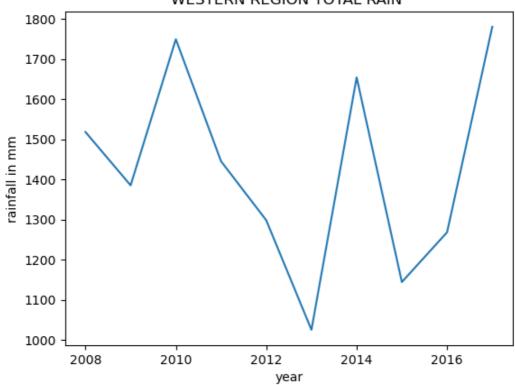
```
In [2]:
import pandas as pd
import numpy as np
import os
import sys
import matplotlib.pylab as plt
import plotly.graph objects as go
import plotly.express as px
                                                    In [3]:
# Useful parameters
DATA DIR = 'datasets/'
prod estimates_dataset = 'PRODUCTION ESTIMATES.csv'
wholesale prices dataset = 'NATIONAL WHOLESALE PRICE OF
SOME SELECTED COMMODITIES.csv'
rainfall dataset = 'RAINFALL.csv'
path to prod est dataset = os.path.join(DATA DIR,
prod estimates dataset)
path to ws prices dataset = os.path.join(DATA DIR,
wholesale prices dataset)
path_to_rainfall_dataset = os.path.join(DATA_DIR,
rainfall dataset)
print('Production estimates dataset is found in:',
path to prod est dataset)
print('Wholesale prices dataset is found in:',
path_to_ws_prices_dataset)
print('Rainfall dataset is found in:',
path to rainfall dataset)
      Production estimates dataset is found in:
      datasets/PRODUCTION ESTIMATES.csv
      Wholesale prices dataset is found in: datasets/
      NATIONAL WHOLESALE PRICE OF SOME SELECTED
      COMMODITIES.csv
      Rainfall dataset is found in: datasets/
      RAINFALL.csv
                                                    In [4]:
data rain= pd.read csv('datasets/RAINFALL.csv')
data rain.dtypes
                              int64
      YEAR
 Out[
                             object
      REGION
  41:
                              int64
      TOTAL RAINFALL(MM)
      dtype: object
                                                    In [5]:
data production = pd.read csv('datasets/production.csv')
data production.dtypes
```

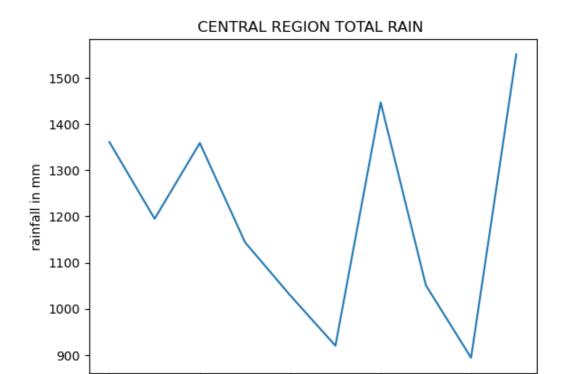
```
object
      REGION
Out[
                         object
      DISTRICT
 51:
      YEAR
                          int64
      CROP
                         object
      AREA (HA)
                         object
      YIELD (MT/HA)
                         object
      PRODUCTION (MT)
                         object
      dtype: object
                                                  In [6]:
df_whole= pd.read_csv('datasets/wholesale.csv',
encoding="ISO-8859-1")
dw= df whole
                                                  In [7]:
data production.rename(columns={'YIELD (MT/HA)':'Yield',
'PRODUCTION (MT)': 'Production'}, inplace=True)
data production= data production.dropna(axis=0)
dp=
data_production.drop(data_production[data_production.Yiel
d.str.contains(r'[-]')].index)
dp= dp.drop(dp[dp.Production.str.contains(r'[,]')].index)
dp['Yield'] = dp.Yield.astype(float)
dp['Production'] = dp.Production.astype(float)
                                                  In [8]:
# SET SIMPLE NAMES FOR DATA SETS. dr=rain data,
dw=wholesale data, dp= production data
dr=data rain
dw=df_whole
                                                 In [17]:
In [15]:
# Check for how rainfall changed for each region for each
month as base to later compare production
regions= dr['REGION'].unique()
year = dp['YEAR'].unique()
crops = dp['CROP'].unique()
for x in range(0,len(regions)):
   R = dr[dr['REGION']== regions[x]].reset_index()
   plt.plot(R['YEAR'],R['TOTAL RAINFALL(MM)'])
   plt.title(regions[x]+ ' REGION TOTAL RAIN ')
   plt.xlabel('year')
```

