

POSCO AI Course: Computer Vision

# Image Filtering

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# OpenCV



# OpenCV

- OpenCV (Open Source Computer Vision Library: <http://opencv.org>) is an open-source library that includes several hundreds of computer vision algorithms.
- C++, Python API
  - OpenCV Python API makes use of Numpy.

# Basic IO

- Read image

`retval=cv.imread(filename[, flags])`

- Returns Numpy ndarray

- **Note:** RGB images are loaded in BGR format!!!

- Need to swap B and R channels if you want to use the image with other libraries.

- Write image

`retval=cv.imwrite(filename, img[, params])`

- For color image, expects BGR format image.

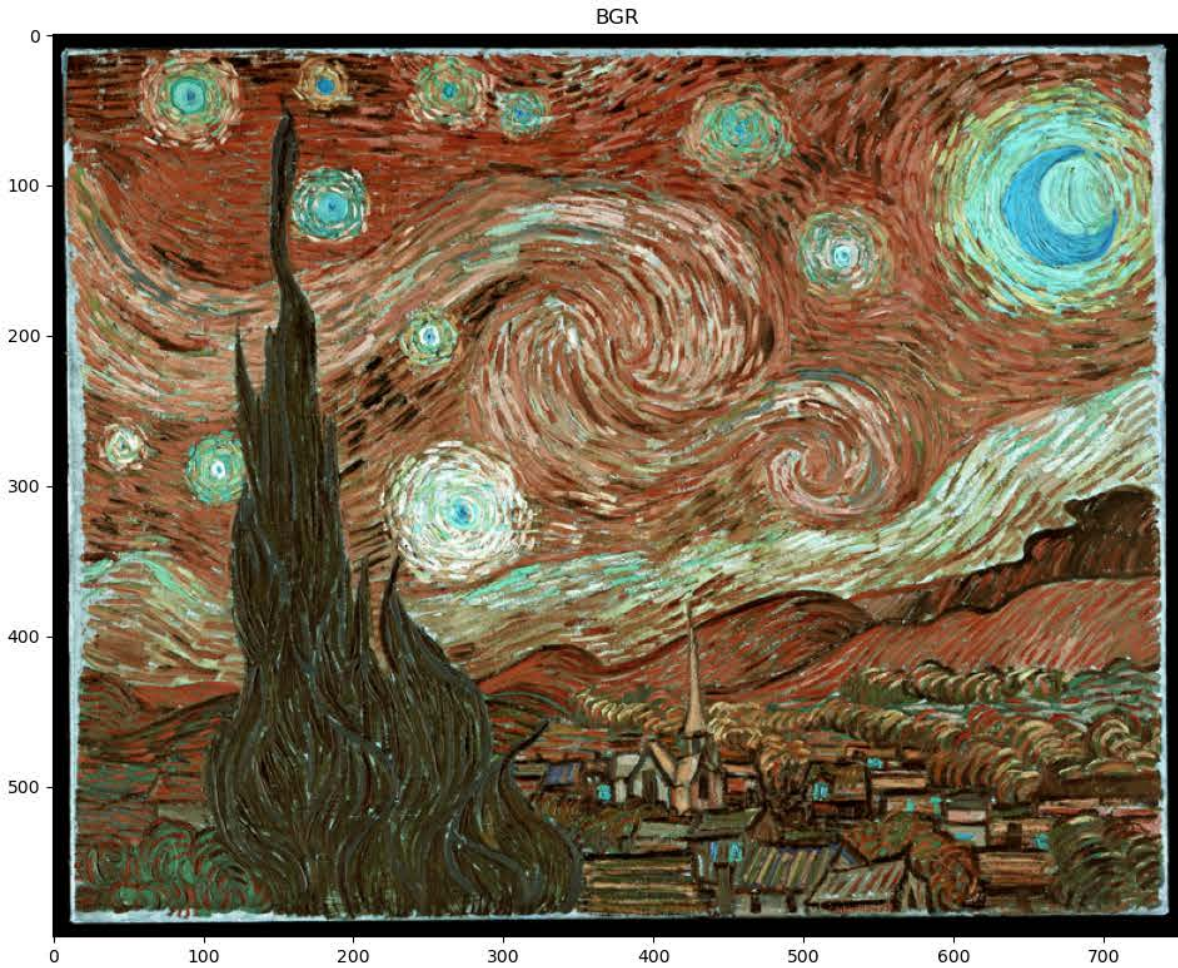
- See official OpenCV documentation

(<https://docs.opencv.org/4.4.0/index.html>) for more information.



