Lab₀₈ E/19/167

Fourier Transformation

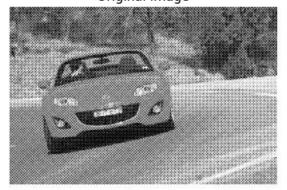
Task 1

Code

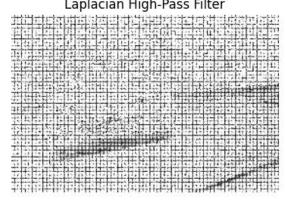
```
import cv2
import numpy as np
import matplotlib.pyplot as plt
image_path = '_/content/car.jpg' # Replace this with the path to your image
image = cv2.imread(image_path, cv2.IMREAD_GRAYSCALE)
# Check if the image is loaded properly
if image is None:
   print("Error loading image")
    laplacian = cv2.Laplacian(image, cv2.CV_64F)
    laplacian = cv2.convertScaleAbs(laplacian)
    plt.figure(figsize=(10, 5))
    plt.subplot(1, 2, 1)
    plt.title('Original Image')
plt.imshow(image, cmap='gray')
    plt.axis('off')
    plt.subplot(1, 2, 2)
plt.title('Laplacian High-Pass Filter')
    plt.imshow(laplacian, cmap='gray')
    plt.axis('off')
    plt.show()
```

Output

Original Image



Laplacian High-Pass Filter

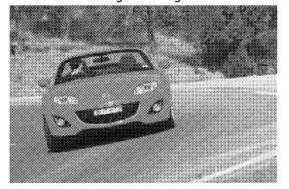


Code

```
def ideal_high_pass_filter(img, D0):
    rows, cols = img.shape
                                                                              image = cv2.imread(image_path, cv2.IMREAD_GRAYSCALE)
   mask = np.ones((rows, cols), np.uint8)
                                                                              # Check if the image is loaded properly
                                                                              if image is None:
                                                                                  print("Error loading image")
    crow, ccol = rows // 2, cols // 2
                                                                                  # Apply the ideal high-pass filter
    # Create the ideal high-pass filter
for u in range(rows):
                                                                                  D0 = 100 # Cut-off frequency
                                                                                  filtered_image = ideal_high_pass_filter(image, D0)
        for v in range(cols):
            D = np.sqrt((u - crow) ** 2 + (v - ccol) ** 2)
            if D <= D0:
                mask[u, v] = 0
                                                                                  plt.figure(figsize=(10, 5))
    # Perform the Fourier Transform
dft = cv2.dft(np.float32(img), flags=cv2.DFT_COMPLEX_OUTPUT)
                                                                                  plt.subplot(1, 2, 1)
                                                                                  plt.title('Original Image')
    dft_shift = np.fft.fftshift(dft)
                                                                                  plt.imshow(image, cmap='gray')
    # Apply the mask
                                                                                  plt.axis('off')
    dft_shift[:, :, 0] = dft_shift[:, :, 0] * mask
dft_shift[:, :, 1] = dft_shift[:, :, 1] * mask
                                                                                  plt.subplot(1, 2, 2)
plt.title('Ideal High-Pass Filter (D0=100)')
    # Perform the inverse Fourier Transform
f_ishift = np.fft.ifftshift(dft_shift)
img_back = cv2.idft(f_ishift)
                                                                                  plt.imshow(filtered_image, cmap='gray')
                                                                                  plt.axis('off')
    img_back = cv2.magnitude(img_back[:, :, 0], img_back[:, :, 1])
                                                                                  plt.show()
    return img_back
```

Output

Original Image



Ideal High-Pass Filter (D0=100)

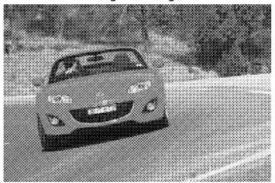


Code

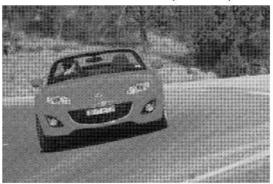
```
ideal_low_pass_filter(img, D0):
# Get the image dimension:
rows, cols = img.shape
# Create a mask with the same size as the image
mask = np.zeros((rows, cols), np.uint8)
                                                                                                                     image = cv2.imread(image_path, cv2.IMREAD_GRAYSCALE)
# Get the center of the image
crow, ccol = rows // 2, cols // 2
                                                                                                                     if image is None:
    print("Error loading image")
 for u in range(rows):
      for v in range(cols):
    D = np.sqrt((u - crow) ** 2 + (v - ccol) ** 2)
    if D <= D0:</pre>
                                                                                                                           filtered_image = ideal_low_pass_filter(image, D0)
                   mask[u, v] = 1
                                                                                                                           plt.figure(figsize=(10, 5))
# Perform the Fourier Transform
dft = cv2.dft(np.float32(img), flags=cv2.DFT_COMPLEX_OUTPUT)
dft_shift = np.fft.fftshift(dft)
                                                                                                                          plt.subplot(1, 2, 1)
plt.title('Original Image')
plt.imshow(image, cmap='gray')
plt.axis('off')
# Apply the mask
dft_shift[:, :, 0] = dft_shift[:, :, 0] * mask
dft_shift[:, :, 1] = dft_shift[:, :, 1] * mask
                                                                                                                           plt.subplot(1, 2, 2)
plt.title('Ideal Low-Pass Filter (D0=100)')
plt.imshow(filtered_image, cmap='gray')
f_ishift = np.fft.ifftshift(dft_shift)
img_back = cv2.idft(f_ishift)
img_back = cv2.magnitude(img_back[:, :, 0], img_back[:, :, 1])
                                                                                                                            plt.axis('off')
                                                                                                                           plt.show()
return img_back
```

