

```
# Importing libraries

from __future__ import print_function
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import classification_report
from sklearn import metrics
from sklearn import tree
import warnings
warnings.filterwarnings('ignore')
```

```
PATH = '/content/Crop_recommendation (1).csv'
df = pd.read_csv(PATH)
```

df.head()

	N	P	K	temperature	humidity	ph	rainfall	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

df.tail()

	N	P	K	temperature	humidity	ph	rainfall	label
2195	107	34	32	26.774637	66.413269	6.780064	177.774507	coffee
2196	99	15	27	27.417112	56.636362	6.086922	127.924610	coffee
2197	118	33	30	24.131797	67.225123	6.362608	173.322839	coffee
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

```
df.size

17600
```

```
df.shape

(2200, 8)
```

```
df.columns

Index(['N', 'P', 'K', 'temperature', 'humidity', 'ph', 'rainfall', 'label'], dtype='object')
```

```
df['label'].unique()

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

```
df.dtypes

N                int64
P                int64
K                int64
temperature     float64
humidity        float64
ph              float64
rainfall        float64
label           object
dtype: object
```

```
df['label'].value_counts()
```

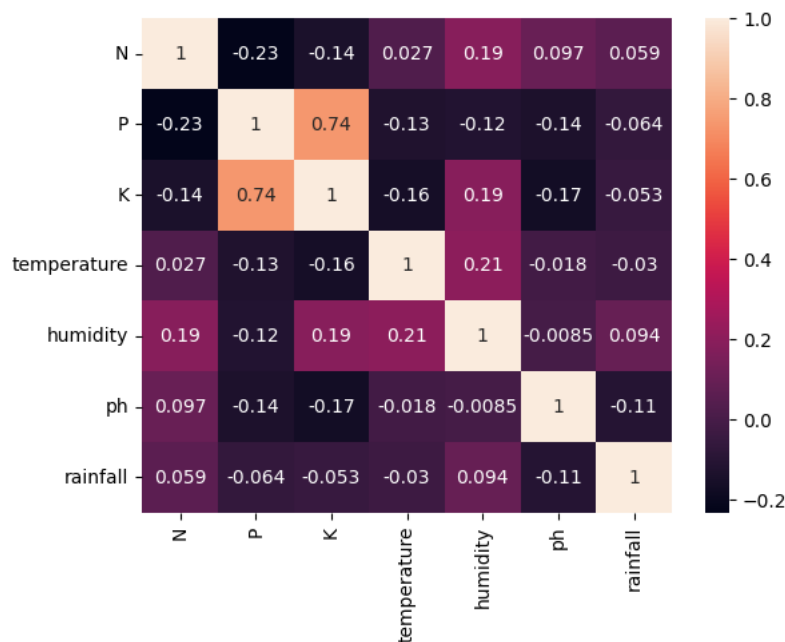
```

rice      100
maize     100
jute      100
cotton    100
coconut   100
papaya    100
orange    100
apple     100
muskmelon 100
watermelon 100
grapes    100
mango     100
banana    100
pomegranate 100
lentil    100
blackgram 100
mungbean  100
mothbeans 100
pigeonpeas 100
kidneybeans 100
chickpea  100
coffee    100
Name: label, dtype: int64

```

```
sns.heatmap(df.corr(),annot=True)
```

```
<Axes: >
```



```

features = df[['N', 'P','K','temperature', 'humidity', 'ph', 'rainfall']]
target = df['label']
labels = df['label']

```

```

# Initializing empty lists to append all model's name and corresponding name
acc = []
model = []

```

```
# Splitting into train and test data
```

```

from sklearn.model_selection import train_test_split
Xtrain, Xtest, Ytrain, Ytest = train_test_split(features,target,test_size = 0.2,random_state =2)

```

```
from sklearn.tree import DecisionTreeClassifier
```

```
DecisionTree = DecisionTreeClassifier(criterion="entropy",random_state=2,max_depth=5)
```

```
DecisionTree.fit(Xtrain,Ytrain)
```

```

predicted_values = DecisionTree.predict(Xtest)
x = metrics.accuracy_score(Ytest, predicted_values)
acc.append(x)
model.append('Decision Tree')
print("DecisionTrees's Accuracy is: ", x*100)

```

