## Project Development Phase Model Performance Test

Date	13 November 2022
Team ID	PNT2022TMID03926
Project Name Project – Web Phishing Detection	
Maximum Marks	10 Marks

### **Model Performance Testing:**

Project team shall fill the following information in model performance testing template.

S.No.	Parameter	Values	Screenshot		
1.	Metrics	Classification Model: Gradient Boosting Classification Accuray Score- 97.4%	In [52]: Accepaning the clossification regard of the model priori(metries aleasification_regard(p_text_p_text_plex)) president result fluores support 1 8.90 0.66 0.67 0.70 0.70 0.70 0.70 0.70 0.7		
2.	Tune the Model	Hyperparameter Tuning - 97% Validation Method – KFOLD & Cross Validation Method	Wilcomo signed-rank test  is (A) emission of creat situation made  for expectable more edicinary  for expectable more edicinary  for expectable more edicinary  for edicinary emission expectations and the edicinary  for edicinary emission of concentration of the edicinary  at a many emission of the edicinary emission of the edicinary  at a many emission of the edicinary emission of the edicinary  at these emission and order programment emission  at the edicinary emission of the edicinary emission emis		

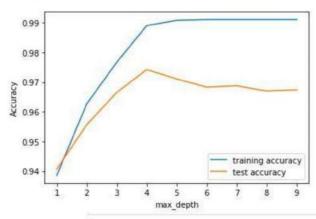
#### 1. METRICS:

### **CLASSIFICATION REPORT:**

In [52]: #computing the classification report of the model
 print(metrics.classification\_report(y\_test, y\_test\_gbc))

	precision	recall	f1-score	support
-1	0.99	0.96	0.97	976
1	0.97	0.99	0.98	1235
accuracy			0.97	2211
macro avg	0.98	0.97	0.97	2211
weighted avg	0.97	0.97	0.97	2211

# PERFORMANCE:



Out[83]:		ML Model	Accuracy	f1_score	Recall	Precision
	0	Gradient Boosting Classifier	0.974	0.977	0.994	0.986
	1	CatBoost Classifier	0.972	0.975	0.994	0.989
	2	Random Forest	0.969	0.972	0.992	0.991
	3	Support Vector Machine	0.964	0.968	0.980	0.965
	4	Decision Tree	0.958	0.962	0.991	0.993
	5	K-Nearest Neighbors	0.956	0.961	0.991	0.989
	6	Logistic Regression	0.934	0.941	0.943	0.927
	7	Naive Bayes Classifier	0,605	0.454	0.292	0.997
	8	XGBoost Classifier	0.548	0.548	0.993	0.984
	9	Multi-layer Perceptron	0.543	0.543	0.989	0.983

### 2. TUNE THE MODEL - HYPERPARAMETER TUNING

```
In [58]: #HYPERPARAMETER TUNING
         grid.fit(X_train, y_train)
Out[58]:
                                                            GridSearchCV
          GridSearchCV(cv=5,
                       estimator=GradientBoostingClassifier(learning_rate=0.7,
                                                           max_depth=4),
                       param_grid={'max_features': array([1, 2, 3, 4, 5]),
                                   'n_estimators': array([ 10, 20, 30, 40, 50, 60, 70, 80, 90, 100, 110, 120, 130,
                 140, 150, 160, 170, 180, 190, 200])})
                                                estimator: GradientBoostingClassifier
                                    GradientBoostingClassifier(learning rate=0.7, max_depth=4)
                                                     GradientBoostingClassifier
                                     GradientBoostingClassifier(learning_rate=0.7, max_depth=4)
  In [59]: print("The best parameters are %s with a score of %0.2f"
                 % (grid.best_params_, grid.best_score_))
           The best parameters are {'max_features': 5, 'n_estimators': 200} with a score of 0.97
```

## VALIDATION METHODS: KFOLD & Cross Folding

## Wilcoxon signed-rank test

```
In [78]: #KFOLD and Cross Validation Model
         from scipy.stats import wilcoxon
         from sklearn.datasets import load_iris
         from sklearn.ensemble import GradientBoostingClassifier
         from xgboost import XGBClassifier
         from sklearn.model_selection import cross_val_score, KFold
         # Load the dataset
         X = load iris().data
         y = load_iris().target
        # Prepare models and select your CV method
        model1 = GradientBoostingClassifier(n_estimators=100)
         model2 = XGBClassifier(n_estimators=100)
         kf = KFold(n_splits=20, random_state=None)
         # Extract results for each model on the same folds
         results_model1 = cross_val_score(model1, X, y, cv=kf)
         results_model2 = cross_val_score(model2, X, y, cv=kf)
         stat, p = wilcoxon(results_model1, results_model2, zero_method='zsplit');
         stat
Out[78]: 95.0
```

#### 5x2CV combined F test