INT3404E 20 - Image Processing: Homeworks 2

Ngô Lê Hoàng - 22028042

1 Image Filtering

1.1 Replicate padding

```
def padding_img(img, filter_size=3):
       Inputs:
           img: cv2 image: original image
           filter_size: int: size of square filter
          padded_img: cv2 image: the padding image
       height, width = img.shape
       # new_height = height + filter_size - 1
       # new_width = width + filter_size - 1
       # must pad each axis (filter_size - 1) padder, and each side in each direction will be (
                                                         filter_size - 1) // 2
       padded_add = (filter_size - 1) // 2
15
       new_height, new_width = height + padded_add * 2, width + padded_add * 2
       padded_img = np.zeros((new_height, new_width), dtype=np.uint8)
       for x in range(padded_add, new_height - padded_add):
           for y in range(padded_add, new_width - padded_add):
               padded_img[x][y] = img[x - padded_add][y - padded_add]
       # for corner pixel
       for x in range(padded_add + 1):
           for y in range(padded_add + 1):
               padded_img[x][y] = img[0][0]
               padded_img[new_height - x - 1][y] = img[-1][0]
               padded_img[x][new_width - y - 1] = img[0][-1]
               padded_img[new\_height - x - 1][new\_width - y - 1] = img[-1][-1]
       # for other border pixel
30
       for x in range(padded_add + 1, new_height - padded_add - 1):
           for y in range(padded_add):
               padded_img[x][y] = img[x - padded_add][0]
               padded_img[x][new_width - y - 1] = img[x - padded_add][-1]
35
       for y in range(padded_add + 1, new_width - padded_add - 1):
           for x in range (padded_add):
               padded_img[x][y] = img[0][y - padded_add]
               padded_img[new_height - x - 1][y] = img[-1][y - padded_add]
       return padded_img
```

1.2 Mean filters

```
def mean_filter(img, filter_size=3):
    """
    Inputs:
    img: cv2 image: original image
```

```
filter_size: int: size of square filter,
       Return:
          smoothed_img: cv2 image: the smoothed image with mean filter.
       padded_img = padding_img(img, filter_size)
10
       height, width = img.shape
       smoothed_img = np.zeros((height, width), dtype=float)
       for x in range(height):
           for y in range(width):
15
               smoothed_img[x][y] += sum(
                   padded_img[xx][yy]
                   for xx in range(x, x + filter_size)
                   for yy in range(y, y + filter_size)
20
               smoothed_img[x][y] /= filter_size**2
       return smoothed_img
```



Figure 1: Image after using mean filter

1.3 Median filters

```
def median_filter(img, filter_size=3):
    """
    Inputs:
        img: cv2 image: original image
        filter_size: int: size of square filter
    Return:
        smoothed_img: cv2 image: the smoothed image with median filter.
    """
```



Figure 2: Image after using median filter

1.4 Peak Signal-to-Noise Ratio (PSNR) metric

```
def psnr(gt_img, smooth_img):
    """
    Calculate the PSNR metric
    Inputs:
        gt_img: cv2 image: groundtruth image
        smooth_img: cv2 image: smoothed image
    Outputs:
        psnr_score: PSNR score
    """
```

Filter Type	PSNR Score
Mean Filter	18.295335205529753
Median Filter	17.835212311092132

Table 1: PSNR scores of the mean and median filters.

Based on the PSNR score, we can conclude that we should choose the median filter.

2 Fourier Transform

2.1 1D Fourier Transform

2.2 2D Fourier Transform

```
def DFT_2D(gray_img):
    """
    Implement the 2D Discrete Fourier Transform
    Note that: dtype of the output should be complex_
    params:
        gray_img: (H, W): 2D numpy array

returns:
        row_fft: (H, W): 2D numpy array that contains the row-wise FFT of the input image
        row_col_fft: (H, W): 2D numpy array that contains the column-wise FFT of the input image
        """
        row_fft = np.fft.fft(gray_img, axis=1)
        row_col_fft = np.fft.fft(row_fft, axis=0)
```

```
return row_fft, row_col_fft
```

2.3 Frequency Removal Procedure

```
def filter_frequency(orig_img, mask):
       Params:
         orig_img: numpy image
        mask: same shape with orig_img indicating which frequency hold or remove
         f_img: frequency image after applying mask
        img: image after applying mask
       transform = np.fft.fft2(orig_img)
       shift_fft = np.fft.fftshift(transform)
       f_img = shift_fft * mask
       shift_f_img = np.fft.ifftshift(f_img)
15
       f_{img} = np.abs(f_{img})
       img = np.fft.ifft2(shift_f_img)
       img = np.abs(img)
20
       return f_img, img
```

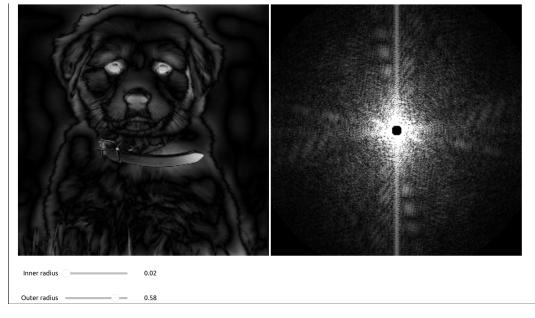


Figure 3: 2D Frequency Removal

2.4 Creating a Hybrid Image

```
def create_hybrid_img(img1, img2, r):
    f1 = fft2(img1)
    f2 = fft2(img2)

f1s = fftshift(f1)
```

```
f2s = fftshift(f2)
       rows, cols = img1.shape
       center_row, center_col = rows // 2, cols // 2
10
       mask = np.ones((rows, cols))
       r2 = r * r
       for i in range(rows):
           for j in range(cols):
               if (i - center_row) ** 2 + (j - center_col) ** 2 < r2:</pre>
                   mask[i, j] = 0
       f1s = f1s * mask
       f2s = f2s * (1 - mask)
       hybrid = f1s + f2s
20
       hybrid = ifftshift(hybrid)
       img_hybrid = ifft2(hybrid)
       img_hybrid = np.abs(img_hybrid)
       return img_hybrid
25
```







Figure 4: Creating hybrid image