



```

        if j + z - indexer < 0 or j + indexer > len(data[0]) - 1:
            temp.append(0)
        else:
            for k in range(filter_size):
                temp.append(data[i + z - indexer][j + k - indexer])

    temp.sort()
    data_final[i][j] = sum(temp[8:17])/9
    temp = []
return data_final

```

```

delnoise_img=AlphaTrimmedMean(img,5)
plt.imshow(delnoise_img,cmap='gray')
plt.show()

```

```

delnoise_img = cv2.copyMakeBorder(delnoise_img, 0, 800, 0, 800, cv2.BORDER_CONSTANT)
#padding
kid_DFT = np.fft.fft2(delnoise_img)
kid_DFT = np.fft.fftshift(kid_DFT)
M, N = kid_DFT.shape
print(kid_DFT.shape)

```

## ## BLPF AND GLPF

```

BLPF = np.zeros((M, N), dtype=np.float32)
GLPF = np.zeros((M, N), dtype=np.float32)
D01 = 200 # Butter worth LPF cutoff frequency (fixed)
D02 = [100, 150, 200, 250] # Gassian LPF cutoff frequency (changed)
B = 0.414
n = 5 # order
for D0 in D02:
    print(D0)
    for u in range(M):
        for v in range(N):
            D = np.sqrt((u-M/2)**2 + (v-N/2)**2)
            BLPF[u,v] = 1 / (1 + B*((D/D01)**(2*n))) # Butter worth LPF
            GLPF[u,v] = np.exp(-D**2/(2*D0*D0)) # Gassian LPF
Gshift = BLPF* (kid_DFT / GLPF)
kid_invshift = np.fft.ifftshift(Gshift)
kid_invDFT = np.abs(np.fft.ifft2(kid_invshift))
img = Image.fromarray(kid_invDFT[0:800, 0:800])
img = img.convert("L")
img.save('output/'+str(D0)+'RESULT.jpg', dpi=(200, 200))

```

(b) Results of noise model and model parameters:

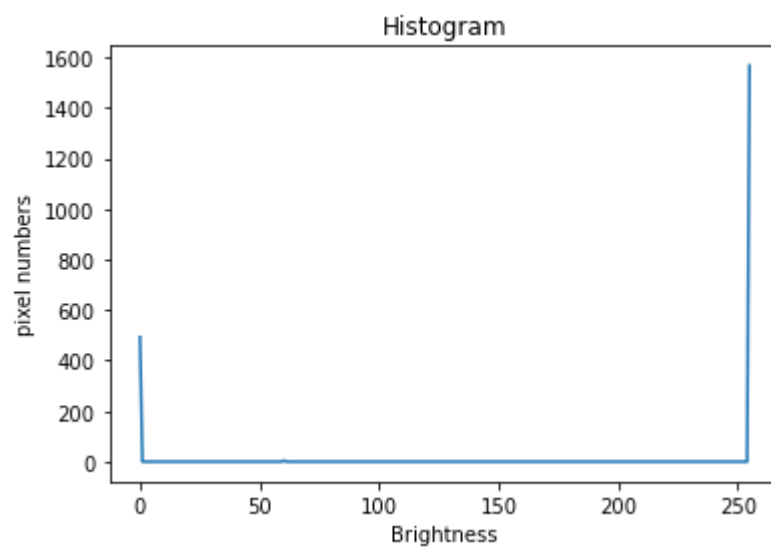
裁切左上角 100x100 的區域:

$P_a$  為黑點佔該區域的 probability

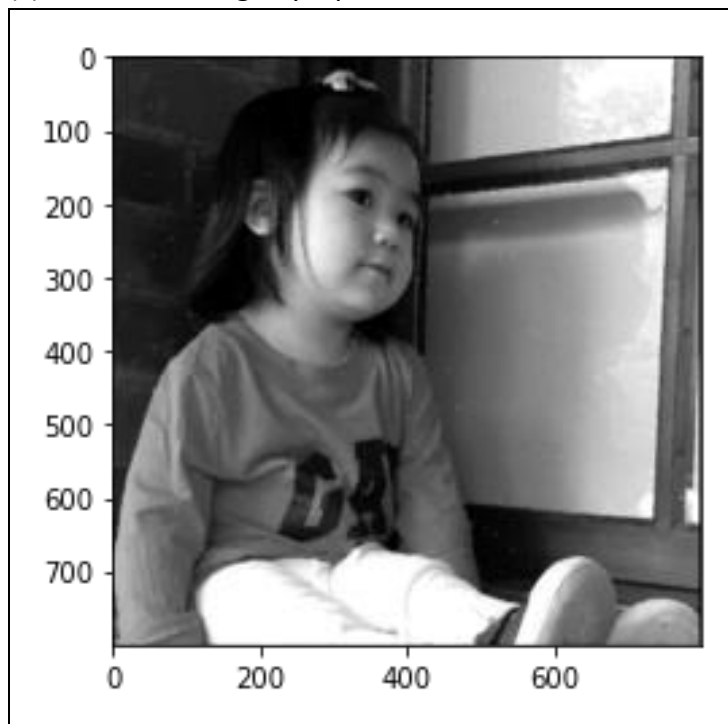
$P_b$  為白點佔該區域的 probability

$P_a = 0.0493$

$P_b = 0.1568$



(c) De-noised image by alpha-trimmed mean filter:

