```
In [1]:
             import pandas as pd
           2 import numpy as np
           3 import seaborn as sns
          4 import matplotlib.pyplot as plt
             from sklearn.metrics import classification_report
           6 from sklearn import metrics
          7
             from sklearn import tree
           8 import warnings
             warnings.filterwarnings('ignore')
In [2]:
             df=pd.read csv('Crop recommendation.csv')
             df
Out[2]:
                        K temperature
                                       humidity
                                                     ph
                                                            rainfall
                                                                    label
                   42 43
            0
                90
                             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
                74
                    35 40
                             26.491096 80.158363 6.980401 242.864034
            3
                                                                     rice
                78 42 42
                             20.130175 81.604873 7.628473
                                                         262.717340
                                                                     rice
                    ...
                             26.774637 66.413269 6.780064 177.774507 coffee
         2195 107 34 32
         2196
                99
                   15 27
                             27.417112 56.636362 6.086922 127.924610 coffee
                             24.131797 67.225123 6.362608 173.322839 coffee
         2197
               118 33 30
         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 × 8 columns
          1 df.shape
In [3]:
Out[3]: (2200, 8)
In [4]:
          1 p = list(df.columns)
           2 print(p)
           3 \mid \#N = NITROGEN
          4 \#P = PHOSPHORUS
          5 #K = POTASSIUM
            #TEMPERATURE OF THE REGION
          7
            # HUMIDITY IN THAT REGION
             # PH OF THE SOIL
          9 # RAINFALL IN THAT REGION
         10 # CROP THAT IS SUITABLE IN THAT REGION
         ['N', 'P', 'K', 'temperature', 'humidity', 'ph', 'rainfall', 'label']
```

```
In [5]:
          1 (df.dtypes)
Out[5]: N
                          int64
                          int64
        Κ
                          int64
                        float64
        temperature
                        float64
        humidity
        ph
                        float64
        rainfall
                        float64
        label
                         object
        dtype: object
          1 print(list(df['label'].unique()))
In [6]:
          2 #LIST OF CROPS IN THE DATASET
        ['rice', 'maize', 'chickpea', 'kidneybeans', 'pigeonpeas', 'mothbeans', 'mungbe
        an', 'blackgram', 'lentil', 'pomegranate', 'banana', 'mango', 'grapes', 'waterm
        elon', 'muskmelon', 'apple', 'orange', 'papaya', 'coconut', 'cotton', 'jute',
         'coffee']
In [7]:
          1 df['label'].value_counts()
          2 #22 DIFFERENT TYPES OF CROPS ARE PRESENT IN THE DATASET
          3 #AND EACH CROP HAS 100 DIFFERENT SAMPLES
Out[7]: maize
                        100
                        100
        watermelon
        banana
                        100
        grapes
                        100
        jute
                        100
        apple
                        100
        pigeonpeas
                        100
                        100
        chickpea
                        100
        cotton
                        100
        mothbeans
        orange
                        100
                        100
        papaya
                        100
        lentil
                        100
        pomegranate
        mungbean
                        100
        muskmelon
                        100
        mango
                        100
        blackgram
                        100
        coffee
                        100
        coconut
                        100
        rice
                        100
        kidneybeans
                        100
        Name: label, dtype: int64
```

In [8]: 1 df.describe()

Out[8]:

	N	Р	K	temperature	humidity	ph	rainfall
count	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000
mean	50.551818	53.362727	48.149091	25.616244	71.481779	6.469480	103.463655
std	36.917334	32.985883	50.647931	5.063749	22.263812	0.773938	54.958389
min	0.000000	5.000000	5.000000	8.825675	14.258040	3.504752	20.211267
25%	21.000000	28.000000	20.000000	22.769375	60.261953	5.971693	64.551686
50%	37.000000	51.000000	32.000000	25.598693	80.473146	6.425045	94.867624
75%	84.250000	68.000000	49.000000	28.561654	89.948771	6.923643	124.267508
max	140.000000	145.000000	205.000000	43.675493	99.981876	9.935091	298.560117

In [9]:

1 df.head()

Out[9]:

	N	Р	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

In [10]:

