# **Assignment 2**

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A report on assignment 2 which consists of blurring and detecting edges with various methods of OpenCV and manually.

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# 1 INTRODUCTION

The original image of a bicycle was blurred using Box Filters and Gaussian both with manually written code and using OpenCV. Additionally, Sobel was applied to detect edges of the image with 1st derivative kernels in both the X-Axis and Y-Axis, as well as the XY-Axis by applying Pythagorean's Theorem. Kernels of different diameters were applied to determine how various diameters affected the amount of blurring was applied and how the edge detection was affected.

## 2 PROGRAM

An in depth analysis of each partition in the program.

#### 2.1 PRE-PROCESSING

OpenCV and Numpy are the only libraries that were imported. Some constants were defined for Window Creation and Filter Application below. No explicit global constants were needed, however. The original image was loaded initially as 'image' to be used by all filters. See Figure 2.1.

```
import cv2 as cv
import numpy as np

Hoad Image
image = cv.imread('bicycle.bmp')
```

Figure 2.1. Lines 1-5: Pre-Processing which includes importing libraries and loading the original image

#### 2.2 METHODS

Methods used in the program that were not imported by another library, although inspired by the cited references. Includes work done for manual filter application.

### 2.2.1 create\_kernel()

This method is called whenever a kernel is needed, particularly for Filter Application. kernel\_diameter is inputted such that a diameter of 5 corresponds to a 5x5 kernel filled with 1's. This is used for manual box filter as it uses all 1's in the kernel to find an unweighted average. See Figure 2.2.

Figure 2.2. Lines 7-12: create\_kernel() method which creates a kernel of specified diameter

#### 2.2.2 convolution()

This method is called whenever convolution needs to be applied to an image. The parameters required are the image, and a kernel. This is manual convolution as opposed to OpenCV-convolution. The Kernel radius is derived from the diameter as it represents the number of pixels on all 4 sides that are not able to be included in the convolution and need to be manually set as black pixels. The condition on line 30 is what determines if the current pixel is outside of the border or not. Line 32 creates an array that represents the pixels in the current mask, which are then averaged and stored in the current pixel. See Figure 2.3.

**Figure 2.3.** Lines 14-36: convolution() method which performs convolution on an image with a specified kernel

## 2.2.3 apply\_threshold()

This method applies a given threshold in a range of [0, 1] to a given image. The threshold represents the % distance from the minimum pixel to the maximum pixel's magnitude. If a threshold is given outside the specified range a Value Error is raised. The image is converted from BGR to Grayscale with the following weights on line 54 [.144, .587, .299]. The min and max pixel magnitude first need to be found in lines 56-65. Afterwards, a final iteration occurs where pixels that meet the threshold are black, and pixels that don't are white. See Figure 2.4.

```
### Apply a threshold to an image by outsetting white if it does not meet the threshold,
### Black if it does. Threshold must range from [0-1]
### Company threshold from the distance from in towards max, from 0 to 100% [0, 1]
### Threshold of .5 refers to the midploit between min and max
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```

Figure 2.4. Lines 38-84: apply\_threshold() method which applies a filter to a specified image

#### 2.3 WINDOW CREATION

All windows are created for each blurring and filter technique, both manual and opency, of unspecified kernel diameter. The windows are shifted horizontally and vertically by the width and height of the image, respectively. See Figure 2.5.

#### 2.4 FILTER APPLICATION

All filters and blurring techniques are applied with inputted kernel\_diameter of either 3 or 5. The original image is loaded without changes. Box filter is applied by OpenCV and manually. Guassian is applied by OpenCV. Sobel is applied for X, Y, and XYAxis manually, and XyAxis with OpenCV. For Sobel manuals, the kernels are also inputted manually as copies found from the cited sources instead of computing it on-site. As such, the kernel\_diameter must be either 3 or 5 in order for Sobel manual to function correctly. For Sobel Manual, the image is convolved using convolution(), then thresholded using apply\_threshold(). For XY-Axis, the X axis and Y axis are computed into XY-Axis using (x\*\*2+y\*\*2)\*\*.5. See Figure 2.6.

