## Assignment 2

This assignment is choose one of Canny edge, Watershed algorithm, Hough transform and SLIC super pixel algorithm apply on 4 to 8 pictures. **Canny Edge** will be performed in my assignment.

The techniques will be used on this assignment:

#### 1. Gaussian Filter:

To filter out noise and maintain the weight(original image) of the image, also can improve the Sobel filter performance. Gaussian filter use gaussian distribution to keep and distribute the original pixel information to others and filter noise at the same time.

## 2. Sobel Filter:

To filter out the edge of image by 2 direction Sobel filter. Gx = [[-1, 0, 1], [-2, 0, 2], [-1, 0, 1]] as horizontal filter to detect the horizontal pixels is the same or not, if the horizontal pixels is the same, the sum of detected area will be 0. Otherwise, it will detected how different of those pixel, that is, to present the edge of image by intensity and also the gradient direction. For vertical pixels, Gy = [[-1, -2, -1], [0, 0, 0], [1, 2, 1]] will be used.

## 3. Canny detector

## 1. Non-maximum suppression

According to the gradient direction/angle calculated by Sobel Filter, a direction of intensity will be known, therefore, the width of edge will be selected to compare and keep the maximum one by compare that gradient direction intensity one by one.

## 2. Connect Weak Edge by select correct threshold

For the real case, we cannot easily distinguish noise by one threshold since there must be some noises or fake edges in the real image after filter, a low and high threshold were choose to reduce most of noises and fake edges, also keep the real significant edge. However, the pixels between high and low threshold will be kept if there are two higher threshold pixels aside the pixel need to be determined, Or it will be count as noises or fake edges.

Try to improve canny edge method:

## 1. Histogram Equalization:

To count the number of pixel in the image and accumulate it before multiplication of 255 to let image distribute much more better.

- 2. Add 2 more direction of Sobel filter.
- 3. Use larger gaussian filter.
- 4. Using Sharpen filter instead of Sobel filter.
- 5. Normalization after Sobel Calculation

## Part 1 Gaussian Filter:

To keep the weight of original pixel and reduce the noise, also color image is not good for edge detection, not easy to read, may confuse by observation and may generate some noises, that is, I use gray image as original image.



Figure 1-1 original image

Figure 1-2 image after gaussian filter

In here, I selected 3x3 gaussian filter by below equation where  $\sigma$  = square root of 0.5 for easy implement.

$$G(x,y) = \frac{1}{2\pi\sigma^2} e^{-\frac{x^2+y^2}{2\sigma^2}}$$

We can obtain the picture become blur and the pixel value will much more average as original one.

#### Part 2 Sobel Filter:

To filter out the edge and present by horizontal and vertical edge, then combine it as image edge detection.



Figure 2-1 Image after Sobel Filter in Horizontal Direction

Figure 2-2 Image after Sobel Filter in vertical Direction

As we can see, Sobel filter Gx is actually a vertical edge detector (Figure2-1) and Gy is horizontal edge detector (Figure2-2). After careful consideration of Sobel filter Gx and Gy from the beginning introduction, you will understand it. For more explanation, horizontal filter will compare left and right pixel, that is, it will produce a result of vertical edge if left and right pixels are not the same. And finally, we will have the image edge by combined both direction edge together as Figure 2-4.



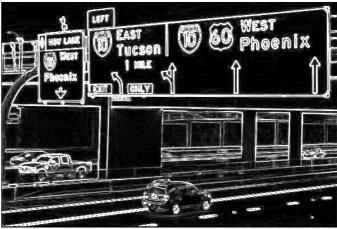


Figure 2-3 Original gray image

Figure 2-4 Combination of Sobel filter in both horizontal and vertical direction

Actually, I believe this edge detection is good enough for drawing the edge of input image, but there may have application need to have more precise edge such as medical application - tumor detection. As a result we need to reduce the edge to have more accurate edge.

## Part 3 Canny detection:

#### Part 3-1:

Non-maximum suppression:

To compare with adjacent pixels and select the maximum intensity by the intensity (G) and direction( $\theta$ ) calculated by Sobel filter as shown as below. We can obviously obtain the different between Sobel edge detector and Figure 3-1, it is significantly thinner.

$$G = \sqrt{G_x^2 + G_y^2}$$
  $\theta = arc \tan(G_y / G_x)$ 

## Part 3-2:

Connect weak edge:

Set a high and low thresholds to maintain the real edge and drop the noise. For the real edge, set it to 255. For the noise, set it to be 0. In the meantime, check pixels between high and low threshold are around the real edge pixel, if so, connect it with real edge pixel, otherwise, drop it.

From the observation, we can found that the edge of image is much more easier to view, especially for the vehicle in the middle bottom. The original curve of vehicle is almost disappear, but it reappear again after connect weak edge. Somehow, the weak edge connectivity is not good enough.

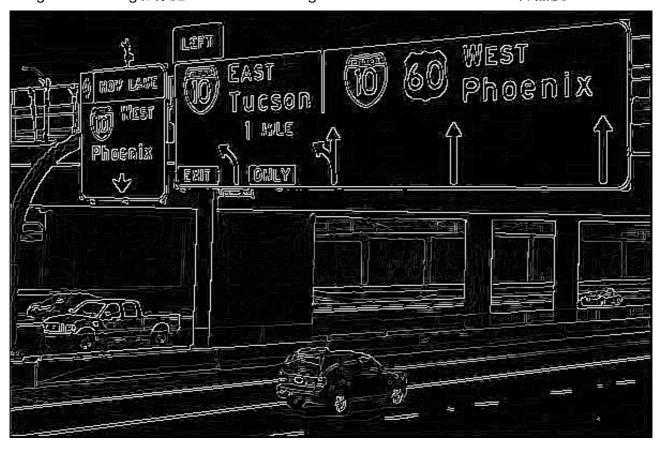


Figure 3-1 After Non-max suppression to minimize the edge



Figure 3-2 After weak edge detection to maximize the edge and drop the noise

## Part 4 Comparison with canny library of internet:





Figure 4-1 Canny Detection from Internet (High threshold = 130, Low threshold = 60)

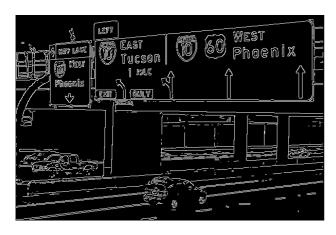
Figure 4-2 Canny Detection of my coding (High threshold = 130, Low threshold = 60)

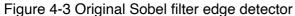
From the figure 4-1, I found that for the same threshold, my image generate more noise and edges detect also not shown completely after process. May be it is because of image pre-process and weak detection.

## **Improvement Trials**

## Trial 1: (Not work!)

Pre-process image before Sobel edge detection by Histogram Equalization(HE) and compare with original one at Canny Edge detection stage. The detail of HE design can review to my HW1.





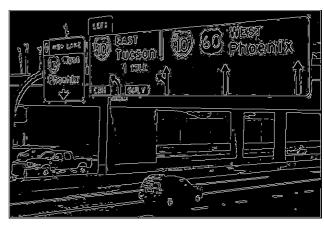


Figure 4-4 Pre-process original by Histogram Equalization

Obviously, Histogram Equalization is not benefit to edge detection, the result even become worst than previous one since some of edge missing and generate more noise.

# Trial 2: (Not work!)

Add detected direction of Sobel filter see if we can improve the Sobel edge detection.

As I added 45 degree and 135 degree Sobel filter, the result seems not improvement. Some of edges even become lighter.



