

LAB: Digital Image Processing

Objective: To apply Spatial processing based filtering operations on digital image for blurring/sharpening using opencv library

Applying the Averaging Filter (Low-Pass Filter):

```
[5]: import cv2

# Load the image (replace with your own image path)
image = cv2.imread('img9.jpg')

# Apply averaging filter with a single kernel size (5, 5)
kernel = (5, 5)
blurred_image = cv2.blur(image, kernel)

# Show the original and blurred images
cv2.imshow('Original Image', image)
cv2.imshow('Blurred Image (5x5)', blurred_image)

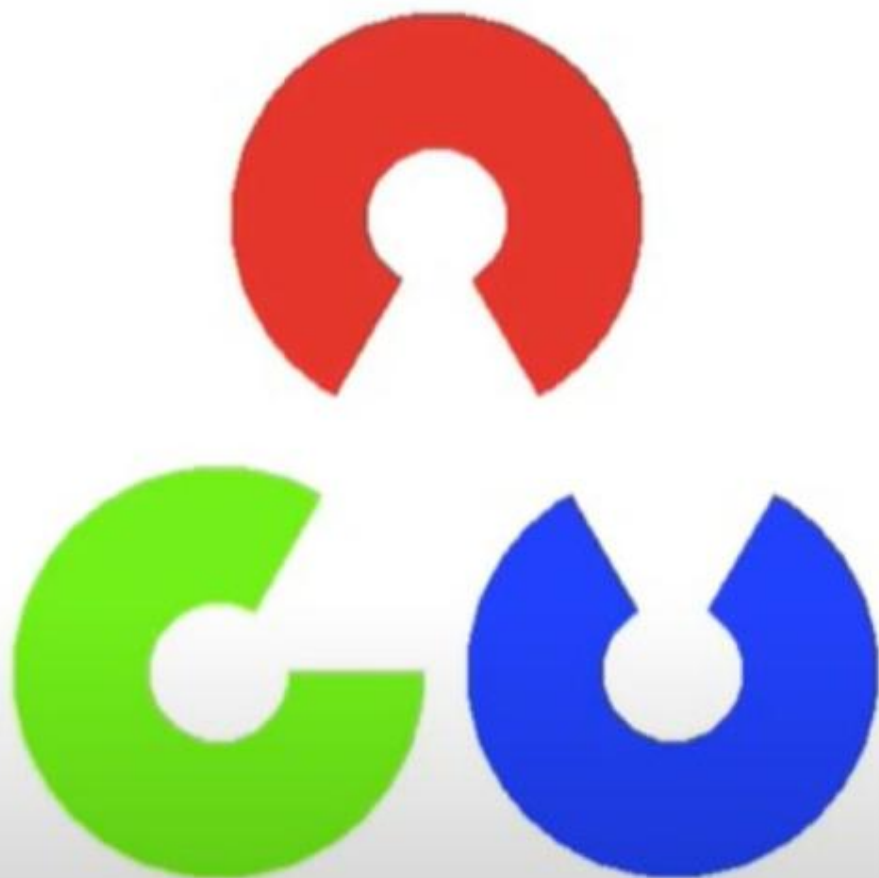
# Wait for a key press and close all windows
cv2.waitKey(0)
cv2.destroyAllWindows()
```

Applying the Averaging Filter (Low-Pass Filter):

- `kernel = (5, 5)`
- `blurred_image = cv2.blur(image, kernel)`
- **kernel = (5, 5):** This defines the size of the kernel (a 5x5 matrix) used for the blurring operation. The kernel size determines how much of the image will be averaged over each pixel. A larger kernel size results in stronger blurring.
- **cv2.blur(image, kernel):** This applies the **averaging filter** (box blur) to the image. It uses the kernel defined earlier to average the pixel values in a **local neighborhood**. The function produces a new image (blurred_image) where high-frequency details are reduced (**the image becomes smoother**).

