```
import numpy as np
import matplotlib.pyplot as plt
from tqdm import tqdm # Displays a progress bar
import pandas as pd
import torch
from torch import nn
from torch import optim
import torch.nn.functional as F
from torchvision import datasets, transforms
from torch.utils.data import Dataset, Subset, DataLoader, random_split
import math
```

```
In [3]: # Load the dataset and train, val, test splits
        print("Loading datasets...")
        dataset_path = "C:/Users/Admin/Desktop/cse803_hw5"
        MNIST_transform = transforms.Compose([
            transforms.ToTensor(),
            transforms.Normalize([0.1307], [0.3081])
        ])
        MNIST_train = datasets.MNIST(
            dataset_path,
            download=True,
            train=True,
            transform=MNIST_transform
        MNIST_test = datasets.MNIST(
            dataset_path,
            download=True,
            train = False,
            transform=MNIST_transform
        )
        FASHION_transform = transforms.Compose([
            transforms.ToTensor(),
            transforms.Normalize([0.2859], [0.3530])
        ])
        FASHION_train = datasets.FashionMNIST(
            dataset_path,
            download=True,
            train=True,
            transform=MNIST_transform
        FASHION_test = datasets.FashionMNIST(
            dataset_path,
            download=True,
            train=False,
            transform=FASHION_transform
        )
```

1 of 7

```
Loading datasets...
Downloading http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz
Failed to download (trying next):
<urlopen error [SSL: CERTIFICATE_VERIFY_FAILED] certificate verify failed: certifica</pre>
te has expired (_ssl.c:1000)>
Downloading https://ossci-datasets.s3.amazonaws.com/mnist/train-images-idx3-ubyte.gz
Downloading https://ossci-datasets.s3.amazonaws.com/mnist/train-images-idx3-ubyte.gz
to C:/Users/Admin/Desktop/cse803_hw5\MNIST\raw\train-images-idx3-ubyte.gz
                                            9912422/9912422 [00:02<00:00, 438600
4.32it/s]
Extracting C:/Users/Admin/Desktop/cse803_hw5\MNIST\raw\train-images-idx3-ubyte.gz to
C:/Users/Admin/Desktop/cse803_hw5\MNIST\raw
Downloading http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz
Failed to download (trying next):
<urlopen error [SSL: CERTIFICATE_VERIFY_FAILED] certificate verify failed: certifica</pre>
te has expired (_ssl.c:1000)>
Downloading https://ossci-datasets.s3.amazonaws.com/mnist/train-labels-idx1-ubyte.gz
Downloading https://ossci-datasets.s3.amazonaws.com/mnist/train-labels-idx1-ubyte.gz
to C:/Users/Admin/Desktop/cse803_hw5\MNIST\raw\train-labels-idx1-ubyte.gz
                                            28881/28881 [00:00<00:00, 92504
100%
5.96it/s]
Extracting C:/Users/Admin/Desktop/cse803_hw5\MNIST\raw\train-labels-idx1-ubyte.gz to
C:/Users/Admin/Desktop/cse803_hw5\MNIST\raw
Downloading http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz
Failed to download (trying next):
<urlopen error [SSL: CERTIFICATE_VERIFY_FAILED] certificate verify failed: certifica</pre>
te has expired ( ssl.c:1000)>
Downloading https://ossci-datasets.s3.amazonaws.com/mnist/t10k-images-idx3-ubyte.gz
Downloading https://ossci-datasets.s3.amazonaws.com/mnist/t10k-images-idx3-ubyte.gz
to C:/Users/Admin/Desktop/cse803_hw5\MNIST\raw\t10k-images-idx3-ubyte.gz
100%
                                    1648877/1648877 [00:00<00:00, 666476
9.54it/s]
Extracting C:/Users/Admin/Desktop/cse803_hw5\MNIST\raw\t10k-images-idx3-ubyte.gz to
C:/Users/Admin/Desktop/cse803_hw5\MNIST\raw
Downloading http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz
Failed to download (trying next):
<urlopen error [SSL: CERTIFICATE_VERIFY_FAILED] certificate verify failed: certifica</pre>
te has expired (_ssl.c:1000)>
Downloading https://ossci-datasets.s3.amazonaws.com/mnist/t10k-labels-idx1-ubyte.gz
Downloading https://ossci-datasets.s3.amazonaws.com/mnist/t10k-labels-idx1-ubyte.gz
to C:/Users/Admin/Desktop/cse803 hw5\MNIST\raw\t10k-labels-idx1-ubyte.gz
100%
                                                 4542/4542 [00:00
<?, ?it/s]
Extracting C:/Users/Admin/Desktop/cse803_hw5\MNIST\raw\t10k-labels-idx1-ubyte.gz to
C:/Users/Admin/Desktop/cse803_hw5\MNIST\raw
```

```
0.000
 In [6]:
         Data Loaders.
         class GridDataset(Dataset):
             def __init__(self, MNIST_dataset, FASHION_dataset): # pass in dataset
                 assert len(MNIST_dataset) == len(FASHION_dataset)
                 self.MNIST dataset, self.FASHION dataset = MNIST dataset, FASHION dataset
                 self.targets = FASHION_dataset.targets
                 torch.manual_seed(442) # Fix random seed for reproducibility
                 N = len(MNIST dataset)
                 self.randpos = torch.randint(low=0,high=4,size=(N,)) # position of the FASH
                 self.randidx = torch.randint(low=0,high=N,size=(N,3)) # indices of MNIST im
             def len (self):
                 return len(self.MNIST_dataset)
             def __getitem__(self,idx): # Get one Fashion-MNIST image and three MNIST images
                 idx1, idx2, idx3 = self.randidx[idx]
                 x = self.randpos[idx]%2
                 y = self.randpos[idx]//2
                 p1 = self.FASHION_dataset.__getitem__(idx)[0]
                 p2 = self.MNIST_dataset.__getitem__(idx1)[0]
                 p3 = self.MNIST_dataset.__getitem__(idx2)[0]
                 p4 = self.MNIST_dataset.__getitem__(idx3)[0]
                 combo = torch.cat((torch.cat((p1,p2),2),torch.cat((p3,p4),2)),1)
                 combo = torch.roll(combo, (x*28,y*28), dims=(0,1))
                 return (combo, self.targets[idx])
         trainset = GridDataset(MNIST train, FASHION train)
         testset = GridDataset(MNIST_test, FASHION_test)
         trainloader = DataLoader(trainset, batch_size=64, shuffle=True)
         testloader = DataLoader(testset, batch_size=64, shuffle=True)
         0.00
In [13]:
         Network class.
         class Network(nn.Module):
             def __init__(self):
                 super().__init__()
                 # TODO: Design your own base module, define layers here
                 k_size = 5
                 self.base = nn.Sequential(
                     nn.Conv2d(1, 32, kernel_size=k_size, stride=1, padding=2),
                     nn.ReLU(),
                     nn.MaxPool2d(kernel_size=2, stride=2),
                     nn.Conv2d(32, 64, kernel_size=k_size, stride=1, padding=2),
                     nn.ReLU(),
                     nn.MaxPool2d(kernel_size=2, stride=2),
                     nn.Conv2d(64, 128, kernel_size=k_size, stride=1, padding=2),
                     nn.ReLU(),
                     nn.MaxPool2d(kernel_size=2, stride=2),
                 out_channel = 128
                 self.avgpool = nn.AdaptiveAvgPool2d(1)
                 self.fc = nn.Linear(out_channel,10)
```

