

20210510攝影測量報告

組員：

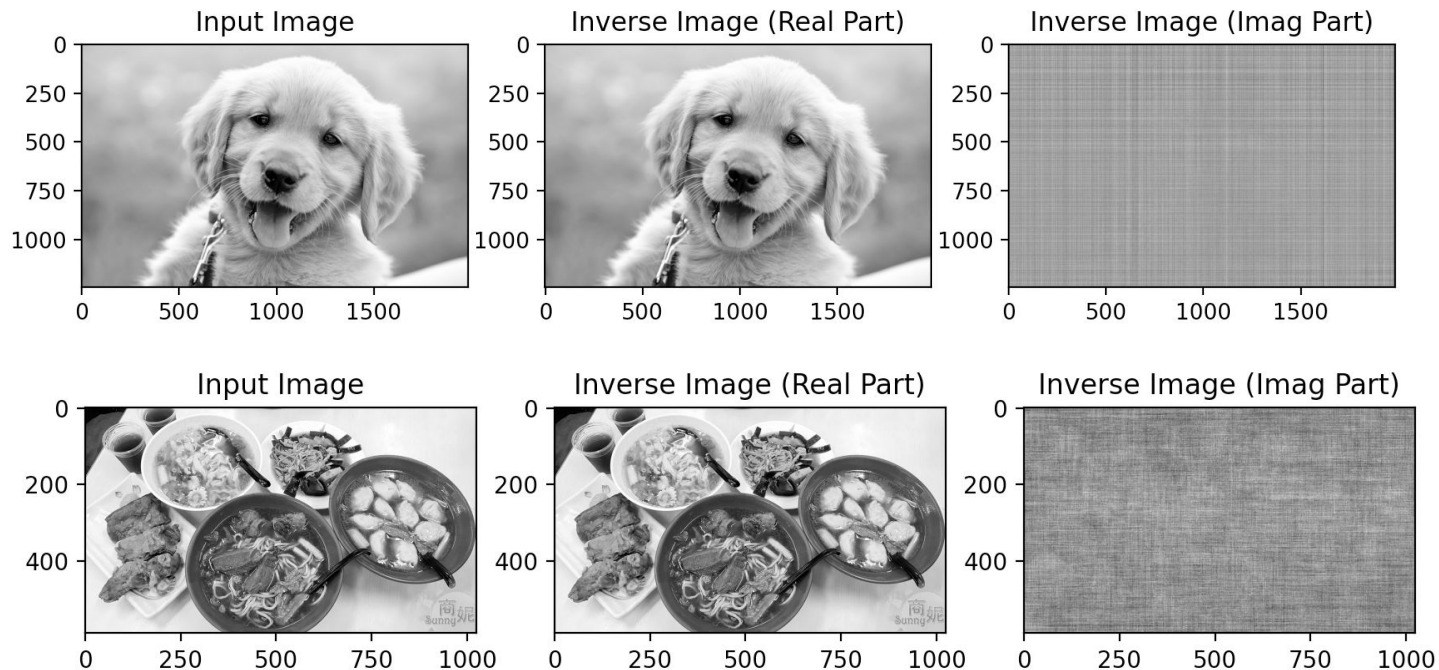
B06501015 土木四 周遠同

B06501103 土木四 何文杰

B06501075 土木四 潘佳旻

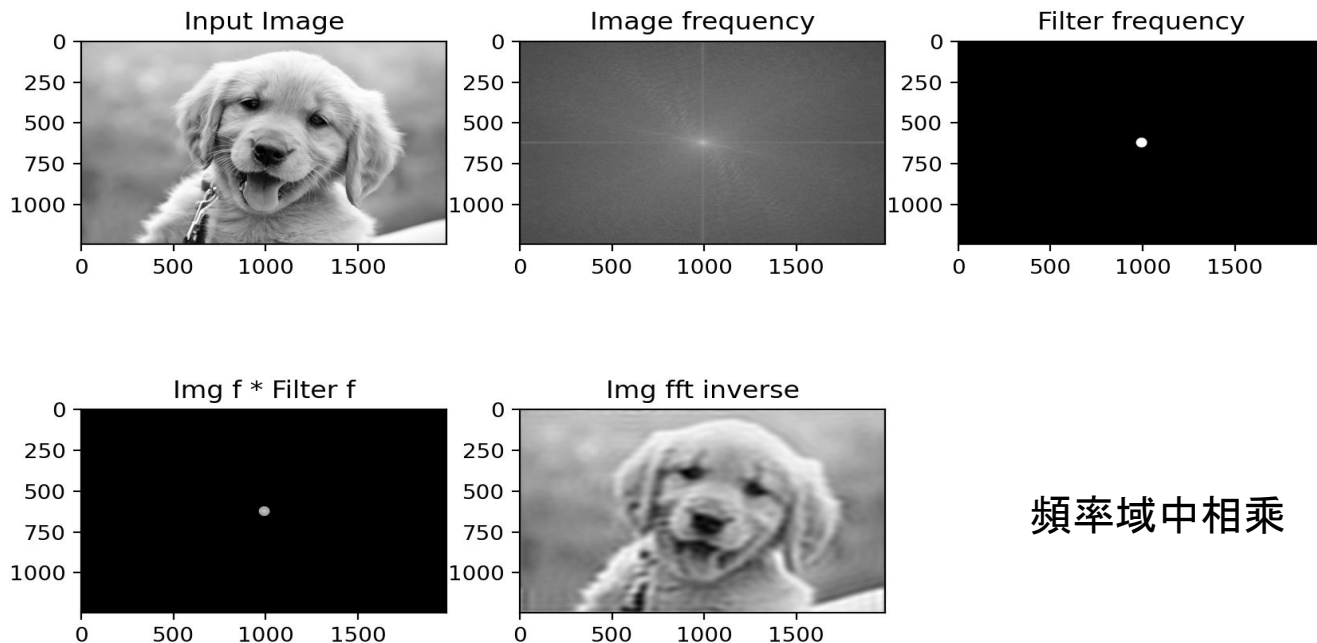
