Project Development Phase Model Performance Test

Date	18 November 2022
Team ID	PNT2022TMID44312
Project Name	Developing a Flight Delay Prediction Model using Machine Learning
Maximum Marks	10 Marks

Model Performance Testing:

S.No.	Parameter	Values	Screenshot				
1. Metrics	Metrics	Classification Model: Confusion Matrix - , Accuray	Classification Report				
	Score- & Classification	<pre>print(classification_report(Y_test, Y_pred_log_test))</pre>					
		Report -		precision	recall	f1-score	support
			\$5,000	0.96			1985
			1.6	0.60	0.73	0.66	262
			accuracy	,		0.91	2247
			macro av		0.83	(A) 2 (A)	2247
			weighted av	g 0.92	0.91	0.92	2247
			Accuracy, Precision,	Recall, F1 Score			
			<pre>print('Accuracy Sci print('Precision = print('Recall =', print('F1 Score =' Accuracy Score = 0</pre>	f1_log, sup_log = prope =', acc_log) ', prec_log[0]) rec_log[0]) rec_log[0]) .9127725856697819		fscore_support(Y_	test, Y_pred_log_test)
		Precision = 0.9632 Recall = 0.9370277 F1 Score = 0.94994	078085643				
		Checking for Overfitting and Underfitting log_train_acc = accuracy_score(Y_train, Y_pred_log_train) log_test_acc = accuracy_score(Y_test, Y_pred_log_test) print('Training Accuracy =', log_train_acc) print('Testing Accuracy =', log_test_acc)					

			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>	
			<pre>Out[57]: GridSearchCV(estimator=LogisticRegression(max_iter=800),</pre>
			Testing the Tuned Model
			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()</pre>
			0.0 7734
			1.0 1250 dtype: int64
			pd.DataFrame(Y_pred_tun_test).value_counts()
			0.0 1922 1.0 325 dtype: int64

Evaluating the Tuned Model using Metrics
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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	
accur	acy			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