User Acceptance Testing (UAT)

Date	18 November 2022	
Team ID	PNT2022TMID21605	
Project Name	Visualizing and Predicting Heart	
	Diseases with an Interactive	
	Dashboard	
Maximum Marks	10 Marks	

Purpose of Document

The purpose of this document is to briefly explain the test coverage and open issues of the (Visualizing and Predicting Heart Diseases) project at the time of the release to User Acceptance Testing (UAT).

1. Using ML model predicting whether the person having heart disease or not

Total Number of Test cases tested: 15

Total Number of Test cases passed:12

Total Number of Test cases failed:3

Using ML model predicting whether the person having heart disease or not

```
new_data1=pd.DataFrame({
              'Age':57,
            'Sex':1,
             'Chest pain type':4,
             'BP':140,
             'Cholesterol':192,
              'FBS over 120':0,
              'EKG results':0,
             'Max HR':148,
              'Exercise angina':0,
              'ST depression':0.4,
              'Slope of ST':2,
              'Number of vessels fluro':0,
              'Thallium':6,
         },index=[0])
/ [44] new_data1
     Age Sex Chest pain type BP Cholesterol FBS over 120 EKG results Max HR Exercise angina ST depression Slope of ST Number of vessels fluro Thallium
    0 67 0 3 115 564 0 2 160 0 1.6 2 0 7
         p1 =rf.predict(new data1)
          if p1[0]==0:
            print("disease")
          else:
            print("No Disease")
          No Disease
```

Machine Learning Model Performance Testing

	Model	Accuracy	Confusion Matrix
0	Logistic Regression	0.870370	[[30, 4], [3, 17]]
1	Support Vector Machine	0.740741	[[26, 8], [6, 14]]
2	Random Forest Classifier	0.777778	[[28, 6], [6, 14]]
3	K-Nearest Neighbour	0.740741	[[26, 8], [6, 14]]
4	Gaussian NB	0.851852	[[30, 4], [4, 16]]
5	Linear Discrimant Analysis	0.888889	[[32, 2], [4, 16]]
6	Adaboost	0.740741	[[28, 6], [8, 12]]

Confusing Matrix

Confusion Matrix as the name suggests gives us a matrix as output and describes the complete performance of the model.

Let's assume we have a binary classification problem. We have some samples belonging to two classes: YES or NO. Also, we have our own classifier which predicts a class for a given input sample. On testing our model on 165 samples ,we get the following result.

	Predicted:	Predicted:
n=165	NO	YES
Actual:		
NO	50	10
Actual:		
YES	5	100

The are 4 important terms:

- True Positives: The cases in which we predicted YES and the actual output was also YES.
- True Negatives: The cases in which we predicted NO and the actual output was NO.
- False Positives: The cases in which we predicted YES and the actual output was NO.
- False Negatives: The cases in which we predicted NO and the actual output was YES.
- 1. Logistic Regression

```
[15] scLR=model.score(X_test,Y_test)
scLR
```

0.8703703703703703

```
[17] ##from explainerdashboard.explainers import precision_score

from sklearn.metrics import confusion_matrix,f1_score
sclr1=confusion_matrix(Y_test,model.predict(X_test))

print(sclr1)

[[25  4]
  [ 3 22]]
```

2. Support Vector Machine

```
[19] model1=SVC()
    model1.fit(X train, Y train)
    sc=model1.score(X_test,Y_test)
    0.6851851851851852
[20] sclr2=confusion_matrix(Y_test,model1.predict(X_test))
    sclr2
    array([[25, 4],
           [13, 12]])
3. Random Forest Classifier
[21] model2=RandomForestClassifier()
     model2.fit(X_train,Y_train)
     scrfc=model2.score(X_test,Y_test)
     scrfc
     0.88888888888888888
[22] sclr3=confusion_matrix(Y_test,model2.predict(X_test))
     sclr3
     array([[26, 3],
            [ 3, 22]])
4. K - Nearest Neighbour
23] model3=KNeighborsClassifier()
    model3.fit(X_train,Y_train)
    sckn=model3.score(X_test,Y_test)
    sckn
    0.6111111111111111
24] sclr4=confusion_matrix(Y_test,model3.predict(X_test))
    sclr4
    array([[18, 11],
            [10, 15]])
```

5. Gaussian NB

```
[25] model4=GaussianNB()
       model4.fit(X_train,Y_train)
       scnb=model4.score(X_test,Y_test)
       0.8518518518518519
[26] sclr5=confusion_matrix(Y_test,model4.predict(X_test))
       sclr5
       array([[25, 4],
[ 4, 21]])
6. Linear Discriminant Analysis
[27] model5=LinearDiscriminantAnalysis()
      model5.fit(X_train,Y_train)
      sclda=model5.score(X_test,Y_test)
      sclda
     0.88888888888888888
[28] sclr6=confusion_matrix(Y_test,model5.predict(X_test))
     sclr6
      array([[26, 3],
             [ 3, 22]])
7. AdaBoost Classifier
291 model6=AdaBoostClassifier()
     model6.fit(X_train,Y_train)
     scabc=model6.score(X_test,Y_test)
     0.8148148148148148
30] sclr7=confusion_matrix(Y_test,model6.predict(X_test))
    sclr7
     array([[23, 6],
            [ 4, 21]])
```

Using ML Deployed model in Cloud predicting whether the person having heart disease or not

Total Number of Test cases tested: 10

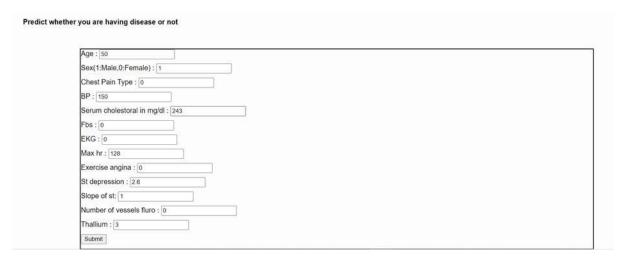
Total Number of Test cases passed:10

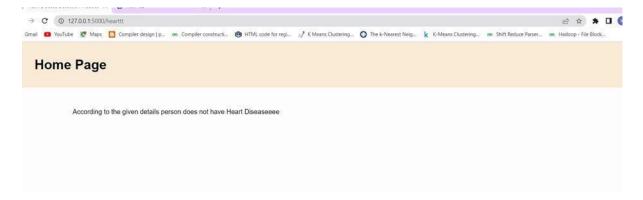
Total Number of Test cases failed:0

Using the form data we can check whether the individual has heart disease or not Individual having disease:



Individual does not have heart disease:





Dashboard Performance Testing:

