

Project Development Phase Performance Test

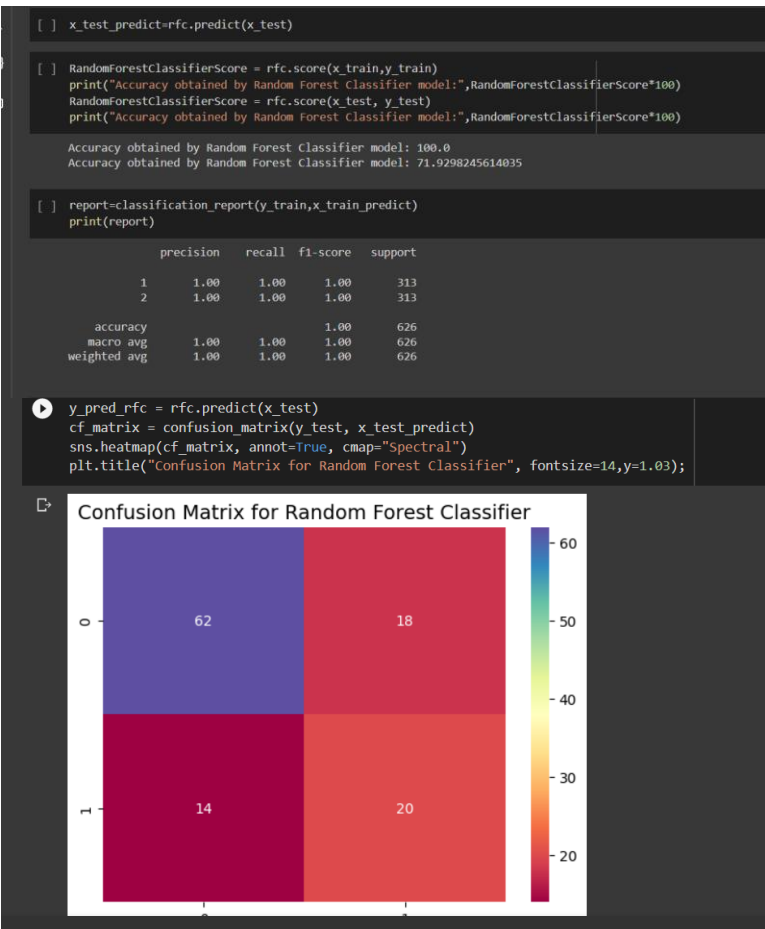
Date	22 November 2022
Team ID	PNT2022TMID00067
Project Name	Statistical Machine Learning Approaches for Liver Disease Prediction
Maximum Marks	10 Marks

Model Performance Testing:

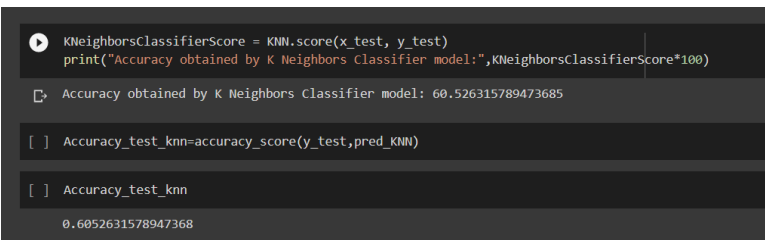
Project team shall fill the following information in model performance testing template.

S No	Parameter	Values	Screenshot																														
1.	Metrics	Classification Model: Confusion Matrix , Accuracy Score & Classification Report	Logistic Regression <div><div>▾ Accuracy Of Logistic Regression</div><div><div><div><div>▶</div><div>Accuracy_test_log=accuracy_score(y_test,pred_lr)</div><div>Accuracy_test_log</div></div><div>0.6929824561403509</div></div><div><div><div><div>▶</div><div>report_test_log=classification_report(y_test, pred_lr)</div><div>print(report_test_log)</div></div><table><thead><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr></thead><tbody><tr><td>1</td><td>0.94</td><td>0.60</td><td>0.73</td><td>80</td></tr><tr><td>2</td><td>0.49</td><td>0.91</td><td>0.64</td><td>34</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.69</td><td>114</td></tr><tr><td>macro avg</td><td>0.72</td><td>0.76</td><td>0.69</td><td>114</td></tr><tr><td>weighted avg</td><td>0.81</td><td>0.69</td><td>0.70</td><td>114</td></tr></tbody></table></div></div><div><div>[] Accuracy_train_lr</div><div>0.7060702875399361</div><div><div>[] confusion_matrix(y_test,pred_lr)</div><div>array([[48, 32], [3, 31]])</div></div></div></div></div>		precision	recall	f1-score	support	1	0.94	0.60	0.73	80	2	0.49	0.91	0.64	34	accuracy			0.69	114	macro avg	0.72	0.76	0.69	114	weighted avg	0.81	0.69	0.70	114
	precision	recall	f1-score	support																													
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accuracy			0.69	114																													
macro avg	0.72	0.76	0.69	114																													
weighted avg	0.81	0.69	0.70	114																													