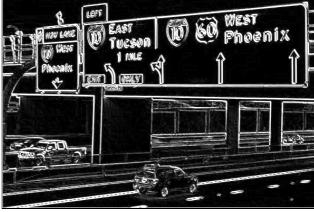


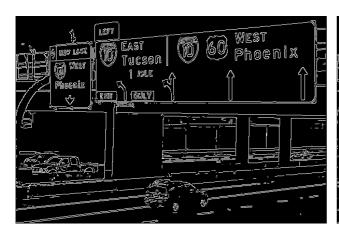
Figure 4-5 Original Sobel filter edge detector

Figure 4-6 Using another 2 angle Sobel edge detection

# Trial 3: (Not work!)

Use larger size of Gaussian filter.

From the observation, it is difficult to distinguish different between two images.





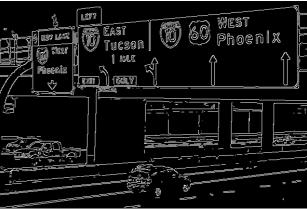


Figure 4-8 Sobel filter edge detector with 7x7
Gaussian filter

Trial 4: (Not work!)

Use sharpen filter to detect more edges before doing canny detection.

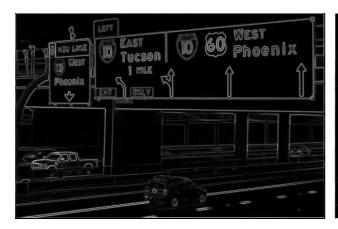




Figure 4-9 Original Sobel image

Figure 4-10 Sharpen filter edge detection

As a result, Sharpen filter also cannot detect more edge from image and seems worst than Sobel.

## Trial 5: (seems work?)

Normalization after Sobel filter calculate magnitude. As the magnitude of Sobel calculation is square root of (Gx^2 + Gy^2), it may exceed possible value of image(255), so I normalized it as divided all pixels by maximum pixel value and multiple 255. Thus, a result shown as below.





Figure 4-11 Comparison between Internet Canny Edge library with threshold 130 and 60 vs my normalized calculation after Sobel with threshold 20 and 5

I am not sure this normalization improve the edge detection of my original although it produces less noise than previous one, but more edge on the image. However, it change the among of threshold values for canny detection, maybe it is because of threshold selection or it reduces noise.

# Part 5 Application of different image:

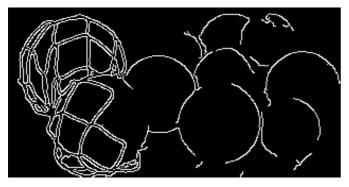


Figure 5-1 Reference Canny detection using High threshold 230 and Low threshold 100

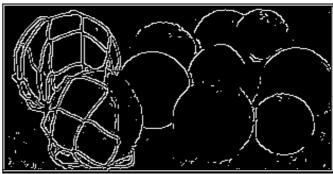


Figure 5-2 My Canny detection using High threshold 120 and Low threshold 45.



Figure 5-3 Reference Canny detection using High threshold 130 and Low threshold 80

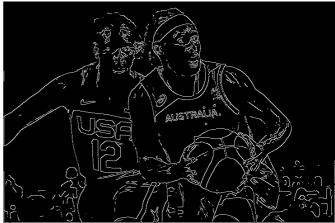


Figure 5-4 My Canny detection using High threshold 130 and Low threshold 80.

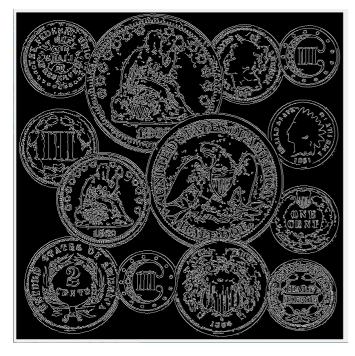


Figure 5-5 Reference Canny detection using High threshold 160 and Low threshold 70



Figure 5-6 My Canny detection using High threshold 160 and Low threshold 70.

2nd testing picture(balls) was shown as below Figure 5-1 and 5-2 by using reference canny detection compare with my canny detection.

