Project 2 – Report

Project 2 is to perform image blending from using Gaussian/Laplacian pyramid computation technique gPyr, lPyr = ComputePyr(input_image, num_layers) where input_image is an input image in grayscale or RGG and num_layers is the number of Gaussian/Laplacian pyramid layers to be computed. As a part of implementing the Gaussian/Laplacian pyramid, a smoothing function, a down sampling function, and a up sampling function need to be implemented as well although OpenCV built-in functions can be used as well. For the smoothing function, built-in functions can be used to generate the Gaussian kernel but the convolution/FFT function from project 1 must be used. A simple GUI is also required to create a black and white binary mask image. The GUI can open the foreground image to select a region of interest (ROI) in rectangular shape, elliptical shape, or polygon (free form) region. As a result, three blended images are generated from three image pairs of two, total of six images (pair A, pair B, and pair C).

The entire code for this project is broken up into three files: main.py, P2_Packages.py, and P1_q1.py. Main.py contains the main code to load input images and display several images including foreground image with marked mask, mask, background image, and background image. It also contains the GUI to draw and create the mask from the foreground image. P2_Pakages.py contains functions to align images, create mask pyramid, downscale/upscale images, and blend images. This file is used as a module that the main file imports the above functions. P1_q1.py contains the convolution and the padding function implemented in project 1. This file is used as a module that the P2_Packages file imports its functions.

```
# true if mouse is pressed

# true if mouse is pressed

# true if mouse is pressed

# if True, draw rectangle, press 'e' to toggle to draw ellipse

# initial x-y position of mouse

# falm = cv2.resize(cv2.imread('lc1.jpg'), (940,620)) # background img as RGB, optional: grasycale

# falm = cv2.resize(cv2.imread('lc2.jpg'), (940,620)) # background img as RGB, optional: grasycale

# initial x-y position of mouse

# falm = cv2.resize(cv2.imread('lc2.jpg'), (940,620)) # background img as RGB, optional: grasycale

# initial x-y position of mouse

# falm = cv2.resize(cv2.imread('lc1.jpg'), (940,620)) # background img as RGB, optional: grasycale

# initial x-y position of mouse

# img as RGB, optional: grasycale

# img as RGB, optional: grasycale

# img as RGB, optional: grasycale

## initial x-y position of mouse

## initial x-y position of mouse

## initial x-y position of mouse

## copy of aligned img as RGB, optional: grasycale

## copy of aligned img as RGB, optional: grasycale

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## copy of aligned img as RGB, optional: gras
```

Figure 1. Code Snip 1, Main.py.

Figure 1 shows some initial setups for the GUI's logics to draw the mask. The original x-y coordinate of the mouse is set to (-1, -1). Once the user starts drawing by pressing left button, the new starting x-y coordinate of the mouse on the image is assigned to the original x-y coordinate (line 59-75, figure 2). Two images are read as grayscale, RGB, or both. The images are resized to be less than 1000 pixels in either width or length, typically about 500x500 in size for pair A and pair B, for faster processing and for the convolution function to work properly with a padding width of two. Pair C is about 940x620. Most tested images are resized to be square images and are even in both width and length (both foreground and background images) before being blended, but the code should work for non-square images (for example, pair C). The foreground image is manually aligned if needed with the background image if its size is smaller than the background image's.

```
def create_mask_fimg(f_imgcp, init_x, init_y, new_x, new_y):
    new_mask = np.zeros(f_imgcp.shape).astype(np.float32)
       new_mask = cv2.ellipse(new_mask, ((init_x+(new_x-init_x)//2),(init_y+(new_y-init_y)//2)), (new_x-init_x,new_y-init_y)
       new_mask = cv2.rectangle(new_mask, (init_x,init_y),(new_x,new_y), (1,1,1), -1)
   return new_mask
                                                    # return new mask to main code
""" function to draw mask on displayed GUI """

def get_mask(event,x,y,f,param):
                                                    # openCV defaut parameters
    global init_x, init_y, new_x, new_y, mouse_pressed # these vars can be used outside of this function
    if event == cv2.EVENT_LBUTTONDOWN:
                                                    # left mouse button is pressed
       init_x,init_y = x,y
   elif event == cv2.EVENT_LBUTTONUP:
           cv2.ellipse(f_imgcp, ((init_x+(x-init_x)//2),(init_y+(y-init_y)//2)), (x-init_x,y-init_y), 0,0,360, (0,0,0), 1)
           cv2.rectangle(f_imgcp, (init_x,init_y), (x,y), (0,0,0), 1) #show drawn shape
        new_y = y;
```

Figure 2. Code Snip 2, Main.py.