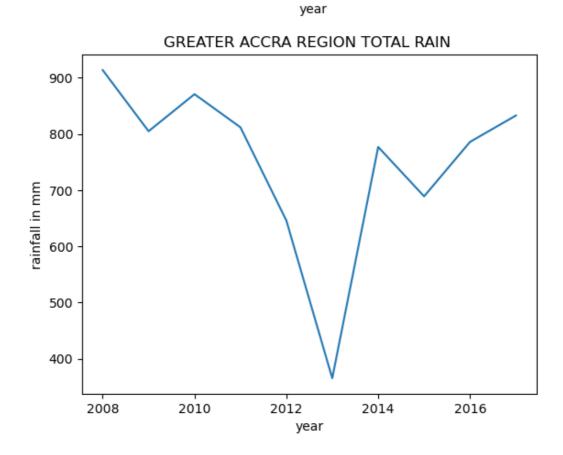
```
plt.ylabel('rainfall in mm')
plt.show()

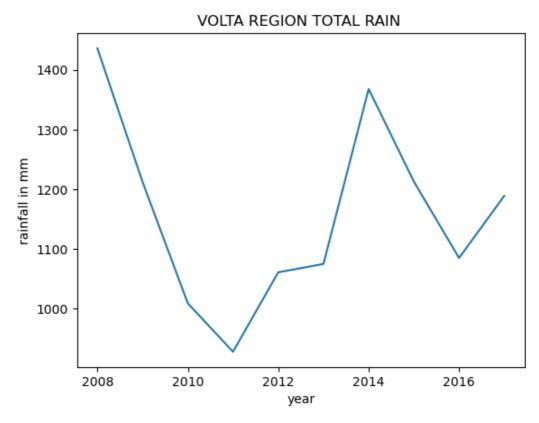
"""compare rainfall directly with production"""
```