一、ifft2後，實部與虛部的比較

- 在頻率域相乘後，轉回空間域需要用ifft2，會產生實數、虛數



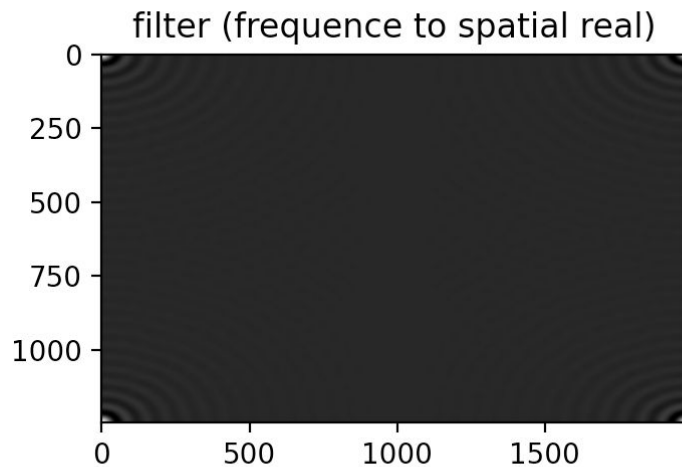
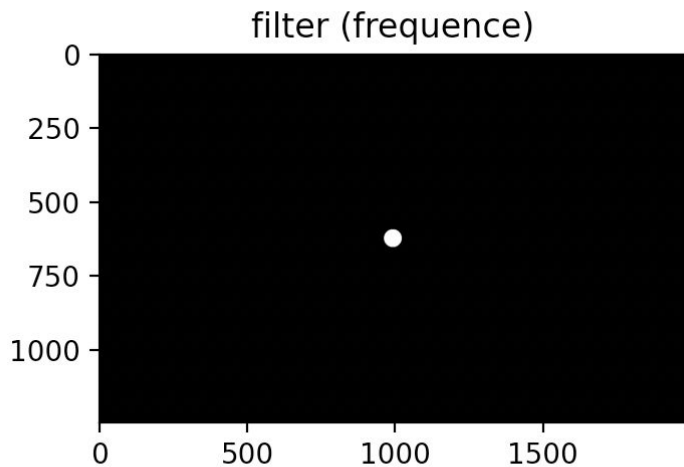
二、濾波函數是在空間域中或頻率域中設計？

