## annotation 구한 unlabeled data + labeled data로 모델 학습

```
In [8]:
         import os
         import numpy as np
         import pandas as pd
         import pickle
In [2]:
         # for modeling
         import torch
         torch.manual_seed(0)
         import matplotlib.patches as patches
         import matplotlib.pyplot as plt
         from PIL import Image
         import torchvision
         from torchvision import transforms, datasets, models
         from torchvision.models.detection.faster_rcnn import FastRCNNPredictor
         import time
         from tgdm import tgdm
         import csv
In [3]:
         os.chdir('/home/work/sample-notebooks/train')
```

# labeled data(train, test) 파일명 list 로드

```
In [4]: # train 데이터
with open('./labeled_data/train_img.pkl', 'rb') as file:
    train_img_list = pickle.load(file)

# test 데이터
with open('./labeled_data/test_img.pkl', 'rb') as file:
    test_img_list = pickle.load(file)
```

### unlabeled data 파일명 list 생성

```
In [5]:
       ul_img_path = './unlabeled_data/'
       ul_img_list = os.listdir(ul_img_path)
       ul_img_list.sort()
       del ul_img_list[-1]
        print(len(ul_img_list))
        print(ul_img_list[:5])
       19000
       I_000004.jpg']
In [6]:
       # train_img_list와 ul_img_list 합치기
       ssl_train_img = train_img_list + ul_img_list
       print(len(ssl_train_img))
       print(ssl_train_img[:2], ssl_train_img[-2:])
       19900
       ['sk_tr_000863.jpg', 'sk_tr_000822.jpg'] ['sk_ul_018998.jpg', 'sk_ul_018999.jpg']
```

## 데이터셋 클래스 정의

```
In [9]: def generate_box(df_obj, size): # 객체 하나씩 (한 이미지에 객체 여러개여도 하나씩)
             W = size[0]
             H = size[1]
             xmin = df_obj['xmin']*W
             ymin = df_obj['ymin']*H
             xmax = df_obj['xmax']*W
             ymax = df_obj['ymax']*H
             return [xmin, ymin, xmax, ymax]
         def generate_label(df_obj): # 원래 label에 1씩 더해서 반환
             adjust_label = 1
             return int(df_obj['class'] + adjust_label)
         def generate target(file, size):
             if 'ul' in file:
                df = pd.read_table(file, sep = ',', header = None, names = ['class','xmin',']
                df = pd.read_table(file, sep = ' ', header = None, names = ['class', 'xmin', ']
             boxes = []
             labels = []
             for obj in range(df.shape[0]):
                 boxes.append(generate_box(df.iloc[obj], size))
                 labels.append(generate_label(df.iloc[obj]))
             boxes = torch.as_tensor(boxes, dtype = torch.float32)
             labels = torch.as_tensor(labels, dtype = torch.int64)
             target = {}
             target["boxes"] = boxes
             target["labels"] = labels
             return target
```

```
In [10]:
          ## <labeled_train + unlabeled (19900개)>, <labeled_test(100개)>
          class MaskDataset(object):
              def __init__(self, transforms, path, imgs):
                  self.transforms = transforms
                  self.path = path # img path
                  self.imgs = imgs # img 파일명 list
              def __getitem__(self, idx):
                  # load image and masks
                  file_image = self.imgs[idx]
                  file_label = self.imgs[idx][:-3] + 'txt'
                  if 'tr' in file_image: ## labeled
                      img_path = os.path.join(self.path, file_image)
                      if 'test' in self.path: ## labeled - test
                          label_path = os.path.join('./labeled_data/test_annotations/', file_lal
                      else:
                                               ## labeled - train
                          label_path = os.path.join('./labeled_data/train_annotations/', file_l
                  elif 'ul' in file_image: ## unlabeled
                      self.path = './unlabeled_data/'
                      img_path = os.path.join(self.path, file_image)
                      label_path = os.path.join('./ul_annots_rate_ssl/', file_label)
```

```
img = Image.open(img_path).convert('RGB')
        size = img.size
        # generate label
        target = generate_target(label_path, size)
        if self.transforms is not None:
            img = self.transforms(img)
        return img, target
   def __len__(self):
        return len(self.imgs)
data_transform = transforms.Compose([
   transforms.ToTensor()
    1)
def collate_fn(batch):
    return tuple(zip(*batch))
dataset = MaskDataset(data_transform, './labeled_data/train_images/', ssl_train_img) #
test_dataset = MaskDataset(data_transform, './labeled_data/test_images/', test_img_lis
data_loader = torch.utils.data.DataLoader(dataset, batch_size = 2, collate_fn = colla
test_data_loader = torch.utils.data.DataLoader(test_dataset, batch_size = 1, collate_
```

### 모델 불러오기

```
In [11]:

def get_model_instance_segmentation(num_classes): # num_classes 는 background 클래스

model = torchvision.models.detection.fasterrcnn_resnet50_fpn(pretrained = True)
in_features = model.roi_heads.box_predictor.cls_score.in_features
model.roi_heads.box_predictor = FastRCNNPredictor(in_features, num_classes)

return model
```

# 전이학습

```
In [12]: model = get_model_instance_segmentation(8) # 실제 클래스 개수 : 7 (0~6)

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

```
RuntimeError
                                          Traceback (most recent call last)
/tmp/ipykernel_10120/2218573399.py in <module>
      3 device = torch.device('cuda') if torch.cuda.is_available() else torch.device(
'cpu')
---> 4 model.to(device)
/opt/conda/lib/python3.7/site-packages/torch/nn/modules/module.py in to(self, *args, *
*kwargs)
    610
                     return t.to(device, dtype if t.is_floating_point() else None, non
_blocking)
    611
                return self._apply(convert)
--> 612
    613
    614
             def register_backward_hook(
```

```
/opt/conda/lib/python3.7/site-packages/torch/nn/modules/module.py in _apply(self, fn)
             def _apply(self, fn):
                 for module in self.children():
     358
--> 359
                    module._apply(fn)
     360
                 def compute_should_use_set_data(tensor, tensor_applied):
     361
/opt/conda/lib/python3.7/site-packages/torch/nn/modules/module.py in _apply(self, fn)
             def _apply(self, fn):
     358
                 for module in self.children():
--> 359
                    module._apply(fn)
     360
     361
                 def compute_should_use_set_data(tensor, tensor_applied):
/opt/conda/lib/python3.7/site-packages/torch/nn/modules/module.py in _apply(self, fn)
     357
             def _apply(self, fn):
     358
                 for module in self.children():
--> 359
                    module._apply(fn)
     360
     361
                 def compute_should_use_set_data(tensor, tensor_applied):
/opt/conda/lib/python3.7/site-packages/torch/nn/modules/module.py in _apply(self, fn)
     379
                         # `with torch.no_grad():`
     380
                         with torch.no_grad():
--> 381
                            param_applied = fn(param)
                         should_use_set_data = compute_should_use_set_data(param, para
     382
m_applied)
     383
                         if should_use_set_data:
/opt/conda/lib/python3.7/site-packages/torch/nn/modules/module.py in convert(t)
     608
                     if convert_to_format is not None and t.dim() == 4:
     609
                         return t.to(device, dtype if t.is_floating_point() else None,
non_blocking, memory_format=convert_to_format)
--> 610
                    return t.to(device, dtype if t.is_floating_point() else None, non_
blocking)
     611
     612
                 return self._apply(convert)
RuntimeError: CUDA error: out of memory
학습
 num_epochs = 1
 params = [p for p in model.parameters() if p.requires_grad]
 optimizer = torch.optim.Adam(params, Ir = 0.005, weight_decay = 0.0005)
```

```
In [13]:
         #optimizer = torch.optim.SGD(params, Ir = 0.005, momentum = 0.9, weight_decay = 0.0005
In [14]:
         # 새로 학습 (start:0)
         print('-----')
         for epoch in range(num_epochs):
            start = time.time()
            model.train()
             i = 0
            epoch_loss = 0
             #numb = 0
             for imgs, annotations in tqdm(data_loader):
                imgs = list(img.to(device) for img in imgs)
                annotations = [{k: v.to(device) for k,v in t.items()} for t in annotations]
                try:
                    loss_dict = model(imgs, annotations)
```

```
losses = sum(loss for loss in loss_dict.values())
                 except:
                    #print(numb)
                    continue
                 optimizer.zero grad()
                 losses.backward()
                 optimizer.step()
                 epoch_loss += losses
                 \#numb += 1
             print(f'epoch : {epoch + 1}, Loss : {epoch_loss}, time : {time.time() - start}')
             torch.save(model.state_dict(), f'../Miso/weight_ssl/model_{epoch+1}.pt')
                  -----train start-----
         100%| 9950/9950 [42:25<00:00, 3.91it/s]
         epoch : 1, Loss : 0, time : 2545.747303247452
In [41]:
         # 학습 과정을 반복해야 하므로 Train 함수 생성
         def Train(start_epoch, end_epoch):
             print('-----')
             # 이전 가중치 불러오기
             model.load_state_dict(torch.load(f'../Miso/weight_ssl/model_{start_epoch}.pt'))
             for epoch in range(start_epoch, end_epoch):
                start = time.time()
                 model.train()
                 i = 0
                 epoch_loss = 0
                 for imgs, annotations in tqdm(data_loader):
                    i += 1
                    imgs = list(img.to(device) for img in imgs)
                    annotations = [\{k: v.to(device) for k, v in t.items()\} for t in annotation
                        loss_dict = model(imgs, annotations)
                        losses = sum(loss for loss in loss_dict.values())
                    except:
                        continue
                    optimizer.zero_grad()
                    losses.backward()
                    optimizer.step()
                    epoch_loss += losses
                 print(f'epoch : {epoch + 1}, Loss : {epoch_loss}, time : {time.time() - start
                 torch.save(model.state_dict(), f'../Miso/weight_ssl/model_{epoch + 1}.pt')
```

### • 학습

### 모델에 weight 로드

```
In [ ]: model.load_state_dict(torch.load(f'../Miso/weight_ssl/model_1.pt'))
```

### 예측

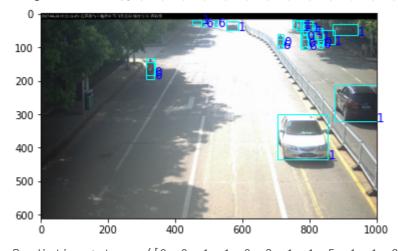
```
In [29]:
          def make_prediction(model, img, threshold):
              model.eval()
              preds = model(img)
              for id in range(len(preds)):
                  idx_list = []
                  for idx, score in enumerate(preds[id]['scores']): # 한 이미지 내에 검출된 객
                      if score > threshold: # 신뢰도가 threshold보다 높은것만 idx_list에 저장
                          idx_list.append(idx)
                  # thr보다 높은 객체들만 boxes, labels, scores 정보 추출해서 덮어쓰기로 저장
                  preds[id]['boxes'] = preds[id]['boxes'][idx_list]
                  preds[id]['labels'] = preds[id]['labels'][idx_list]
                                                                       -1 # label : 0-6
                  preds[id]['scores'] = preds[id]['scores'][idx_list]
              return preds
In [33]:
          with torch.no_grad():
              for imgs, annotations in test_data_loader:
                  im = imgs
                  imgs = list(img.to(device) for img in imgs)
                  pred = make_prediction(model, imgs, 0.4) # threshold
                  print(pred)
                  break
                           # test_data_loader의 첫 번째 배치만 결과 출력
         [{'boxes': tensor([[ 703.4944,
                                          59.9619,
                                                    725.3139,
                                                                 99.68711.
                 [ 774.2667,
                               55.1289,
                                         796.3938,
                                                     99.5487],
                              209.7026, 1001.0000,
                 [ 880.0817,
                                                    321.65331.
                               20.5485,
                                         590.6553,
                 [ 554.2569,
                                                     49.1402],
                 [ 312.9951,
                               132.5156,
                                         339.6205,
                                                     185.6905],
                 775.7050,
                               69.8275.
                                         794.6923.
                                                     108.3002].
                 [ 449.7556,
                               18.2737,
                                         483.4883,
                                                     32.0022],
                 [ 752.9364,
                               19.0609,
                                         782.9688,
                                                     44.3429],
                 [ 956.2925,
                               91.0808,
                                         991.5442,
                                                     129.3998],
                 [ 485.5266,
                               18.3976.
                                         520.0612.
                                                      33.1010].
                                                     49.7759],
                 [ 769.5856,
                               21.2229,
                                         811.9795,
                 [ 312.1575,
                                         337.2722,
                               145.5818,
                                                     194.1991],
                 [709.4665,
                              296.4266.
                                                     434.6549],
                                         855.6778.
                 [ 822.0977,
                               48.4249,
                                         878.2283,
                                                     94.57531.
                 [ 866.3963,
                               40.7793,
                                         886.5568,
                                                     65.0910],
                 [ 796.1738,
                               42.0485,
                                         839.4514,
                                                      79.5353],
                               74.3117,
                                         722.3051,
                 [ 705.5055,
                                                     102.3828],
                 [ 868.8712,
                               32.0100,
                                         946.5863,
                                                     65.4043],
                 [ 827.4374,
                               80.7260,
                                         838.3604,
                                                     101.3921],
                 [ 477.8173,
                               19.3210.
                                         488.2294.
                                                     35.6006].
                 [ 514.8702,
                               18.9247,
                                         527.0136.
                                                      33.82351.
                 [ 557.6417,
                               19.4193,
                                         587.9046,
                                                      31.1005].
                 515.4282,
                               19.2760.
                                         526.1896.
                                                      35.8974].
                               47.7967.
                                         792.6917.
                 773.1207.
                                                      86.2238].
                 [ 777.9705,
                               23.8157,
                                         813.5848,
                                                      56.7420],
                 [ 816.8560,
                               52.9844,
                                         830.9999,
                                                     89.8972]], device='cuda:0'), 'labels': ten
         sor([0, 0, 1, 1, 0, 6, 1, 1, 5, 1, 1, 6, 1, 1, 1, 1, 6, 1, 6, 1, 1, 1, 6, 0,
                 3, 0], device='cuda:0'), 'scores': tensor([0.9992, 0.9981, 0.9974, 0.9972, 0.9
         970, 0.9969, 0.9951, 0.9930, 0.9924,
```

```
0.9920, 0.9903, 0.9899, 0.9897, 0.9854, 0.9786, 0.9756, 0.9632, 0.9591, 0.9361, 0.7834, 0.7581, 0.6922, 0.6107, 0.5900, 0.4690, 0.4012], device='cuda:0')}]
```

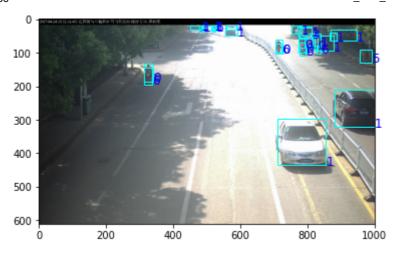
```
In [34]:
          def plot_image_from_output(img, annotation):
              img = img.cpu().permute(1,2,0) # 0위치:2에 있던게 옴, 1위치:0, 2위치:1 으로 차원
              fig, ax = plt.subplots(1)
              ax.imshow(img) # (M,N,3) : M=row, N=col
              for idx in range(len(annotation["boxes"])):
                  xmin, ymin, xmax, ymax = annotation["boxes"][idx]
                  xmin = xmin.cpu().data.numpy()
                  ymin = ymin.cpu().data.numpy()
                  xmax = xmax.cpu().data.numpy()
                  ymax = ymax.cpu().data.numpy()
                  #print("xmin,ymin,xmax,ymax :", xmin,ymin,xmax,ymax)
                  rect = patches.Rectangle((xmin,ymin),(xmax-xmin),(ymax-ymin), linewidth=1, ed
                  label = int(annotation['labels'][idx])
                  plt.text(xmax, ymax, label, color = 'blue', size = 12)
                  ax.add_patch(rect)
              plt.show()
```

```
__idx = 0
print("Target :", annotations[_idx]['labels']) # Ground Truth 값 (-1 해야됨)
plot_image_from_output(imgs[_idx], annotations[_idx]) # imgs : test_data_loader에서 t
print("Prediction :", pred[_idx]['labels'])
plot_image_from_output(imgs[_idx], pred[_idx])
```

Target: tensor([0, 0, 0, 0, 0, 1, 1, 6, 6, 6, 1, 1, 1, 1, 1, 1, 1, 6, 6, 1, 6])



Prediction: tensor([0, 0, 1, 1, 0, 6, 1, 1, 5, 1, 1, 6, 1, 1, 1, 6, 1, 6, 1, 1, 1, 6, 0, 3, 0], device='cuda:0')



In []: