

Project Development Phase Model Performance Test

Date	03 November 2023
Team ID	592974
Project Name	Airline Review Classification Using ML
Maximum Marks	10 Marks

Model Performance Testing:

S.No.	Parameter	Values	Screenshot																																													
1	Metrics	<p>Classification Model(KNN):</p> <p>Confusion Matrix - [[2861, 192], [156, 2937]]</p> <p>Accuracy Score- 0.943</p> <p>Classification Report -</p> <table><thead><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr></thead><tbody><tr><td>0</td><td>0.95</td><td>0.94</td><td>0.94</td><td>3053</td></tr><tr><td>1</td><td>0.94</td><td>0.95</td><td>0.94</td><td>3093</td></tr></tbody></table>		precision	recall	f1-score	support	0	0.95	0.94	0.94	3053	1	0.94	0.95	0.94	3093	<pre>model5=KNeighborsClassifier(n_neighbors=10) model5.fit(x_train,y_train) KNeighborsClassifier KNeighborsClassifier(n_neighbors=10) pred5=model5.predict(x_test) print(classification_report(y_test,pred5)) print(accuracy_score(y_test,pred5)) print(confusion_matrix(y_test,pred5)) print(roc_auc_score(y_test,pred5))</pre> <table><thead><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr></thead><tbody><tr><td>0</td><td>0.95</td><td>0.94</td><td>0.94</td><td>3053</td></tr><tr><td>1</td><td>0.94</td><td>0.95</td><td>0.94</td><td>3093</td></tr></tbody></table> <table><tbody><tr><td>accuracy</td><td></td><td></td><td>0.94</td><td>6146</td></tr><tr><td>macro avg</td><td>0.94</td><td>0.94</td><td>0.94</td><td>6146</td></tr><tr><td>weighted avg</td><td>0.94</td><td>0.94</td><td>0.94</td><td>6146</td></tr></tbody></table>		precision	recall	f1-score	support	0	0.95	0.94	0.94	3053	1	0.94	0.95	0.94	3093	accuracy			0.94	6146	macro avg	0.94	0.94	0.94	6146	weighted avg	0.94	0.94	0.94	6146
	precision	recall	f1-score	support																																												
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2	Tune the Model	Hyperparameter Tuning - Default Validation Method - Default																																														

For all the models:

	A	B	C	D	E	F	G	H	I
1	Classification Model	Accuracy Score	Precision	Recall	F1-Score	True Positives	True Negatives	False Positives	False Negatives
2	DecisionTreeClassifier	0.847	0.79	0.95	0.86	2253	2759	800	142
3	RandomForestClassifier	0.895	0.84	0.97	0.9	2491	3010	562	83
4	LogisticRegression	0.878	0.87	0.89	0.88	2639	2759	414	334
5	GaussianNB	0.855	0.87	0.84	0.85	2664	2589	389	504
6	KNeighborsClassifier	0.943	0.94	0.95	0.94	2861	2937	192	156
7	SVC	0.929	0.92	0.94	0.93	2813	2899	240	194
8	XGBClassifier	0.912	0.89	0.94	0.92	2701	2907	352	186

KNN performed best of all, so we used it in deployment.

Screenshots for all the models:

DecisionTreeClassifier:

```
DecisionTreeClassifier
DecisionTreeClassifier(criterion='entropy', random_state=0)
```

```
pred1=model1.predict(x_test)
pred1
```

```
array([1, 1, 0, ..., 0, 1, 1])
```

```
from sklearn.metrics import classification_report, confusion_matrix, accuracy_score, roc_auc_score
print(classification_report(y_test, pred1))
print(accuracy_score(y_test, pred1))
print(confusion_matrix(y_test, pred1))
print(roc_auc_score(y_test, pred1))
```

	precision	recall	f1-score	support
0	0.94	0.74	0.83	3053
1	0.79	0.95	0.86	3093
accuracy			0.85	6146
macro avg	0.86	0.85	0.84	6146
weighted avg	0.86	0.85	0.84	6146

```
0.8467295802147738
```

```
[[2253 800]
```

```
 [ 142 2951]]
```

```
0.8460262700270225
```