Original Image

OpenCV



Blurred Image (5x5)

OpenCV



Applying the Gaussian Filter (Low-Pass Filter):

```
[*]: import cv2

# Load the image (replace with your own image path)
image = cv2.imread('img9.jpg')

# Apply Gaussian filter with a single kernel size (5, 5)
kernel = (5, 5)
blurred_image = cv2.GaussianBlur(image, kernel, 0)

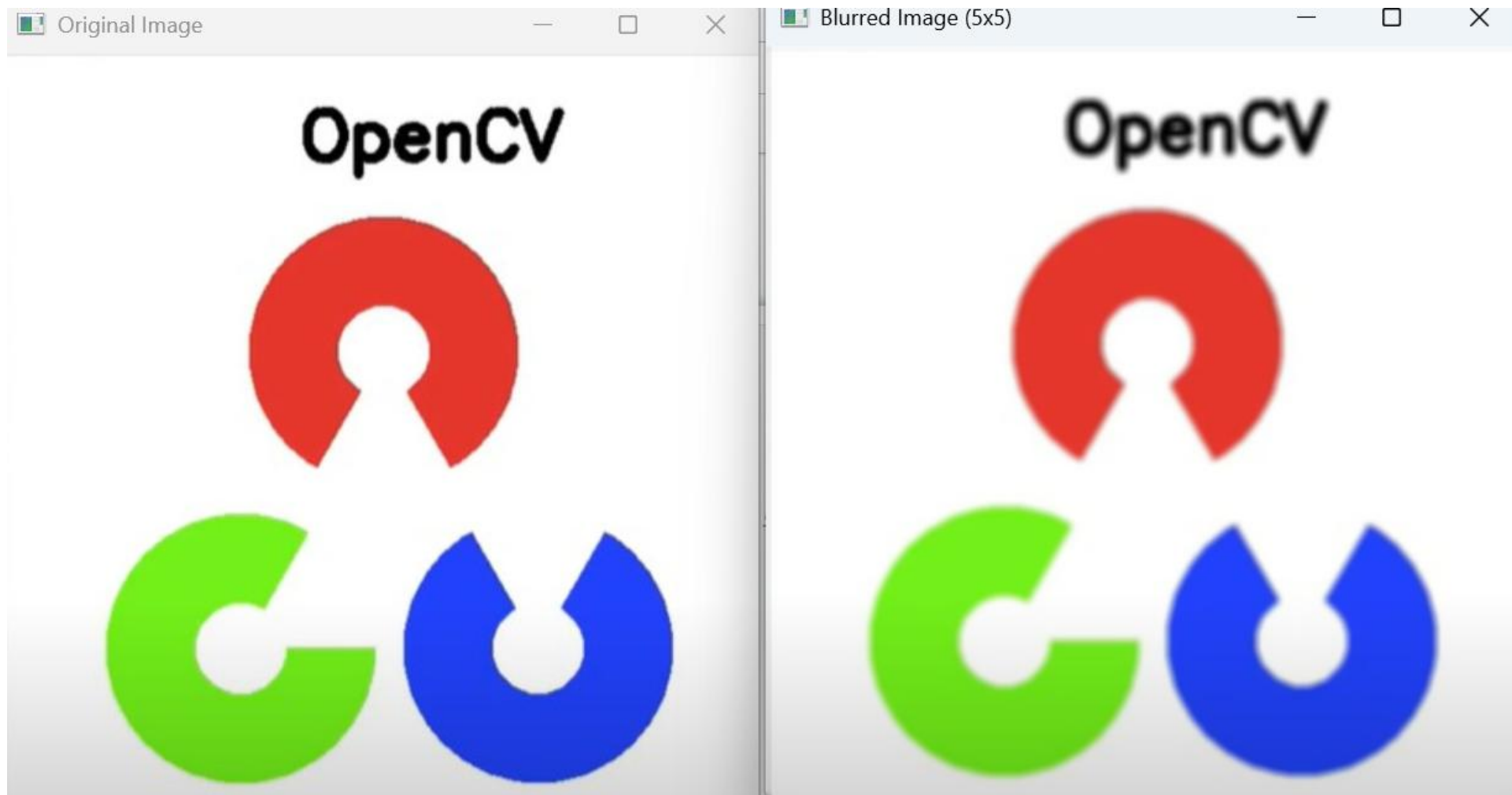
# Show the original and blurred images
cv2.imshow('Original Image', image)
cv2.imshow('Gaussian Blurred Image (5x5)', blurred_image)

# Wait for a key press and close all windows
cv2.waitKey(0)
cv2.destroyAllWindows()
```

