code

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[12]: import pandas as pd
      import numpy as np
      from sklearn.model_selection import train_test_split
      from sklearn.preprocessing import StandardScaler
      from sklearn.metrics import accuracy_score, precision_score, recall_score
[13]: # Load the data from the Excel file
      data = pd.read_excel("~/dataset/logistic-regression/Pumpkin_Seeds_Dataset.xlsx")
      # Preprocess the data
      X = data.drop(columns=['Class'])
      y = (data['Class'] == 'Çerçevelik').astype(int) # Convert 'Çerçevelik' to 1, __
       ⇔else 0
      # Split the data into training, validation, and testing sets
      X_train, X_temp, y_train, y_temp = train_test_split(X, y, test_size=0.5,_
       →random_state=42)
      X_val, X_test, y_val, y_test = train_test_split(X_temp, y_temp, test_size=0.6,_
       ⇒random state=42)
      # Normalize/standardize the data
      scaler = StandardScaler()
      X_train_scaled = scaler.fit_transform(X_train)
      X_val_scaled = scaler.transform(X_val)
      X_test_scaled = scaler.transform(X_test)
[14]: # Implement Logistic Regression
      def sigmoid(z):
          return 1 / (1 + np.exp(-z))
      def predict(X, weights):
          z = np.dot(X, weights)
          return sigmoid(z)
      def logistic_regression(X, y, learning_rate, num_iterations):
         num_samples, num_features = X.shape
          weights = np.zeros(num_features)
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for _ in range(num_iterations):
              y_pred = predict(X, weights)
              gradient = np.dot(X.T, (y_pred - y)) / num_samples
              weights -= learning_rate * gradient
          return weights
[15]: # Calculate accuracy, precision, and recall
      def calculate_metrics(y_true, y_pred):
          y_pred_binary = (y_pred >= 0.5).astype(int)
          accuracy = accuracy_score(y_true, y_pred_binary)
          precision = precision_score(y_true, y_pred_binary, zero_division=1)
          recall = recall_score(y_true, y_pred_binary, zero_division=1)
          return accuracy, precision, recall
[16]: def logistic_regression_calc(learning_rate,num_iterations):
          weights = logistic_regression(X_train_scaled, y_train, learning_rate,_
       →num_iterations)
          # Make predictions on test set
          test_predictions = sigmoid(np.dot(X_test_scaled, weights))
          test_accuracy, test_precision, test_recall = calculate metrics(y_test,__
       →test_predictions)
          print(f"Test Data evaluation(Learning rate = {learning_rate} and Number of ⊔
       →iterations = {num_iterations}):")
          print("Accuracy:", test_accuracy)
          print("Precision:", test_precision)
          print("Recall:", test_recall)
[17]: logistic_regression_calc(0.01,100)
     Test Data evaluation(Learning rate = 0.01 and Number of iterations = 100):
     Accuracy: 0.8573333333333333
     Precision: 0.8601583113456465
     Recall: 0.8578947368421053
[18]: logistic_regression_calc(0.001,100)
     Test Data evaluation(Learning rate = 0.001 and Number of iterations = 100):
     Accuracy: 0.8453333333333333
     Precision: 0.8473684210526315
     Recall: 0.8473684210526315
[19]: logistic_regression_calc(0.0001,100)
     Test Data evaluation(Learning rate = 0.0001 and Number of iterations = 100):
     Accuracy: 0.844
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Precision: 0.8469656992084432 Recall: 0.8447368421052631

[20]: logistic_regression_calc(0.01,200)

Test Data evaluation(Learning rate = 0.01 and Number of iterations = 200):

Accuracy: 0.858666666666667 Precision: 0.8605263157894737 Recall: 0.8605263157894737

[21]: logistic_regression_calc(0.001,200)

Test Data evaluation(Learning rate = 0.001 and Number of iterations = 200):

Accuracy: 0.846666666666667 Precision: 0.8496042216358839 Recall: 0.8473684210526315

[22]: logistic_regression_calc(0.0001,200)

Test Data evaluation(Learning rate = 0.0001 and Number of iterations = 200):

Accuracy: 0.844

Precision: 0.8469656992084432 Recall: 0.8447368421052631