Project Development Phase Model Performance Test

Date	18 November 2022
Team ID	PNT2022TMID52891
Project Name	Project - Developing a Flight Delay Prediction Model using Machine Learning
Maximum Marks	10 Marks

Model Performance Testing:

S.No.	Parameter	Values	Screenshot				
1.	Metrics	Classification Model: Confusion Matrix - , Accuray Score- & Classification	<pre>Classification Report print(classification_report(Y_test, Y_pred_log_test))</pre>				
			0.0	0.96	0.94	0.95	1985
			1.0	0.60	0.73	0.66	262
			accuracy			0.91	2247
			macro avg	9.78	0.83		
			weighted avg		0.91		2247
			accuracy, Precision, Rec : acc_log = accuracy_scc prec_log, rec_log, fl print('Accuracy Score print('Precision = , , print('Recall = ', rec print('F1 Score = ', f1 Accuracy Score = 0.912	re(Y_test, Y_pred_ log, sup_log = pre =', acc_log) rec_log[0]) log[0]) _log[0])	g = precision_recall_fscore_support(Y_t)		test, Y_pred_log_
			Precision = 0.96323148 Recall = 0.93702770786 F1 Score = 0.949948927	862765406 985643			
			Checking for Overfitting and Underfitting				
			<pre>log_train_acc = log_test_acc = print('Training print('Testing</pre>	;, Y_pred_lo min_acc)			
			Training Accuracy = 0.9205253784505788 Testing Accuracy = 0.9127725856697819				

			Confusion Matrix		
			<pre>pd.crosstab(Y_test.ravel(), Y_pred_log_test)</pre>		
			col_0 0.0 1.0		
			row_0		
			0.0 1860 125		
			1.0 71 191		
2.	Tune the Model	Hyperparameter Tuning - Validation Method -	Tuning the Hyper Parameters of Logistic Regression		
			<pre>parameters = { 'solver':['newton-cg', 'lbfgs', 'liblinear']</pre>		
			<pre>In [57]: tuned_model = GridSearchCV(LogisticRegression(max_iter=800), param_grid=parameters, verbose=2) tuned_model.fit(X_train, Y_train.ravel())</pre>		
			Out[57]: GridSearchCV(estimator=LogisticRegression(max_iter=800), param_grid={'C': [100, 10, 1.0, 0.1, 0.01], 'penalty': ['12'], 'solver': ['newton-cg', 'lbfgs', 'liblinear']}, verbose=2)		
			Testing the Tuned Model		
			Y_pred_tun_train = tuned_model.predict(X_train) Y_pred_tun_test = tuned_model.predict(X_test)		
			pd.DataFrame(Y_pred_tun_train).value_counts()		
			0.0 7734 1.0 1250 dtype: int64		
			pd.DataFrame(Y_pred_tun_test).value_counts()		
			0.0 1922 1.0 325 dtype: int64		
			acyper incor		

Evaluating the Tuned Model using Metrics

Classification Report

<pre>print(classification_report(Y_test, Y_pred_tun_test))</pre>				
	precision	recall	f1-score	support
0.0	0.97	0.94	0.95	1985
1.0	0.61	0.76	0.68	262
accuracy			0.92	2247
macro avg	0.79	0.85	0.81	2247
weighted avg	0.93	0.92	0.92	2247

Accuracy, Precision, Recall, F1 Score

```
: acc_tun = accuracy_score(Y_test, Y_pred_tun_test)
prec_tun, rec_tun, f1_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('F1 Score =', f1_tun[0])
```

Accuracy Score = 0.9158878504672897 Precision = 0.9672216441207075 Recall = 0.9365239294710328 F1 Score = 0.9516252879447147

Checking for Overfitting and Underfitting

```
tun_train_acc = accuracy_score(Y_train, Y_pred_tun_train)
tun_test_acc = accuracy_score(Y_test, Y_pred_tun_test)
print('Training Accuracy =', tun_train_acc)
print('Testing Accuracy =', tun_test_acc)
```

Training Accuracy = 0.9213045414069457 Testing Accuracy = 0.9158878504672897

Confusion Matrix

```
pd.crosstab(Y_test.ravel(), Y_pred_tun_test)
```

col_0 0.0 1.0

row_0

0.0 1859 126

1.0 63 199