Applying the Gaussian Filter (Low-Pass Filter):

- Notes

- `cv2.GaussianBlur(image, kernel, 0)` → 0 is the standard deviation in X direction (auto-calculated if set to 0).
- You can change kernel size (3,3), (7,7), (9,9) to observe different levels of blurring.



Task : Change the Kernel Size

Objective: Modify the kernel size from 5x5 to other values (e.g., 3x3, 7x7, 9x9).

Instructions:

- Adjust the size of the kernel and observe how the blurring effect changes.
- Compare the results with different kernel sizes (e.g., 3x3, 7x7, 9x9).

Task: Experiment with Different Image Types

Objective: Use different images (e.g., grayscale, color, or noisy images).

Instructions:

- Load different types of images, such as grayscale or images with noise, and apply the blurring effect.
- Observe how blurring affects images differently based on the type (grayscale vs color).

Task: Change the Image Format

- Objective: Convert an image from one format (e.g., JPG) into other formats such as PNG, BMP, and TIFF.

Instructions:

- Load an image in JPG format.
- Save and convert the image into PNG, BMP, and TIFF formats.
- Compare the results in terms of file size, compression, and image quality.

Applying the Bilateral Filter:

```
[4]: import cv2

# Load the image
image = cv2.imread('img9.jpg')

# Apply Bilateral Filter (with 9x9 neighborhood, sigma for color and space)
bilateral_filtered_image = cv2.bilateralFilter(image, 12, 125, 125)

# Display the original and bilateral filtered images
cv2.imshow('Original Image', image)
cv2.imshow('Bilateral Filtered Image', bilateral_filtered_image)
cv2.waitKey(0)
cv2.destroyAllWindows()
```

Task : use different neighborhood, sigma color, space value to compare results

Applying the Bilateral Filter:

- `cv2.bilateralFilter()` is an advanced image filtering function that applies the bilateral filter to the image.
- It preserves edges while smoothing the image (unlike Gaussian blur, which also smooths edges). The function takes four parameters.

image: The input image to which the filter will be applied.

12: This is the diameter of the pixel neighborhood that will be considered for filtering. A larger value will result in more smoothing, while a smaller value keeps more details.

125 (sigmaColor): This controls how much the filter considers pixel color differences. Larger values allow for more color variations, which means more smoothing. Smaller values preserve more color details.

125 (sigmaSpace): This controls how much the filter considers the spatial distance between pixels. Larger values allow for more distant pixels to be used in the calculation, resulting in a more significant blur effect.).