- (1) 在頻率域中設計：

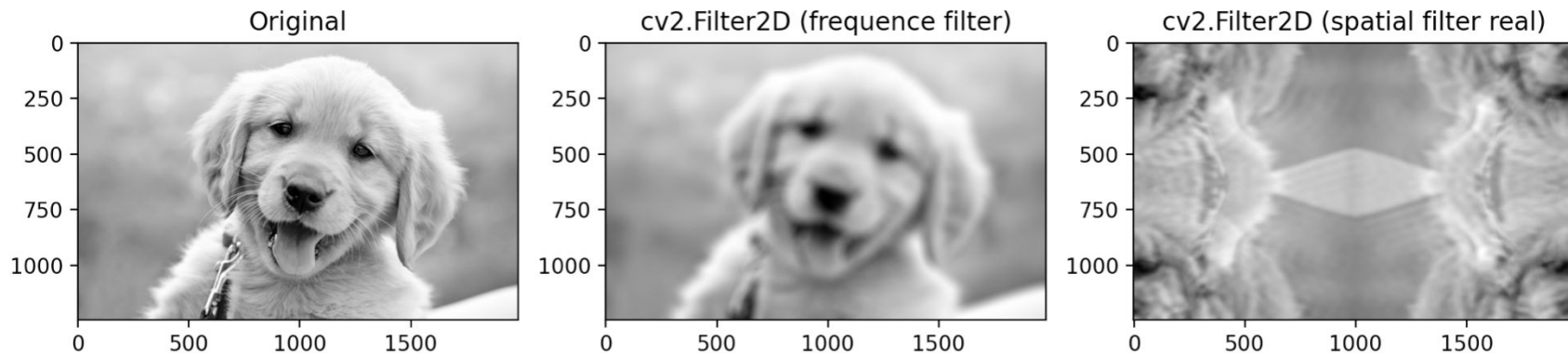


頻率域中相乘

- 在空間域做convolution時，不確定要將filter轉換到空間域還是維持在頻率域 → 都試試看

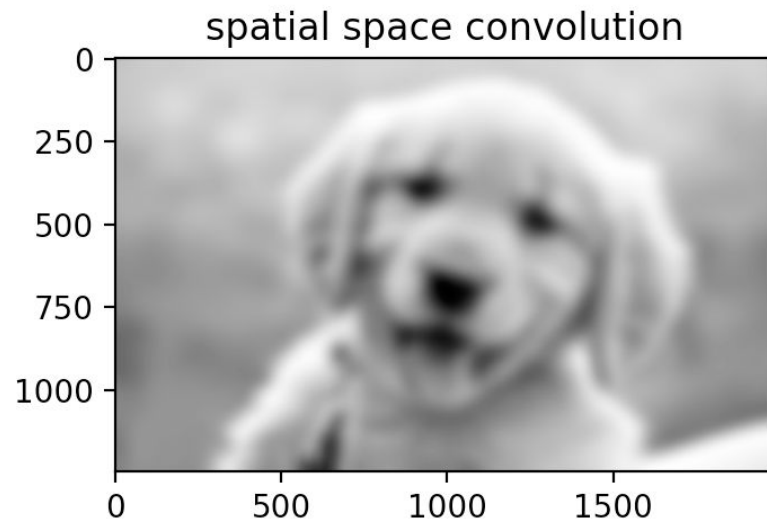
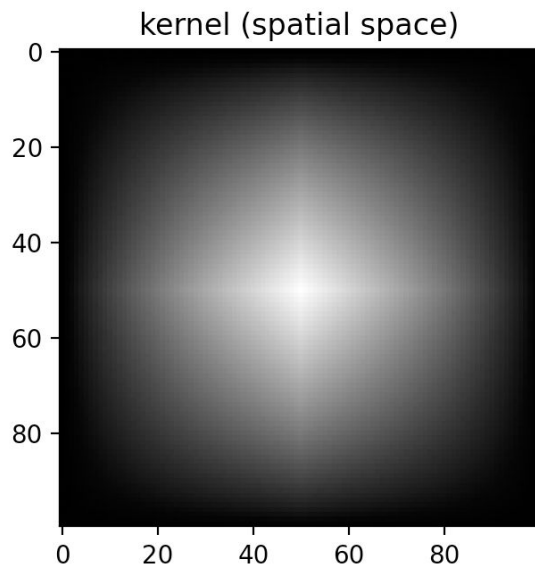


