# INT3404E 20 - Image Processing: Homeworks 2

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# 1 Image Filtering

The exercise will help to understand basic image filters by manipulating box/mean and median filters by implement the first two problems in the provided Python script file (ex1.py)

# 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
   filter_size //= 2
   padded_img = np.zeros((height + 2 * filter_size, width + 2 * filter_size), dtype=img.dtype)
   padded_imq[filter_size:filter_size + height, filter_size:filter_size + width] = imq
   # Padding top
   padded_img[:filter_size, filter_size:filter_size + width] = img[0]
   #Padding bottom
   padded_img[-filter_size:, filter_size:filter_size + width] = img[-1]
   for k in range (filter_size):
   # Padding left
     padded_img[:, k] = padded_img[:, filter_size]
   # Padding right
     padded_img[:, - k -1] = padded_img[:, - filter_size - 1]
   return padded_img
```

### Result:

```
1 sample = np.random.randint(10, size=(5, 5))
2 sample_pad = padding_img(sample, 3)
3 sample_pad

array([[5, 5, 4, 2, 1, 7, 7],
       [5, 5, 4, 2, 1, 7, 7],
       [6, 6, 3, 5, 5, 4, 4],
       [8, 8, 3, 3, 6, 5, 5],
       [6, 6, 9, 3, 4, 7, 7],
       [3, 3, 0, 9, 5, 3, 3],
       [3, 3, 0, 9, 5, 3, 3]])
```

Figure 1: Replicate padding function result

# 1.2 Box/mean filter

```
def mean_filter(img, filter_size=3):
    Inputs:
        img: cv2 image: original image
        filter_size: int: size of square filter,
       smoothed_img: cv2 image: the smoothed image with mean filter.
  # Need to implement here
   height, width = img.shape
   padded_img = padding_img(img, filter_size)
    \# Assume stride = 1
   kernel = np.ones((filter_size, filter_size), dtype=np.float32) / (filter_size ** 2)
   filtered_img = np.zeros(img.shape, dtype = np.float32)
   for i in range (height):
     for j in range (width):
       neighbor = padded_img[i:i+filter_size, j:j+filter_size]
        filtered_img[i, j] = np.sum(neighbor * kernel)
    filtered_img = filtered_img.astype(img.dtype)
   return filtered_img
```

### **Result:**

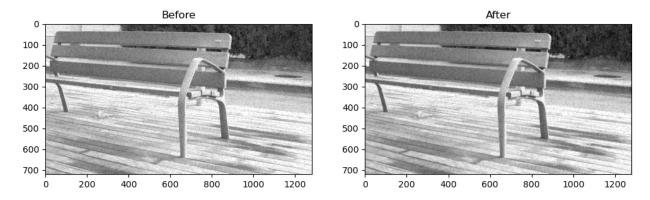


Figure 2: Mean filter function result

### 1.3 Median filter

```
padded_img = padding_img(img, filter_size)

# Assume stride = 1

kernel = np.ones((filter_size, filter_size), dtype=np.float32) / (filter_size ** 2)

filtered_img = np.zeros(img.shape, dtype = np.float32)

for i in range (height):
    for j in range (width):
        neighbor = padded_img[i:i+filter_size, j:j+filter_size]

    filtered_img[i, j] = np.median(neighbor)

filtered_img = filtered_img.astype(img.dtype)

return filtered_img
```

#### **Result:**

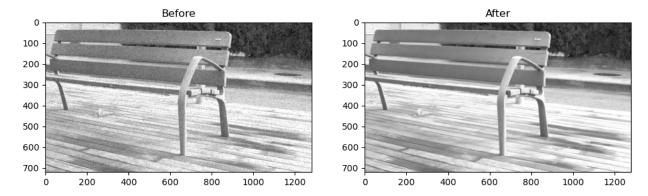


Figure 3: Median filter function result

# 2 Fourier Transform

In this exercise, we will implement the Discrete Fourier Transform (DFT) algorithm from scratch. The goal is to familiarize the fundamental concepts and procedural steps involved in applying this algorithm. first two problems will be done in the provided Python script file (ex212.py), and the remaining two problems will be completed in the provided Jupyter notebook (ex234.ipynb).

#### 2.1 1D Fourier Transform

