

Mini Project 3

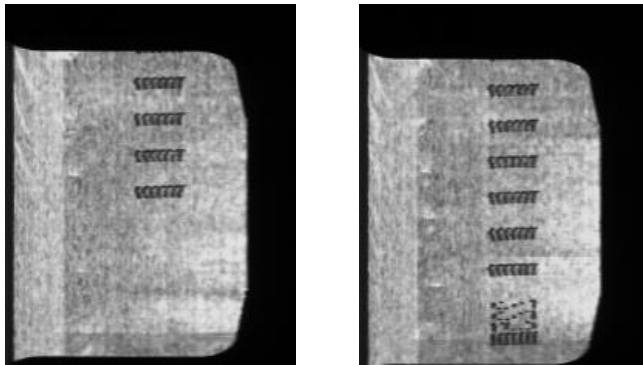
ResNet을 이용하여 현업에서 얻어진 이미지로 분류 학습하기

2021254019 김지현

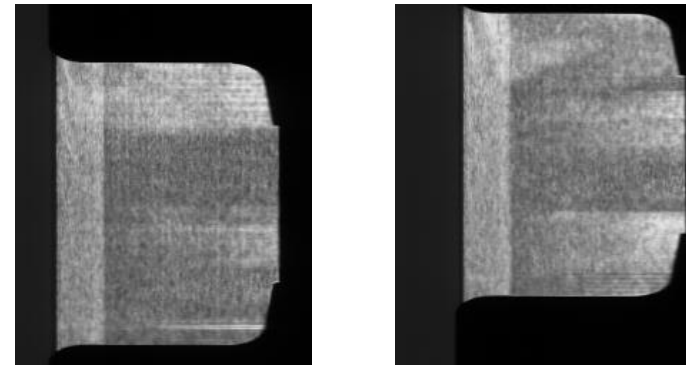
사용한 이미지 데이터

1. 이미지 데이터

학습에 사용된 이미지는 2차 전지 배터리의 전극이미지로서 불량 셀에는 NG 마킹을 한다.



<NG마킹>



<양품>

실제 코드(Colab)



```
1 %matplotlib inline
2 from __future__ import print_function, division
3
4 import torch
5 import torch.nn as nn
6 import torch.optim as optim
7 from torch.optim import lr_scheduler
8 import numpy as np
9 import torchvision
10 from torchvision import datasets, models, transforms
11 import matplotlib.pyplot as plt
12 import time
13 import os
14 import copy
15 plt.ion()
16
17 from google.colab import drive
18 drive.mount('/content/drive')
19
20 data_transforms = {
21     'train': transforms.Compose([
22         transforms.RandomResizedCrop(224),
23         transforms.RandomHorizontalFlip( ),
24         transforms.ToTensor(),
25         transforms.Normalize([0.485, 0.456, 0.406], [0.229, 0.224, 0.225])
26     ]),
27     'val': transforms.Compose([
28         transforms.Resize(256),
29         transforms.CenterCrop(224),
30         transforms.ToTensor(),
31         transforms.Normalize([0.485, 0.456, 0.406], [0.229, 0.224, 0.225])
32     ]),
33 }
34
35 }
```

```

37 data_dir = '/content/drive/My Drive/hymenoptera_data'
38 image_datasets = {x : datasets.ImageFolder(os.path.join(data_dir, x), data_transforms[x]) for x in ['train', 'val']}
39 dataloaders = {x : torch.utils.data.DataLoader(image_datasets[x], batch_size=4, shuffle=True, num_workers=4) for x in ['train', 'val']}
40 dataset_sizes = {x : len(image_datasets[x]) for x in ['train', 'val']}
41 class_names = image_datasets['train'].classes
42 device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
43
44 def imshow(inp, title=None):
45     inp = inp.numpy().transpose((1,2,0))
46     mean = np.array([0.485, 0.456, 0.406])
47     std = np.array([0.229, 0.224, 0.225])
48     inp = std * inp + mean
49     inp = np.clip(inp, 0, 1)
50     plt.imshow(inp)
51     if title is not None :
52         plt.title(title)
53     plt.pause(0.001)
54
55 inputs, classes = next(iter(dataloaders['train']))
56 out = torchvision.utils.make_grid(inputs)
57
58 imshow(out, title = [class_names[x] for x in classes])
59
60 def train_model(model, criterion, optimizer, scheduler, num_epochs=25) :
61     since = time.time()
62     best_model_wts = copy.deepcopy(model.state_dict())
63     best_acc = 0.0
64
65     for epoch in range(num_epochs) :
66         print('Epoch {}/{}'.format(epoch, num_epochs - 1))
67         print('-'*10)
68         for phase in ['train', 'val'] :
69             if phase == 'train' :
70                 scheduler.step()
71                 model.train()
72             else:
73                 model.eval()
74                 running_loss = 0.0
75                 running_corrects = 0
76                 for inputs, labels in dataloaders[phase] :
77                     inputs = inputs.to(device)

```

→ 학습에 사용할 이미지데이터를 가져오는 경로

```

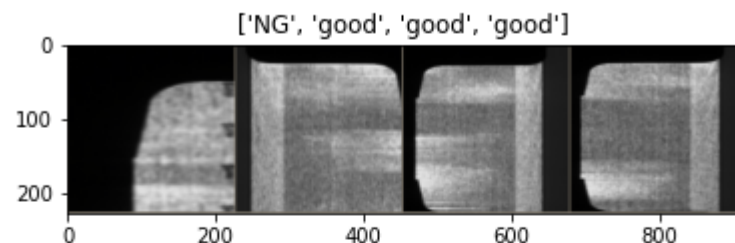
78     labels = labels.to(device)
79     optimizer.zero_grad()
80     with torch.set_grad_enabled(phase == 'train') :
81         outputs = model(inputs)
82         _, preds = torch.max(outputs, 1)
83         loss = criterion(outputs, labels)
84         if phase == 'train' :
85             loss.backward()
86             optimizer.step()
87         running_loss += loss.item() * inputs.size(0)
88         running_corrects += torch.sum(preds == labels.data)
89     epoch_loss = running_loss / dataset_sizes[phase]
90     epoch_acc = running_corrects.double() / dataset_sizes[phase]
91     print('{} Loss: {:.4f} Acc: {:.4f}'.format(phase, epoch_loss, epoch_acc))
92     if phase == 'val' and epoch_acc > best_acc :
93         best_acc = epoch_acc
94         best_model_wts = copy.deepcopy(model.state_dict())
95     print()
96     time_elapsed = time.time() - since
97     print('Training complete in {:.0f}m {:.0f}s'.format(time_elapsed // 60, time_elapsed % 60))
98     print('Best val Acc: {:.4f}'.format(best_acc))
99     model.load_state_dict(best_model_wts)
100     return model
101
102 def visualize_model(model, num_images = 6) :
103     was_training = model.training
104     model.eval()
105     images_so_far = 0
106     fig = plt.figure()
107
108     with torch.no_grad() :
109         for i, (inputs, labels) in enumerate(dataloaders['val']) :
110             inputs = inputs.to(device)
111             labels = labels.to(device)
112
113             outputs = model(inputs)
114             _, preds = torch.max(outputs, 1)

```

```
116     for j in range(inputs.size()[0]):
117         images_so_far += 1
118         ax = plt.subplot(num_images//2, 2, images_so_far)
119         ax.axis('off')
120         ax.set_title('predicted: {}'.format(class_names[preds[j]]))
121         imshow(inputs.cpu().data[j])
122
123     if images_so_far == num_images :
124         model.train(mode = was_training)
125         return
126     model.train(mode = was_training)
127
128 model_ft = models.resnet18(pretrained = True)
129 num_fts = model_ft.fc.in_features
130 model_ft.fc = nn.Linear(num_fts, 2)
131
132 model_ft = model_ft.to(device)
133 criterion = nn.CrossEntropyLoss()
134
135 optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)
136 exp_lr_scheduler = lr_scheduler.StepLR(optimizer_ft, step_size=7, gamma=0.1)
137 model_ft = train_model(model_ft, criterion, optimizer_ft, exp_lr_scheduler, num_epochs=25)
138
139 visualize_model(model_ft)
```

학습 결과

Drive already mounted at /content/drive; to attempt to forcibly
/usr/local/lib/python3.7/dist-packages/torch/utils/data/data
cpuset_checked))



Epoch 0/24

/usr/local/lib/python3.7/dist-packages/torch/optim/lr_scheduler
"<https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate>
train Loss: 0.4922 Acc: 0.7746
val Loss: 0.0495 Acc: 0.9833

Epoch 1/24

train Loss: 0.5255 Acc: 0.8279
val Loss: 0.0095 Acc: 1.0000

Epoch 2/24

train Loss: 0.5791 Acc: 0.8607
val Loss: 0.0155 Acc: 1.0000

Epoch 3/24

train Loss: 0.2617 Acc: 0.9344
val Loss: 0.1846 Acc: 0.9333

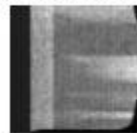
Training complete in 32m 9s

Best val Acc: 1.000000

predicted: NG



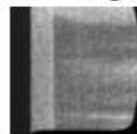
predicted: good



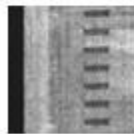
predicted: NG



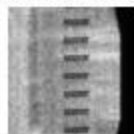
predicted: good



predicted: NG



predicted: NG



학습 시간 : 약 32분

정확도 : 1(100% 정확하게 분류)