- 頻率域、空間域的filter都convolution試試看



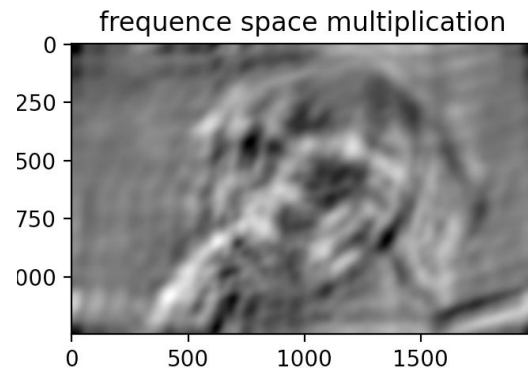
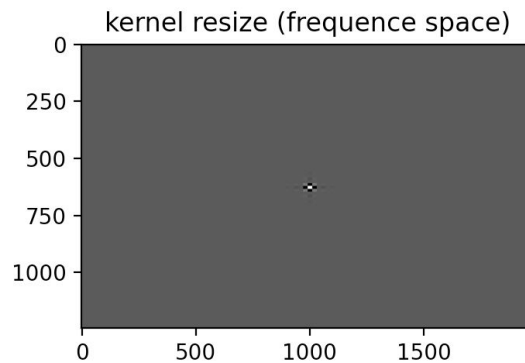
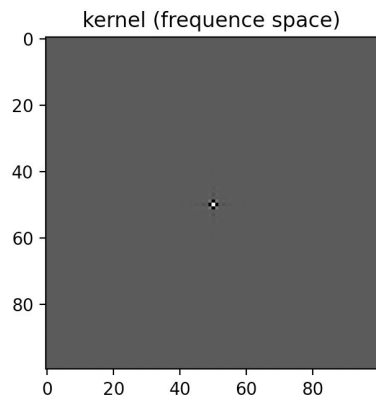
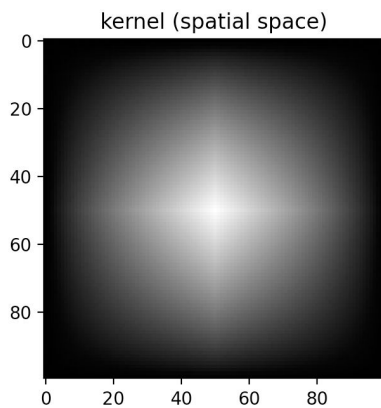
- 頻率域中的filter, convolution後還算正常, 但與頻率域中相乘的仍有差別
- 轉到空間域中的filter, convolution後變得很奇怪

