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In [1]: from torchvision import transforms
        from PIL import Image
        # Image Preprocessing: Resize, Crop, Normalize
        transform = transforms.Compose([
           transforms.Resize((224, 224)),
           transforms.ToTensor(),
           transforms.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.225]),
        # Load Image
        image_path = "cat.jpg" # Replace with your image file path
        image = Image.open(image_path).convert("RGB")
        input_tensor = transform(image).unsqueeze(0) # Add batch dimension
In [2]: import torch
       from torchvision.models import alexnet
        # Load Pretrained AlexNet Model
       model = alexnet(pretrained=True)
       model.eval() # Set to evaluation mode
        # Forward Pass
        with torch.no_grad():
           output = model(input_tensor)
           predicted_class = torch.argmax(output, 1)
           print(f"Predicted class: {predicted_class.item()}")
      c:\Users\jerin\AppData\Local\Programs\Python\Python311\Lib\site-packages\torchvision\models\_utils.py:208: UserWarning: The parameter 'pretrained' is deprecated since 0.13 and may be removed in the future, please use 'weights' instead.
        warnings.warn(
       c:\Users\jerin\AppData\Local\Programs\Python\Python311\Lib\site-packages\torchvision\models\_utils.py:223: UserWarning: Arguments other than a weight enum or `None` for 'weights' are deprecated since 0.13 and may be removed in the future. The current behavior is equivalent to passing `weights=AlexN
       et_Weights.IMAGENET1K_V1`. You can also use `weights=AlexNet_Weights.DEFAULT` to get the most up-to-date weights.
        warnings.warn(msg)
      Predicted class: 285
In [3]: from torchvision.datasets import CIFAR10
        from torch.utils.data import DataLoader
        # Data Loading and Augmentation
       transform_train = transforms.Compose([
           transforms.Resize((224, 224)),
           transforms.RandomHorizontalFlip(),
           transforms.ToTensor(),
           transforms.Normalize(mean=[0.5, 0.5, 0.5], std=[0.5, 0.5, 0.5]),
       train_dataset = CIFAR10(root="./data", train=True, download=True, transform=transform_train)
       train_loader = DataLoader(train_dataset, batch_size=32, shuffle=True)
       test_dataset = CIFAR10(root="./data", train=False, download=True, transform=transform_train)
       test_loader = DataLoader(test_dataset, batch_size=32, shuffle=False)
      Files already downloaded and verified
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In [4]: import torch.nn as nn
       import torch.optim as optim
       # Modify AlexNet for CIFAR-10 (10 classes)
       model.classifier[6] = nn.Linear(4096, 10) # Output layer for 10 classes
        device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
        model.to(device)
        # Loss Function and Optimizer
        criterion = nn.CrossEntropyLoss()
        optimizer = optim.Adam(model.parameters(), lr=0.001)
        # Training Loop
        epochs = 5
        for epoch in range(epochs):
           model.train()
           running_loss = 0.0
           for inputs, labels in train_loader:
               inputs, labels = inputs.to(device), labels.to(device)
               optimizer.zero_grad()
               outputs = model(inputs)
               loss = criterion(outputs, labels)
               loss.backward()
               optimizer.step()
               running_loss += loss.item()
           print(f"Epoch {epoch+1}/{epochs}, Loss: {running_loss/len(train_loader)}")
       Epoch 1/5, Loss: 2.0272153523665395
      Epoch 2/5, Loss: 1.5890526312021438
      Epoch 3/5, Loss: 1.3956197938397383
      Epoch 4/5, Loss: 1.273427989874905
      Epoch 5/5, Loss: 1.192788677343709
In [5]: # Validation Loop
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model.eval()
correct = 0
total = 0
with torch.no_grad():
 for inputs, labels in test_loader:
 inputs, labels = inputs.to(device), labels.to(device)
 outputs = model(inputs)
 _, predicted = torch.max(outputs, 1)
 total += labels.size(0)
 correct += (predicted == labels).sum().item()

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

Accuracy: 63.76%