```
Linear Classification - Logistic Regression Model
      # Import the necessary libraries
       import pandas as pd
       import numpy as np
       from sklearn.model_selection import train_test_split
      from sklearn.linear_model import LogisticRegression
       from sklearn.metrics import accuracy_score
      from sklearn.metrics import accuracy_score, classification_report
       from sklearn.tree import DecisionTreeClassifier
       from sklearn.ensemble import RandomForestClassifier
In [... # Load the dataset into a Pandas dataframe
       # data = pd.read_csv('.\\Dataset #2 - Classification\\Classification.csv')
      data=pd.read_csv('fetal_health.csv')
      data['fetal_health'] = data['fetal_health'].replace(1.0,0)
      data['fetal_health'] = data['fetal_health'].replace(2.0,0)
data['fetal_health'] = data['fetal_health'].replace(3.0,1)
      # Split the dataset into features (X) and target (y)
      X = data.drop('fetal_health', axis = 1)
      Y = data['fetal_health']
       # Split the data into training and testing sets
      xtr, xts, ytr, yts = train_test_split(X, Y, test_size=0.2, random_state=0)
In [... # Create a logistic regression model
      model = LogisticRegression(penalty='12', max_iter=3000, verbose=0)
       # Train the model on the training data
       model.fit(xtr, ytr)
     LogisticRegression(max_iter=3000)
       # Make predictions on the test data
      predictions = model.predict(xts)
       # Calculate the accuracy of the model
       accuracy = accuracy score(yts, predictions)
       # Print the accuracy of the model
      print("Accuracy:", accuracy)
      Accuracy: 0.960093896713615
In [... # taking predictions on 10 test points
      predictions_20 = model.predict(xts[:20])
      print(' Actual labels: {} \nPredicted labels: {}'.format(np.array(yts[:20]), predictions_20))
       Actual labels: [0. 0. 0. 0. 0. 0. 0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0.]
      Predicted labels: [0. 0. 0. 0. 0. 0. 0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
      predictions = model.predict(xts)
      print("\nClassification Report: \n\n", classification_report(yts, predictions, zero_division=1))
      Classification Report:
                                  recall f1-score
                     precision
                                                     support
               0.0
                         0.97
                                   0.98
                                             0.98
                                                        384
               1.0
                         0.84
                                   0.74
                                             0.78
                                                         42
                                                        426
          accuracy
                                             0.96
                         0.90
                                   0.86
                                             0.88
                                                        426
         macro avg
      weighted avg
                         0.96
                                   0.96
                                             0.96
                                                        426
      import matplotlib.pyplot as plt
      from sklearn import metrics
       metrics.plot_roc_curve(model, xts, yts)
      plt.show()
      B.0 B
                       LogisticRegression (AUC = 0.98)
      Classification Trees - Basic Model
```

# Create a decision tree classifier

Accuracy: 98.12 %

model\_tree = DecisionTreeClassifier(splitter='best', max\_depth=15) # Split the data into training and testing sets xtr, xts, ytr, yts = train\_test\_split(X, Y, test\_size=0.2, random\_state=0) # Train the model on the training data model\_tree.fit(xtr, ytr) Out[... DecisionTreeClassifier(max\_depth=15)

# Make predictions on the test data predictions = model\_tree.predict(xts) # Calculate the accuracy of the model accuracy = accuracy\_score(yts, predictions) # Print the accuracy of the model print("Accuracy: {:.2f} %".format(accuracy\*100))

Actual labels: [0. 0. 0. 0. 0. 0. 0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0.] Predicted labels: [0. 0. 0. 0. 0. 0. 0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0.] print("\nClassification Report: \n\n", classification\_report(yts, predictions, zero\_division=1)) Classification Report:

support

print(' Actual labels: {} \nPredicted labels: {}'.format(np.array(yts[:20]), predictions\_20))

0.0 0.98 0.99 0.99 384 1.0 0.95 0.86 0.90 42 426 0.98 accuracy macro avg 0.97 0.93 0.94 426 weighted avg 0.98

recall f1-score

metrics.plot\_roc\_curve(model\_tree, xts, yts) plt.show()

predictions\_20 = model\_tree.predict(xts[:20])

precision

DecisionTreeClassifier (AUC = 0.93) 0.2 0.4 0.6 0.8 False Positive Rate (Positive label: 1.0) Classification Trees - "Random" Model

## # Create Random Forest classifier with 100 trees

rfc = RandomForestClassifier(n\_estimators=100, random\_state=0) # Fit the model to the training data

rfc.fit(xtr, ytr) RandomForestClassifier(random\_state=0)

In [... # Make predictions on test data y\_pred = rfc.predict(xts)

# Evaluate the model performance accuracy = accuracy\_score(yts, y\_pred) print("Accuracy:", accuracy)

Accuracy: 0.9906103286384976

print("\nClassification Report: \n\n", classification\_report(yts, y\_pred)) Classification Report: recall f1-score

Predicted labels: [0. 0. 0. 0. 0. 0. 0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]

support

precision 0.0

0.99 1.00 0.99 384 1.0 0.97 0.93 0.95 accuracy 0.99 0.98 0.97 macro avg weighted avg 0.99 0.99 0.99

426 y\_pred = rfc.predict(xts[:20]) print(' Actual labels: {} \nPredicted labels: {}'.format(np.array(yts[:20]), y\_pred)) Actual labels: [0. 0. 0. 0. 0. 0. 0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0.]

plt.show()

False Positive Rate (Positive label: 1.0)

metrics.plot\_roc\_curve(rfc, xts, yts)