- (2) 在空間域中設計

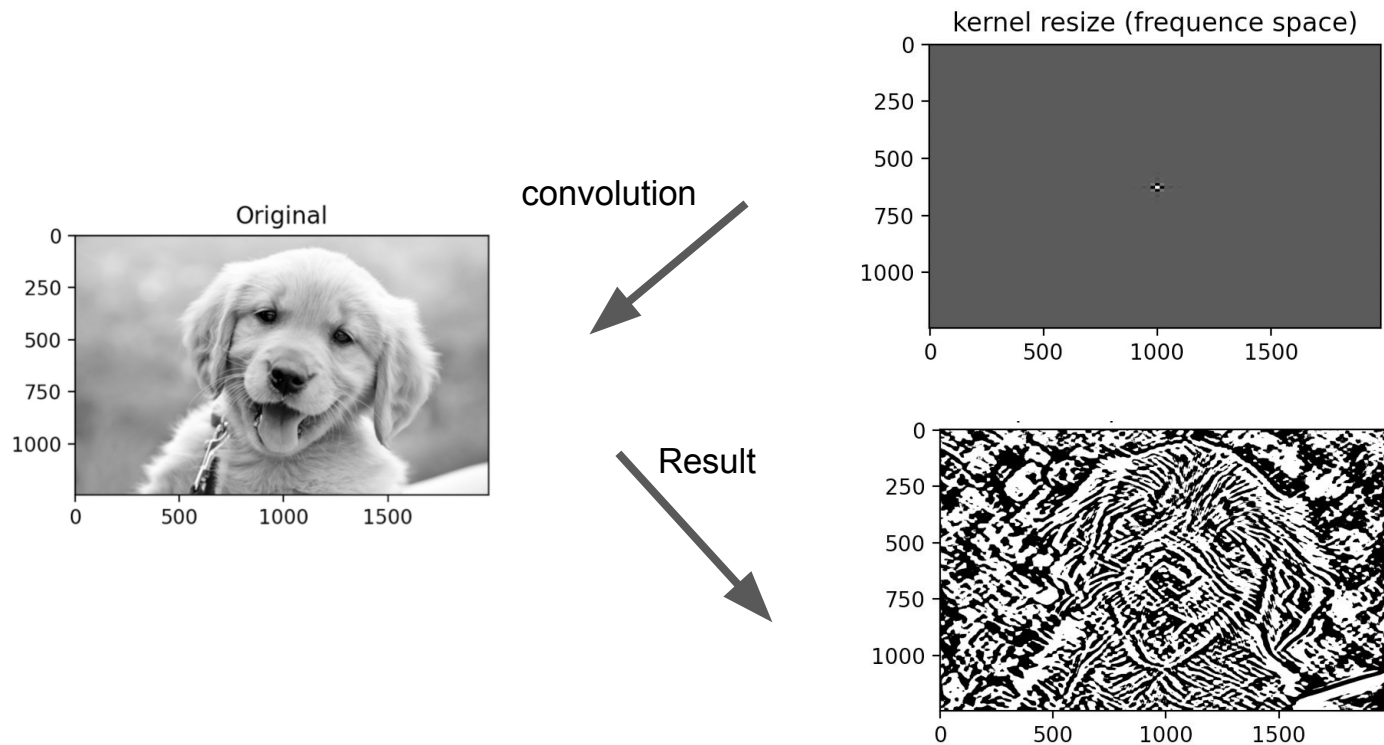


空間域做convolution

- 將kernel 轉換到頻率域
- 在頻率域中相乘, 再轉回空間域



- 亂試試看：
- 將轉到頻率域的kernel, 與image做convolution



三、比較不同濾波函數的過濾效果與差異

- 濾波器

Input Image



high pass filter

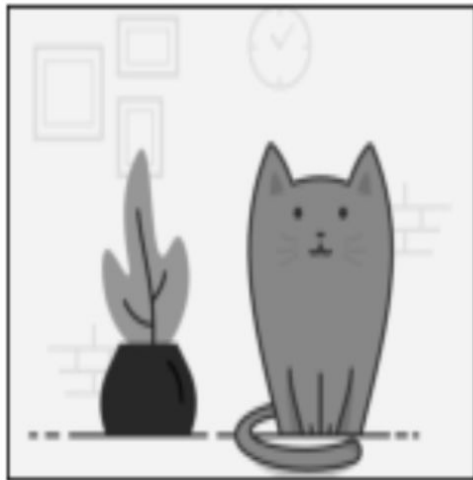


low pass filter



濾波器

Input Image



bandreject filters



濾波器

Input Image



guais low pass

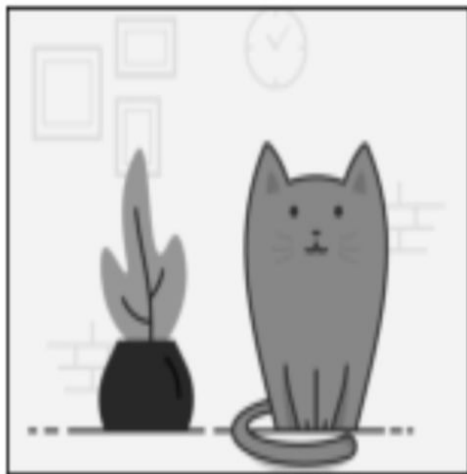


guais high pass



濾波器

Input Image



laplacian filter



濾波器混合

LF+GHPF



BF+GHPF

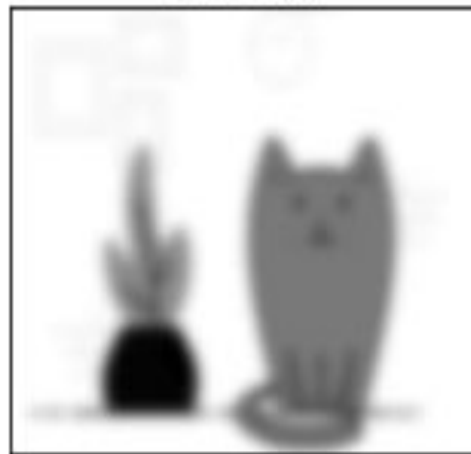


混合濾波器

LF+GLPF



BF+GLPF



混合濾波器

LF+BF



LF+BF+GLPF

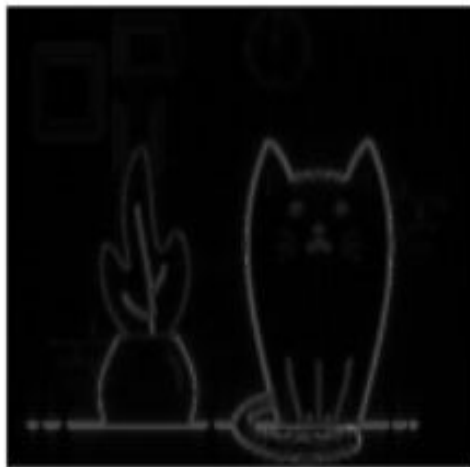


HPF 濾波半徑不同

HPF $r=40$



HPF $r=150$



LPF 濾波半徑不同

LPF $r=40$



LPF $r=150$



四、比較numpy與OpenCV的計算、效能差異



1. Developer: Travis Oliphant
2. Functional: A large collection of high-level mathematical functions to operate large and multi-dimensional arrays.
