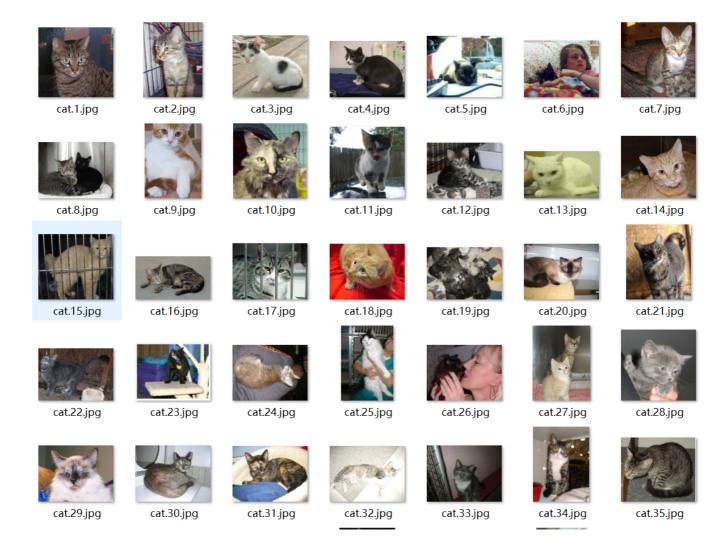
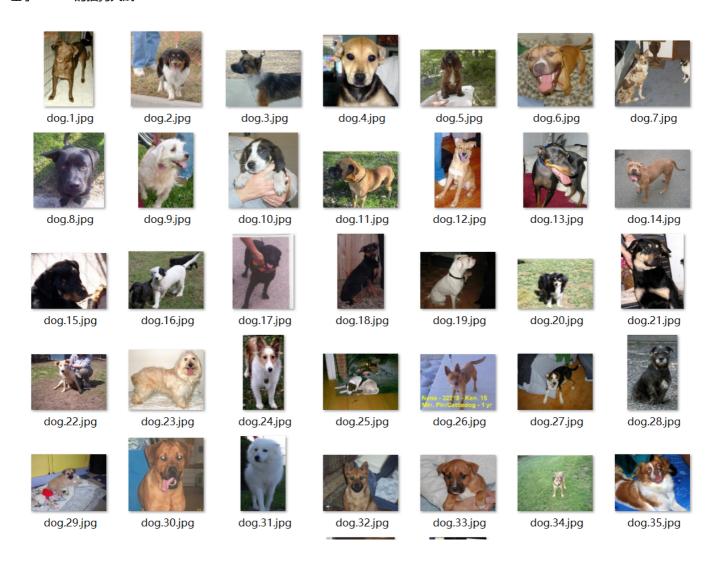
基于Alexnet的猫狗大战

数据来源

数据来自于kaggle,网址为https://www.kaggle.com/tongpython/cat-and-dog。测试集包含4000张猫、4005张狗,测试集包含了1011张猫、1012张狗。部分图片如下。





可以看到数据较为复杂,图片尺寸不一旦部分图片存在干扰。

网络结构

网络利用python中pytorch建立,更改分类数为2

```
class AlexNet(nn.Module):
   def __init__(self, num_classes=2):
        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),
            nn.ReLU(inplace=True),
            nn.Conv2d(384, 256, kernel_size=3, padding=1),
            nn.ReLU(inplace=True),
            nn.Conv2d(256, 256, kernel_size=3, padding=1),
            nn.ReLU(inplace=True),
            nn.MaxPool2d(kernel_size=3, stride=2),
        )
```

训练过程

数据预处理

由于图片尺寸不一致,因此先将图片较短的一边缩放到224像素,再居中裁剪形成224*224*3,最后进行归一 化,使得数据性质较好

建立网络并设置参数

batch_size设置较大,加快收敛速度,多线程读取加速IO,学习率设置为0.0001。

```
device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')
  batch_size = 256
  lr= 0.0001
  net = AlexNet().to(device)
    trainloader = torch.utils.data.DataLoader(dataset, batch_size=batch_size,
  shuffle=True, num_workers=12)
  testloader = torch.utils.data.DataLoader(testset, batch_size=batch_size,
  shuffle=False, num_workers=12)
  criterion = nn.CrossEntropyLoss()
  optimizer = torch.optim.Adam(net.parameters(), lr=lr)
```

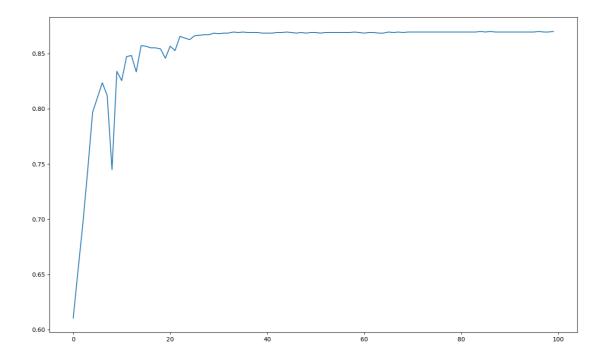
开始训练

```
for epoch in range(100):
    running_loss = []
    t1 = time.time()
    for i, data in enumerate(trainloader, ∅):
        inputs, labels = data
        inputs, labels = inputs.to(device), labels.to(device)
        optimizer.zero grad()
        outputs = net(inputs)
        loss = criterion(outputs, labels)
        loss.backward()
        optimizer.step()
        running_loss.append(loss.item())
        loss_seq.append(loss.item())
    t2 = time.time()
    t = t2 - t1
    print('epoch:%d loss: %.8f time:%.2fs' % (epoch + 1, np.mean(running_loss),
t))
    ac = ac_rate(testloader, net, device)
    ac_seq.append(ac)
```

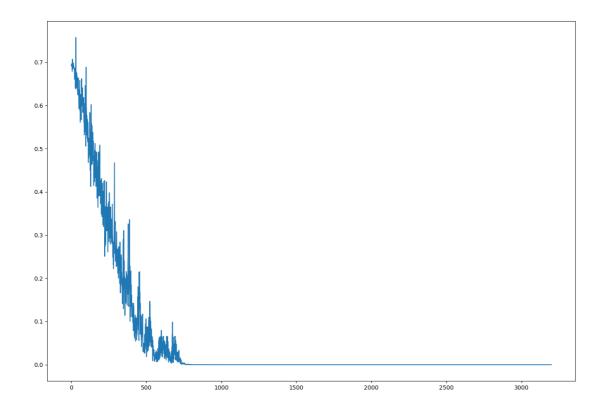
训练结果

下面是训练过程中部分输出 (删去了中间部分)

测试集准确率 (每epoch求一次) 如下图, 随着训练轮次逐步上升, 慢慢收敛到87%左右



loss (每batch求平均) 如下图,随着训练轮次逐渐下降至接近0



完整代码

```
import matplotlib.pyplot as plt
import torch
from torch import nn
import torchvision.datasets as dset
from torchvision import transforms
import numpy as np
import time
class AlexNet(nn.Module):
    def __init__(self, num_classes=2):
        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),
            nn.ReLU(inplace=True),
            nn.Conv2d(384, 256, kernel_size=3, padding=1),
            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.Linear(256 * 6 * 6, 4096),
            nn.ReLU(inplace=True),
            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
def ac_rate(test_dataloader, net, device):
    total = 0
    ac = 0
    for i, data in enumerate(test_dataloader, ∅):
        inputs, labels = data
        inputs, labels = inputs.to(device), labels.to(device)
        outputs = net(inputs)
        outputs = outputs.argmax(dim=1)
        ac += int((outputs == labels).sum())
        total += len(inputs)
    print(ac, "/", total, sep='')
```

```
return ac / total
def main():
   transform = transforms.Compose([
       transforms.Resize(224), #缩放图片,保持长宽比不变,最短边的长为224像素,
       transforms.CenterCrop(224), # 从中间切出 224*224的图片
       transforms.ToTensor(), # 将图片转换为Tensor,归一化至[0,1]
       transforms.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224,
0.225]) # 标准化至[-1,1]
   ])
   dataset = dset.ImageFolder("./training_set", transform=transform)
   testset = dset.ImageFolder("./test_set", transform=transform)
   device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')
   batch_size = 256
   lr = 0.0001
   net = AlexNet().to(device)
   trainloader = torch.utils.data.DataLoader(dataset, batch_size=batch_size,
shuffle=True, num_workers=12)
   testloader = torch.utils.data.DataLoader(testset, batch_size=batch_size,
shuffle=False, num_workers=12)
   criterion = nn.CrossEntropyLoss()
   optimizer = torch.optim.Adam(net.parameters(), lr=lr)
   print("start training")
   t0 = time.time()
   for epoch in range(100):
        running_loss = []
       t1 = time.time()
       for i, data in enumerate(trainloader, ∅):
           inputs, labels = data
           inputs, labels = inputs.to(device), labels.to(device)
           optimizer.zero grad()
           outputs = net(inputs)
           loss = criterion(outputs, labels)
           loss.backward()
           optimizer.step()
           running_loss.append(loss.item())
           loss seq.append(loss.item())
       t2 = time.time()
       t = t2 - t1
        print('epoch:%d loss: %.8f time:%.2fs' % (epoch + 1,
np.mean(running loss), t))
        ac = ac_rate(testloader, net, device)
        ac_seq.append(ac)
   print('Finished Training')
   t = time.time() - t0
    print("total time:%.1fs" % (t))
loss_seq = []
ac seq = []
```

```
if __name__ == '__main__':
    main()
```