# Basic IO

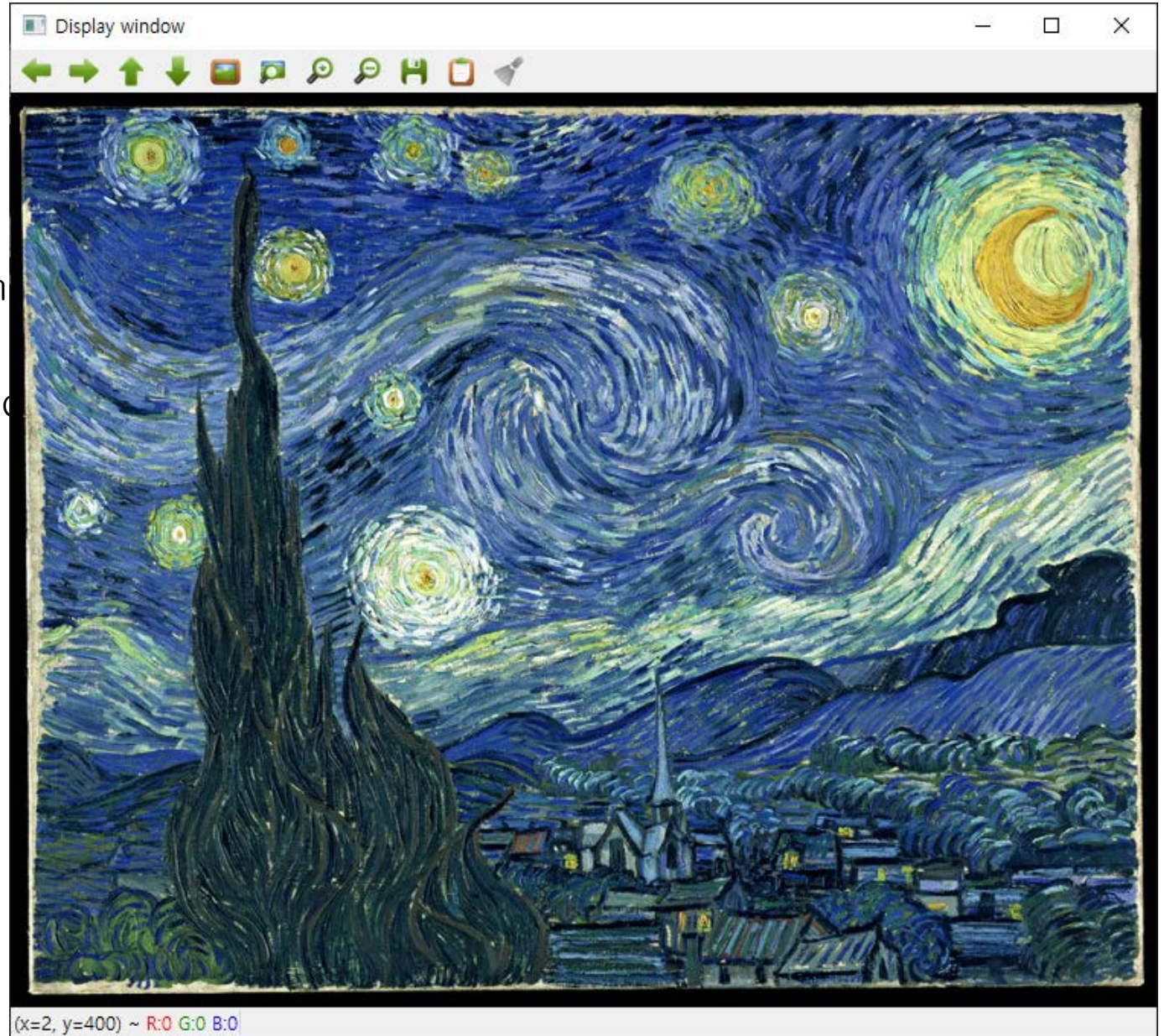
- BGR vs RGB





# Show Image using OpenCV

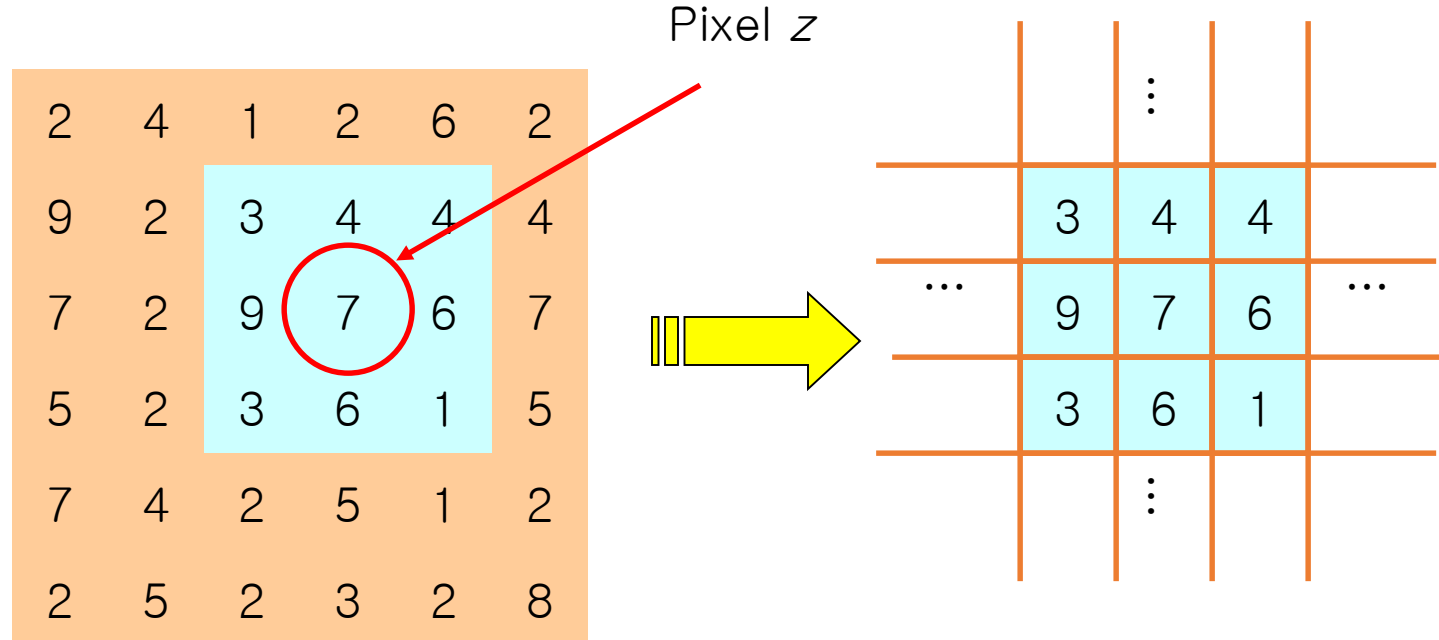
- Show image
  - `None=cv.imshow(winname, mat)`
  - Displays image in a window.
  - No need for convert images to RGB format
  - Able to specify GUI behavior (not covered)
  - **Note:** some issues with Jupyter





# Linear filters

Step 1. Select only needed pixels



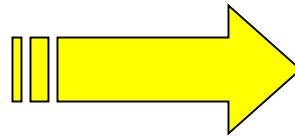
# Linear filters

Step 2. Multiply every pixel by kernel and then sum up the values

		⋮		
	3	4	4	
...	9	7	6	...
	3	6	1	
		⋮		

x

$\frac{1}{9}$	1	1	1
	1	1	1
	1	1	1



$$\begin{aligned} y = & \frac{1}{9} \cdot 3 + \frac{1}{9} \cdot 4 + \frac{1}{9} \cdot 4 \\ & + \frac{1}{9} \cdot 9 + \frac{1}{9} \cdot 7 + \frac{1}{9} \cdot 6 \\ & + \frac{1}{9} \cdot 3 + \frac{1}{9} \cdot 6 + \frac{1}{9} \cdot 1 \end{aligned}$$



# Linear filters

Step 1: Move the window to the first location where we want to compute the average value and then select only pixels inside the window.

2	4	1	2	6	2
9	2	3	4	4	4
7	2	9	7	6	7
5	2	3	6	1	5
7	4	2	5	1	2

Original image



2	4	1
9	2	3
7	2	9

Sub image  $p$



	4.3		

Output image

Step 4: Move the window to the next location and go to Step 2

Example : 3x3 averaging kernel

$\frac{1}{9}$	1	1	1
	1	1	1
	1	1	1

Step 2: Compute the filtered value

$$y(x, y) = \sum_{i, j=-\infty}^{\infty} f(i, j) I(x - i, y - j)$$

Step 3: Place the result at the pixel in the output image

# Image Filtering using OpenCV

- 2D Correlation

$$y(x, y) = \sum_{i, j=-\infty}^{\infty} f(i, j)I(i - x, j - x)$$

- `dst=cv.filter2D(src, ddepth, kernel[, dst[, anchor[, delta[, borderType]]])`
- Compute 2D correlation of image `src` and `kernel`. The filter (or the image) is not flipped.
- `ddepth`: bit depth of outout, set `ddepth` to `-1` to retain bit depth of input