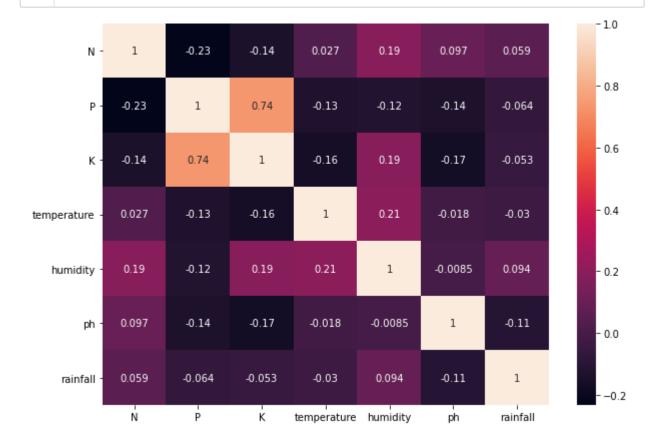
1 df.tail()

Out[10]:

	N	Р	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

In [11]:

- plt.figure(figsize=(10,7))
- 2 sns.heatmap(df.corr(),annot = True)
- 3 plt.show()
- 4 #shows the correalation b/w to features in the dataset
- 5 #annot = writes the data value in each cell



```
In [12]: 1 features = df[['N','P','K','temperature','humidity','ph','rainfall']]
2 features
3 #THESE ARE THE FEATURES WHICH WE USE TO RECOMMEND A
4 #CROP IN A CERTAIN RANGE
```

Out[12]:

	N	Р	K	temperature	humidity	ph	rainfall
0	90	42	43	20.879744	82.002744	6.502985	202.935536
1	85	58	41	21.770462	80.319644	7.038096	226.655537
2	60	55	44	23.004459	82.320763	7.840207	263.964248
3	74	35	40	26.491096	80.158363	6.980401	242.864034
4	78	42	42	20.130175	81.604873	7.628473	262.717340
2195	107	34	32	26.774637	66.413269	6.780064	177.774507
2196	99	15	27	27.417112	56.636362	6.086922	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
2199	104	18	30	23.603016	60.396475	6.779833	140.937041

2200 rows × 7 columns

```
In [13]: 1 target = df[['label']]
2 target
3 #BY USING ABOVE FEATURES WE DECIDE
4 #WHICH CROP HAS TO BE GROWN AND HARVESTED IN THAT REGION
```

Out[13]:

	label
0	rice
1	rice
2	rice
3	rice
4	rice
2195	coffee
2196	coffee
2197	coffee
2198	coffee
2199	coffee

2200 rows × 1 columns

Out[16]:

	N	Р	K	temperature	humidity	ph	rainfall
1863	1	12	30	27.754298	95.946438	5.562224	131.090008
987	11	10	45	22.630452	88.455772	6.397996	109.035760
375	5	74	21	16.244692	21.357939	5.591704	66.970533
1071	105	88	54	25.787498	84.511942	6.020445	114.200546
37	95	39	36	23.863305	83.152508	5.561399	285.249365
960	1	27	36	23.985988	93.342366	5.684995	104.991282
905	31	25	38	24.962732	92.405014	6.497367	109.416919
1096	92	81	52	28.010680	76.528081	5.891414	103.704078
235	28	58	81	17.475010	16.543148	6.180427	93.350343
1061	95	74	50	25.901131	80.471527	6.002482	110.103230

1760 rows × 7 columns

In [17]:

1 features_test

2 #TESTING DATASET

Out[17]:

	N	Р	K	temperature	humidity	ph	rainfall
1276	25	129	195	17.986678	81.177121	5.777271	72.371277
1446	106	20	51	29.730197	90.970157	6.342573	20.490356
335	33	59	22	22.642369	21.593961	5.947000	122.388601
1458	89	9	47	29.471563	90.770696	6.668383	28.752261
2038	62	49	37	24.217446	82.852840	7.479248	166.136589
1508	22	144	196	21.911913	91.687481	6.499227	117.076128
1595	40	120	197	23.805938	92.488795	5.889481	119.633555
1032	105	74	45	25.145176	81.382041	6.098369	119.218154
1330	118	15	45	24.214957	84.205770	6.538006	48.011385
1263	37	135	205	11.827682	80.282719	5.510925	74.102251

440 rows × 7 columns

In [18]:

- 1 target_train
- 2 #LABEL OR OUTPUT FOR TRAINING DATA AS INPUT

Out[18]:

	label
1863	coconut
987	pomegranate
375	kidneybeans
1071	banana
37	rice
960	pomegranate
905	pomegranate
1096	banana
235	chickpea
1061	banana

1760 rows × 1 columns

```
In [19]: 1 target_test
2 #LABEL OR OUTPUT FOR TESTING DATA AS INPUT
```

