4/25/2021 AML_final

```
import torch
In [1]:
         import torchvision
         import numpy as np
         import matplotlib.pyplot as plt
         import torch.nn as nn
         import torch.nn.functional as F
         from torchvision.datasets import CIFAR100
         import torchvision.transforms as transforms
         from torchvision.utils import make grid
         from torch.utils.data.dataloader import DataLoader
         from torch.utils.data import random_split,ConcatDataset
         import model
         import data
In [2]:
         device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
         net = model.resnet34().to(device)
In [3]:
In [ ]:
         trainDataLoader, testDataLoader = data.loadData(128)
         loss = nn.CrossEntropyLoss()
In [5]:
         optimizer = torch.optim.Adam(net.parameters(), lr=0.0001)
         def evaluate(model,dataloader):
In [6]:
           acc = 0.0
           rights = 0
           wrongs = 0
           for i, test_examples in enumerate(dataloader, 0):
             #predicting using the nets
             inputs, labels = test examples
             predicted outputs = model(inputs.float().cuda())
             #Selecting the label which has the largest outputs
             outputs = torch.argmax(predicted outputs, 1)
             #Counting successfully and unsuccessfully predicted cases
             for j, n in enumerate(outputs):
               if n == labels[j]:
                 rights += 1
               else:
                 wrongs += 1
           #calculate accuracy with the cases we recorded
           acc = rights/(rights+wrongs)
           #return the accuracy
           return acc
         def train(model,train,test,loss_fn,optimizer,watch_iter):
In [7]:
             total iter = 0
             loss = 0.0
             while total iter < 10000:
                 for batch in train:
                     total iter += 1
                     train inputs, train labels = batch
                     train inputs, train labels = train inputs.to(device), train labels.to(devic
                     train outputs = model(train inputs)
                     1 = loss_fn(train_outputs, train_labels)
```

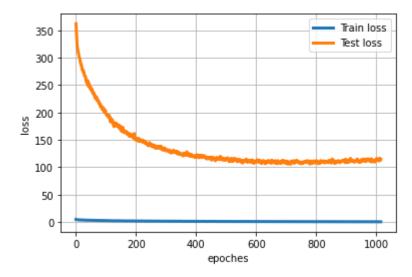
4/25/2021 AML_final

```
loss += l.item()
        optimizer.zero grad()
        1.backward()
        optimizer.step()
        if total iter % watch iter == 0:
            train loss = loss / watch iter
            train loss his.append(train loss)
            loss = 0.0
            for batch in test:
                test inputs, test labels = batch
                test inputs, test labels = test inputs.to(device), test labels.to(d
                test outputs = model(test inputs)
                1 = loss_fn(test_outputs, test_labels)
                loss += l.item()
            test_loss_his.append(loss)
            txt = f'iter: {total iter: 6d}, train loss: {train loss}, test loss: {1
            print(txt)
            print('accuracy: ' + str(evaluate(model,test)*100) + '%')
            loss = 0.0
return
```

```
In [ ]: train_loss_his = []
    test_loss_his = []
    train(net,trainDataLoader,testDataLoader,loss,optimizer,10)
```

```
In [11]: plt.plot(range(len(train_loss_his)),train_loss_his,'-',linewidth=3,label='Train_loss')
    plt.plot(range(len(train_loss_his)),test_loss_his,'-',linewidth=3,label='Test_loss')
    plt.xlabel('epoches')
    plt.ylabel('loss')
    plt.grid(True)
    plt.legend()
```

Out[11]: <matplotlib.legend.Legend at 0x1460fc111ac0>



```
In [13]: print('accuracy: ' + str(evaluate(net,testDataLoader)*100) + '%')
accuracy: 63.94%
```