```
print(classification_report(Ytest,predicted_values))
```

```
DecisionTrees's Accuracy is:  90.0
precision    recall  f1-score   support

   apple      1.00      1.00      1.00      13
  banana      1.00      1.00      1.00      17
blackgram      0.59      1.00      0.74      16
 chickpea      1.00      1.00      1.00      21
   coconut      0.91      1.00      0.95      21
   coffee      1.00      1.00      1.00      22
   cotton      1.00      1.00      1.00      20
   grapes      1.00      1.00      1.00      18
    jute       0.74      0.93      0.83      28
kidneybeans      0.00      0.00      0.00      14
   lentil      0.68      1.00      0.81      23
   maize      1.00      1.00      1.00      21
   mango      1.00      1.00      1.00      26
mothbeans      0.00      0.00      0.00      19
 mungbean      1.00      1.00      1.00      24
 muskmelon      1.00      1.00      1.00      23
   orange      1.00      1.00      1.00      29
   papaya      1.00      0.84      0.91      19
pigeonpeas      0.62      1.00      0.77      18
pomegranate      1.00      1.00      1.00      17
    rice       1.00      0.62      0.77      16
 watermelon      1.00      1.00      1.00      15

 accuracy            0.90      440
 macro avg           0.84      440
weighted avg           0.86      440
```

```
from sklearn.model_selection import cross_val_score
```

```
# Cross validation score (Decision Tree)
score = cross_val_score(DecisionTree, features, target,cv=5)
```

```
score
```

```
array([0.93636364, 0.90909091, 0.91818182, 0.87045455, 0.93636364])
```

```
import pickle
# Dump the trained Naive Bayes classifier with Pickle
DT_pkl_filename = 'DecisionTree.pkl'
# Open the file to save as pkl file
DT_Model_pkl = open(DT_pkl_filename, 'wb')
pickle.dump(DecisionTree, DT_Model_pkl)
# Close the pickle instances
DT_Model_pkl.close()
```

```
from sklearn.naive_bayes import GaussianNB
```

```
NaiveBayes = GaussianNB()
```

```
NaiveBayes.fit(Xtrain,Ytrain)
```

```
predicted_values = NaiveBayes.predict(Xtest)
x = metrics.accuracy_score(Ytest, predicted_values)
acc.append(x)
model.append('Naive Bayes')
print("Naive Bayes's Accuracy is: ", x)
```

```
print(classification_report(Ytest,predicted_values))
```

```
Naive Bayes's Accuracy is:  0.990909090909091
precision    recall  f1-score   support

   apple      1.00      1.00      1.00      13
  banana      1.00      1.00      1.00      17
blackgram      1.00      1.00      1.00      16
 chickpea      1.00      1.00      1.00      21
   coconut      1.00      1.00      1.00      21
   coffee      1.00      1.00      1.00      22
   cotton      1.00      1.00      1.00      20
   grapes      1.00      1.00      1.00      18
    jute       0.88      1.00      0.93      28
kidneybeans      1.00      1.00      1.00      14
   lentil      1.00      1.00      1.00      23
   maize      1.00      1.00      1.00      21
   mango      1.00      1.00      1.00      26
mothbeans      1.00      1.00      1.00      19
 mungbean      1.00      1.00      1.00      24
```

muskmelon	1.00	1.00	1.00	23
orange	1.00	1.00	1.00	29
papaya	1.00	1.00	1.00	19
pigeonpeas	1.00	1.00	1.00	18
pomegranate	1.00	1.00	1.00	17
rice	1.00	0.75	0.86	16
watermelon	1.00	1.00	1.00	15
accuracy			0.99	440
macro avg	0.99	0.99	0.99	440
weighted avg	0.99	0.99	0.99	440

