## **AlexNet**

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import os
os. getcwd()
     '/content'
import torch as T
import torch.nn as nn
import torch.optim as optim
import torch.nn.functional as F
import torchvision
import torchvision.transforms as transforms
device = 'cuda' if T. cuda. is available() else 'cpu'
device
     'cuda'
transform = transforms.Compose(
        [transforms. Resize((224, 224)),
          transforms. ToTensor(),
          transforms. Normalize (mean = (0.5, 0.5, 0.5), std = (0.5, 0.5, 0.5))
batch size = 32
image size = (32,
                     32,
                          3)
train set = torchvision.datasets.CIFAR10(root='./data', train=True, download=True, transform=t
train loader = T.utils.data.DataLoader(train set, batch size=batch size, shuffle=True,
                                                                                            num wor
test_set = torchvision.datasets.CIFAR10(root='./data', train=False, download=True, transform=t
test loader = T.utils.data.DataLoader(test set, batch size=batch size, shuffle=False, num work
     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/cifa
                                                  170499072/? [00:11<00:00, 17129467.27it/s]
     Extracting ./data/cifar-10-python.tar.gz to ./data
     Files already downloaded and verified
class AlexNet(nn.Module):
        def __init__(self, num_classes=1000):
                super(AlexNet, self). init ()
                self.features = nn.Sequential(
                        nn.Conv2d(3, 64, kernel_size=11, stride=4, padding=2),
                        nn.ReLU(inplace=True),
                        nn. MaxPool2d (kernel size=3, stride=2),
                        nn. Conv2d(64, 192, kernel size=5, padding=2),
                        nn. ReLU(inplace=True),
                        nn. MaxPool2d(kernel size=3, stride=2),
                        nn. Conv2d(192, 384, kernel size=3, padding=1),
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nn. ReLU(inplace=True),
                         nn. Conv2d (384,
                                         256,
                                                kernel size=3,
                         nn. ReLU(inplace=True),
                         nn. Conv2d (256, 256, kernel size=3,
                                                               padding=1),
                         nn. ReLU(inplace=True),
                         nn. MaxPool2d(kernel size=3, stride=2),
                )
                self.classifier = nn.Sequential(
                         nn. Dropout(),
                         nn. Linear (256 * 6 * 6,
                                                     4096),
                         nn. ReLU(inplace=True),
                         nn. Dropout (),
                         nn. Linear (4096, 4096),
                         nn. ReLU(inplace=True),
                         nn. Linear (4096, num classes),
                )
        def forward(self, x):
                x = self. features(x)
                x = x. view(x. size(0), 256 * 6 * 6)
                x = self.classifier(x)
                return x
net = AlexNet()
net. to (device)
     AlexNet(
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        (features): Sequential(
          (0): Conv2d(3, 64, \text{kernel size}=(11, 11), \text{stride}=(4, 4), padding}=(2, 2))
          (1): ReLU(inplace=True)
          (2): MaxPool2d(kernel_size=3, stride=2, padding=0, dilation=1, ceil_mode=False)
          (3): Conv2d(64, 192, kernel_size=(5, 5), stride=(1, 1), padding=(2, 2))
          (4): ReLU(inplace=True)
          (5): MaxPool2d(kernel size=3, stride=2, padding=0, dilation=1, ceil mode=False)
          (6): Conv2d(192, 384, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
          (7): ReLU(inplace=True)
          (8): Conv2d(384, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
          (9): ReLU(inplace=True)
          (10): Conv2d(256, 256, \text{ kernel size}=(3, 3), \text{ stride}=(1, 1), \text{ padding}=(1, 1))
          (11): ReLU(inplace=True)
          (12): MaxPool2d(kernel size=3, stride=2, padding=0, dilation=1, ceil mode=False)
       )
        (classifier): Sequential(
          (0): Dropout (p=0.5, inplace=False)
          (1): Linear(in_features=9216, out_features=4096, bias=True)
          (2): ReLU(inplace=True)
          (3): Dropout (p=0.5, inplace=False)
          (4): Linear (in features=4096, out features=4096, bias=True)
          (5): ReLU(inplace=True)
          (6): Linear(in features=4096, out features=1000, bias=True)
     )
criterion = nn. CrossEntropyLoss()
# also the ontimizer
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optimizer = optim. SGD (net. parameters (), 1r=0.001, momentum=0.9)
train_loss = []
test_loss = []
train acc = []
test_acc = []
for epoch in range (1, 33): #32 epoch
       running_loss = .0
       correct = 0
       total = 0
       for i, data in enumerate(train_loader):
               # get the inputs
               inputs, labels = data
               if device == 'cuda':
                      inputs, labels = inputs. to(device), labels. to(device)
               # reset the parameter gradients
               optimizer.zero grad()
               # forward
               outputs = net(inputs)
               loss = criterion(outputs, labels)
               # backward
               loss. backward()
               # optimize
               optimizer. step()
               running loss += loss.item()
               , predicted = T. max(outputs. data, 1)
               total += labels.size(0)
               correct += (predicted == labels).sum().item()
       running_loss /= len(train_loader)
       train_loss.append(running_loss)
       running acc = correct / total
       train_acc. append(running_acc)
       if epoch \% 4 == 0:
               print('\nEpoch: {}'.format(epoch))
               print('Train Acc. => {:.3f}%'.format(100 * running_acc), end=' ')
               print('Train Loss => {:.5f}'.format(running_loss))
       with T.no_grad():
               correct = 0
               total = 0
               test running loss = .0
               for data in test loader:
                      inputs, labels = data
                      if device == 'cuda':
                              inputs, labels = inputs.to(device), labels.to(device)
                      outputs = net(inputs)
                      loss = criterion(outputs, labels)
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test_running_loss += loss.item()
                        , predicted = T. max(outputs. data, 1)
                        total += labels. size(0)
                        correct += (predicted == labels).sum().item()
                test running loss /= len(test loader)
                test_loss.append(test_running_loss)
                test_running_acc = correct / total
                test acc. append (test running acc)
                if epoch % 4 == 0:
                        print('Test Acc. => \{:.3f}\%'.format(100 * test_running_acc),
                        print('Test Loss
                                             => \{\text{:.5f}}'\text{.format(test running loss)}
print('Finished Training')
     Epoch: 4
     Train Acc. => 55.230% | Train Loss => 1.23943
     Test Acc. => 57.490% | Test Loss => 1.16939
     Epoch: 8
     Train Acc. => 73.354% | Train Loss => 0.76147
     Test Acc. => 72.400% | Test Loss => 0.79087
     Epoch: 12
     Train Acc. => 81.438% | Train Loss => 0.53863
     Test Acc. \Rightarrow 76.890% | Test Loss \Rightarrow 0.67115
     Epoch: 16
     Train Acc. => 86.848% | Train Loss => 0.37812
     Test Acc. \Rightarrow 80.370% | Test Loss \Rightarrow 0.60078
     Epoch: 20
     Train Acc. => 90.918% | Train Loss => 0.25543
     Test Acc. => 81.050\% | Test Loss => 0.60969
     Epoch: 24
     Train Acc. => 93.958% | Train Loss => 0.16896
     Test Acc. => 81.040% | Test Loss => 0.69141
     Epoch: 28
     Train Acc. => 95.906% | Train Loss => 0.11479
     Test Acc. => 82.110% | Test Loss => 0.71463
     Epoch: 32
     Train Acc. => 97.228% | Train Loss => 0.07889
     Test Acc. => 82.770% | Test Loss => 0.75348
     Finished Training
```

T. save(AlexNet() , './trained model')

✓ 1秒 完成時間: 下午7:26

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