3 of 7

```
self.conv = nn.Conv2d(out_channel,10,1) # 1x1 conv layer (substitutes fc)
             def transfer(self): # Copy weights of fc layer into 1x1 conv layer
                  self.conv.weight = nn.Parameter(self.fc.weight.unsqueeze(2).unsqueeze(3))
                 self.conv.bias = nn.Parameter(self.fc.bias)
             def visualize(self,x):
                 x = self.base(x)
                 x = self.conv(x)
                 return x
             def forward(self,x):
                 x = self.base(x)
                 x = self.avgpool(x)
                 x = x.view(x.size(0), -1)
                 x = self.fc(x)
                 return x
In [17]:
         Hyperparameters.
         # configure device
         device = "cuda" if torch.cuda.is_available() else "cpu"
         # init model
         model = Network().to(device)
         # specify the loss layer
         criterion = nn.CrossEntropyLoss()
         # TODO: Modify the line below, experiment with different optimizers and parameters
         optimizer = optim.Adam(
             model.parameters(),
             lr=0.001,
             weight_decay=1e-4
         )
         # TODO: choose an appropriate number of training epochs
         num epoch = 6
         0.00
In [18]:
         Train & evaluation functions.
         def train(model, loader, num_epoch = 10): # Train the model
             print("Start training...")
             model.train() # Set the model to training mode
             for i in range(num_epoch):
                 running_loss = []
```

loss = criterion(pred, label) # Calculate the loss

optimizer.zero_grad() # Clear gradients from the previous iteration
pred = model(batch) # This will call Network.forward() that you impleme

for batch, label in tqdm(loader):
 batch = batch.to(device)
 label = label.to(device)

