

Department of Computer Science and Engineering (Data Science) Academic Year 2022-2023

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AIM: To implement Ideal Low Pass & High Pass Filtering on an image

THEORY:

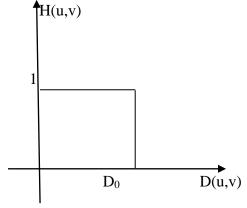
1. Ideal Low Pass Filter

This filter cuts off all high frequency components of the Fourier transform that are at a distance greater than a specified distance D_0 .

$$H(u,v) = 1$$
; if $D(u,v) < D_0$
= 0; if $D(u,v) > D_0$

Where,

 D_0 is the specified non negative distance.



Response of Ideal Low Pass Filter

D(u,v) is the distance from the point (u,v) to the origin of the frequency rectangle for an M X N image.

$$D(u,v)=[(u-M/2)^2+(v-N/2)^2]^{\frac{1}{2}}$$

Therefore,

For an image, when u=M/2, v=N/2

$$D(u,v)=0$$

This formula centers our H(u,v).

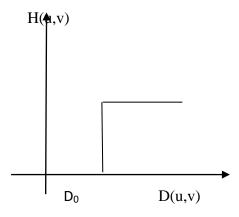


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D(u,v) gives us concentric rings with each ring having a fixed value.

2. Ideal High Pass Filter

This filter cuts off all high frequency components of the Fourier transform that are at a distance greater than a specified distance D_0 .



$$H(u,v) = 0$$
; if $D(u,v) < D0$
= 1; if $D(u,v) > D0$

Where,

D0 is the specified non negative distance.

D(u,v) is the distance from the point (u,v) to the origin of the frequency rectangle for an M X N image.

RESULT:

An ideal low-pass filter completely eliminates all frequencies above the cutoff frequency while passing those below unchanged. Ideal low pass filters blurs the image where the High pass filters only shows/highlights the part of the image which is important

Ideal low pass filter and High pass filter Sarvayga Singh -- 60009200030 (K1) Lab8 In [1]: import cv2 import numpy as np from google.colab.patches import cv2_imshow import matplotlib.pyplot as plt In [2]: def step(num): num_list = num.split('.') **if** num_list[1] > 5: return int(num_list(1))+1 else: return int(num_list[0]) In []: img = cv2.imread('/content/image.jpg',0) cv2_imshow(img) In []: M, N = img.shape[0], img.shape[1] M, N (564, 849) Out[]: np.max(img), np.min(img) (218, 169) Out[]: def dist(u,v): return np.sqrt((u - M/2)**2+(v-N/2)**2) In []: F = np.fft.fft2(img) Fshift = np.fft.fftshift(F) In []: h = [[0 for i in range(N)] for j in range(M)] h = np.array(h)threshold = 70print(len(h[0]),len(h)) 849 564 In []: for i in range(0, M): #y axis for j in range(0,N): # x axis d = dist(i,j)if d<=threshold:</pre> h[i,j] = 1else: h[i,j] = 0plt.imshow(h, cmap='gray') plt.axis('off') plt.show() Gshift = Fshift * h G = np.fft.ifftshift(Gshift) g = np.abs(np.fft.ifft2(G)) plt.imshow(g, cmap='gray') plt.axis('off') plt.show() In []: np.shape(h) Out[]: (564, 849) In []: i = [[1 for i in range(N)] for j in range(M)] In []: # cv2_imshow(np.array(h)) # Filter: High pass filter h = 1 - hplt.imshow(h, cmap='gray') plt.axis('off') plt.show() Gshift = Fshift * h G = np.fft.ifftshift(Gshift) g = np.abs(np.fft.ifft2(G)) plt.imshow(g, cmap='gray') plt.axis('off') plt.show() In []: # calculating the discrete Fourier transform DFT = cv2.dft(np.float32(img), flags=cv2.DFT_COMPLEX_OUTPUT) # reposition the zero-frequency component to the spectrum's middle shift = np.fft.fftshift(DFT) row, col = img.shapecenter_row, center_col = row // 2, col // 2 # create a mask with a centered square of 1s mask = np.zeros((row, col, 2), np.uint8) mask[center_row - 30:center_row + 30, center_col - 30:center_col + 30] = 1 # put the mask and inverse DFT in place. fft_shift = shift * mask

Input Image

fft_ifft_shift = np.fft.ifftshift(fft_shift)

calculate the magnitude of the inverse DFT

plt.subplot(121), plt.imshow(img, cmap='gray')

imageThen = cv2.magnitude(imageThen[:,:,0], imageThen[:,:,1])

visualize the original image and the magnitude spectrum

plt.title('Input Image'), plt.xticks([]), plt.yticks([])
plt.subplot(122), plt.imshow(imageThen, cmap='gray')

plt.title('Low pass filter'), plt.xticks([]), plt.yticks([])

imageThen = cv2.idft(fft_ifft_shift)

plt.figure(figsize=(10,10))

Low pass filter