## ▼ BHARAT INTERN TASK-2

#### NUMBER RECOGNITION

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Handwritten digit recognition using MNIST dataset to detect the scanned images of handwritten digits

```
import torch
import torchvision
from torch import nn
import torch.optim as optim
import torchvision.transforms as transforms
import matplotlib.pyplot as plt
from tqdm.notebook import tqdm
```

#### Load the MNIST dataset

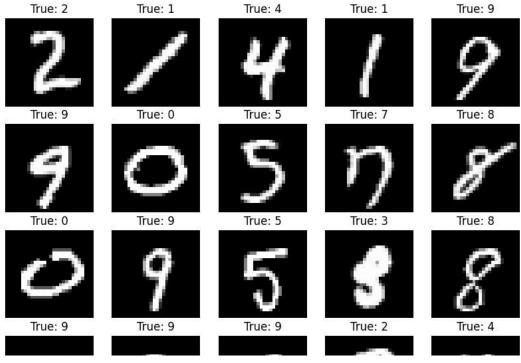
```
trainset = torchvision.datasets.MNIST(root='./data', train=True, download=True, transform=transforms.ToTensor())
trainloader = torch.utils.data.DataLoader(trainset, batch_size=64, shuffle=True)
testset = torchvision.datasets.MNIST(root='./data', train=False, download=True, transform=transforms.ToTensor())
testloader = torch.utils.data.DataLoader(testset, batch_size=64, shuffle=False)
                 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>
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                 4
```

### Visualize Dataset

images, labels = next(iter(trainloader))

show\_images(images, labels)

```
def show_images(images, labels, label_header="True"):
    figure = plt.figure(figsize=(10, 10))
    rows, cols = 5, 5
    for i in range(1, rows*cols+1):
        figure.add_subplot(rows, cols, i)
        plt.axis(False)
        plt.title(f"(label_header): {labels[i-1].item()}")
        plt.imshow(images[i-1].permute(1,2,0), cmap='gray')
```



# **Defining CNN model**

```
10.00
class CNN(nn.Module):
    def __init__(self):
        super(CNN, self).__init__()
        self.conv = nn.Sequential(
            nn.Conv2d(1, 32, kernel_size=3, padding=1),
            nn.ReLU(),
            nn.MaxPool2d(2, 2),
            nn.Conv2d(32, 64, kernel_size=3, padding=1),
            nn.ReLU(),
            nn.MaxPool2d(2, 2),
        )
        self.fc = nn.Sequential(
            nn.Flatten(),
            nn.Linear(64 * 7 * 7, 128),
nn.Linear(128, 10)
    def forward(self, x):
        x = self.conv(x)
        x = self.fc(x)
        return x
```

device = torch.device("cuda" if torch.cuda.is\_available() else "cpu")

# Initializing the model

```
net = CNN().to(device)
criterion = nn.CrossEntropyLoss()
optimizer = optim.SGD(net.parameters(), lr=0.001, momentum=0.9)
```

# Training the CNN Model

```
for epoch in range(5):
    running_loss = 0.0
    for idx, (inputs, labels) in tqdm(enumerate(trainloader), total=len(trainloader)):

    inputs = inputs.to(device)
    labels = labels.to(device)

    optimizer.zero_grad()

    outputs = net(inputs)
    loss = criterion(outputs, labels)
    loss.backward()
```

```
optimizer.step()
        running_loss += loss.item()
    print(f'Epoch: {epoch + 1}, Loss: {running_loss / 100:.3f}')
print('Training is done!')
     100%
                                                      938/938 [01:11<00:00, 10.23it/s]
     Epoch: 1, Loss: 8.940
                                                      938/938 [01:12<00:00, 14.74it/s]
     100%
     Epoch: 2, Loss: 2.734
                                                      938/938 [01:11<00:00, 14.48it/s]
     100%
     Epoch: 3, Loss: 1.921
                                                      938/938 [01:11<00:00, 15.58it/s]
     100%
     Epoch: 4, Loss: 1.426
                                                      938/938 [01:11<00:00, 14.80it/s]
     100%
     Epoch: 5, Loss: 1.133
     Training is done!
```

### **Evaluating the CNN Model on test data**

```
correct = 0
total = 0
with torch.no_grad():
    for (images, labels) in testloader:
        images = images.to(device)
        labels = labels.to(device)

        outputs = net(images)
        predicted = outputs.argmax(1)
        total += labels.size(0)
        correct += (predicted == labels).sum().item()

accuracy = 100 * correct / total
print(f'Accuracy on test set: {accuracy:.2f}%')

Accuracy on test set: 97.12%
```

```
images, labels = next(iter(testloader))
with torch.no_grad():
    pred = net(images.to(device))
    pred = pred.argmax(1)
```

show\_images(images, pred, "Predicted")

Predicted: 7	Predicted: 2	Predicted: 1	Predicted: 0	Predicted: 4
7	2	/	0	4
Predicted: 1	Predicted: 4	Predicted: 9	Predicted: 5	Predicted: 9
1	4	٩	4	9
Predicted: 0	Predicted: 6	Predicted: 9	Predicted: 0	Predicted: 1
0	0	9	0	j
Predicted: 5	Predicted: 9	Predicted: 7	Predicted: 3	Predicted: 4
5	9	7	$\mathcal{O}$	4
Predicted: 9	Predicted: 6	Predicted: 6	Predicted: 5	Predicted: 4
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