

木薯叶病分类直播答疑

导师:

本次答疑内容

Course content



2、学员问题解答

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重难点知识串讲

梳理本阶段的重难点知识



学员问题解答

对大家提出的问题进行解答



@于东阳:

想输出分类的结果,每个类别的概率是多少,应该怎么操作呢?

```
#for epoch in range(CFG['epochs']-3):
    for i, epoch in enumerate(CFG['used_epochs']):
        print(i)
        print(epoch)
        print(fold)
        model.load_state_dict(torch.load('../input/tf-efficientnet-b4-ns0102/{}_fold_{}_{{}_{-}}^{*}'.format(CFG['model_arch'], i, epoch)))

    with torch.no_grad():
        for _ in range(CFG['tta']):
            val_preds += [CFG['weights'][i]/sum(CFG['weights'])/CFG['tta']*inference_one_epoch(model, val_loader, device)]
            tst_preds += [CFG['weights'][i]/sum(CFG['weights'])/CFG['tta']*inference_one_epoch(model, tst_loader, device)]

val_preds = np.mean(val_preds, axis=0)

tst_preds = np.mean(tst_preds, axis=0)
```



@明天去见你:

请问为什么我的训练速度会这么慢,是没有使用GPU吗,本机GPU应该怎么使用

```
C:\ProgramData\Anaconda3\lib\site-packages\torch\cuda\amp\grad_scaler.py:115: UserWar ning: torch.cuda.amp.GradScaler is enabled, but CUDA is not available. Disabling. warnings.warn("torch.cuda.amp.GradScaler is enabled, but CUDA is not available. Disabling.")

0%|
0/8559 [00:00<?, ?it/s]C:\ProgramData\Anaconda3\lib\site-packages\torch\cuda\amp\au tocast_mode.py:114: UserWarning: torch.cuda.amp.autocast only affects CUDA ops, but CUDA is not available. Disabling.

warnings.warn("torch.cuda.amp.autocast only affects CUDA ops, but CUDA is not available. Disabling.")
epoch 0 loss: 1.0118: 6%|
```



@马建华:

- 1. 老师能不能具体再说一下在baseline哪个地方修改fmix, cutmix 这些数据增强?
- 2. 打卡作业二中提到的"修改训练方法", 没太懂具体要怎么修改?
- 3. NNI是不是只能用于.py脚本文件提交任务,不能用于jupyter notebook?
- 4. 第一次课和第二次课给的baseline中train_one_epoch函数有不同。with autocast(): 包含的命令行不同, 会导致不同的输出结果吗? (如图所示)



@马建华:

```
, image_labels) in pbar: # 遍历每个 batch
.to(device).float()
s = image labels.to(device).long()
st(): # 开启自动混精度
reds = model(imgs) # 前向传播, 计算预测值
loss_fn(image_preds, image_labels) # 计算 loss
e(loss).backward() # 对 loss scale, scale梯度
使用指数平均
loss is None:
loss = loss.item()
_loss = running_loss * .99 + loss.item() * .01
1) % accum_iter == 0) or ((step + 1) == len(train_loader)):
step(
imizer) # unscale 梯度,如果梯度没有 overflow,使用 opt 更新梯度,否则不更新
update() # 等着下次 scale 梯度
er.zero_grad() # 梯度清空
duler is not None and schd_batch_update: # 学习率调整策略
eduler.step()
= f'epoch {epoch} loss: {running_loss:.4f}'
scription(description)
```

```
for step. (imas, image labels) in phar:
   imas = imas.to(device).float()
   image_labels = image_labels.to(device).long()
   with autocast():
       image preds = model(imgs) # output = model(input)
       loss = loss fn(image preds, image labels)
       scaler.scale(loss).backward()
       if running loss is None:
           running loss = loss.item()
           running_loss = running_loss * .99 * loss.item() * .01
       if ((step + 1) % CFG['accum_iter'] == 0) or ((step + 1) == len(train_loader)):
           # may unscale_ here if desired (e.g., to allow clipping unscaled gradients)
           scaler.step(optimizer)
           scaler.update()
           optimizer.zero_grad()
           if scheduler is not None and schd_batch_update:
               scheduler.step()
       if ((step + 1) % CFG['verbose_step'] == 0) or ((step + 1) == len(train_loader)):
           description = f'epoch {epoch} loss: {running_loss:.4f}'
           pbar.set_description(description)
if scheduler is not None and not schd_batch_update:
   scheduler.step()
```



@正心, 正己:

1: 先不考虑网络更改网络后要重新训练,请问代码上如何实现从网络中抽取某一层的特征图输出,接入到一个模块;

举例而言,比如代码上如何实现,把SiLu()的输出特征图,接入到RFB模块中输出后再接入到classifier中,我直接如图将RFB_small()这个类写进去,Pycharm中可以跑通,但是kaggle kernel上出错;

这是RFB模块---:

https://github.com/pprp/SimpleCVReproduction/blob/master/Plug-and-play%20module/ReceptiveFieldModule.py

这是我改的方法: https://www.kaggle.com/chuyunxinlan/cassava-rfb-train 请老师您帮忙看一下, 代码上我该如何实现将网络中某一层的特征输出, 接入到RFB模块这个类中, 然后再经过他输出; 谢谢;