Output

Original Image



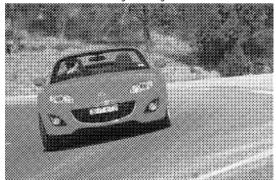
Ideal Low-Pass Filter (D0=100)



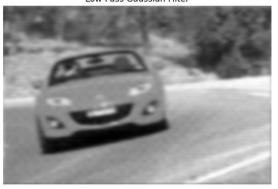
```
import cv2
import numpy as np
import matplotlib.pyplot as plt
image_path = '/content/car.jpg' # Replace this with the path to your image
image = cv2.imread(image_path, cv2.IMREAD_GRAYSCALE)
# Check if the image is loaded properly
if image is None:
   print("Error loading image")
    f = np.fft.fft2(image)
    fshift = np.fft.fftshift(f)
    # IFFT2
    ishift = np.fft.ifftshift(fshift)
    img_back = np.fft.ifft2(ishift)
    img_back = np.abs(img_back)
    # Apply low-pass Gaussian filter
    rows, cols = image.shape
    crow, ccol = rows // 2 , cols // 2
    def create_gaussian_mask(shape, sigma):
        rows, cols = shape
crow, ccol = rows // 2 , cols // 2
        x = np.linspace(-cols/2, cols/2, cols)
        y = np.linspace(-rows/2, rows/2, rows)
        x, y = np.meshgrid(x, y)
        d = np.sqrt(x*x + y*y)
g = np.exp(-(d**2 / (2.0 * sigma**2)))
        return g
```

```
mask = create_gaussian_mask((rows, cols), sigma)
fshift_lp = fshift * mask
ishift_lp = np.fft.ifftshift(fshift_lp)
img_lp = np.fft.ifft2(ishift_lp)
img_lp = np.abs(img_lp)
# Apply high-pass Laplacian filter
laplacian = cv2.Laplacian(img_lp, cv2.CV_64F)
laplacian = cv2.convertScaleAbs(laplacian)
# Display the results
plt.figure(figsize=(15, 10))
plt.subplot(2, 2, 1)
plt.title('Original Image')
plt.imshow(image, cmap='gray')
plt.axis('off')
plt.subplot(2, 2, 2)
plt.title('Reconstructed Image (IFFT2)')
plt.imshow(img_back, cmap='gray')
plt.axis('off')
plt.subplot(2, 2, 3)
plt.title('Low-Pass Gaussian Filter')
plt.imshow(img_lp, cmap='gray')
plt.axis('off')
plt.subplot(2, 2, 4)
plt.title('High-Pass Laplacian Filter')
plt.imshow(laplacian, cmap='gray')
plt.axis('off')
plt.show()
```

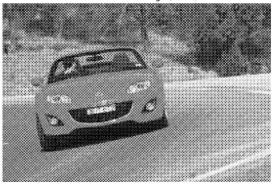
Original Image



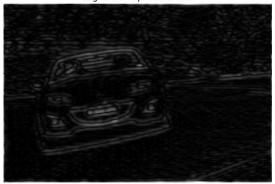
Low-Pass Gaussian Filter



Reconstructed Image (IFFT2)



