# 案例7: 使用ResNet5对Mnist做分类

## 2023年4月26日

```
[1]: import sys
    sys.path.append(r"D:\Rhitta_GPU")
    import numpy as np
    import cupy as cp
    import rhitta.nn as nn
    from sklearn.preprocessing import OneHotEncoder
    onehot_encoder = OneHotEncoder(sparse_output=False)
```

### 第一步:载入数据集

```
[2]: loader=nn.MnistLoader()
train_x,number_labels=loader.load(r"D:\Rhitta_GPU\data\dataset")
labels = cp.array(onehot_encoder.fit_transform(cp.asnumpy(number_labels).

→reshape(-1, 1)))

train_x.shape,number_labels.shape,labels.shape
```

[2]: ((60000, 784), (60000,), (60000, 10))

#### 第二步: 构造模型

```
[3]: model=nn.ResNet_simple(in_channels=1) print(model)
```

#### Model:

Layer 1: Conv2D(in\_channels=1,out\_channels=4,kernel\_size=5,stride=1,padding=0)

kernel: [4,5,5] num\_params: 104

Layer 2: LayerNorm()

```
Layer 3: ReLU()
    Layer 4: Pooling(in_channels=4,window_size=2,stride=2,mode=MaxPooling)
    Layer 5: Conv2D(in_channels=4,out_channels=4,kernel_size=3,stride=1,padding=1)
    kernel: [4,3,3] num_params: 40
    Layer 6: LayerNorm()
    Layer 7: ReLU()
    Layer 8: Conv2D(in_channels=4,out_channels=4,kernel_size=3,stride=1,padding=1)
    kernel : [4,3,3] num_params: 40
    Layer 9: LayerNorm()
    Layer 10: ResAdd()
    Layer 11: Conv2D(in_channels=4,out_channels=4,kernel_size=5,stride=1,padding=0)
    kernel: [4,5,5] num_params: 104
    Layer 12: Pooling(in_channels=4,window_size=4,stride=4,mode=AveragePooling)
                             num_params: 170
    Layer 13: Linear(16,10)
    Total params: 458
    构造完整计算图:输入输出节点,模型,损失
[4]: x=nn.to_tensor(size=(28,28))
    label=nn.to_tensor(size=(1,10))
    out=model(x)
    predict=nn.Softmax(out)
    loss=nn.CrossEntropyLoss(out,label)
    第三步: 选择并初始化优化器
[5]: learning_rate = 0.01
    optimizer = nn.Adam(nn.default_graph, loss, learning_rate=learning_rate)
    第四步: 开始训练
    本框架慢的离谱,就拿16个样本跑通试一试
[6]: epochs = 20
    batch_size = 4
```

for epoch in range(epochs):

```
N = 16
count = 0
# 遍历样本训练
for i in range(N):
    x.set_value(train_x[i].reshape(28,28))
    label.set_value(labels[i])
    optimizer.one_step()
    count+=1
    if count >= batch_size:
        optimizer.update()
        count=0
# 遍历样本求准确率
pred=[]
for i in range(N):
   x.set_value(train_x[i].reshape(28,28))
   label.set_value(labels[i])
   predict.forward()
    pred.append(predict.value.flatten())
temp=(cp.array(pred).argmax(axis=1) == number_labels[:N])
accuracy=temp.sum()/N
print("epoch:{} accuracy:{}".format(epoch+1,accuracy))
```

```
epoch:1 accuracy:0.125
epoch:2 accuracy:0.125
epoch:3 accuracy:0.25
epoch:4 accuracy:0.1875
epoch:5 accuracy:0.1875
epoch:6 accuracy:0.125
epoch:7 accuracy:0.0625
epoch:8 accuracy:0.125
epoch:9 accuracy:0.125
epoch:10 accuracy:0.25
epoch:11 accuracy:0.25
epoch:12 accuracy:0.25
epoch:13 accuracy:0.25
epoch:14 accuracy:0.25
epoch:15 accuracy:0.25
```

epoch:16 accuracy:0.4375

epoch:17 accuracy:0.5

epoch:18 accuracy:0.5

epoch:19 accuracy:0.5625

epoch:20 accuracy:0.5625