《计算机视觉》实验报告

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实验 10

一. 任务1

a) 核心代码:

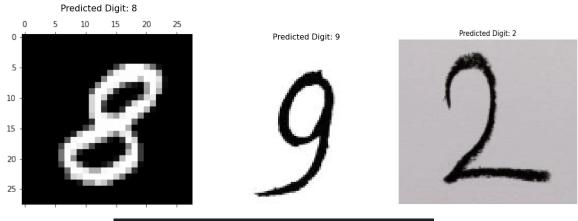
```
# 定义 CNN 模型
class Net(torch.nn.Module):
         super(Net, self).__init__()
         self.model = torch.nn.Sequential(
              torch.nn.Conv2d(in_channels=1, out_channels=16, kernel_size=3, stride=1, padding=1),
              torch.nn.ReLU(),
              torch.nn.MaxPool2d(kernel_size=2, stride=2),
              torch.nn.Conv2d(in_channels=16, out_channels=32, kernel_size=3, stride=1, padding=1),
              torch.nn.ReLU(),
              torch.nn.MaxPool2d(kernel_size=2, stride=2),
              torch.nn.Conv2d(in_channels=32, out_channels=64, kernel_size=3, stride=1, padding=1),
              torch.nn.ReLU(),
              torch.nn.Flatten(),
              torch.nn.Linear(in_features=7 * 7 * 64, out_features=128),
              torch.nn.ReLU(),
              torch.nn.Softmax(dim=1)
    def forward(self, input):
         output = self.model(input)
         return output
lossF = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(net.parameters())
history = { 'Test Loss': [], 'Test Accuracy': []}
```

```
for epoch in range(1, EPOCHS + 1):
     processBar = tqdm(trainDataLoader, unit='step')
     net.train(True)
     for step, (trainImgs, labels) in enumerate(processBar):
         trainImgs = trainImgs.to(device)
         labels = labels.to(device)
         net.zero_grad()
         outputs = net(trainImgs)
         loss = lossF(outputs, labels)
         predictions = torch.argmax(outputs, dim=1)
         accuracy = torch.sum(predictions == labels).item() / labels.shape[0]
         loss.backward()
         optimizer.step()
         processBar.set\_description(f"[\{epoch\}/\{EPOCHS\}]\ Loss:\ \{loss.item():.4f\},\ Acc:\ \{accuracy:.4f\}")
         if step == len(processBar) - 1:
              correct, totalLoss = 0, 0
              net.train(False)
              with torch.no_grad():
                    for testImgs, labels in testDataLoader:
                        testImgs = testImgs.to(device)
                        labels = labels.to(device)
                        outputs = net(testImgs)
                        loss = lossF(outputs, labels)
                        predictions = torch.argmax(outputs, dim=1)
                        totalLos += loss
                        correct += torch.sum(predictions == labels)
                   testAccuracy = correct.item() / (BATCH_SIZE * len(testDataLoader))
                   testLoss = totalLoss / len(testDataLoader)
                   history['Test Loss'].append(testLoss.item())
                   history['Test Accuracy'].append(testAccuracy)
              processBar.set_description(f"[{epoch}/{EPOCHS}] Loss: {loss.item():.4f}, Acc: {accuracy:.4f}, Test Loss:
{testLoss.item():.4f}, Test Acc: {testAccuracy:.4f}"
torch.save(net.state_dict(), './model.pth')
```

b) 实验结果截图

网络结构示意图

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Conv2d-1	[-1, 16, 28, 28]	160
ReLU-2	[-1, 16, 28, 28]	0
MaxPool2d-3	[-1, 16, 14, 14]	0
Conv2d-4	[-1, 32, 14, 14]	4,640
ReLU-5	[-1, 32, 14, 14]	0
MaxPool2d-6	[-1, 32, 7, 7]	0
Conv2d-7	[-1, 64, 7, 7]	18,496
ReLU-8	[-1, 64, 7, 7]	0
Flatten-9	[-1, 3136]	0
Linear-10	[-1, 128]	401,536
ReLU-11	[-1, 128]	0
Linear-12	[-1, 10]	1,290
Softmax-13	[-1, 10]	0



Finished Training Accuracy on test images: 98 %

c) 实验小结

通过这次实验,我学习到了卷积神经网络的简单应用,这次实验构建了一个简单的网络进行训练,一进行了 5 轮训练,最后准确率都高达 98%,很好的完成了实验目的,对深度学习有了直观的感受。