## National Tsing Hua University Fall 2023 11210IPT 553000 Deep Learning in Biomedical Optical Imaging Homework report

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這次我是用 transfer learning 的方法進行分類的。

## Swin\_B\_Weights.IMAGENET1K\_V1:

These weights reproduce closely the results of the paper using a similar training recipe. Also available as Swin\_B\_Weights.DEFAULT.

acc@1 (on ImageNet-1K)	83.582
acc@5 (on ImageNet-1K)	96.64
categories	tench, goldfish, great white shark, (997 omitted)
num_params	87768224
min_size	height=224, width=224
recipe	link
GFLOPS	15.43
File size	335.4 MB

## 圖 1:此次使用的 pre train model

除此之外,為了防止出現 overfitting 的情況,我又多加入了 dropout 層避免 overfitting。

圖 2:本次的 model ● Dropout 0.3

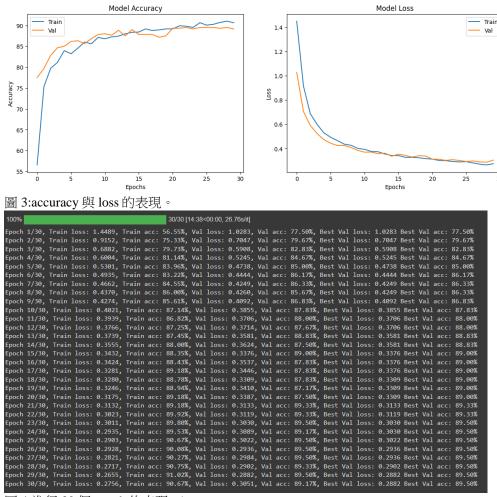
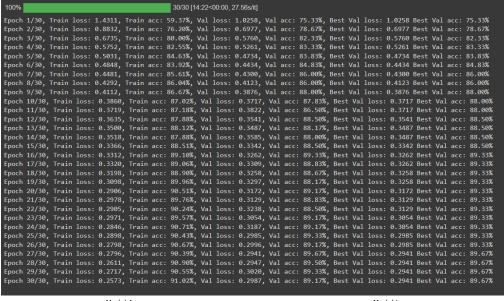
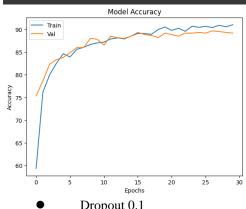
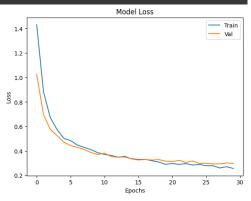