Task : change parameters and observe the results

Apply low pass-filters using numpy:

```
[*]: import cv2
import numpy as np

# Load the image
image = cv2.imread('img9.jpg')

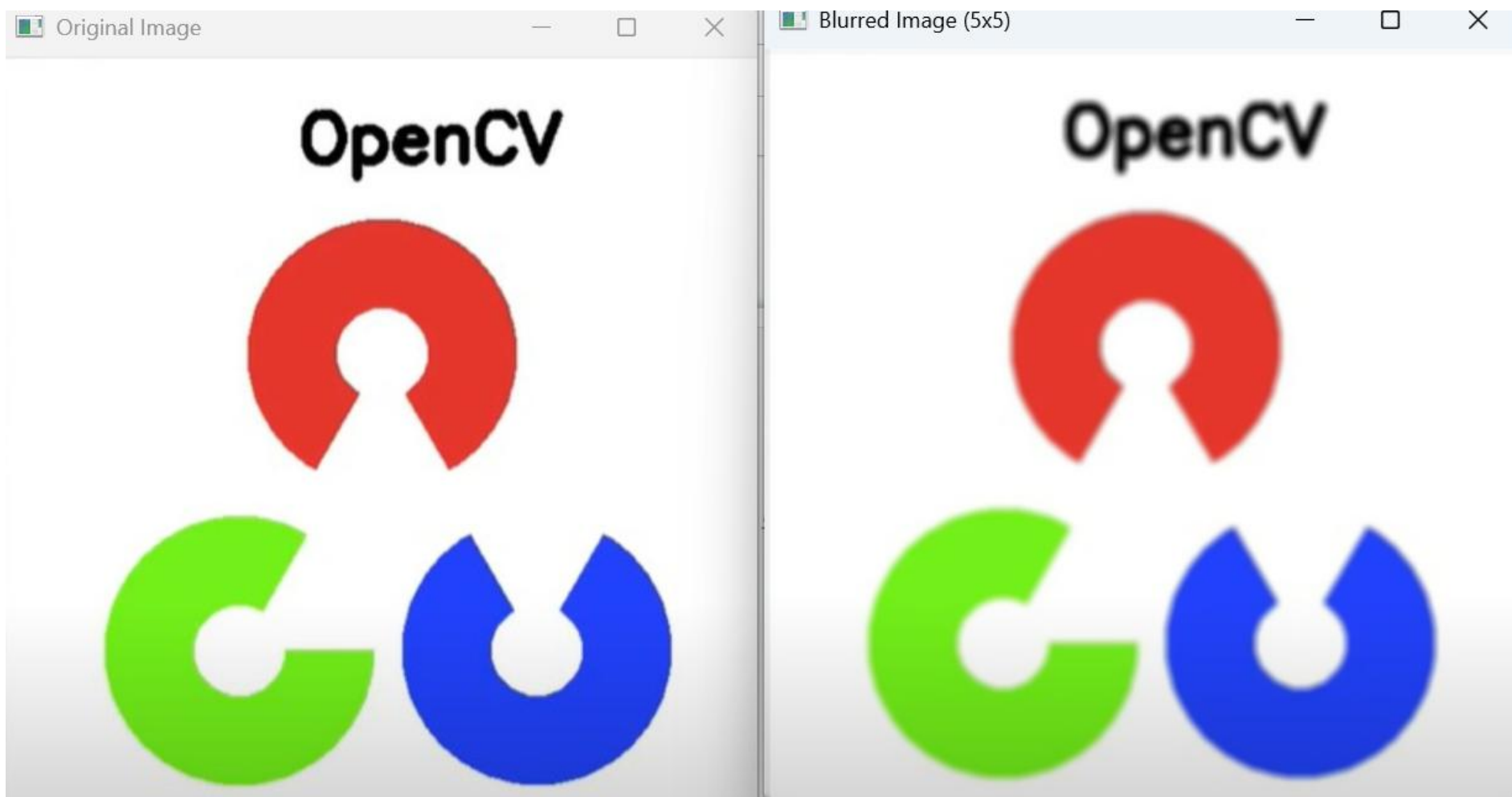
# Define a kernel for blurring (5x5 matrix of ones)
kernel = np.ones((7, 7), np.float32) / 49

# Apply the kernel to the image using filter2D for blurring
blurred = cv2.filter2D(image, -1, kernel)

# Display the original and blurred image
cv2.imshow("Original", image)
cv2.imshow("Blurred", blurred)

# Wait for a key press and close windows
```

- `np.ones((7, 7))` creates a 7x7 matrix where each element is 1.
- This matrix is then divided by 49 (`np.float32 / 49`), which is the total sum of the kernel (7x7 matrix of 1s). This creates a kernel where each element is $1/49$, which averages out pixel values in a 7x7 region, effectively blurring the image.