Out[19]:

	label
1276	grapes
1446	muskmelon
335	kidneybeans
1458	muskmelon
2038	jute
1508	apple
1595	apple
1032	banana
1330	watermelon
1263	grapes

440 rows × 1 columns

MODELS

DECISION TREE MODEL

```
In [23]:
           1 model dt.fit(features train ,target train)
Out[23]: DecisionTreeClassifier()
In [24]:
             q = model dt.score(features test, target test)
             #GETTING THE ACCURACY VALUE OF THE MODEL
In [25]:
           1 x = q*100
           2
             print("ACCURACY OF DECISION TREE MODEL IS: ",x)
         ACCURACY OF DECISION TREE MODEL IS: 99.54545454545455
In [26]:
             models accuracy.append(x)
           2 ml_models.append('DECISION_TREE')
In [27]:
             print(list(model dt.predict([[70,42,63,30.879744,82.002744,16.502985,45.9355
         ['banana']
In [28]:
             def predict dt crop():
           2
                 lst=list(range(7))
           3
                 lst[0]=int(input("PLEASE ENTER VALUE OF NITROGEN: "))
           4
                 lst[1]=int(input("PLEASE ENTER VALUE OF PHOSPHORUS: "))
           5
                 lst[2]=int(input("PLEASE ENTER VALUE OF POTASSIUM: "))
           6
                 lst[3]=float(input("PLEASE ENTER VALUE OF TEMPERATURE: "))
           7
                 lst[4]=float(input("PLEASE ENTER VALUE OF HUMIDITY: "))
           8
                 lst[5]=float(input("PLEASE ENTER VALUE OF PH: "))
           9
                 lst[6]=float(input("PLEASE ENTER AMOUNT OF RAINFALL: "))
          10
                 print("THE PREFERED CROP IN YOUR REGION IS: ",
                       model dt.predict([lst]))
          11
          12
                 In [29]:
             predict_dt_crop()
         PLEASE ENTER VALUE OF NITROGEN: 3
         PLEASE ENTER VALUE OF PHOSPHORUS: 49
         PLEASE ENTER VALUE OF POTASSIUM: 27
         PLEASE ENTER VALUE OF TEMPERATURE: 64
         PLEASE ENTER VALUE OF HUMIDITY: 3.69
         PLEASE ENTER VALUE OF PH: 3
         PLEASE ENTER AMOUNT OF RAINFALL: 32
         THE PREFERED CROP IN YOUR REGION IS: ['kidneybeans']
         👍 👍 HAPPY FARMING 📛 🤭
```

NAIVE BAYES

```
In [30]: 1 from sklearn.naive_bayes import GaussianNB
2 model_nb = GaussianNB()
```

```
In [31]:
           1 model nb.fit(features train, target train)
Out[31]: GaussianNB()
In [32]:
             y1 = model_nb.score(features_test, target_test)
           2
             y = y1*100
             print("ACCURACY OF NAIVE BAYES MODEL IS: ",y)
         ACCURACY OF NAIVE BAYES MODEL IS: 99.54545454545455
In [33]:
             target test[:10]
Out[33]:
                     label
          1276
                   grapes
          1446
                muskmelon
           335
               kidneybeans
          1458
                muskmelon
          2038
                      jute
                watermelon
          1314
           389
               kidneybeans
          1639
                   orange
          2004
                      jute
           403
                pigeonpeas
In [34]:
             print(list(model nb.predict(features test[:10])))
         ['grapes', 'muskmelon', 'kidneybeans', 'muskmelon', 'jute', 'watermelon', 'kidn
         eybeans', 'orange', 'jute', 'pigeonpeas']
In [35]:
           1
              def predict_nb_crop():
           2
                  lst=list(range(7))
           3
                  lst[0]=int(input("PLEASE ENTER VALUE OF NITROGEN: "))
                  lst[1]=int(input("PLEASE ENTER VALUE OF PHOSPHORUS: "))
           4
           5
                  lst[2]=int(input("PLEASE ENTER VALUE OF POTASSIUM: "))
                  lst[3]=float(input("PLEASE ENTER VALUE OF TEMPERATURE: "))
           6
                  lst[4]=float(input("PLEASE ENTER VALUE OF HUMIDITY: "))
           7
           8
                  lst[5]=float(input("PLEASE ENTER VALUE OF PH: "))
                  lst[6]=float(input("PLEASE ENTER AMOUNT OF RAINFALL: "))
           9
                  print("THE PREFERED CROP IN YOUR REGION IS: ",
          10
                        model nb.predict([lst]))
          11
          12
```