@正心, 正己:

```
BatchNorm2d(2304, eps=1e-05, momentum=0.1, affine=True
): SiLU(inplace=True)
SqueezeExcite(
nv reduce): Conv2d(2304, 96, kernel size=(1, 1), stride=
t1): SiLU(inplace=True)
nv expand): Conv2d(96, 2304, kernel size=(1, 1), stride=
pwl): Conv2d(2304, 384, kernel size=(1, 1), stride=(1,
 BatchNorm2d(384, eps=1e-05, momentum=0.1, affine=True,
 Conv2d(384, 1536, kernel size=(1, 1), stride=(1, 1), b
nNorm2d(1536, eps=1e-05, promentum=0.1, affine=True, trac
J(inplace=True∰
l): SelectAdaptivePool2d (pool_type=avg, flatten=True)
): Linear(in features=1536, out features=1000, bias=True
```

```
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# 模型构建
class CassvaImgClassifier(nn.Module):
    def init (self, model arch, n class, pretrained=False):
        super(). init ()
        self.model = timm.create model(model arch.
pretrained=pretrained)
        # h features = self.model.classifier.in features
        self.model.global pool = nn.Sequential(
            BasicFRB small(in planes=1792, out planes=512,
stride=1,scale=0.1),
            SelectAdaptivePool2d(output size=1, flatten=True)
        self.model.classifier = nn.Linear(512, n class)
    def forward(self, x):
        x = self.model(x)
        return x
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```



@嘉Tarus:

1. 关于梯度累加的问题 图片1是 accum_iter=2, loss没有除以2 图片2是accum_iter=4 并且loss除以4 但是感觉最后结果都差不多呢? 但是loss差了4 倍啊

2. 看了下其他kaggle上的baseline 为什么他们batchsize 32 都可以, 我到16就不行了 还是用了梯度累加。都是在kaggle自带的GPU运行。我们的baseline哪里特别吃显存呢?

deepshare.net 深度之眼

@嘉Tarus:

```
epoch 2 loss: 0.3791: 100%| | 1070/1070 [18:25<00:00. 1.03s/it]
epoch 2 loss: 0.3358: 100%| | 268/268 [01:55<00:00, 2.32it/s]
             | 0/1070 [00:00<?, ?it/s]
validation multi-class accuracy = 0.8895
epoch 3 loss: 0.3484: 100%| | 1070/1070 [18:26<00:00, 1.03s/it]
epoch 3 loss: 0.3367: 100%| | 268/268 [01:54<00:00, 2.34it/s]
             | 0/1070 [00:00<?, ?it/s]
validation multi-class accuracy = 0.8909
epoch 4 loss: 0.3226: 100% | 1070/1070 [18:25<00:00, 1.03s/it]
epoch 4 loss: 0.3395: 100% | | 268/268 [01:55<00:00. 2.33it/s]
             | 0/1070 [00:00<?, ?it/s]
validation multi-class accuracy = 0.8909
epoch 5 loss: 0.3070: 100%| | 1070/1070 [18:29<00:00, 1.04s/it]
epoch 5 loss: 0.3439: 100% | | 268/268 [01:54<00:00, 2.34it/s]
             | 0/1070 [00:00<?, ?it/s]
validation multi-class accuracy = 0.8897
epoch 6 loss: 0.2956: 100% | 1070/1070 [18:26<00:00, 1.03s/it]
epoch 6 loss: 0.3427: 100%| | 268/268 [01:54<00:00, 2.33it/s]
             | 0/1070 [00:00<?, ?it/s]
```

```
validation multi-class accuracy = 0.8729
epoch 1 loss: 0.1063: 100%| | 1070/1070 [17:57<00:00, 1.01s/it]
epoch 1 loss: 0.3666: 100%| | 268/268 [01:59<00:00, 2.25it/s]
             | 0/1070 [00:00<?, ?it/s]
validation multi-class accuracy = 0.8776
epoch 2 loss: 0.0973: 100%[ | 1070/1070 [17:29<00:00, 1.02it/s]
epoch 2 loss: 0.3439: 100%| | 268/268 [01:52<00:00, 2.381t/s]
             | 0/1070 [00:00<?, ?it/s]
validation multi-class accuracy = 0.8843
epoch 3 loss: 0.0899: 100%| | 1070/1070 [17:30<00:00, 1.02it/s]
epoch 3 loss: 0.3340: 100%| | 268/268 [01:53<00:00. 2.37it/s]
             | 0/1070 [00:00<?, ?it/s]
 0%1
validation multi-class accuracy = 0.8951
epoch 4 loss: 0.0818: 100%| | 1070/1070 [17:30<00:00, 1.02it/s]
epoch 4 loss: 0.3364: 100%| | 268/268 [01:52<00:00, 2.38it/s]
             | 0/1070 [00:00<?, ?it/s]
validation multi-class accuracy = 0.8925
epoch 5 loss: 0.0786: 100%|
                                | 1070/1070 [17:30<00:00, 1.02it/s]
epoch 5 loss: 0.3386: 100%| | 268/268 [01:52<00:00, 2.38it/s]
validation multi-class accuracy = 0.8902
```



互动时间

结语-

感谢同学们参加今晚的直播答疑!

课下,请好好总结和回顾知识点