## **Model Performance Test**

Date	21 / 11 / 2022	
Team ID	PNT2022TMID06211	
Project Name	Efficient Water Quality Analysis and prediction using Machine learning	
Maximum Marks	10 Marks	

## **Model Performance Testing:**

Project team shallfill the following information in model performance testing template.

S.No.	Parameter	Values	Screenshot
1.	Model Summary	Total params: 3279 rows,16 columns Trainable params: 3279 rows,16 columns Non- trainable params: 0	0.575 (0.029) with: {'criterion: 'entropy', 'min_samples_split': 2, 'splitten': 'best'} 0.575 (0.024) with: {'criterion: 'entropy', 'min_samples_split': 2, 'splitten': 'random'} 0.575 (0.034) with: {'criterion: 'entropy', 'min_samples_split': 4, 'splitten': 'best'} 0.571 (0.032) with: {'criterion: 'entropy', 'min_samples_split': 6, 'splitten': 'random'} 0.576 (0.028) with: {'criterion: 'entropy', 'min_samples_split': 6, 'splitten': 'best'} 0.576 (0.023) with: {'criterion: 'entropy', 'min_samples_split': 6, 'splitten': 'best'} 0.576 (0.023) with: {'criterion: 'entropy', 'min_samples_split': 8, 'splitten': 'best'} 0.580 (0.023) with: {'criterion: 'entropy', 'min_samples_split': 10, 'splitten': 'best'} 0.580 (0.028) with: {'criterion: 'entropy', 'min_samples_split': 10, 'splitten': 'random'} 0.576 (0.028) with: {'criterion: 'entropy', 'min_samples_split': 12, 'splitten': 'random'} 0.586 (0.029) with: {'criterion: 'entropy', 'min_samples_split': 12, 'splitten': 'random'} 0.586 (0.024) with: {'criterion: 'entropy', 'min_samples_split': 14, 'splitten': 'best'} 0.586 (0.024) with: {'criterion: 'entropy', 'min_samples_split': 14, 'splitten': 'best'} 0.586 (0.024) with: {'criterion: 'entropy', 'min_samples_split': 14, 'splitten': 'best'} 0.586 (0.024) with: {'criterion: 'entropy', 'min_samples_split': 14, 'splitten': 'best'} 0.586 (0.026) with: {'criterion: 'entropy', 'min_samples_split': 14, 'splitten': 'best'} 0.586 (0.026) with: {'criterion: 'entropy', 'min_samples_split': 14, 'splitten': 'best'} 0.586 (0.026) with: {'criterion: 'entropy', 'min_samples_split': 14, 'splitten': 'best'} 0.586 (0.026) with: {'criterion: 'entropy', 'min_samples_split': 14, 'splitten': 'best'} 0.586 (0.026) with: {'criterion: 'entropy', 'min_samples_split': 14, 'splitten': 'best'} 0.586 (0.026) with: {'criterion: 'entropy', 'min_samples_split': 14, 'splitten': 'best'} 0.586 (0.026) with: {'criterion: 'entropy', 'min_samples_split': 14, 'splitten': 'best'}
2.	Accuracy	TrainingAccuracy – 12.68	In [28]: prediction-dt.predict[/(test) print(f**Coursey Score = (accuracy_score(Y_test,prediction)*188)**) print(f**Confusion Matrix = In (confusion_matrix(Y_test,prediction))*) print(f**Classification_Report = In (classification_report(Y_test,predict
		ValidationAccuracy – 13.07	Accuracy Score - 56.859756897568975  Convision Hamilton - [1274 128]   [1274 128]

## ModelSummary

```
0.575 (0.029) with: {'criterion': 'entropy', 'min_samples_split': 2, 'splitter': 'best'}
0.573 (0.024) with: {'criterion': 'entropy', 'min_samples_split': 2, 'splitter': 'random'}
0.575 (0.034) with: {'criterion': 'entropy', 'min_samples_split': 4, 'splitter': 'best'}
0.571 (0.032) with: {'criterion': 'entropy', 'min_samples_split': 4, 'splitter': 'random'}
0.577 (0.034) with: {'criterion': 'entropy', 'min_samples_split': 6, 'splitter': 'best'}
0.578 (0.028) with: {'criterion': 'entropy', 'min_samples_split': 6, 'splitter': 'random'}
0.574 (0.033) with: {'criterion': 'entropy', 'min_samples_split': 8, 'splitter': 'best'}
0.578 (0.023) with: {'criterion': 'entropy', 'min_samples_split': 8, 'splitter': 'random'}
0.580 (0.029) with: {'criterion': 'entropy', 'min_samples_split': 10, 'splitter': 'best'}
0.582 (0.026) with: {'criterion': 'entropy', 'min_samples_split': 10, 'splitter': 'random'}
0.576 (0.028) with: {'criterion': 'entropy', 'min_samples_split': 12, 'splitter': 'best'}
0.584 (0.026) with: {'criterion': 'entropy', 'min_samples_split': 12, 'splitter': 'random'}
0.576 (0.024) with: {'criterion': 'entropy', 'min_samples_split': 14, 'splitter': 'random'}
0.585 (0.036) with: {'criterion': 'entropy', 'min_samples_split': 14, 'splitter': 'random'}
Training Score: 90.11450381679388
Testing Score: 59.29878048780488
```

## **Accuracy**

```
In [28]: prediction=dt.predict(X_test)
             prediction=dt.predict(\_lest)
print(f"Accuracy Score = {accuracy_score(Y_test,prediction)*100}")
print(f"Confusion Matrix =\n {confusion_matrix(Y_test,prediction)}")
print(f"Classification Report =\n {classification_report(Y_test,predict
              Accuracy Score = 56.859756097560975
              Confusion Matrix =
               [[274 128]
                        99]]
               T155
              Classification Report =
                                   precision
                                                       recall f1-score
                              0
                                         0.64
                                                       0.68
                                                                      0.66
                                                                                       402
                                         0.44
                                                                                       254
                                                       0.39
                                                                      0.41
                                                                       0.57
                                                                                       656
                    accuracy
                                        0.54
                                                      0.54
                                                                      0.54
                                                                                        656
                  macro avg
```