## Problem2

## November 1, 2024

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[1]: #Library Imports
     from torchvision import datasets
     from torchvision.transforms import ToTensor
     from torch.utils.data import DataLoader
     import torch
     from torch import nn
     import numpy as np
     import matplotlib.pyplot as plt
[2]: | train_data = datasets.USPS(root='usps', download=True, transform=ToTensor(),
     otrain=True)
     test_data = datasets.USPS(root='usps', download=True, transform=ToTensor(),__
     ⇔train=False)
     # Create DataLoaders for training and testing
     train_loader = DataLoader(train_data, batch_size=1024, shuffle=True)
     test_loader = DataLoader(test_data, batch_size=len(test_data), shuffle=False)
[3]: # Define the MLP model with 1 hidden layer of width 65, following the specified
     \hookrightarrowstructure
     class MLP(nn.Module):
         def __init__(self):
             super().__init__()
             self.flatten = nn.Flatten()
             self.mlp = nn.Sequential(
                 nn.Linear(16 * 16, 64), # Input layer to hidden layer
                 nn.ReLU(),
                 nn.Linear(64, 64), # Hidden layer
                 nn.ReLU(),
                 nn.Linear(64, 10) # Output layer
             )
         def forward(self, X):
             return self.mlp(self.flatten(X))
     model = MLP().to('cuda:1')
```

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loss_fn = nn.CrossEntropyLoss()
optimizer = torch.optim.SGD(model.parameters(), lr=0.5)
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[4]: %%time
    epochs = 1000
    training_accuracy = []
    test_accuracy = []
    epoch_loss = []
    for epoch in range(epochs):
         cumulative accuracy = 0
         cumulative_loss = 0
        for X, Y in train_loader:
            X, Y = X.to('cuda:1'), Y.to('cuda:1')
            out = model(X)
            loss = loss_fn(out, Y)
            optimizer.zero_grad()
            loss.backward()
            optimizer.step()
            cumulative_loss += loss.item()
             cumulative_accuracy += (out.argmax(axis=1) == Y).sum().item()
         epoch_loss.append(cumulative_loss / len(train_loader))
        training_accuracy.append(cumulative_accuracy / len(train_data))
        with torch.no_grad():
             for Xt, Yt in test_loader:
                Xt, Yt = Xt.to('cuda:1'), Yt.to('cuda:1')
                test_out = model(Xt)
                test_accuracy_epoch = (test_out.argmax(axis=1) == Yt).sum().item() /
      → len(test_data)
            test_accuracy.append(test_accuracy_epoch)
        if (epoch + 1) \% 100 == 0:
            print(f"Epoch {epoch + 1}/{epochs} | Loss: {epoch_loss[-1]:.4f} |
      →Training Accuracy: {training_accuracy[-1]:.4f} | Test Accuracy:
      Epoch 100/1000 | Loss: 0.0351 | Training Accuracy: 0.9900 | Test Accuracy:
    0.9352
    Epoch 200/1000 | Loss: 0.8745 | Training Accuracy: 0.7106 | Test Accuracy:
    0.7937
    Epoch 300/1000 | Loss: 0.0164 | Training Accuracy: 0.9971 | Test Accuracy:
    0.9387
    Epoch 400/1000 | Loss: 0.0073 | Training Accuracy: 0.9993 | Test Accuracy:
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Epoch 500/1000 | Loss: 1.8209 | Training Accuracy: 0.3395 | Test Accuracy:

0.4190

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Epoch 600/1000 | Loss: 0.2941 | Training Accuracy: 0.9093 | Test Accuracy:
    0.8321
    Epoch 700/1000 | Loss: 0.5172 | Training Accuracy: 0.8287 | Test Accuracy:
    0.7833
    Epoch 800/1000 | Loss: 0.1645 | Training Accuracy: 0.9468 | Test Accuracy:
    0.8999
    Epoch 900/1000 | Loss: 1.5180 | Training Accuracy: 0.3973 | Test Accuracy:
    0.3234
    Epoch 1000/1000 | Loss: 0.5628 | Training Accuracy: 0.8098 | Test Accuracy:
    0.7434
    CPU times: user 10min 37s, sys: 844 ms, total: 10min 38s
    Wall time: 10min 38s
[5]: plt.figure(figsize=(10, 5))
     plt.plot(np.arange(1, epochs + 1), epoch loss, label="Cross Entropy Loss")
     plt.xlabel("Epoch")
     plt.ylabel("Loss")
     plt.title("Cross Entropy Loss over Epochs")
     plt.legend()
     plt.show()
     plt.figure(figsize=(10, 5))
     plt.plot(np.arange(1, epochs + 1), training_accuracy, label="Training Accuracy")
     plt.plot(np.arange(1, epochs + 1), test accuracy, label="Test Accuracy")
     plt.xlabel("Epoch")
     plt.ylabel("Accuracy")
     plt.title("Training and Test Accuracy over Epochs")
     plt.legend()
     plt.show()
```



