```
import torch as t
import torch.nn as nn
import torch.optim as optim
import torchvision
import torchvision.transforms as transforms
from torch.utils.data import DataLoader
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns
from sklearn.metrics import confusion_matrix, classification_report
transforms. To Tensor(), transforms. Normalize((0.5,),(0.5,))]) \\
train dataset=torchvision.datasets.MNIST(root='./data',train=True,download=True,transform=transform)
test\_dataset = torchvision.datasets. MNIST (root = './data', train = False, download = True, transform = transform)
image,label=train_dataset[0]
print("Image shape:",image.shape)
print("Number of training samples:",len(train_dataset))
image,label=test_dataset[0]
print("Image shape:",image.shape)
print("Number of testing samples:",len(test_dataset))
train_loader=DataLoader(train_dataset,batch_size=32,shuffle=True)
test_loader=DataLoader(test_dataset,batch_size=32,shuffle=False)
class CNNClassifier(nn.Module):
 def __init__(self):
   super(CNNClassifier,self).__init__()
    self.conv1=nn.Conv2d(in_channels=1,out_channels=32,kernel_size=3,padding=1)
    self.conv2=nn.Conv2d(in_channels=32,out_channels=64,kernel_size=3,padding=1)
    self.conv3=nn.Conv2d(in_channels=64,out_channels=128,kernel_size=3,padding=1)
   self.pool=nn.MaxPool2d(kernel_size=2,stride=2)
   self.fc1=nn.Linear(128*3*3,128)
   self.fc2=nn.Linear(128,64)
    self.fc3=nn.Linear(64,10)
 def forward(self,x):
    x=self.pool(t.relu(self.conv1(x)))
   x=self.pool(t.relu(self.conv2(x)))
    x=self.pool(t.relu(self.conv3(x)))
   x=x.view(x.size(0),-1)
   x=nn.functional.relu(self.fc1(x))
   x=nn.functional.relu(self.fc2(x))
   x=self.fc3(x)
   return x
from torchsummary import summary
model=CNNClassifier()
if t.cuda.is_available():
 device=t.device("cuda")
 model.to(device)
print("Name: Yashwini M")
print("Reg.no: 212223230249")
summary(model,input_size=(1,28,28))
criterion=nn.CrossEntropyLoss()
optimizer=optim.Adam(model.parameters(),lr=0.001)
def train_model(model,train_loader,num_epochs):
 for epoch in range(num_epochs):
   model.train()
    running_loss=0.0
    for images,labels in train_loader:
     if t.cuda.is_available():
       images,labels=images.to(device),labels.to(device)
     optimizer.zero_grad()
     outputs=model(images)
     loss=criterion(outputs,labels)
     loss.backward()
     optimizer.step()
     running_loss+=loss.item()
    print("Name: Yashwini M")
print("Reg.no: 212223230249")
train_model(model,train_loader,num_epochs=10)
def test_model(model, test_loader):
 model.eval()
 correct = 0
 total = 0
 all_preds = []
 all_labels = []
```

```
with t.no_grad():
   for images, labels in test_loader:
      if t.cuda.is_available():
       images, labels = images.to(device), labels.to(device)
     outputs = model(images)
      _, predicted = t.max(outputs, 1)
      total += labels.size(0)
     correct += (predicted == labels).sum().item()
     all_preds.extend(predicted.cpu().numpy())
      all_labels.extend(labels.cpu().numpy())
 accuracy = correct/total
 print("Name: Yashwini M")
 print("Register No: 212223230249")
 print(f"Test Accuracy: {accuracy:.4f}")
 cm = confusion_matrix(all_labels, all_preds)
 plt.figure(figsize=(8, 6))
 print("Name: Yashwini M")
 print("Register No: 212223230249")
 sns.heatmap(cm, annot=True, fmt="d", cmap="Blues", xticklabels=test_dataset.classes, yticklabels=test_dataset.classes)
 plt.xlabel("Predicted")
 plt.ylabel("Actual")
 plt.title("Confusion Matrix")
 plt.show()
 print("Name: Yashwini M")
 print("Register No: 212223230249")
 print("Classification Report:")
 print(classification_report(all_labels, all_preds, target_names=[str(i) for i in range(10)]))
test_model(model, test_loader)
def predict_image(model,image_index,dataset):
 model.eval()
 image,label=dataset[image_index]
 if t.cuda.is_available():
   image=image.to(device)
 with t.no_grad():
   output=model(image.unsqueeze(0))
    _,predicted=t.max(output,1)
 class_names=[str(i) for i in range(10)]
 print("Name: Yashwini M")
 print("Reg no: 212223230249")
 plt.imshow(image.cpu().squeeze(0),cmap='gray')
 plt.title(f"Actual: {class_names[label]}\nPredicted: {class_names[predicted.item()]}")
 plt.axis("off")
 plt.show()
 print(f"Actual: {class_names[label]}\nPredicted: {class_names[predicted.item()]}")
predict_image(model,image_index=80,dataset=test_dataset)
```

Image shape: torch.Size([1, 28, 28])
Number of training samples: 60000
Image shape: torch.Size([1, 28, 28])
Number of testing samples: 10000

Name: Yashwini M Reg.no: 212223230249

Layer (type)	Output Shape	Param #
Conv2d-1 MaxPool2d-2 Conv2d-3 MaxPool2d-4 Conv2d-5 MaxPool2d-6 Linear-7 Linear-8	[-1, 32, 28, 28] [-1, 32, 14, 14] [-1, 64, 14, 14] [-1, 64, 7, 7] [-1, 128, 7, 7] [-1, 128, 3, 3] [-1, 128] [-1, 64] [-1, 64]	320 0 18,496 0 73,856 0 147,584 8,256

Total params: 249,162 Trainable params: 249,162 Non-trainable params: 0

.

Input size (MB): 0.00

Forward/backward pass size (MB): 0.42

Params size (MB): 0.95

Estimated Total Size (MB): 1.37

Name: Yashwini M Reg.no: 212223230249

Epoch [1/10], Loss: 0.1547
Epoch [2/10], Loss: 0.0464
Epoch [3/10], Loss: 0.0338
Epoch [4/10], Loss: 0.0262
Epoch [5/10], Loss: 0.0217
Epoch [6/10], Loss: 0.0190
Epoch [7/10], Loss: 0.0137
Epoch [8/10], Loss: 0.0141
Epoch [9/10], Loss: 0.0120
Epoch [10/10], Loss: 0.0122

Name: Yashwini M
Register No: 212223230249
Test Accuracy: 0.9917
Name: Yashwini M
Register No: 212223230249

Confusion Matrix 0 - zero -1 - one -2 - two -3 - three -4 - four -Actual 5 - five -- 400 6 - six -7 - seven -8 - eight -9 - nine -- 0 zero 2 - two nine four five 3 - three ×. 7 - seven - 9 4-2 -

Name: Yashwini M Register No: 212223230249 Classification Report:

	precision	recall	t1-score	support
0	0.99	1.00	1.00	980
1	1.00	1.00	1.00	1135
2	1.00	1.00	1.00	1032
3	0.99	1.00	0.99	1010
4	1.00	0.98	0.99	982
5	0.99	0.99	0.99	892

Predicted