Project Development Phase Model Performance Test

Date	13 November 2022
Team ID	PNT2022TMID28334
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.

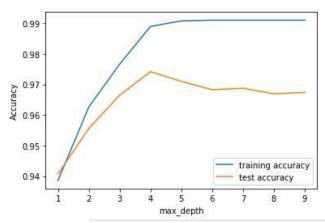
S.No.	Parameter	Values	Screenshot			
1.	Metrics	Classification Model: Gradient Boosting Classification Accuray Score- 97.4%	In [52]: **computing the classification report of the model print(metrics classification report(y_text, y_text_gbc)) **precision recall fiscore apport -1 8.59 8.58 0.57 575 -1 0.77 8.59 0.59 2135 **accuracy ag 0.58 0.57 0.57 2211 **marry ang 0.58 0.57 0.57 2211 **marry ang 0.57 0.57 0.77 2211			
2.	Tune the Model	Hyperparameter Tuning - 97% Validation Method – KFOLD & Cross Validation Method	Wilcowon signed-rank test 1s [78] MONDS and Cours to their ten model from scipy state impert militaron from scipy state impert militaron from scipy state impert militaron from scips state impert militaron impert militaron from spines impert militaron impert militaron impert from spines impert militaron impert militaron from spines impert militaron impert militaron from spines impert militaron impert militaron from spines from spines impert militaron from spines			

1. METRICS:

CLASSIFICATION REPORT:

In [52]:	#computing th	ne classifica	tion repo	ort of the i	nodel
	<pre>print(metrics.classification_report(y_test, y_test_gbc))</pre>				
		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

```
In [89]: from mlxtend.evaluate import combined_ftest_5x2cv
         from sklearn.tree import DecisionTreeClassifier, ExtraTreeClassifier
         from sklearn.ensemble import GradientBoostingClassifier
         from mlxtend.data import iris_data
         # Prepare data and clfs
         X, y = iris_data()
         clf1 = GradientBoostingClassifier()
         clf2 = DecisionTreeClassifier()
         # Calculate p-value
         f, p = combined_ftest_5x2cv(estimator1=clf1,
                                   estimator2=clf2,
                                   X=X, y=y,
                                   random_seed=1)
         print('f-value:', f)
         print('p-value:', p)
         f-value: 1.727272727272733
         p-value: 0.2840135734291782
```