Imports and Data Loading

```
In [60]: # Load data from CSV files
    import pandas as pd
    data_df = pd.read_csv('train.csv')

# Split data into features (X) and labels (y)
X = data_df.drop('label', axis=1).values
y = data_df['label'].values

# Split data into 70% training and 30% testing
    from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state)

# Convert data to PyTorch tensors
    import torch
X_train_tensor = torch.tensor(X_train, dtype=torch.float32)
y_train_tensor = torch.tensor(y_train, dtype=torch.float32)
y_test_tensor = torch.tensor(X_test, dtype=torch.float32)
y_test_tensor = torch.tensor(y_test, dtype=torch.long)
```

Define Custom Dataset and Transformations

```
# Define a custom dataset class
In [61]:
         from torch.utils.data import DataLoader, Dataset
         class CustomDataset(Dataset):
             def __init__(self, data, labels):
                 self.data = data
                 self.labels = labels
             def __len__(self):
                 return len(self.data)
             def __getitem__(self, idx):
                 sample = {'image': self.data[idx], 'label': self.labels[idx]}
                 return sample
         # Create datasets and dataloaders
         batch size = 64
         train dataset = CustomDataset(X train tensor, y train tensor)
         test_dataset = CustomDataset(X_test_tensor, y_test_tensor)
         train_loader = DataLoader(train_dataset, batch_size=batch_size, shuffle=True)
         test loader = DataLoader(test dataset, batch size=batch size)
```

Define Neural Structure

```
In [78]: # Define a simple neural network architecture
import torch.nn as nn

class Net(nn.Module):
    def __init__(self):
        super(Net, self).__init__()
        self.fc1 = nn.Linear(784, 128)
        self.fc2 = nn.Linear(128, 64)
        self.fc3 = nn.Linear(64, 10)
```

```
def forward(self, x):
    x = x.view(-1, 784)
    x = torch.relu(self.fc1(x))
    x = torch.relu(self.fc2(x))
    x = self.fc3(x)
    return x

# Create an instance of the model
net = Net()

# Define Loss function and optimizer
import torch.optim as optim

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

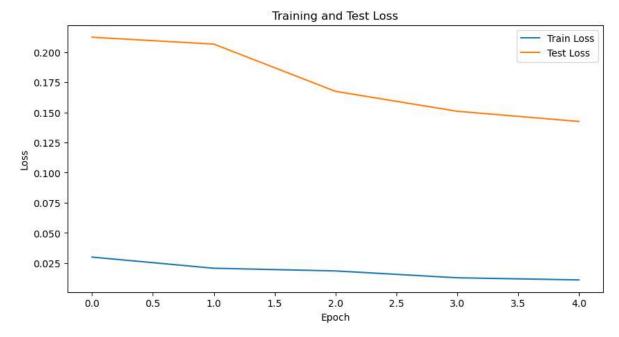
Training Loop with Progress Updates

```
In [79]: import matplotlib.pyplot as plt
         # Training loop with progress updates and loss tracking
         num epochs = 5
         train_losses = []
         test losses = []
         for epoch in range(num_epochs):
              running_loss = 0.0
             for i, data in enumerate(train_loader, 0):
                  inputs, labels = data['image'], data['label']
                 optimizer.zero_grad()
                 outputs = net(inputs)
                 loss = criterion(outputs, labels)
                 loss.backward()
                 optimizer.step()
                  running_loss += loss.item()
                 if i % 100 == 99:
                      print(f"Epoch [{epoch+1}/{num_epochs}], Batch [{i+1}/{len(train_loader
                      running_loss = 0.0
             # Calculate and store train loss for each epoch
             train_loss = running_loss / len(train_loader)
             train_losses.append(train_loss)
             # Calculate and store test loss for each epoch
             test loss = 0.0
             total_samples = 0
             with torch.no_grad():
                  for data in test loader:
                      inputs, labels = data['image'], data['label']
                      outputs = net(inputs)
                      loss = criterion(outputs, labels)
                      test_loss += loss.item() * labels.size(0)
                      total samples += labels.size(0)
                  test_loss /= total_samples
                  test_losses.append(test_loss)
         print("Training finished.")
```

```
Epoch [1/5], Batch [100/460], Loss: 1.1128
Epoch [1/5], Batch [200/460], Loss: 0.3548
Epoch [1/5], Batch [300/460], Loss: 0.2653
Epoch [1/5], Batch [400/460], Loss: 0.2666
Epoch [2/5], Batch [100/460], Loss: 0.1682
Epoch [2/5], Batch [200/460], Loss: 0.1890
Epoch [2/5], Batch [300/460], Loss: 0.1651
Epoch [2/5], Batch [400/460], Loss: 0.1833
Epoch [3/5], Batch [100/460], Loss: 0.1326
Epoch [3/5], Batch [200/460], Loss: 0.1274
Epoch [3/5], Batch [300/460], Loss: 0.1393
Epoch [3/5], Batch [400/460], Loss: 0.1286
Epoch [4/5], Batch [100/460], Loss: 0.1061
Epoch [4/5], Batch [200/460], Loss: 0.0987
Epoch [4/5], Batch [300/460], Loss: 0.0923
Epoch [4/5], Batch [400/460], Loss: 0.1061
Epoch [5/5], Batch [100/460], Loss: 0.0709
Epoch [5/5], Batch [200/460], Loss: 0.0792
Epoch [5/5], Batch [300/460], Loss: 0.0751
Epoch [5/5], Batch [400/460], Loss: 0.0874
Training finished.
```

Graph Plot

```
In [80]: # Plot the loss values
    plt.figure(figsize=(10, 5))
    plt.plot(train_losses, label='Train Loss')
    plt.plot(test_losses, label='Test Loss')
    plt.xlabel('Epoch')
    plt.ylabel('Loss')
    plt.title('Training and Test Loss')
    plt.legend()
    plt.show()
```



Testing the Model

```
In [81]: # Evaluate the model
    correct = 0
    total = 0
    with torch.no_grad():
```

```
for data in test_loader:
    inputs, labels = data['image'], data['label']
    outputs = net(inputs)
    _, predicted = torch.max(outputs.data, 1)
    total += labels.size(0)
    correct += (predicted == labels).sum().item()

print(f"Accuracy on test data: {100 * correct / total:.2f}%")
```

Accuracy on test data: 96.15%