FusionDeepCNNTest

February 26, 2023

```
[1]: %reload ext autoreload
     %autoreload 2
     %matplotlib inline
     import torch
     import numpy as numpy
     import cv2
     import PIL
     from PIL import Image
     import numpy as np
     import matplotlib.pyplot as plt
     from tqdm import tqdm
     import matplotlib.cm as cm
     import torch.nn as nn
     import torchvision.transforms as transforms
     import imageio
     from torch.autograd import Variable
```

```
nn.Conv2d(1, 64, kernel_size=3, padding=1),
    nn.BatchNorm2d(64),
    nn.LeakyReLU(0.1, inplace=True),
)
self.conv2_1 = nn.Sequential(
    nn.Conv2d(64, 128, kernel_size=3, padding=1),
    nn.BatchNorm2d(128),
    nn.LeakyReLU(0.1, inplace=True),
    nn.MaxPool2d(2)
)
self.conv2_2 = nn.Sequential(
    nn.Conv2d(64, 128, kernel_size=3, padding=1),
    nn.BatchNorm2d(128),
    nn.LeakyReLU(0.1, inplace=True),
    nn.MaxPool2d(2)
)
self.conv2_3 = nn.Sequential(
    nn.Conv2d(64, 128, kernel_size=3, padding=1),
    nn.BatchNorm2d(128),
    nn.LeakyReLU(0.1, inplace=True),
    nn.MaxPool2d(2)
)
self.conv3_1 = nn.Sequential(
    nn.Conv2d(128, 128, kernel_size=3, padding=1),
    nn.BatchNorm2d(128),
    nn.LeakyReLU(0.1, inplace=True),
    nn.MaxPool2d(2)
)
self.conv3_2 = nn.Sequential(
    nn.Conv2d(128, 128, kernel_size=3, padding=1),
    nn.BatchNorm2d(128),
    nn.LeakyReLU(0.1, inplace=True),
    nn.MaxPool2d(2)
)
self.conv3_3 = nn.Sequential(
    nn.Conv2d(128, 128, kernel_size=3, padding=1),
    nn.BatchNorm2d(128),
    nn.LeakyReLU(0.1, inplace=True),
    nn.MaxPool2d(2)
```

```
self.conv4 = nn.Sequential(
        nn.Conv2d(128, 256, kernel_size=3, padding=1),
        nn.BatchNorm2d(256),
        nn.LeakyReLU(0.1, inplace=True),
        nn.MaxPool2d(2)
    )
    self.conv5 = nn.Sequential(
        nn.Conv2d(128*2, 256, kernel_size=3, padding=1),
        nn.BatchNorm2d(256),
        nn.LeakyReLU(0.1, inplace=True),
        nn.MaxPool2d(2)
    )
    self.fc1 = nn.Linear(256*8*4*2, 2)
def forward(self, x, y, z):
    outx = self.conv1_1(x)
    outx = self.conv2_1(outx)
    outx = self.conv3_1(outx)
    outx = self.conv4(outx)
    outx = outx.view(outx.size(0), -1)
    outy = self.conv1_2(y)
    outy = self.conv2_2(outy)
    outy = self.conv3_2(outy)
    outz = self.conv1_3(z)
    outz = self.conv2_3(outz)
    outz = self.conv3_3(outz)
    oyz=torch.cat([outy,outz],1)
    oyz = self.conv5(oyz)
    oyz = oyz.view(oyz.size(0), -1)
```

```
oo=torch.cat([outx,oyz],1)
             out = self.fc1(oo)
             return out
[3]: model=CNN()
     model
[3]: CNN(
       (conv1_1): Sequential(
         (0): Conv2d(1, 64, kernel size=(3, 3), stride=(1, 1), padding=(1, 1))
         (1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
     track_running_stats=True)
         (2): LeakyReLU(negative_slope=0.1, inplace=True)
       (conv1_2): Sequential(
         (0): Conv2d(1, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
         (1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
     track_running_stats=True)
         (2): LeakyReLU(negative_slope=0.1, inplace=True)
       (conv1 3): Sequential(
         (0): Conv2d(1, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
         (1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
     track_running_stats=True)
         (2): LeakyReLU(negative_slope=0.1, inplace=True)
       (conv2_1): Sequential(
         (0): Conv2d(64, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
         (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
     track_running_stats=True)
         (2): LeakyReLU(negative_slope=0.1, inplace=True)
         (3): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
     ceil_mode=False)
       (conv2_2): Sequential(
         (0): Conv2d(64, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
         (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
     track running stats=True)
         (2): LeakyReLU(negative_slope=0.1, inplace=True)
         (3): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
     ceil_mode=False)
       )
       (conv2_3): Sequential(
