

Programming Assignment #2

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Introduction / Objectives:

The purpose of this program is to implement Canny Edge Detection to find the particular edge points in an image. One can observe the difference of results by applying different settings and experiences (filter, threshold... etc).

Method:

Noise reduction: Gaussian filter, Bilateral filter

Canny edge detection

Adaptation Threshold of Canny Edge Detection

Review of the methods I have implemented:

I. Noise reduction

1. Gaussian filter: A Gaussian has the advantage of being smooth in both the spatial and frequency domains.

2. Bilateral filter: A non-linear, edge-preserving, and noise-reducing smoothing filter for images, based on Gaussian distribution.

II. Sobel filter: A filter for calculating image gradients in horizontal and vertical direction to find the details of edge.

III. Non maximum suppression: Keep only the “strongest” edge points along the gradient direction.

IV. Double thresholding: Retain all the “strong” edge points as well as the “weak” edge points that are connected to the “strong” edge points.

3. Adaptive Threshold of Canny Edge Detection:

- Calculate the gradient and the max value(maxv) of a gray scale image
- Set a histogram of gradient image ranged in [0, maxv] and calculate the histogram.
- Set the ratio of non-edge pixels to the entire image pixels as Percent Of Pixels Not Edges.
- $\text{Total threshold} = \text{img.height} * \text{img.width} * \text{PercentOfPixelsNotEdges}$.
- Traverse the histogram, the number of pixels corresponding to each gradient value, and save the sum in a variable x.
- If x is greater than the value of total, exit the loop of hist traversal.
- Calculate lower and upper thresholds ($\text{High} = 2.5 * \text{Low}$).
- Double Thresholding

Picture 1:

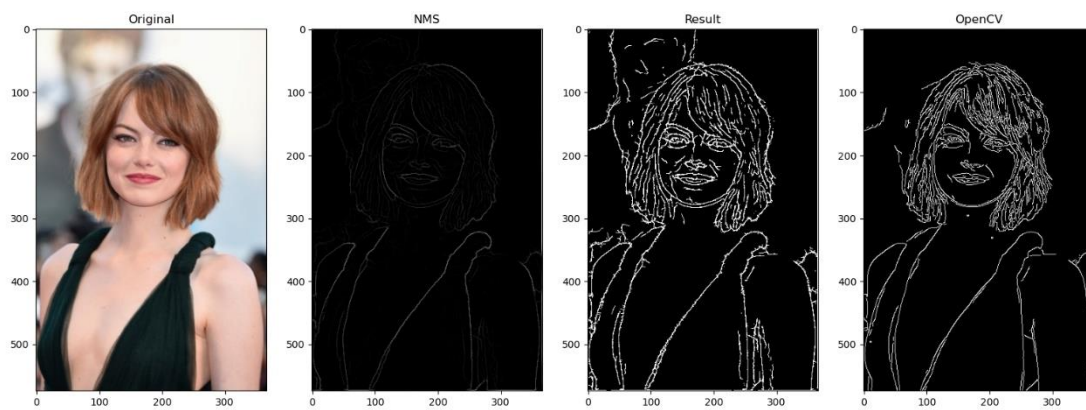
I. Original Picture:



II. Implement Method: Canny Edge Detection, Bilateral Filter

III. Experiments & Results:

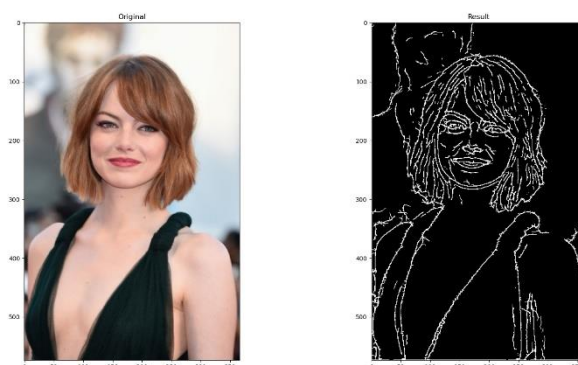
1. Implement and comparison



IV. Discussions:

The original picture is a RGB image. The result of non maximum suppression is the same image with thinner edges. Double threshold then links the strong edge points and weak edge points which is the neighbor of strong edge points. The comparison of OpenCV and our result shows that our implementation can retain more edges in the same criteria of thresholding.