Get_mask and **create_mask_fimg** are called from the main code with in **main.py** to display the GUI for drawing the mask's shape (or ROI) and create the mask from ROI. In **get_mask**, after the user release the mouse [left] button, the ending x-y coordinate is assigned to new_x and new_y. Then, init_x (see figure 1), init_y (line 91, figure 1), new_x, and new_y are fed to **create_mask_fimg**. Using OpenCV's built-in functions, a mask is created in either rectangular shape or elliptical shape. The user has the option to press "e" to toggle to ellipse from the default rectangle.

```
""" display f_img, b_img, and mask """

cv2.namedkindow('Foreground Image', cv2.WINDOW_NORMAL)

cv2.namedkindow('Mask', cv2.WINDOW_NORMAL)

cv2.namedkindow('Mask', cv2.WINDOW_NORMAL)

cv2.namedkindow('Background Image', cv2.WINDOM_NORMAL)

cv2.namedkindow('Background Image', b_img)

cv2.waltkey('0)

# after dislaying, press any key to close windows

cv2.waltkey('0)

# after dislaying, press any key to close windows

cv2.waltkey('0)

# after dislaying, press any key to close windows

cv2.waltkey('0)

# after dislaying, press any key to close windows

""" compute Gaussian/Laplacian pyramid, create Gaussian mask, and blend images

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""" tempto lime, lbyr_fimg = ComputePyr(f_img, 18)

gPyr_bimg, lbyr_bimg = ComputePyr(f_img, 18)

gPyr_bimg, lbyr_bimg = ComputePyr(f_img, 18)

gPyr_mask = create_gPyr_mask(mask,lan(gPyr_fimg))

blended_img = imgBlending(lPyr_fimg, lPyr_bimg, gPyr_mask, len(gPyr_fimg))

""" testing - print to see images """

# testing - print to see images """

# cv2.immrite('gPyr_fimg(i).png'.format(i-i), gPyr_fimg[i])

# cv2.immrite('gPyr_fimg(i).png'.format(i-i), gPyr_bimg[i])

# cv2.immrite('gPyr_bimg(i).png'.format(i-i), gPyr_mask[i])

# cv2.immrite('gPyr_mask(i).png'.format(i-i), gPyr_mask[i])

# cv3.immrite('gPyr_mask(i).png'.format(i-i), gPyr_mask[i])

# cv3.i
```

Figure 3. Code Snip 3, Main.py.

After being pre-processed, the two input images are sent to **P2_Packges.py** to compute their Gaussian/Laplacian pyramids by calling **ComputePyr**. The recently created mask is also sent to compute its pyramid. The number of layers (line 138 and 139, figure 3) are entered manually, in this case is 10 layers. However, the actual possible (maximum and minimum) layers computed can be less than the desired number of layers entered depending on the images' sizes and the kernel's size respectively (line 151-158, figure 4).

```
""" function compute Gaussian/Laplacian pyramid """

def ComputePyr(img, num_layers):

global pyr_layers, w  # minimum layers should be 1 and w is Gaussian kernel

pyr_layers = 1

shape = img.shape[0]  # since we're using square images, use either shape is alright

for i in range(num_layers-1):
    if (shape//2 < 5):  # floor division, cannot be less than kernel size 5x5

print("Moximum no. of layers computed from the image is %d "%pyr_layers)

inum_layers = pyr_layers  # num_layers = last increment of pyr_layers in else statement

break else:
    shape = shape // 2  # as long as shape is floored by two and > 5, increment no. layers in the

g kernel = cv2.getGaussianKernel(5,2)  # 1-d Gaussian array, std deviation is 2

w = g_kernel*(g_kernel.T)  # create 2-d gaussian filter

""" Gaussian pyramid """
    g = img.copy().astype(np.float32)
    g_pyr_img = [g]  # first layer of the Gaussian pyr = original img

for i in range(num_layers-1):
    g = downSampler(conv2(g,w))  # pyramid of smaller images - call downSampler, convolve then downscale

g = downSampler(conv2(g,w))  # pyramid of saussian images

""" Laplacian pyramid """

l.pyr_img = [g_pyr_img[num_layers-1]]  # first layer of Laplacian pyr = last layer in Gaussian list of images

for i in range(num_layers-1]  # first layer of Laplacian pyr = last layer in Gaussian list of images

for i in range(num_layers-1]  # first layer of Laplacian pyr = last layer in Gaussian list of images

for i in range(num_layers-1]  # first layer of Laplacian pyr = last layer in Gaussian list of images

for i in range(num_layers-1]  # first layer of Laplacian pyr = last layer in Gaussian list of images

for i in range(num_layers-1]  # first layer of Laplacian images

"" Lapr_img = [g_pyr_img[i-1], G)

l.pyr_img, append()  # add layers to list of Laplacian images

return g_pyr_img, l_pyr_img  # return to main code
```

Figure 4. Code Snip 4, P2_Packages.py.