```
def DFT_slow(data):
    """

Implement the discrete Fourier Transform for a 1D signal
    params:
    data: Nx1: (N, ): 1D numpy array
    returns:
        DFT: Nx1: 1D numpy array
    """

N = len(data)
```

```
DFT = np.zeros(N, dtype=np.complex128)

for s in range (N):
    for n in range (N):
        DFT[s] += data[n] * np.exp(-1j * 2 * np.pi * s * n / N)
        # print(DFT[s])
    DFT[s] /= N

    """
    Note:
    Some Discrete Fourier Transform formula doesn't include the 1/N part
    Therefore, removing the DFT[s] /= N part will match this fucntion result to the np.fft.fft()
    """

return DFT
```

### 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

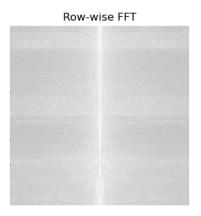
returns:
        row_fft: (H, W): 2D numpy array that contains the row-wise FFT of the input image
        row_col_fft: (H, W): 2D numpy array that contains the column-wise FFT of the input image
        """

img_row_fft = np.fft.fft(gray_img, axis = 1)
    img_col_fft = np.fft.fft(img_row_fft, axis = 0)

return img_row_fft, img_col_fft
```

### Result:





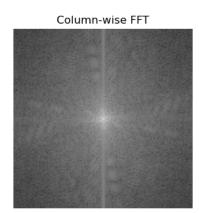


Figure 4: 2D Fourier Transform result

## 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
Output:
  f_img: frequency image after applying mask
  img: image after applying mask
"""

f_img = np.fft.fft2(orig_img)
f_img = np.fft.fftshift(f_img)
f_img = f_img * mask
img = np.fft.ifftshift(f_img)
img = np.fft.ifftshift(f_img)
return np.abs(f_img), np.abs(img)
```

#### **Result:**

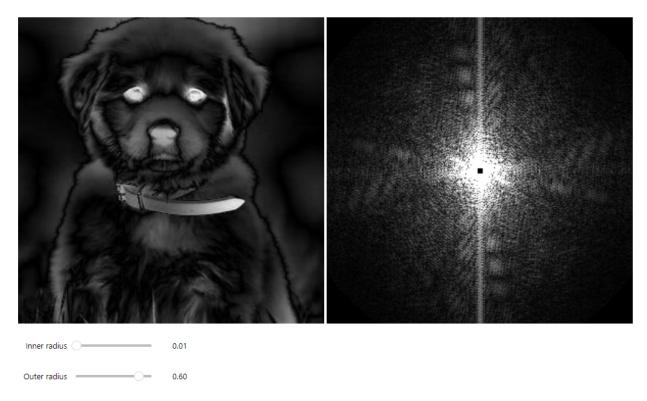


Figure 5: Frequency removal result

## 2.4 Creating a Hybird Image

```
def create_hybrid_img(img1, img2, r):
    """
    Create hydrid image
    Params:
    img1: numpy image 1
    img2: numpy image 2
    r: radius that defines the filled circle of frequency of image 1
    """
    # You need to implement the function
    img1_fft = np.fft.fft2(img1)
    img2_fft = np.fft.fft2(img2)
```

```
imgl_fft_shifted = np.fft.fftshift(imgl_fft)
img2_fft_shifted = np.fft.fftshift(img2_fft)

height, width = imgl.shape
mask = np.zeros(imgl.shape, dtype=np.float32)
center = (height // 2, width // 2)
for i in range(height):
    for j in range(width):
        if (i - center[0])**2 + (j - center[1])**2 <= r**2:
            mask[i, j] = 1

hybrid_fft_shifted = imgl_fft_shifted * (mask) + img2_fft_shifted * (mask - 1)

hybrid_fft = np.fft.ifftshift(hybrid_fft_shifted)

hybrid_img = np.fft.ifft2(hybrid_fft)

resulst = np.real(hybrid_img)

return resulst</pre>
```

### Result:







Figure 6: Hybird image result