Random Forest Classifier



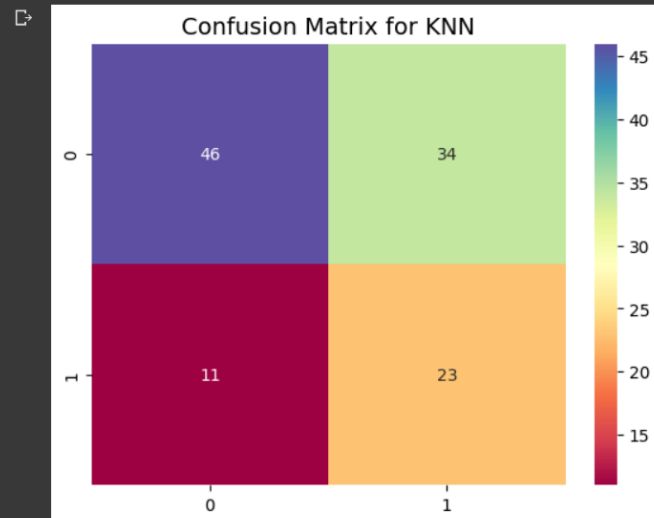
KNN



```

cf_matrix = confusion_matrix(y_test, pred_KNN)
sns.heatmap(cf_matrix, annot=True, cmap="Spectral")
plt.title("Confusion Matrix for KNN", fontsize=14,y=1.03);

```



Decision Tree Classifier

```

[ ] Accuracy_test_DT=accuracy_score(y_test,x_test_pred_DT)
Accuracy_test_DT

```

0.6754385964912281

```

[ ] report_test_DT=classification_report(y_test, x_test_pred_DT)#
print(report_test_DT)

```

	precision	recall	f1-score	support
1	0.78	0.75	0.76	80
2	0.46	0.50	0.48	34
accuracy			0.68	114
macro avg	0.62	0.62	0.62	114
weighted avg	0.68	0.68	0.68	114

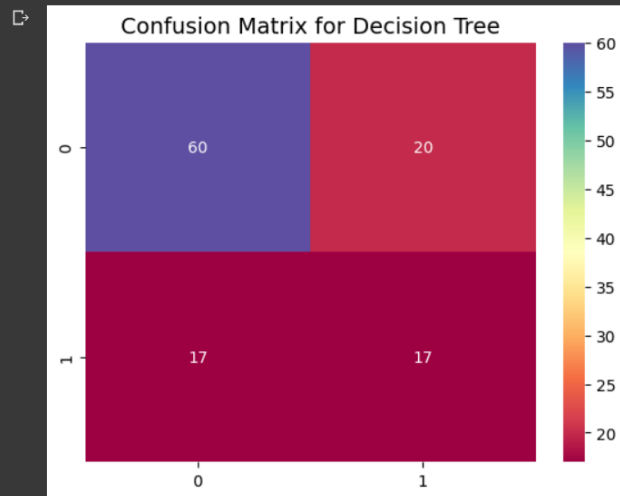
```

[ ] print('Testing score for DT regression is',DT.score(x_test,y_test))

```

Testing score for DT regression is 0.6754385964912281

```
plt.title("Confusion Matrix for Decision Tree", fontsize=14,y=1.03);
```



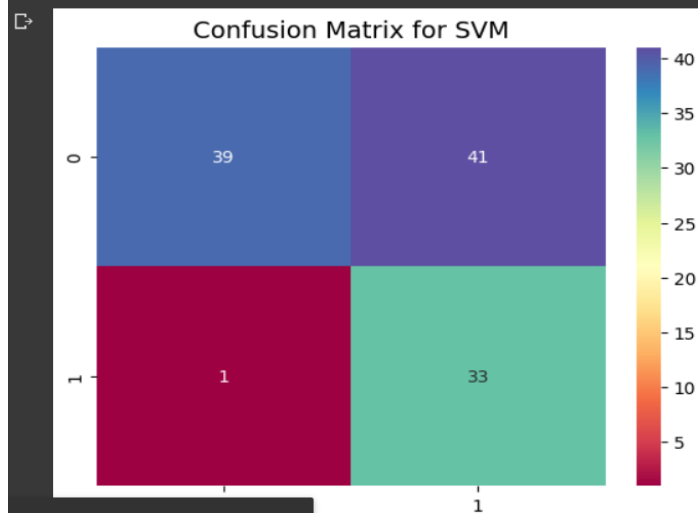
Support Vector Machine

```
[75] x_test_predict=svclassifier.predict(x_test)
```

```
[76] report=classification_report(y_test,x_test_predict)
print(report)
```

	precision	recall	f1-score	support
1	0.97	0.49	0.65	80
2	0.45	0.97	0.61	34
accuracy			0.63	114
macro avg	0.71	0.73	0.63	114
weighted avg	0.82	0.63	0.64	114

```
cf_matrix = confusion_matrix(y_test, x_test_predict)
sns.heatmap(cf_matrix, annot=True, cmap="Spectral")
plt.title("Confusion Matrix for SVM", fontsize=14,y=1.03);
```



			Conclusion Random Forest Classifier as the highest accuracy but it is overfitting the training data. Therefore, we used Logistic Regression model which gives second best accuracy.
2.	Tune the Model	Hyperparameter Tuning, Validation Method	Grid Search CV For Logistic Regression <pre> Hyperparameter Tuning [] from sklearn.model_selection import GridSearchCV [] model = LogisticRegression() solvers = ['newton-cg', 'lbfgs', 'liblinear'] penalty = ['l2'] c_values = [100, 10, 1.0, 0.1, 0.01] # define grid search grid = dict(solver=solvers,penalty=penalty,C=c_values) [] grid_search = GridSearchCV(estimator=model, param_grid=grid, n_jobs=-1, scoring='accuracy',error_score=0) grid_result = grid_search.fit(x_train, y_train) # summarize results print("Best: %f using %s" % (grid_result.best_score_, grid_result.best_params_)) means = grid_result.cv_results_['mean_test_score'] stds = grid_result.cv_results_['std_test_score'] params = grid_result.cv_results_['params'] for mean, stdev, param in zip(means, stds, params): print("%f (%f) with: %r" % (mean, stdev, param)) Best: 0.702933 using {'C': 100, 'penalty': 'l2', 'solver': 'newton-cg'} 0.702933 (0.039790) with: {'C': 100, 'penalty': 'l2', 'solver': 'newton-cg'} 0.702933 (0.039790) with: {'C': 100, 'penalty': 'l2', 'solver': 'lbfgs'} 0.702933 (0.039790) with: {'C': 100, 'penalty': 'l2', 'solver': 'liblinear'} 0.699733 (0.039575) with: {'C': 10, 'penalty': 'l2', 'solver': 'newton-cg'} 0.699733 (0.039575) with: {'C': 10, 'penalty': 'l2', 'solver': 'lbfgs'} 0.696546 (0.040268) with: {'C': 10, 'penalty': 'l2', 'solver': 'liblinear'} 0.693333 (0.015540) with: {'C': 1.0, 'penalty': 'l2', 'solver': 'newton-cg'} 0.693333 (0.015540) with: {'C': 1.0, 'penalty': 'l2', 'solver': 'lbfgs'} 0.690146 (0.018940) with: {'C': 1.0, 'penalty': 'l2', 'solver': 'liblinear'} {'C': 100, 'penalty': 'l2', 'solver': 'newton-cg'} ▶ model_lor= LogisticRegression(C=0.01, penalty= 'l2', solver= 'liblinear') model_lor.fit(x_train,y_train) LogisticRegression(C=0.01, solver='liblinear') [] x_test_predict_lor=model_lor.predict(x_test) [] accuracy = accuracy_score(y_test,x_test_predict_lor) accuracy 0.6403508771929824 </pre> Grid Search CV For SVM

			<pre>[] parameters = {'kernel':['linear','rbf'], 'C':[0.1,0.5,1.0], 'gamma':['scale','auto']} [] clf=GridSearchCV(SVC(),param_grid=parameters,verbose=2) [] clf.fit(x_train,y_train) Fitting 5 folds for each of 12 candidates, totalling 60 fits [CV] ENDC=0.1, gamma=scale, kernel=linear; total time= 0.0s [CV] ENDC=0.1, gamma=scale, kernel=linear; total time= 0.0s [CV] ENDC=0.1, gamma=scale, kernel=linear; total time= 0.0s [CV] ENDC=0.1, gamma=scale, kernel=linear; total time= 0.0s</pre>
			<pre>[] clf.best_score_ 0.6901079365079363 [] clf.best_params_ {'C': 1.0, 'gamma': 'scale', 'kernel': 'linear'} [] model_svc=SVC(C=1,gamma='scale',kernel='linear') model_svc.fit(x_train,y_train) SVC(C=1, kernel='linear') [] y_=model_svc.predict(x_test) [] accuracy_score(y_test,y_) 0.6578947368421053</pre>
			<pre>[] pd.crosstab(y_test,y_) col_0 1 2 Dataset 1 42 38 2 1 33 [] print(classification_report(y_test,y_)) precision recall f1-score support 1 0.98 0.53 0.68 80 2 0.46 0.97 0.63 34 accuracy 0.66 macro avg 0.72 weighted avg 0.82</pre>