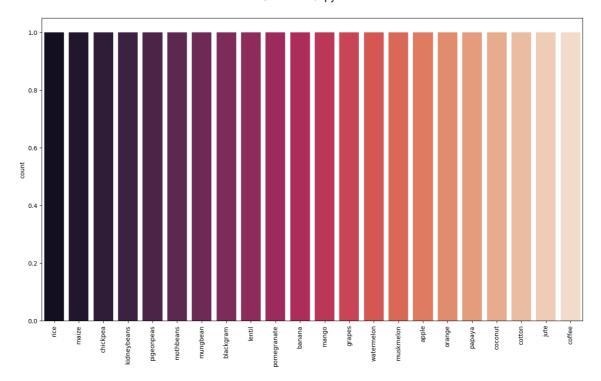
```
import numpy as np
import pandas as pd
from sklearn.neighbors import KNeighborsClassifier
from matplotlib import pyplot as plt
from scipy.interpolate import make_interp_spline
from pylab import rcParams
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy_score as acc
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
import seaborn as sns

import warnings
warnings.filterwarnings('ignore')
```

```
data = pd.read_csv("C:/Users/saiteja143/Downloads/Crop_recommendation.csv")
In [55]:
         col = list(data.columns)
         classes = data["label"].unique()
         from sklearn.preprocessing import LabelEncoder
         le=LabelEncoder()
         trans=['N','P','K','temperature','humidity','ph','rainfall']
         for i in trans:
             data[i]=le.fit_transform(data[i])
         data
         xdata = data.iloc[:, 0:7].values
         ydata = data.iloc[:, 7].values
         plt.figure(figsize = (16, 9))
         sns.countplot(classes, palette = 'rocket')
         plt.xticks(rotation=90)
Out[55]: (array([ 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 1
         6,
                  17, 18, 19, 20, 21]),
           [Text(0, 0, 'rice'),
           Text(1, 0, 'maize'),
           Text(2, 0, 'chickpea'),
           Text(3, 0, 'kidneybeans'),
           Text(4, 0, 'pigeonpeas'),
           Text(5, 0, 'mothbeans'),
           Text(6, 0, 'mungbean'),
           Text(7, 0, 'blackgram'),
           Text(8, 0, 'lentil'),
           Text(9, 0, 'pomegranate'),
           Text(10, 0, 'banana'),
           Text(11, 0, 'mango'),
           Text(12, 0, 'grapes'),
           Text(13, 0, 'watermelon'),
           Text(14, 0, 'muskmelon'),
           Text(15, 0, 'apple'),
           Text(16, 0, 'orange'),
           Text(17, 0, 'papaya'),
           Text(18, 0, 'coconut'),
           Text(19, 0, 'cotton'),
           Text(20, 0, 'jute'),
Text(21, 0, 'coffee')])
```



In [56]: data

## Out[56]:

	N	P	K	temperature	humidity	ph	rainfall	label
0	90	37	38	360	1234	1211	2074	rice
1	85	53	36	436	1085	1753	2128	rice
2	60	50	39	582	1262	2117	2174	rice
3	74	30	35	1278	1067	1710	2153	rice
4	78	37	37	306	1201	2055	2171	rice
2195	107	29	27	1327	763	1523	1920	coffee
2196	99	10	22	1441	483	666	1666	coffee
2197	118	28	25	788	793	1012	1895	coffee
2198	117	27	29	1236	387	1493	1663	coffee
2199	104	13	25	693	555	1521	1723	coffee

2200 rows × 8 columns

In [57]: data.shape

Out[57]: (2200, 8)

## In [58]: data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2200 entries, 0 to 2199
Data columns (total 8 columns):

#	Column	Non-Null Count	Dtype
0	N	2200 non-null	int64
1	P	2200 non-null	int64
2	K	2200 non-null	int64
3	temperature	2200 non-null	int64
4	humidity	2200 non-null	int64
5	ph	2200 non-null	int64
6	rainfall	2200 non-null	int64
7	label	2200 non-null	object

dtypes: int64(7), object(1)
memory usage: 137.6+ KB

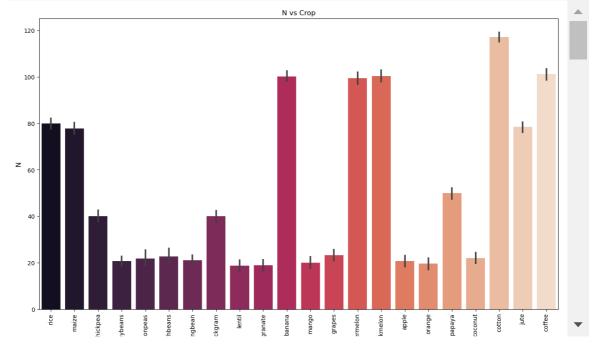
## In [59]: data.describe()

