```
# for manipulation
In [4]:
          import numpy as np
          import pandas as pd
          # for data visulization
          import matplotlib.pyplot as plt
          import seaborn as sns
          #for data interactivity
          from ipywidgets import interact
         #read the dataset
In [5]:
          data=pd.read csv('data.csv')
         # check the shape of data set
In [6]:
          print("shape of the dataset :",data.shape)
         shape of the dataset : (2200, 8)
         #check the head of the data set
In [7]:
          data.head
        <bound method NDFrame.head of</pre>
                                                         K temperature
                                                                           humidity
                                                                                            ph
Out[7]:
                                                 N
                                                                                                   rai
         nfall
                 label
         0
                90
                   42
                        43
                               20.879744
                                          82.002744
                                                     6.502985
                                                                 202.935536
                                                                                rice
         1
                85
                    58
                        41
                               21.770462
                                          80.319644
                                                     7.038096
                                                                 226.655537
                                                                               rice
         2
                60
                    55
                        44
                               23.004459
                                          82.320763
                                                     7.840207
                                                                 263.964248
                                                                               rice
         3
                74
                    35
                        40
                               26.491096
                                          80.158363
                                                      6.980401
                                                                 242.864034
                                                                                rice
         4
                78
                    42
                        42
                               20.130175
                                          81.604873
                                                      7.628473
                                                                 262.717340
                                                                                rice
         . . .
               . . .
                         . .
                                     . . .
                                                 . . .
                                                            . . .
                                                      6.780064
         2195
               107
                    34
                        32
                               26.774637
                                          66.413269
                                                                 177.774507
                                                                             coffee
                99
                   15
                        27
                               27.417112
                                                      6.086922
                                                                             coffee
         2196
                                          56.636362
                                                                 127.924610
         2197
               118
                    33
                        30
                               24.131797
                                          67.225123
                                                      6.362608
                                                                173.322839
         2198
               117
                    32
                        34
                               26.272418
                                          52.127394
                                                      6.758793
                                                                127.175293
                                                                             coffee
         2199
               104
                    18
                        30
                               23.603016 60.396475
                                                      6.779833 140.937041
                                                                             coffee
         [2200 rows x \in columns]>
         #check the missing values
In [8]:
          data.isnull().sum()
                        0
Out[8]: N
                         0
                         0
         temperature
                         0
         humidity
                         0
         ph
                         0
         rainfall
                         0
         label
                         0
         dtype: int64
In [9]:
         #check the crops present in the dataset
          data['label'].value_counts()
Out[9]: kidneybeans
                         100
                         100
         papaya
                         100
         banana
         mango
                         100
                         100
         rice
                        100
         blackgram
```

```
chickpea
                      100
                      100
        orange
        grapes
                      100
                      100
        cotton
                      100
        apple
                      100
        muskmelon
        mothbeans
                      100
        pomegranate
                      100
                      100
        mungbean
                      100
        jute
                      100
        maize
        coconut
                      100
                      100
        watermelon
                      100
        lentil
        pigeonpeas
                      100
        coffee
                      100
        Name: label, dtype: int64
         # The Average condition of all crops
In [10]:
         print("Average ratio of the nitrogen in the soil: {0:2f}".format(data['N'].mean()))
         print("Average ratio of the Phosphorous in the soil: {0:2f}".format(data['P'].mean()))
         print("Average ratio of the Potassium in the soil: {0:2f}".format(data['K'].mean()))
         print("Average Temperature in celsius: {0:2f}".format(data['temperature'].mean()))
         print("Average Relative Humidity in the soil: {0:2f}".format(data['humidity'].mean()))
         print("Average PH value in the soil: {0:2f}".format(data['ph'].mean()))
         print("Average rainfall in mm: {0:2f}".format(data['rainfall'].mean()))
        Average ratio of the nitrogen in the soil: 50.551818
        Average ratio of the Phosphorous in the soil: 53.362727
        Average ratio of the Potassium in the soil: 48.149091
        Average Temperature in celsius: 25.616244
        Average Relative Humidity in the soil: 71.481779
        Average PH value in the soil: 6.469480
        Average rainfall in mm: 103.463655
In [11]:
         #check the summary statistics for each of the crops
         @interact
         def summary(crops=list(data['label'].value counts().index)):
             x=data[data['label']==crops]
             print("----")
             print("Statics for Nitrogen")
             print("Minimum Nitrogen required :",x['N'].min())
             print("Average Nitrogen required :",x['N'].mean())
             print("Maximum Notrogen required :",x['N'].max())
             print("-----")
             print("Statics for Phosphorous")
             print("Minimum phosphorous required :",x['P'].min())
             print("Average phosphorous required :",x['P'].mean())
             print("Maximum phosphorous required :",x['P'].max())
             print("-----")
             print("Statics for Potassium")
             print("Minimum Potassium required :{0:2f}",x['K'].min())
             print("Average Potassium required :{0:2f}",x['K'].mean())
             print("Maximum Potassium required :{0:2f}",x['K'].max())
             print("----")
             print("Statics for Temperature")
             print("Minimum Temperature required :{0:2f}",x['temperature'].min())
             print("Average Temperature required :{0:2f}",x['temperature'].mean())
```

```
#compare the average requirement for each crops with average condition
In [12]:
          @interact
          def compare(conditions =['N','P','K','temperature','ph','humidity','rainfall']):
              print("Average value for",conditions,"is{0:2f}".format(data[conditions].mean()))
              print("Rice :{0:2f}".format(data[(data['label']=='rice')][conditions].mean()))
              print("Black Grams :{0:2f}".format(data['data['label']=='blackgrams')][conditions].