         (0): Conv2d(64, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
```

```
(1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (2): LeakyReLU(negative_slope=0.1, inplace=True)
    (3): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
ceil_mode=False)
  (conv3 1): Sequential(
    (0): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
    (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (2): LeakyReLU(negative slope=0.1, inplace=True)
    (3): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
ceil_mode=False)
 )
  (conv3 2): Sequential(
    (0): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
    (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (2): LeakyReLU(negative_slope=0.1, inplace=True)
    (3): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
ceil_mode=False)
  )
  (conv3_3): Sequential(
    (0): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1), padding=(1, 1))
    (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (2): LeakyReLU(negative_slope=0.1, inplace=True)
    (3): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1,
ceil_mode=False)
 )
  (conv4): Sequential(
    (0): Conv2d(128, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
    (1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (2): LeakyReLU(negative_slope=0.1, inplace=True)
    (3): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
ceil mode=False)
  (conv5): Sequential(
    (0): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
    (1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (2): LeakyReLU(negative slope=0.1, inplace=True)
    (3): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
ceil_mode=False)
  (fc1): Linear(in_features=16384, out_features=2, bias=True)
```

```
[8]: model_path='ECNN_network_wights_1.pth'
      use_gpu=torch.cuda.is_available()
      if use_gpu:
          print('GPU Mode Acitavted')
          model = model.cuda()
          model.cuda()
          model = torch.nn.DataParallel(model, device_ids=range(torch.cuda.

device count()))
          model.load_state_dict(torch.load(model_path), strict=False)
      else:
          print('CPU Mode Acitavted')
          state_dict = torch.load(model_path,map_location='cpu')
          # create new OrderedDict that does not contain `module.`
          from collections import OrderedDict
          new_state_dict = OrderedDict()
          for k, v in state_dict.items():
              name = k[7:] # remove `module.`
              new_state_dict[name] = v
          # load params
          model.load_state_dict(new_state_dict, strict=False)
     CPU Mode Acitavted
 [9]: def to_var(x, volatile=False):
          if torch.cuda.is_available():
              x = x.cuda()
          return Variable(x, volatile=volatile)
[26]: original_path1= 'data/bunker_r.bmp'
      original_path2= 'data/IR_bunker_g.bmp'
 []: tfms1 = transforms.Compose([
          transforms.Resize((64, 32)),
          transforms.ToTensor(),
          transforms.Normalize([0.45], [0.1])
      ])
      tfms2 = transforms.Compose([
          transforms.Resize((64, 32)),
          transforms.ToTensor(),
```

)

```
transforms.Normalize([ 0.050], [ 0.09])
])
tfms3 = transforms.Compose([
   transforms.Resize((64, 32)),
   transforms.ToTensor(),
   transforms.Normalize([0.06], [ 0.09])
])
img1_org = Image.open(original_path1)
img2_org = Image.open(original_path2)
img1_org = np.asarray(img1_org)
img2_org = np.asarray(img2_org)
height=img1_org.shape[0]
width=img2_org.shape[1]
windows_size=32
# stride can be set as 2 or 4 or 8 based on the size of input images
if width>= 500 and height>=500:
   factor=1
   stride=4
else:
   factor=2
   stride=8
dim1=(width, height)
dim2 = (int(width*factor), int(height*factor))
img1 = cv2.resize(img1_org, dim2, interpolation = cv2.INTER_AREA)
img2 = cv2.resize(img2_org, dim2, interpolation = cv2.INTER_AREA)
kernel=np.array([[-1, -2, -1], [0,
                                           0 , 0], [1 , 2, 1]])
img1_gray =img1 #for rqb convert gray
img1_GY = cv2.filter2D(img1_gray,-1,kernel)
img1_GX = cv2.filter2D(img1_gray,-1,np.transpose(kernel))
