

```
import os
import numpy as np
import copy

import torch
import torchvision
from torch import nn
from torch.autograd import Variable
from torchvision import datasets, transforms
from torch.utils.data import DataLoader
from torchvision import transforms
from torchvision.utils import save_image
import torch.nn.functional as F
from sklearn.model_selection import KFold
from sklearn.model_selection import StratifiedKFold
import torchvision.models as models
```

```
def train(train_loader):
    model.train()
    running_loss = 0
    for batch_idx, (images, labels) in enumerate(train_loader):
        images = images.to(device)
        labels = labels.to(device)
        print(batch_idx, end=" ")
        optimizer.zero_grad()
        outputs = model(images)

        loss = criterion(outputs, labels)
        running_loss += loss.item()

        loss.backward()
        optimizer.step()

    train_loss = running_loss / len(train_loader)
```

```
return train_loss
```

```
def valid(test_loader):
```

```
    model.eval()
```

```
    running_loss = 0
```

```
    correct = 0
```

```
    total = 0
```

```
    with torch.no_grad():
```

```
        for batch_idx, (images, labels) in enumerate(test_loader):
```

```
            images = images.to(device)
```

```
            labels = labels.to(device)
```

```
            outputs = model(images)
```

```
            loss = criterion(outputs, labels)
```

```
            running_loss += loss.item()
```

```
            predicted = outputs.max(1, keepdim=True)[1]
```

```
            correct += predicted.eq(labels.view_as(predicted)).sum().item()
```

```
            total += labels.size(0)
```

```
    val_loss = running_loss / len(test_loader)
```

```
    val_acc = correct / total
```

```
    return val_loss, val_acc
```

```
if __name__ == "__main__":
```

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

```
    print(device)
```

```
    np.random.seed(1234)
```

```
    torch.manual_seed(1234)
```

```
    num_epochs = 50
```

```
    batch_size = 128
```

```

learning_rate =0.00025

for fold in range(3):
    resnet = models.resnet50(pretrained=False)
    resnet.fc = nn.Linear(2048, 2)
    model = resnet.to(device)
    criterion = nn.CrossEntropyLoss()
    optimizer = torch.optim.Adam(model.parameters(), lr=learning_rate)

    train_file_dir=str(fold)+'_train_dir\\'
    validation_file_dir=str(fld)+'_validation_dir\\'
    logging_dir=str(fld)+'_logging_dir\\'

    transform = transforms.Compose([
        transforms.Resize((224, 224)),
        transforms.ToTensor()
    ])
    train_dataset = datasets.ImageFolder(train_file_dir, transform)
    train_data_loader = DataLoader(train_dataset, batch_size=batch_size, shuffle=True)

    validation_dataset = datasets.ImageFolder(validation_file_dir, transform)
    validation_data_loader = DataLoader(validation_dataset, batch_size=batch_size,
shuffle=False)

    best_val_acc=0

    for epoch in range(num_epochs):
        loss = train(train_data_loader)
        val_loss, val_acc = valid(validation_data_loader)
        print('epoch %d, loss: %.4f val_loss: %.4f val_acc: %.4f' % (epoch, loss, val_loss,
val_acc))

        if val_acc > best_val_acc :
            best_val_acc = val_acc
            print("Saving best model")
            best_model_wts = copy.deepcopy(model.state_dict())

```

```
print("Best validation accuracy: {:.4f}".format(best_val_acc))  
torch.save(best_model_wts, os.path.join(logging_dir, "model_resnet50.pkl"))
```