ex1-2

October 15, 2024

1 ex1: Inverse Filtering

```
[1]: from scipy.fft import fft2, fftshift
     from pathlib import Path
     import numpy as np
     import skimage
     import matplotlib.pyplot as plt
[2]: ASSETS FOLDER PATH = "./assets"
     OUTPUT_FOLDER_PATH = "."
[3]: Path(OUTPUT_FOLDER_PATH).mkdir(parents=True, exist_ok=True)
    Cargamos la imagen
[4]: my_image = skimage.io.imread(fname=f"{ASSETS_FOLDER_PATH}/lena_gray.tif")
     imsize = my_image.shape
    Definimos las funciones a utilizar:
[5]: def fft2c(x: np.ndarray) -> np.ndarray:
         return fftshift(fft2(fftshift(x)))
[6]: def ifft2c(x: np.ndarray) -> np.ndarray:
         return fftshift(np.fft.ifft2(fftshift(x)))
[7]: def lowpass_kernel(kernel_size, image_dimensions):
         height, width = image_dimensions
         kernel = np.zeros((height, width))
         center_x, center_y = (height // 2, width // 2)
         kernel[center_x - kernel_size // 2: center_x + kernel_size // 2 + 1,
                center_y - kernel_size // 2: center_y + kernel_size // 2 + 1] = 1 / __
      ⇔(kernel_size ** 2)
         return kernel
[8]: def plot_image(img, title):
         plt.imshow(img, cmap='gray')
         plt.axis('off')
```

```
plt.title(title)
plt.show()
```

```
[9]: def plot_images(img1, title1, img2, title2):
    fig = plt.figure(figsize=(6, 7))

fig.add_subplot(1, 2, 1)
    plt.imshow(img1, cmap='gray')
    plt.axis('off')
    plt.title(title1)

fig.add_subplot(1, 2, 2)
    plt.imshow(img2, cmap='gray')
    plt.axis('off')
    plt.title(title2)

plt.subplots_adjust(wspace=0.05, hspace=0)
    plt.show()
```

```
[10]: def mse(original, variation):
    return np.mean((original - variation) ** 2)
```

1.1 Blur

Calculamos el kernel:

```
[11]: gks = 80
gk = np.exp(-(np.arange(max(imsize)) - max(imsize) // 2) ** 2 / (2 * gks ** 2))
gauss_kernel = np.outer(gk, gk)

gauss_kernel = gauss_kernel[
          (max(imsize) - imsize[0]) // 2:(max(imsize) - imsize[0]) // 2 + imsize[0],
               (max(imsize) - imsize[1]) // 2:(max(imsize) - imsize[1]) // 2 + imsize[1]
]
```

```
[12]: # blur_kernel = gauss_kernel
blur_kernel = lowpass_kernel(5, imsize)
```

Obtenemos la imagen con blur:

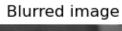
```
[13]: blurred_image = np.abs(ifft2c(fft2c(my_image) * fft2c(blur_kernel)))
plot_images(my_image, 'Original image', blurred_image, 'Blurred image')
```

Original image

Blurred image



Aplicando la operación inversa sobre la imagen borrosa:





Restored image



mse: 1.2837269174207216e-21

1.2 Blur + Noise

Definimos el signal-to-noise ratio:

```
[15]: SNR_dB = 10
```

Calculamos la desviación estándar de la imagen borrosa para el ruido:

```
[16]: sigma_blurred_image = np.std(blurred_image)
sigma_noise = np.sqrt((sigma_blurred_image ** 2) * (10 ** (-SNR_dB / 10)))
noise = np.random.normal(0, sigma_noise, (imsize[0], imsize[1]))
```

```
[17]: blurred_noise_image = blurred_image + noise plot_images(my_image, 'Original image', blurred_noise_image, 'Blurred image_u with noise')
```

Original image



Blurred image with noise



Aplicando la operación inversa:

Blurred image with noise



Restored image



mse: 273.99873726572713

2 ex2: Wiener

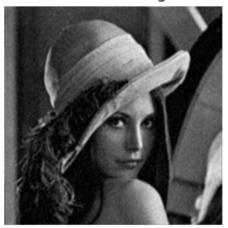
Definimos un filtro de Wiener:

```
def wiener_filter_restoration(blurred_image, blur_kernel, noise):
    blurred_fft = fft2c(blurred_image)
    kernel_fft = fft2c(blur_kernel)
    noise_psd = np.square(np.abs(fft2c(noise)))
    wiener_filter = np.conj(kernel_fft) / (np.abs(kernel_fft) ** 2 + noise_psd /
    np.square(np.abs(blurred_fft)))
    restored_fft = wiener_filter * blurred_fft
    restored_image = np.abs(ifft2c(restored_fft)))
    return restored_image
```

Blurred image with noise



Restored Image



mse: 68.87812877972368