```
# Create Windows
     # Set dimensions for window separation
     height, width, channel = image.shape # Get height/width quickly for window size and offset
     height=height+30 # Add 28 pixels of height for the window tab size
     # Original
     cv.namedWindow('Original')
     cv.moveWindow('Original', width*2, height*1)
     # Box Filter
     cv.namedWindow('BoxFilter')
     cv.moveWindow('BoxFilter', width*0, height*0)
     # Box Filter Manual
     cv.namedWindow('BoxFilterManual')
     cv.moveWindow('BoxFilterManual', width*0, height*1)
     # Gaussian
     cv.namedWindow('Gaussian')
     cv.moveWindow('Gaussian', width*0, height*2)
     # Sobel
     cv.namedWindow('Sobelmanualxaxis')
     cv.moveWindow('Sobelmanualxaxis', width*1, height*0)
     cv.namedWindow('Sobelmanualyaxis')
     cv.moveWindow('Sobelmanualyaxis', width*1, height*1)
     cv.namedWindow('Sobelmanualxyaxis')
     cv.moveWindow('Sobelmanualxyaxis', width*1, height*2)
     cv.namedWindow('Sobelxyaxis')
116 cv.moveWindow('Sobelxyaxis', width*2, height*2)
```

Figure 2.5. Lines 87-116: Creation of all windows

#### 2.5 SHOW IMAGES

The images for original, boxfilter, gaussian, and sobel for both manual and opency are displayed to the screen at once with specified kernel\_diameter. After any key is pressed, the windows are closed. See Figure 2.7.

```
kernel_diameter = 5
      image = image #no effect
box = cv.boxFilter(image, -1, (kernel_diameter,kernel_diameter))
     boxManual = convolution(image, create_kernel(kernel_diameter))
gauss = cv.GaussianBlur(image,(kernel_diameter,kernel_diameter),0)
     if kernel diameter==3:
       vert_kernel = [[-1, 0, 1], [-2, 0, 2], [-1, 0, 1]] #vertical edge detector (product of deriv and gaussian filters)
        threshold = .2 #halved for xyaxis
      elif kernel_diameter==5:
        vert_kernel = [[1,2,0,-2,-1],
                         [4,8,0,-8,-4],
                         [6,12,0,-12,-6],
                         [4,8,0,-8,-4],
        threshold = .15 #halved for xyaxis
     raise ValueError('Sobel manual kernel_diameter must be 3 or 5')
horiz_kernel = np.transpose(vert_kernel) #horizontal edge detector
      sobelmanual_xaxis = apply_threshold(convolution(image, vert_kernel), threshold)
      sobelmanual_yaxis = apply_threshold(convolution(image, horiz_kernel), threshold)
      sobelmanual_xyaxis = apply_threshold((convolution(image, vert_kernel)**2 + convolution(image, horiz_kernel)**2)**.5, threshold) #sqrt(x^2+y^2)
      sobel_xyaxis = cv.cvtColor(image, cv.COLOR_BGR2GRAY) # convert from bgr to grayscale automatically
sobel_xyaxis = cv.GaussianBlur(sobel_xyaxis,(kernel_diameter,kernel_diameter), sigmaX=0, sigmaY=0)
      sobel_xyaxis = cv.Sobel(src=sobel_xyaxis, ddepth=cv.CV_64F, dx=1, dy=1, ksize=kernel_diameter)
```

Figure 2.6. Lines 120-156: Filter Application

```
# Original
      cv.imshow('Original', image)
      # BoxFilter
      cv.imshow('BoxFilter', box)
      cv.imshow('BoxFilterManual', boxManual)
      # Gaussian
      cv.imshow('Gaussian', gauss)
170
      # Sobel
171
      cv.imshow('Sobelmanualxaxis', sobelmanual_xaxis)
      cv.imshow('Sobelmanualyaxis', sobelmanual_yaxis)
172
173
      cv.imshow('Sobelmanualxyaxis', sobelmanual xyaxis)
174
      cv.imshow('Sobelxyaxis', sobel_xyaxis)
175
176
      # Wait to close
      cv.waitKey(0)
178
      cv.destroyAllWindows()
```

Figure 2.7. Lines 159-178: Show Images

# **3 FUNCTIONALITY**

### 3.1 ORIGINAL

The original image is displayed without any blurring or filters applied. Original 3.1.

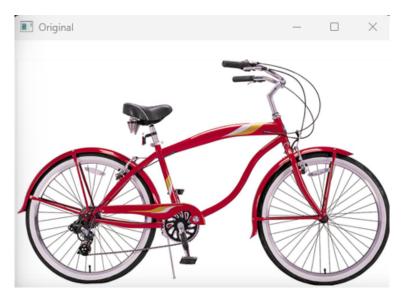


Figure 3.1. Original image without blurring/filtering

### 3.2 BOX FILTERS

The manual box filter is computed by calling convolution() with a 1-filled kernel using create\_kernel(), then displayed to the screen. OpenCV box filter is also created and displayed. Both are displayed for both 3x3 and 5x5 kernels.

# 3.2.1 Manual

3x3 <mark>3.4</mark>.

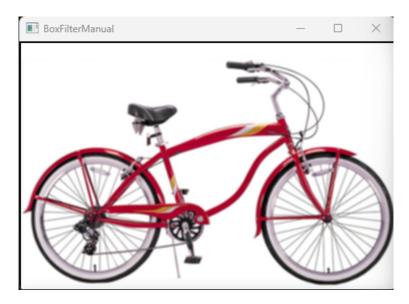


Figure 3.2. Box Filter 3x3 Manually computed

5x5 3.5.

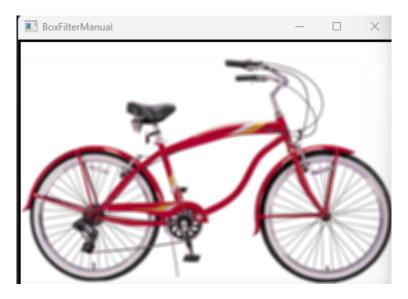


Figure 3.3. Box Filter 5x5 Manually computed

# 3.2.2 OpenCV

3x3 <mark>3.4</mark>.

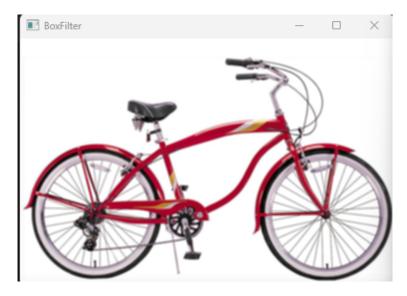


Figure 3.4. Box Filter 3x3 using OpenCV library

5x5 3.5.

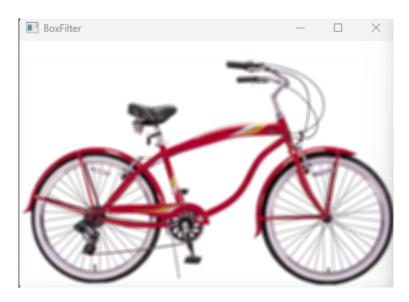


Figure 3.5. Box Filter 5x5 using OpenCV library

### 3.3 GAUSSIAN

Gaussian blurring is applied strictly with OpenCV and not manually.

## 3.3.1 OpenCV

3x3 3.6.

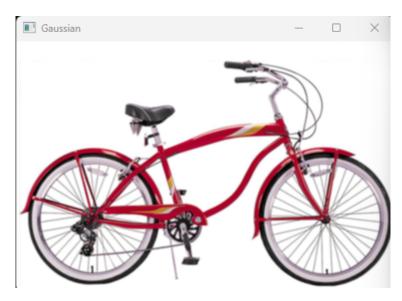


Figure 3.6. Gaussian 3x3 using OpenCV libary

5x5 3.7.

### 3.4 SOBEL

Sobel filters are applied manually and with OpenCV for X-Axis, Y-Axis, and XY-Axis. The XY-Axis referred to simply equates to  $(X^{**2}+Y^{**2})^{**}.5$ , or rather, Pythagorean's Theorem applied to the X-Axis and Y-Axis either manually or with OpenCV.

#### 3.4.1 X-Axis

#### Manual

3x3 3.8.

5x5 3.9.

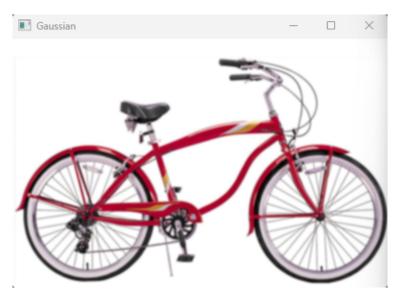


Figure 3.7. Gaussian 5x5 using OpenCV libary



Figure 3.8. Sobel 3x3 X-Axis Manually computed

# 3.4.2 Y-Axis

## Manual

3x3 3.10.

5x5 **3.11**.

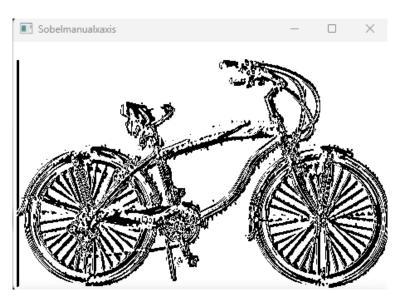


Figure 3.9. Sobel 5x5 X-Axis Manually computed

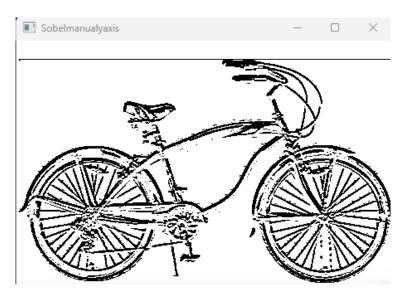


Figure 3.10. Sobel 3x3 Y-Axis Manually computed

## 3.4.3 XY-Axis

### Manual

3x3 3.12.

5x5 **3.13**.

## OpenCV

3x3 <mark>3.14.</mark>

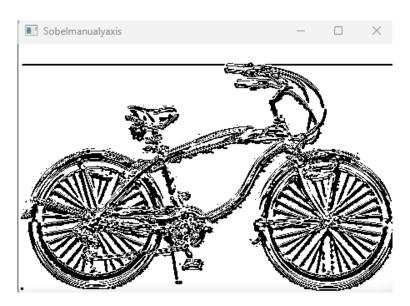


Figure 3.11. Sobel 5x5 Y-Axis Manually computed

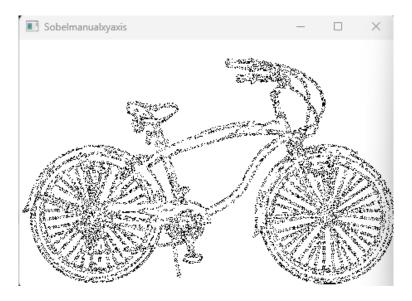


