## 1. 알약 분류 유효성 검증 준비

알약 분류 기능의 유효성을 검증하기 위해서는 아래와 같이 사전에 필요한 구성요소들이 필요하다.

- 1. 실행에 필요한 python package 설치.
- 2. 폴더로 구분되어진 알약 이미지들
- 3. 실행에 필요한 file 들

그리고 기본적으로 실행환경은 Ubuntu 18.04, Python3.8 을 기준으로 한다. (다른 버전으로 해도 크게 문제 되지 않는다.)

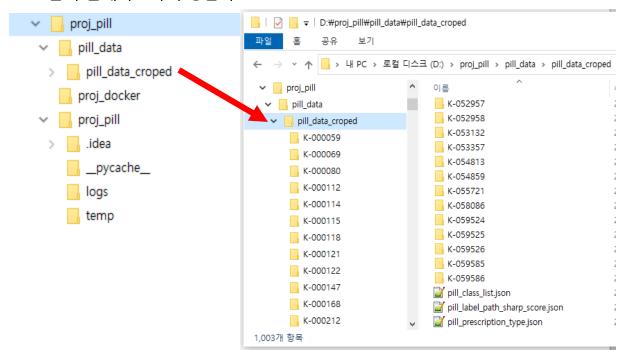
#### ■ 실행에 필요한 python package 설치

Ubuntu 18.04와 Python 3.8이상이 설치 된 것을 가정하고, 아래 package을 설치한다.

- Nvidia cuda driver 10.2 (혹은 11.0 이상)설치
- pip install torch torchvision (https://pytorch.org/ 참조)
- pip install numpy
- pip install opency-python
- pip install imgaug
- pip install PIL
- pip install tqdm
- pip install codecs
- pip install json
- pip install matplotlib

#### ■ 폴더로 구분되어진 알약 이미지들

알약을 구분할 수 있는 ID이름으로 폴더를 구성하고 그 안에 해당 알약 이미지들이 존재하도록 구성한다.



위 그림처럼 proj\_pill/pill\_data/pill\_data\_croped 안에 알약 folder을 구성한다.

추가적으로 제공하는 pill\_label\_path\_sharp\_score.json과 pill\_class\_list.json이 그림과 같이 위치해 있어야 한다.

## 1. 알약 분류 유효성 검증 준비

pill\_label\_path\_sharp\_score.json 의 목적은 알약에 대한 ID와 Label을 정의하고 실행시 참조하게 되어 있다.

아래는 그 내용이다.

```
알약 Label
🔚 pill_label_path_sharp_score,json 🗵
         1 F{
2 F
                                                                                                                                                                                                                                   알약 ID
                                                    "pill label path s
                                                                                                                                                                                                                 (알약 폴더이름)

  Image: Control of the 
                                                                                          "K-037589",
                                                                                                                                                                                                                                                                         알약 선명도
         6
                                                                                         149.2305576160079,
                                                                                                                                                                                                                                                              (실행시 사용안함)
                                                                                         45.62416263502471,
                                                                                         422.67862429732025
        9
                                                                     ],
     10
     11
     12
                                                                                         "K-029534",
    13
                                                                                         103.72038138092651,
    14
                                                                                         29.223249162946423,
     15
                                                                                         289.5755319448373
    16
     17
    18
    19
                                                                                         "K-015270",
    20
                                                                                         74.87973794032236,
    21
                                                                                         27.717132812636635,
    22
                                                                                         157.15786420281157
    23
                                                                     ],
    24
    25
    26
                                                                                         "K-040531",
    27
                                                                                         136.75974612027386,
    28
                                                                                         27.235786729929398,
    29
                                                                                         435.2421273529804
     30
                                                                     ],
    31
     32
    33
                                                                                         "K-041153",
    34
                                                                                         52.66275258542345,
    35
                                                                                         25.46791224576989,
    36
                                                                                         99.80926907057565
    37
```

pill\_class\_list.json 의 목적은 알약의 class와 용도을 정의하고 실행시 참조하게 되어 있다.

Class0: 카메라 각도가 90도, 75도로 캡쳐된 알약이미지의 list Class1: 카메라 각도가 70도, 60도로 캡쳐된 알약이미지의 list

```
"pngfile class0 train": [
 "/home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/K-026788/K-026788_0_0_0_2_90_240_200.png",
 "/home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/K-046428/K-046428_0_2_0_2_75_340_200.png",
 "/home/ubuntu/proj/proj pill/pill data/pill data croped/K-001438/K-001438 0 0 1 0 75 140 200.png",
 "pngfile class0 valid": [
 "/home/ubuntu/proj/proj pill/pill data/pill data croped/K-046428/K-046428 0 2 0 1 90 140 200.png",
 "/home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/K-046428/K-046428_0_1_0_0_90_220_200.png",
 "/home/ubuntu/proj/proj pill/pill data/pill data croped/K-001728/K-001728 0 2 1 0 90 020 200.png",
 "pngfile class0 test": [
 "/home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/K-052619/K-052619_0_2_0_2_75_140_200.png",
 "/home/ubuntu/proj/proj pill/pill data/pill data croped/K-006250/K-006250 0 1 0 1 75 000 200.png",
 "/home/ubuntu/proj/proj pill/pill data/pill data croped/K-038576/K-038576 0 0 0 2 75 120 200.png",
 "pngfile class1 train": [
 "/home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/K-044085/K-044085_0_0_1_1_70_340_200.png",
 "/home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/K-009458/K-009458_0_2_0_0_60_240_200.png",
 "/home/ubuntu/proj/proj pill/pill data/pill data croped/K-023720/K-023720 0 1 1 2 60 060 200.png",
 "pngfile class1 valid": [
 "/home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/K-024752/K-024752_0_2_1_2_60_320_200.png",
 "/home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/K-037589/K-037589_0_2_0_2_70_140_200.png",
 "/home/ubuntu/proj/proj pill/pill data/pill data croped/K-037589/K-037589 0 2 1 2 60 180 200.png",
 "pngfile_class1_train": [
 "/home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/K-037043/K-037043_0_2_0_1_60_080_200.png",
 "/home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/K-018254/K-018254_0_2_1_0_70_200_200.png",
 "/home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/K-005949/K-005949_0_0_1_0_60_260_200.png",
위의 내용에서
'pngfile_class0_train' 은 class0에 해당하고, 모델 학습용임을 의미함. 'pngfile_class0_valid' 은 class0에 해당하고, 모델 검증용임을 의미함.
```

## 1. 알약 분류 유효성 검증 준비

또한 위의 내용에서 'pngfile\_class0\_test' 은 class0에 해당하고, 모델 시험용임을 의미함. 'pngfile\_class1\_train' 은 class1에 해당하고, 모델 학습용임을 의미함. 'pngfile\_class1\_valid' 은 class1에 해당하고, 모델 검증용임을 의미함. 'pngfile\_class1\_test' 은 class1에 해당하고, 모델 시험용임을 의미함.

그리고 이미지 파일이 절대경로이므로, 이 경로를 맞추어야 함. "/home/ubuntu/proj/proj\_pill/pill\_data/pill\_data\_croped/K-026788/K-026788\_0\_0\_0\_2\_90\_240\_200.png",

- \*\* 절대경로를 python file에서 바꿀 수 있지만, 편의상 고정경로만 설명함.
- \*\* 학습용, 검증용, 시험용 폴더는 존재하지 않음. 'pill\_class\_list.json' file이 기능을 대신함.

#### ■ 실행에 필요한 file 들

- 실행할 python file들은 이미 학습한 모델 파일이 필요한다. 아래와 같다.

pill\_resnet152\_dataclass0\_aug0.pt : 알약 class0으로 학습한 모델 파일 pill\_resnet152\_dataclass1\_aug0.pt : 알약 class1으로 학습한 모델 파일 pill\_resnet152\_dataclass01\_aug0.pt : 알약 class0, class1으로 학습한 모델 파일

- 실행할 python file은 아래와 같다.

main\_cls0.py : 알약 class0을 시험할 실행 파일. main\_cls1.py : 알약 class1을 시험할 실행 파일.

main\_cls01.py: 알약 class0, class1을 시험할 실행 파일.

- 그 외 필요한 파일들.

get\_cli\_args.py: 환경변수을 만들거나, 수정할 수 있다.

gen\_pill.py : 이미지들을 읽고, 학습할 수 있게 dataset을 만든다.

pill\_classifier.py : 학습과 검증을 시행하는 code.

make\_label\_sharpness.py : data 준비용 make\_pill\_class\_list.py : data 준비용.

### 2. 알약 분류 유효성 검증 시행

- 2.1 알약 분류 검증 전 아래와 같이 환경 폴더가 지정되어 있다고 가정한다.
- ~/proj/proj\_pill/proj\_pill: python 파일과 모델 파일이 위치한 경로.
- ~/proj/proj\_pill/pill\_data/pill\_data\_cropped:
  - → 이미지가 폴더로 구분되어 위치한 경로
  - → pill\_class\_list.json, pill\_label\_path\_sharp\_score.json 가 존재함.
- 2.2 알약 class0에 대해 아래와 같이 실행한다.

```
(base) root@55e94ecbb291:/# cd /home/ubuntu/proj/proj pill/proj pill
(base) root@55e94ecbb291:/home/ubuntu/proj/proj pill/proj pill# python main_cls0.py
job=resnet152 run_phase:test aug_level:0, dataclass:0
BATCH SIZE:8, num workers:4, num threads:2
model_path_in is /home/ubuntu/proj/proj_pill/proj_pill/pill_resnet152_dataclass0_aug0.pt
dataset dir is /home/ubuntu/proj/proj_pill/pill_data/pill_data_croped/pill_class_list.json
run phase is test, aug level is 0
gen type is read only image, loading data ...
label path was loaded from <<< pngfile_class0_test >>>
data loading done. dataset'length is 64740
valid dataset was loaded
dataset loading time is 3.2785534858703613
optimizer was selected as type:sgd
model path will be loaded from:/home/ubuntu/proj/proj pill/proj pill/pill resnet152 dataclass0 aug0.pt
model was loaded from state
Mon Dec 6 03:08:52 2021
resnet152 aug level:0 :test Epoch #0: 100%
                                                                                                                                          8093/8093 [58:21<00:00, 2.31it/s, loss=0.0111, top1=99.7, top5=100]
Epoch: [0][8092/8093] Data time 3501.450 (3501.450 Now:2021-12-06 04:07:14.246878)
                                                                                       Loss 718.4540 (0.0111)Accuracy top1:99.7312, top5:100.0000
Mon Dec 6 04:07:14 2021
(base) root@55e94ecbb291:/home/ubuntu/proj/proj_pill/proj_pill#
```

- → 위의 실행은 모델명:reset152, Batch size:8, 모델파일:pill\_resnet152\_dataclass0\_aug0.pt, dataset길이:64740, optimizer:sgd
- → top1:99.7, top5:100

## 2. 알약 분류 유효성 검증 시행

2.3 특정한 알약 이미지를 모델이 예측한 label을 확인하려면 실행폴더에서 'dir\_testimage' sub 폴더를 만들고, 확인하려는 알약이미지를 넣어둔다.

그리고 아래 처럼 실행한다.

```
(horovod) ubuntu@gpu-1:~/proj/proj pill/proj pill$ python main cls01 dir.py
job=resnet152 run phase:test aug level:0, dataclass:01
BATCH SIZE:64, num workers:4, num threads:2
model_path_in is /home/ubuntu/proj/proj_pill/proj_pill/prill_resnet152_dataclass01_aug0.pt
valid dataset was loaded
2021-12-09 15:14:48.593789: I tensorflow/stream executor/platform/default/dso loader.cc:49] Successfully opened dynamic library libcudart.so.10.1
model was loaded from state
Thu Dec 9 15:14:58 2021
resnet152 aug level:0 :test Epoch #0: 0%
                                                                                                                                                                                     | 0/1 [00:00<?, ?it/s]
/home/ubuntu/anaconda3/envs/horovod/lib/python3.9/site-packages/torch/nn/functional.py:718: UserWarning: Named tensors and all their associated APIs are an experimental feature and subject to change. Pl
ease do not use them for anything important until they are released as stable. (Triggered internally at /pytorch/c10/core/TensorImpl.h:1156.)
 return torch.max pool2d(input, kernel size, stride, padding, dilation, ceil mode)
resnet152 aug level:0 :test Epoch #0: 100%
                                                                                                                                                   1/1 [00:01<00:00, 1.06s/it, loss=29.4, top1=0, top5=0]
Epoch: [0][0/1] Data time 1.064 (1.064 Now: 2021-12-09 15:14:59.822028) Loss 58.7487 (29.3743) Accuracy top1:0.0000, top5:0.0000
Thu Dec 9 15:14:59 2021
('K-045583 0 0 0 0 60 340 200.png', 'K-051669 0 0 0 0 60 100 200.png')
[743, 494]
job done
(horovod) ubuntu@gpu-1:~/proj/proj pill/proj pill$ ■
```

→ 위의 실행은 'K-045583 0 0 0 0 60 340 200.png', 'K-051669 0 0 0 0 60 100 200.png' 이미지가 각각 [743, 494] 으로 예측되었음을 표시한다.

# 2. 알약 분류 유효성 검증 시행(Docker이용시)

# 1. Docker 설치하기

```
~$> sudo apt-get update

~$> sudo apt-get install apt-transport-https ca-certificates curl gnupg-
agent software-properties-common

~$> curl -fsSL https://download.docker.com/linux/ubuntu/gpg | sudo apt-key
add -

~$> sudo add-apt-repository "deb [arch=amd64]
https://download.docker.com/linux/ubuntu $(lsb_release -cs) stable"

~$> sudo apt-get update

~$> sudo apt-get install docker-ce docker-ce-cli

설치 확인.

~$> sudo docker version
```

# 2. 알약용 image을 Docker에 load 하기

```
ubuntu@gpu-1:~$ cd proj/proj pill/proj docker/
ubuntu@gpu-1:~/proj/proj pill/proj docker$ dir
total 4964700
drwxrwxr-x 2 ubuntu ubuntu
                           4096 Nov 15 15:27 .
drwxrwxr-x 5 ubuntu ubuntu
                           4096 Nov 15 15:20 ...
-rw-rw-r-- 1 ubuntu ubuntu 5083836928 Nov 15 15:07 pill class.tar
ubuntu@gpu-1:~/proj/proj pill/proj docker$ sudo docker load -i
pill class.tar
824bf068fd3d: Loading layer
[========] 65.51MB/65.51MB
0677e35507df: Loading layer
[=======>] 3.576GB/3.576GB
dce55ae465d9: Loading layer
[=========] 1.326GB/1.326GB
cc1427064650: Loading layer
[========] 116.3MB/116.3MB
Loaded image: ubuntu:pill
ubuntu@gpu-1:~/proj/proj_pill/proj_docker$ sudo docker images
REPOSITORY
          TAG
                  IMAGE ID
                               CREATED
                                             SIZE
          pill
                  8d50ddab7868
ubuntu
                               45 minutes ago
                                             4.96GB
ubuntu@gpu-1:~/proj/proj pill/proj docker$
```

# 2. 알약 분류 유효성 검증 시행(Docker이용시)

# 3. 알약용 docker image을 실행하기

~\$> ubuntu@gpu-1:~/proj/proj\_pill/proj\_docker\$ sudo docker run -it -v /home/ubuntu/proj/proj\_pill:/home/ubuntu/proj/proj\_pill ubuntu:pill

(base) root@7cce986fb0b5:/#

Docker 내부의 작업위치

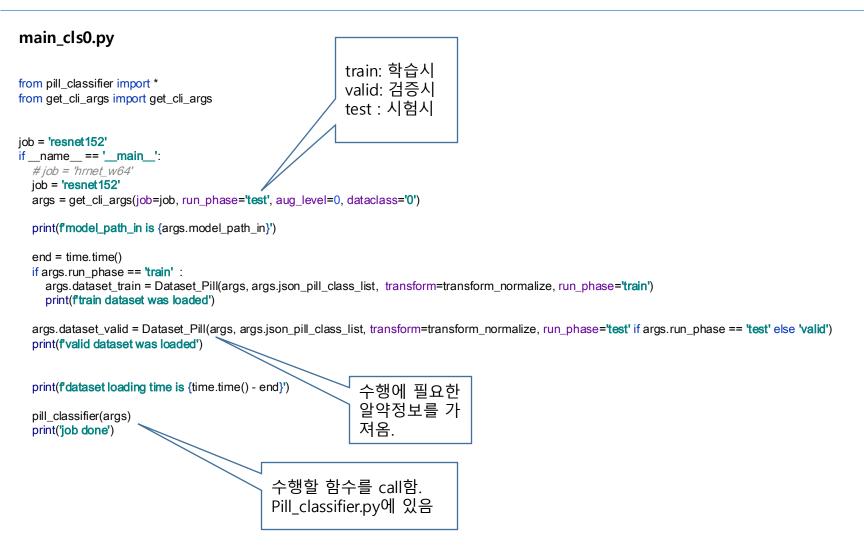
Docker 내부의 작업위치

외부 director에 python file과 시험할 이미지가 있음.

### 4. 알약 class 0 에 대해 시험해보기

```
(base) root@7cce986fb0b5:/# cd /home/ubuntu/proj/proj pill/proj pill
(base) root@7cce986fb0b5:/home/ubuntu/proj/proj_pill/proj_pill# python main_cls0.py _
job=resnet152 run phase:test aug level:0, dataclass:0
BATCH_SIZE:8, num_workers:4, num_threads:2
model path in is /home/ubuntu/proj/proj pill/proj pill/pill resnet152 dataclass0 aug0.pt
                                                                                                                 알약 class 1 에 대해 estimate 하려면,
dataset dir is /home/ubuntu/proj/proj pill/pill data/pill data croped/pill class list.json
                                                                                                                         'python main cls1.py'
run phase is test, aug level is 0
gen_type is read_only_image, loading data ...
label path was loaded from <<< pngfile class0 test >>>
data loading done. dataset'length is 64740
valid dataset was loaded
dataset loading time is 1.5846672058105469
optimizer was selected as type:sgd
model path will be loaded from:/home/ubuntu/proj/proj pill/proj pill/pill resnet152 dataclass0 aug0.pt
model was loaded from state
resnet152 aug level:0 :test Epoch #0:
                                                                         | 14/8093 [00:48<7:43:26, 3.44s/it, loss=0.00693, top1=100, top5=100]
```

- 주의: 1. 사용되는 docker image은 GPU을 지원하지 않음.
  - 2. docker은 시험환경만 제공하고, 내부에 source code와 이미지를 포함하지 않는다(외부 연결 사용).



from get\_cli\_args import get\_cli\_args

#### pill\_classifier.py

import torch.nn as nn

import torch

```
from torch.utils.data import Dataset, DataLoader from torchvision import transforms, models import torch.optim as optim import torch.backends.cudnn as cudnn from torch.optim.lr_scheduler import ReduceLROnPlateau

from gen_pill import Gen_Digit from hrnet import get_hrnet from utils import model_load, model_save, accuracy, get_optimizer, transform_normalize, AverageMeter import time from tqdm import tqdm from torch.utils.tensorboard import SummaryWriter import os import datetime
```

```
class Dataset_Pill(Dataset):
  def init (self, args, dir dataset, transform=None, target transform=None, run phase='train'):
     self.args = args
     self.gen_digit = Gen_Digit(args, dir_dataset, run_phase)
     self.transform = transform
     self.target_transform = target_transform
     self.run_phase = run_phase
  def len (self):
     return self.gen_digit.len_total
  def getitem (self, idx):
     image, label, path img, aug name = self.gen_digit.generate_digits_by_index(self.args, idx)
     if self transform is not None:
       image = self.transform(image)
     if self.target_transform is not None:
       label = self.target transform(label)
     if self.run phase == 'valid' or self.run phase == 'test':
       return image, label, path_img, aug_name
     else:
       return image, label
def get_pill_model(args):
  if args.cnn_name == 'resnet152':
     model = models.resnet152(num_classes=args.num_classes)
  elif args.cnn name == 'hrnet w64':
     model = get hrnet()
     model.classifier = nn.Linear(in_features=2048, out_features=args.num_classes, bias=True)
     raise Exception('No Found CNN Name')
  if args.cuda == True:
     if args.gpu is not None:
       model.cuda(args.gpu)
     else:
       model.cuda()
  else:
     model.cpu()
  return model
```

```
pill classifier.py
def train(args, dataloader, sampler, model, criterion, optimizer, epoch, log_writer=None, verbose=True):
  model.train()
  if sampler != None:
                                                 학습시 수행
     sampler.set_epoch(epoch)
  metric train loss = AverageMeter()
  ametric data time = AverageMeter()
  top1 = AverageMeter()
  top5 = AverageMeter()
  end = time.time()
  Ir = optimizer.param groups[0]['Ir']
  with tqdm(total=len(dataloader), desc=args.tqdm_desc_head + Train Epoch #t]*.format(epoch), disable=not verbose) as t:
    for batch idx, (img, target) in enumerate(dataloader):
       if args.cuda:
          img = img.cuda()
          target = target.cuda()
       optimizer.zero_grad()
       output = model(img)
       loss = criterion(output, target)
       loss.backward()
       optimizer.step()
        prec1, prec5 = accuracy(output, target, (1, 5))
       count_try = img.cpu().shape[0]
       top1.update(prec1[0].detach().cpu().item(), count_try)
       top5.update(prec5[0].detach().cpu().item(), count_try)
       metric_train_loss.update(loss.detach().cpu().item(), count_try)
       t.set_postfix({"loss": metric_train_loss.avg, "lr":lr, "top1":top1.avg, "top5":top5.avg })
       t.update(1)
  ametric_data_time.update(time.time() - end)
  if loa writer:
    log_writer.add_scalar('train/loss', metric_train_loss.avg, epoch)
  try:
     print_string = 'Epoch: [{0}][{1}/{2}]\t'.format(epoch, batch_idx, len(dataloader))
     print_string += 'Data time {data_time.val:.3f} ({data_time.avg:.3f} Now:{Now})\t'.format(data_time=ametric_data_time,
Now=datetime.datetime.now())
     print_string += 'Loss {loss.val:.4f} ({loss.avg:.4f})'.format(loss=metric_train_loss)
     print_string += 'Accuracy top1:{top1.avg:.4f}, top5:{top5.avg:.4f}'.format(top1=top1, top5=top5)
     print(print_string)
  except:
```

pass

return metric\_train\_loss.avg

```
def valid(args, dataloader, sampler, model, criterion, epoch, log_writer=None, verbose=True):
  metric_train_loss = AverageMeter()
  ametric data time = AverageMeter()
  top1 = AverageMeter()
                                                                           검증시 수행
  top5 = AverageMeter()
  if sampler != None:
     sampler.set_epoch(epoch)
  model.eval()
  end = time.time()
  args.list_preds = []
  args.list_target = []
  args.count_correct = 0
  with torch.no_grad():
     with tqdm(total=len(dataloader), desc=args.tqdm_desc_head + '{} Epoch #{}'.format( args.run_phase, epoch), disable=not verbose)
as t:
       for i, (img, target, path_img, aug_name) in enumerate(dataloader):
          if args.cuda:
            img = img.cuda()
            target = target.cuda()
          output = model(img)
          loss = criterion(output, target)
          prec1, prec5 = accuracy(output, target, (1, 5))
          if ( prec1[0].detach().cpu().item() != 100.):
             if args.run_phase == 'valid': print(f'<---- class valid fail file: {path_img[0]}, aug_name:{aug_name}')
          preds = output.data.max(dim=1, keepdim=True)[1]
          count_correct = preds.eq(target.data.view_as(preds)).cpu().sum()
          list_preds = preds.view(-1).tolist()
          args.list_preds = list_preds
          args.list_target = target.detach().cpu().tolist()
          args.count_correct = count_correct.item()
          args.path_img = path_img
          count_try = img.cpu().shape[0]
          top1.update(prec1[0].detach().cpu().item(), count_try)
          top5.update(prec5[0].detach().cpu().item(), count_try)
          metric_train_loss.update(loss.detach().cpu().item(), count_try)
          t.set_postfix({'loss': metric_train_loss.avg, 'top1': top1.avg, 'top5': top5.avg})
          t.update(1)
  ametric_data_time.update(time.time() - end)
  if log writer:
     log_writer.add_scalar('validation/loss', metric_train_loss.avg, epoch)
  try:
     print_string = 'Epoch: [{0}][{1}/{2}]\t'.format(epoch, i, len(dataloader))
     print_string += 'Data time {data_time.val:.3f} ({data_time.avg:.3f} Now:{Now})\t'.format(data_time=ametric_data_time)
Now=datetime.datetime.now())
     print_string += 'Loss {loss.val:.4f} ({loss.avg:.4f})'.format(loss=metric_train_loss)
     print_string += 'Accuracy top1:{top1.avg:.4f}, top5:{top5.avg:.4f}'.format(top1=top1, top5=top5)
     print(print string)
  except:
     pass
  return metric train loss.avg
```

#### pill\_classifier.py

```
def run_model(args, model, dataloader_train, dataloader_valid, sampler_train, sampler_valid,
criterion.optimizer, epoch begin, log writer, verbose=True ):
  if args.run phase == 'valid' or args.run phase == 'test':
     print(time.asctime())
    valid(args, dataloader valid, sampler valid, model, criterion, 0, log writer, verbose)
     print(time.asctime())
     return
  Ir scheduler = ReduceLROnPlateau(optimizer, mode='min', factor=0.8, patience=2, verbose=True,
threshold=0.0001, threshold mode='rel', cooldown=3, min Ir=0, eps=1e-08)
  best_perf = 1000
  for epoch in range(epoch_begin, args.epochs): #(0:100)
     # adjust_learning_rate(args, optimizer, epoch)
     # train for one epoch
     perf indicator = train(args, dataloader train, sampler train, model, criterion, optimizer, epoch, log writer,
verbose)
    if epoch > 10:
       perf_indicator = valid(args, dataloader_valid,sampler_valid, model, criterion, epoch,
log_writer, verbose)
       if (args.gpu == 0):
          print(f perf_indicator:{perf_indicator}, best_perf:{best_perf}')
       if perf_indicator < best_perf:
          model_save(args.model_path, epoch, model, optimizer, args.rank)
          best_perf = perf_indicator
     else:
       model_save(args.model_path, epoch, model, optimizer, args.rank)
       best_perf = perf_indicator
    Ir_scheduler.step(perf_indicator)
model = None
criterion = None
optimizer = None
epoch begin = 0
log writer = None
```

```
def pill_classifier(args):
  global model, criterion, optimizer, epoch begin, log writer
  if args.dataset_valid != None:
    dataloader_valid = DataLoader(args.dataset_valid, batch_size=args.batch_size, shuffle= False,
num_workers=args.num_workers)
  else:
     dataloader valid = None
  if args.run_phase == 'train' and args.dataset_train != None:
     dataloader_train = DataLoader(args.dataset_train, batch_size=args.batch_size, shuffle=True,
num workers=args.num workers)
  else:
     dataloader train = None
  if model == None:
    log_writer = SummaryWriter(args.dir_log)
    if args.cuda == False or torch.cuda.device_count() == 0 :
       args.gpu = None
     else:
       args.gpu = 0
     args.rank = args.gpu
    cudnn.benchmark = True
    torch backends cudnn deterministic = False
    torch.backends.cudnn.enabled = True
     model = get_pill_model(args)
     # define loss function (criterion) and optimizer
    criterion = torch.nn.CrossEntropyLoss()
    if args.cuda:
       criterion = criterion.cuda()
    optimizer = get_optimizer(args,model)
    epoch_begin, dict_checkpoint, success = model_load(args, model, optimizer)
  run model(args, model, dataloader train, dataloader valid, None, None, criterion, optimizer, epoch begin,
log_writer)
```

#### main\_cls01\_dir.py

```
from pill_classifier import *
from get_cli_args import get_cli_args
from pathlib import Path
from PIL import Image
import os
class Dataset Dir(Dataset):
  def __init__(self, args, dir_dataset, transform=None, target_transform=None, run_phase='train'):
    self.args = args
    self.dir_dataset = dir_dataset
     self.transform = transform
    self.target_transform = target_transform
    self.list_images = [ png.name_for png in Path(dir_dataset).iterdir() if png.suffix == '.png']
    self.run_phase = run_phase
  def len (self):
    return len(self.list_images)
  def __getitem__(self, idx):
    image = Image.open(os.path.join(self.dir_dataset, self.list_images[idx]))
     label = 0
    path_img = self.list_images[idx]
    aug_name = ""
     if self.transform is not None:
       image = self.transform(image)
    if self.target_transform is not None:
       label = self.target_transform(label)
    if self.run_phase == 'valid' or self.run_phase == 'test':
       return image, label, path_img, aug_name
     else:
       return image, label
```