## Out[59]:

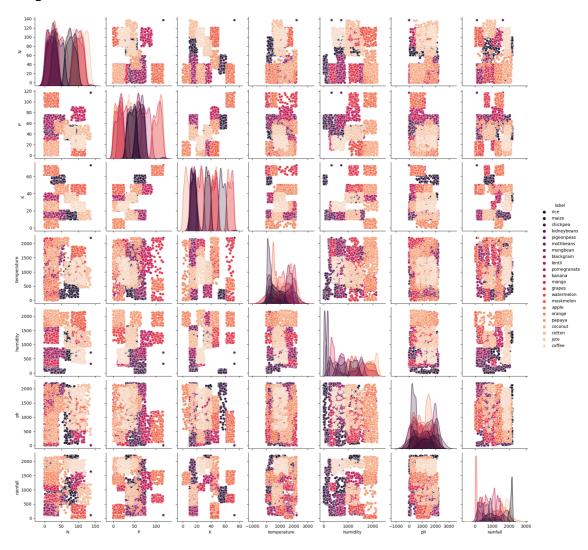
r	ph	humidity	temperature	K	Р	N	
2200.0	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000	count
1099.5	1099.500000	1099.500000	1099.499545	30.700909	46.241818	50.630000	mean
635.2	635.229617	635.229617	635.228831	18.065758	28.122979	36.927241	std
0.0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	min
549.7	549.750000	549.750000	549.750000	15.000000	23.000000	21.000000	25%
1099.5	1099.500000	1099.500000	1099.500000	27.000000	46.000000	37.000000	50%
1649.2	1649.250000	1649.250000	1649.250000	44.000000	63.000000	85.000000	75%
2199.0	2199.000000	2199.000000	2198.000000	73.000000	117.000000	137.000000	max
							4

```
In [60]: all_col = data.columns[:-1]

for col in all_col:
    plt.figure(figsize = (16, 9))
    sns.barplot(x = 'label', y = col, data = data, palette = 'rocket')
    plt.xlabel('label', fontsize = 12)
    plt.ylabel(col, fontsize = 12)
    plt.xticks(rotation=90)
    plt.title(f'{col} vs Crop')
    plt.show()
```

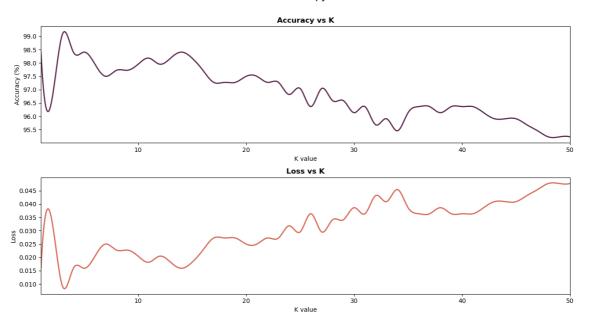


<Figure size 1000x1700 with 0 Axes>



```
In [62]:
    xtrain, xtest, ytrain, ytest = train_test_split(xdata, ydata, test_size=0.2
    x_st = StandardScaler()
    xtrain = x_st.fit_transform(xtrain)
    xtest = x_st.fit_transform(xtest)
```

```
acc_list = []
In [63]:
         err_rate = []
         neighbors = np.linspace(1, 50, 50)
         neighbors = neighbors.astype(int)
         for K in neighbors:
           classifier = KNeighborsClassifier(n_neighbors = K)
           classifier.fit(xtrain, ytrain)
           y_pred = classifier.predict(xtest)
           accuracy = round(acc(ytest, y_pred)*100, 3)
           acc_list.append(accuracy)
           err_rate.append(np.mean(y_pred != ytest))
         xy = make_interp_spline(neighbors, acc_list)
         xz = make_interp_spline(neighbors, err_rate)
         x = np.linspace(1, 50, 1000)
         y = xy(x)
         z = xz(x)
         plt.figure(figsize = (13, 7))
         plt.subplot(2, 1, 1)
         sns.lineplot(x, y, linewidth = 2, color = '#5C284F')
         plt.xlabel('K value')
         plt.ylabel('Accuracy (%)')
         plt.title('Accuracy vs K', fontweight = 'bold')
         plt.xlim(min(neighbors), max(neighbors))
         plt.subplot(2, 1, 2)
         sns.lineplot(x, z, linewidth = 2, color = '#D96856')
         plt.xlabel('K value')
         plt.ylabel('Loss')
         plt.title('Loss vs K', fontweight = 'bold')
         plt.xlim(min(neighbors), max(neighbors))
         plt.tight_layout()
         plt.show()
         K opt = acc list.index(max(acc list))
         print('\nOptimal value of K = ', K_opt)
```