Figure 3.12. Sobel 3x3 XY-Axis Manually computed

5x5 **3.15**.

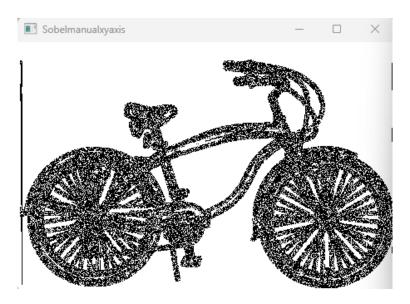


Figure 3.13. Sobel 5x5 XY-Axis Manually computed



Figure 3.14. Sobel 3x3 XY-Axis using OpenCV

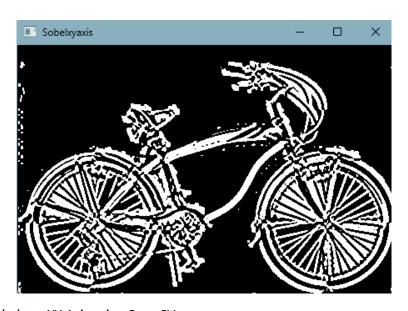


Figure 3.15. Sobel 5x5 XY-Axis using OpenCV

# 4 DISCUSSION

When applying the filter to images in BGR instead of Grayscale, some channels would meet the threshold while others weren't. As a result, one pixel could be outputted as [255, 0, 0]. Even though the pixels are outputted as either 0 or 255, the outputted pixel is blue using these values. All combinations were created, such as [255, 255, 0], [0, 255, 0], etc. The apply\_threshold() method was adjusted first by calculating the average value between the 3 channels. Later, it was learned that gray scale can be computed as just a weighted average of the 3 channels. This weighted average was used in place of the unweighted average, and the BGR image, to represent the image in Grayscale.

There were difficulties making the apply\_threshold work simply. In its current form, the threshold parameter represents the % distance from the min to the max gray-scale value in the image, that way the threshold is based on a normalized image. However, it seems that min and max aren't great values to base it off, as most pixels that aren't the background share very similar grayscale values. If this were to be redone, the threshold would instead be related to the # of z-scores from the mean, with negative values being below the mean, and positive values above the mean. Even this doesn't seem like it would detect edges nearly as well as the cv-built in Sobel method.

Additionally, there were issues getting the cv-Sobel to have the same formatting as all other images. The color of the edges and the background are inverted. An attempt was made to replace 'sobel' with '255 – sobel' to invert the colors, but the resulting image was just a white background with no detected images. The solution was never found, and the cv-Sobel was left as incorrectly inverted colors.

Lastly, the manual xy-axis Sobel may be calculated from the x-axis Sobel and the y-axis Sobel incorrectly. After applying  $(x^{**}2+y^{**}2)^{**}.5$ , the output seems fragmented. Especially for the 5x5 kernel. Different thresholds were attempted to get around the fragmentation, but did not have any significant benefit.

### **REFERENCES**

- [1] Blur Image using cv2.blur(), indianaiproduction, 2024, https://indianaiproduction.com
- [2] Edge Detection using OpenCV, learnopencv, 2024, https://learnopencv.com
- [3] Image Blurring, OpenCV, 2024, docs.opencv.org