Actual: 0.0
Date: 2020-08-25 - Test prediction: 1.851 | Actual: 0.0
Date: 2020-08-26 - Test prediction: 1.711 | Actual: 12.0
Date: 2020-08-27 - Test prediction: 3.749 | Actual: 0.0
Date: 2020-08-28 - Test prediction: -0.030 | Actual: 0.0
 ------Errors------
MAPE: 85.742
MSE: 105.865
```

------ AR(5) ------

```
----- Predicting Pennsylvania deaths ------
Date: 2020-08-22 - Test prediction: 21.082 | Actual: 18.0
Date: 2020-08-23 - Test prediction: 14.471 | Actual: 2.0
Date: 2020-08-24 - Test prediction: 18.595 | Actual: 1.0
Date: 2020-08-25 - Test prediction: 24.901 | Actual: 26.0
Date: 2020-08-26 - Test prediction: 29.952 | Actual: 19.0
Date: 2020-08-27 - Test prediction: 24.052 | Actual: 11.0
Date: 2020-08-28 - Test prediction: 24.223 | Actual: 20.0
-----Errors-----
MAPE: 371.691
MSE: 111.996
----- Predicting Rhode Island deaths -----
Date: 2020-08-22 - Test prediction: 2.226 | Actual: 0.0
Date: 2020-08-23 - Test prediction: 2.228 | Actual: 0.0
Date: 2020-08-24 - Test prediction: 0.909 | Actual: 0.0
Date: 2020-08-25 - Test prediction: 2.938 | Actual: 0.0
Date: 2020-08-26 - Test prediction: 2.548 | Actual: 12.0
Date: 2020-08-27 - Test prediction: 4.419 | Actual: 0.0
Date: 2020-08-28 - Test prediction: -0.202 | Actual: 0.0
-----Errors------
MAPE: 78.769
MSE: 89.346
In [ ]:
covid_PA_confirmed_feb21 = covid_PA_confirmed[377:404].to_numpy()
covid_RI_confirmed_feb21 = covid_RI_confirmed[377:404].to_numpy()
covid_PA_deaths_feb21 = covid_PA_deaths[377:404].to_numpy()
covid RI deaths feb21 = covid RI deaths[377:404].to numpy()
In [ ]:
covid PA confirmed mar21 = covid PA confirmed[404:435].to numpy()
covid RI confirmed mar21 = covid RI confirmed[404:435].to numpy()
covid PA deaths mar21 = covid PA deaths[404:435].to numpy()
covid RI deaths mar21 = covid RI deaths[404:435].to numpy()
# print(covid RI deaths mar21)
In[18]:
In [ ]:
```

def corrected variance(arr):

```
square_sum = 0
mean = np.mean(arr)
n = len(arr)
for i in range(n):
    square_sum = square_sum + (arr[i] -mean)*(arr[i]-mean)
return square_sum/(n-1)
```

In[19]:

```
In [ ]:
```

```
def walds_test_1sample(dist1, dist2, threshold, descr):
    dist1_mean = np.mean(dist1)
    dist2_mean = np.mean(dist2)
    numerator = dist2_mean - dist1_mean
    denominator = np.sqrt(dist2_mean/dist2.size)
    result = numerator/denominator
    w = np.abs(result)
    if(w>threshold):
        print("walds 1 sample testing for mean of "+str(descr)+" cases is w="+str(w) +" wh
ich is greater than z_alpha/2 = "+str(threshold)+" so reject the NULL hypothesis");
    else:
        print("walds 1 sample testing for mean of "+str(descr)+" cases is w="+str(w)+" which is less than z_alpha/2 = "+str(threshold)+" so accept the NULL hypothesis")
```

In[20]:

```
In [ ]:
```

```
def walds_test_2sample(dist1, dist2, threshold, descr):
    dist1_mean = np.mean(dist1)
    dist2_mean = np.mean(dist2)

numer = dist2_mean-dist1_mean
    denom = np.sqrt(dist2_mean/dist2.size + dist1_mean/dist1.size)
    w = abs(numer/denom)
    if(w>threshold):
        print("walds 2 sample testing for mean of "+str(descr)+" cases is w="+str(w) +" wh
    ich is greater than z_alpha/2 = "+str(threshold)+" so reject the NULL hypothesis");
    else:
        print("walds 2 sample testing for mean of "+str(descr)+" cases is w="+str(w)+" w
    hich is less than z_alpha/2 = "+str(threshold)+" so accept the NULL hypothesis")
```

In[21]:

z test