High-Pass Laplacian Filter



Task 5

```
sigma = 30
mask = create_gaussian_mask((rows, cols), sigma)
   mport cv2
 import numpy as np
                                                                                                                                                          fshift_lp = fshift * mask
ishift_lp = np.fft.ifftshift(fshift_lp)
img_lp = np.fft.ifft2(ishift_lp)
img_lp = np.abs(img_lp)
import matplotlib.pyplot as plt
# Load the image
image_path = '/content/car.jpg' # Path to the uploaded image
image = cv2.imread(image_path, cv2.IMREAD_GRAYSCALE)
                                                                                                                                                           # Apply high-pass Laplacian filter
laplacian = cv2.Laplacian(image_denoised, cv2.CV_64F)
laplacian = cv2.convertScaleAbs(laplacian)
if image is None:
    print("Error loading image")
                                                                                                                                                                nsharp Masking wusharp.mask(image, sigma=1.0, strength=1.5, threshold=0): blurred = cv2.GaussianBlur(image, (0, 0), sigma) sharpened = float(strength + 1) * image - float(strength) * blurred sharpened = np.maximum(sharpened, np.zeros(sharpened.shape)) sharpened = np.minimum(sharpened, 255 * np.ones(sharpened.shape)) sharpened = sharpened.round().astype(np.uint8)
         # Apply Gaussian Blur to remove noise
image_denoised = cv2.GaussianBlur(image, (5, 5), 0)
                                                                                                                                                                if threshold > 0:
    low_contrast_mask = np.absolute(image - blurred) < threshold
    np.copyto(sharpened, image, where=low_contrast_mask)</pre>
          f = np.fft.fft2(image_denoised)
                                                                                                                                                                 return sharpened
          ishift = np.fft.ifftshift(fshift)
                                                                                                                                                           sharp_image = unsharp_mask(image_denoised)
         img_back = np.fft.ifft2(ishift)
img_back = np.abs(img_back)
                                                                                                                                                          # Display the results
plt.figure(figsize=(15, 10))
         # Apply low-pass Gaussian filter
rows, cols = image.shape
crow, ccol = rows // 2 , cols // 2
                                                                                                                                                          plt.subplot(2, 3, 1)
plt.title('Original Image')
plt.imshow(image, cmap='gray')
plt.axis('off')
         # Create a Gaussian mask
def create_gaussian_mask(shape, sigma):
                                                                                                                                                          plt.subplot(2, 3, 2)
plt.title('Denoised Image')
plt.imshow(image_denoised, cmap='gray')
                  rows, cols = shape
                  x = np.linspace(-cols/2, cols/2, cols)
y = np.linspace(-rows/2, rows/2, rows)
                                                                                                                                                          plt.axis('off'
                                                                                                                                                          plt.subplot(2, 3, 3)
plt.title('Reconstructed Image (IFFT2)')
plt.imshow(img_back, cmap='gray')
plt.axis('off')
                  x, y = np.meshgrid(x, y)
                 d = np.sqrt(x*x + y*y)
g = np.exp(-(d**2 / (2.0 * sigma**2)))
```

```
plt.subplot(2, 3, 4)
plt.title('Low-Pass Gaussian Filter')
plt.imshow(img_lp, cmap='gray')
plt.axis('off')

plt.subplot(2, 3, 5)
plt.title('High-Pass Laplacian Filter')
plt.imshow(laplacian, cmap='gray')
plt.axis('off')

plt.subplot(2, 3, 6)
plt.title('Sharpened Image (Unsharp Mask)')
plt.imshow(sharp_image, cmap='gray')
plt.axis('off')

plt.show()
```

Original Image



Denoised Image

Reconstructed Image (IFFT2)



Low-Pass Gaussian Filter



High-Pass Laplacian Filter

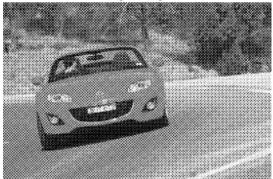


Sharpened Image (Unsharp Mask)



```
image_path = '/content/car.jpg' # Path to the uploaded image
image = cv2.imread(image_path, cv2.IMREAD_GRAYSCALE)
# Check if the image is loaded properly
if image is None:
    print("Error loading image")
    image_denoised = cv2.GaussianBlur(image, (5, 5), 0)
    sobelx = cv2.Sobel(image_denoised, cv2.CV_64F, 1, 0, ksize=3)
    sobely = cv2.Sobel(image_denoised, cv2.CV_64F, 0, 1, ksize=3)
    sobel = np.hypot(sobelx, sobely) # Combine Sobel X and Y
    # FFT2 of the Sobel-filtered image
    f_sobel = np.fft.fft2(sobel)
    fshift_sobel = np.fft.fftshift(f_sobel)
    magnitude_spectrum = np.log(1 + np.abs(fshift_sobel))
    plt.figure(figsize=(15, 10))
    plt.subplot(2, 2, 1)
    plt.title('Original Image')
plt.imshow(image, cmap='gray')
    plt.axis('off')
    plt.subplot(2, 2, 2)
    plt.title('Denoised Image')
    plt.imshow(image_denoised, cmap='gray')
    plt.axis('off')
    plt.subplot(2, 2, 3)
plt.title('Sobel Filtered Image')
plt.imshow(sobel, cmap='gray')
    plt.axis('off')
    plt.subplot(2, 2, 4)
    plt.title('Magnitude Spectrum (Fourier Domain)')
    plt.imshow(magnitude_spectrum, cmap='gray')
    plt.axis('off')
    plt.show()
```





