

MultifocusRGBFusionDeepCNNTTest

February 27, 2023

```
[82]: %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
from collections import OrderedDict

import warnings
warnings.filterwarnings('ignore')
```

```
[83]: #Module-----
class CNN(nn.Module):

    def __init__(self):
        super(CNN, self).__init__()

        self.conv1_1 = nn.Sequential(
            nn.Conv2d(1, 64, kernel_size=3, padding=1),
            nn.BatchNorm2d(64),

            nn.LeakyReLU(0.1, inplace=True),

        )

        self.conv1_2 = nn.Sequential(
            nn.Conv2d(1, 64, kernel_size=3, padding=1),
            nn.BatchNorm2d(64),
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        nn.LeakyReLU(0.1, inplace=True),
    )

    self.conv1_3 = nn.Sequential(
        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)
    )

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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)

#forward
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)

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        oo=torch.cat([outx,oyz],1)

        out = self.fc1(oo)

        return out

```

```

[84]: #model
model=CNN()
model

```

```

[84]: 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))

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        (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)

```

)

```
[85]: #Model path
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')
    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

```
[86]: def to_var(x, volatile=False):
        if torch.cuda.is_available():
            x = x.cuda()
        return Variable(x, volatile=volatile)
```

```
[87]: #data path
original_path1= 'data/nf.jpg'
original_path2= 'data/ff.jpg'
```

```
[88]: #data augmentation
tfms1 = transforms.Compose([
    transforms.Resize((64, 32)),
    transforms.ToTensor(),
    transforms.Normalize([0.45 ], [0.1])
])

tfms2 = transforms.Compose([
    transforms.Resize((64, 32)),
    transforms.ToTensor(),
```

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        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 should 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]])
    #for rgb convert to gray:cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
    img1_gray =cv2.cvtColor(img1, cv2.COLOR_BGR2GRAY)
    img1_GY = cv2.filter2D(img1_gray,-1,kernel)
    img1_GX = cv2.filter2D(img1_gray,-1,np.transpose(kernel))

    #for rgb convert to gray: cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
    img2_gray = cv2.cvtColor(img2, cv2.COLOR_BGR2GRAY)
    img2_GY = cv2.filter2D(img2_gray,-1,kernel)
    img2_GX = cv2.filter2D(img2_gray,-1,np.transpose(kernel))

    #gradient estimation
    test_image1_1=img1_gray
    test_image1_2=img1_GX
    test_image1_3=img1_GY

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test_image2_1=img2_gray
test_image2_2=img2_GX
test_image2_3=img2_GY

#map generation
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)

window_size_r = window_size-1
window_size_c = window_size-1

map1=np.zeros([img1.shape[0], img1.shape[1]])
map2=np.zeros([img2.shape[0], img2.shape[1]])

for r in tqdmm(range(0,img1.shape[0] - window_size_r, stride)):
    for c in range(0,img1.shape[1] - window_size_c, stride):

        block_test1_1 = test_image1_1[r:r+window_size_r+1,c:c+window_size_c+1]
        block_test1_2 = test_image1_2[r:r+window_size_r+1,c:c+window_size_c+1]
        block_test1_3 = test_image1_3[r:r+window_size_r+1,c:c+window_size_c+1]

        block_test2_1 = test_image2_1[r:r+window_size_r+1,c:c+window_size_c+1]
        block_test2_2 = test_image2_2[r:r+window_size_r+1,c:c+window_size_c+1]
        block_test2_3 = test_image2_3[r:r+window_size_r+1,c:c+window_size_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)

#modelevaluation
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+window_size_r+1,c:c+window_size_c+1] += 1

if score1 <= score2:
    map1[r:r+window_size_r+1,c:c+window_size_c+1] += +1
else:

```

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        map1[r:r+window_size_r+1,c:c+window_size_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:
            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)

```

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```

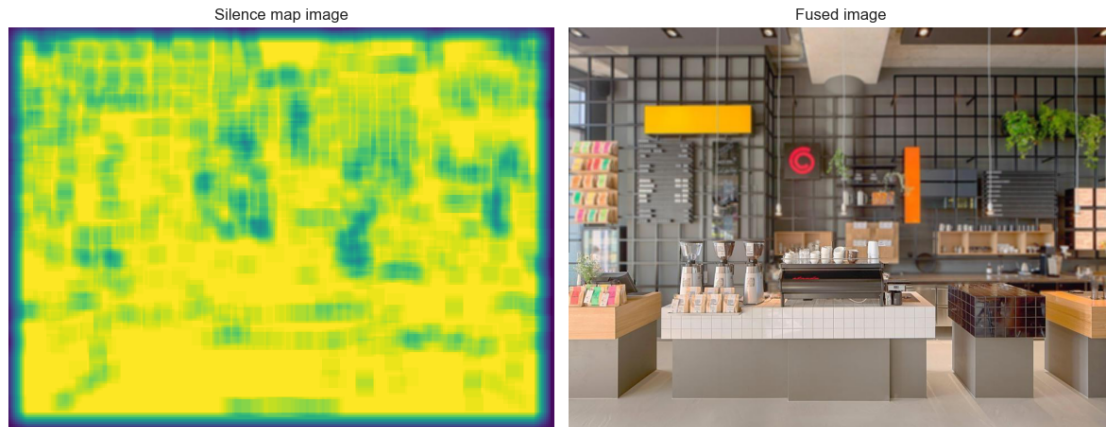
[99]: #Show output
plt.rcParams['savefig.dpi'] = 300
fig, (ax0, ax1) = plt.subplots(ncols=2, figsize=(12,10))

ax0.imshow(map1, cmap='viridis')
ax0.set_title("Silence map image")
ax0.axis('off')

ax1.imshow(FUSED_8)
ax1.set_title("Fused image")
ax1.axis('off')

plt.subplots_adjust(wspace=0, hspace=0)
fig.tight_layout()
#save output
imageio.imwrite('data/output.tif', FUSED_8)
imageio.imwrite('data/output.jpg', FUSED_8)

```



```
[101]: #Fused results
plt.style.use('seaborn-white')
plt.rcParams['savefig.dpi'] = 300
fig, (ax0, ax1, ax2) = plt.subplots(ncols=3, figsize=(14, 12))

ax0.imshow(img1, cmap='brg')
ax0.set_title("Near focus image")
ax0.axis('off')

ax1.imshow(img2, cmap='brg')
ax1.set_title("Far focus image")
ax1.axis('off')

ax2.imshow(FUSED_8, cmap='brg')
ax2.set_title("Fused image")
ax2.axis('off')

plt.subplots_adjust(wspace=0, hspace=0)
fig.tight_layout()
plt.show();
```

