

Lecture Notes of NOAI Training

Day 2, June 8, 2024

1. MNIST

```
1  # 定义数据集
2  test_dataset=FashionDataset(
3      datadir='FashionMNIST/raw',
4      transform=T.Compose([
5          T.ToTensor(),
6          T.Normalize((0.5,), (0.5,))
7      ]),
8      is_train=True
9  )
10
11 # 定义测试集
12 test_dataset=FashionDataset(
13     datadir='FashionMNIST/raw',
14     transform=T.Compose([
15         T.ToTensor(),
16         T.Normalize((0.5,), (0.5,))
17     ]),
18     is_train=False
19 )
20
21 # 网络结构
22 def __init__(self):
23     super(LeNet, self).__init__()
24     self.conv1=nn.Conv2d(1,6,5,1,2)
25     self.pool1=nn.MaxPool2d(2,2)
26     self.conv2=nn.Conv2d(6,16,5)
27     self.pool2=nn.MaxPool2d(2,2)
28     self.fc1=nn.Linear(16*5*5,120)
29     self.fc2=nn.Linear(120,84)
30     self.fc3=nn.Linear(84,10)
31
32 # 前向传播
33 def forward(self, x):
34     x=F.relu(self.conv1(x))
35     x=self.pool1(x)
36     x=F.relu(self.conv2(x))
37     x=self.pool2(x)
38     x=x.view(x.size()[0],-1)
39     x=F.relu(self.fc1(x))
```

```

40     x=F.relu(self.fc2(x))
41     x=self.fc3(x)
42     return x
43
44 # 训练
45 def train(epoch):
46     model.train()
47     total_loss=0
48     for iter,(data,target) in enumerate(train_loader):
49         opt.zero_grad()
50         output=model(data)
51         loss=loss_fn(output,target)
52         loss.backward()
53         opt.step()
54         total_loss+=loss.item()
55     total_loss/=len(train_loader)
56     print(f'Epoch: {epoch} Loss: {total_loss:.3f}')
57
58 # 测试
59 def test(epoch):
60     model.eval()
61     correct=0
62     tot=0
63     for data,target in test_loader:
64         output=model(data)
65         pred=[]
66         for i in range(output.size(0)):
67             max_index=0
68             max_value=output[i][0]
69             for j in range(1,output.size(1)):
70                 if output[i][j]>max_value:
71                     max_value=output[i][j]
72                     max_index=j
73             pred.append(max_index)
74         for i in range(len(pred)):
75             if pred[i]==target[i]:
76                 correct+=1
77         tot+=data.shape[0]
78     print(f'Test Epoch: {epoch} Accuracy: {correct/tot*100:.2f}%')
79
80 # 训练模型
81 for epoch in range(10):
82     train(epoch)
83     test(epoch)

```

2. 多变量

```
1
2 # 训练
3 def train(epoch):
4     #...
5     #for...
6         output=model(data)
7         output=torch.sigmoid(output)
8
9 # 测试
10 def test(epoch):
11     #...
12     #for...
13         output=model(data)
14         output=torch.sigmoid(output)
15         _,output_indices=torch.topk(output,k=2,dim=-1,largest=True,sorted=True)
16         _,target_indices=torch.topk(target,k=2,dim=-1,largest=True,sorted=True)
17         output_indices=output_indices.numpy()
18         output_indices=target_indices.numpy()
19     #for...
20         if np.array_equal(output_indices[i],target_indices[i]):
21             correct+=1
```

3. GAN 生成对抗网络

流程：

1. 设置随机种子
2. dataloader
3. 判别器网络结构
4. 生成器网络结构
5. 实例化判别器和生成器
6. 使用未训练的生成器生成数据并可视化
7. 超参数
8. 损失函数
9. 优化器设置
10. 训练判别器和生成器
11. 模型测评

$$\begin{aligned}
 & \text{nn.Linear} \leftarrow 2\text{D} \\
 & \text{nn.Linear} \rightarrow \text{nn.ReLU} \rightarrow \text{nn.Dropout} \\
 & \text{nn.Linear } 2\text{D} \rightarrow 1\text{D}
 \end{aligned}
 \tag{1}$$