Apply low pass-filters using matplotlib:

```
[1]: import cv2
import numpy as np
import matplotlib.pyplot as plt

# Load the image
image = cv2.imread('img9.jpg')

# Convert the image from BGR (OpenCV default) to RGB (Matplotlib format)
image_rgb = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)

# Define a kernel for blurring (5x5 matrix of ones)
kernel = np.ones((7, 7), np.float32) / 49

# Apply the kernel to the image using filter2D for blurring
blurred = cv2.filter2D(image_rgb, -1, kernel)
```

```
# Create a figure with two subplots (original and blurred)
fig, axes = plt.subplots(1, 2, figsize=(10, 5))

# Display the original image
axes[0].imshow(image_rgb)
axes[0].set_title("Original")
axes[0].axis("off") # Hide axes

# Display the blurred image
axes[1].imshow(blurred)
axes[1].set_title("Blurred")
axes[1].axis("off") # Hide axes

# Show the plot with both images
plt.show()
```

Task: Adjust the Normalization of the Kernel

Objective: Experiment with different normalization techniques, such as dividing by other values like 10, 100, or even 255.

Instructions:

- Change the kernel normalization value and observe how it affects the blurring strength.
- Try using larger values like 100 or smaller ones like 10.

Applying High-Pass Filter:

```
[*]: import cv2

# Load the image (replace with your own image path)
image = cv2.imread('img9.jpg')

# Define a simple high-pass filter (sharpening kernel)
sharpen_kernel = np.array([[ -1,  -1,  -1],
                           [ -1,   9,  -1],
                           [ -1,  -1,  -1]])

# Apply the high-pass filter (sharpening)
sharpened_image = cv2.filter2D(image, -1, sharpen_kernel)

# Show the original and sharpened images
cv2.imshow('Original Image', image)
cv2.imshow('Sharpened Image (High-pass filter)', sharpened_image)

# Wait for a key press and close all windows
cv2.waitKey(0)
cv2.destroyAllWindows()
```

Applying High-Pass Filter:

```
sharpen_kernel = np.array([[ -1, -1, -1],  
                           [ -1,  9, -1],  
                           [ -1, -1, -1]])
```

`np.array()` creates a NumPy array that defines a sharpening kernel (a 3x3 matrix).

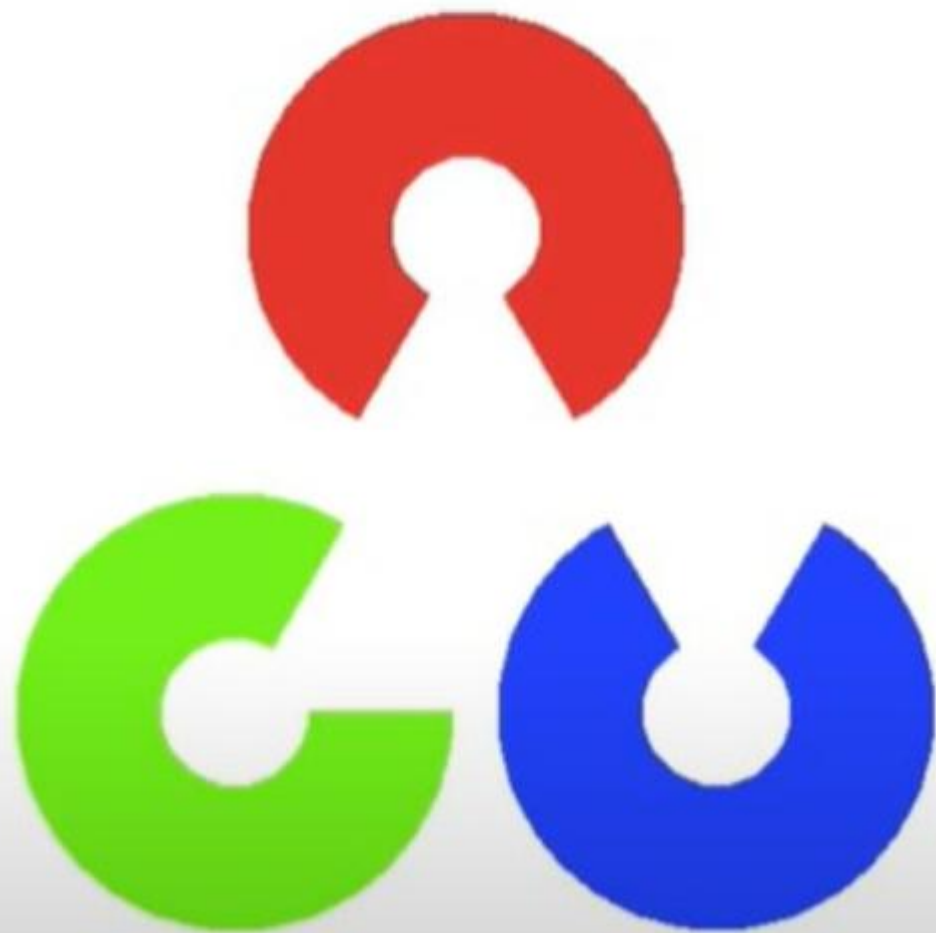
This kernel is a **high-pass filter** that is used to sharpen an image. The center value (9) amplifies the central pixel, and the surrounding values (-1) subtract from the neighboring pixels.

