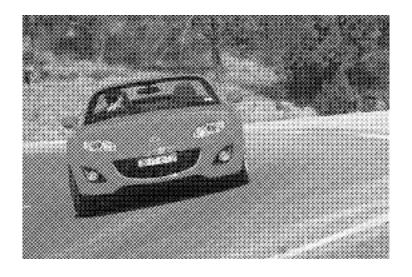
CO543 Lab05

March 24, 2022

- 1 LAB 05
- 2 K.G.I.S. Nawarathna
- 3 E/17/219

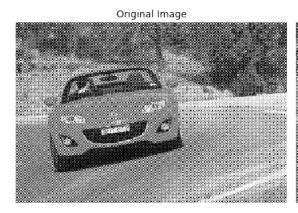
```
[66]: #importing necessary libraries
      import numpy as np
      import pandas as pd
      import cv2
      import matplotlib.pyplot as plt
      # #imporot the patch for the google colab
      from google.colab.patches import cv2_imshow
[67]: #specifying the folder in the google drive
      folder_name = '/content/drive/MyDrive/C0543-Image Processing/Lab5/'
[68]: #load the image
      img = cv2.imread(folder_name+'car-2.jpg',0)
      img
[68]: array([[ 0, 196, 0, ..., 3, 197, 102],
             [210,
                     0, 148, ..., 255,
                                       2, 98],
             [ 0, 184, 0, ..., 14, 123, 101],
             [201, 12, 204, ..., 255, 117, 199],
             [ 2, 254, 32, ..., 8, 255, 165],
             [210, 24, 175, ..., 241, 117, 219]], dtype=uint8)
[69]: #display the image
      cv2_imshow(img)
```

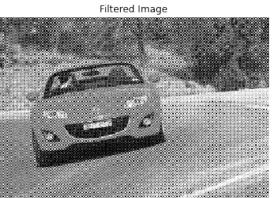


```
[70]: #function definition for comparing two images
      def compare_2_imgs(img1,img2):
        #set the figure size
       plt.figure(figsize=(10,7))
        #make subplots for two plots
       plt.subplot(1,2,1)
        #draw the image on the first plot
       plt.imshow(img1, cmap='gray', vmin = 0, vmax = 255,interpolation='none')
        #set the title of the original image
       plt.title('Original Image')
        #turn off the axis
       plt.axis('off')
        #make the second plot
       plt.subplot(1,2,2)
        #draw the second plot
       plt.imshow(img2, cmap='gray', vmin = 0, vmax = 255,interpolation='none')
        #set the title of the filtered image
       plt.title('Filtered Image')
        #turn off the axis
       plt.axis('off')
        #make the tight layout
```

plt.tight_layout()

[71]: #testing the function with the imported image compare_2_imgs(img,img)





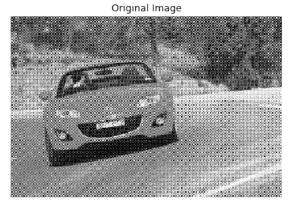
 $\#\#\mathrm{Task}$ 01

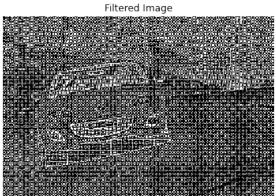
Apply high pass laplacian filter on Car.jpg image.

[72]: #apply the laplaian filter using the in-built function in the cv2

def laplace_img(img,kernal_size):
 laplace_filtered_img = cv2.Laplacian(img, ddepth=cv2.CV_64F, ksize=3)
 return laplace_filtered_img

[73]: #display the results
laplace = laplace_img(img,3)
compare_2_imgs(img,laplace)