```
# Cross validation score (NaiveBayes)
```

```
score = cross_val_score(NaiveBayes, features, target, cv=5)
```

```
score
```

```
array([0.99772727, 0.99545455, 0.99545455, 0.99545455, 0.99090909])
```

```
import pickle
```

```
# Dump the trained Naive Bayes classifier with Pickle
```

```
NB_pkl_filename = 'NBClassifier.pkl'
```

```
# Open the file to save as pkl file
```

```
NB_Model_pkl = open(NB_pkl_filename, 'wb')
```

```
pickle.dump(NaiveBayes, NB_Model_pkl)
```

```
# Close the pickle instances
```

```
NB_Model_pkl.close()
```

```
from sklearn.svm import SVC
```

```
SVM = SVC(gamma='auto')
```

```
SVM.fit(Xtrain, Ytrain)
```

```
predicted_values = SVM.predict(Xtest)
```

```
x = metrics.accuracy_score(Ytest, predicted_values)
```

```
acc.append(x)
```

```
model.append('SVM')
```

```
print("SVM's Accuracy is: ", x)
```

```
print(classification_report(Ytest, predicted_values))
```

```
SVM's Accuracy is: 0.10681818181818181
```

	precision	recall	f1-score	support
apple	1.00	0.23	0.38	13
banana	1.00	0.24	0.38	17
blackgram	1.00	0.19	0.32	16
chickpea	1.00	0.05	0.09	21
coconut	1.00	0.05	0.09	21
coffee	0.00	0.00	0.00	22
cotton	1.00	0.05	0.10	20
grapes	1.00	0.06	0.11	18
jute	1.00	0.07	0.13	28
kidneybeans	0.03	1.00	0.07	14
lentil	0.00	0.00	0.00	23
maize	0.00	0.00	0.00	21
mango	0.00	0.00	0.00	26
mothbeans	0.00	0.00	0.00	19
mungbean	1.00	0.12	0.22	24
muskmelon	1.00	0.30	0.47	23
orange	1.00	0.03	0.07	29
papaya	1.00	0.05	0.10	19
pigeonpeas	0.00	0.00	0.00	18
pomegranate	1.00	0.12	0.21	17
rice	0.50	0.06	0.11	16
watermelon	1.00	0.13	0.24	15
accuracy			0.11	440
macro avg	0.66	0.13	0.14	440
weighted avg	0.66	0.11	0.13	440

```
# Cross validation score (SVM)
```

```
score = cross_val_score(SVM, features, target, cv=5)
```

```
score
```

```
array([0.27727273, 0.28863636, 0.29090909, 0.275      , 0.26818182])
```

```
from sklearn.linear_model import LogisticRegression
```

```
LogReg = LogisticRegression(random_state=2)
```

```
LogReg.fit(Xtrain,Ytrain)

predicted_values = LogReg.predict(Xtest)

x = metrics.accuracy_score(Ytest, predicted_values)
acc.append(x)
model.append('Logistic Regression')
print("Logistic Regression's Accuracy is: ", x)

print(classification_report(Ytest,predicted_values))
```

	precision	recall	f1-score	support
apple	1.00	1.00	1.00	13
banana	1.00	1.00	1.00	17
blackgram	0.86	0.75	0.80	16
chickpea	1.00	1.00	1.00	21
coconut	1.00	1.00	1.00	21
coffee	1.00	1.00	1.00	22
cotton	0.86	0.90	0.88	20
grapes	1.00	1.00	1.00	18
jute	0.84	0.93	0.88	28
kidneybeans	1.00	1.00	1.00	14
lentil	0.88	1.00	0.94	23
maize	0.90	0.86	0.88	21
mango	0.96	1.00	0.98	26
mothbeans	0.84	0.84	0.84	19
mungbean	1.00	0.96	0.98	24
muskmelon	1.00	1.00	1.00	23
orange	1.00	1.00	1.00	29
papaya	1.00	0.95	0.97	19
pigeonpeas	1.00	1.00	1.00	18
pomegranate	1.00	1.00	1.00	17
rice	0.85	0.69	0.76	16
watermelon	1.00	1.00	1.00	15
accuracy			0.95	440
macro avg	0.95	0.95	0.95	440
weighted avg	0.95	0.95	0.95	440

```
# Cross validation score (Logistic Regression)
score = cross_val_score(LogReg,features,target,cv=5)
score

array([0.95      , 0.96590909, 0.94772727, 0.96590909, 0.94318182])
```

```
import pickle
# Dump the trained Naive Bayes classifier with Pickle
LR_pkl_filename = 'LogisticRegression.pkl'
# Open the file to save as pkl file
LR_Model_pkl = open(DT_pkl_filename, 'wb')
pickle.dump(LogReg, LR_Model_pkl)
# Close the pickle instances
LR_Model_pkl.close()
```

```
from sklearn.ensemble import RandomForestClassifier

RF = RandomForestClassifier(n_estimators=20, random_state=0)
RF.fit(Xtrain,Ytrain)

predicted_values = RF.predict(Xtest)

x = metrics.accuracy_score(Ytest, predicted_values)
acc.append(x)
model.append('RF')
print("RF's Accuracy is: ", x)