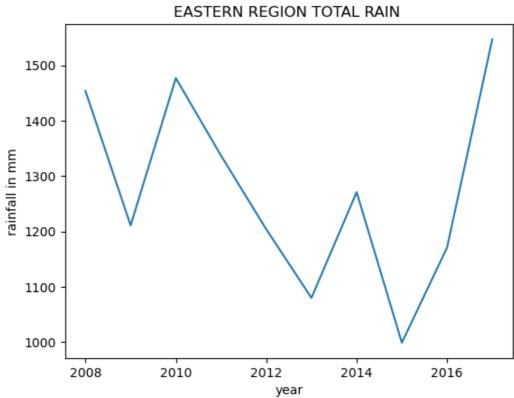
## WESTERN REGION TOTAL RAIN

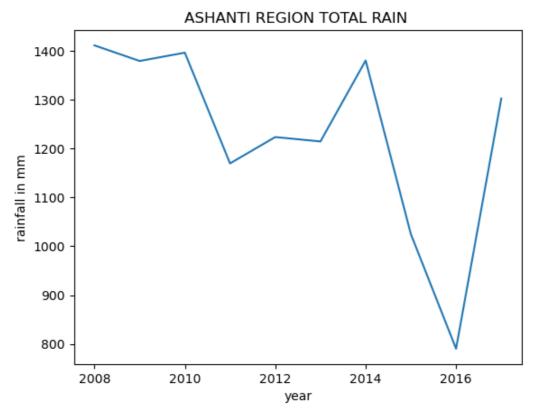


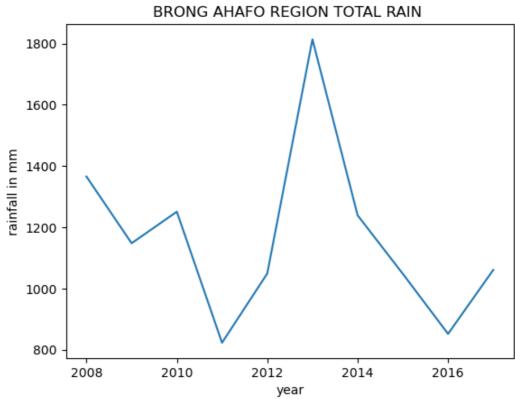


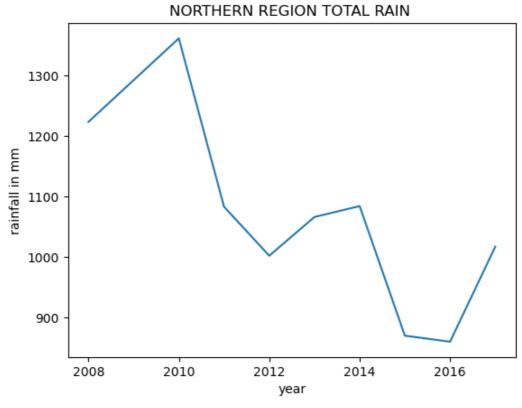


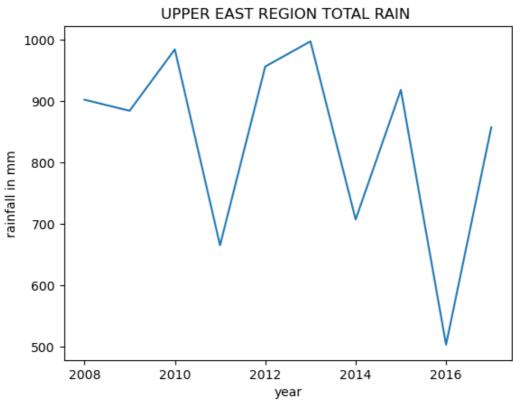




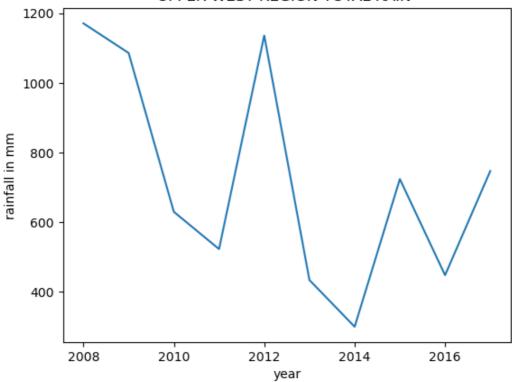












0ut[ 'compare rainfall directly with production' 15]:

```
In [18]:
crops= dw.COMMODITY.unique()
h={}
for i in range(len(year)):
    for n in range (len(crops)):
        x2=dw[dw.COMMODITY==crops[n]]
        x1=x2[x2.YEAR==year[i]]
        x=x1["Price"].mean()
        h[crops[n]]={year[i]:x }
print(h)
"""xaxis = list(h.keys())
yaxis = list(h.values())
plt.plot(xaxis,yaxis)
plt.xlabel('year')
plt.ylabel('average price')
plt.show() """
```

```
{'MAIZE': {2017: 104.47209166666666}, 'MILLET':
       {2017: 19.230500000000003}, 'LOCAL RICE': {2017:
       209.91091666666668}, 'SORGHUM': {2017:
       0.032499999999999994}, 'YAM': {2017:
       516.4208333333333}, 'COCOYAM': {2017: 204.38}}
      "xaxis = list(h.keys())\nyaxis = list(h.values())
 Out[
      \nplt.plot(xaxis,yaxis)\nplt.xlabel('year')
 181:
      \nplt.ylabel('average price')\nplt.show() "
                                                    In [9]:
month switch={'JANUARY': 1, 'FEBRUARY': 2, 'MARCH':
3, 'APRIL': 4, 'MAY': 5, 'JUNE': 6, 'JULY': 7, 'AUGUST':
8, 'SEPTEMBER': 9, 'OCTOBER': 10, 'NOVEMBER': 11, 'DECEMBER':
12}
                                                    In [20]:
dw['monthnum'] = (pd.Series(dw.MONTH)).map(month switch)
crops=dp.CROP.unique()
                                                    In [11]:
dateframe=pd.DataFrame({'yy':dw.YEAR, 'mm':dw.monthnum,
'dd': np.ones(len(dw.YEAR))})
dateframe.columns=['year','month','day']
dw['Date']=pd.to datetime(dateframe)
                                                   In [21]:
data crops=[]
for i in range(len(crops)):
    data crops
```

