

LAB 4 > lab10.py > ...

```
1  # -----
2  # LAB 10: AUTOENCODER ON MNIST
3  # -----
4
5  import torch
6  import torch.nn as nn
7  import torch.optim as optim
8  from torch.utils.data import DataLoader
9  from torchvision import datasets, transforms
10 import matplotlib.pyplot as plt
11
12 # -----
13 # Step 1: Load the MNIST Dataset
14 # -----
15 transform = transforms.Compose([
16     transforms.ToTensor(),
17     transforms.Normalize((0.5,), (0.5,))
18 ])
19
20 train_dataset = datasets.MNIST(root='./data', train=True, transform=transform, download=True)
21 test_dataset = datasets.MNIST(root='./data', train=False, transform=transform, download=True)
22
23 train_loader = DataLoader(train_dataset, batch_size=128, shuffle=True)
24 test_loader = DataLoader(test_dataset, batch_size=128, shuffle=False)
25
26 # -----
27 # Step 2: Define the Autoencoder
28 # -----
29 class Autoencoder(nn.Module):
30     def __init__(self):
31         super(Autoencoder, self).__init__()
32         # Encoder
33         self.encoder = nn.Sequential(
34             nn.Linear(28 * 28, 128),
35             nn.ReLU(),
36             nn.Linear(128, 64),
37             nn.ReLU(),
38             nn.Linear(64, 12),
39             nn.ReLU(),
```

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```
29 class Autoencoder(nn.Module):
30     def __init__(self):
31         # Encoder
32         self.encoder = nn.Sequential(
33             nn.ReLU(),
34             nn.Linear(128, 64),
35             nn.ReLU(),
36             nn.Linear(64, 12),
37             nn.ReLU(),
38             nn.Linear(12, 3) # compressed representation
39         )
40     # Decoder
41     self.decoder = nn.Sequential(
42         nn.Linear(3, 12),
43         nn.ReLU(),
44         nn.Linear(12, 64),
45         nn.ReLU(),
46         nn.Linear(64, 128),
47         nn.ReLU(),
48         nn.Linear(128, 28 * 28),
49         nn.Tanh()
50     )
51
52     def forward(self, x):
53         x = self.encoder(x)
54         x = self.decoder(x)
55         return x
56
57 # -----
58 # Step 3: Initialize Model, Loss, Optimizer
59 # -----
60 device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
61 print("Using device:", device)
62
63 model = Autoencoder().to(device)
64 criterion = nn.MSELoss()
65 optimizer = optim.Adam(model.parameters(), lr=1e-3)
66
67 # -----
68 # Step 4: Train the Autoencoder
69 # -----
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69 # -----
70 # Step 4: Train the Autoencoder
71 # -----
72 num_epochs = 10
73
74 for epoch in range(num_epochs):
75     for data, _ in train_loader:
76         img = data.view(-1, 28 * 28).to(device)
77         output = model(img)
78         loss = criterion(output, img)
79
80         optimizer.zero_grad()
81         loss.backward()
82         optimizer.step()
83
84     print(f'Epoch [{epoch+1}/{num_epochs}], Loss: {loss.item():.4f}')
85
86 print("Training complete!")
87
88 # -----
89 # Step 5: Visualize Reconstruction
90 # -----
91 with torch.no_grad():
92     dataiter = iter(test_loader)
93     images, _ = next(dataiter)
94     images = images.view(-1, 28*28).to(device)
95     outputs = model(images)
96
97 # move to CPU for plotting
98 images = images.cpu()
99 outputs = outputs.cpu()
100
101 # show original and reconstructed images
102 fig, axes = plt.subplots(2, 8, figsize=(12, 3))
103 for i in range(8):
104     # original
105     axes[0, i].imshow(images[i].view(28, 28), cmap='gray')
106     axes[0, i].axis('off')
107     # reconstructed
108     axes[1, i].imshow(outputs[i].view(28, 28), cmap='gray')
```

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```
93     images, _ = next(dataiter)
94     images = images.view(-1, 28*28).to(device)
95     outputs = model(images)
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97     # move to CPU for plotting
98     images = images.cpu()
99     outputs = outputs.cpu()
100
101     # show original and reconstructed images
102     fig, axes = plt.subplots(2, 8, figsize=(12, 3))
103     for i in range(8):
104         # original
105         axes[0, i].imshow(images[i].view(28, 28), cmap='gray')
106         axes[0, i].axis('off')
107         # reconstructed
108         axes[1, i].imshow(outputs[i].view(28, 28), cmap='gray')
109         axes[1, i].axis('off')
110
111     plt.suptitle("Top: Original Images | Bottom: Reconstructed Images", fontsize=10)
112     plt.show()
113
```

```
PS C:\Users\aarus\OneDrive\Desktop\SRM\DLT> & C:/Python313/python.exe "c:/Users/aarus/OneDrive/Desktop/SRM/DLT/LAB 4/lab10.py"
● Using device: cpu
Epoch [1/10], Loss: 0.2091
Epoch [2/10], Loss: 0.1624
Epoch [3/10], Loss: 0.1530
Epoch [4/10], Loss: 0.1477
Epoch [5/10], Loss: 0.1408
Epoch [6/10], Loss: 0.1370
Epoch [7/10], Loss: 0.1502
Epoch [8/10], Loss: 0.1398
Epoch [9/10], Loss: 0.1231
Epoch [10/10], Loss: 0.1359
Training complete!
PS C:\Users\aarus\OneDrive\Desktop\SRM\DLT>
```