圖 4:進行 30 個 epoch 的表現。\

Dropout 0.2

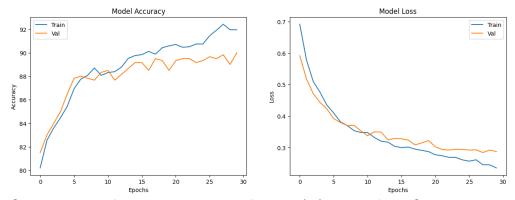






## Dropout 0.1

100% 30/30 [14:01<00:00, 28.64s/it] Epoch 1/30, Train loss: 0.6923, Train acc: 80.20%, Val loss: 0.5921, Val acc: 81.50%, Best Val loss: 0.5921 Best Val acc: 81.50% Epoch 2/30, Train loss: 0.5781, Train acc: 82.55%, Val loss: 0.5178, Val acc: 83.00%, Best Val loss: 0.5178 Best Val acc: 83.00% Epoch 3/30, Train loss: 0.5093, Train acc: 83.61%, Val loss: 0.4709, Val acc: 84.00%, Best Val loss: 0.4709 Best Val acc: 84.00% Epoch 4/30, Train loss: 0.4757, Train acc: 84.47%, Val loss: 0.4432, Val acc: 85.00%, Best Val loss: 0.4432 Best Val acc: 85.00% Epoch 5/30, Train loss: 0.4349, Train acc: 85.45%, Val loss: 0.4229, Val acc: 86.56%, Best Val loss: 0.4229 Best Val acc: 86.56% Epoch 5/30, Train loss: 0.4349, Train acc: 85.45%, Val loss: 0.4229, Val acc: 86.56%, Best Val loss: 0.4229 Best Val acc: 86.50%
Epoch 6/30, Train loss: 0.3019, Train acc: 86.94%, Val loss: 0.3919, Val acc: 87.83%, Best Val loss: 0.3919 Best Val acc: 87.83%
Epoch 7/30, Train loss: 0.3817, Train acc: 87.76%, Val loss: 0.3795, Val acc: 87.83%, Best Val loss: 0.3795 Best Val acc: 88.00%
Epoch 8/30, Train loss: 0.3698, Train acc: 88.08%, Val loss: 0.3696, Val acc: 87.67%, Best Val loss: 0.3696 Best Val acc: 88.00%
Epoch 9/30, Train loss: 0.35478, Train acc: 88.71%, Val loss: 0.3573, Val acc: 87.67%, Best Val loss: 0.3696 Best Val acc: 88.00%
Epoch 10/30, Train loss: 0.3478, Train acc: 88.08%, Val loss: 0.3573, Val acc: 88.33%, Best Val loss: 0.3592 Best Val acc: 88.33%
Epoch 11/30, Train loss: 0.3474, Train acc: 88.33%, Val loss: 0.3373, Val acc: 88.50%, Best Val loss: 0.3373 Best Val acc: 88.50%
Epoch 13/30, Train loss: 0.3317, Train acc: 88.39%, Val loss: 0.3496, Val acc: 88.576%, Best Val loss: 0.3373 Best Val acc: 88.50%
Epoch 13/30, Train loss: 0.33199, Train acc: 88.76%, Val loss: 0.3496, Val acc: 88.17%, Best Val loss: 0.3373 Best Val acc: 88.50%
Epoch 14/30, Train loss: 0.3317, Train acc: 89.53%, Val loss: 0.3243, Val acc: 88.17%, Best Val loss: 0.3243 Best Val acc: 88.50%
Epoch 14/30, Train loss: 0.3377, Train acc: 89.53%, Val loss: 0.3243, Val acc: 88.17%, Best Val loss: 0.3243 Best Val acc: 88.50%
Epoch 15/30, Train loss: 0.3377, Train acc: 89.75%, Val loss: 0.3243, Val acc: 89.17%, Best Val loss: 0.3243 Best Val acc: 89.17% Epoch 15/30, Train loss: 0.3037, Train acc: 89.76%, Val loss: 0.3284, Val acc: 89.17%, Best Val loss: 0.3243 Best Val acc: 89.17%
Epoch 15/30, Train loss: 0.3037, Train acc: 89.76%, Val loss: 0.3284, Val acc: 89.17%, Best Val loss: 0.3243 Best Val acc: 89.17%
Epoch 16/30, Train loss: 0.2996, Train acc: 89.84%, Val loss: 0.3276, Val acc: 89.17%, Best Val loss: 0.3243 Best Val acc: 89.17%
Epoch 18/30, Train loss: 0.3014, Train acc: 89.88%, Val loss: 0.325, Val acc: 89.50%, Best Val loss: 0.3082 Best Val acc: 89.50%
Epoch 18/30, Train loss: 0.2948, Train acc: 89.88%, Val loss: 0.3082, Val acc: 89.53%, Best Val loss: 0.3082 Best Val acc: 89.50% Epoch 20/30, Train loss: 0.2867, Epoch 21/30, Train loss: 0.2769, Train acc: 90.59%, Val loss: 0.3219, Val acc: 88.50%, Best Val loss: 0.3082 Best Val acc: 89.50% Train acc: 90.71%, Val loss: 0.3022, Val acc: 89.33%, Best Val loss: 0.3022 Best Val acc: 89.50% Epoch 22/39, Train loss: 0.2736, Train acc: 90.47%, Val loss: 0.2935, Val acc: 89.50%, Best Val loss: 0.2935 Best Val acc: 89.50% Epoch 23/30, Train loss: 0.2687, Train acc: 90.51%, Val loss: 0.2919, Val acc: 89.50%, Best Val loss: 0.2919 Best Val acc: 89.50% Epoch 24/30, Train loss: 0.2681, Train acc: 90.75%, Val loss: 0.2943, Val acc: 89.17%, Best Val loss: 0.2919 Best Val acc: 89.50% Epoch 25/30, Train loss: 0.2605, Train acc: 90.75%, Val loss: 0.2941, Val acc: 89.33%, Best Val loss: 0.2919 Best Val acc: 89.50% Epoch 26/30, Train loss: 0.2604, Train acc: 91.79%, Val loss: 0.2917, Val acc: 89.33%, Best Val loss: 0.2919 Best Val acc: 89.50% Epoch 26/30, Train loss: 0.2664, Train acc: 91.45%, Val loss: 0.2927, Val acc: 89.67%, Best Val loss: 0.2917 Best Val acc: 89.67% Epoch 27/30, Train loss: 0.2608, Train acc: 91.92%, Val loss: 0.2925, Val acc: 89.50%, Best Val loss: 0.2917 Best Val acc: 89.67% Epoch 28/30, Train loss: 0.2449, Train acc: 91.92%, Val loss: 0.2839, Val acc: 89.83%, Best Val loss: 0.2839 Best Val acc: 89.83% Epoch 29/30, Train loss: 0.2450, Train acc: 91.96%, Val loss: 0.2917, Val acc: 89.06%, Best Val loss: 0.2839 Best Val acc: 89.83% Epoch 30/30, Train loss: 0.2349, Train acc: 91.96%, Val loss: 0.2872, Val acc: 89.00%, Best Val loss: 0.2839 Best Val acc: 89.00%



最終選擇 dropout 為 0.1 進行 evaluate,因為從 loss 來看 dropout 為 0.1 最小,並且在進行 validation data 測試的時候分數也是最高的。

最終 test 的結果為 accuracy 為 92%

```
images = images.cuda()
    images = images / 255.

labels = labels.cuda()
    labels = labels.long()

outputs = model(images)

predicted = outputs.argmax(-1)
    test_correct += (predicted.float() == labels).sum().item()
    test_total += labels.size(0)

print(f'Test accuracy is {100. * test_correct / test_total}%')
Test accuracy is 92.16666666666666667%
```