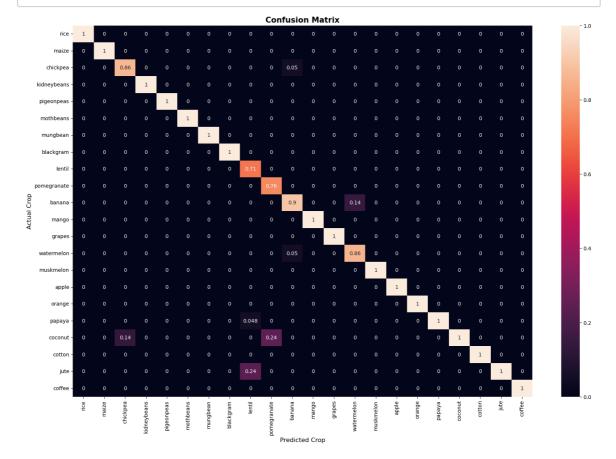
Optimal value of K = 2

```
In [64]: cm = confusion_matrix(ytest, y_pred, normalize = 'pred')

fig, ax = plt.subplots(figsize=(20,13))
sns.heatmap(cm, annot = True)
plt.xlabel('Predicted Crop', fontsize = 12)
plt.ylabel('Actual Crop', fontsize = 12)
plt.title('Confusion Matrix', fontweight = 'bold', fontsize = 15)

plt.xticks(rotation=90)
plt.yticks(rotation=0)

ax.xaxis.set_ticklabels(classes)
ax.yaxis.set_ticklabels(classes)
plt.show()
```



In [65]: y\_pred

Out[65]: array(['muskmelon', 'muskmelon', 'coffee', 'muskmelon', 'kidneybeans', 'pomegranate', 'maize', 'coffee', 'jute', 'cotton', 'muskmelon', 'mungbean', 'grapes', 'watermelon', 'blackgram', 'banana', 'watermelon', 'grapes', 'grapes', 'coffee', 'grapes', 'kidneybeans', 'coffee', 'maize', 'grapes', 'mothbeans', 'pigeonpeas', 'papaya', 'cotton', 'rice', 'watermelon', 'banana', 'muskmelon', 'blackgram', 'blackgram', 'coconut', 'coconut', 'pomegranate', 'blackgram', 'mango', 'jute', 'orange', 'muskmelon', 'blackgram', 'chickpea', 'kidneybeans', 'chickpea', 'coconut', 'pomegranate', 'coffee', 'mango', 'lentil', 'mango', 'mango', 'maize', 'orange', 'kidneybeans', 'chickpea', 'rice', 'pigeonpeas', 'rice', 'chickpea', 'jute', 'chickpea', 'pigeonpeas', 'muskmelon', 'pomegranate', 'blackgram', 'coconut', 'blackgram', 'grapes', 'mungbean', 'papaya', 'blackgram', 'jute', 'cotton', 'cotton', 'maize', 'pigeonpeas', 'coffee', 'banana', 'maize', 'jute', 'banana', 'rice', 'muskmelon', 'mango', 'pomegranate', 'chickpea', 'maize', 'jute', 'mothbeans', 'lentil', 'mango', 'coconut', 'maize', 'jute', 'apple', 'maize', 'rice', 'coffee', 'maize', 'jute', 'blackgram', 'orange', 'kidneybeans', 'banana', 'grapes', 'coconut', 'kidneybeans', 'papaya', 'coffee', 'coconut', 'kidneybeans', 'coffee', 'apple', 'cotton', 'blackgram', 'kidneybeans', 'cotton', 'kidneybeans', 'mango', 'coconut', 'lentil', 'lentil', 'papaya', 'banana', 'maize', 'cotton', 'banana', 'coffee', 'kidneybeans', 'cotton', 'kidneybeans', 'coconut', 'apple', 'grapes', 'lentil', 'banana', 'muskmelon', 'orange', 'kidneybeans', 'pomegranate', 'mango', 'muskmelon', 'banana', 'chickpea', 'banana', 'grapes', 'rice', 'pomegranate', 'kidneybeans', 'lentil', 'kidneybeans', 'apple', 'jute', 'banana', 'jute', 'rice', 'cotton', 'mungbean', 'pomegranate', 'watermelon', 'watermelon', 'chickpea', 'grapes', 'jute', 'coffee', 'watermelon', 'watermelon', 'kidneybeans', 'mungbean', 'lentil', 'grapes', 'rice', 'chickpea', 'banana', 'watermelon', 'pomegranate', 'orange', 'banana', 'chickpea', 'apple', 'kidneybeans', 'maize', 'kidneybeans', 'mungbean', 'banana', 'apple', 'cotton', 'maize', 'mungbean', 'banana', 'rice', 'kidneybeans', 'chickpea', 'grapes', 'blackgram', 'watermelon', 'pomegranate', 'mango', 'mungbean', 'cotton', 'coconut', 'blackgram', 'chickpea', 'jute', 'blackgram', 'cotton', 'watermelon', 'coconut', 'mango', 'lentil', 'banana', 'maize', 'cotton', 'lentil', 'chickpea', 'papaya', 'mothbeans', 'maize', 'banana', 'muskmelon', 'pomegranate', 'cotton', 'watermelon', 'coffee', 'orange', 'lentil', 'coffee', 'muskmelon', 'pomegranate', 'mothbeans', 'chickpea', 'apple', 'chickpea', 'pigeonpeas', 'papaya', 'blackgram', 'banana', 'muskmelon', 'watermelon', 'chickpea', 'pigeonpeas', 'grapes', 'banana', 'orange', 'apple', 'mothbeans', 'grapes', 'grapes', 'rice', 'mothbeans', 'orange', 'chickpea', 'orange', 'kidneybeans', 'chickpea', 'rice', 'maize', 'muskmelon', 'grapes', 'coconut', 'lentil', 'chickpea', 'lentil', 'muskmelon', 'jute', 'coffee', 'banana', 'cotton', 'maize', 'apple', 'grapes', 'cotton', 'grapes', 'mango', 'mungbean', 'muskmelon', 'blackgram', 'coconut', 'banana', 'kidneybeans', 'jute', 'blackgram', 'cotton', 'mango', 'jute', 'cotton', 'watermelon', 'mango', 'mothbeans', 'orange', 'maize', 'chickpea', 'cotton', 'cotton', 'mungbean', 'mothbeans', 'kidneybeans', 'blackgram', 'lentil', 'muskmelon', 'papaya', 'pomegranate', 'mungbean', 'muskmelon', 'maize', 'coconut', 'banana', 'pigeonpeas', 'mothbeans', 'mungbean', 'rice', 'coconut', 'mothbeans', 'banana', 'chickpea', 'kidneybeans', 'kidneybeans', 'coffee', 'papaya', 'lentil', 'chickpea', 'orange', 'kidneybeans', 'mango', 'papaya', 'lentil', 'papaya', 'kidneybeans', 'blackgram', 'cotton', 'mungbean', 'muskmelon', 'cotton', 'coffee', 'banana', 'muskmelon', 'apple', 'apple', 'maize', 'kidneybeans',

