Project Development Phase Model Performance Test

Date	13 November 2022
Team ID	PNT2022TMID48973
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.

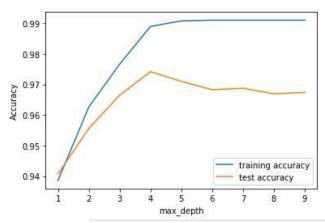
S.No.	Parameter	Values	Screenshot				
1.	Metrics	Classification Model: Gradient Boosting Classification Accuray Score- 97.4%	In [52]: **computing the classification report of the model print(metrics.classification_report(y_test_y_t_cest_gbc)) practision recall fi-score support 1 0.99 0.56 0.97 978 1 0.77 0.99 0.58 0.235 accuracy 0 0.90 0.79 2211 metero ag 0.00 0.97 0.79 2211 suggleted ang 0.77 0.77 0.79 0.77 2211				
2.	Tune the Model	Hyperparameter Tuning - 97% Validation Method – KFOLD & Cross Validation Method	Wilcoxon signed-rank test le [70] #87620 and Cross testination Americ for a size, relate inpert ultimose for a size, relate inpert ultimose for a size and a size				

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

```
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
```