V. Final Output:



Picture 2:

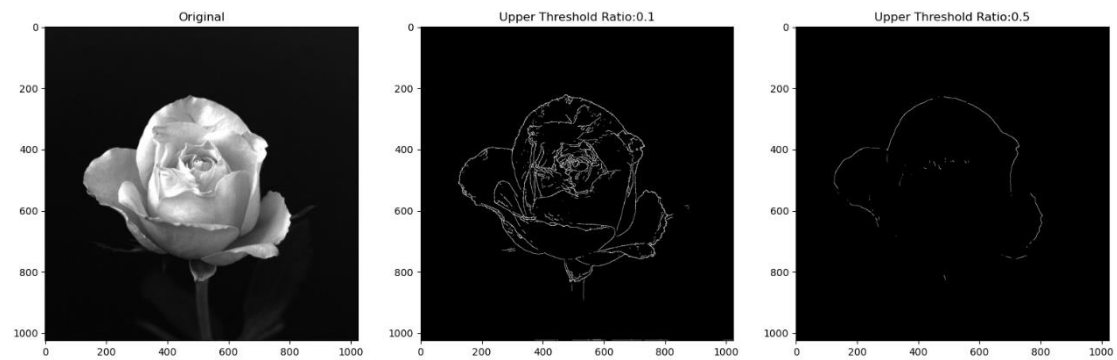
I. Original Picture:



II. Implement Method: Canny Edge Detection, Bilateral Filter

III. Experiments & Results:

1. Experiment different settings of upper threshold



IV. Discussions:

The original picture is a gray scale image. To test the performance of different upper threshold, we can observe that when the value of upper threshold becomes higher, the less the 'strong' edge points it can retain. The contour of rose become more blurred.

Picture 3:

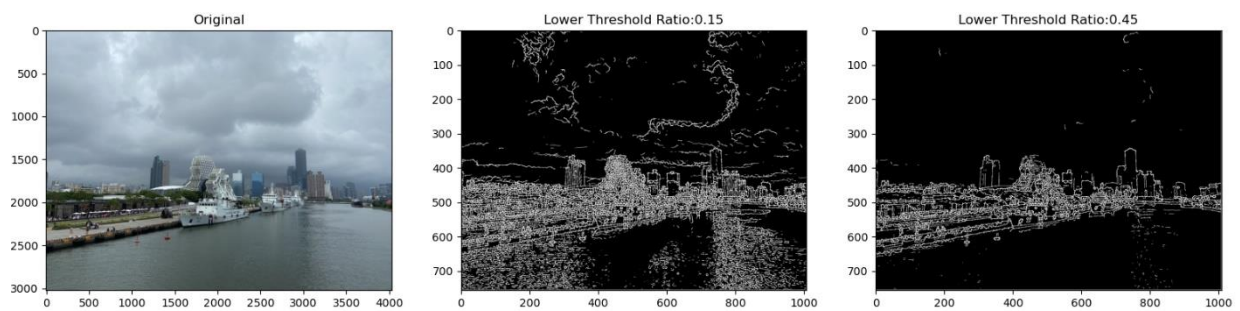
I. Original Picture:



II. Implement Method: Canny Edge Detection, Bilateral Filter

III. Experiments & Results:

1. Experiment different settings of lower threshold



IV. Discussions:

The original picture is a RGB image. To test the performance of different lower threshold, we can observe that when the value of lower threshold becomes higher, it can reduce more 'weak' edge points. The image has less edge points.

Picture 4:

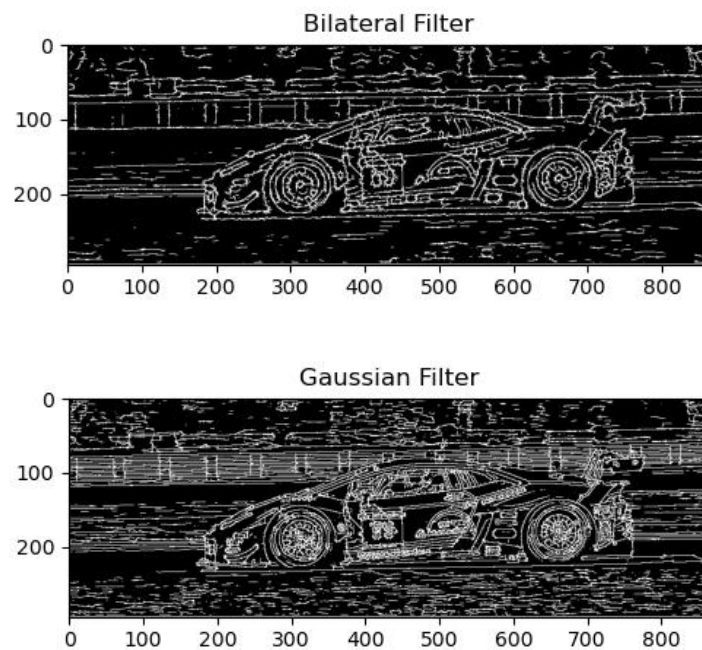
I. Original Picture:



II. Implement Method: Canny Edge Detection, Bilateral Filter, Gaussian Filter

III. Experiments & Results:

1. Experiment different smoothing filters with the same criteria of upper and lower threshold



IV. Discussions:

The original picture is a RGB image. To test the performance of different smoothing filter, we can observe that the result of using Gaussian filter can retain more edge points than bilateral filter. The result of Canny edge detection using Gaussian filter is more detailed.

Picture 5:

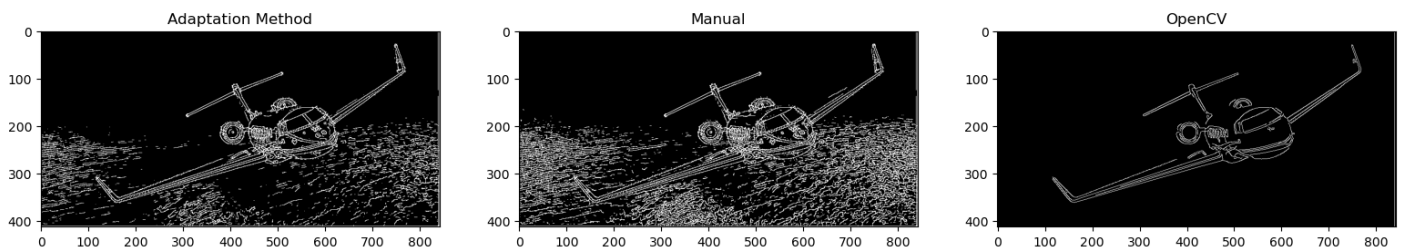
I. Original Picture:



II. Implement Method: Adaptive threshold of Canny Edge Detection, Gaussian Filter

III. Experiments & Results:

1. Experiment adaptive threshold of Canny Edge Detection



IV. Discussions:

The original picture is a RGB image. To test the performance of adaptive threshold of Canny Edge Detection, we can observe that the result of adaptive method had a better edge retain at the part of ocean. One can output a high performance of edge detection in the circumstance of do not need to set the upper and lower bound of threshold.

Code:

1. HW2.py

```
import cv2
import math
import numpy as np
import time
from scipy.signal import convolve2d as conv2
from matplotlib import pyplot as plt
from ProcFunc import *

class CannyEdgeDetection:
    def __init__(self, img, smooth_method = 'Bilateral',
                  threshold_mode = 'auto', l_threshold_ratio = 0.05,
                  u_threshold_ratio = 0.15):
        self.img = img
        #Obtain the time
        self.stime = time.time()
        self.angles = [0,45,90,135]
        self.nang = len(self.angles)
        self.smooth_method = smooth_method
        self.threshold_mode = threshold_mode
        #self.l_threshold = l_threshold
        #self.u_threshold = u_threshold
        self.highThreshold = u_threshold_ratio
        self.lowThreshold = l_threshold_ratio
        self.result = { }

    def deroGauss(self, w=5, s=1, angle=0):
        wlim = (w-1)/2
        y,x = np.meshgrid(np.arange(-wlim,wlim+1),np.arange(-wlim,wlim+1))
        G = np.exp(-np.sum((np.square(x),np.square(y)),axis=0)/(2*np.float64(s)**2))
        G = G/np.sum(G)
        dGdx = -np.multiply(x,G)/np.float64(s)**2
        dGdy = -np.multiply(y,G)/np.float64(s)**2
        angle = angle*math.pi/180
        dog = math.cos(angle)*dGdx + math.sin(angle)*dGdy
```

```

return dog

def bilatfilt(self, I, w, sd, sr):
    dim = I.shape
    Iout= np.zeros(dim)
    wlim = (w-1)//2
    y,x = np.meshgrid(np.arange(-wlim,wlim+1),np.arange(-wlim,wlim+1))
    g = np.exp(-np.sum((np.square(x),np.square(y)),axis=0)/(2*(np.float64(sd)**2)))
    Ipad = np.pad(I,(wlim,),'edge')
    for r in range(wlim,dim[0]+wlim):
        for c in range(wlim,dim[1]+wlim):
            Ix = Ipad[r-wlim:r+wlim+1,c-wlim:c+wlim+1]
            s = np.exp(-np.square(Ix-Ipad[r,c])/(2*(np.float64(sr)**2)))
            k = np.multiply(g,s)
            Iout[r-wlim,c-wlim] = np.sum(np.multiply(k,Ix))/np.sum(k)
    return Iout

# define function
# Use the gaussian filter to denoise
def get_edges(self, I, sd):
    dim = I.shape
    Idog2d = np.zeros((self.nang,dim[0],dim[1]))
    for i in range(self.nang):
        dog2d = self.deroGauss(5,sd, self.angles[i])
        Idog2dtemp = abs(conv2(I,dog2d,mode='same',boundary='fill'))
        Idog2dtemp[Idog2dtemp<0]=0
        Idog2d[i,:,:] = Idog2dtemp
    return Idog2d

# compute the gradient
def calc_sigt(self, I,threshval):
    M,N = I.shape
    ulim = np.uint8(np.max(I))
    N1 = np.count_nonzero(I>threshval)
    N2 = np.count_nonzero(I<=threshval)
    w1 = np.float64(N1)/(M*N)
    w2 = np.float64(N2)/(M*N)

```



```

try:
    u1 = sum(i*np.count_nonzero(np.multiply(I>i-0.5,I<=i+0.5))/N1 for i
in range(threshval+1,ulim))
    u2 = sum(i*np.count_nonzero(np.multiply(I>i-0.5,I<=i+0.5))/N2 for i
in range(threshval+1))

    uT = u1*w1+u2*w2
    sigt = w1*w2*(u1-u2)**2
except:
    return 0
return sigt

```

NMS

```

def nonmaxsup(self, I, gradang):
    dim = I.shape
    Inms = np.zeros(dim)
    xshift = int(np.round(math.cos(gradang*np.pi/180)))
    yshift = int(np.round(math.sin(gradang*np.pi/180)))
    Ipad = np.pad(I,(1,),'constant',constant_values = (0,0))
    for r in range(1,dim[0]+1):
        for c in range(1,dim[1]+1):
            maggrad = [Ipad[r-xshift,c-yshift],Ipad[r,c],Ipad[r+xshift,c+yshift]]
            if Ipad[r,c] == np.max(maggrad):
                Inms[r-1,c-1] = Ipad[r,c]
    return Inms

```

Obtain the best threshold

```

def get_threshold(self, I):
    max_sigt = 0
    opt_t = 0
    ulim = np.uint8(np.max(I))
    print(ulim,)
    for t in range(ulim+1):
        sigt = self.calc_sigt(I,t)
        if sigt > max_sigt:
            max_sigt = sigt
            opt_t = t
    print ('optimal high threshold: ',opt_t,)