A Gaussian kernel is obtained by generating a 1-D Gaussian array multiplied by its own transpose using **cv2.getGaussianKernel**. Line 160-176 in figure 4 shows the logics in computing the Gaussian and Laplacian pyramid following the diagram shown in figure 5.

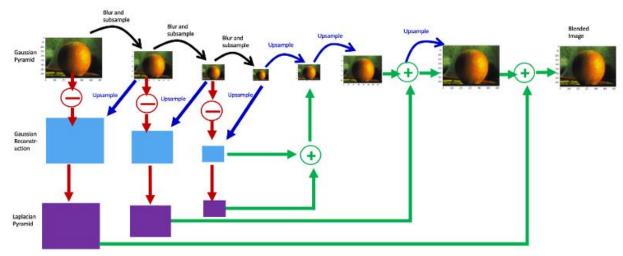


Figure 5. Gaussian/Laplacian Pyramid Logic [1].

The up-sampling and down-sampling function are implemented by referring to the built-in OpenCV functions. These functions are called upSampler and downSampler and are shown in figure 6.

```
### function to downscale ing using mearest neighbor interpolation """

### function to downscale ing using mearest neighbor interpolation """

### function to downscale ing using nearest neighbor interpolation """

### function to downscale ing using nearest neighbor interpolation """

### function to downscale ing using nearest neighbor interpolation """

### function to downscale ing using nearest neighbor interpolation """

### function to downscale ing using nearest neighbor interpolation """

### function to downscale ing using nearest neighbor interpolation """

### function to downscale ing using nearest neighbor interpolation """

#### def downSampler(ing):

### ago as ing.shape[0]

### ago as ing.shape
```

Figure 6. Code Snip 5, P2_Packages.py.

The 2-D convolution function used in this project has been implemented in project 1 since the computation is performed in the spatial domain. Though the computation in the spatial domain is not as fast as in the frequency domain, it works well as expected. "Reflect-acrossedge" padding type is used for all tested images in this project. By the end of **ComputePyr**, two pyramids are obtained each for the foreground image and the background image: the Gaussian pyramid that contains a list of bigger images to smaller images with the first image being the original image and the Laplacian pyramid that contains a list of smaller images to bigger images with the first image is the last image in the Gaussian pyramid.

In the meantime, the created mask in line 123 of figure 1 is also sent to **create_gPyr_mask** to compute its Gaussian pyramid list of images (bigger to smaller). The list is then reversed to smaller images to bigger images, so that the list can be applied to the Laplacian list of images for blending images.

```
""" function to blend images """
def imgBlending(l_fimg, l_bimg, gMask, layers):
     """ blend images """
    LS = []
for la,lb,mask in zip(l_fimg,l_bimg,gMask):
    ls = la * mask + lb * (1 - mask)
    LS.append(np.float32(ls))
     """ reconstruct final image """
     lap_bl = LS[0]
for i in range(1,layers):
    lap_bl = conv2(upSampler(lap_bl,LS[i]),w)
    lap_bl = cv2.add(lap_bl,LS[i])
     final = np.clip(lap_bl,0,255).astype(np.uint8)
     return final
""" function to create Gaussian pyramid for mask """
def create_gPyr_mask(mask,num_layers):
     gMask = np.copy(mask).astype(np.float32)
g_pyr_mask = [gMask]
                                                                           # create list of masks
     for i in range(num_layers-1):
                                                                          # pyramid of smaller masks - call downSampler, convolve then downscale # add layers to list of Gaussian masks
          gMask = downSampler(conv2(gMask,w))
g_pyr_mask.append(gMask)
     g_pyr_mask.reverse()
     return g_pyr_mask
```

Figure 7. Code Snip 6, P2_Packages.py.

Upon obtaining the Laplacian pyramids of the foreground image, background image, and the reversed Gaussian pyramid of the mask, the three pyramids are sent to **imgBlending** to blend and reconstruct the final blended image, which is returned to the main code for displaying. Three pairs of two images are blended as required and shown below.