'pigeonpeas', 'coffee', 'lentil', 'mango', 'pomegranate', 'papaya', 'rice', 'blackgram', 'coconut', 'muskmelon', 'mango', 'lentil', 'grapes', 'watermelon', 'banana', 'pigeonpeas', 'coconut', 'mango', 'lentil', 'pomegranate', 'grapes', 'coffee', 'mothbeans', 'apple', 'blackgram', 'cotton', 'pigeonpeas', 'kidneybeans', 'coconut', 'jute', 'lentil', 'banana', 'pigeonpeas', 'chickpea', 'kidneybeans', 'apple', 'coffee', 'chickpea', 'banana', 'orange', 'mungbean', 'kidneybeans', 'jute', 'jute', 'cotton', 'kidneybeans', 'papaya', 'mango', 'blackgram', 'maize', 'papaya', 'chickpea', 'rice', 'apple', 'banana', 'mungbean', 'mothbeans', 'coconut', 'mungbean', 'muskmelon', 'mothbeans', 'mungbean', 'jute', 'banana', 'mango', 'mungbean', 'jute', 'kidneybeans', 'watermelon', 'rice', 'cotton', 'rice', 'lentil', 'kidneybeans', 'papaya', 'coconut', 'pomegranate', 'mothbeans', 'muskmelon', 'kidneybeans', 'watermelon', 'coffee', 'muskmelon', 'orange', 'banana', 'jute', 'mango', 'mungbean', 'maize', 'coconut', 'coconut', 'mango'], 'dtype=object)

In [66]: ytest

Out[66]: array(['muskmelon', 'muskmelon', 'coffee', 'muskmelon', 'kidneybeans', 'pomegranate', 'maize', 'coffee', 'jute', 'cotton', 'muskmelon', 'mungbean', 'grapes', 'watermelon', 'blackgram', 'banana', 'watermelon', 'grapes', 'grapes', 'coffee', 'grapes', 'kidneybeans', 'coffee', 'maize', 'grapes', 'mothbeans', 'pigeonpeas', 'papaya', 'cotton', 'rice', 'watermelon', 'banana', 'muskmelon', 'blackgram', 'blackgram', 'coconut', 'coconut', 'pomegranate', 'pigeonpeas', 'mango', 'rice', 'orange', 'muskmelon', 'blackgram', 'chickpea', 'kidneybeans', 'chickpea', 'coconut', 'pomegranate', 'coffee', 'mango', 'lentil', 'mango', 'mango', 'mango', 'mango', 'mango', ' 'mango', 'maize', 'orange', 'kidneybeans', 'chickpea', 'rice', 'pigeonpeas', 'rice', 'chickpea', 'jute', 'chickpea', 'pigeonpeas', 'muskmelon', 'pomegranate', 'blackgram', 'coconut', 'pigeonpeas', 'grapes', 'mungbean', 'papaya', 'blackgram', 'jute', 'cotton', 'cotton', 'maize', 'pigeonpeas', 'coffee', 'banana', 'maize', 'jute', 'banana', 'rice', 'muskmelon', 'mango', 'pomegranate', 'chickpea', 'maize', 'jute', 'mothbeans', 'lentil', 'mango', 'coconut', 'maize', 'jute', 'apple', 'maize', 'rice', 'coffee', 'maize', 'papaya', 'blackgram', 'orange', 'kidneybeans', 'banana',
'grapes', 'coconut', 'pigeonpeas', 'papaya', 'coffee', 'coconut', 'kidneybeans', 'coffee', 'apple', 'cotton', 'blackgram', 'pigeonpeas', 'cotton', 'kidneybeans', 'mango', 'coconut', 'lentil', 'lentil', 'papaya', 'banana', 'maize', 'cotton', 'banana', 'coffee', 'kidneybeans', 'cotton', 'kidneybeans', 'coconut', 'apple', 'grapes', 'lentil', 'banana', 'muskmelon', 'orange', 'kidneybeans', 'pomegranate', 'mango', 'muskmelon', 'banana', 'chickpea', 'banana', 'grapes', 'rice', 'pomegranate', 'kidneybeans', 'lentil', 'kidneybeans', 'apple', 'jute', 'banana', 'jute', 'rice', 'cotton', 'mungbean', 'pomegranate', 'watermelon', 'watermelon', 'chickpea', 'grapes', 'jute', 'coffee', 'watermelon', 'watermelon', 'kidneybeans', 'mungbean', 'lentil', 'grapes', 'rice', 'chickpea', 'banana', 'watermelon', 'pomegranate', 'orange', 'banana', 'chickpea', 'apple', 'kidneybeans', 'maize', 'kidneybeans', 'mungbean', 'banana', 'apple', 'cotton', 'maize', 'mungbean', 'banana', 'rice', 'pigeonpeas', 'chickpea', 'grapes', 'blackgram', 'watermelon', 'pomegranate', 'mango', 'mungbean', 'cotton', 'coconut', 'blackgram', 'chickpea', 'rice', 'blackgram', 'cotton', 'watermelon', 'coconut', 'mango', 'lentil', 'banana', 'maize', 'cotton', 'lentil', 'chickpea', 'papaya', 'lentil', 'maize', 'banana', 'muskmelon', 'pomegranate', 'cotton', 'watermelon', 'coffee', 'orange', 'lentil', 'coffee', 'muskmelon', 'pomegranate', 'lentil', 'chickpea', 'apple', 'chickpea', 'pigeonpeas', 'papaya', 'blackgram', 'banana', 'muskmelon', 'watermelon', 'chickpea', 'pigeonpeas', 'grapes', 'banana', 'orange', 'apple', 'mothbeans', 'grapes', 'grapes', 'rice', 'mothbeans', 'orange', 'chickpea', 'orange', 'kidneybeans', 'chickpea', 'rice', 'maize', 'muskmelon', 'grapes', 'coconut', 'blackgram', 'chickpea', 'lentil', 'muskmelon', 'jute', 'coffee', 'banana', 'cotton', 'maize', 'apple', 'grapes', 'cotton', 'grapes', 'mango', 'mungbean', 'muskmelon', 'blackgram', 'coconut', 'banana', 'pigeonpeas', 'jute', 'blackgram', 'cotton', 'mango', 'jute', 'cotton', 'watermelon', 'mango', 'mothbeans', 'orange', 'maize', 'chickpea', 'cotton', 'cotton', 'mungbean', 'mothbeans', 'kidneybeans', 'blackgram', 'lentil', 'muskmelon', 'papaya', 'pomegranate', 'mungbean', 'muskmelon', 'maize', 'coconut', 'banana', 'pigeonpeas', 'mothbeans', 'mungbean', 'rice', 'coconut', 'mothbeans', 'banana', 'chickpea', 'pigeonpeas', 'kidneybeans', 'coffee', 'papaya', 'lentil', 'chickpea', 'orange', 'kidneybeans', 'mango', 'papaya', 'lentil', 'papaya', 'kidneybeans', 'blackgram', 'cotton', 'mungbean', 'muskmelon', 'cotton', 'coffee', 'banana', 'muskmelon', 'apple', 'apple', 'maize', 'pigeonpeas', 'pigeonpeas',

