

## Project Development Phase Model Performance Test

|               |                                                                               |
|---------------|-------------------------------------------------------------------------------|
| Date          | 18 November 2022                                                              |
| Team ID       | PNT2022TMID20879                                                              |
| Project Name  | Project - Statistical Machine learning Approaches to Liver Disease Prediction |
| Maximum Marks | 10 Marks                                                                      |

### Model Performance Testing:

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

➤ **Metrics:**

❖ **Regression Model:**

- *Random Forest:*

**Values:**

r2\_score: -0.38965517241379266

MSE: 0.26495726495726496

MAE: 0.26495726495726496

RMSE: 0.5147399974329419

### Screenshot:

#### Performance Metrics

```
In [15]: from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
```

#### Random Forest

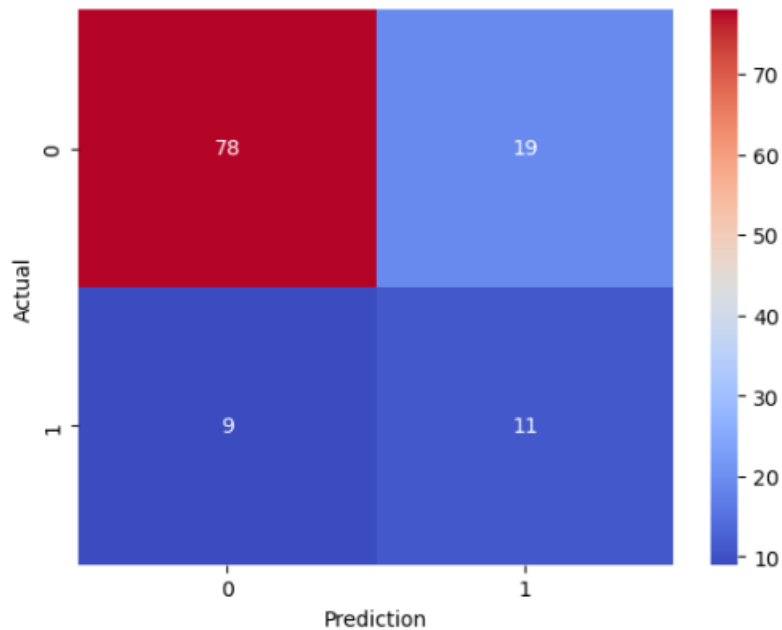
```
In [18]: RFmodel_pred = RFmodel.predict(X_test)
```

```
In [19]: print("r2_score :", r2_score(Y_test, RFmodel_pred))
print("MSE :", mean_squared_error(Y_test, RFmodel_pred))
print("MAE :", mean_absolute_error(Y_test, RFmodel_pred))
print("RMSE :", np.sqrt(mean_squared_error(Y_test, RFmodel_pred)))
```

```
r2_score : -0.38965517241379266
MSE : 0.26495726495726496
MAE : 0.26495726495726496
RMSE : 0.5147399974329419
```

### Confusion Matrix:

```
In [39]: cf=confusion_matrix(RFmodel_pred, Y_test)
sb.heatmap(cf, cmap='coolwarm', annot=True)
plt.xlabel("Prediction")
plt.ylabel("Actual")
plt.show()
```



### Accuracy Score:

```
In [35]: RFmodel_pred = RFmodel.predict(X_test)

In [36]: RFmodel_accuracy = accuracy_score(RFmodel_pred,Y_test)

In [37]: print("RF Accuracy:",RFmodel_accuracy)
RF Accuracy: 0.7606837606837606
```

### ❖ Classification Model:

- *Support Vector Machine Model:*

#### Values:

```
r2_score: -0.34482758620689613
MSE: 0.2564102564102564
MAE: 0.2564102564102564
RMSE: 0.5063696835418333
```