3. Program: Python



1. Developer: Intel
2. Functional: real-time ray tracing and 3D displaying walls.
3. Program: C/C++

OpenCV2 optimization

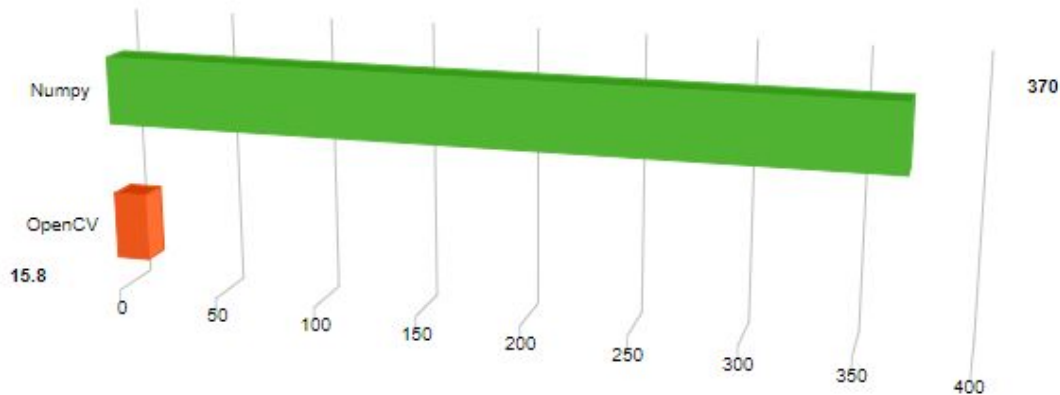
- **cv2.useoptimized()**
 - Recalled this function to activate optimized version of CV2.
- **cv2.cv2.setUseOptimized(False)**
 - Recalled this function to de-activate optimized version of CV2.
- **Optimized version able to shorten the duration by half.**

```
In [30]: runfile('D:/Civil Engineering/Senior/  
Photogrammetry/cv2 with optimization performance.py'  
wdir='D:/Civil Engineering/Senior/Photogrammetry')  
0.0195031
```

```
In [31]: runfile('D:/Civil Engineering/Senior/  
Photogrammetry/cv2 without optimization (median).py'  
wdir='D:/Civil Engineering/Senior/Photogrammetry')  
0.0389865
```