3rd testing picture(basketball) was shown as below Figure 5-3 and 5-4 by using reference canny detection compare with my canny detection.

4th testing picture(coins) was shown as below Figure 5-5 and 5-6 by using reference canny detection compare with my canny detection.



Figure 5-7 Reference Canny detection using High threshold 160 and Low threshold 60

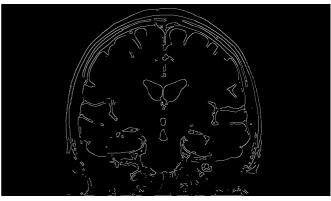


Figure 5-8 My Canny detection using High threshold 160 and Low threshold 60.

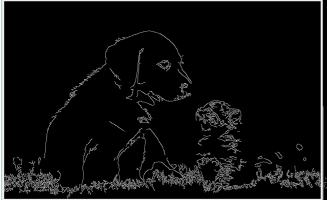


Figure 5-9 Reference Canny detection using High threshold 180 and Low threshold 70

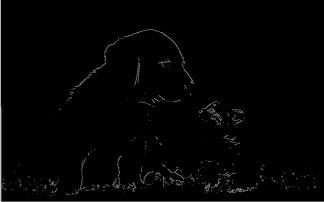


Figure 5-10 My Canny detection using High threshold 180 and Low threshold 70.

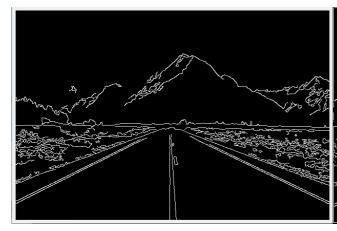


Figure 5-11 Reference Canny detection using High threshold 210 and Low threshold 70



Figure 5-12 My Canny detection using High threshold 130 and Low threshold 50.

5th testing picture(MRI) was shown as below Figure 5-7 and 5-8 by using reference canny detection compare with my canny detection.

6th testing picture(pets) was shown as below Figure 5-9 and 5-10 by using reference canny detection compare with my canny detection.

7th testing picture(scene) was shown as below Figure 5-11 and 5-12 by using reference canny detection compare with my canny detection.

8th testing picture(CT) was shown as below Figure 5-13 and 5-14 by using reference canny detection compare with my canny detection.

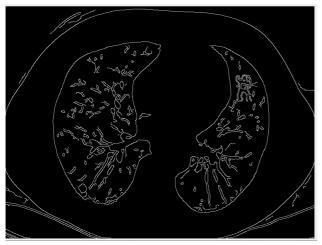


Figure 5-13 Reference Canny detection using High threshold 210 and Low threshold 70

Figure 5-14 My Canny detection using High threshold 130 and Low threshold 40.

## **Part 6 Conclusion:**

This assignment let me more understand canny edge detection operation, it can be adaptable to various environment. However, my canny edge detector is not as good as website library reference. By the method and observation, I believe it is because of pre-processing related problems - noise reduction. In advance, my weak edge detection seems not good enough.

As the reference canny detection can have much more edge. I believe it is related to weak edge detection method, I think my weak edge detection is not the best and remain many improvement. For example, if there are several weak edge pixels connected together and one of them connected to real edge, my weak edge detection cannot consider it as real edge, but actually, it can be a real edge.

In my trials, I failed to improve my image processing program, but actually, I found that normalization during Sobel calculation may improve edge detection, it may be related to threshold selection, but also may related to noise reduction after my consideration.

As a result, if I would like to have better result, maybe I should try below direction:

- 1. To enlarge the kernel of weak edge detection and improve the detection method.
- 2. To find a better way to determine a better threshold.
- 3. Image pre-processing as noise reduction or better edge detection rather than Sobel.

At last, I will briefly explain my coding. First, I create a GUI for easy implementation, and I use Library code for comparison, also I had included my trial code in below like Histogram Equalization, Sharpen filter, etc.