```

```

    return opt_t

# Threshold
def threshold(self, I, uth):
    if self.threshold_mode == 'auto':
        lth = uth/2.5
        Ith = np.zeros(I.shape)
        Ith[I>=uth] = 255
        Ith[I<lth] = 0
        Ith[np.multiply(I>=lth, I<uth)] = 100
    else:
        highThreshold = I.max() * self.highThreshold
        lowThreshold = highThreshold * self.lowThreshold

        Ith = np.zeros(I.shape)
        Ith[I >= highThreshold] = 255
        Ith[I < lowThreshold] = 0
        Ith[np.multiply(I>=lowThreshold, I<highThreshold)] = 100

    return Ith

# hysteresis
def hysteresis(self, I):
    r,c = I.shape
    Ipad = np.pad(I,(1,),'edge')
    c255 = np.count_nonzero(Ipad==255)
    imgchange = True
    for i in range(1,r+1):
        for j in range(1,c+1):
            if Ipad[i,j] == 100:
                if np.count_nonzero(Ipad[r-1:r+1,c-1:c+1]==255)>0:
                    Ipad[i,j] = 255
            else:
                Ipad[i,j] = 0
    Ih = Ipad[1:r+1,1:c+1]
    return Ih

def detector(self):

```

```

self.result['Original'] = self.img
# Resize the image
while self.img.shape[0] > 1100 or self.img.shape[1] > 1100:
    self.img = cv2.resize(self.img, None, fx=0.5, fy=0.5, interpolation =
cv2.INTER_AREA)
# translate into gray scale
gimg = cv2.cvtColor(self.img, cv2.COLOR_BGR2GRAY)
# Obtain the size of image
dim = self.img.shape

## Start canny
#Smoothing
if self.smooth_method == 'Bilateral':
    #Bilateral filtering
    print ('Bilateral filtering...')
    gimg = self.bilatfilt(gimg, 5, 3, 10)
    print ('after bilat: ', np.max(gimg))
elif self.smooth_method == 'Gaussain':
    print ('Gaussain filtering...')
    gimg = gs_filter(gimg)
    print ('after bilat: ', np.max(gimg))

self.result['Smoothed'] = gimg

#Gradient of Image
print ('Calculating Gradient...')
img_edges = self.get_edges(gimg, 2)
print ('after gradient: ', np.max(img_edges))

#Non-max suppression : ( NMS )
print ('Suppressing Non-maximas...')
for n in range(self.nang):
    img_edges[n, :, :] = self.nonmaxsup(img_edges[n, :, :], self.angles[n])

print ('after nms: ', np.max(img_edges),)

img_edge = np.max(img_edges, axis=0)
lim = np.uint8(np.max(img_edge))

```

```

self.result['NMS'] = img_edge

# Compute the threshold
print ('Calculating Threshold...')
th = self.get_threshold(gimg)
the = self.get_threshold(img_edge)

# Obtain the best threshold
print ('Thresholding...')
img_edge = self.threshold(img_edge, the*0.25)

print ('Applying Hysteresis...')
img_edge = self.nonmaxsup(self.hysteresis(img_edge),90)

print( 'Time taken :: ', str(time.time()-self.stime)+' seconds...')

self.result['Result'] = img_edge
# cv2.imwrite(img_name+'3'+'.jpg', img_edge)
# cv2.imwrite(img_name+'4'+'.jpg', img_canny)

return self.result

if __name__=='__main__':
    parser = argparse.ArgumentParser()
    parser.add_argument("-image1" , type = str , default='Face.jpg')
    parser.add_argument("-image2" , type = str , default='rose.tif')
    parser.add_argument("-image3" , type = str , default='Koa2.jpg')
    parser.add_argument("-image4" , type = str , default='car2.jpg')
    parser.add_argument("-image5" , type = str , default='jet.jpg')
    args = parser.parse_args()

    image = cv2.imread(args.image4)

    result = { }
    # #My implementation
    detector = CannyEdgeDetection(image, smooth_method = 'Gaussain',