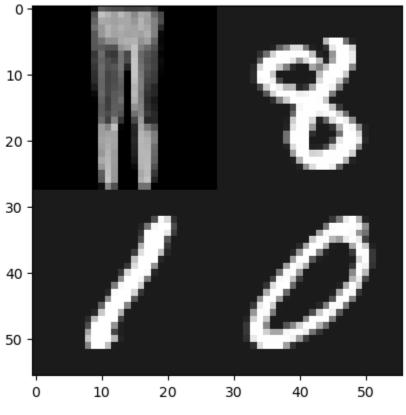
running_loss.append(loss.item())

```
loss.backward() # Backprop gradients to all tensors in the network
                     optimizer.step() # Update trainable weights
                 print("Epoch {} loss:{}".format(i+1,np.mean(running_loss))) # Print the ave
             print("Done!")
         def evaluate(model, loader): # Evaluate accuracy on validation / test set
             model.eval() # Set the model to evaluation mode
             correct = 0
             with torch.no grad(): # Do not calculate grident to speed up computation
                 for batch, label in tqdm(loader):
                     batch = batch.to(device)
                     label = label.to(device)
                     pred = model(batch)
                     correct += (torch.argmax(pred,dim=1)==label).sum().item()
             acc = correct/len(loader.dataset)
             print("Evaluation accuracy: {}".format(acc))
             return acc
In [19]:
         Train and evaluate model.
         0.00
         # train
         train(model, trainloader, num_epoch)
        Start training...
        100%
                                                                 | 938/938 [05:06<00:00,
        3.06it/s
        Epoch 1 loss:0.9900219045214053
        100%
                                                                 938/938 [05:13<00:00,
        3.00it/s]
        Epoch 2 loss:0.5521289967715359
        100%
                                                                938/938 [06:18<00:00,
        2.48it/s]
        Epoch 3 loss:0.43913425810174395
        100%
                                                            938/938 [08:40<00:00,
        1.80it/s]
        Epoch 4 loss:0.3843382164232258
        100%
                                                            938/938 [14:30<00:00,
        1.08it/s]
        Epoch 5 loss:0.3522641232717774
        Done!
        TypeError
                                                 Traceback (most recent call last)
        Cell In[19], line 6
              4 # train & evaluate
              5 train(model, trainloader, num_epoch)
        ---> 6 evaluate(model, testloader, num_epoch)
              8 # Copy the weights from fc layer to 1x1 conv layer
              9 model.transfer()
        TypeError: evaluate() takes 2 positional arguments but 3 were given
In [21]: train(model, trainloader, 1)
```

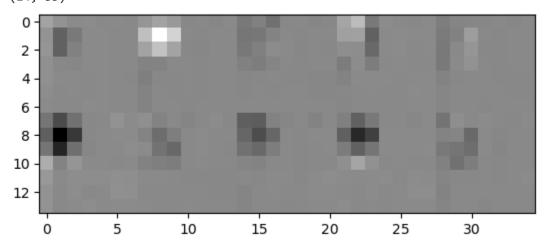
```
Start training...
         100%
                                                                   | 938/938 [14:42<00:00,
         1.06it/s]
         Epoch 1 loss:0.3304809541590432
          0.00
In [22]:
          Evaluate.
          evaluate(model, testloader)
          # Copy the weights from fc layer to 1x1 conv layer
          model.transfer()
         100%
                                                             157/157 [02:05<00:00,
         1.25it/s]
         Evaluation accuracy: 0.803
In [114...
          Visualize.
          # TODO: Choose a correctly classified image and visualize it
          predloader = DataLoader(testset, batch_size=64, shuffle=False)
          model.eval() # Set the model to evaluation mode
          ....
          correct_img = None
          while correct_img is None:
              batch, label = next(iter(predloader))
              batch = batch.to(device)
              label = label.to(device)
              pred = model(batch)
              correct_imgs = batch[torch.argmax(pred,dim=1) == label]
              if len(correct_imgs):
                  correct_img = correct_imgs[0]
          ....
          # select correctly predicted image
          batch, label = next(iter(predloader))
          batch = batch.to(device)
          label = label.to(device)
          correct_idx = 3
          correct_img = batch[correct_idx] # pre-calculated correct image
          print(correct_img.shape)
          plt.imshow(correct_img.squeeze(), cmap="gray")
          plt.show()
          # visualize activation map
          activated_img = model.visualize(correct_img).detach().numpy()
          print(activated_img.shape)
          grid_img = np.zeros((14, 35))
          for i in range(2):
              for j in range(5):
                  idx = (i * 5) + j
                  grid_img[i * 7: (i + 1) * 7, j * 7: (j + 1) * 7] = activated_img[idx, :, :]
```

```
print(grid_img.shape)
plt.imshow(grid_img, cmap="gray")
plt.show()
```

torch.Size([1, 56, 56])



(10, 7, 7) (14, 35)



In []:

7 of 7