'coffee', 'lentil', 'mango', 'pomegranate', 'papaya', 'rice',
'blackgram', 'coconut', 'muskmelon', 'mango', 'mothbeans',
'grapes', 'watermelon', 'banana', 'pigeonpeas', 'coconut', 'mango',
'lentil', 'pomegranate', 'grapes', 'coffee', 'mothbeans', 'apple',
'pigeonpeas', 'cotton', 'pigeonpeas', 'chickpea',
'kidneybeans', 'apple', 'coffee', 'chickpea', 'banana', 'orange',
'mungbean', 'kidneybeans', 'jute', 'rice', 'cotton', 'kidneybeans',
'papaya', 'mango', 'blackgram', 'maize', 'papaya', 'chickpea',
'rice', 'apple', 'banana', 'mungbean', 'mothbeans', 'coconut',
'mungbean', 'muskmelon', 'mothbeans', 'mungbean', 'blackgram',
'watermelon', 'mungbean', 'jute', 'kidneybeans', 'watermelon',
'rice', 'cotton', 'rice', 'lentil', 'kidneybeans', 'papaya',
'coconut', 'pomegranate', 'mothbeans', 'muskmelon', 'pigeonpeas',
'watermelon', 'coffee', 'muskmelon', 'orange', 'banana', 'rice',
'mango', 'mungbean', 'maize', 'coconut', 'coconut', 'mango'],
dtype=object)

```
In [67]: classifier = KNeighborsClassifier(n_neighbors=K_opt+1)
    classifier.fit(xtrain, ytrain)
    y_pred = classifier.predict(xtest)

accuracy = acc(ytest, y_pred)*100
    print('Accuracy of the training Model : ', round(accuracy, 3), '%')
```

Accuracy of the training Model: 99.091 %

```
In [68]: from sklearn.ensemble import RandomForestClassifier
    rfc = RandomForestClassifier(n_estimators=200)
    rfc.fit(xtrain, ytrain)
```

Out[68]: RandomForestClassifier(n estimators=200)

