Lin Yu STAT37710 HW2

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[1]: import numpy as np
     import matplotlib.pyplot as plt
     from sklearn.model_selection import KFold
[2]: def load_data(digits_file, labels_file=None):
         # Load the files
         digits = np.loadtxt(digits_file).astype(np.float32)
         if labels_file:
             labels = np.loadtxt(labels_file).astype(int)
             return digits, labels
         return digits
     X_train, Y_train = load_data('train35.digits', 'train35.labels')
     X_test = load_data('test35.digits')
[3]: def Batch_perceptron_train(X, y, epochs=1):
         Train a perceptron model using batch updates.
         Args:
         X (np.array): Input feature matrix where each row represents a data point.
         y (np.array): Output label vector where each element is the label for \Box
      ⇔corresponding data point.
         epochs (int): Number of times the training data is processed.
         Returns:
         np.array: Final weights after training.
         float: Final bias value after training.
         list: List of cumulative mistakes over all epochs.
         list: List of Number of mistakes made in each epoch.
         n_samples, n_features = X.shape # Access the number of data points and the
      →number of dimensions
         weights = np.zeros(n_features) # Initialize weights as zeros
         bias = 0 # Initialize bias as zero
         cumulative mistakes = [] # List to store the cumulative number of mistakes
         total_mistakes = 0 # Total number of mistakes made across all epochs
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mistakes_per_epoch = [] # List to store the number of mistakes per epoch
    # Iterate over the number of epochs
   for epoch in range(epochs):
       mistakes = 0 #Reset mistakes for this epoch
       for i in range(n_samples):
            prediction = np.dot(X[i], weights) + bias
            if np.sign(prediction) != np.sign(y[i]):
                weights += y[i] * X[i]
                bias += y[i]
                mistakes += 1
                total mistakes += 1
            # Record the total mistakes made after k examples
            cumulative_mistakes.append(total_mistakes)
        # Record the number of mistakes made this epoch
       mistakes_per_epoch.append(mistakes)
   return weights, bias, cumulative_mistakes,mistakes_per_epoch
def cross_validate(X, y, k_fold_num=5, max_epochs=10):
   Perform k-fold cross-validation to determine the optimal number of training
 ⇔epochs.
   Arqs:
   X (np.array): Input feature matrix.
   y (np.array): Output label vector.
   k_fold_num (int): Number of folds for cross-validation.
   max_epochs (int): Maximum number of epochs to test.
   Returns:
    int: The best epoch number found during cross-validation.
    np.array: Array of average accuracies for each epoch.
    int: Number of training samples in each fold.
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   kf = KFold(n_splits=k_fold_num)
   epoch_performance = np.zeros(max_epochs) # Array to store the performance_
 ⇔per epoch
    # Iterate over each split
   for train_index, test_index in kf.split(X):
       X_train, X_test = X[train_index], X[test_index]
       y_train, y_test = y[train_index], y[test_index]
        # Test each epoch count from 1 to maximal epochs
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for epoch in range(1, max_epochs + 1):
                 weights, bias, _ ,_ = Batch_perceptron_train(X_train, y_train,_
      ⇔epochs=epoch)
                 predictions = np.sign(X_test.dot(weights) + bias)
                 accuracy = np.mean(predictions == y_test)
                 epoch performance[epoch-1] += accuracy
         # Calculate average performance over all k-folds
         epoch_performance /= k_fold_num
         best_epoch = np.argmax(epoch_performance) + 1
         n_train=len(X_train)
         return best_epoch, epoch_performance,n_train
     def plot_cumulative_mistakes(mistakes):
         Plot the cumulative number of mistakes over training.
         mistakes (list): Cumulative mistakes counted during training.
         plt.figure(figsize=(10, 6))
         plt.plot(mistakes, marker='o')
         plt.title("Cumulative Number of Mistakes Over Training")
         plt.xlabel("Number of Examples Seen")
         plt.ylabel("Cumulative Mistakes")
         plt.grid(True)
         plt.show()
[4]: def plot_mistakes_per_epoch(mistakes_per_epoch, n_samples, n_epoch):
         Plots the frequency of mistakes made per epoch during training.
         # Calculate the frequency of mistakes (mistakes per data point)
         mistake_freq = np.array(mistakes_per_epoch) / n_samples
         # Create the epochs array (starting from 1 to n_epoch)
         epochs = np.arange(1, n_epoch + 1)
         # Plotting
         plt.figure(figsize=(10, 6))
         plt.plot(epochs, mistake_freq, marker='o', linestyle='-', color='b')
         plt.title("Frequency of Mistakes Per Epoch at the Best Epoch")
         plt.xlabel("Epoch(Number of Feeding Test)")
         plt.ylabel("Frequency of Mistakes")
         plt.grid(True)
         plt.show()
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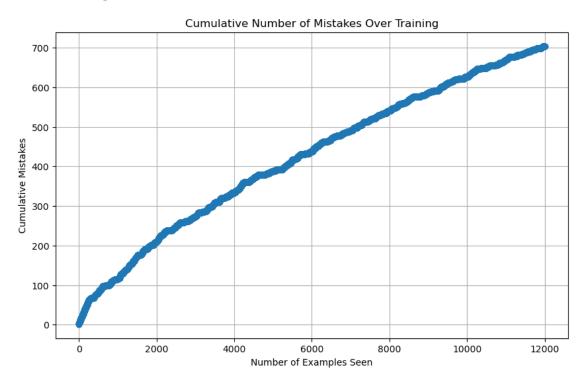
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[5]: # Main Execution
if __name__ == "__main__":
    # Determine the best number of epochs via cross-validation
    best_epoch, performance,n_train = cross_validate(X_train, Y_train)
    print("Best number of epochs:", best_epoch)

# Train with the best number of epochs
    weights, bias, cul_mistakes, mistakes_per_epoch =__
Batch_perceptron_train(X_train, Y_train, epochs=best_epoch)

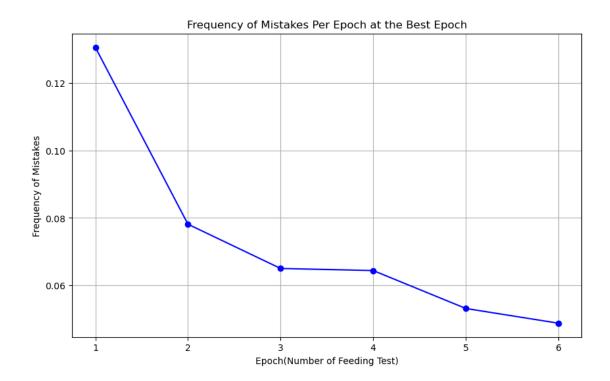
# Plot cumulative mistakes
    plot_cumulative_mistakes(cul_mistakes)

# Generate predictions for the test set
    predictions = np.sign(X_test.dot(weights) + bias)
    np.savetxt('test35.predictions', predictions, fmt='%d')
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Best number of epochs: 6



[6]: plot_mistakes_per_epoch(mistakes_per_epoch, n_train, best_epoch)



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[11]: count=0
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print("The number of 3 in the test set:" ,len(predictions)-count)

The number of 3 in the test set: 102 The number of 5 in the test set: 98

print("The number of 5 in the test set:", count)

for i in range(len(predictions)):
 if predictions[i]<0:
 count+=1</pre>

[7]: print (predictions)