Scene Recognition

前提資訊

- 決定開始訓練的權重 torch.manual_seed
- 進行一次iteration所訓練資料的數量 batch size
- 將所有訓練資料訓練次數 epoch
- · 各種對於梯度下降演算法的優化方法 optimizer
- 根據梯度下降的方向決定步長 learning rate

預設參數

Seed=123, epochs=10, batch_size=32, SGD, lr=0.01

```
16
      torch.manual seed (123)
      torch.backends.cudnn.deterministic = True
17
18
      torch.backends.cudnn.benchmark = False
19
40
        #print(DATASET ROOT)
41
        train set = IMAGE Dataset (Path (DATASET ROOT), data transform)
42
        data loader = DataLoader(dataset=train set, batch size=32, shuffle=True, num workers=1)
        #print(train set.num classes)
43
        model = VGG16(num_classes=train_set.num classes)
44
45
        model = model.cuda (CUDA DEVICES)
46
        model.train()
48
          best model params = copy.deepcopy(model.state dict())
49
          best acc = 0.0
50
          count = 100.0
51
          num epochs = 10
52
          criterion = nn.CrossEntropyLoss()
53
          optimizer = torch.optim.SGD(params=model.parameters(), lr=0.01, momentum=0.9)
54
```

程式碼新增部分

每10個將model存成一個檔案,並在用python test.py執行

```
86
              if (epoch+1) %10==0:
87
                  #torch.save(model, f'model.pth')
88
                  #tmp1, tmp2=test.test(f'model.pth')
                  #print(str(tmp1)+'\t'+str(tmp2))
89
90
                  torch.save(model, f'model-{count:.02f}-train.pth')
91
                  #result test loss.append(str(tmp1))
92
                  #result test acc.append(str(tmp2))
93
                  count=count+1
94
```

· 將訓練過程中得到的loss和accuracy分別 存在檔案裡

```
□if name == ' main ':
107
108
          train()
109
          train loss file=open('train loss file.txt', 'w')
110
          train acc file=open('train acc file.txt', 'w')
111
          #test loss file=open('test loss file.txt', 'w')
          #test acc file=open('test acc file.txt', 'w')
112
113
114
          for i in result train loss:
115
              train loss file.write(i)
116
              train loss file.write('\n')
          for i in result train acc:
117
118
              train acc file.write(i)
119
              train acc file.write('\n')
120
          #for i in result test loss:
121
              test loss file.write(i)
122
              test loss file.write('\n')
123
          #for i in result test acc:
124
              test acc file.write(i)
125
              test acc file.write('\n')
          train loss file.close()
126
          train acc file.close()
127
128
          #test loss file.close()
129
          #test acc file.close()
```

訓練過程與結果

• 第10個模型進行測試結果

```
(mylab) waynell16@gslave01:~/preset$ CU
Accuracy on the ALL test images: 85 %
Accuracy of street : 83 %
Accuracy of forest : 96 %
Accuracy of sea : 86 %
Accuracy of mountain : 79 %
Accuracy of glacier : 84 %
Accuracy of buildings : 82 %
test_loss: 0.4336245342095693
```

。 最佳訓練的模型測試結果

```
Accuracy on the ALL test images: 85 %
Accuracy of street : 83 %
Accuracy of forest : 96 %
Accuracy of sea : 86 %
Accuracy of mountain : 79 %
Accuracy of glacier : 84 %
Accuracy of buildings : 82 %
test_loss: 0.43362453015645347

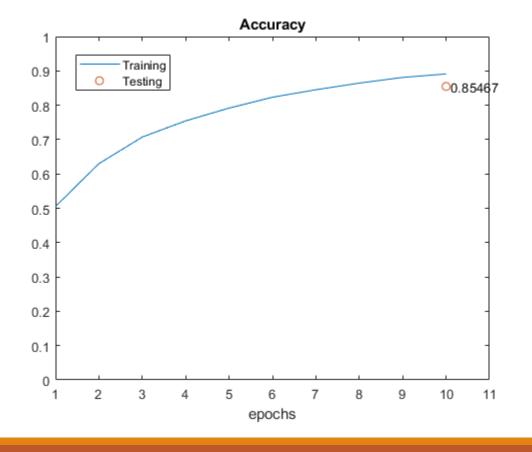
test_acc: 0.8546666666666666
```