I try to name the function more easier to read and add some explanation, please let me know if you curious or you have any questions about this assignment.import itertools

```
import math
import tkinter as tk
from PIL import Image, ImageTk
import math as m
import numpy as np
import copy
import cv2
from tkinter import filedialog
```

```
# create a layout
def define_layout(obj, cols=1, rows=1):
    def method(trg, col, row):

        for c in range(cols):
            trg.columnconfigure(c, weight=1)
        for r in range(rows):
            trg.rowconfigure(r, weight=1)

    if type(obj) == list:
        [method(trg, cols, rows) for trg in obj]
    else:
        trg = obj
        method(trg, cols, rows)
```

```
# open file

def openfn():
    filename = filedialog.askopenfilename(title='open')
    return filename
```

```
# open image
def open_img_left():
```

```
x = openfn()
   img = Image.open(x)
   img = img.resize((img_before_size[1], img_before_size[0]), Image.ANTIALIAS)
   img = ImageTk.PhotoImage(img)
   panel = tk.Label(div2, image=img)
   panel.image = img
   panel.grid(column=1, row=1, sticky=align mode)
lef open_img_right():
   x = openfn()
   img = Image.open(x)
   img = img.resize((img_after_size[1], img_after_size[0]), Image.ANTIALIAS)
   img = ImageTk.PhotoImage(img)
   panel = tk.Label(div3, image=img)
   panel.image = img
   panel.grid(column=1, row=1, sticky=align mode)
def s_print(text): # print scale variable
   print('Gaussian Filter:', value1.get(), 'High threshold:', value3.get(), 'Low thresh
ld:', value4.get())
 create a window
window = tk.Tk()
window.title('Canny Edge')
align mode = 'nswe
pad = 5
value1 = tk.IntVar()
value2 = tk.IntVar()
value3 = tk.IntVar()
value4 = tk.IntVar()
```

```
read image before = './pic4PR2/CT.jpg' # read image you would like to process, can be
changed
write image before = "CT gray.png" # write image you would like to process, can
read_image_after = './CT_gray.png' # preview gray image you would like to process,
image before = cv2.imread(read image before)
img_before_gray = cv2.cvtColor(image_before, cv2.COLOR_RGB2GRAY)
arr = np.array(img_before_gray)
img before size = arr.shape
cv2.imwrite(write image before, img before gray
image after = copy.deepcopy(image before)
img_after_temp = copy.deepcopy(image_before)
img after size = img before size
div size = 200
div1 = tk.Frame(window, width=img before size[1], height=div size, bg='blue')
bg='blue'
div4 = tk.Frame(window, width=img before size[1], height=div size) # bg='yellow'
div2 = tk.Frame(window, width=img_before_size[1], height=img_before_size[0], bg='orange')
div3 = tk.Frame(window, width=img after size[1], height=img after size[0], bg='gre
window.update()
win size = min(window.winfo width(), window.winfo height())
div1.grid(column=0, row=0, padx=pad, pady=pad, sticky=align mode)
div4.grid(column=1, row=0, padx=pad, pady=pad, sticky=align mode)
div2.grid(column=0, row=1, padx=pad, pady=pad, rowspan=2, sticky=align_mode)
div3.grid(column=1, row=1, padx=pad, pady=pad, rowspan=2, sticky=align mode)
define layout(window, cols=2, rows=2)
define layout([div1, div2, div3, div4]
im = Image.open(write image before) # read original image
imTK L = ImageTk.PhotoImage(im.resize((img before size[1], img before size[0])
```

filter size = []

return fil

for i in range(fNum):

fil = np.array(filter\_size)

filter\_size.append([0.0] \* fNum)

```
image_main = tk.Label(div2, image=imTK_L)
image main['height'] = img before size[0]
image main['width'] = img before size[1]
image main.grid(column=0, row=1, sticky=align mode)
image main.grid(column=1, row=1, sticky=align mode)
define layout(window, cols=3, rows=3)
define_layout(div1)
define layout(div4)
define layout(div2, rows=2)
define_layout(div3, rows=2)
 read image in right hand side
im = Image.open(read image after)
imTK R = ImageTk.PhotoImage(im.resize((img after size[1], img after size[0]
image main = tk.Label(div3, image=imTK R)
image main['height'] = img after size[0]
image_main['width'] = img_after_size[1]
image_main.grid(column=0, row=1, sticky=align_mode)
image main.grid(column=1, row=1, sticky=align mode)
 filter design
def initial filter(fNum):
```

```
# create a gaussian filter

def gaussian_filter_creator():
    sum_of_filter = 0

    gaussian_size = int(value1.get()) # gaussain filter size(3) can be changed to val-
```

```
def conv_full image(img, fil):
    temp = []
    img_temp = copy.deepcopy(img)
    img_new_c = np.zeros(img.shape[0] * img.shape[1])
    img_new_c = img_new_c.reshape(img.shape[0], img.shape[1])
    img_size = img.shape
    fil_size = fil.shape
    for i in range(img_size[0]):
        for j in range(img_size[1]):
        img_new_c = conv(img_new_c, img_temp, img_size, fil, fil_size, temp, i, j)
```

return img new c

temp sum = np.sum(temp)

```
cen_pointx = int(fil_size[0] / 2)
cen_pointy = int(fil_size[1] / 2)
num = int(temp_sum)

if (i + fil_size[0] < img_size_c[0]) and (j + fil_size[1] < img_size_c[1]):
    img_new_c[i + cen_pointx][j + cen_pointy] = num
    temp.clear()
return img_new_c</pre>
```

```
def sobel_filter(phase):
    if phase > 112.5: # 135 degree
        sobel_fil = [[2, 1, 0], [1, 0, -1], [0, -2, -1]]
    elif phase > 67.5: # 90 degree
        sobel_fil = [[1, 2, 1], [0, 0, 0], [-1, -2, -1]]
    elif phase > 22.5: # 45 degree
        sobel_fil = [[0, 1, 2], [-1, 0, 1], [-2, -1, 0]]
    else: # 0 degree
        sobel_fil = [[-1, 0, 1], [-2, 0, 2], [-1, 0, 1]]
```

```
def sobel_calculation(image_x, image_y):
    image_combine = np.zeros(image_x.shape[0] * image_x.shape[1])
    image_combine = image_combine.reshape(image_x.shape[0], image_x.shape[1])
    theta = np.zeros(image_x.shape[0] * image_x.shape[1])
    theta = theta.reshape(image_x.shape[0], image_x.shape[1])
    for i, j in itertools.product(range(image_x.shape[0]), range(image_x.shape[1])):
        theta[i][j] = math.atan2(image_y[i][j], image_x[i][j]) * 180 / math.pi
        theta[i][j] = theta[i][j] * 180
        if theta[i][j] > 112.5:
            theta[i][j] > 67.5:
            theta[i][j] = 90
        elif theta[i][j] > 22.5:
            theta[i][j] = 45
        else:
```

<mark>return</mark> image after sharper

```
theta[i][j] = 0
       image_combine[i][j] = int(math.sqrt(image_x[i][j] ** 2 + image_y[i][j] ** 2))
   # for i, j in itertools.product(range(image x.shape[0]), range(image x.shape[1])):
normalize
   print('np.max(image):', np.max(image_combine))
   print('done sobel calculation')
   return image combine, theta
lef smooth filter creator():
   average_of_filter = value1.get() * value1.get()
   smooth size = int(value1.get()) # gaussain filter size(3) can be changed to
ralue1.get()
   matrix var = int(smooth size / 2)
   initial smooth filter = initial filter(smooth size)
   final_smooth_filter = initial_filter(smooth_size)
   for i, j in itertools.product(range(smooth size), range(smooth size)):
       final_smooth_filter[i][j] = (initial_smooth_filter[i][j] + 1) / average_of_filter
   print('done Smooth')
   return final smooth filter
lef sharpen filter(source image):
   smooth filter = smooth filter creator()
   smooth filter size = smooth filter.shape
   print('smooth_filter:', smooth_filter_size, 'sum:', np.sum(smooth_filter))
   img_after_smooth = conv_full_image(source_image, smooth_filter)
   image after sharpen = np.zeros(source image.shape[0] * source image.shape[1])
   image after sharpen = image after sharpen.reshape(source image.shape[0], source im
age.shape[1])
   for i in range(source image.shape[0]):
       for j in range(source image.shape[1]):
            image_after_sharpen[i][j] = int(source_image[i][j]) - int(img_after_smooth[:
(ii)
            if image after sharpen[i][j] < 0: # confirm the value will not less than</pre>
zero
                image_after_sharpen[i][j] =
```

```
lef canny edge detector(image source, canny theta):
   canny image = copy.deepcopy(image source)
   for i, j in itertools.product(range(canny theta.shape[0]);
range(canny_theta.shape[1])):
        if (i + 1) < canny_image.shape[0] and (j + 1) < canny_image.shape[1]:</pre>
           if canny theta[i][j] > 112.5: # 135 degree, compare two dimension value with
core one
                if image source[i][j] < image source[i - 1][j - 1] or '</pre>
                         image_source[i][j] < image_source[i + 1][j +</pre>
                    canny_image[i][j] = 0
            elif canny theta[i][j] > 67.5: # 90 degree
                if image_source[i][j] < image_source[i - 1][j] or \</pre>
                         image source[i][j] < image source[i + 1][j</pre>
                    canny image[i][j] = 0
            elif canny_theta[i][j] > 22.5: # 45 degree
                if image_source[i][j] < image_source[i + 1][j - 1] or '</pre>
                         image_source[i][j] < image_source[i - 1][j +</pre>
                    canny image[i][j] =
                if image_source[i][j] < image_source[i - 1][j - 1] or '</pre>
                         image source[i][j] < image source[i + 1][j</pre>
                    canny image[i][j] = (
   return canny image
ef canny edge threshold(source image, high threshold, low threshold, filter in):
   canny_image = np.zeros(source_image.shape[0] * source_image.shape[1])
   canny_image = canny_image.reshape(source_image.shape[0], source_image.shape[1])
   print('high threshold:', high threshold, 'low threshold', low threshold)
   count = 0
   for i, j in itertools.product(range(source image.shape[0])
range(source_image.shape[1])):
        if source image[i][j] >= high threshold:
            canny image[i][j] = 255
        elif source image[i][j] <= low threshold:</pre>
```

```
# Histogram Equalization

def CountPixel(original_image, ArrayG, imageH, imageW):
    for i in range(imageH - 1): # use i as index of image to determine which pixels need to be filtered

        for j in range(imageW - 1): # use j as index of image to determine which pixels need to be filtered

        ArrayG[original_image[i][j]] += 1 # count the corresponding pixel number in the image from 0 to 255

    return ArrayG
```

```
def Accumulate(Array):
    AccumArray = [0] * 256
    j = 0
    for i in range(256):
        AccumArray[i] = Array[i] + j # accumulate the appear pixel
        j = AccumArray[i]
    return AccumArray
```

```
def ImageHeq(original_image, ACPAG, imageH, imageW):
    img_tmp = copy.deepcopy(original_image)
    for i in range(imageH):
```

```
for j in range(imageW):
           img tmp[i][j] = ACPAG[img tmp[i][j]] * 255 # do the histogram equalizat
   return img tmg
ef Histogram Equalization(source image):
   image_temp = copy.deepcopy(source_image)
   IGray = [0] * 256 # initial the RGB value for counting the corresponding number
Lmage
   AcPAG = [0] * 256
   CountPixel(image temp, IGray, source image.shape[0], source image.shape[1])
   EveryPAG = np.array(IGray) / (source image.shape[0] * source image.shape[1])
   AcPAG = Accumulate(EveryPAG) # accumulate the number of appear pixel for each chan
   image temp = ImageHeq(image temp, AcPAG, source image.shape[0],
                          source image.shape[1]) # do the histogram equalizati
   return image temp
 ef main():
   gaussian_filter = gaussian_filter_creator()
   gaussian filter size = gaussian filter.shape
   print('gaussian filter size:', gaussian filter size, 'sum', np.sum(gaussian filter)
   img_after_temp = conv_full_image(img_before_gray, gaussian_filter)
   cv2.imwrite("img after gaussian.png", img after temp)
   img_after_sobel_x = conv_full_image(img_after_temp, sobel_filter(0))
   img_after_sobel_y = conv_full_image(img_after_temp, sobel_filter(90))
img_after_sobel_combine_x_y, img_after_sobel_combine_theta =
sobel_calculation(img_after_sobel_x, img_after_sobel_y)
```

```
cv2.imwrite("img_after_sobel_x.png", img_after_sobel_x)
   cv2.imwrite("img_after_sobel_y.png", img_after_sobel_y)
   cv2.imwrite("img after sobel combine x y.png", img after sobel combine x y.png",
   img_after_sharpen = sharpen_filter(img_before_gray) # method 2 to get edg
   cv2.imwrite("img after sharpen.png", img after sharpen)
sobel_calculation(img_after_sobel_yx, img_after_sobel_yx)
   # cv2.imwrite("img after sobel xy.png", img after sobel xy)
p.shape[1])
mg after sobel combine final.reshape(img after temp.shape[0] * img after temp.shape[1])
   # img_after_sobel_combine_final = sobel calculation(img after sobel combine vx
img_after_sobel_combine_x_y)
   img_after_canny = canny_edge_detector(img_after_sobel_combine_x y, img_after_sobel
combine_theta)
   cv2.imwrite("img after canny.png", img after canny)
   img after canny threshold = canny edge threshold(img after canny, value3.get(), val
ue4.get(), gaussian filter)
   cv2.imwrite("img after canny threshold.png", img after canny threshold
   print('done, please open image
def library canny():
   image source = cv2.imread(read image before)
   gray = cv2.cvtColor(image source, cv2.COLOR BGR2GRAY)
   blurred = cv2.GaussianBlur(gray, (value1.get(), value1.get()), 0)
   img_after_library_canny = cv2.Canny(blurred, value3.get(), value4.get())
   cv2.imwrite("img after library canny.png", img after library canny)
```

print('done library canny')

```
scale bar function
lbl title1 = tk.Scale(div1, label='Gaussian Filter', from =3, to=9, orient=tk.HORIZONTAL,
                     resolution=1, variable=value1, command=s print)
lbl title3 = tk.Scale(div1, label='High threshold', from =0, to=255, orient=tk.HORIZON-
TAL,
                     resolution=5, variable=value3, command=s print)
lbl title4 = tk.Scale(div1, label='Low threshold', from =0, to=250, orient=tk.HORIZONTAL
                  resolution=5, variable=value4, command=s print)
lbl title1.grid(column=0, row=1, sticky=align mode)
lbl title3.grid(column=0, row=3, sticky=align mode)
lbl_title4.grid(column=0, row=4, sticky=align_mode)
btn1 = tk.Button(div4, text='Run', command=main).grid(column=0, row=1, sticky=align mode)
btn2 = tk.Button(div4, text='Run library_canny', command=library_canny).grid(column=0,
row=2, sticky=align_mode)
btn3 = tk.Button(div4, text='open before image', command=open_img left).grid(column=0,
row=3, sticky=align_mode)
btn4 = tk.Button(div4, text='open after image', command=open img_right).grid(column=0)
row=4, sticky=align mode)
rint('initial'
vindow.mainloop()
```