SUPPORT VECTOR MACHINE

ACCURACY OF SUPPORT VECTOR MACHINE MODEL IS: 97.95454545454545

```
In [42]:
           1 target test[:10]
Out[42]:
                     label
          1276
                   grapes
          1446
                muskmelon
           335
               kidneybeans
          1458
                muskmelon
          2038
                      jute
          1314
                watermelon
           389
               kidneybeans
          1639
                   orange
          2004
                      jute
           403
                pigeonpeas
In [43]:
             print(list(model_svm.predict(features_test[:10])))
         ['grapes', 'muskmelon', 'kidneybeans', 'muskmelon', 'jute', 'watermelon', 'kidn
         eybeans', 'orange', 'jute', 'pigeonpeas']
In [44]:
              models accuracy.append(z)
             ml models.append('SVM')
In [45]:
              def predict svm crop():
           1
           2
                  lst=list(range(7))
           3
                  lst[0]=int(input("PLEASE ENTER VALUE OF NITROGEN: "))
           4
                  lst[1]=int(input("PLEASE ENTER VALUE OF PHOSPHORUS: "))
           5
                  lst[2]=int(input("PLEASE ENTER VALUE OF POTASSIUM: "))
                  lst[3]=float(input("PLEASE ENTER VALUE OF TEMPERATURE: "))
           6
           7
                  lst[4]=float(input("PLEASE ENTER VALUE OF HUMIDITY: "))
           8
                  lst[5]=float(input("PLEASE ENTER VALUE OF PH: "))
           9
                  lst[6]=float(input("PLEASE ENTER AMOUNT OF RAINFALL: "))
          10
                  print("THE PREFERED CROP IN YOUR REGION IS: ",
          11
                        model svm.predict([lst]))
          12
                  In [46]:
             predict_nb_crop()
         PLEASE ENTER VALUE OF NITROGEN: 3
         PLEASE ENTER VALUE OF PHOSPHORUS: 49
         PLEASE ENTER VALUE OF POTASSIUM: 18
         PLEASE ENTER VALUE OF TEMPERATURE: 27
         PLEASE ENTER VALUE OF HUMIDITY: 64
         PLEASE ENTER VALUE OF PH: 3
         PLEASE ENTER AMOUNT OF RAINFALL: 32
         THE PREFERED CROP IN YOUR REGION IS: ['mothbeans']
          👍 👍 HAPPY FARMING 📛 🤭
```

