#https://github.com/jhelto12/4105_Intro_to_ML/blob/main/Homework%207

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import torch
import torch.nn as nn
import torch.nn.functional as F
import torchvision
import torchvision.transforms as transforms
from torch.utils.data import DataLoader
class ResBlock(nn.Module):
    def __init__(self, n_chans):
        super().__init__()
        self.conv1 = nn.Conv2d(n_chans, n_chans, kernel_size=3, padding=1)
        self.conv2 = nn.Conv2d(n_chans, n_chans, kernel_size=3, padding=1)
    def forward(self, x):
        identity = x
        out = F.relu(self.conv1(x))
        out = self.conv2(out)
        out += identity
        return F.relu(out)
# Define NetResDeep
class NetResDeep(nn.Module):
    def __init__(self, n_chans1=32, n_blocks=10):
        super().__init__()
        self.n\_chans1 = n\_chans1
        self.conv1 = nn.Conv2d(3, n_chans1, kernel_size=3, padding=1)
        self.resblocks = nn.Sequential(
            *(n_blocks * [ResBlock(n_chans=n_chans1)])
        self.fc1 = nn.Linear(8 * 8 * n chans1, 32)
        self.fc2 = nn.Linear(32, 10) # 10 classes for CIFAR-10
    def forward(self, x):
        out = F.max_pool2d(torch.relu(self.conv1(x)), 2)
        out = self.resblocks(out)
        out = F.max_pool2d(out, 2)
        out = out.view(-1, 8 * 8 * self.n_chans1)
        out = torch.relu(self.fc1(out))
        out = self.fc2(out)
        return out
device = torch.device("cuda" if torch.cuda.is available() else "cpu")
transform = transforms.Compose([
    transforms.ToTensor().
    transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5)) # Normalize to [-1, 1]
1)
train_dataset = torchvision.datasets.CIFAR10(
    root='./data', train=True, transform=transform, download=True
test_dataset = torchvision.datasets.CIFAR10(
    root='./data', train=False, transform=transform, download=True
train_loader = DataLoader(train_dataset, batch_size=64, shuffle=True)
test_loader = DataLoader(test_dataset, batch_size=64, shuffle=False)
 Downloading <a href="https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz">https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz</a> to ./data/cifar-10-python.tar.gz
                   | 170M/170M [00:03<00:00, 47.9MB/s]
     Extracting ./data/cifar-10-python.tar.gz to ./data
     Files already downloaded and verified
# Initialize the model, loss function, and optimizer
model = NetResDeep(n_chans1=32, n_blocks=10).to(device)
criterion = nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.parameters(), lr=0.001)
# Training loop
num epochs = 200
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for epoch in range(num_epochs):
   model.train()
   running_loss = 0.0
   correct = 0
   total = 0
   for images, labels in train_loader:
        images, labels = images.to(device), labels.to(device)
       # Forward pass
       outputs = model(images)
       loss = criterion(outputs, labels)
       # Backward pass
       optimizer.zero_grad()
       loss.backward()
       optimizer.step()
       # Statistics
       running loss += loss.item()
        _, predicted = outputs.max(1)
       total += labels.size(0)
       correct += predicted.eq(labels).sum().item()
   # Print epoch stats
   train_accuracy = 100.0 * correct / total
   print(f"Epoch = 1}/{num_epochs}], Loss: {running_loss/len(train_loader):.4f}, Accuracy: {train_accuracy:.2f}%")
→ Epoch [1/200], Loss: 1.4736, Accuracy: 46.60%
    Epoch [2/200], Loss: 1.0955, Accuracy: 61.08%
    Epoch [3/200], Loss: 0.9493, Accuracy: 66.29%
    Epoch [4/200], Loss: 0.8719, Accuracy: 69.29%
    Epoch [5/200], Loss: 0.8082, Accuracy: 71.47%
    Epoch [6/200], Loss: 0.7579, Accuracy: 73.18%
    Epoch [7/200], Loss: 0.7250, Accuracy: 74.30%
    Epoch [8/200], Loss: 0.6924, Accuracy: 75.56%
    Epoch [9/200], Loss: 0.6685, Accuracy: 76.40%
    Epoch [10/200], Loss: 0.6433, Accuracy: 77.37%
    Epoch [11/200], Loss: 0.6230, Accuracy: 77.84%
    Epoch [12/200], Loss: 0.6045, Accuracy: 78.64%
    Epoch [13/200], Loss: 0.5843, Accuracy: 79.18%
    Epoch [14/200], Loss: 0.5709, Accuracy: 79.74%
    Epoch [15/200], Loss: 0.5588, Accuracy: 80.01%
    Epoch [16/200], Loss: 0.5548, Accuracy: 80.06%
    Epoch [17/200], Loss: 0.5400, Accuracy: 80.69%
    Epoch [18/200], Loss: 0.5294, Accuracy: 81.09%
    Epoch [19/200], Loss: 0.5208, Accuracy: 81.26%
    Epoch [20/200], Loss: 0.5126, Accuracy: 81.53%
    Epoch [21/200], Loss: 0.5071, Accuracy: 81.86%
    Epoch [22/200], Loss: 0.4995, Accuracy: 82.11%
    Epoch [23/200], Loss: 0.4925, Accuracy: 82.43%
    Epoch [24/200], Loss: 0.4829, Accuracy: 82.58%
    Epoch [25/200], Loss: 0.4840, Accuracy: 82.64%
    Epoch [26/200], Loss: 0.4767, Accuracy: 82.73%
    Epoch [27/200], Loss: 0.4742, Accuracy: 82.83%
    Epoch [28/200], Loss: 0.4706, Accuracy: 83.00%
    Epoch [29/200], Loss: 0.4597, Accuracy: 83.50%
    Epoch [30/200], Loss: 0.4555, Accuracy: 83.44%
    Epoch [31/200], Loss: 0.4547, Accuracy: 83.39%
    Epoch [32/200], Loss: 0.4496, Accuracy: 83.72%
    Epoch [33/200], Loss: 0.4450, Accuracy: 83.97%
    Epoch [34/200], Loss: 0.4416, Accuracy: 84.01%
    Epoch [35/200], Loss: 0.4372, Accuracy: 84.13%
    Epoch [36/200], Loss: 0.4393, Accuracy: 84.13%
    Epoch [37/200], Loss: 0.4336, Accuracy: 84.23%
    Epoch [38/200], Loss: 0.4292, Accuracy: 84.45%
    Epoch [39/200], Loss: 0.4271, Accuracy: 84.49%
    Epoch [40/200], Loss: 0.4248, Accuracy: 84.53%
    Epoch [41/200], Loss: 0.4150, Accuracy: 84.88%
    Epoch [42/200], Loss: 0.4183, Accuracy: 84.90%
    Epoch [43/200], Loss: 0.4123, Accuracy: 85.08%
    Epoch [44/200], Loss: 0.4112, Accuracy: 85.21%
    Epoch [45/200], Loss: 0.4082, Accuracy: 85.14%
    Epoch [46/200], Loss: 0.4017, Accuracy: 85.53%
    Epoch [47/200], Loss: 0.4060, Accuracy: 85.27%
    Epoch [48/200], Loss: 0.4025, Accuracy: 85.52%
    Epoch [49/200], Loss: 0.3980, Accuracy: 85.29%
    Epoch [50/200], Loss: 0.3952, Accuracy: 85.61%
    Epoch [51/200], Loss: 0.3948, Accuracy: 85.53%
    Epoch [52/200], Loss: 0.4006, Accuracy: 85.22%
```