```
In [ ]:
```

```
def z_test(dist1, dist2, full, zthreshold, descr):
    dist1_mean = np.mean(dist1)
    dist2_mean = np.mean(dist2)

numer = dist2_mean - dist1_mean
    denom = np.sqrt(corrected_variance(full)/full.size)

z = abs(numer/denom)

if(z>zthreshold):
    print("z test 1 sample testing for mean of "+str(descr)+" cases is w="+str(z) +" w
hich is greater than z_alpha/2 = "+str(zthreshold)+" so reject the NULL hypothesis");
    else:
        print("z test 1 sample testing for mean of "+str(descr)+" cases is w="+str(z)+"
which is less than z_alpha/2 = "+str(zthreshold)+" so accept the NULL hypothesis")
```

```
t test
```

```
In []:

def t_test_lsample(dist1, dist2, tthreshold, descr):
    dist1_mean = np.mean(dist1)
    dist2_mean = np.mean(dist2)

numer = dist2_mean - dist1_mean
    denom = np.sqrt(np.var(dist2)/dist2.size)

t = abs(numer/denom)

if(t>tthreshold):
    print("t test 1 sample testing for mean of "+str(descr)+" cases is w="+str(t) +" w
hich is greater than t threshold = "+str(tthreshold)+" so reject the NULL hypothesis");
    else:
        print("t test 1 sample testing for mean of "+str(descr)+" cases is w="+str(t)+" which is less than t thresfold = "+str(tthreshold)+" so accept the NULL hypothesis")
```

In[23]:

t test

```
In [ ]:
```

```
def t_test_2sample_unpaired(dist1, dist2, tthreshold, descr):
    dist1_mean = np.mean(dist1)
    dist2_mean = np.mean(dist2)

numer = dist2_mean - dist1_mean
    denom = np.sqrt(np.var(dist2)/dist2.size + np.var(dist1)/dist1.size)

t = abs(numer/denom)

if(t>tthreshold):
    print("t test 2 sample testing for mean of "+str(descr)+" cases is w="+str(t) +" w
hich is greater than t threshold = "+str(tthreshold)+" so reject the NULL hypothesis");
    else:
        print("t test 2 sample testing for mean of "+str(descr)+" cases is w="+str(t)+"
which is less than t thresfold = "+str(tthreshold)+" so accept the NULL hypothesis")
```

In[24]:

walds 1 cases

```
In [ ]:
```

```
walds_test_1sample(covid_PA_confirmed_feb21, covid_PA_confirmed_mar21, 1.962, "PA confirm
ed")
walds_test_1sample(covid_RI_confirmed_feb21, covid_RI_confirmed_mar21, 1.962, "RI confirm
ed")
```

```
In [ ]:
```

```
walds_test_1sample(covid_PA_deaths_feb21, covid_PA_deaths_mar21, 1.962, "PA_death")
walds_test_1sample(covid_RI_deaths_feb21, covid_RI_deaths_mar21, 1.962, "RI_death")
```

In[25]:

z test cases

```
In [ ]:
```

```
z_test(covid_PA_confirmed_feb21, covid_PA_confirmed_mar21, covid_PA_confirmed.to_numpy(),
1.962, "PA confirmed")
```

```
In []:

z_test(covid_PA_deaths_feb21, covid_PA_deaths_mar21, covid_PA_deaths.to_numpy(), 1.962, "
PA_death")
z_test(covid_RI_deaths_feb21, covid_RI_deaths_mar21, covid_RI_deaths.to_numpy(), 1.962, "
RI_death")
```

z_test(covid_RI_confirmed_feb21, covid_RI_confirmed_mar21, covid_RI_confirmed.to_numpy(),

In[26]:

t test 1 sample cases