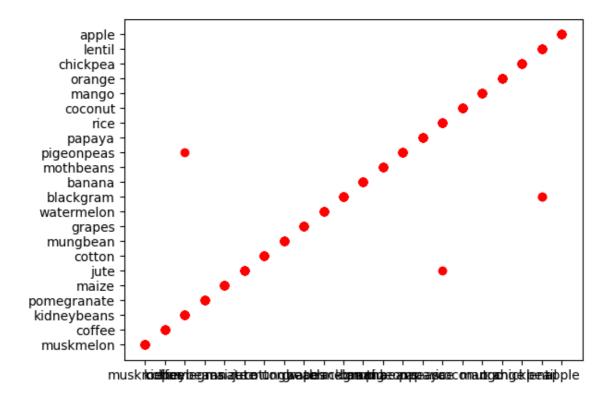
In [69]: s=rfc.predict(xtest)
s

Out[69]: array(['muskmelon', 'muskmelon', 'coffee', 'muskmelon', 'pigeonpeas', 'pomegranate', 'maize', 'coffee', 'jute', 'cotton', 'muskmelon', 'mungbean', 'grapes', 'watermelon', 'blackgram', 'banana', 'watermelon', 'grapes', 'grapes', 'coffee', 'grapes', 'kidneybeans', 'coffee', 'maize', 'grapes', 'mothbeans', 'pigeonpeas', 'papaya', 'cotton', 'rice', 'watermelon', 'banana', 'muskmelon', 'blackgram', 'blackgram', 'coconut', 'coconut', 'pomegranate', 'pigeonpeas', 'mango', 'rice', 'orange', 'muskmelon', 'blackgram', 'chickpea', 'kidneybeans', 'chickpea', 'coconut', 'pomegranate', 'coffee', 'mango', 'lentil', 'mango', 'mango', 'maize', 'orange', 'pigeonpeas', 'chickpea', 'rice', 'pigeonpeas', 'rice', 'chickpea', 'jute', 'chickpea', 'pigeonpeas', 'muskmelon', 'pomegranate', 'blackgram', 'coconut', 'pigeonpeas', 'grapes', 'mungbean', 'papaya', 'blackgram', 'jute', 'cotton', 'cotton', 'maize', 'pigeonpeas', 'coffee', 'banana', 'maize', 'jute', 'banana', 'rice', 'muskmelon', 'mango', 'pomegranate', 'chickpea', 'maize', 'jute', 'mothbeans', 'lentil', 'mango', 'coconut', 'maize', 'jute', 'apple', 'maize', 'rice', 'coffee', 'maize', 'papaya', 'blackgram', 'orange', 'kidneybeans', 'banana', 'grapes', 'coconut', 'pigeonpeas', 'papaya', 'coffee', 'coconut', 'pigeonpeas', 'coffee', 'apple', 'cotton', 'blackgram', 'pigeonpeas', 'cotton', 'pigeonpeas', 'mango', 'coconut', 'lentil', 'lentil', 'papaya', 'banana', 'cotton', 'cotton', 'banana', 'coffee', 'pigeonpeas', 'cotton', 'kidneybeans', 'coconut', 'apple', 'grapes', 'lentil', 'banana', 'muskmelon', 'orange', 'kidneybeans', 'pomegranate', 'mango', 'muskmelon', 'banana', 'chickpea', 'banana', 'grapes', 'rice', 'pomegranate', 'kidneybeans', 'lentil', 'pigeonpeas', 'apple', 'jute', 'banana', 'papaya', 'rice', 'cotton', 'mungbean', 'pomegranate',
'watermelon', 'watermelon', 'chickpea', 'grapes', 'jute', 'coffee',
'watermelon', 'watermelon', 'kidneybeans', 'mungbean', 'lentil', 'grapes', 'rice', 'chickpea', 'banana', 'watermelon', 'pomegranate', 'orange', 'banana', 'chickpea', 'apple', 'kidneybeans', 'maize', 'kidneybeans', 'mungbean', 'banana', 'apple', 'cotton', 'maize', 'mungbean', 'banana', 'rice', 'pigeonpeas', 'chickpea', 'grapes', 'blackgram', 'watermelon', 'pomegranate', 'mango', 'mungbean', 'cotton', 'coconut', 'blackgram', 'chickpea', 'rice', 'blackgram', 'cotton', 'watermelon', 'coconut', 'mango', 'lentil', 'banana', 'maize', 'cotton', 'lentil', 'chickpea', 'papaya', 'lentil', 'maize', 'banana', 'muskmelon', 'pomegranate', 'cotton', 'watermelon', 'coffee', 'orange', 'lentil', 'coffee', 'muskmelon', 'pomegranate', 'lentil', 'chickpea', 'apple', 'chickpea', 'pigeonpeas', 'papaya', 'blackgram', 'banana', 'muskmelon', 'watermelon', 'chickpea', 'pigeonpeas', 'grapes', 'banana', 'orange', 'apple', 'mothbeans', 'grapes', 'grapes', 'rice', 'mothbeans', 'orange', 'chickpea', 'orange', 'kidneybeans', 'chickpea', 'rice', 'maize', 'muskmelon', 'grapes', 'coconut', 'blackgram', 'chickpea', 'lentil', 'muskmelon', 'jute', 'coffee', 'banana', 'cotton', 'maize', 'apple', 'grapes', 'cotton', 'grapes', 'mango', 'mungbean', 'muskmelon', 'blackgram', 'coconut', 'banana', 'pigeonpeas',
'jute', 'blackgram', 'cotton', 'mango', 'jute', 'cotton', 'watermelon', 'mango', 'mothbeans', 'orange', 'maize', 'chickpea', 'cotton', 'cotton', 'mungbean', 'mothbeans', 'pigeonpeas', 'blackgram', 'lentil', 'muskmelon', 'papaya', 'pomegranate', 'mungbean', 'muskmelon', 'maize', 'coconut', 'banana', 'pigeonpeas', 'mothbeans', 'mungbean', 'rice', 'coconut', 'mothbeans', 'banana', 'chickpea', 'pigeonpeas', 'kidneybeans', 'coffee', 'papaya', 'lentil', 'chickpea', 'orange', 'pigeonpeas', 'mango', 'papaya', 'lentil', 'papaya', 'pigeonpeas', 'blackgram', 'cotton', 'mungbean', 'muskmelon', 'cotton', 'coffee', 'banana',

'muskmelon', 'apple', 'apple', 'maize', 'pigeonpeas', 'pigeonpeas',
'coffee', 'lentil', 'mango', 'pomegranate', 'papaya', 'rice',
'blackgram', 'coconut', 'muskmelon', 'mango', 'mothbeans',
'grapes', 'watermelon', 'banana', 'pigeonpeas', 'coconut', 'mango',
'lentil', 'pomegranate', 'grapes', 'coffee', 'mothbeans', 'apple',
'pigeonpeas', 'cotton', 'pigeonpeas', 'chickpea',
'kidneybeans', 'apple', 'coffee', 'chickpea', 'banana', 'orange',
'mungbean', 'kidneybeans', 'jute', 'rice', 'cotton', 'kidneybeans',
'papaya', 'mango', 'blackgram', 'maize', 'papaya', 'chickpea',
'rice', 'apple', 'banana', 'mungbean', 'mothbeans', 'coconut',
'mungbean', 'muskmelon', 'mothbeans', 'mungbean', 'rice',
'banana', 'mango', 'mungbean', 'jute', 'pigeonpeas', 'watermelon',
'rice', 'cotton', 'rice', 'lentil', 'pigeonpeas', 'yapaya',
'coconut', 'pomegranate', 'mothbeans', 'muskmelon', 'pigeonpeas',
'watermelon', 'coffee', 'muskmelon', 'orange', 'banana', 'jute',
'mango', 'mungbean', 'maize', 'coconut', 'coconut', 'mango'],
dtype=object)