- 2D Convolution

$$y(x, y) = \sum_{i, j=-\infty}^{\infty} f(i, j)I(x - i, y - j)$$

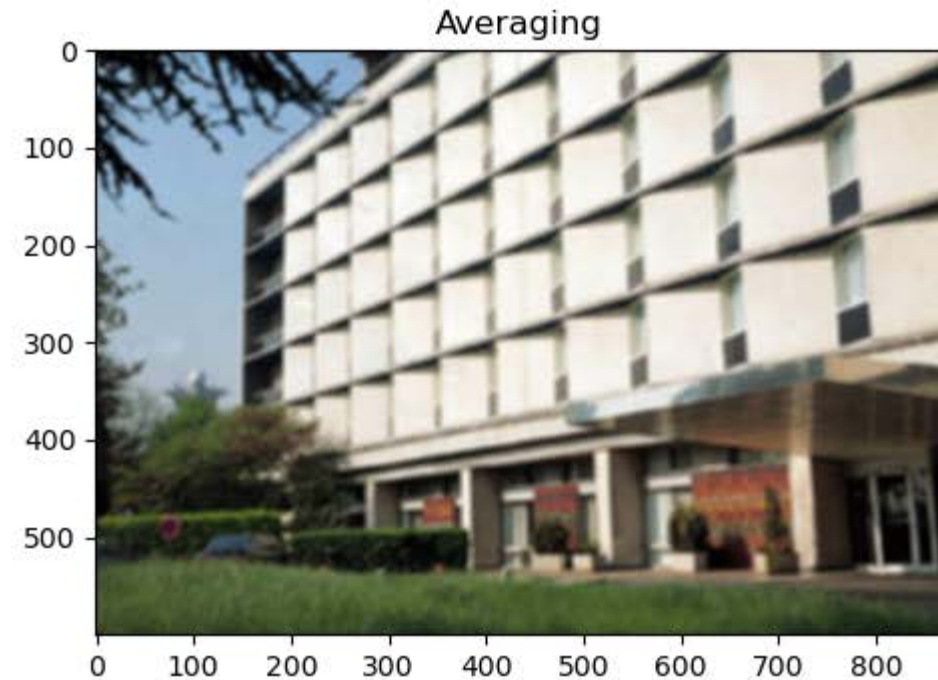
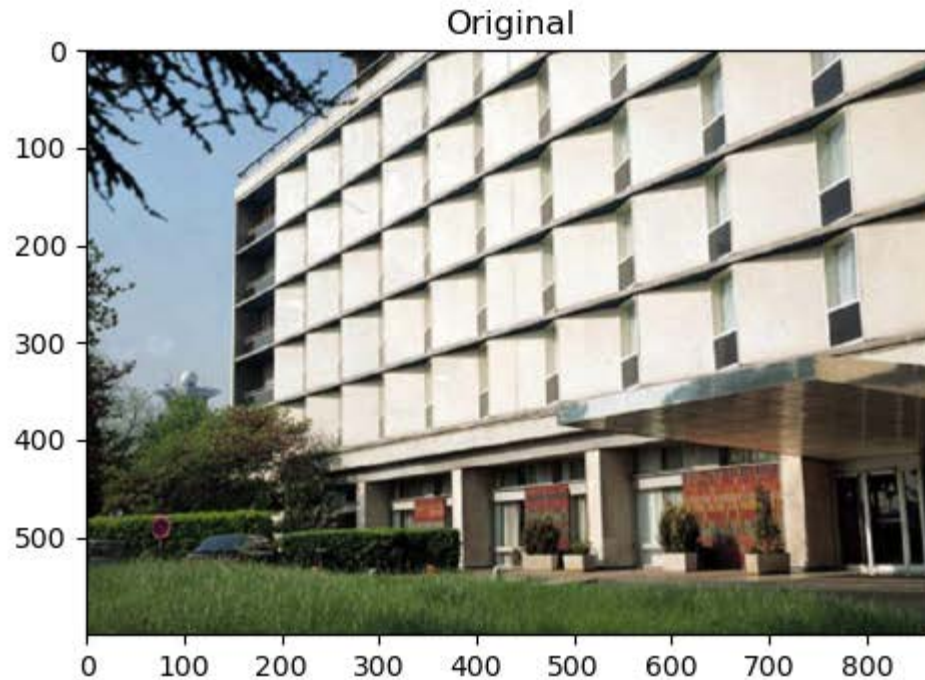
- There are no convolution function in OpenCV.
- The filter (or the image) is flipped.
- If you need convolution, you need to flip kernel and use `filter2D` function.

# Box filter

- Box filter (Average filter)
  - Averages pixels in a box shaped window.
  - Sum of kernel should be 1