print(classification_report(Ytest,predicted_values))
```

	precision	recall	f1-score	support
apple	1.00	1.00	1.00	13
banana	1.00	1.00	1.00	17
blackgram	0.94	1.00	0.97	16
chickpea	1.00	1.00	1.00	21
coconut	1.00	1.00	1.00	21
coffee	1.00	1.00	1.00	22
cotton	1.00	1.00	1.00	20
grapes	1.00	1.00	1.00	18
jute	0.90	1.00	0.95	28

kidneybeans	1.00	1.00	1.00	14
lentil	1.00	1.00	1.00	23
maize	1.00	1.00	1.00	21
mango	1.00	1.00	1.00	26
mothbeans	1.00	0.95	0.97	19
mungbean	1.00	1.00	1.00	24
muskmelon	1.00	1.00	1.00	23
orange	1.00	1.00	1.00	29
papaya	1.00	1.00	1.00	19
pigeonpeas	1.00	1.00	1.00	18
pomegranate	1.00	1.00	1.00	17
rice	1.00	0.81	0.90	16
watermelon	1.00	1.00	1.00	15
accuracy			0.99	440
macro avg	0.99	0.99	0.99	440
weighted avg	0.99	0.99	0.99	440

```
# Cross validation score (Random Forest)
score = cross_val_score(RF, features, target, cv=5)
score
```

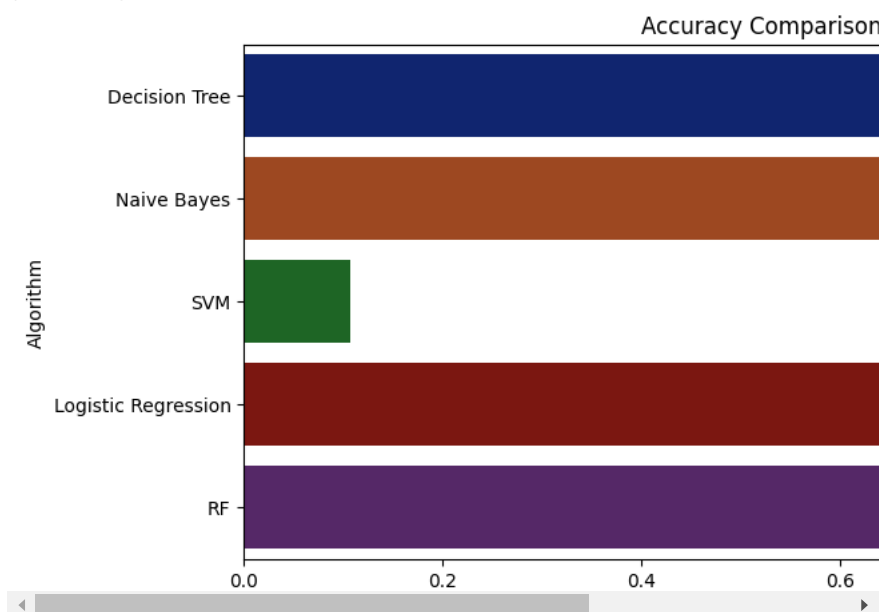
```
array([0.99772727, 0.99545455, 0.99772727, 0.99318182, 0.98863636])
```

```
import pickle
# Dump the trained Naive Bayes classifier with Pickle
RF_pkl_filename = 'reported.pkl'
# Open the file to save as pkl file
RF_Model_pkl = open(RF_pkl_filename, 'wb')
pickle.dump(RF, RF_Model_pkl)
# Close the pickle instances
RF_Model_pkl.close()
```

```
import pickle
# Dump the trained Naive Bayes classifier with Pickle
XB_pkl_filename = 'XGBoost.pkl'
# Open the file to save as pkl file
XB_Model_pkl = open(XB_pkl_filename, 'wb')
pickle.dump(XB, XB_Model_pkl)
# Close the pickle instances
XB_Model_pkl.close()
```

```
plt.figure(figsize=[10,5],dpi = 100)
plt.title('Accuracy Comparison')
plt.xlabel('Accuracy')
plt.ylabel('Algorithm')
sns.barplot(x = acc,y = model,palette='dark')
```

```
<Axes: title={'center': 'Accuracy Comparison'}, xlabel='Accuracy',
ylabel='Algorithm'>
```



```
accuracy_models = dict(zip(model, acc))
for k, v in accuracy_models.items():
    print (k, '-->', v)
```

```
Decision Tree --> 0.9
Naive Bayes --> 0.990909090909091
SVM --> 0.10681818181818181
Logistic Regression --> 0.9522727272727273
RF --> 0.990909090909091
```

```
data = np.array([[104,18, 30, 23.603016, 60.3, 6.7, 140.91]])
prediction = RF.predict(data)
print(prediction)
```

```
['coffee']
```

```
data = np.array([[83, 45, 60, 28, 70.3, 7.0, 150.9]])
prediction = RF.predict(data)
print(prediction)
```

```
['jute']
```