```

```

        threshold_mode    =    'auto',    l_threshold_ratio    =    0.15,
u_threshold_ratio = 0.10)
    result_auto = detector.detector()
    result['Adaptation Method'] = result_auto['Result']

    #Manual
    detector = CannyEdgeDetection(image, smooth_method = 'Gaussain',
        threshold_mode    =    'Manual',    l_threshold_ratio    =    0.15,
u_threshold_ratio = 0.10)
    result_m = detector.detector()
    result['Manual'] = result_m['Result']

    # #OpenCV
    gray = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
    # blurred = cv2.medianBlur(gray, 11)
    blurred = cv2.GaussianBlur(gray, (5, 5), 0)
    # blurred = cv2.bilateralFilter(gray, 5, 3, 10)
    canny_blurred = cv2.Canny(blurred, 30, 150)
    result['OpenCV'] = canny_blurred

    show_all_img(result)

```

```

2. ProcFunc.py
from copy import copy
from pickletools import uint8
import cv2
import argparse
import random
import numpy as geek
import skimage.util.noise as noise
import numpy as np
import matplotlib.pyplot as plt
import math
from scipy import ndimage

#Resize
def img_resize(image, resize_height, resize_width):

    image_shape=np.shape(image)
    height=image_shape[0]
    width=image_shape[1]

    if (resize_height is None) and (resize_width is None):
        return image

    #Resize height
    if resize_height is not None:
        resize_width=int(width* (resize_height/height) )
    #Resize width
    elif resize_width is not None:
        resize_height=int(height* (resize_width/width) )

    img = cv2.resize(image, dsize=(resize_width, resize_height))

    return img

#RGB -> Gray
def RGB2GRAY(image):
    return cv2.cvtColor(image , cv2.COLOR_BGR2GRAY)

```

#Padding

```
def padding(image , kernel_size):
```

```
    padding_size = (kernel_size[0] // 2, kernel_size[1] // 2)
```

```
    height = image.shape[0]
```

```
    width = image.shape[1]
```

```
    output_h = height + padding_size[0] * 2
```

```
    output_w = width + padding_size[1] * 2
```

```
    output = np.zeros( (output_h , output_w), dtype="uint8")
```

```
    for i in range(height):
```

```
        for j in range(width):
```

```
            output[i + padding_size[0]][j + padding_size[1]] = image[i][j]
```

```
    return output
```

#Convolution

```
def convolution(image, kernel):
```

```
    kernel_size = kernel.shape
```

```
    output = np.zeros(image.shape, dtype = "uint8")
```

```
    pad_img = padding(image , kernel_size)
```

```
    for i in range(image.shape[0]):
```

```
        for j in range(image.shape[1]):
```

```
            temp = []
```

```
            result = 0
```

```
            temp.append( pad_img[ i:i+kernel_size[0], j:j+kernel_size[1] ] )
```

```
            image_data = np.row_stack(temp).flatten()
```

```
            kernel_data = np.row_stack(kernel).flatten()
```

```
            #Calculate
```

```
            for k in range(len(kernel_data)):
```

```
                result += image_data[k] * kernel_data[k]
```

```
            output[i][j] = result
```

```
    return output
```

#Sobel Filter

def sobel_filter(image):

 Kx = np.array([[-1, 0, 1], [-2, 0, 2], [-1, 0, 1]], np.float32)

 Ky = np.array([[1, 2, 1], [0, 0, 0], [-1, -2, -1]], np.float32)

 gx = convolution (image, Kx)

 gy = convolution (image, Ky)

 G = np.hypot(gx, gy)

 G = G / G.max() * 255

 theta = np.arctan2(gx, gy)

 return (G, theta)

#Smoothing

#Average

def box_filter(image, kernel_size = (3, 3)):

 all = kernel_size[0]* kernel_size[1]

 kernel = np.array([1]*all).reshape(kernel_size)

 avg_kernel = (1/ all)*kernel

 if len(image.shape) == 2:

 output = convolution (image, avg_kernel)

 else :

 image_r = image[:, :, 0]

 image_g = image[:, :, 1]

 image_b = image[:, :, 2]

 output_r = convolution (image_r , avg_kernel)

 output_g = convolution (image_g , avg_kernel)

 output_b = convolution (image_b , avg_kernel)

 output = np.dstack((output_r, output_g, output_b))