Figure 8. Blended Image from Pair A.

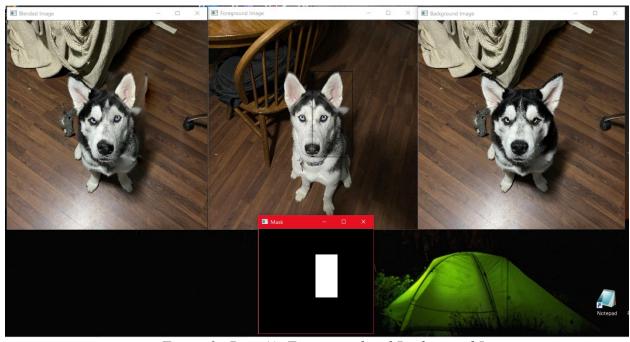


Figure 9. Pair A's Foreground and Background Image.



Figure 10. Blended Image from Pair B.

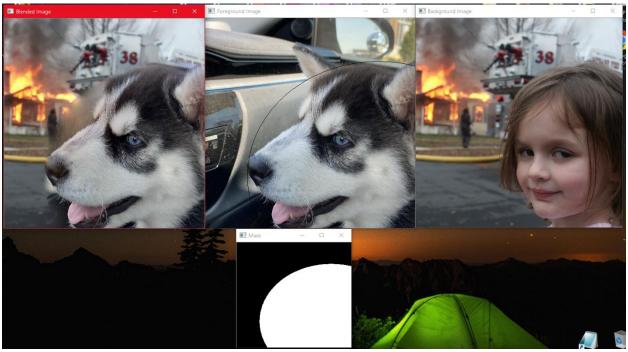


Figure 11. Pair B's Foreground and Background Image.



Figure 12. Blended Image from Pair C.

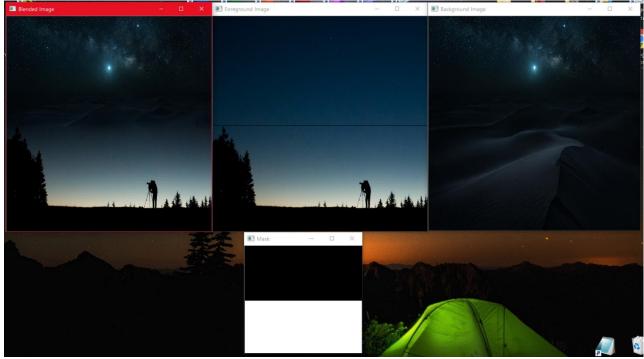


Figure 13. Pair C's Foreground and Background Image.

Attached to this report:

- 1) kddo_code
 - main.py
 - P1_q1.py
 - P2_Packages.py
- 2) kddo_images

Pair A

- Original Images
 - o bella,jpg
 - o lucas.jpg
- Result Images
 - o b_img.png
 - o bl_img.png
 - o f_imgcp.png
 - o sum1.png

Pair B

- Original Images
 - o dgirl,jpg
 - o lucasface.jpg
- Result Images
 - o b_img.png
 - o bl_img.png
 - o f_imgcp.png
 - o sum2.png

Pair C

- Original Images
 - o lc1,jpg
 - o lc2.jpg
- Result Images
 - o b_img.png
 - o bl_img.png
 - o f_imgcp.png
 - o sum3.png

Reference

- [1] M. Zhao, "Image Blending Using Laplacian Pyramids," Becominghuman.ai, 14-May-2020. [Online]. Available: https://becominghuman.ai/image-blending-using-laplacian-pyramids-2f8e9982077f. [Accessed: 28-Oct-2021].
- [2] F. Projcheski, "How to Blend Images Using Gaussian and Laplacian Pyramid," Laconicml.com, 25-Jul-2020. [Online]. Available: https://laconicml.com/blend-images-laplacian-pyramid/. [Accessed: 28-Oct-2021].
- [3] I. Nader, "Image Blending Using Pyramids in OpenCV Python," Morioh.com, Jun-2021. [Online]. Available: https://morioh.com/p/1e6a4d2d950c. [Accessed: 28-Oct-2021].
- [4] Kang and Atul, "Image Pyramids OpenCV Python," Theailearner.com, 19-Aug-2019. [Online]. Available: https://theailearner.com/tag/image-pyramids-opencv-python/. [Accessed: 28-Oct-2021].
- [5] https://github.com/PadovaY/pyramid_blend.
- [6] https://github.com/jagracar/OpenCV-python-tests.