 $\#\#\mathrm{Task}$ 02

Apply ideal high-pass filter on Car.jpg image for D0=100

```
[74]: #function defintion for high pass filter
def high_pass_filter(img,size):

    #getting the height and the width
    height , width = img.shape[0:2]

    #getting the mid of the image
    mid_height = int(height/2)
    mid_width = int(width/2)

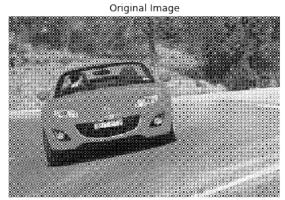
    #setting the corresponding section to 0
    img[mid_height-int(size/2):mid_height+int(size/2), mid_width-int(size/2):
    mid_width+int(size/2)] = 0

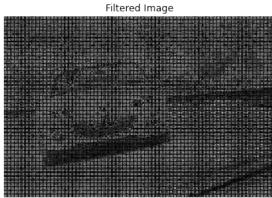
    return img
```

```
[75]: def display_pass_filtered(img,option=0):
        #apply fourier transform
        ft_img = np.fft.fft2(img)
        #move the frequency domain
        shifted_img = np.fft.fftshift(ft_img)
        if option == 1:
          #apply high pass filter
          shifted_img=high_pass_filter(shifted_img,100)
        elif option == 2:
          shifted_img=low_pass_filter(shifted_img,100)
        resulted_img = np.log(np.abs(shifted_img))
        #apply inverse fourier transform
        inv_tr = np.fft.ifftshift(shifted_img)
        #geeting the image
        ifimg = np.fft.ifft2(inv_tr)
        ifimg = np.abs(ifimg)
        #compare results
        compare_2_imgs(img,ifimg)
```

```
[76]: #show high pass filter display_pass_filtered(img,1)
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:14: RuntimeWarning:





3.1 Task 03

Apply ideal low-pass filter on Car.jpg image for D0=100

```
[77]: #function definition for low pass filter
def low_pass_filter(img,D):
    #getting the height and the width
    height , width = img.shape[0:2]

#getting the mid of the image
mid_height = int(height/2)
mid_width = int(width/2)

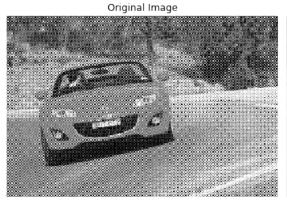
#definine a black image similar to the given image
img_zeros = np.zeros((height, width), np.uint8)

#set the corresponding section to 1
img_zeros[mid_height-int(D/2):mid_height+int(D/2), mid_width-int(D/2):
-mid_width+int(D/2)] = 1

#return the low pass filter
return img_zeros*img
```

```
[78]: #show high pass filter display_pass_filtered(img,2)
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:14: RuntimeWarning: divide by zero encountered in log



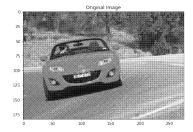


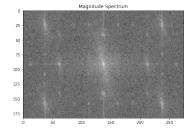
 $\#\#\mathrm{Task}~04$

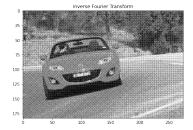
Apply FFT2, IFFT2, low-pass Gaussian filter, and high-pass laplacian filter on Car.jpg image.

```
[79]: #transforming the image
      transformed_img = np.fft.fft2(img)
      #shifting the image
      shifted_img = np.fft.fftshift(transformed_img)
      #getting the magnitude specturnm of the image
      magnitude_spectrum = np.abs(shifted_img)
      #inverse
      inv_img = np.fft.ifft2(shifted_img)
      #plotting three images
      fig, (axis1, axis2, axis3) = plt.subplots(1, 3, figsize=(25, 15))
      axis1.set_title('Original Image')
      axis1.imshow(img, cmap='gray')
      axis2.set_title('Magnitude Spectrum')
      axis2.imshow(20 * np.log(magnitude_spectrum), cmap='gray')
      axis3.set_title('Inverse Fourier Transform')
      axis3.imshow(np.abs(inv_img), cmap='gray')
```

