```
import torchvision.datasets as datasets
import torchvision.transforms as transforms
from torch.utils.data import DataLoader
import torch
import torch.nn as nn
import torch.nn.functional as F
import torch.optim as optim
import matplotlib.pyplot as plt

transform = transforms.Compose([
    transforms.ToTensor(),
    transforms.Normalize((0.5, 0.5, 0.5),
                         (0.5, 0.5, 0.5))
])

train_dataset = datasets.CIFAR10(
    root='./data',
    train=True,
    download=True,
    transform=transform
)

test_dataset = datasets.CIFAR10(
    root='./data',
    train=False,
    download=True,
    transform=transform
)

train_loader = DataLoader(train_dataset, batch_size=64, shuffle=True)
test_loader = DataLoader(test_dataset, batch_size=64, shuffle=False)

images, labels = next(iter(train_loader))
```

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```

```
class SimpleCNN(nn.Module):
    def __init__(self):
        super(SimpleCNN, self).__init__()
        self.conv1 = nn.Conv2d(in_channels=3, out_channels=32, kernel_size=3, padding=1)
        self.bn1 = nn.BatchNorm2d(32)
        self.conv2 = nn.Conv2d(32, 64, kernel_size=3, padding=1)
        self.bn2 = nn.BatchNorm2d(64)
        self.pool = nn.MaxPool2d(kernel_size=2, stride=2)
        self.dropout = nn.Dropout(0.5)
        self.fc1 = nn.Linear(4096, 10)

    def forward(self, x):
        x = self.pool(F.relu(self.bn1(self.conv1(x))))
        x = self.pool(F.relu(self.bn2(self.conv2(x))))
        x = x.view(x.size(0), -1)
        x = self.dropout(x)
        x = self.fc1(x)
        return x
```

```
model = SimpleCNN()
device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
model.to(device)

criterion = nn.CrossEntropyLoss()
optimizer = optim.Adam(model.parameters(), lr=0.001)

num_epochs = 10
train_losses = []

for epoch in range(num_epochs):
    model.train()
    running_loss = 0.0

    for images, labels in train_loader:
        images, labels = images.to(device), labels.to(device)

        outputs = model(images)
        loss = criterion(outputs, labels)

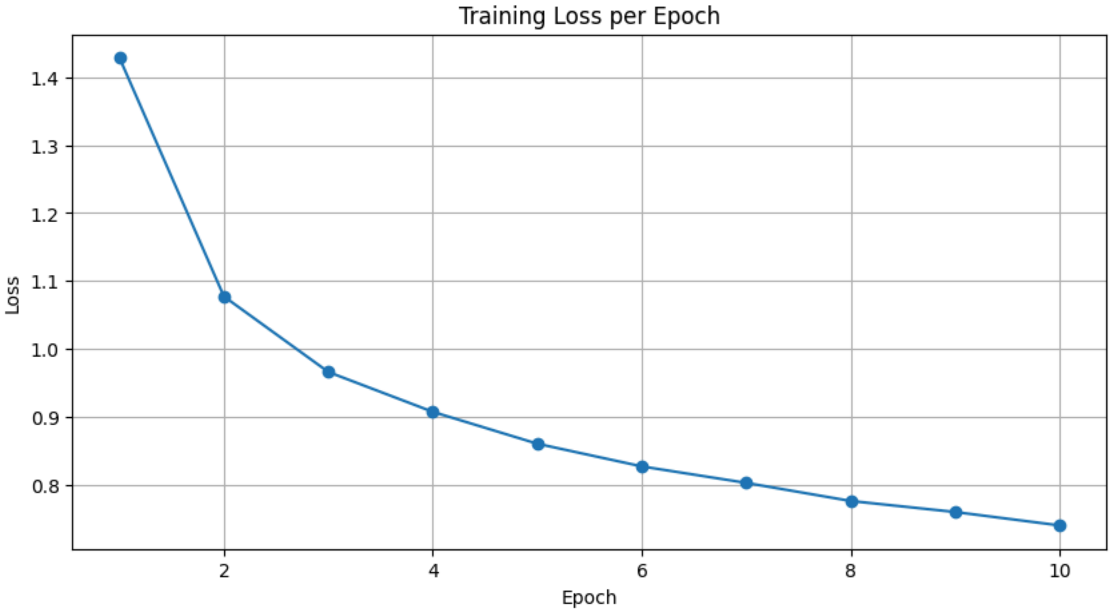
        optimizer.zero_grad()
        loss.backward()
        optimizer.step()

        running_loss += loss.item()
    epoch_loss = running_loss / len(train_loader)
    train_losses.append(epoch_loss)

    print(f"Эпоха [{epoch+1}/{num_epochs}], Потеря: {epoch_loss:.4f}")
```

```
Эпоха [1/10], Потеря: 1.4286
Эпоха [2/10], Потеря: 1.0773
Эпоха [3/10], Потеря: 0.9661
Эпоха [4/10], Потеря: 0.9075
Эпоха [5/10], Потеря: 0.8605
Эпоха [6/10], Потеря: 0.8272
Эпоха [7/10], Потеря: 0.8029
Эпоха [8/10], Потеря: 0.7763
Эпоха [9/10], Потеря: 0.7601
Эпоха [10/10], Потеря: 0.7403
```

```
plt.figure(figsize=(10, 5))
plt.plot(range(1, num_epochs + 1), train_losses, marker='o')
plt.title('Training Loss per Epoch')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.grid(True)
plt.show()
```



```
model.eval()
correct = 0
total = 0
test_loss = 0.0

with torch.no_grad():
    for images, labels in test_loader:
        images, labels = images.to(device), labels.to(device)

        outputs = model(images)
        loss = criterion(outputs, labels)
        test_loss += loss.item()

        _, predicted = torch.max(outputs, 1)
        total += labels.size(0)
        correct += (predicted == labels).sum().item()

avg_loss = test_loss / len(test_loader)
accuracy = 100 * correct / total

print(f"Test Loss: {avg_loss:.4f}, Test Accuracy: {accuracy:.2f}%")
```

```
Test Loss: 0.7431, Test Accuracy: 74.33%
```