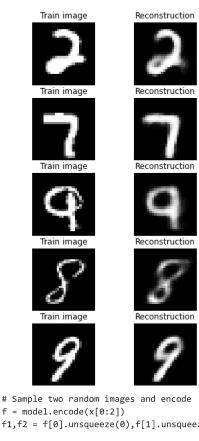
## 37 - Mrunmayee Naik

```
import torch
import torchvision
from torch import nn
import torch.nn.functional as F
import matplotlib.pyplot as plt
import numpy as np
rng = np.random.default_rng(123456)
data = torchvision.datasets.MNIST(root='~/data', download=True)
data = data.data
data = data.float() / 255.
data = data.view(-1, 1, 28, 28)
print(data.shape)
      Downloading <a href="http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz">http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz</a>
      Downloading <a href="http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz">http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz</a> to /root/data/Mi
                                                           9913344/? [00:00<00:00, 27427523.14it/s]
      Extracting /root/data/MNIST/raw/train-images-idx3-ubyte.gz to /root/data/MNIST/raw
      Downloading <a href="http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz">http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz</a>
      Downloading <a href="http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz">http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz</a> to /root/data/MN
                                                           29696/? [00:00<00:00, 848016.04it/s]
      Extracting /root/data/MNIST/raw/train-labels-idx1-ubyte.gz to /root/data/MNIST/raw
      Downloading <a href="http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz">http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz</a>
      Downloading <a href="http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz">http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz</a> to /root/data/MNI
                                                           1649664/? [00:00<00:00, 7308248.03it/s]
      Extracting /root/data/MNIST/raw/t10k-images-idx3-ubyte.gz to /root/data/MNIST/raw
      Downloading <a href="http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz">http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz</a>
      Downloading <a href="http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz">http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz</a> to /root/data/MNI
                                                           5120/? [00:00<00:00, 63263.81it/s]
      Extracting /root/data/MNIST/raw/t10k-labels-idx1-ubyte.gz to /root/data/MNIST/raw
      /usr/local/lib/python3.7/dist-packages/torchvision/datasets/mnist.py:498: UserWarning: 1
        return torch.from_numpy(parsed.astype(m[2], copy=False)).view(*s)
      torch.Size([60000, 1, 28, 28])
class AutoEncoder(nn.Module):
     def __init__(self):
          super().__init__()
          self.encoder = nn.Sequential(
               nn.Flatten(),
               nn.Linear(28*28, 100),
               nn.ReLU(),
               nn.Linear(100, 10),
               nn.ReLU(),
          self.decoder = nn.Sequential(
               nn.Linear(10, 100),
               nn.ReLU(),
               nn.Linear(100, 28*28),
               nn.Sigmoid()
     def encode(self, x):
          return self.encoder(x)
     def decode(self, x):
          x = self.decoder(x)
          return x.view(-1,1,28,28)
     def forward(self, x):
          return self.decode(self.encode(x))
model = AutoEncoder().cuda()
opt = torch.optim.Adam(model.parameters())
```

```
for epoch in range(25):
   print(f'Epoch {epoch+1}/25')
    for i in range(0, data.shape[0], 32):
       x = data[i:i+32].cuda()
       x rec = model(x)
       loss = F.binary_cross_entropy(x_rec, x)
        opt.zero_grad()
       loss.backward()
       opt.step()
   data = data[rng.permutation(len(data))]
    print(f'\tloss: {loss.item():.4f}')
    Epoch 1/25
             loss: 0.1552
    Epoch 2/25
             loss: 0.1270
    Epoch 3/25
             loss: 0.1206
    Epoch 4/25
             loss: 0.1324
    Epoch 5/25
             loss: 0.1237
     Epoch 6/25
             loss: 0.1256
    Epoch 7/25
             loss: 0.1357
     Epoch 8/25
             loss: 0.1197
    Epoch 9/25
             loss: 0.1368
    Epoch 10/25
             loss: 0.1163
    Epoch 11/25
             loss: 0.1418
    Epoch 12/25
             loss: 0.1164
    Epoch 13/25
             loss: 0.1288
    Epoch 14/25
             loss: 0.1217
    Epoch 15/25
             loss: 0.1252
    Epoch 16/25
             loss: 0.1132
    Epoch 17/25
             loss: 0.1180
    Epoch 18/25
             loss: 0.1194
    Epoch 19/25
             loss: 0.1155
    Epoch 20/25
             loss: 0.1237
    Epoch 21/25
             loss: 0.1141
    Epoch 22/25
             loss: 0.1259
    Epoch 23/25
             loss: 0.1217
    Epoch 24/25
             loss: 0.1208
     Epoch 25/25
             loss: 0.1151
Double-click (or enter) to edit
plt.figure(figsize=(5,10))
for i in range(5):
   plt.subplot(5, 2, i*2+1, title=f'Train image')
   plt.imshow(np.squeeze(x[i].cpu()), cmap='gray')
   plt.axis('off')
   plt.subplot(5, 2, i*2+2, title='Reconstruction')
   with torch.no_grad(): plt.imshow(np.squeeze(x_rec[i].cpu()), cmap='gray')
   plt.axis('off')
```



```
# Sample two random images and encode
f = model.encode(x[0:2])
f1,f2 = f[0].unsqueeze(0),f[1].unsqueeze(0)

# Show reconstructions of interpolated codes
plt.figure(figsize=(20,5))
reconstructions = []
for i in range(20):
    v = i/19.
    f_interp = f1*(1-v) + f2*v
    with torch.no_grad():
        x_rec_interp = np.squeeze(model.decode(f_interp).cpu())
        reconstructions.append(x_rec_interp)

plt.subplot(2,10,i+1)
    plt.imshow(x_rec_interp, cmap='gray')
    plt.axis('off')
```



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