              print("Banana :{0:2f}".format(data[(data['label']=='banana')][conditions].mean()))
              print("Jute :{0:2f}".format(data[(data['label']=='jute')][conditions].mean()))
              print("Coconut :{0:2f}".format(data['label']=='coconut')][conditions].mean())
              print("Apple :{0:2f}".format(data[(data['label']=='apple')][conditions].mean()))
              print("Papaya :{0:2f}".format(data[(data['label']=='papaye')][conditions].mean()))
              print("Muskmelon :{0:2f}".format(data[(data['label']=='muskmelon')][conditions].mea
              print("Grapes :{0:2f}".format(data[(data['label']=='grapes')][conditions].mean()))
              print("Watermelon :{0:2f}".format(data[(data['label']=='watermelon')][conditions].m
              print("Kidney Beans :{0:2f}".format(data[(data['label']=='kidneybeans'))][conditions
              print("Mung Beans :{0:2f}".format(data[(data['label']=='mungbeans'))][conditions].me
              print("Oranges :{0:2f}".format(data[(data['label']=='oranges')][conditions].mean())
              print("Chick peas :{0:2f}".format(data[(data['label']=='chickpeas'))][conditions].me
              print("Lentil :{0:2f}".format(data[(data['label']=='lentil')][conditions].mean()))
              print("Cotton :{0:2f}".format(data[(data['label']=='cotton')][conditions].mean()))
              print("Maize :{0:2f}".format(data[(data['label']=='maize')][conditions].mean()))
              print("Moth Beans :{0:2f}".format(data[(data['label']=='moth beans')][conditions].m
              print("Pigeon Peas :{0:2f}".format(data['label']=='pigeonpeas')][conditions].
              print("Mango :{0:2f}".format(data[(data['label']=='mango')][conditions].mean()))
              print("Pomegranate :{0:2f}".format(data[(data['label']=='pomegranate')][conditions]
              print("Coffee :{0:2f}".format(data[(data['label']=='coffee')][conditions].mean()))
```

```
#Makes this function more intuitive
@interact
def compare(conditions =['N','P','K','temperature','ph','humidity','rainfall']):
    print("crops which require greater than average",conditions,'\n')
    print(data[data[conditions]>data[conditions].mean()]['label'].unique())
    print("crops which require less than average",'\n')
    print(data[data[conditions]<=data[conditions].mean()]['label'].unique())</pre>
```

```
In [32]:
         #Some interesting facts
          print("----")
          print("crops which require very high ratio of Nitrogen content in soil:",data[data['N']
          print("crops which require very high ratio of Phosphorous content in soil:",data[data['
          print("crops which require very high ratio of Potassium content in soil:",data[data['K'
          print("crops which require very high Rainfall:",data[data['rainfall']>200]['label'].uni
          print("crops which require very low Temperature:",data[data['temperature']<10]['label']</pre>
          print("crops which require very high Temperature:",data[data['temperature']>40]['label'
          print("crops which require very Low Humidity:",data[data['humidity']<20]['label'].uniqu</pre>
          print("crops which require very Low pH:",data[data['ph']<4]['label'].unique())</pre>
          print("crops which require very high pH:",data[data['ph']>9]['label'].unique())
         crops which require very high ratio of Nitrogen content in soil: ['cotton']
         crops which require very high ratio of Phosphorous content in soil: ['grapes' 'apple']
         crops which require very high ratio of Potassium content in soil: ['grapes' 'apple']
         crops which require very high Rainfall: ['rice' 'papaya' 'coconut']
         crops which require very low Temperature: ['grapes']
         crops which require very high Temperature: ['grapes' 'papaya']
         crops which require very Low Humidity: ['chickpea' 'kidneybeans']
         crops which require very Low pH: ['mothbeans']
         crops which require very high pH: ['mothbeans']
         #crops which Grown in summer season, winter season and Rainy season
In [38]:
          print("Summer crops")
          print(data[(data['temperature']>30)&(data['humidity']>50)]['label'].unique())
          print("----")
          print("winter crops")
          print(data[(data['temperature']<20)&(data['humidity']>30)]['label'].unique())
          print("----")
          print("Rainy crops")
          print(data[(data['rainfall']>200)&(data['humidity']>30)]['label'].unique())
         Summer crops
         ['pigeonpeas' 'mothbeans' 'blackgram' 'mango' 'grapes' 'orange' 'papaya']
         ['maize' 'pigeonpeas' 'lentil' 'pomegranate' 'grapes' 'orange']
         Rainy crops
         ['rice' 'papaya' 'coconut']
In [14]: | #split of data set for predective modelling
          y=data['label']
          x=data.drop(['label'],axis=1)
          print("shape of x:",x.shape)
          print("shape of y:",y.shape)
         shape of x: (2200, 7)
         shape of y: (2200,)
         #creating traning and testing sets for validation of result
In [18]:
          from sklearn.model_selection import train_test_split
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=0)
          print("The shape of x train:",x_train.shape)
```

```
print("The shape of x test:",x_test.shape)
          print("The shape of y train:",x_train.shape)
          print("The shape of y test:",x_test.shape)
         The shape of x train: (1760, 7)
         The shape of x test: (440, 7)
         The shape of y train: (1760, 7)
         The shape of y test: (440, 7)
In [19]: | #Lets create a predective model
          from sklearn.linear_model import logisticRegression
          model=logisticRegression()
          model.fit(x_train,y_train)
          y_pred=model.predict(x_test)
                                                    Traceback (most recent call last)
         ImportError
         <ipython-input-19-f3148d4f6f65> in <module>
               1 #lets create a predective model
          ----> 3 from sklearn.linear model import logisticRegression
               5 model=logisticRegression()
         ImportError: cannot import name 'logisticRegression' from 'sklearn.linear model' (C:\pyt
         hon anaconda mohit\lib\site-packages\sklearn\linear model\ init .py)
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