### Screenshot:

#### Performance Metrics

```
In [15]: from sklearn.metrics import mean_absolute_error,mean_squared_error,r2_score
```

#### SVM Model

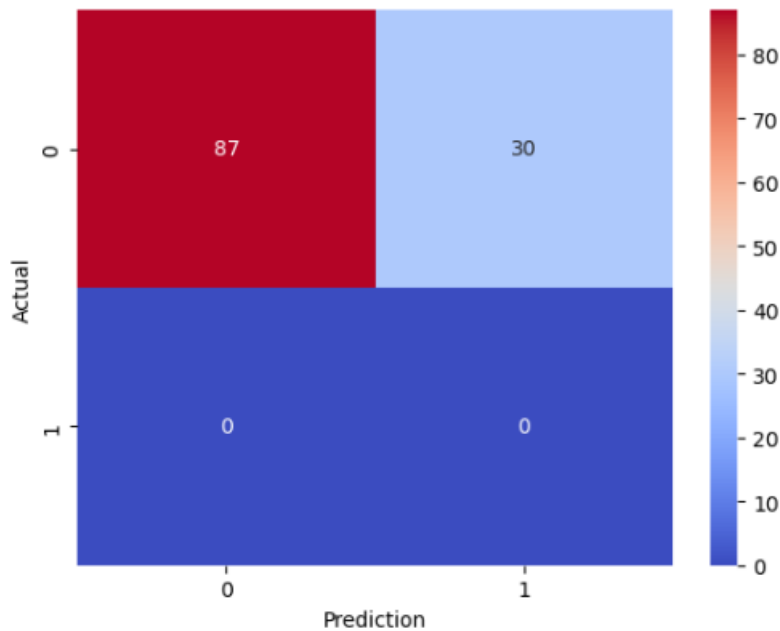
```
In [16]: svm_pred = svm.predict(X_test)
```

```
In [17]: print("r2_score :",r2_score(Y_test,svm_pred))
print("MSE :",mean_squared_error(Y_test,svm_pred))
print("MAE :",mean_absolute_error(Y_test,svm_pred))
print("RMSE :",np.sqrt(mean_squared_error(Y_test,svm_pred)))

r2_score : -0.34482758620689613
MSE : 0.2564102564102564
MAE : 0.2564102564102564
RMSE : 0.5063696835418333
```

## Confusion Matrix:

```
In [34]: cf=confusion_matrix(svm_pred,Y_test)
sb.heatmap(cf,cmap='coolwarm',annot=True)
plt.xlabel("Prediction")
plt.ylabel("Actual")
plt.show()
```



## Accuracy Score:

```
In [30]: svm_pred = svm.predict(X_test)
```

```
In [31]: svm_accuracy = accuracy_score(svm_pred,Y_test)
```

```
In [32]: print("SVM Accuracy:",svm_accuracy)
```

SVM Accuracy: 0.7435897435897436

- **KNN Model:**

### Values:

r2\_score: -0.6586206896551721  
MSE: 0.3162393162393162  
MAE: 0.3162393162393162  
RMSE: 0.5623515948579823

## Screenshot:

### Performance Metrics

```
In [15]: from sklearn.metrics import mean_absolute_error,mean_squared_error,r2_score
```

### KNN Model

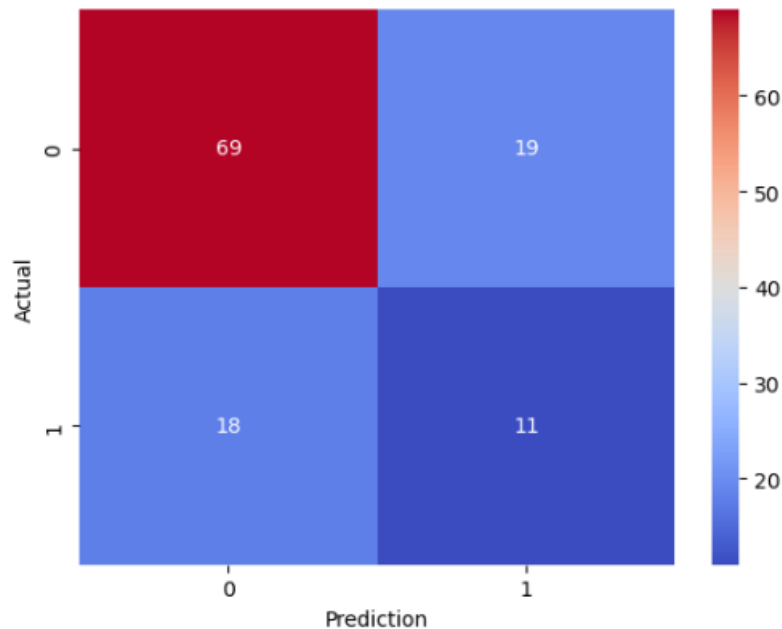
```
In [20]: KNN_pred = KNN.predict(X_test)
```

```
In [21]: print("r2_score :",r2_score(Y_test,KNN_pred))
print("MSE :",mean_squared_error(Y_test,KNN_pred))
print("MAE :",mean_absolute_error(Y_test,KNN_pred))
print("RMSE :",np.sqrt(mean_squared_error(Y_test,KNN_pred)))
```

```
r2_score : -0.6586206896551721
MSE : 0.3162393162393162
MAE : 0.3162393162393162
RMSE : 0.5623515948579823
```

