Project Development Phase Model Performance Test

Date	10 NOvember 2022	
Team ID	PNT2022TMID15570	
Project Name	Project - Developing a Flight Delay Prediction Model using Machine Learning	
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	Metrics	Regression Model:	Classification Report
		MAE - , MSE - , RMSE - , R2 score - Classification Model: Confusion Matrix - , Accuray Score- & Classification Report -	<pre>print(classification_report(Y_test, Y_pred_twn_test)) precision recall f1-score support</pre>
			0.0 0.5% 0.5% 0.5% 1974 1.0 0.57 0.72 0.59 273 sccurety macro and 0.81 0.84 0.82 2287 weighted avg 0.93 0.92 0.92 2287
			Accuracy, Precision, Recall, F1 Score
			<pre>acctum = accuracy_score(\text{test}, Y_pred_tum_test) prec_tum, rec_tum_f_tum_sup_um_precision_recall_fscore_support(Y_test, Y_pred_tum_test) print('Recuracy Score ', acc_tum) print('Precision ', prec_tum(D)) print('Recall ', rec_tum(D)) print('Recall ', rec_tum(D))</pre>
			Accuracy Score = 0.922118800623953 Prestsion = 0.9503493573318935 Recall = 0.948849627361894 F1 Score = 0.9554140127388534
			Checking for Overfitting and Underfitting
			<pre>tun_train_acc = accuracy_score('\texture')_train, Y_pred_tun_train) tun_text_acc = accuracy_score('\text_Y_pred_tun_text) print('\texture')_naccuracy = '\texture', tun_train_acc) print('\texture')_train_accuracy = '\texture', tun_text_acc)</pre>
			Training Accuracy = 0.9194122885129118 Testing Accuracy = 0.9221183800623053
			Confusion Matrix pd.crosstab(Y_test.rave1(), Y_pred_tum_test)
			ex(2 - 0.0 - 1.0 exx(3 - 0.0 - 1.0 - 0.0 - 1.0 - 1.0 - 1.0 - 70 - 107
2.	Tune the Model	Hyperparameter Tuning - Validation Method -	Tuning the Hyper Parameters of Logistic Regression
			parameters * ('solwe:'['mexton-cg', 'lbfgg', 'lblinear'], 'C':[80, [0, 1, 0, 0.1, 0.01], 'pesalty':['12'])
			<pre>tuned_model = GridGearch(V[(ogisticRegression(max_iter=800), param_gridoparameters, verbose=2) tuned_model.fit(X_train,train.reve())</pre>
			Fitting 5 folia for each of 15 condidates, totalling 75 fits [or] 800
			<pre>Y_pred_tun_train = tuned_model.predict(X_train) Y_pred_tun_test = tuned_model.predict(X_test)</pre>
			pd.DataFrame(Y_pred_tun_train).value_counts()
			0.0 7772 1.0 1212 dtype: int64
			pd.DataFrame(Y_pred_tun_test).value_counts()
			0.0 1951 1.0 296 dtype: int64

Evaluating the Tuned Model using Metrics
Classification Report
<pre>print(classification_report(Y_test, Y_pred_tun_test))</pre>
precision recall fi-score support
0.0 0.96 0.95 0.96 1974 1.0 0.67 0.72 0.69 273
accuracy 0.92 2247 macro avg 0.81 0.84 0.82 2247 weighted avg 0.93 0.92 0.92 2247
Accuracy, Precision, Recall, F1 Score
<pre>acc_tun = accuracy_score(Y_test, Y_pred_tun_test) prec_tun, rec_tun, fl_tun, sup_tun = precision_recall_fscore_support(Y_test, Y_pred_tun_test) print('Accuracy Score =', acc_tun) print('Precision =', prec_tun[0]) print('Recall =', rec_tun[0]) print('Fl Score =', fl_tun[0])</pre>
Accuracy Score = 0.922183800523053 Precision = 0.9510456176319835 Recall = 0.496480243161094 Fl Score = 0.9554140127386534
Checking for Overfitting and Underfitting
tm_train_acc = accoray_assor(\tau_train, \tau_ped_tm_train) tm_test_acc = accorate_terror(\tau_train) tm_test_acc = accorate_terror(\tau_train) print(\tau_train_acc) = \tau_train_acc) print(\tau_train_acc) = \tau_train_acc)
Training Accuracy = 0.3194122881229813 Testing Accuracy = 0.222118388023933 Testing Accuracy = 0.222118388023933
There is no big variation in the training and testing accuracy. Therefore, the Tuned Logistic Regression model is not overfit or underfit.
Confusion Matrix
pd.crosstab(Y_test.ravel(), Y_pred_ten_test) ext0 00 10
900.03 90 80 0005 90 10 76 197