Time Performance (Numpy vs OpenCV)

- Same condition: images, task, best of three results.
- `cv2.countNonZero()`
 - Recalled this to calculate non-zero values by openCV.
- `np.count_nonzero()`
 - Recalled this to calculate non-zero values by numpy.
- **OpenCV 23.42x faster than Numpy.**

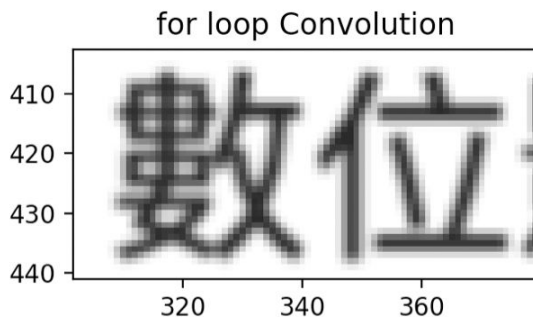
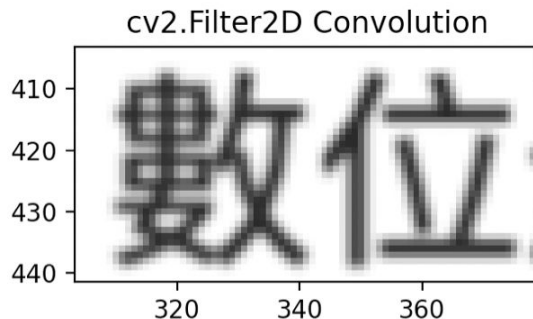
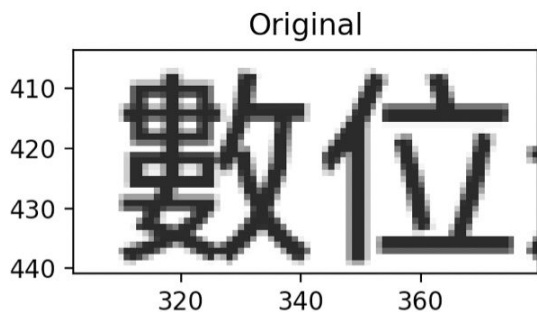


Conclusion & Limitations

- **OpenCV functions over Numpy for same calculations. However, Numpy are faster on views instead of copies.**
- **Both functions are designed for vector optimization, thus try to vectorize the code to the maximum extent.**
- **Copying an array significantly increase the time of calculations. Therefore, try to use view over copy.**
- **OpenCV has faster rate of computation by 20x faster.**

五、cv2.filter2D vs. 自己寫的convolution

```
kernal = np.array((  
    [0.0625, 0.125, 0.0625],  
    [0.125, 0.25, 0.125],  
    [0.0625, 0.125, 0.0625]))  
convolution1 = cv2.filter2D(img,-1,kernal)  
convolution2 = convolution_func.my_convolution(img, kernal)
```



