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In [ ]: import numpy as np
import torch
import torch.nn as nn
import torch.nn.functional as F
import torch.optim as optim
import torch.utils.data
import time
import os

# Load the dataset
data = np.load('lab2_dataset.npz')
train_feats = torch.tensor(data['train_feats'])
test_feats = torch.tensor(data['test_feats'])
train_labels = torch.tensor(data['train_labels'])
test_labels = torch.tensor(data['test_labels'])
phone_labels = data['phone_labels']

print(train_feats.shape)
print(test_feats.shape)
print(train_labels.shape)
print(test_labels.shape)
print(phone_labels.shape)

device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
print(device)
```

```
torch.Size([44730, 11, 40])
torch.Size([4773, 11, 40])
torch.Size([44730])
torch.Size([4773])
(48,)
cpu
```

```
In [ ]: # Set up the dataloaders
train_dataset = torch.utils.data.TensorDataset(train_feats, train_labels)
train_loader = torch.utils.data.DataLoader(train_dataset, batch_size=8, shuffle=True)

test_dataset = torch.utils.data.TensorDataset(test_feats, test_labels)
test_loader = torch.utils.data.DataLoader(test_dataset, batch_size=8, shuffle=False)
```

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In [ ]: print(phone_labels)

['sil' 's' 'ao' 'l' 'r' 'iy' 'vcl' 'd' 'eh' 'cl' 'p' 'ix' 'z' 'ih' 'sh'
'n' 'v' 'aa' 'y' 'uw' 'w' 'ey' 'dx' 'b' 'ay' 'ng' 'k' 'epi' 'ch' 'dh'
'er' 'en' 'g' 'aw' 'hh' 'ae' 'ow' 't' 'ax' 'm' 'zh' 'ah' 'el' 'f' 'jh'
'uh' 'oy' 'th']
```

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In [ ]: # Define the model architecture
class MyModel(nn.Module):
    def __init__(self):
        super(MyModel, self).__init__()

        self.fc1 = nn.Linear(11 * 40, 1024)
        self.fc2 = nn.Linear(1024, 1024)
        self.fc3 = nn.Linear(1024, 1024)
        self.fc4 = nn.Linear(1024, 512)
        self.fc5 = nn.Linear(512, 256)
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self.fc6 = nn.Linear(256, 128)
self.fc7 = nn.Linear(128, 64)
self.fc8 = nn.Linear(64, 48)

self.dropout = nn.Dropout(0.25)

self.relu = nn.ReLU() # activation function

def forward(self, x):
    x = torch.reshape(x, (-1, 11 * 40))
    #print(x.shape)

    x = self.fc1(x)
    x = self.relu(x)

    x = self.fc2(x)
    x = self.relu(x)

    x = self.fc3(x)
    x = self.relu(x)

    x = self.fc4(x)
    x = self.relu(x)

    x = self.dropout(x)

    x = self.fc5(x)
    x = self.relu(x)

    x = self.fc6(x)
    x = self.relu(x)

    x = self.fc7(x)
    x = self.relu(x)

    x = self.fc8(x)

    return x

```

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In [ ]: # Instantiate the model, loss function, and optimizer
model = MyModel()
modeltrained = False
try :
    model.load_state_dict(torch.load('model.pt'))
    model.eval()
    modeltrained = True
    print('Model loaded')
except:
    print('No model found')
    pass
criterion = nn.CrossEntropyLoss()
optimizer = optim.SGD(model.parameters(), lr=0.001, momentum=0.9)

def train_network(model, train_loader, criterion, optimizer):
    # TODO: fill in
    for epoch in range(10):
        # running_loss = 0.0
        time1 = time.time()
        for i, (inputs, labels) in enumerate(train_loader, 0):

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optimizer.zero_grad()
outputs = model(inputs)
loss = criterion(outputs, labels)
loss.backward()
optimizer.step()
# running_loss += loss.item()
# if i % 1000 == 999:
#     print('[%d, %5d] loss: %.3f' % (epoch + 1, i + 1, running_loss))
#     running_loss = 0.0
#
time2 = time.time()
print('Epoch %d' % (epoch + 1))
print(time2 - time1)
test_network(model, test_loader)

label_acc = {}

missclassifications = {}

def test_network(model, test_loader):
    correct = 0
    total = 0
    with torch.no_grad():
        for data in test_loader:
            inputs, labels = data
            outputs = model(inputs)
            # outputs
            _, predicted = torch.max(outputs.data, 1)
            total += labels.size(0)
            correct += (predicted == labels).sum().item()
            for i in range(len(labels)):
                if labels[i].item() not in label_acc:
                    label_acc[labels[i].item()] = [0, 0]
                label_acc[labels[i].item()][1] += 1
                if predicted[i] == labels[i]:
                    label_acc[labels[i].item()][0] += 1
                else:
                    if labels[i].item() not in missclassifications:
                        missclassifications[labels[i].item()] = {}
                    if predicted[i].item() not in missclassifications[labels[i].item()]:
                        missclassifications[labels[i].item()][predicted[i].item()] = 0
                    missclassifications[labels[i].item()][predicted[i].item()] += 1

    print('Test accuracy: %d %%' % (100 * correct / total))

if not modeltrained:
    print('Training model')
    time1 = time.time()
    train_network(model, train_loader, criterion, optimizer)
    time2 = time.time()
    print(time2 - time1)
    print('Finished Training')
    torch.save(model.state_dict(), 'model.pt')

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Model loaded

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In [ ]: print('Finished Training')
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```
Finished Training
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In [ ]: test_network(model, test_loader)
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Test accuracy: 58 %
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In [ ]: # sort by accuracy
sorted_acc = sorted(label_acc.items(), key=lambda x: x[1][0] / x[1][1], reverse=True)

print("Top 5 most accurate phones:")
for i in range(5):
    print(phone_labels[sorted_acc[i][0]], sorted_acc[i][1][0] / sorted_acc[i][1][1])

print("Top 5 least accurate phones:")

for i in range(1, 6):
    print(phone_labels[sorted_acc[-i][0]], sorted_acc[-i][1][0] / sorted_acc[-i][1][1])
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```
Top 5 most accurate phones:
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```
s 0.91
```

```
sil 0.85
```

```
b 0.83
```

```
y 0.82
```

```
r 0.81
```

```
Top 5 least accurate phones:
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```
vcl 0.23
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```
ah 0.25
```

```
ih 0.27
```

```
zh 0.2876712328767123
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```
p 0.29
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In [ ]: common_missclassifications = []
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common_missclassifications.append(np.where(phone_labels == 'sh')[0][0])
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```
common_missclassifications.append(np.where(phone_labels == 'p')[0][0])
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```
common_missclassifications.append(np.where(phone_labels == 'm')[0][0])
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```
common_missclassifications.append(np.where(phone_labels == 'r')[0][0])
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```
common_missclassifications.append(np.where(phone_labels == 'ae')[0][0])
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for i in common_missclassifications:
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    sorted_miss = sorted(missclassifications[i].items(), key=lambda x: x[1], reverse=True)
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    print(phone_labels[i], "is commonly misclassified as:", phone_labels[sorted_miss[0][0]])
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sh is commonly misclassified as: s with 13 instances
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p is commonly misclassified as: b with 26 instances
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m is commonly misclassified as: ng with 8 instances
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```
r is commonly misclassified as: er with 3 instances
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ae is commonly misclassified as: eh with 20 instances
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