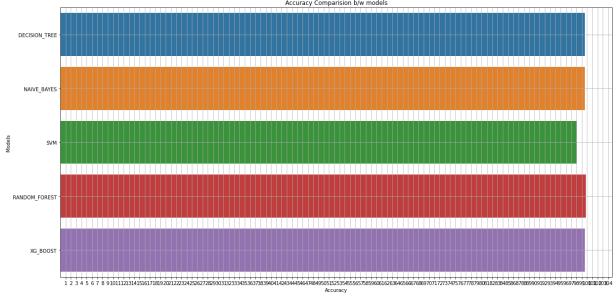
RANDOM FOREST

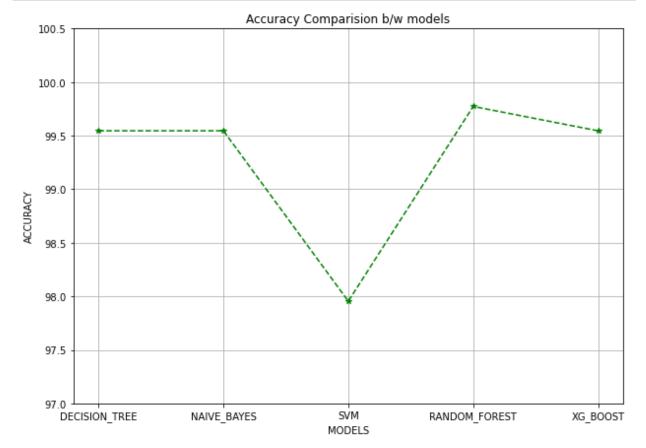
```
In [47]:
              from sklearn.ensemble import RandomForestClassifier
In [48]:
              model rf = RandomForestClassifier()
In [49]:
              model rf.fit(features train, target train)
Out[49]: RandomForestClassifier()
In [50]:
              a1 = model rf.score(features test, target test)
           2 \mid a = a1*100
              print("ACCURACY OF RANDOM FOREST MODEL IS: ",a)
         ACCURACY OF RANDOM FOREST MODEL IS: 99.77272727272727
In [51]:
           1 target_test[:5]
Out[51]:
                     label
          1276
                   grapes
          1446
                muskmelon
           335
               kidneybeans
          1458
                muskmelon
          2038
                     jute
In [52]:
              print(list(model rf.predict(features test[:5])))
         ['grapes', 'muskmelon', 'kidneybeans', 'muskmelon', 'jute']
In [53]:
              models accuracy.append(a)
              ml models.append('RANDOM FOREST')
In [54]:
           1
              def predict rf crop():
           2
                  lst=list(range(7))
           3
                  lst[0]=int(input("PLEASE ENTER VALUE OF NITROGEN: "))
           4
                  lst[1]=int(input("PLEASE ENTER VALUE OF PHOSPHORUS: "))
           5
                  lst[2]=int(input("PLEASE ENTER VALUE OF POTASSIUM: "))
           6
                  lst[3]=float(input("PLEASE ENTER VALUE OF TEMPERATURE: "))
                  lst[4]=float(input("PLEASE ENTER VALUE OF HUMIDITY: "))
           7
           8
                  lst[5]=float(input("PLEASE ENTER VALUE OF PH: "))
                  lst[6]=float(input("PLEASE ENTER AMOUNT OF RAINFALL: "))
           9
          10
                  print("THE PREFERED CROP IN YOUR REGION IS: ",
          11
                        model rf.predict([lst]))
          12
```

XGBOOST

```
In [56]:
              import xgboost as xgb
In [57]:
              model xb = xgb.XGBClassifier()
In [58]:
           1 model xb.fit(features train, target train)
           2 b1 = model xb.score(features test, target test)
         [19:41:23] WARNING: C:/Users/Administrator/workspace/xgboost-win64 release 1.5.
         1/src/learner.cc:1115: Starting in XGBoost 1.3.0, the default evaluation metric
         used with the objective 'multi:softprob' was changed from 'merror' to 'mloglos
         s'. Explicitly set eval metric if you'd like to restore the old behavior.
In [59]:
           1 b = b1*100
           2 print("ACCURACY OF XGBOOST MODEL IS: ",b)
           3 #WE GOT ACCURACY SAME AS DECISION TREE AND NAIVE BAYES
         ACCURACY OF XGBOOST MODEL IS: 99.5454545454545
           1 target test[5:10]
In [60]:
Out[60]:
                     label
          1314
                watermelon
           389
               kidneybeans
          1639
                    orange
          2004
                      jute
           403
                pigeonpeas
In [61]:
              print(list(model xb.predict(features test[5:10])))
         ['watermelon', 'kidneybeans', 'orange', 'jute', 'pigeonpeas']
```

```
In [62]:
           1 models accuracy.append(b)
             ml models.append('XG BOOST')
In [63]:
              ml_models
Out[63]: ['DECISION_TREE', 'NAIVE_BAYES', 'SVM', 'RANDOM_FOREST', 'XG_BOOST']
In [65]:
              print(models accuracy)
         [99.545454545455, 99.545454545455, 97.954545454545, 99.77272727272727, 9
         9.54545454545455]
In [90]:
              plt.figure(figsize = [20,10])
           2 plt.title('Accuracy Comparision b/w models')
           3 plt.xlabel('Accuracy')
           4 plt.ylabel('Models')
           5 sns.barplot(x = models_accuracy,y = ml_models)
           6 plt.grid()
           7 plt.xticks(np.arange(1,105,1))
             plt.show()
                                               Accuracy Comparision b/w models
```





FINAL PREDICTION

```
In [97]: 1 predict_rf_crop()

PLEASE ENTER VALUE OF NITROGEN: 90
PLEASE ENTER VALUE OF PHOSPHORUS: 42
PLEASE ENTER VALUE OF POTASSIUM: 43
PLEASE ENTER VALUE OF TEMPERATURE: 23.603
PLEASE ENTER VALUE OF HUMIDITY: 60.3
PLEASE ENTER VALUE OF PH: 6.7
PLEASE ENTER AMOUNT OF RAINFALL: 140.91
THE PREFERED CROP IN YOUR REGION IS: ['coffee']

HAPPY FARMING 
TIN []: 1
```