。 訓練過程

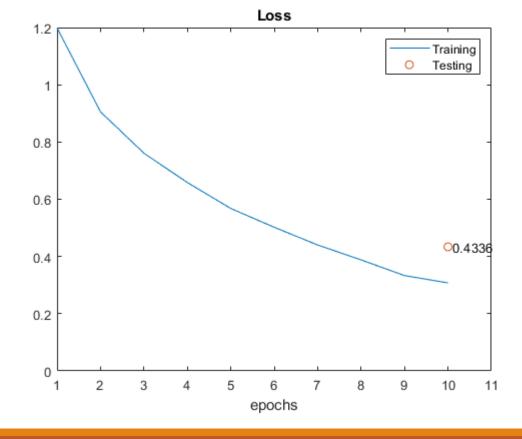
```
Epoch: 1/10
Training loss: 1.1987
                      accuracy: 0.5053
Epoch: 2/10
Training loss: 0.9055
                      accuracy: 0.6299
Epoch: 3/10
Training loss: 0.7609 accuracy: 0.7074
Epoch: 4/10
Training loss: 0.6584 accuracy: 0.7545
Epoch: 5/10
Training loss: 0.5678 accuracy: 0.7917
Epoch: 6/10
Training loss: 0.5021 accuracy: 0.8233
Epoch: 7/10
Training loss: 0.4401 accuracy: 0.8451
Epoch: 8/10
Training loss: 0.3880
                      accuracy: 0.8645
Epoch: 9/10
Training loss: 0.3332 accuracy: 0.8810
Epoch: 10/10
Training loss: 0.3077 accuracy: 0.8911
```

Training+Testing Accuracy/Loss matlab 繪圖

Training+Testing Accuracy



Training+Testing Loss



調整參數-考量因素seed

- 調整seed, 其他因素不變(batch size=32, epochs=10, SGD, Ir=0.01)
- ✓ 當seed值從123->0時, seed值=0的訓練最佳結果為90%, 而測試整體準確率為83%
- ✓ 當seed值從123->1024時, seed值=1024的訓練最佳結果為89%,而測試整體準確率為85%
- ▶ 結論:儘管seed值=O訓練準確率最好,但測試準確率卻不如預期
- 。 用最佳訓練模型測試seg_test結果

```
Training loss: 1.2017
                       accuracy: 0.5077
Epoch: 2/10
Training loss: 0.8726
                       accuracy: 0.6584
Training loss: 0.7347
                       accuracy: 0.7239
Epoch: 4/10
Training loss: 0.6311
                       accuracy: 0.7725
Epoch: 5/10
Training loss: 0.5516
                       accuracy: 0.8038
                       accuracy: 0.8279 test acc: 0.838
Training loss: 0.4786
Epoch: 7/10
Training loss: 0.4158
Epoch: 8/10
Training loss: 0.3797
                       accuracy: 0.8655
Training loss: 0.3534
                       accuracy: 0.8735
Epoch: 10/10
Training loss: 0.2845
                       accuracy: 0.9008
```

```
^[[AAccuracy on the ALL test images: 83 %
Accuracy of street : 83 %
Accuracy of forest : 96 %
Accuracy of sea : 76 %
Accuracy of mountain : 82 %
Accuracy of glacier : 79 %
Accuracy of buildings : 84 %
test_loss: 0.5311218338807424

test_acc: 0.838
```

• seed值=0

```
Accuracy on the ALL test images: 85 %
Accuracy of street : 91 %
Accuracy of forest : 96 %
Accuracy of sea : 92 %
Accuracy of mountain : 83 %
Accuracy of glacier : 69 %
Accuracy of buildings : 79 %
test_loss: 0.42865071817239125

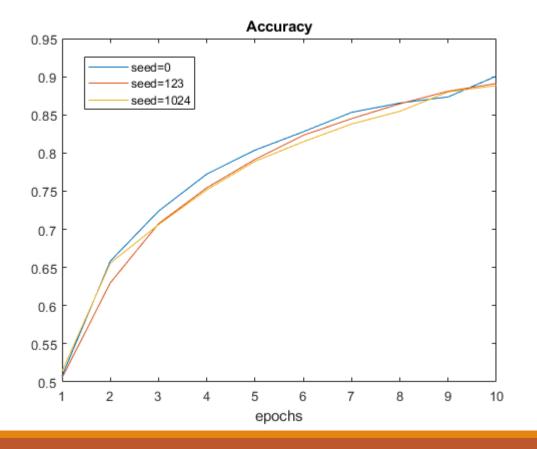
test_acc: 0.85133333333333334
```

