

## ▼ Deep Learning

```
1 import pandas as pd
2 import os
3 import numpy as np
4 import matplotlib
5 import matplotlib.pyplot as plt
6 import torch
7 import torch.nn as nn
8 import torch.optim as optim

1 device = 'cuda' if torch.cuda.is_available() else 'cpu'
```

## ▼ Data import

```
1 traindata_url = 'https://bitbucket.org/hyuk125/lg_dic/raw/889649d1bc273bf53967'
2 testdata_url = 'https://bitbucket.org/hyuk125/lg_dic/raw/889649d1bc273bf53967c'
3 train_data = pd.read_csv(traindata_url)
4 test_data = pd.read_csv(testdata_url)

1 from sklearn.preprocessing import LabelEncoder
2
3 le = LabelEncoder()
4
5 le.fit(train_data.label == 5)
6
7 train_data.label = le.transform(train_data.label == 5)
8 test_data.label = le.transform(test_data.label == 5)
```

## ▼ Deep learning - classification 모델

### ▼ Pytorch 모델에 입력하기 위한 데이터 변환

```
1 train_data = torch.from_numpy(train_data.values).float()
2 test_data = torch.from_numpy(test_data.values).float()

1 BATCH_SIZE = 15
2 epochs = 2
3 learning_rate = 0.001
4 drop_prob = 0.5
```

```

1
2 data_loader = torch.utils.data.DataLoader(train_data,
3                                           batch_size=BATCH_SIZE,
4                                           shuffle=True,
5                                           num_workers=0)

```

## ▼ Deep learning 모델 정의 with Xavier initializer & dropout

```

1 class DNNModel(nn.Module):
2     def __init__(self):
3         super(DNNModel, self).__init__()
4         self.layer1 = nn.Linear(28 * 28, 300)
5         self.layer2 = nn.Linear(300, 300)
6         self.layer3 = nn.Linear(300, 2)
7         self.relu = nn.ReLU()
8
9         self.dropout = nn.Dropout(p=drop_prob)
10
11         nn.init.xavier_uniform_(self.layer1.weight)
12         nn.init.xavier_uniform_(self.layer2.weight)
13         nn.init.xavier_uniform_(self.layer3.weight)
14
15
16     def forward(self, x):
17
18         layers = nn.Sequential(self.layer1, self.relu, self.dropout,
19                               self.layer2, self.relu, self.dropout,
20                               self.layer3).to(device)
21         out = layers(x)
22         return out
23
24 model = DNNModel()
25 model

```

```

DNNModel(
  (layer1): Linear(in_features=784, out_features=300, bias=True)
  (layer2): Linear(in_features=300, out_features=300, bias=True)
  (layer3): Linear(in_features=300, out_features=2, bias=True)
  (relu): ReLU()
  (dropout): Dropout(p=0.5, inplace=False)
)

```

## ▼ 학습 시작

```

1 criterion = nn.CrossEntropyLoss().to(device)
2 optimizer = optim.Adam(model.parameters(), lr = learning_rate)

1 for epoch in range(epochs):

```

```

2     running_cost = 0.0
3
4     for step, (batch_data) in enumerate(data_loader):
5         batch_x = batch_data[:, 1:].view(-1, 28*28).to(device)
6         batch_y = batch_data[:, 0].to(device).long()
7
8         optimizer.zero_grad()
9
10        outputs = model(batch_x)
11        cost = criterion(outputs, batch_y)
12
13        cost.backward()
14        optimizer.step()
15
16        running_cost += cost.item()
17        if step % 200 == 199:
18            print('[%d, %5d] cost: %.3f' % (epoch + 1, step + 1, running_cost
19            running_cost = 0.0
20

```

```

☞ [1, 200] cost: 13.980
   [1, 400] cost: 4.697
   [1, 600] cost: 2.456
   [1, 800] cost: 2.250
   [1, 1000] cost: 1.086
   [1, 1200] cost: 1.153
   [1, 1400] cost: 0.812
   [1, 1600] cost: 0.751
   [1, 1800] cost: 0.519
   [1, 2000] cost: 0.516
   [1, 2200] cost: 0.383
   [1, 2400] cost: 0.303
   [1, 2600] cost: 0.226
   [1, 2800] cost: 0.199
   [1, 3000] cost: 0.184
   [1, 3200] cost: 0.204
   [1, 3400] cost: 0.199
   [1, 3600] cost: 0.148
   [1, 3800] cost: 0.188
   [1, 4000] cost: 0.190
   [2, 200] cost: 0.146
   [2, 400] cost: 0.146
   [2, 600] cost: 0.141
   [2, 800] cost: 0.179
   [2, 1000] cost: 0.217
   [2, 1200] cost: 0.154
   [2, 1400] cost: 0.097
   [2, 1600] cost: 0.162
   [2, 1800] cost: 0.132
   [2, 2000] cost: 0.177
   [2, 2200] cost: 0.241
   [2, 2400] cost: 0.195
   [2, 2600] cost: 0.183
   [2, 2800] cost: 0.251
   [2, 3000] cost: 0.265
   [2, 3200] cost: 0.174
   [2, 3400] cost: 0.176
   [2, 3600] cost: 0.162

```

```
[2, 3800] cost: 0.194
[2, 4000] cost: 0.107
```

## ▼ 정확도 판단

## ▼ Confusion matrix

```
1 from sklearn.metrics import confusion_matrix
2 from sklearn.metrics import precision_score, recall_score

1 with torch.no_grad():
2     X_test = test_data[:, 1:].view(-1, 28 * 28).float().to(device)
3     y_test = test_data[:, 0].float()
4
5     prediction = model(X_test).cpu()
6     print(confusion_matrix(torch.argmax(prediction, 1), y_test))
7     print("==Precision==")
8     print(precision_score(torch.argmax(prediction, 1), y_test, average=None))
9     print(precision_score(torch.argmax(prediction, 1), y_test, average='weight
10    print("Recall")
11    print(recall_score(torch.argmax(prediction, 1), y_test, average=None))
12    print(recall_score(torch.argmax(prediction, 1), y_test, average='weighted')

[[9046 301]
 [ 62 591]]
==Precision==
[0.9931928 0.66255605]
0.9716022181751223
Recall
[0.96779715 0.9050536 ]
0.9637
```

✓ 0초 오전 11:42에 완료됨