```
In [ ]:
```

```
t_test_1sample(covid_PA_confirmed_feb21, covid_PA_confirmed_mar21, 1.695, "PA confirmed")
t_test_1sample(covid_RI_confirmed_feb21, covid_RI_confirmed_mar21, 1.695, "RI confirmed")
```

In []:

```
t_test_1sample(covid_PA_deaths_feb21, covid_PA_deaths_mar21, 1.695, "PA_death")
t_test_1sample(covid_RI_deaths_feb21, covid_RI_deaths_mar21, 1.695, "RI_death")
```

In[27]:

walds 2 sample cases

In []:

```
walds_test_2sample(covid_PA_confirmed_feb21, covid_PA_confirmed_mar21, 1.962, "PA confirm
ed")
walds_test_2sample(covid_RI_confirmed_feb21, covid_RI_confirmed_mar21, 1.962, "RI confirm
ed")
```

In []:

```
walds_test_2sample(covid_PA_deaths_feb21, covid_PA_deaths_mar21, 1.962, "PA_death")
walds_test_2sample(covid_RI_deaths_feb21, covid_RI_deaths_mar21, 1.962, "RI_death")
```

In[28]:

t test 1 sample cases

```
In [ ]:
```

```
t_test_2sample_unpaired(covid_PA_confirmed_feb21, covid_PA_confirmed_mar21, 1.672, "PA confirmed")
t_test_2sample_unpaired(covid_RI_confirmed_feb21, covid_RI_confirmed_mar21, 1.672, "RI confirmed")
```

```
In [ ]:
```

```
t_test_2sample_unpaired(covid_PA_deaths_feb21, covid_PA_deaths_mar21, 1.672, "PA_death")
t_test_2sample_unpaired(covid_RI_deaths_feb21, covid_RI_deaths_mar21, 1.672, "RI_death")
```

Applicability Of Tests:

Wald's Test:

- 1. We require an asymptomatical normal estimator for wald's test and as n = 28, 30 is fairly low, this is not the ideal case for assumption of CLT for sample mean. Therefore, this test is not applicable.
- 2. The above reasoning works well for 2 sample test as well, since we need both estimators to be asymptomatically Normal. Therfore this test is also not applicable.

Z-test:

3. As we need to use true standard deviation in z-test, the dataset size is just 438, which is not particularly large. Neither the datasets are normally distributed. Therefore this testis not applicable.

T-test:

- 4. For one sample, similarly to above tests, the data inferred here is neither normally distributed nor large enough. Therefore the test is not applicable.
- 5. For unpaired 2 sample test, distributions need to be independent and normally distributed. But that doesnt seem the same in above case. Therefore the test is not applicable.

```
In [ ]:
```

```
data = covid_data
data_oct_to_dec = data[data["Date"] >= '2020-10-01'][data["Date"] <= '2020-12-31']</pre>
```

In []:

```
pa_confirmed = np.array(data_oct_to_dec["PA confirmed"]).astype('int')
ri_confirmed = np.array(data_oct_to_dec["RI confirmed"]).astype('int')
```

In []:

```
pa_deaths = np.array(data_oct_to_dec["PA deaths"]).astype('int')
ri_deaths = np.array(data_oct_to_dec["RI deaths"]).astype('int')
```

In[461]:

```
In [ ]:
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
In [ ]:
```

```
from scipy.stats import poisson, binom, geom
```

K-S Test 1-sample and 2-Sample

For 1-sample we are using MME to find the parameter estimates for PA data and testing if that is same for RI data

In[462]:

```
In [ ]:
```

```
def get_eCDF(data, s, e):
    n = len(data)
    cdf = [0.0]
    for i in range(0, n):
        cdf = cdf + [cdf[len(cdf)-1] + (1/n)]
    return cdf

def plot_eCDF(data, cdf, label, s, e):
    # Since the ecdf has two values for some x, we have to create new x and y lists from cdf
    n = len(data)
    x, y = [s], [0]
```

```
for i in range (0, n):
       x = x + [data[i], data[i]]
       y = y + [cdf[i], cdf[i+1]]
   x = x + [e]
   y = y + [1.0]
   plt.plot(x, y, label=label)
   plt.xlabel('x')
   plt.ylabel('CDF')
   plt.legend(loc='best')
def cdf at x(data, cdf, x):
   # finding the first element larger than the x and then we take cdf value at that poin
t
   x i = -1
   for i, d in enumerate(data):
       if d >= x:
           x i = i
           break
   return cdf[x i]
```

In []:

```
def k s test 2 sample (pa, ri, c = 0.05):
    # Sort the data to get the CDFs
   pa, ri = np.sort(pa), np.sort(ri)
    s = min(pa[0], ri[0]) - 100
    e = max(pa[len(pa) - 1], ri[len(ri) - 1]) + 100
    F pa = get eCDF(pa, s, e)
    F ri = get eCDF(ri, s, e)
    fig= plt.figure(figsize=(12,9))
    plt.grid(True)
    plt.title("K-S Test 2-Sample PA vs RI")
    plot_eCDF(pa, F_pa, "PA", s, e)
   plot_eCDF(ri, F_ri, "RI", s, e)
    # pa CDF at ri change points
    F_pa = [cdf_at_x(pa, F_pa, cp) for cp in ri]
    F ri minus, F ri plus = F ri[0:-1], F ri[1:]
    ks stat = ks index = 0
    # y value for the vertical line at ks index
    ymin = 0
    for i in range(0, len(F pa)):
        if abs(F_pa[i] - F_ri_minus[i]) > ks_stat:
            ks_stat = abs(F_pa[i] - F_ri_minus[i])
            ks_index = i
            ymin = min(F ri minus[ks index], F pa[ks index])
        if abs(F pa[i] - F_ri_plus[i]) > ks_stat:
            ks stat = abs(F_pa[i] - F_ri_plus[i])
            ks index = i
            ymin = min(F ri plus[ks index], F pa[ks index])
    print("ks stat is {} at {}, where as c is {}".format(ks stat, ri[ks index], c))
    if ks stat > c:
       print("d > c, So, we reject Null Hypothesis")
    else:
       print("d <= c, So, we accept Null Hypothesis")</pre>
    plt.plot([ri[ks index], ri[ks index]],[ymin, ymin + ks stat])
```

```
plt.annotate("K-S statistic = " + str(ks_stat), xy = [ri[ks_index] + int((e-s)/100),
ymin + ks_stat/4], rotation = 90)
plt.show()
return
```

```
In [ ]:
```

```
def k s test 1 sample(pa, ri, dist='poisson', c = 0.05):
    pa mean = np.mean(pa)
   pa_var = np.var(pa)
    # Sort the data to get the CDFs
   pa, ri = np.sort(pa), np.sort(ri)
    s = min(pa[0], ri[0]) - 100
    e = max(pa[len(pa) - 1], ri[len(ri) - 1]) + 100
    F pa = []
    F ri = get eCDF(sorted ri confirmed, s, e)
    # pa CDF at ri change points
    if dist == 'poisson':
        lam mme = pa_mean
        F pa = [poisson.cdf(cp, lam mme) for cp in ri]
    elif dist == 'binomial':
       p mme = 1 - pa_var/pa_mean
       n mme = pa mean/p mme
        F_pa = [binom.cdf(cp, n_mme, p_mme) for cp in ri]
    elif dist == 'geometric':
       p mme = 1/pa mean
        F pa = [geom.cdf(cp, p_mme) for cp in ri]
    F ri minus, F ri plus = F ri[0:-1], F ri[1:]
    ks stat = ks index = 0
    for i in range(0, len(X)):
        if abs(F pa[i] - F ri minus[i]) > ks stat:
            ks_stat = abs(F_pa[i] - F_ri_minus[i])
            ks_index = i
        if abs(F pa[i] - F ri plus[i]) > ks stat:
            ks stat = abs(F pa[i] - F ri plus[i])
            ks index = i
    print("ks stat is {} at {}, where as c is {}".format(ks stat, ri[ks index], c))
    if ks stat > c:
        print("d > c, So, we reject Null Hypothesis")
    else:
       print("d <= c, So, we accept Null Hypothesis")</pre>
    return
```

K-S Test 1 Sample - PA Confimed assumed to be poisson distribution vs RI Confimed, threshold, c = 0.05

Null Hypothesis: H_o = RI Confimed is poisson distribution with parameters obtained from MME on PA Confimed.

Alternate Hypothesis: H_1 = RI Confimed is poisson distribution with parameters obtained from MME on PA Confimed.

In[463]:

```
In [ ]:
```

```
k_s_test_1_sample(pa_confirmed, ri_confirmed)
```

K-S Test 1 Sample - PA Deaths assumed to be poisson distribution vs RI Deaths, threshold, c = 0.05

Null Hypothesis: H_o = RI Deaths is poisson distribution with parameters obtained from MME on PA Deaths.

Alternate Hypothesis: H_1 = RI Deaths is poisson distribution with parameters obtained from MME on PA Deaths.

In[464]:

```
In [ ]:
```

```
k s test 1 sample (pa deaths, ri deaths)
```

K-S Test 1 Sample - PA Confirmed aussmed to be binomial distribution vs RI Confirmed, threshold, c = 0.05

Null Hypothesis: H_o = RI Confirmed is binomial distribution with parameters obtained from MME on PA Confirmed.

Alternate Hypothesis: H_1 = RI Confirmed is binomial distribution with parameters obtained from MME on PA Confirmed.

In[465]:

```
In [ ]:
```

```
k s test 1 sample(pa confirmed, ri confirmed, dist='binomial')
```

K-S Test 1 Sample - PA Deaths assumed to be binomial distribution vs RI Deaths, threshold, c = 0.05

Null Hypothesis: H_o = RI Deaths is binomial distribution with parameters obtained from MME on PA Deaths.

Alternate Hypothesis: H_1 = RI Deaths is binomial distribution with parameters obtained from MME on PA Deaths.

In[466]:

```
In [ ]:
```

```
k s test 1 sample(pa deaths, ri deaths, dist='binomial')
```

K-S Test 1 Sample - PA Confirmed assumed to be geometric distribution vs RI Confirmed, threshold, c = 0.05

Null Hypothesis: H_o = RI Confirmed is geometric distribution with parameters obtained from MME on PA Confimred.

Alternate Hypothesis: H_1 = RI Confirmed is geometric distribution with parameters obtained from MME on PA Confirmed.

In[467]:

```
In [ ]:
```

```
k s test 1 sample(pa confirmed, ri confirmed, dist='geometric')
```

K-S Test 1 Sample - PA Deaths assumed to be geometric distribution vs RI Deaths, threshold, c = 0.05

Null Hypothesis: H_o = RI Deaths is geometric distribution with parameters obtained from MME on PA Deaths.

Alternate Hypothesis: H_1 = RI Deaths is geometric distribution with parameters obtained from MME on PA Deaths.

In[468]:

```
In [ ]:
```

```
k_s_test_1_sample(pa_deaths, ri_deaths, dist='geometric')
```

K-S Test 2 Sample - threshold, c = 0.05

Null Hypothesis: H_o = PA Confirmed and RI Confirmed have similar distributions.

Alternate Hypothesis: H_1 = PA Confirmed and RI Confirmed have different distributions.

In[469]:

```
In [ ]:
```

```
k_s_test_2_sample(pa_confirmed, ri_confirmed)
```

K-S Test 2 Sample - threshold, c = 0.05

Null Hypothesis: H_o = PA Deaths and RI Deaths have similar distributions.

Alternate Hypothesis: H_1 = PA Deaths and RI Deaths have different distributions.

In[470]:

```
In [ ]:
```

```
k_s_test_2_sample(pa_deaths, ri_deaths)
```

Permutation Test - 1000 permutations, threshold, c = 0.05

In[473]:

```
In [ ]:
```

```
def perm_test(pa, ri, no_of_perm):
    count =0

    perm_diff = abs(np.mean(pa) - np.mean(ri))
    covid = np.array(pa.tolist() + ri.tolist())

    pa_n = pa.shape[0]
    N = covid.shape[0]

    for i in range(no_of_perm):
        perm = np.random.permutation(covid)
        d1 = perm[:pa_n+1]
        d2 = perm[pa_n+1:]
        if (abs(d1.mean() - d2.mean()) > perm_diff):
             count += 1

    p_val = count/no_of_perm

    return p val
```

Null Hypothesis: H_o = PA Confirmed and RI Confirmed have similar distributions.

Alternate Hypothesis: H_1 = PA Confirmed and RI Confirmed have different distributions.

In[474]:

```
In [ ]:
```

```
c = 0.05
p_val_confirmed = perm_test(pa_confirmed, ri_confirmed, 1000)
```

```
print("p-value for confirmed cases = \{\}\ <=\ c=\{\}\ . So, we reject Null Hypothesis.".format (p_{val}_{confirmed},\ c))
```

Null Hypothesis: H_o = PA Deaths and RI Deaths have similar distributions.

Alternate Hypothesis: H_1 = PA Deaths and RI Deaths have different distributions.

In[475]:

```
In [ ]:
```

```
p_val_deaths = perm_test(pa_deaths, ri_deaths, 1000)
print("p-value for deaths = {} <= c = {}. So, we reject Null Hypothesis.".format(p_val_de aths, c))</pre>
```



```
In [ ]:
```

```
covid_data_actual = covid_data
covid_combined = pd.DataFrame()
covid_combined["date"] = covid_data_actual["Date"]
covid_combined['date'] = pd.to_datetime(covid_combined['date'])
covid_combined["cases"] = covid_data_actual["PA confirmed"] + covid_data_actual["RI confirmed"]
covid_combined["deaths"] = covid_data_actual["PA deaths"] + covid_data_actual["RI deaths"]
covid_combined = covid_combined.set_index(['date'])
```

In[139]:

```
In [ ]:
```

```
covid_jun_four = covid combined.loc['2020-6-1':'2020-6-28']
al = covid combined.loc['2020-6-29':'2020-7-05']
# print(al['deaths'].mean())
# covid jun four.reset index(drop=False, inplace=True)
lambda mme = covid jun four['deaths'].mean()
# print(lambda_mme)
beta = lambda_mme
week 5 = \text{covid} \text{ combined.loc}['2020-6-29':'2020-7-05']
# week 5.reset index(drop=False, inplace=True)
deaths 5 = list(week 5['deaths'])
week 6 = covid combined.loc['2020-7-06':'2020-7-12']
# week 6.reset index(drop=False, inplace=True)
deaths 6 = list(week 6['deaths'])
week_7 = covid_combined.loc['2020-7-13':'2020-7-19']
# week_7.reset_index(drop=False, inplace=True)
deaths_7 = list(week_7['deaths'])
week 8 = \text{covid combined.loc}['2020-7-20':'2020-7-26']
# week 8.reset index(drop=False, inplace=True)
deaths 8 = list(week 8['deaths'])
```

print(np.mean(deaths_5), np.mean(deaths_6), np.mean(deaths_7), np.mean(deaths_8))

```
In [ ]:
```

```
deaths = [deaths_5,deaths_6,deaths_7,deaths_8]
deaths
# deaths_alt = deaths_5+deaths_6+deaths_7+deaths_8
# print(deaths_alt)
```

```
In [ ]:
```

```
plt.figure(figsize=(16,8))
death_sum = 0
```

```
i=0
for d_i in deaths:
    death_sum += sum(d_i)
    alpha = death_sum + 1
    b = (i+1)*7 + (1/beta)
    x = np.linspace(gamma.ppf(0.01, alpha, scale=1/b), gamma.ppf(0.99, alpha, scale=1/b)
, 100)
    plt.title("Posterior Gamma distributions")
    label= "Week-" + str(i+5) + " MAP(mean): " + str(alpha/b)
    plt.plot(x, gamma.pdf(x, alpha, scale=1/b), label=label)
    plt.xlabel("Deaths")
    plt.ylabel("PDF of Gamma distribution")
    plt.legend()
    i+=1
plt.show()
```

```
In [ ]:
```

```
In [ ]:
```

Observations:

- From the above graphs, we can say that as the weeks progress, MAP is reducing indicating a decrease in number of deaths
- We can also infer that as the time progresses, the number of deaths might saturate if the trend follows a similar pattern (rate)