seed值=1024

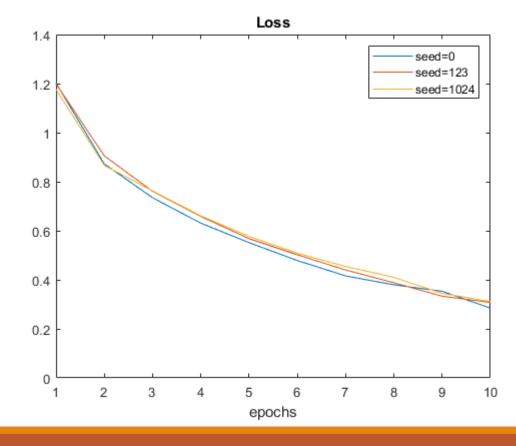
```
Training loss: 1.1761
                        accuracy: 0.5137
Epoch: 2/10
Training loss: 0.8666
                        accuracy: 0.6557
Training loss: 0.7624
                        accuracy: 0.7062
Epoch: 4/10
Training loss: 0.6609
                        accuracy: 0.7518
poch: 5/10
Training loss: 0.5768
                        accuracy: 0.7892
Fraining loss: 0.5085
                        accuracy: 0.8148
Epoch: 7/10
Training loss: 0.4543
                        accuracy: 0.8380
Epoch: 8/10
Training loss: 0.4100
                        accuracy: 0.8547
Training loss: 0.3448
                        accuracy: 0.8801
Epoch: 10/10
Training loss: 0.3114 accuracy: 0.8884
```

Training Accuracy & loss

Training Accuracy



Training loss



調整參數-考量因素batch_size

- 。 調整batch_size, 其他因素不變(Seed=123, epochs=10, SGD, Ir=0.01)
- ✓ 當batch_size從32->64時,batch_size=64的訓練最佳結果為88%,而測試整體準確率為82%
- ✓ 當batch_size從32->16時, batch_size=16的訓練最佳結果為87%,而測試整體準確率為83%
- ▶ 結論:使用batch size=32測試結果會比較好
- · 用最佳訓練模型測試seg_test結果

```
(mylab) waynell16@gslave01:~/hwl_batch
Accuracy on the ALL test images: 83 %
Accuracy of street : 87 %
Accuracy of forest : 91 %
Accuracy of sea : 80 %
Accuracy of mountain : 76 %
Accuracy of glacier : 81 %
Accuracy of buildings : 87 %
test_loss: 0.4472797746658325
```

```
• Batch_size=16
```

```
(mylab) waynell16@gslave01:~/hwl_v4$ C
Accuracy on the ALL test images: 85 %
Accuracy of street : 83 %
Accuracy of forest : 96 %
Accuracy of sea : 86 %
Accuracy of mountain : 79 %
Accuracy of glacier : 84 %
Accuracy of buildings : 82 %
test_loss: 0.4336245404879252
```

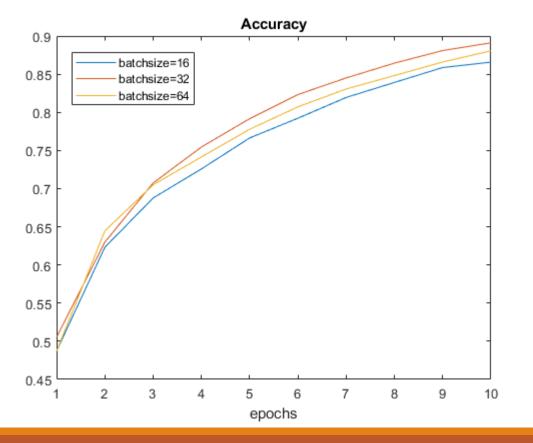
• Batch size=32

```
(mylab) waynell16@gslave01:~/hw1_v2$ py
Accuracy on the ALL test images: 82 %
Accuracy of street : 85 %
Accuracy of forest : 90 %
Accuracy of sea : 96 %
Accuracy of mountain : 62 %
Accuracy of glacier : 81 %
Accuracy of buildings : 76 %
test_loss: 0.5457783784866334
test_acc: 0.82033333333333334
```

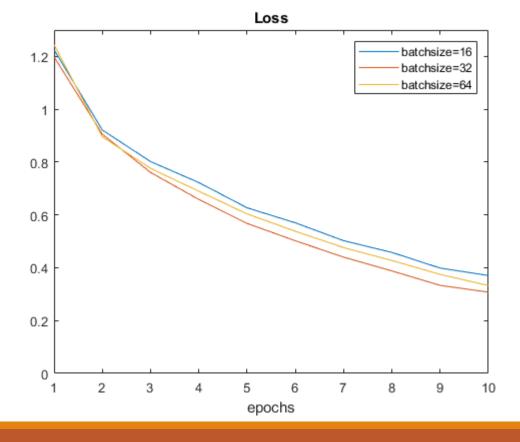
• Batch size=64

Training Accuracy & loss

Training Accuracy



Training loss

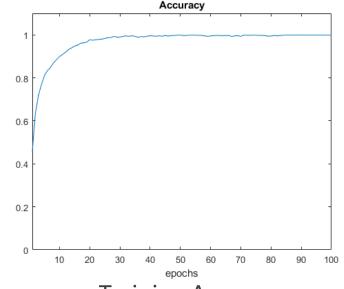


調整參數-考量因素epochs

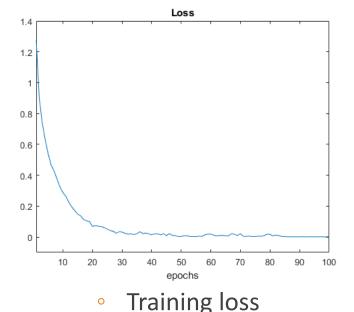
- 。 調整epoch,其他因素不變(Seed=123, batch_size=32, SGD, lr=0.01)
- ✓ 將epoch=10改為epoch=100,觀察訓練次數增加時 Accuracy曲線、Loss曲線狀況
- ✓ 結論: 當epochs到達25的時候,訓練準確率已到99%,且Loss也愈接近0

Accuracy on the ALL test images: 85 %
Accuracy of sea : 89 %
Accuracy of buildings : 83 %
Accuracy of forest : 96 %
Accuracy of street : 88 %
Accuracy of glacier : 77 %
Accuracy of mountain : 81 %

Best testing result



Training Accuracy



調整參數-考量因素learning rate

- 調整Ir,其他因素不變(Seed=123, batch_size=32, epoch=10, SGD)
- ✓ 當Ir從0.01->0.1,發現train_loss一直處於nan, Ir=0.1的訓練最佳結果為17%,而測試整體準確率為16%
- ✓ 當Ir從0.01->0.001, Ir=0.001的訓練最佳結果為85%, 而測試整體準確率為85%
- ✓ 結論: lr=0.01在訓練模型時雖然準確率及loss比lr=0.001好,但測試時相差無幾,而lr=0.01 loss的部分相對較好
- 。 用最佳訓練模型測試seg_test結果

```
Training loss: nan
                        accuracy: 0.1693
Epoch: 2/10
Training loss: nan
                        accuracy: 0.1697
Epoch: 3/10
Training loss: nan
                        accuracy: 0.1697
Epoch: 4/10
Training loss: nan
                        accuracy: 0.1697
Epoch: 5/10
Training loss: nan
                        accuracy: 0.1697
Epoch: 6/10
                        accuracy: 0.1697
Training loss: nan
Epoch: 7/10
Training loss: nan
                        accuracy: 0.1697
Epoch: 8/10
Training loss: nan
                        accuracy: 0.1697
Epoch: 9/10
Training loss: nan
                        accuracy: 0.1697
Epoch: 10/10
Training loss: nan
                        accuracy: 0.1697
```

```
(mylab) waynellI6@gslave01:~/hwl_lr$ CUI
Accuracy on the ALL test images: 16 %
Accuracy of street : 100 %
Accuracy of forest : 0 %
Accuracy of sea : 0 %
Accuracy of mountain : 0 %
Accuracy of glacier : 0 %
Accuracy of buildings : 0 %
test_loss: nan
```

Learning rate=0.1

```
Accuracy on the ALL test images: 85 %
Accuracy of street : 91 %
Accuracy of forest : 95 %
Accuracy of sea : 91 %
Accuracy of mountain : 79 %
Accuracy of glacier : 77 %
Accuracy of buildings : 74 %
test_loss: 0.4236939247449239

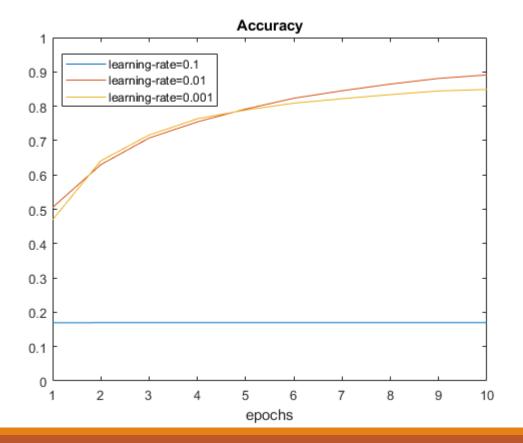
test_acc: 0.8523333333333334
```

• Learning rate=0.001

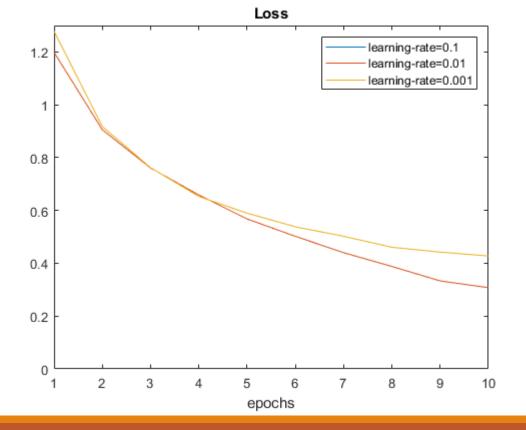
```
Training loss: 1.2785
                        accuracy: 0.4692
Training loss: 0.9177
                        accuracy: 0.6412
Epoch: 3/10
Training loss: 0.7619
                        accuracy: 0.7162
Epoch: 4/10
Training loss: 0.6530
                        accuracy: 0.7635
Epoch: 5/10
Training loss: 0.5897
                        accuracy: 0.7889
Training loss: 0.5382
                        accuracy: 0.8088
Epoch: 7/10
Training loss: 0.5022
                        accuracy: 0.8219
Epoch: 8/10
Training loss: 0.4607
                        accuracy: 0.8338
Training loss: 0.4423
                        accuracy: 0.8446
Epoch: 10/10
Training loss: 0.4277 accuracy: 0.8487
```

Training Accuracy & loss

Training Accuracy



Training loss(learning rate=0.1的loss都是NAN)



調整參數-考量因素optimizer

調整optimizer,其他因素不變(Seed=123, batch_size=32, epoch=10)

```
optimizer = torch.optim.Adam(params=model.parameters(), lr=0.001, betas=(0.9,0.999), eps=le-08, weight_decay=0);
```

- ✓ 當optimizer改為Adam, Adam的訓練最佳結果為18%,而測試準確率為17%
- ✓ 結論:雖然訓練時間較SGD短,但準確率卻不如SGD
- · 用最佳訓練模型測試seg_test結果

```
Accuracy on the ALL test images: 17 %
Accuracy of street : 0 %
Accuracy of forest : 0 %
Accuracy of sea : 0 %
Accuracy of mountain : 100 %
Accuracy of glacier : 0 %
Accuracy of buildings : 0 %
test_loss: 1.7899835192362468
```

```
Training loss: 2.6425
                        accuracy: 0.1712
Training loss: 1.7912
                        accuracy: 0.1791
Epoch: 3/10
Training loss: 1.7912
                        accuracy: 0.1751
Epoch: 4/10
Training loss: 1.7911
                        accuracy: 0.1786
Epoch: 5/10
Training loss: 1.7910
                        accuracy: 0.1790
Epoch: 6/10
Training loss: 1.7911
                        accuracy: 0.1786
Training loss: 1.7910
                        accuracy: 0.1770
Epoch: 8/10
Training loss: 1.7909
                        accuracy: 0.1789
Training loss: 1.7910
                        accuracy: 0.1790
Epoch: 10/10
 raining loss: 1.7910
                        accuracy: 0.1790
```

統整調整參數後結果

- 。 最佳結果參數設定
 - seed=1024, batch_size=32, epoch=35, SGD, learning_rate=0.01
 - learning rate每30個epoch會乘上0.1做為調整

```
torch.manual seed (1024)
        torch.backends.cudnn.deterministic = True
17
18
        torch.backends.cudnn.benchmark = False
45
        #print(DATASET ROOT)
46
        train set = IMAGE Dataset(Path(DATASET ROOT), data transform)
47
       data loader = DataLoader(dataset=train set, batch size=32, shuffle=True, num workers=1)
48
        #print(train set.num classes)
54
         best acc = 0.0
55
         count = 100.0
56
         num epochs = 35
57
         criterion = nn.CrossEntropyLoss()
58
         optimizer = torch.optim.SGD(params=model.parameters(), lr=0.01, momentum=0.9)
```

```
34
     □def adjust learning rate(optimizer, epoch):
35
           lr=0.01*(0.1**(epoch//30))
 36
           for param group in optimizer.param groups:
 37
               param group['lr']=lr
 38
60
        for epoch in range (num epochs):
61
            print(f'Epoch: {epoch + 1}/{num epochs}')
62
            print('-' * len(f'Epoch: {epoch + 1}/{num epochs}'))
63
            adjust learning rate (optimizer, num epochs);
```

最佳結果顯示&過程

。 最佳結果顯示:86%>預設參數85%

Accuracy on the ALL test images: 86 % Accuracy of street : 91 % Accuracy of forest : 95 % Accuracy of sea : 89 % Accuracy of mountain : 83 % Accuracy of glacier : 84 % Accuracy of buildings : 72 % test_loss: 0.7284157524108886 test acc: 0.8606666666666667

• 執行過程

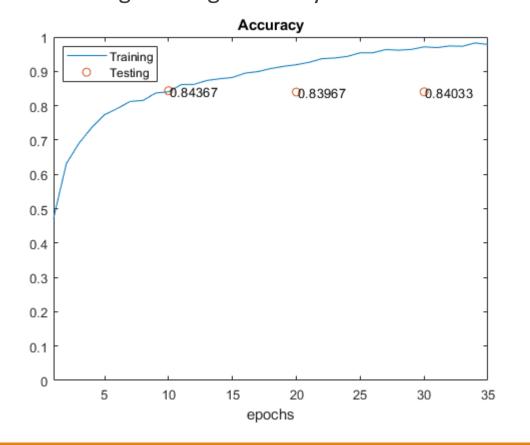
Training loss:	1.2857	accuracy:	0.4722
Epoch: 2/35			
Training loss:	0.9374	accuracy:	0.6312
Epoch: 3/35			
Training loss:	0.8162	accuracy:	0.6913
Epoch: 4/35			
Training loss:	0.7131	accuracy:	0.7370
Epoch: 5/35			
Training loss:	0.6247	accuracy:	0.7741
Epoch: 6/35			
Training loss:	0.5644	accuracy:	0.7922
Epoch: 7/35			
Training loss:	0.5246	accuracy:	0.8123
Epoch: 8/35			
Training loss:	0.4986	accuracy:	0.8154
Epoch: 9/35			
Training loss:	0.4610	accuracy:	0.8370
Epoch: 10/35			
Training loss:	0.4401	accuracy:	0.8405
Epoch: 11/35			
Training loss:	0.3985	accuracy:	0.8616
Epoch: 12/35			
Training loss:	0.3836	accuracy:	0.8621
	The state of the s		

Epoch: 13/35			
Training loss:	0.3558	accuracy:	0.8735
Epoch: 14/35			
Training loss:	0.3338	accuracy:	0.8784
-			
Epoch: 15/35			
Training loss:	0.3296	accuracy:	0.8822
Epoch: 16/35			
Training loss:	0.2998	accuracy:	0.8949
Epoch: 17/35			
Training loss:	0.2771	accuracy:	0.8993
Epoch: 18/35			
Training loss:	0.2591	accuracy:	0.9082
Epoch: 19/35			
Training loss:	0.2389	accuracy:	0.9147
Epoch: 20/35			
Training loss:	A 2268	accuracy:	A 9194
	0.2200	accuracy.	0.3134
Epoch: 21/35			
Training loss:	0.2008	accuracy:	0.9263
Epoch: 22/35			
Training loss:	0.1766	accuracy:	0.9369
Epoch: 23/35			
Training loss.	0 1650	20011520111	0.0000
Training loss:	0.1659	accuracy:	0.9390
Epoch: 24/35			
Training loss:	0.1603	accuracy:	0.9436
Epoch: 25/35			
Training loss:	0 1204	accuracy:	0.0542
maining toss:	0.1294	accuracy:	0.9342

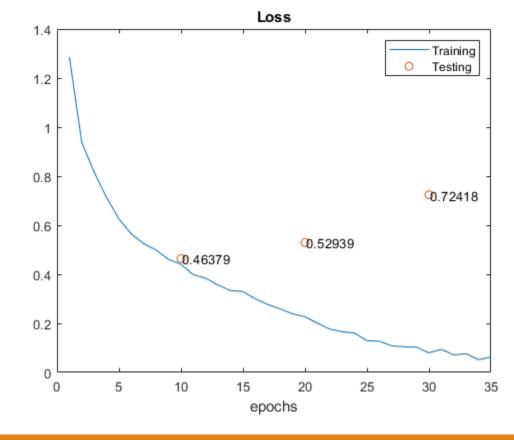
Epoch: 26/35			
Training loss:	0.1271	accuracy:	0.9541
Epoch: 27/35			15 -0.00 000 0000
Training loss:	0.1081	accuracy:	0.9635
Epoch: 28/35			
Training loss:	0.1047	accuracy:	0.9617
Epoch: 29/35			
Training loss:	0.1027	accuracy:	0.9639
Epoch: 30/35			
Training loss:	0.0798	accuracy:	0.9714
Epoch: 31/35			
Training loss:	0.0941	accuracy:	0.9692
Epoch: 32/35			
Training loss:	0.0714	accuracy:	0.9740
Epoch: 33/35			
Training loss:	0.0762	accuracy:	0.9731
Epoch: 34/35			
Training loss:	0.0518	accuracy:	0.9828
Epoch: 35/35			
Training loss:	0.0626	accuracy:	0.9782

Training+Testing Accuracy/Loss matlab 繪圖

Training+Testing Accuracy



Training+Testing Loss



結論Discussion

- 1. 根據預設參數以及調整參數的數據顯示,對於森林的分類有較高的準確率
- 2. 根據自己調整參數數據顯示,每10次epoch對現在model做一次testing的loss值明顯上升
- 3.根據數據顯示,儘管自己調整參數training accuracy達到97%,而預設參數的training accuracy只有89%,但是兩者的testing accuracy卻沒有明顯的差異(86%,85%)
- 4.Batch size越大,訓練時間越短
- 5.當訓練次數越多,training accuracy就越高

問題與困難Problem & difficulties

- 1. 使用工學院的深度學習伺服器訓練模型,原本電腦都要一直開著,而且有時候還會中斷連線,而且那時候epoch是50,所以每一次都需要重跑一次,後來SA提供了nohup的方法
- 2.我一開始以為training accuracy越高,testing accuracy應該也要越高才對,但跑完模型並testing後發現不是這樣
- 3.對於optimizer的選擇,對於Adam optimizer結合了AdaGrad依照梯度去調整learning rate的特性,以及SGD中momentum對梯度方向做梯度速度調整,理論上應該會比這兩個optimizer結果還要來的好,但是經由實驗數據顯示卻不如預期
- 4. 因為當時我想要測試最佳model的準確率,但是深度學習伺服器都有人使用,所以我將model從伺服器下載下來並使用專題實驗室附有GPU的電腦測試最佳model準確率,但是結果很奇怪都是30%,最後我重新回到伺服器測試才比較正常,準確率為85%

参考文獻Reference

• [機器學習ML NOTE]SGD,Momentum,AdaGrad,Adam optimizer

https://medium.com/%E9%9B%9E%E9%9B%9E%E8%88%87%E5%85%94%E5%85%94%E7%9A%84%E5%B7%A5%E7%A8%8B%E4%B8%96%E7%95%8C/%E6%A9%9F%E5%99%A8%E5%AD%B8%E7%BF%92ml-note-sgd-momentum-adagrad-adam-optimizer-f20568c968db

https://blog.csdn.net/aliceyangxi1987/article/details/73210204

Batch_size

https://www.leiphone.com/news/201710/RIIIL7LdIIT1Mvm8.html

https://www.zhihu.com/question/32673260

Deep learning

https://ithelp.ithome.com.tw/articles/10189086