[79]: <matplotlib.image.AxesImage at 0x7f6ec4667ed0>

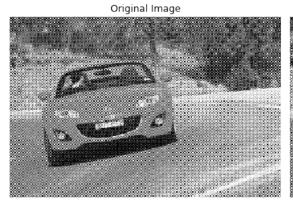




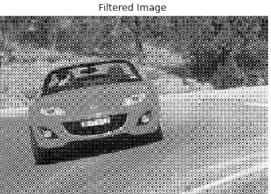


```
[80]: #function definition for the gaussian kernal
      #this is a supporting function for the gaussian filter
      def gaussian_kernel(img, sigma_value):
       return np.exp(-(img ** 2) / (2 * sigma_value ** 2)) / (2 * np.pi *_
       ⇒sigma_value ** 2)
[81]: #function definition for the guasiisan filter
      def gaussian_lowpass(img, mask_size, sigma_value):
        #get the kernal as a set of zeros
        kernel = np.zeros(mask_size, dtype=np.float32)
        #get the radius
        radius = mask_size[0]
        #get the low pass gaussian
        for i in range(mask_size[0]):
          for j in range(mask_size[1]):
            kernel[i, j] = gaussian_kernel((i - radius) ** 2 + (j - radius)_u
       →**2,sigma_value)
        #get the kernal
        kernel /= np.sum(kernel)
        #return the low pass gaussian filter
```

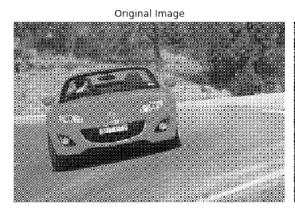
[82]: #apply low pass gaussian filter and display
gaussian_lowpass_img = gaussian_lowpass(img,(3,3),1)
compare_2_imgs(img, gaussian_lowpass_img)

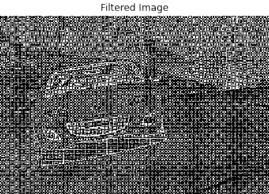


return cv2.filter2D(img, -1, kernel)



[83]: #apply high pass laplacian filter and display laplace = laplace_img(img,3) compare_2_imgs(img,laplace)





 $\#\#\mathrm{Task}~05$

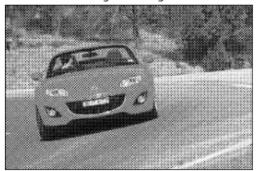
Apply the necessary filter and correct the noise in the image. Image file is uploaded.

```
[88]: #plotting the results

#set the fitler size
plt.figure(figsize=(10,7))

#plot the results in subplot
```

Original Image



Filtered Image



 $\#\#\mathrm{Task}$ 06

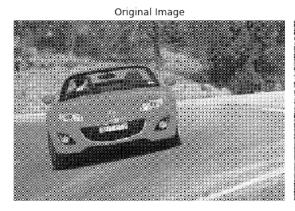
Apply the sobel operator (filter) on Car.jpg in the Fourier domain to detect edges

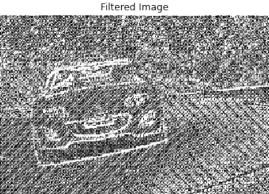
```
[93]: def sobel_filter(img):
                                             # array for the Sx
                                            x_{=} = np.array([[1.0, 0.0, -1.0], [2.0, 0.0, -2.0], [1.0, 0.0, -1.0]])
                                             # array for the Sy
                                            y_{-} = np.array([[1.0, 2.0, 1.0], [0.0, 0.0, 0.0], [-1.0, -2.0, -1.0]])
                                            #defining the padding values
                                            padding_x = (img.shape[0] - x_.shape[0], img.shape[1] - x_.shape[1])
                                           padding_y = (img.shape[0] - y_.shape[0], img.shape[1] - y_.shape[1])
                                            #getting the Sx and Sy which are incoporated with the padding
                                            Sx = np.pad(x_{,} ((padding_x[0] + 1) // 2, padding_x[0] // 2), ((padding_x[1] +_{\sqcup} +_{
                                         Sy = np.pad(y_{,} ((padding_y[0] + 1) // 2, padding_y[0] //2), ((padding_y[1] +_{L}))
                                         #shift Sx and Sy
                                            Sx_shifted = np.fft.ifftshift(Sx)
                                            Sy_shifted = np.fft.ifftshift(Sy)
                                             #return the filtered image
```

```
return np.abs(np.fft.ifft2(np.fft.fft2(img) * np.fft.fft2(Sx_shifted)) + np. 

offt.ifft2(np.fft.fft2(img) * np.fft.fft2(Sy_shifted)))
```

```
[94]: sobel_img = sobel_filter(img)
compare_2_imgs(img, filtered_img)
```





