

## Project Development Phase Model Performance Test

Date	15 November 2022
Team ID	PNT2022TMID40372
Project Name	Statistical Machine Learning Approaches to Liver Disease Prediction
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	<b>Classification Model:</b>  Confusion Matrix, Accuracy Score, Classification Report – Random Forest	<pre>In [27]: from sklearn.metrics import accuracy_score, confusion_matrix, classification_report</pre> <pre>In [28]: from sklearn.ensemble import RandomForestClassifier           RandomForest = RandomForestClassifier()           RandomForest = RandomForest.fit(X_train,y_train)            y_pred = RandomForest.predict(X_test)            print('Accuracy:', accuracy_score(y_test,y_pred))           print(confusion_matrix(y_test,y_pred))           print(classification_report(y_test,y_pred))</pre> <pre>Accuracy: 0.6929824561403509 [[13 25]  [10 66]]</pre> <pre>               precision    recall  f1-score   support        0       0.57      0.34      0.43       38       1       0.73      0.87      0.79       76   accuracy          0.69       114  macro avg       0.65       0.61       0.61       114  weighted avg    0.67       0.69       0.67       114</pre>

	Metrics	<b>Classification Model:</b>  Confusion Matrix, Accuracy Score, Classification Report-KNN	<div>In [30]: <pre>from sklearn.neighbors import KNeighborsClassifier knn_classifier = KNeighborsClassifier() knn_classifier = knn_classifier.fit(X_train, y_train)  y_pred = knn_classifier.predict(X_test)  print('Accuracy:', accuracy_score(y_test,y_pred)) print(confusion_matrix(y_test,y_pred)) print(classification_report(y_test,y_pred))</pre></div> <div>Accuracy: 0.5964912280701754 [[13 25]  [21 55]]</div> <table><tr><th></th><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td></td><td>0</td><td>0.38</td><td>0.34</td><td>0.36</td><td>38</td></tr><tr><td></td><td>1</td><td>0.69</td><td>0.72</td><td>0.71</td><td>76</td></tr><tr><td></td><td>accuracy</td><td></td><td></td><td>0.60</td><td>114</td></tr><tr><td></td><td>macro avg</td><td>0.53</td><td>0.53</td><td>0.53</td><td>114</td></tr><tr><td></td><td>weighted avg</td><td>0.59</td><td>0.60</td><td>0.59</td><td>114</td></tr></table>			precision	recall	f1-score	support		0	0.38	0.34	0.36	38		1	0.69	0.72	0.71	76		accuracy			0.60	114		macro avg	0.53	0.53	0.53	114		weighted avg	0.59	0.60	0.59	114
		precision	recall	f1-score	support																																		
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	Metrics	<b>Classification Model:</b>  Confusion Matrix, Accuracy Score, Classification Report-SVM	<div>In [32]: <pre>from sklearn.svm import SVC svm_classifier = SVC() svm_classifier = svm_classifier.fit(X_train, y_train)  y_pred = svm_classifier.predict(X_test)  print('Accuracy:', accuracy_score(y_test,y_pred)) print(confusion_matrix(y_test,y_pred)) print(classification_report(y_test,y_pred))</pre></div> <div>Accuracy: 0.6666666666666666 [[ 0 38]  [ 0 76]]</div> <table><tr><th></th><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td></td><td>0</td><td>0.00</td><td>0.00</td><td>0.00</td><td>38</td></tr><tr><td></td><td>1</td><td>0.67</td><td>1.00</td><td>0.80</td><td>76</td></tr><tr><td></td><td>accuracy</td><td></td><td></td><td>0.67</td><td>114</td></tr><tr><td></td><td>macro avg</td><td>0.33</td><td>0.50</td><td>0.40</td><td>114</td></tr><tr><td></td><td>weighted avg</td><td>0.44</td><td>0.67</td><td>0.53</td><td>114</td></tr></table>			precision	recall	f1-score	support		0	0.00	0.00	0.00	38		1	0.67	1.00	0.80	76		accuracy			0.67	114		macro avg	0.33	0.50	0.40	114		weighted avg	0.44	0.67	0.53	114
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2.	Tune the Model	Hyperparameter Tuning - Grid SearchCV, finding best estimators for each algorithm in ensemble model Validation Method -cross validation	<div><p>ROC AUC</p><table><tr><th>Model</th><th>ROC AUC</th></tr><tr><td>Decision Tree</td><td>86.64%</td></tr><tr><td>Random Forest</td><td>94.08%</td></tr><tr><td>SVM</td><td>93.85%</td></tr><tr><td>KNN</td><td>66.11%</td></tr><tr><td>ANN</td><td>50.0%</td></tr></table></div>	Model	ROC AUC	Decision Tree	86.64%	Random Forest	94.08%	SVM	93.85%	KNN	66.11%	ANN	50.0%
Model	ROC AUC														
Decision Tree	86.64%														
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3.			<div><pre>In [ ]: from sklearn.model_selection import GridSearchCV  # Create the parameter grid based on the results of random search param_grid = {     'bootstrap': [True],     'max_depth': [10,15],     'max_features': [2, 3],     'min_samples_leaf': [3, 4, 5,6],     'min_samples_split': [3,4,5,6],     'n_estimators': [1150, 1200, 1250, 1300,1350] }  # Create a based model rf = RandomForestClassifier()  # Instantiate the grid search model grid_search = GridSearchCV(estimator = rf, param_grid = param_grid,                            cv = 3, n_jobs = -1, verbose = 2)  # Fit the grid search to the data grid_search.fit(x_train,y_train)</pre></div> <div>best parameter from gridseachCV</div>												