- **cv2.filter2D()**
 - Actually it is correlation, not convolution
 - If the kernel is too big (kernel > 11x11), use DTF algorithm

$$\text{dst}(x, y) = \sum_{\substack{0 \leq x' < \text{kernel.cols}, \\ 0 \leq y' < \text{kernel.rows}}} \text{kernel}(x', y') * \text{src}(x + x' - \text{anchor.x}, y + y' - \text{anchor.y})$$

- my for loop convolution

```
def my_convolution(image, kernel):
    m, n = image.shape[0], image.shape[1]
    a, b = kernel.shape[0], kernel.shape[1]
    h, w = m-a+1, n-b+1
    new = np.zeros((h,w))

    # for every point in the conv matrix
    for i in range(h):
        for j in range(w):
            # calculate the aggregated multiplication value in the kernel
            new[i, j] = np.multiply(image[i:i+a, j:j+b], kernel).sum()

    return new
```

- cv2.filter2D, My Convolution效能差異、比較:

```
(base) zhouyuantongdeMacBook-Pro:Lab2 tony$ python3 filter2D_test.py
Total Round: 10
cv2.filter2D time : 0.000661
my convolution time: 1.867547
-----
(base) zhouyuantongdeMacBook-Pro:Lab2 tony$ python3 filter2D_test.py
Total Round: 100
cv2.filter2D time : 0.005684
my convolution time: 21.286446
-----
(base) zhouyuantongdeMacBook-Pro:Lab2 tony$ python3 filter2D_test.py
Total Round: 1000
cv2.filter2D time : 0.050415
my convolution time: 224.596234
-----
```


- **cv2.filter2D, FFT 效能差異、比較:**

```
(base) zhouyuantongdeMacBook-Pro:Lab2 tony$ python3 TimeComparison.py
Total Round: 10
cv2.filter2D time      : 0.017779
Frequency space multiply: 0.033480
-----
(base) zhouyuantongdeMacBook-Pro:Lab2 tony$ python3 TimeComparison.py
Total Round: 100
cv2.filter2D time      : 0.165138
Frequency space multiply: 0.316796
-----
(base) zhouyuantongdeMacBook-Pro:Lab2 tony$ python3 TimeComparison.py
Total Round: 1000
cv2.filter2D time      : 1.648850
Frequency space multiply: 3.681419
-----
```

cv2.filter2D	>	FFT	>	my convolution
4000	>	2000	>	1

- Why OpenCV function can be so fast !?
- The Convolution Theorem, Seperable Convolution

1. The Convolution Theorem

- 偷偷轉到頻率域做相乘, 再轉回來

2. Seperable Convolution

- 將kernal矩陣拆成兩個向量相乘
- $\text{img} * (\text{v} * \text{h}) = (\text{img} * \text{v}) * \text{h}$
- 原先: $O(M*N*P*Q)$
- 變成: $O(M*N*(P+Q))$

$$\frac{1}{256} \cdot \begin{bmatrix} 1 & 4 & 6 & 4 & 1 \\ 4 & 16 & 24 & 16 & 4 \\ 6 & 24 & 36 & 24 & 6 \\ 4 & 16 & 24 & 16 & 4 \\ 1 & 4 & 6 & 4 & 1 \end{bmatrix} = \frac{1}{256} \cdot \begin{bmatrix} 1 \\ 4 \\ 6 \\ 4 \\ 1 \end{bmatrix} \cdot [1 \quad 4 \quad 6 \quad 4 \quad 1]$$

5x5 Seperable Gaussian Kernel

參考資料

<https://docs.opencv.org/2.4/modules/imgproc/doc/filtering.html#filter2d>

<https://stackoverflow.com/questions/31336186/opencv-computational-efficiency-of-filter2d-function>

<https://makerpro.cc/2019/06/the-convolution-of-opencv/>

<https://www.cnblogs.com/lfri/p/10599420.html>

<https://dsp.stackexchange.com/questions/8471/where-does-convolution-fit-in-dft>

<https://zh.wikipedia.org/wiki/快速傅里叶变换>

分工

- 依照上述投影片五個題目來分：
- 一、周遠同
- 二、周遠同
- 三、潘佳旻
- 四、何文杰
- 五、周遠同

Thanks for listening