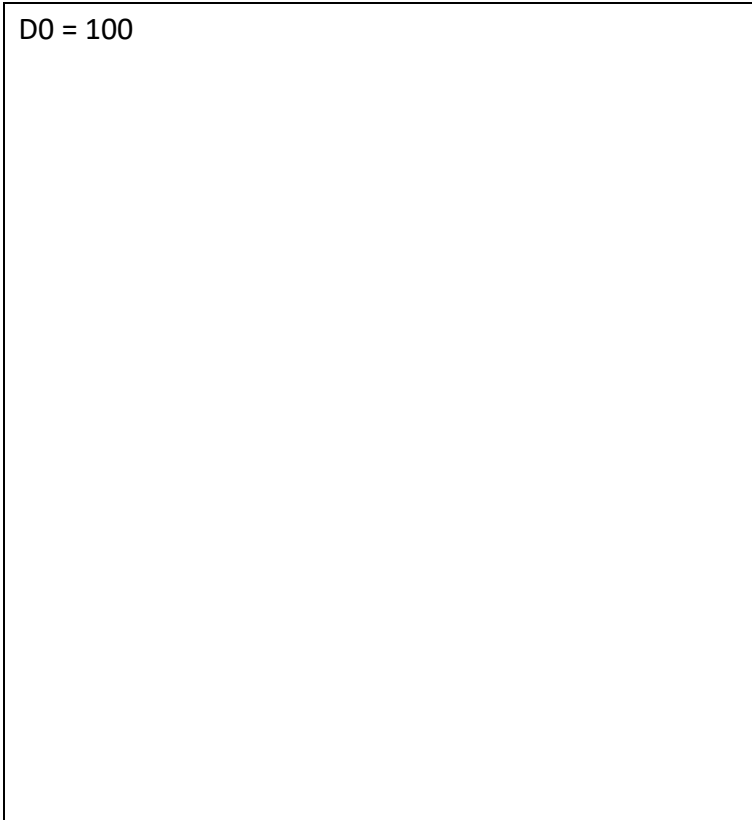


(d) Output image、parameters:

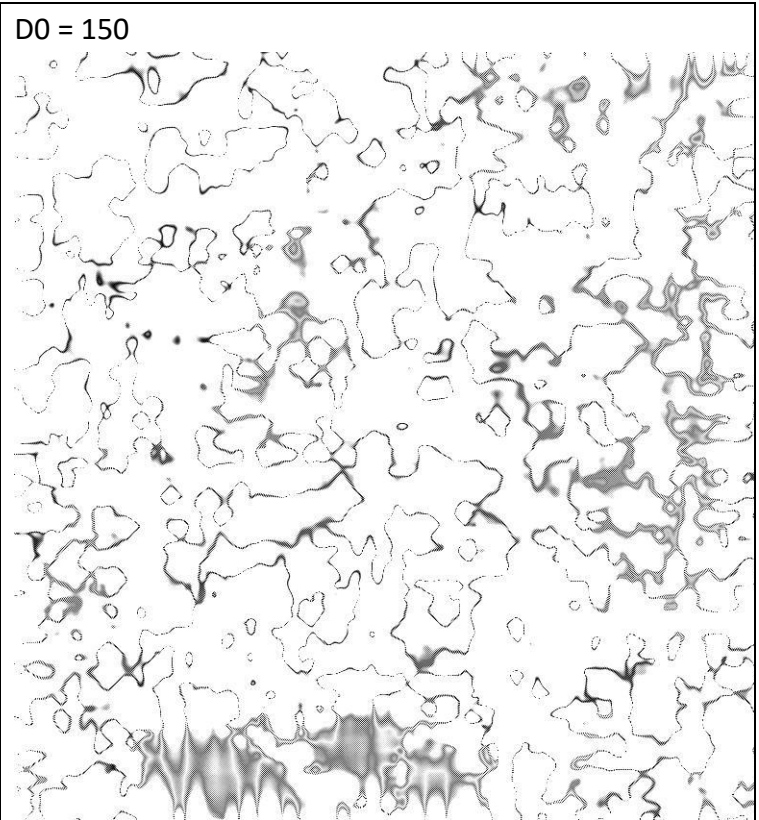
固定 Butter worth LPF (  $n = 5$ , and  $D0 = 200$  )

使用 Gaussian LPF 做 inverse filter，分別使用  $D0 = 100, 150, 200, 250$  來做，從上圖的結果可以得知當 cutoff frequency 越小的時候，所能得到的頻率範圍越小，所以只能得出一張白色的圖案，將  $D0$  條大之後，所能得到的頻率範圍變大，所得到的圖片也會越清晰。

$D0 = 100$



$D0 = 150$



$D0 = 200$



$D0 = 250$