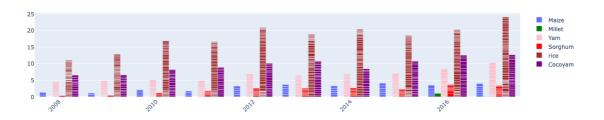
## Normalize yield and find most produced plant and the change in region and years

```
In [22]:
dp_maize=dp[dp['CROP']=='MAIZE']
m=dp_maize.Yield.max()
dp_maize['Yield']=dp_maize['Yield'].div(m).round(2)
dp_millet=dp[dp['CROP']=='MILLET']
mi=dp_millet.Yield.max()
dp_millet['Yield']=dp_millet['Yield'].div(mi).round(2)
dp_rice=dp[dp['CROP']=='RICE']
r=dp_rice.Yield.max()
dp_rice['Yield']=dp_rice['Yield'].div(r).round(2)
dp_sorghum=dp[dp['CROP']=='SORGHUM']
```

```
s=dp sorghum.Yield.max()
dp sorghum['Yield']=dp sorghum['Yield'].div(s).round(2)
dp_yam=dp[dp['CROP']=='YAM']
y=dp yam.Yield.max()
dp yam['Yield']=dp yam['Yield'].div(y).round(2)
dp cocoyam=dp[dp['CROP']=='COCOYAM']
c=dp cocoyam.Yield.max()
dp_cocoyam['Yield']=dp_cocoyam['Yield'].div(c)
                                                   In [23]:
fig = go.Figure()
fig.add trace(go.Bar(
    x=dp maize["YEAR"],
    y=dp_maize['Yield'],
    name='Maize'
))
fig.add trace(go.Bar(
    x=dp_millet["YEAR"],
    y=dp_millet["Yield"],
    name='Millet',
   marker color='green'
))
fig.add_trace(go.Bar(
    x=dp yam["YEAR"],
    y=dp yam["Yield"],
    name='Yam',
    marker color='pink'
))
fig.add trace(go.Bar(
    x=dp sorghum["YEAR"],
    y=dp_sorghum["Yield"],
    name='Sorghum',
    marker_color='red'))
fig.add trace(go.Bar(
    x=dp rice["YEAR"],
    y=dp_rice["Yield"],
    name='rice',
    marker color='brown'))
fig.add trace(go.Bar(
    x=dp_cocoyam["YEAR"],
    y=dp cocoyam["Yield"],
```

```
name='Cocoyam',
    marker_color='purple'))

fig.update_layout(barmode='group', xaxis_tickangle=-45)
fig.show()
```



## Use merged data in pivot

In [28]: #pip install pivottablejs import pandas as pd from pivottablejs import pivot ui pivot ui(df merged) REGION ▼ TOTAL RAINFALL(MM) ▼ YEAR -CROP ▼ 2015 CROP сосоуам COWPEA 11.55 218.37 GROUNDNUT 8.10 114.45 MAIZE MILLET PLANTAIN 38.14 32.40 40.20 40.44 RICE 78.08 SORGHUM 1.40 1.44 SOYABEAN 7.55 8.88 9.02 103.43 65.93 93.75 88.25 93.39 227.39 237.59 309.77 311.10 431.87 443.43 438.55 450.72

In [27]:

```
import seaborn as sns

# Create heatmap using seaborn
plt.figure(figsize=(10, 8))
sns.heatmap(df_merged.corr(), cmap='coolwarm',
annot=True, fmt=".2g", linewidths=0.5)

Out[ <AxesSubplot:>
27]:
```

