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In [2]: # Import necessary libraries
        import pandas as pd
        from sklearn.model selection import train test split
        from sklearn.preprocessing import StandardScaler
        from sklearn.ensemble import RandomForestClassifier
        from sklearn.metrics import accuracy score, classification report
In [4]: # Load your dataset (replace 'your data.csv' with your dataset file)
        data = pd.read csv('heart failure clinical records dataset.csv')
In [5]: # Define features and target variable
        X = data.drop('DEATH EVENT', axis=1) # Features (input variables)
        y = data['DEATH EVENT'] # Target variable (0 or 1)
In [6]: # Split the dataset into training and test sets
        X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=42)
In [7]: # Preprocess data (scaling)
        scaler = StandardScaler()
        X train = scaler.fit transform(X train)
        X test = scaler.transform(X test)
In [8]: # Initialize and train the model (Random Forest Classifier)
        model = RandomForestClassifier(random state=42)
        model.fit(X_train, y_train)
Out[8]:
                  RandomForestClassifier
        RandomForestClassifier(random_state=42)
In [9]: # Make predictions on the test set
        y pred = model.predict(X test)
         # Fyaluate the model.
        accuracy = accuracy score(y test, y pred)
        print(f"Accuracy: {accuracy:.2f}")
        print(classification report(y test, y pred))
```

```
Accuracy: 0.75
                       precision
                                    recall f1-score support
                    0
                            0.72
                                      0.94
                                                0.81
                                                            35
                    1
                            0.86
                                      0.48
                                                0.62
                                                            25
                                                0.75
             accuracv
                                                            60
            macro avg
                            0.79
                                      0.71
                                                0.72
                                                            60
         weighted avg
                            0.78
                                      0.75
                                                0.73
                                                            60
In [15]: # Load your training dataset (replace 'training_data.csv' with your dataset file)
         training data = pd.read csv('heart failure clinical records dataset.csv')
         # Define features and target variable for training data
         X train = training data.drop('DEATH EVENT', axis=1)
         y train = training data['DEATH EVENT']
In [16]: # Preprocess training data (scaling)
         scaler = StandardScaler()
         X_train = scaler.fit_transform(X_train)
         # Initialize and train the model (Random Forest Classifier)
          model = RandomForestClassifier(random state=42)
         model.fit(X train, y train)
Out[16]:
                   RandomForestClassifier
         RandomForestClassifier(random state=42)
In [22]: # Load the new patient data from an Excel file (replace 'new_patient_data.xlsx' with the file path)
         new patient data = pd.read excel('newdata.xlsx')
In [23]: # Preprocess the new patient data (scaling, same as the training data)
         new_patient_data = scaler.transform(new_patient_data)
In [24]: # Make a prediction for the new patient
          prediction = model.predict(new patient data)
         # Interpret the prediction: 1 for heart failure predicted, 0 for no heart failure predicted
          if prediction[0] == 1:
             print("Heart failure predicted for the new patient.")
```

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else:
   print("No heart failure predicted for the new patient.")
```

Heart failure predicted for the new patient.

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In [ ]:
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