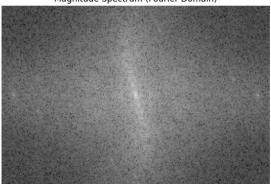
Sobel Filtered Image



Denoised Image

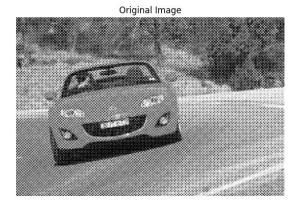


Magnitude Spectrum (Fourier Domain)

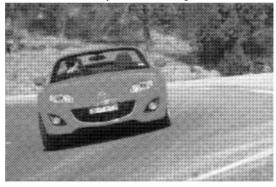


Task 7

```
# Load the image
image_path = '/content/car.jpg' # Adjust the path accordingly
image = cv2.imread(image_path, cv2.IMREAD_GRAYSCALE)
                                                                                                                                                                                                               # Apply filters
butterworth_mask = butterworth_filter(image.shape, 50, 2)
chebyshev_mask = chebyshev_filter(image.shape, 50, 0.5)
gaussian_mask = gaussian_filter(image.shape, 50)
 if image is None:
    print("Error loading image")
    exit()
                                                                                                                                                                                                                fshift_butterworth = fshift * butterworth_mask
fshift_chebyshev = fshift * chebyshev_mask
fshift_gaussian = fshift * gaussian_mask
# FFT2
f = np.fft.fft2(image)
fshift = np.fft.fftshift(f)
                                                                                                                                                                                                                  # Inverse FFT to get images
def apply_inverse_fft(fshift):
   ishift = np.fft.ifftshift(fshift)
   img_back = np.abs(img_back)
   return img_back)
 # Butterworth filter
def butterworth_filter(shape, cutoff, order):
           butterworth_filter(shape, cutoff, order):
rows, cols = shape
crow, ccol = rows // 2, cols // 2
x = np.linspace(-ccol, ccol, cols)
y = np.linspace(-crow, crow, rows)
X, Y = np.meshgrid(x, y)
distance = np.sqrt((X - 0)**2 + (Y - 0)**2)
filter = 1 / (1 + (distance / cutoff)**(2 * order))
return filter
                                                                                                                                                                                                                img_butterworth = apply_inverse_fft(fshift_butterworth)
img_chebyshev = apply_inverse_fft(fshift_chebyshev)
img_gaussian = apply_inverse_fft(fshift_gaussian)
                                                                                                                                                                                                                # Plot the results plt.figure(figsize=(15, 10))
                                                                                                                                                                                                               plt.subplot(2, 2, 1)
plt.title('Original Image')
plt.imshow(image, cmap='gray')
plt.axis('off')
 # Chebyshev filter
def chebyshev_filter(shape, cutoff, ripple):
            chedyshev_filter(shape, cutoff, ripple):
rows, cols = shape
crow, ccol = rows // 2, cols // 2
x = np.linspace(-ccol, ccol, cols)
y = np.linspace(-crow, crow, rows)
X, Y = np.meshgrid(X, y)
distance = np.sqrt((X - 0)**2 + (Y - 0)**2)
filter = 1 / (1 + ripple * (distance / cutoff)**2)
return filten
                                                                                                                                                                                                               plt.subplot(2, 2, 2)
plt.title('Butterworth Filtered Image')
plt.imshow(img_butterworth, cmap='gray')
               return filter
                                                                                                                                                                                                               plt.subplot(2, 2, 3)
plt.title('Chebyshev Filtered Image')
plt.imshow(img_chebyshev, cmap='gray')
           aussian filter
gaussian_filter(shape, sigma):
rows, cols = shape
crow, ccol = rows // 2, cols // 2
x = np.linspace(-ccol, ccol, cols)
y = np.linspace(-crow, crow, rows)
X, Y = np.meshgrid(x, y)
d = np.sqrt(XYX + Y*Y)
g = np.exp(-(d**2 / (2.0 * sigma**2)))
return a
                                                                                                                                                                                                                plt.axis('off')
                                                                                                                                                                                                              plt.subplot(2, 2, 4)
plt.title('Gaussian Filtered Image')
plt.imshow(img_gaussian, cmap='gray')
plt.axis('off')
```







Butterworth Filtered Image



Gaussian Filtered Image



The Butterworth-filtered image shows a moderate level of blurring, with the edges and details slightly softened. The noise is reduced, but the image retains a fair amount of its original sharpness.

The Chebyshev-filtered image is noticeably blurred with a slightly higher degree of smoothness compared to the Butterworth filter. The edges and finer details are less pronounced, indicating a stronger noise reduction effect.

The Gaussian-filtered image exhibits the highest level of blurring among the three filters. The edges and fine details are significantly softened, providing a very smooth and clean appearance. This makes the Gaussian filter effective for removing high-frequency noise but can also result in loss of important image details.