img2_gray = img2 #for rgb convert gray
img2_GY = cv2.filter2D(img2_gray,-1,kernel)
img2_GX = cv2.filter2D(img2_gray,-1,np.transpose(kernel))
test image1 1=img1 gray
test_image1_2=img1_GX
test_image1_3=img1_GY
test_image2_1=img2_gray
test_image2_2=img2_GX
test_image2_3=img2_GY
```

```
source1=img1
source2=img2
j=0
MAP=np.zeros([img1.shape[0], img1.shape[1]])
score1=0
score2=0
FUSED=np.zeros(test_image1_1.shape)
windowsize_r = windows_size_1
windowsize_c = windows_size-1
map1=np.zeros([img1.shape[0], img1.shape[1]])
map2=np.zeros([img2.shape[0], img2.shape[1]])
for r in tqdm(range(0,img1.shape[0] - windowsize_r, stride)):
   for c in range(0,img1.shape[1] - windowsize_c, stride):
       block_test1_1 = test_image1_1[r:r+windowsize_r+1,c:c+windowsize_c+1]
       block test1 2 = test image1 2[r:r+windowsize r+1,c:c+windowsize c+1]
        block_test1_3 = test_image1_3[r:r+windowsize_r+1,c:c+windowsize_c+1]
       block_test2_1 = test_image2_1[r:r+windowsize_r+1,c:c+windowsize_c+1]
       block_test2_2 = test_image2_2[r:r+windowsize_r+1,c:c+windowsize_c+1]
       block_test2_3 = test_image2_3[r:r+windowsize_r+1,c:c+windowsize_c+1]
       block1_1= np.concatenate((block_test1_1, block_test2_1), axis=0)
       block2_1= np.concatenate((block_test2_1, block_test1_1), axis=0)
       block1_1 = Image.fromarray(block1_1, 'L')
       block2_1 = Image.fromarray(block2_1, 'L')
       block1_2= np.concatenate((block_test1_2, block_test2_2), axis=0)
       block2_2= np.concatenate((block_test2_2, block_test1_2), axis=0)
       block1 2 = Image.fromarray(block1 2, 'L')
       block2_2 = Image.fromarray(block2_2, 'L')
       block1_3= np.concatenate((block_test1_3, block_test2_3), axis=0)
        block2_3= np.concatenate((block_test2_3, block_test1_3), axis=0)
        block1 3 = Image.fromarray(block1 3, 'L')
       block2_3 = Image.fromarray(block2_3, 'L')
        imout1_1=tfms1(block1_1)
        imout2_1=tfms1(block2_1)
        imout1_2=tfms2(block1_2)
        imout2_2=tfms2(block2_2)
```

```
imout1_3=tfms3(block1_3)
    imout2_3=tfms3(block2_3)
    if use_gpu:
        imout1_1=to_var(imout1_1)
        imout2_1=to_var(imout2_1)
        imout1_2=to_var(imout1_2)
        imout2_2=to_var(imout2_2)
        imout1_3=to_var(imout1_3)
        imout2_3=to_var(imout2_3)
    imout1_1=(imout1_1)
    imout2_1=(imout2_1)
    imout1_2=(imout1_2)
    imout2_2=(imout2_2)
    imout1_3=(imout1_3)
    imout2_3=(imout2_3)
    inputs1_1 = imout1_1.unsqueeze(0)
    inputs2_1 = imout2_1.unsqueeze(0)
    inputs1_2 = imout1_2.unsqueeze(0)
    inputs2_2 = imout2_2.unsqueeze(0)
    inputs1 3 = imout1 3.unsqueeze(0)
    inputs2_3 = imout2_3.unsqueeze(0)
    model.eval()
    outputs1 = model(inputs1_1,inputs1_2,inputs1_3)
    _, predicted1 = torch.max(outputs1.data, 1)
    score1=predicted1.detach().cpu().numpy()
    model.eval()
    outputs2 = model(inputs2_1,inputs2_2,inputs2_3)
    _, predicted2 = torch.max(outputs2.data, 1)
    score2=predicted2.detach().cpu().numpy()
    map2[r:r+windowsize_r+1,c:c+windowsize_c+1] += 1
    if score1 <= score2:</pre>
        map1[r:r+windowsize_r+1,c:c+windowsize_c+1] += +1
else:
        map1[r:r+windowsize_r+1,c:c+windowsize_c+1] += -1
```

```
map1 = cv2.resize(map1, dim1, interpolation = cv2.INTER_AREA)
     test_image1 = img1_org
     test_image2 = img2_org
     map3=np.zeros([img1_org.shape[0], img2_org.shape[1]])
     FUSED=np.zeros(img1_org.shape)
     for r in range(0,img1_org.shape[0], 1):
         for c in range(0,img1_org.shape[1], 1):
             if map1[r,c] < 0:</pre>
                 map3[r,c] = 0
                 FUSED[r,c]=img2_org[r,c]
             else:
                 map3[r,c] = 1
                 FUSED[r,c]=img1_org[r,c]
     FUSED_8=FUSED.astype(np.uint8)
     plt.imshow(map3, cm.gray)
     plt.show()
     plt.imshow(FUSED_8)
     plt.show()
     imageio.imwrite('./output.tif', FUSED_8)
     imageio.imwrite('./output.jpg', FUSED_8)
     32%|
    | 44/137 [05:16<14:08, 9.12s/it]
[]: plt.imshow(img1, cmap='gray')
    plt.show()
     plt.imshow(img1, cmap='gray')
     plt.show()
     plt.imshow(FUSED_8)
     plt.show()
[]:
```