kernel  $\frac{1}{9}$

1	1	1
1	1	1
1	1	1



# Other filters

input



filter

0	0	0
0	1	0
0	0	0

output



unchanged

input



filter

0	0	0
0	0	1
0	0	0

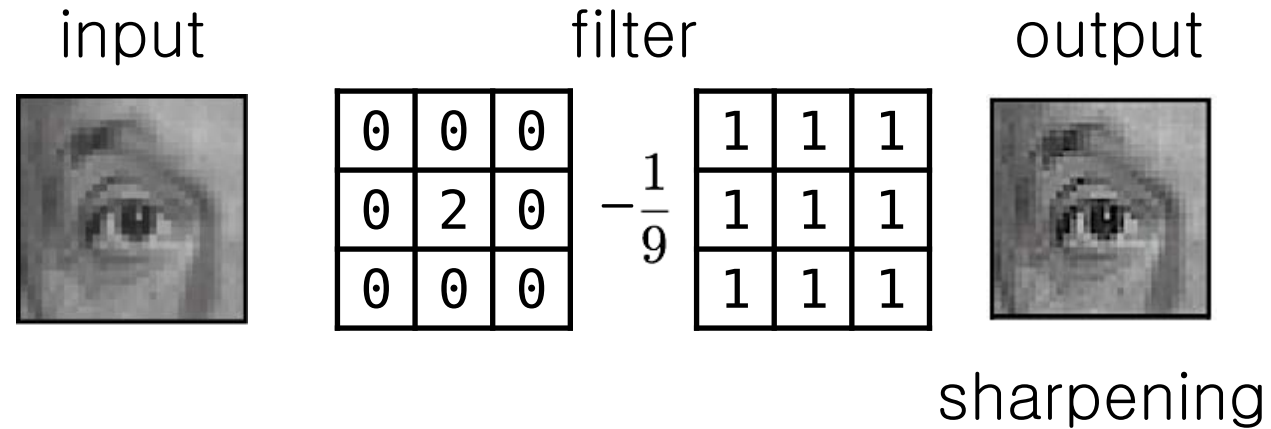
output



shift to  
left by  
one



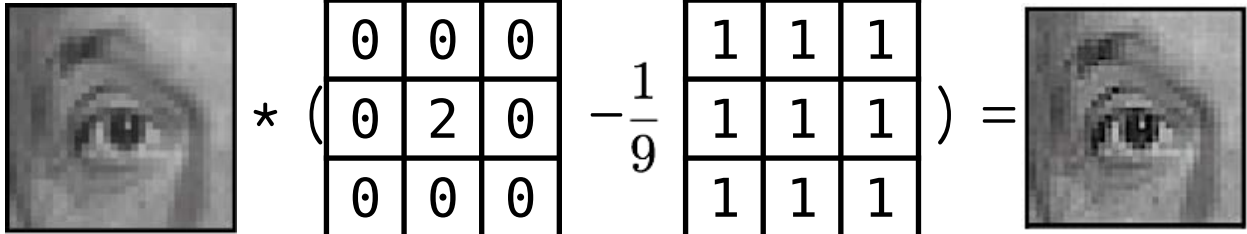
# Sharpening filters



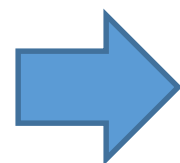
- do nothing for flat areas
- stress intensity peaks

# Sharpening filters

Input  $g$                       filter  $f$                       output ( $g * f$ )



$$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 0 \end{bmatrix} - \frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} =$$



$$\begin{aligned}
 (g * f) &= g * \left( \begin{bmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 0 \end{bmatrix} - \frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \right) \\
 &= g * \left( \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} - \frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \right) \\
 &= g * \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} + g * \left( \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} - \frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \right)
 \end{aligned}$$

High-pass filter  
To obtain high-frequency details

\* A high-pass filter can be obtained by subtracting a low-pass filter from a delta function.

# Quiz 1. Sharpening filters

- Quiz: Sharpening 필터 구현

- 박스 필터를 이용하여 sharpening 필터를 구현

- 박스 필터의 크기는 7으로 구현.

- 커널의 합이 1이어야 함

- 구현 결과 이미지를 미리 저장된 이미지와 비교하여 테스트

input



filter

0	0	0
0	2	0
0	0	0

 $-\frac{1}{9}$ 

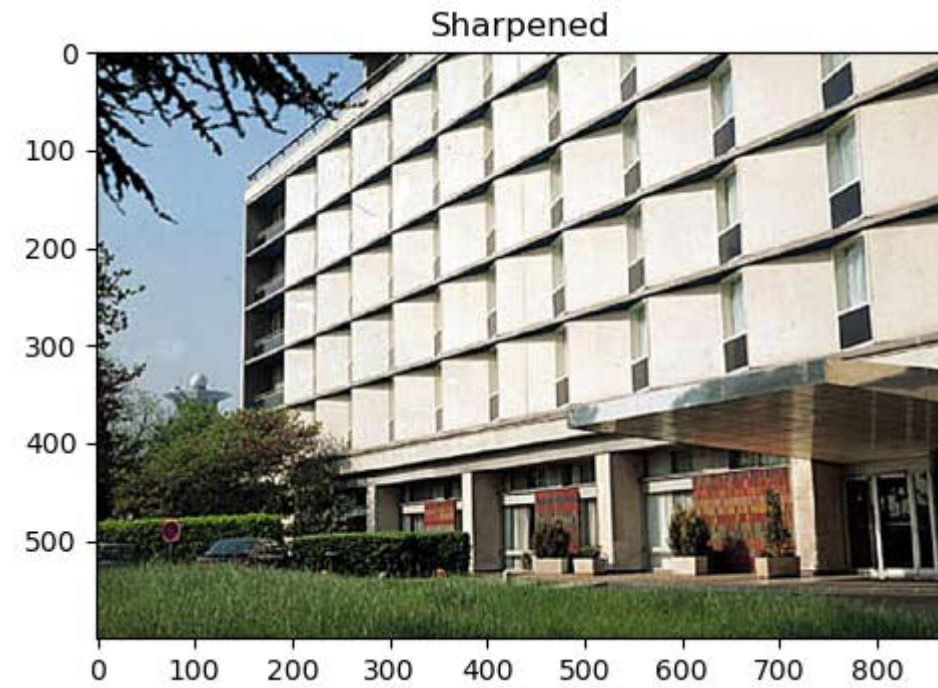
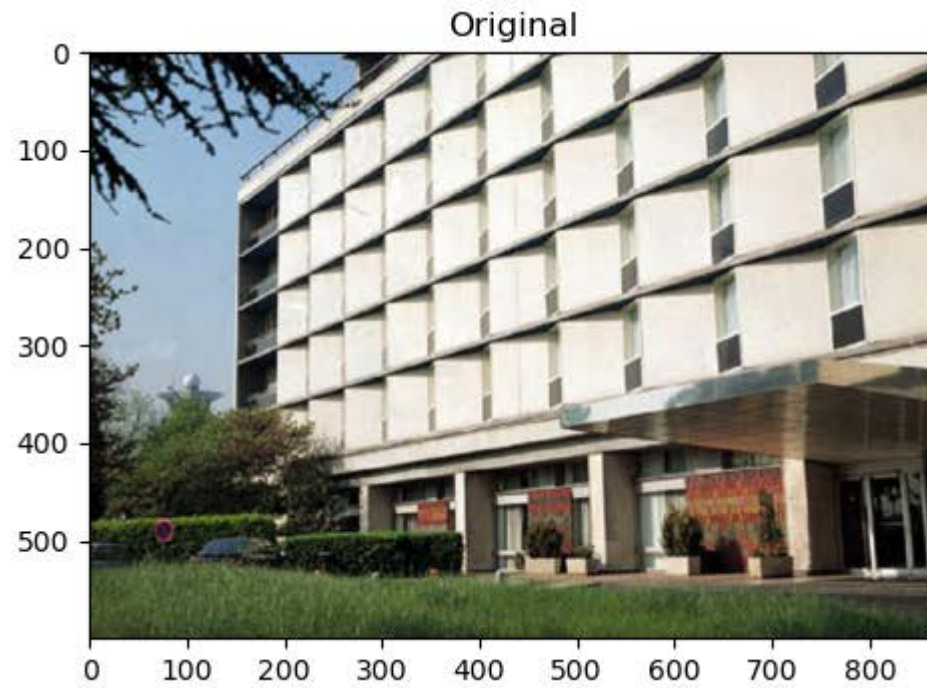
1	1	1
1	1	1
1	1	1

output



sharpening

# Sharpening examples



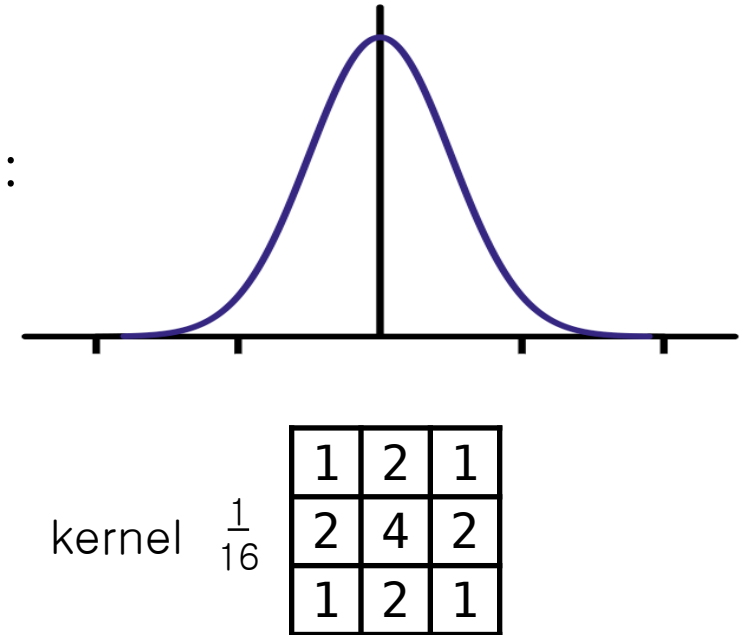


# The Gaussian filter

- named (like many other things) after Carl Friedrich Gauss
- Most representative low-pass filter
- kernel values sampled from the 2D Gaussian function:

$$f(i,j) = \frac{1}{2\pi\sigma^2} e^{-\frac{i^2+j^2}{2\sigma^2}}$$

- weight falls off with distance from center pixel
- theoretically infinite, in practice truncated to some maximum distance
  - Any heuristics for selecting where to truncate?
    - usually at  $2-3\sigma$



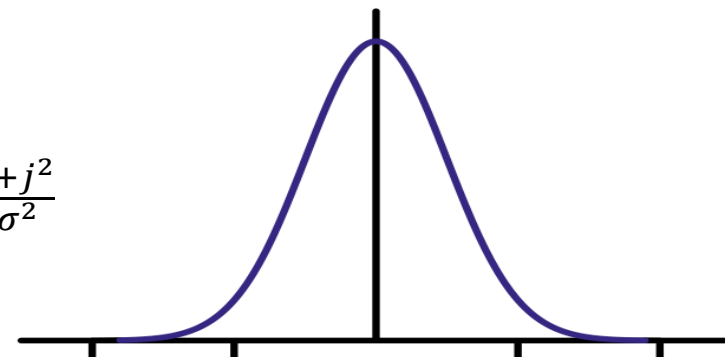
# Quiz 2. Gaussian filter

- Quiz: Gaussian filter 필터 구현

- filter2D 를 사용하여 Gaussian filter를 구현

- cv.GaussianBlur 사용 금지

$$f(i,j) = \frac{1}{2\pi\sigma^2} e^{-\frac{i^2+j^2}{2\sigma^2}}$$

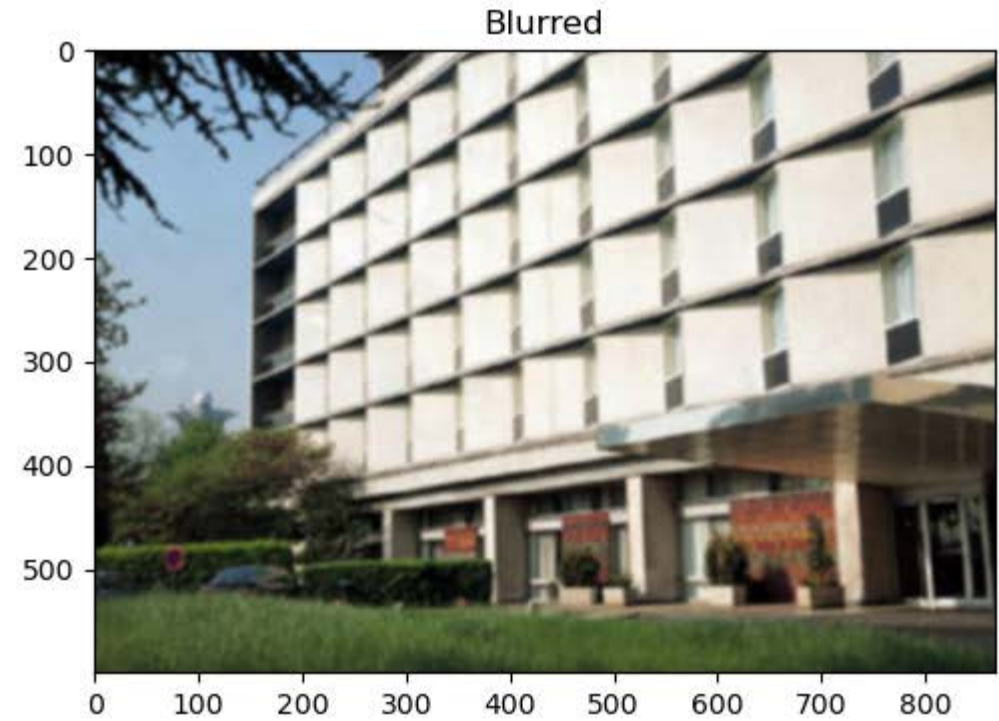
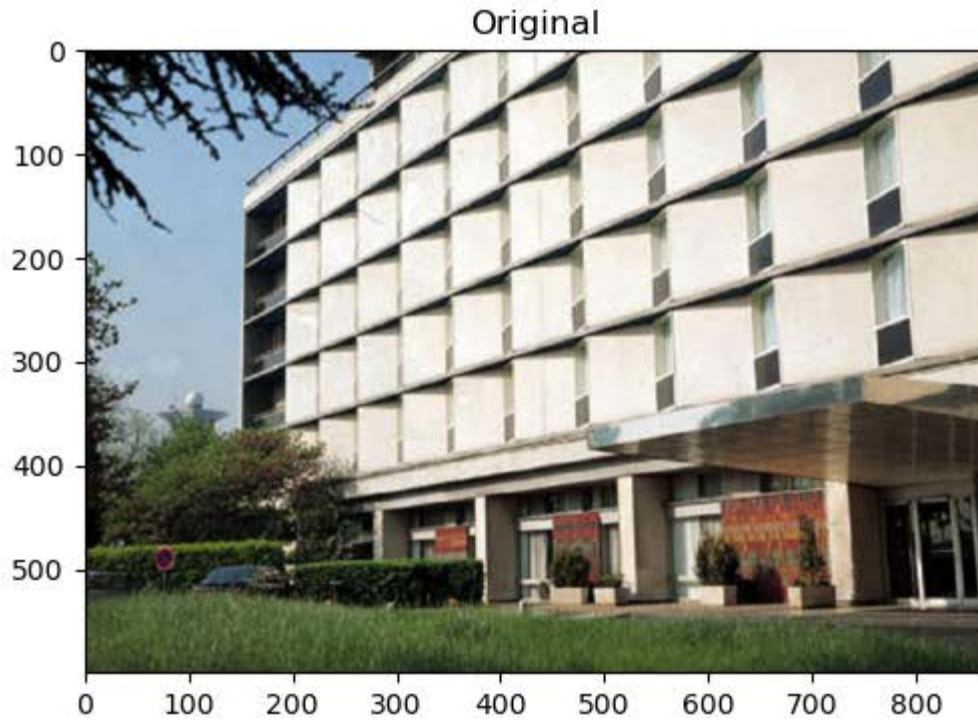


- 구현에 따라 결과 이미지가 차이가 있을수 있으므로 다음과 같이 구현

- $\sigma = 2$
- 커널의 크기는 7
- 커널의 합이 1이어야 함

- 구현 결과 이미지를 미리 저장된 이미지와 테스트

# Gaussian filtering example



# Gradients

Computing finite differences can be implemented using convolution operations

$$\frac{\partial f(x, y)}{\partial x} = f(x + 1, y) - f(x, y)$$

Forward finite difference

1	-1	0
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Note that the kernel is flipped because of the definition of convolution!

$$\frac{\partial f(x, y)}{\partial x} = f(x, y) - f(x - 1, y)$$

Backward finite difference

0	-1	1
---	----	---

$$\frac{\partial f(x, y)}{\partial x} = \frac{f(x + 1, y) - f(x - 1, y)}{2}$$

Central finite difference

0.5	0	-0.5
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# The Sobel filter

A combination of central finite difference and Gaussian filters

Horizontal Sobel filter:

$$\begin{bmatrix} 1 & 0 & -1 \\ 2 & 0 & -2 \\ 1 & 0 & -1 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} * \begin{bmatrix} 1 & 0 & -1 \end{bmatrix}$$

Vertical Sobel filter:

$$\begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} * \begin{bmatrix} 1 & 2 & 1 \end{bmatrix}$$

# Computing image gradients

1. Select your favorite derivative filters.

$$\mathbf{S}_x = \begin{bmatrix} 1 & 0 & -1 \\ 2 & 0 & -2 \\ 1 & 0 & -1 \end{bmatrix} \quad \mathbf{S}_y = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$$

2. Convolve with the image to compute derivatives.

$$\frac{\partial \mathbf{f}}{\partial x} = \mathbf{S}_x \otimes \mathbf{f} \quad \frac{\partial \mathbf{f}}{\partial y} = \mathbf{S}_y \otimes \mathbf{f}$$

3. Form the image gradient, and compute its direction and amplitude.

$$\begin{array}{ccc} \nabla \mathbf{f} = \left[ \frac{\partial \mathbf{f}}{\partial x}, \frac{\partial \mathbf{f}}{\partial y} \right] & \theta = \tan^{-1} \left( \frac{\partial \mathbf{f}}{\partial y} / \frac{\partial \mathbf{f}}{\partial x} \right) & \|\nabla \mathbf{f}\| = \sqrt{\left( \frac{\partial \mathbf{f}}{\partial x} \right)^2 + \left( \frac{\partial \mathbf{f}}{\partial y} \right)^2} \\ \text{gradient} & \text{direction} & \text{amplitude} \end{array}$$

# Quiz 3. Image gradients

- Quiz: Image gradients 구현
  - filter2D 를 사용하여 Image gradients 를 구현
    - cv.Sobel 사용 금지
  - 오른쪽 그림과 같이 구현
    - Sobel filter
  - 구현 결과 이미지를 미리 저장된 이미지와 테스트
    - Horizontal derivative, Vertical derivative
    - Amplitude

$$S_x = \begin{bmatrix} 1 & 0 & -1 \\ 2 & 0 & -2 \\ 1 & 0 & -1 \end{bmatrix}$$

$$S_y = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$$

$$\|\nabla f\| = \sqrt{\left(\frac{\partial f}{\partial x}\right)^2 + \left(\frac{\partial f}{\partial y}\right)^2}$$

amplitude

# Image gradients examples