3.2 Task 07

Discuss applying Butterworth and Chebyshev filters and compare the output image with the Gaussian Filter image (You may use a preferred image to discuss the characteristics of the output images in Q7.)

```
[100]: def butterworth(img, n, d0):
    #define the kernal
    kernel = np.array([[1 / (1 + ((x ** 2 + y ** 2) / (d0 ** 2)) ** n))

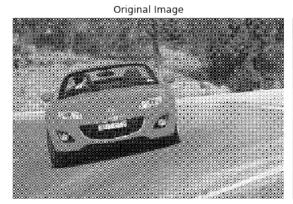
#getting the x and y
for x in range(-img.shape[1] // 2, img.shape[1] // 2)]
for y in range(-img.shape[0] //2, img.shape[0] // 2)], dtype=np.float32)

#fft shifting the image
fftshift_img = np.fft.fftshift(np.fft.fft2(img))

#filtering the image
filtered = np.fft.ifftshift(fftshift_img * kernel)

#returning the filtered image
return np.abs(np.fft.ifft2(filtered))
```

```
[101]: #displaying the results
butterworth_img = butterworth(img, 5, 50)
compare_2_imgs(img,butterworth_img)
```



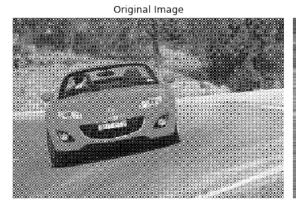


```
[103]: #function definition for the lowpass filter
def chebyshev_lowpass(img, n, rs, Wn):
    #importing the library
    from scipy.signal import cheby2, filtfilt

#retuning the filtered image
b_val, a_val = cheby2(n, rs, Wn)
    return filtfilt(b_val, a_val, img)
```

[104]: #getting the output image
 chebyshev_img = chebyshev_lowpass(img, 2, 30, 0.3)

#printing the image
 compare_2_imgs(img,chebyshev_img)





[107]: #plots all three images for comparison
#setting the figure size

```
plt.figure(figsize=(12,10))
#in the first sub plot, draw the buttorworth fitler
plt.subplot(1,3,1)
plt.imshow(butterworth_img, cmap='gray', vmin = 0, vmax =__
 →255,interpolation='none')
plt.title('Butterworth filtered image')
plt.axis('off')
#in the second sub plot, draw the chebyshev filter
plt.subplot(1,3,2)
plt.imshow(chebyshev_img, cmap='gray', vmin = 0, vmax =__
 ⇔255,interpolation='none')
plt.title('Chebyshev filtered image')
plt.axis('off')
#in the thrid sub plot, draw the low pass filter
plt.subplot(1,3,3)
plt.imshow(gaussian_lowpass_img, cmap='gray', vmin = 0, vmax =__
 ⇔255,interpolation='none')
plt.title('Gaussian Filter')
plt.axis('off')
```

[107]: (-0.5, 274.5, 182.5, -0.5)





Chebyshev filtered image



Gaussian Filter



It can be seen that the chebyshev's filter output is more fuzzy than the butterworth lowpass filter output. The noise is removed using a Gaussian filter. The uncertainty relation is optimized using the Gaussian filter. Without ringing, it gives the sharpest cutoff imaginable. However, the noise may still be observed in the output of the gaussian filter. To reduce this type of noise, the Butterworth filter is more effective.