### Confusion Matrix:

```
In [44]: cf=confusion_matrix(KNN_pred,Y_test)
sb.heatmap(cf,cmap='coolwarm',annot=True)
plt.xlabel("Prediction")
plt.ylabel("Actual")
plt.show()
```



### Accuracy Score:

```
In [41]: KNN_accuracy = accuracy_score(KNN_pred,Y_test)
```

```
In [42]: print("KNN Accuracy:",KNN_accuracy)
```

```
KNN Accuracy: 0.6837606837606838
```

### Report:

Best Accuracy Model:

Compare to the three models RF has more accuracy compared to other models. So, RF accuracy is 76% while testing in Jupiter but the actual accuracy of RF is 81% deployed in IBM Cloud Watson Studio.

- Tune the Model:
  - ❖ Hyperparameter Tuning:

## Hyperparameter Tuning

```
In [22]: from sklearn.model_selection import GridSearchCV, RandomizedSearchCV
```

```
In [23]: forest = RandomForestClassifier(random_state = 42)
```

```
In [24]: n_estimators = [10, 15, 5, 8, 20, 50]
max_features = ['auto', 'sqrt', 'log2']
max_depth = [10, 25, 35, 55]
max_depth.append(None)
min_samples_split = [2, 5, 10, 15, 20]
min_samples_leaf = [1, 2, 5, 10, 15]

hyperF = dict(n_estimators = n_estimators, max_features=max_features, max_depth = max_depth,
              min_samples_split = min_samples_split,
              min_samples_leaf = min_samples_leaf)

gridF = RandomizedSearchCV(forest, hyperF, cv = 5, verbose = 2,
                           n_jobs = -1, n_iter=50)
bestF = gridF.fit(X_train, Y_train)
```

Fitting 5 folds for each of 50 candidates, totalling 250 fits

```
C:\Users\Nithiya Devi S\anaconda3\lib\site-packages\sklearn\ensemble\_forest.py:427: FutureWarning: `max_features='auto'` has been deprecated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'` or remove this parameter as it is also the default value for RandomForestClassifiers and ExtraTreesClassifiers.
  warn(
```

```
In [25]: print(bestF.best_params_)
```

```
{'n_estimators': 20, 'min_samples_split': 10, 'min_samples_leaf': 1, 'max_features': 'auto', 'max_depth': 55}
```

```
In [26]: rfe_trial = bestF.predict(X_train)
         rfe_pred = bestF.predict(X_test)
```

```
In [27]: print('Random Forest Hyperparameter:')
          print('\n')
          print(Y_test, rfe_pred)
```

### Random Forest Hyperparameter:

```

355      2
407      1
90       1
402      1
268      1
..
516      1
305      2
167      1
312      2
329      2
Name: Dataset, Length: 117, dtype: int64 [1 1 1 1 1 1 2 1 2 1 2 1 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 2 1 1 1 2 1 1 1 1
1 1 1 1 1 2 1 1 1 1 2 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 2 1 1 2
1 1 2 1 2 1 2 1 1 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 2 2 1 1 1 1 1 1 1 1 1 2
1 2 1 1 1 1]
```

- ❖ Validation Method:

## Validation testing

```
In [28]: from sklearn.model_selection import cross_val_score, cross_val_predict
```

```
In [30]: RFmodel = RandomForestClassifier(random_state = 42)
```

```
In [31]: accuracy = cross_val_score(RFmodel,X_train,Y_train,cv=5,scoring="accuracy")
```

```
In [32]: np.mean(accuracy)
```

Out[32]: 0.6931137039579044

```
In [34]: y_pred = cross_val_predict(RFmodel,X_train,Y_train,cv=5)
         y_pred
```

[illegible]

```
In [39]: sb.distplot(y_pred)
```

```
C:\Users\Withiya Devi S\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).
  warnings.warn(msg, FutureWarning)
```

```
Out[39]: <AxesSubplot:ylabel='Density'>
```