RandomForestClassifier:

```
RandomForestClassifier
RandomForestClassifier(criterion='entropy', n_estimators=20, random_state=0)
```

```
pred2=model2.predict(x_test)
pred2
```

```
array([1, 1, 0, ..., 0, 1, 1])
```

```
print(classification_report(y_test,pred2))
print(accuracy_score(y_test,pred2))
print(confusion_matrix(y_test,pred2))
print(roc_auc_score(y_test,pred2))
```

	precision	recall	f1-score	support
0	0.97	0.82	0.89	3053
1	0.84	0.97	0.90	3093
accuracy			0.90	6146
macro avg	0.91	0.89	0.89	6146
weighted avg	0.90	0.90	0.89	6146

```
0.8950536934591604
```

```
[[2491 562]
```

```
 [ 83 3010]]
```

```
0.894541990096505
```

LogisticRegression:

▼ LogisticRegression

LogisticRegression()

```
pred3=model3.predict(x_test)
```

```
print(classification_report(y_test,pred3))
print(accuracy_score(y_test,pred3))
print(confusion_matrix(y_test,pred3))
print(roc_auc_score(y_test,pred3))
```

	precision	recall	f1-score	support
0	0.89	0.86	0.88	3053
1	0.87	0.89	0.88	3093
accuracy			0.88	6146
macro avg	0.88	0.88	0.88	6146
weighted avg	0.88	0.88	0.88	6146

0.8782948259030263

[[2639 414]

[334 2759]]

0.8782049510273772

GaussianNB:

```
model4=GaussianNB()  
model4.fit(x_train,y_train)
```

```
▼ GaussianNB  
GaussianNB()
```

```
pred4=model4.predict(x_test)
```

```
print(classification_report(y_test,pred4))  
print(accuracy_score(y_test,pred4))  
print(confusion_matrix(y_test,pred4))  
print(roc_auc_score(y_test,pred4))
```

	precision	recall	f1-score	support
0	0.84	0.87	0.86	3053
1	0.87	0.84	0.85	3093
accuracy			0.85	6146
macro avg	0.86	0.85	0.85	6146
weighted avg	0.86	0.85	0.85	6146

```
0.8547022453628376
```

```
[[2664 389]
```

```
 [ 504 2589]]
```

```
0.8548178748352337
```

KNeighborsClassifier:

```
▼ KNeighborsClassifier  
KNeighborsClassifier(n_neighbors=10)
```

```
pred5=model5.predict(x_test)
```

```
print(classification_report(y_test,pred5))  
print(accuracy_score(y_test,pred5))  
print(confusion_matrix(y_test,pred5))  
print(roc_auc_score(y_test,pred5))
```

	precision	recall	f1-score	support
0	0.95	0.94	0.94	3053
1	0.94	0.95	0.94	3093
accuracy			0.94	6146
macro avg	0.94	0.94	0.94	6146
weighted avg	0.94	0.94	0.94	6146

```
0.943377806703547  
[[2861 192]  
 [ 156 2937]]  
0.9433372844379111
```

SVC:

```
model6= SVC()  
model6.fit(x_train,y_train)
```

▼ SVC
SVC()

```
pred6=model6.predict(x_test)
```

```
print(classification_report(y_test,pred6))  
print(accuracy_score(y_test,pred6))  
print(confusion_matrix(y_test,pred6))  
print(roc_auc_score(y_test,pred6))
```

	precision	recall	f1-score	support
0	0.94	0.92	0.93	3053
1	0.92	0.94	0.93	3093
accuracy			0.93	6146
macro avg	0.93	0.93	0.93	6146
weighted avg	0.93	0.93	0.93	6146

```
0.929384965831435  
[[2813 240]  
 [ 194 2899]]  
0.9293332608981811
```

XGBClassifier:

```
pred7=model7.predict(x_test)
```

```
print(classification_report(y_test,pred7))
print(accuracy_score(y_test,pred7))
print(confusion_matrix(y_test,pred7))
print(roc_auc_score(y_test,pred7))
```

	precision	recall	f1-score	support
0	0.94	0.88	0.91	3053
1	0.89	0.94	0.92	3093
accuracy			0.91	6146
macro avg	0.91	0.91	0.91	6146
weighted avg	0.91	0.91	0.91	6146

0.9124633908232997

[[2701 352]

[186 2907]]

0.9122838898820482