In [70]: plt.scatter(ytest,y\_pred,color='red',label=True)

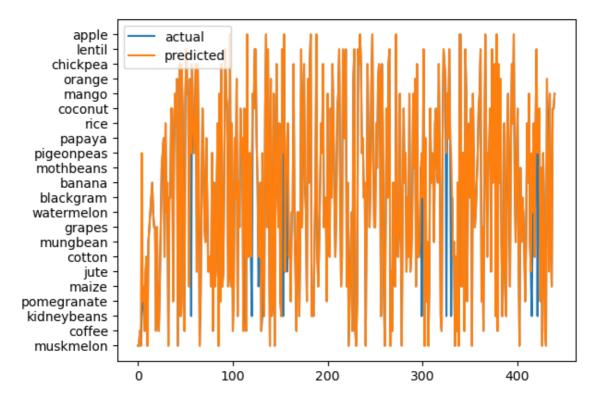




```
In [71]: import matplotlib.pyplot as plt

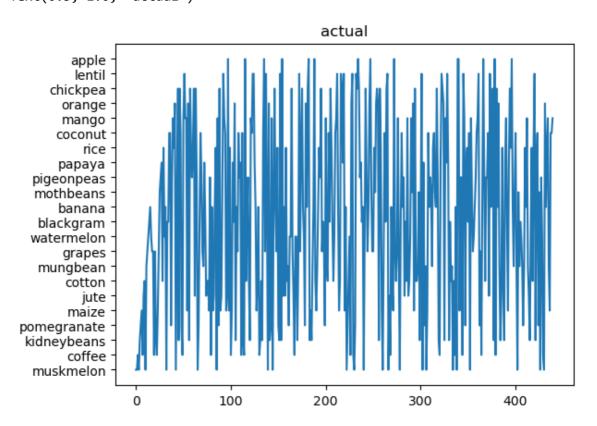
plt.plot(np.arange(0,len(ytest)),ytest,label='actual')
    plt.plot(np.arange(0,len(ytest)),rfc.predict(xtest),label='predicted')
    plt.legend()
```

Out[71]: <matplotlib.legend.Legend at 0x2576afdb1f0>



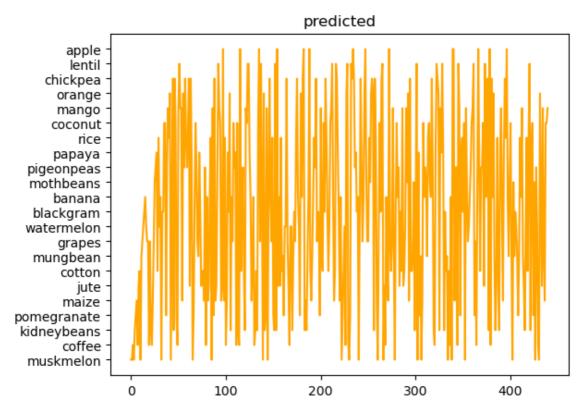
```
In [72]: plt.plot(ytest)
plt.title('actual')
```

Out[72]: Text(0.5, 1.0, 'actual')



```
In [73]: plt.plot(y_pred,color='orange')
plt.title('predicted')
```

Out[73]: Text(0.5, 1.0, 'predicted')



```
from sklearn.metrics import accuracy_score
In [74]:
         accuracy_score(s,ytest)
Out[74]: 0.9681818181818181
In [75]:
         classifier.predict([[85,58,41,21.770462,80.319644,7.038096,226.655537]])[0]
Out[75]:
         'papaya'
In [77]:
         import pickle
         path='C:/Users/saiteja143/Desktop/streamlit/crop_pred.sav'
         bo=pickle.dump(classifier,open(path,"wb"))
 In [ ]: # open("diab_pred.sav","wb")
In [78]: pickle.load(open(path, 'rb'))
Out[78]: KNeighborsClassifier(n_neighbors=3)
 In [ ]:
         xtest
```

xtest[:36,:]

In [ ]:

In [ ]: