8번째 미팅발표

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코멘트

- 1. Teacher model로부터 Feature 정보를 받아와서 Distillation
- 2. Generator <-> Discriminator 수정

knowledge distillation(KL), feature distillation(FT)

```
self.bce_criterion = nn.BCELoss()
self.mse_criterion=nn.MSELoss()
self.kd_criterion=DistillKL(opt)
self.ft_criterion=DistillFT(opt)
```

```
#knowledge distillation
class DistillKL(nn.Module):
    def init (self, opt):
        super(DistillKL, self). init ()
        opt = Options().parse()
        self.T = opt.temperature
        #self.alpha= opt.alpha
    def forward(self, y s, y t):
        \#B, C, H, W = y s.size()
        y s=torch.from numpy(y s)
        y t=torch.from numpy(y t)
        y_s=y_s.reshape(1,len(y_s))
        y_t=y_t.reshape(1,len(y_t))
        p s = F.log softmax(y s/self.T, dim=1)
        p t = F.softmax(y t/self.T, dim=1)
        loss = self.alpha*F.kl_div(p_s, p_t.detach(), reduction='sum') * (self.T**2) / y_s.shape[0]
        #loss= (-1)*p s*p t.detach()
        return loss
```

knowledge distillation(KL), feature distillation(FT)

```
self.bce_criterion = nn.BCELoss()
self.mse_criterion=nn.MSELoss()
self.kd_criterion=DistillKL(opt)
self.ft_criterion=DistillFT(opt)
```

```
#feature distillation
class DistillFT(nn.Module):
   def init (self, opt):
       super(DistillFT, self). init ()
       opt = Options().parse()
       self.p = 2
   def forward(self, g s, g t):
       loss = sum([self.at_loss(f_s, f_t.detach()) for f_s, f_t in zip(g_s, g_t)])
       return loss
   def at loss(self, f s, f t):
       return (self.at(f_s) - self.at(f_t)).pow(2).mean()
   def at(self, f):
       return F.normalize(f.pow(self.p).mean(1).view(f.size(0), -1))
```

Feature Distillation

* 저번 미팅시간에서 제기된 문제:

Teacher-> Student로 feature distillation이 이뤄지지 않음 (teacher model로부터 feature 정보를 받아오자)

- → Discriminator이 아닌 Generator에서 feature list를 받아오는 것으로 변경
- → Teacher model로부터 feature list 정보를 받아오자

```
class Decoder(nn.Module):
class Encoder(nn.Module):
                                                                              def __init__(self, ngpu,opt):
   def __init (self, ngpu,opt,out z):
                                                                                  super(Decoder, self).__init__()
        super(Encoder, self).__init__()
                                                                                  self.ngpu = ngpu
       self.ngpu = ngpu
                                                                                  self.conv1=nn.Sequential(
       self.conv1=nn.Sequential(
                                                                                      # input is Z, going into a convolution
                                                                                      nn.ConvTranspose1d(opt.nz,opt.ngf*16,10,1,0,bias=False),
            nn.Conv1d(opt.nc,opt.ndf,4,2,1,bias=False),
                                                                                      nn.BatchNorm1d(opt.ngf*16),
            nn.LeakyReLU(0.2, inplace=True)
                                                                                      nn.ReLU(True)
                                                                                  # state size. (ngf*16) x10
       self.conv2=nn.Sequential(
                                                                                  self.conv2=nn.Sequential(
            nn.Conv1d(opt.ndf, opt.ndf * 2, 4, 2, 1, bias=False),
                                                                                      nn.ConvTranspose1d(opt.ngf * 16, opt.ngf * 8, 4, 2, 1, bias=False),
            nn.BatchNorm1d(opt.ndf * 2),
                                                                                      nn.BatchNorm1d(opt.ngf * 8),
            nn.LeakyReLU(0.2, inplace=True)
                                                                                      nn.ReLU(True)
                                                                                  # state size. (ngf*8) x 20
                                                                                  self.conv3=nn.Sequential(
       self.conv3=nn.Sequential(
                                                                                      nn.ConvTranspose1d(opt.ngf * 8, opt.ngf * 4, 4, 2, 1, bias=False),
            nn.Conv1d(opt.ndf * 2, opt.ndf * 4, 4, 2, 1, bias=False),
                                                                                      nn.BatchNorm1d(opt.ngf * 4),
            nn.BatchNorm1d(opt.ndf * 4),
                                                                                      nn.ReLU(True)
            nn.LeakyReLU(0.2, inplace=True)
       # state size. (ndf*4) x 40
                                                                                  self.conv4=nn.Sequential(
       self.conv4=nn.Sequential(
                                                                                      nn.ConvTranspose1d(opt.ngf * 4, opt.ngf*2, 4, 2, 1, bias=False),
            nn.Conv1d(opt.ndf * 4, opt.ndf * 8, 4, 2, 1, bias=False),
                                                                                      nn.BatchNorm1d(opt.ngf*2),
            nn.BatchNorm1d(opt.ndf * 8),
                                                                                      nn.ReLU(True)
            nn.LeakyReLU(0.2, inplace=True)
                                                                                  self.conv5=nn.Sequential(
       self.conv5=nn.Sequential(
                                                                                      nn.ConvTranspose1d(opt.ngf * 2, opt.ngf , 4, 2, 1, bias=False),
            nn.Conv1d(opt.ndf * 8, opt.ndf * 16, 4, 2, 1, bias=False),
                                                                                      nn.BatchNorm1d(opt.ngf ),
           nn.BatchNorm1d(opt.ndf * 16),
                                                                                      nn.ReLU(True)
            nn.LeakyReLU(0.2, inplace=True)
                                                                                  # state size. (ngf) x 160
       # state size. (ndf*16) x 10
                                                                                  self.conv6=nn.Sequential(
       self.conv6=nn.Sequential(
                                                                                      nn.ConvTranspose1d(opt.ngf , opt.nc, 4, 2, 1, bias=False),
            nn.Conv1d(opt.ndf * 16, out z, 10, 1, 0, bias=False)
                                                                                      nn.Tanh()
```

저번 시간의 코드 Discriminator

```
class Discriminator(nn.Module):
   def init (self, opt):
        super(Discriminator, self). init ()
       model = Encoder(opt.ngpu,opt,1)
        layers = list(model.main.children())
       self.features = nn.Sequential(*layers[:-1])
       self.classifier = nn.Sequential(layers[-1])
        self.classifier.add module('Sigmoid', nn.Sigmoid())
   def forward(self, x):
       features = self.features(x)
       features = features
       classifier = self.classifier(features)
        classifier = classifier.view(-1, 1).squeeze(1)
       return classifier, features
```

```
def forward(self, x):
    #feature 1~6 (6: features)
    feat1=self.feat1(x)
    feat1=feat1
    feat2=self.feat2(x)
    feat2=feat2
    feat3=self.feat3(x)
    feat3=feat3
    feat4=self.feat4(x)
    feat4=feat4
    feat5=self.feat5(x)
    feat5=feat5
    features = self.features(x)
    features = features
    feat list=[feat1,feat2,feat3,feat4,feat5,features]
    classifier = self.classifier(features)
    classifier = classifier.view(-1, 1).squeeze(1)
    return classifier, features ,feat list
```

Discriminator 수정

저번주: feature도 차례로 받아왔고 리스트 feat_list를 정의 변경: feat_list 제거

```
class Discriminator(nn.Module):
   def init (self, opt):
        super(Discriminator, self).__init__()
       model = Encoder(opt.ngpu,opt,1)
        layers = list(model.conv1.children())
        layers.extend(list(model.conv2.children()))
        layers.extend(list(model.conv3.children()))
        layers.extend(list(model.conv4.children()))
        layers.extend(list(model.conv5.children()))
        layers.extend(list(model.conv6.children()))
        self.features = nn.Sequential(*layers[:-1])
        self.classifier = nn.Sequential(layers[-1])
        self.classifier.add_module('Sigmoid', nn.Sigmoid())
   def forward(self, x):
        features = self.features(x)
        features = features
        classifier = self.classifier(features)
        classifier = classifier.view(-1, 1).squeeze(1)
        return classifier, features
```

```
class Generator(nn.Module):

    def __init__(self, opt):
        super(Generator, self).__init__()
        self.encoder1 = Encoder(opt.ngpu,opt,opt.nz)
        self.decoder = Decoder(opt.ngpu,opt)

def forward(self, x):
    latent_i = self.encoder1(x)
    gen_x = self.decoder(latent_i)
    return gen_x, latent_i
```

Generator 수정

기존: layer을 한번에 받아와서 features 정의

변경: layer 차례로 받아오면서 feature도 차례로 받아왔고 리스트 feat_list를 정의

```
class Generator(nn.Module):
   def init (self, opt):
       super(Generator, self). init ()
       self.encoder1 = Encoder(opt.ngpu,opt,opt.nz)
       self.decoder = Decoder(opt.ngpu,opt)
       layers=list(self.encoder1.conv1.children())
       self.feat1=nn.Sequential(*layers[:-1])
        layers.extend(list(self.encoder1.conv2.children()))
       self.feat2=nn.Sequential(*layers[:-1])
        layers.extend(list(self.encoder1.conv3.children()))
       self.feat3=nn.Sequential(*layers[:-1])
        layers.extend(list(self.encoder1.conv4.children()))
       self.feat4=nn.Sequential(*layers[:-1])
        layers.extend(list(self.encoder1.conv5.children()))
       self.feat5=nn.Sequential(*layers[:-1])
        layers.extend(list(self.encoder1.conv6.children()))
       self.feat6=nn.Sequential(*layers[:-1])
```

```
def forward(self, x):
    latent i = self.encoder1(x)
    gen x = self.decoder(latent i)
    #feature 1~6 (6: features)
    feat1=self.feat1(x)
    feat1=feat1
    feat2=self.feat2(x)
    feat2=feat2
    feat3=self.feat3(x)
    feat3=feat3
    feat4=self.feat4(x)
    feat4=feat4
    feat5=self.feat5(x)
    feat5=feat5
    feat6=self.feat6(x)
   feat6=feat6
   feat list=[feat1,feat2,feat3,feat4,feat5,feat6]
    return gen_x, latent i, feat list
```

손실 함수 통합

```
① f(x)와 f(x')의 손실
def update netg(self):
                                                                      ② x와 x'의 손실
   self.G.zero grad()
                                                                      ③ knowledge distillation 손실
   self.label.data.resize_(self.opt.batchsize).fill_(self.real_label)
                                                                      ④ features 손실
   self.fake, self.latent_i, self.feat_list = self.G(self.input)
   self.out g, self.feat fake = self.D(self.fake)
   _, self.feat_real = self.D(self.input)
   # self.err g adv = self.bce criterion(self.out g, self.label) # loss for ce
1 self.err g adv=self.mse criterion(self.feat_fake,self.feat_real) # loss for feature matching
2 self.err g rec = self.mse criterion(self.fake, self.input) # constrain x' to look like x
   #self.err g = self.err g rec + self.err g adv * self.opt.w adv
   self.err g = self.err g rec + self.err g adv * self.opt.w adv
3 self.err_g+= self.kd_criterion(self.y_,self.teacher.y_pred_t) #knowledge distillation
 4 self.err g+= self.ft criterion(self.feat list, self.teacher.feat list)# feature distillation
   self.err g.backward()
   self.optimizerG.step()
```

Train함수 일부 (teacher model로부터 정보 받아오는 코드)

```
def train(self):
    self.train_hist = {}
    self.train_hist['D_loss'] = []
    self.train_hist['G_loss'] = []
    self.train_hist['per_epoch_time'] = []
    self.train_hist['total_time'] = []

self.teacher.copy() #load teacher G model and save to student path
    self.y_,self.y_pred,self.y_feat_list=self.predict(self.dataloader["train"]) ##
    self.teacher.y_t,self.teacher.y_pred_t,self.teacher_feat_list=self.teacher.predict(self.dataloader["train"]) ##
```

update_netd 함수

```
def update netd(self):
    ##
   self.D.zero grad()
   # Train with real
   self.label.data.resize (self.opt.batchsize).fill (self.real_label)
   self.out_d_real, self.feat_real = self.D(self.input)
   # Train with fake
   self.label.data.resize (self.opt.batchsize).fill (self.fake label)
   self.fake, self.latent_i, self.feat_list = self.G(self.input)
   self.out_d_fake, self.feat_fake = self.D(self.fake)
   self.err_d_real = self.bce_criterion(self.out_d_real, torch.full((self.batchsize,), self.real_label, device=self.device,dtype=torch.float))
   self.err d fake = self.bce criterion(self.out d fake, torch.full((self.batchsize,), self.fake label, device=self.device,dtype=torch.float))
   self.err d=self.err d real+self.err d fake
   self.err d.backward()
   self.optimizerD.step()
```

train/eval teacher model & train/eval student model

predict 함수 (y_, y_pred, y_feat_list 리턴)

```
def predict(self,dataloader ,scale=True):
   with torch.no grad():
       self.an scores = torch.zeros(size=(len(dataloader .dataset),), dtype=torch.float32, device=self.device)
       self.gt labels = torch.zeros(size=(len(dataloader .dataset),), dtype=torch.long, device=self.device)
       self.latent i = torch.zeros(size=(len(dataloader .dataset), self.opt.nz), dtype=torch.float32, device=self.device)
       self.dis feat = torch.zeros(size=(len(dataloader .dataset), self.opt.ndf*16*10), dtype=torch.float32,
                                   device=self.device)
       self.feat list = torch.zeros(size=(len(dataloader .dataset), self.opt.ndf*16*10), dtype=torch.float32,
                                   device=self.device)
       for i, data in enumerate(dataloader , 0):
           self.set input(data)
           self.fake, latent_i, self.feat_list = self.G(self.input)
           # error = torch.mean(torch.pow((d feat.view(self.input.shape[0],-1)-d gen feat.view(self.input.shape[0],-1)), 2), dim=1)
           error = torch.mean(
               torch.pow((self.input.view(self.input.shape[0], -1) - self.fake.view(self.fake.shape[0], -1)), 2),
               dim=1)
           self.an scores[i*self.opt.batchsize : i*self.opt.batchsize+error.size(0)] = error.reshape(error.size(0))
           self.gt labels[i*self.opt.batchsize : i*self.opt.batchsize+error.size(0)] = self.gt.reshape(error.size(0))
           self.latent i [i*self.opt.batchsize : i*self.opt.batchsize+error.size(0), :] = latent i.reshape(error.size(0), self.opt.nz)
```

predict 함수 (y_, y_pred, y_feat_list 리턴)

```
def predict(self,dataloader_,scale=True):
   with torch.no grad():
       self.an scores = torch.zeros(size=(len(dataloader .dataset),), dtype=torch.float32, device=self.device)
       self.gt_labels = torch.zeros(size=(len(dataloader_.dataset),), dtype=torch.long,
                                                                                           device=self.device)
       self.latent_i = torch.zeros(size=(len(dataloader_.dataset), self.opt.nz), dtype=torch.float32, device=self.device)
       self.dis feat = torch.zeros(size=(len(dataloader .dataset), self.opt.ndf*16*10), dtype=torch.float32,
                                   device=self.device)
       self.feat list = torch.zeros(size=(len(dataloader .dataset), self.opt.ndf*16*10), dtype=torch.float32,
                                   device=self.device)
       for i, data in enumerate(dataloader , 0): ...
       # Scale error vector between [0, 1]
       if scale:
           self.an scores = (self.an scores - torch.min(self.an scores)) / (torch.max(self.an scores) - torch.min(self.an scores))
       y_=self.gt_labels.cpu().numpy()
       y pred=self.an scores.cpu().numpy()
       y feat list=self.feat list
       return y_,y_pred,y_feat_list
```