High-pass filters enhance the edges in the image, making them appear more defined by emphasizing sharp transitions

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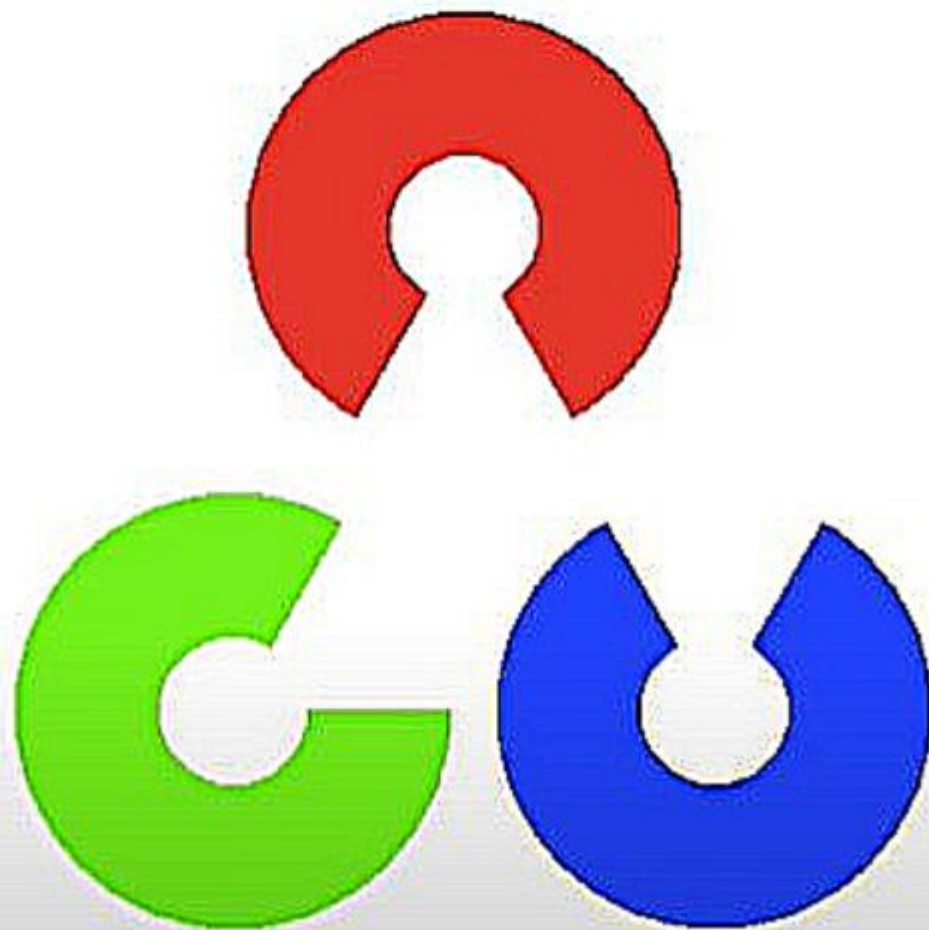
Original Image

OpenCV



Sharpened Image (High-pass filter)

OpenCV



Task: Apply different kernels matrix sizes/ parameters and observe the effects