 return output

#Gaussian Filter

```
def gs_filter(image, kernel_size = (3, 3)):
```

```
    #Gaussian filter
```

```
    x1 = 1 - math.ceil(kernel_size[0] / 2)
```

```
    x2 = math.ceil(kernel_size[0] / 2)
```

```
    y1 = 1 - math.ceil(kernel_size[1] / 2)
```

```
    y2 = math.ceil(kernel_size[1] / 2)
```

```
    x, y = np.mgrid[x1:x2, y1:y2]
```

```
    gaussian_kernel = np.exp(-(x**2+y**2))
```

```
    #Normalization
```

```
    gaussian_kernel = gaussian_kernel / gaussian_kernel.sum()
```

```
    if len(image.shape) == 2:
```

```
        output = convolution (image, gaussian_kernel)
```

```
    else :
```

```
        image_r = image[:, :, 0]
```

```
        image_g = image[:, :, 1]
```

```
        image_b = image[:, :, 2]
```

```
        output_r = convolution (image_r , gaussian_kernel)
```

```
        output_g = convolution (image_g , gaussian_kernel)
```

```
        output_b = convolution (image_b , gaussian_kernel)
```

```
        output = np.dstack((output_r, output_g, output_b))
```

```
    return output
```

#Median filter

```
def median_img_filter(image, kernel_size = (3, 3)):
```

```
    def median_filter(image, kernel_size):
```

```
        height = image.shape[0]
```

```
        width = image.shape[1]
```

```

output = np.zeros((height, width), dtype = "uint8")

#Padding
padding_image = padding(image , kernel_size)

#Find Median
for i in range(height):
    for j in range(width):
        temp = []
        temp.append( padding_image[ i:i+kernel_size[0],
j:j+kernel_size[1] ] )
        #Median
        median = np.median(np.row_stack(temp).flatten())
        output[i][j] = median
    return output

if len(image.shape) == 2:
    output = median_filter(image, kernel_size)
else:
    image_r = image[:, :, 0]
    image_g = image[:, :, 1]
    image_b = image[:, :, 2]

    output_r = median_filter(image_r , kernel_size )
    output_g = median_filter(image_g , kernel_size )
    output_b = median_filter(image_b , kernel_size )
    output = np.dstack((output_r, output_g, output_b))

return output

# Histogram Equalization
def hist_equa_img(image):

    def hist_equa(image):
        data = np.zeros(256).astype(np.int64)
        image_f = image.flatten()
        for i in image_f:
            data[int(i)] += 1

```

```

p = data/image_f.size
p_sum = geek.cumsum(p)
equal = np.around(p_sum * 255).astype('uint8')
output = equal[image_f].reshape(image.shape)

return equal[image]

if len(image.shape) == 2:
    output = hist_equa(image)
else:
    image_r = image[:, :, 0]
    image_g = image[:, :, 1]
    image_b = image[:, :, 2]

    output_r = hist_equa(image_r )
    output_g = hist_equa(image_g )
    output_b = hist_equa(image_b )
    output = np.dstack((output_r, output_g, output_b))

return output

```

#Bilateral Filter

```

def bilatfilt(I, w, sd, sr):
    dim = I.shape
    Iout= np.zeros(dim)
    wlim = (w-1)//2
    y,x = np.meshgrid(np.arange(-wlim,wlim+1),np.arange(-wlim,wlim+1))
    g = np.exp(-np.sum((np.square(x),np.square(y)),axis=0)/(2*(np.float64(sd)**2)))
    Ipad = np.pad(I,(wlim,),'edge')
    for r in range(wlim,dim[0]+wlim):
        for c in range(wlim,dim[1]+wlim):
            Ix = Ipad[r-wlim:r+wlim+1,c-wlim:c+wlim+1]
            s = np.exp(-np.square(Ix-Ipad[r,c])/(2*(np.float64(sr)**2)))
            k = np.multiply(g,s)
            Iout[r-wlim,c-wlim] = np.sum(np.multiply(k,Ix))/np.sum(k)
    return Iout

```

```

#Show Image
def show_all_img(result, columns = None, rows = 1):

    if columns is None:
        columns = len(result)
    fig = plt.figure(figsize=(20, 20))

    for i, img_n in enumerate(result):
        ax = fig.add_subplot(rows, columns, i+1)
        ax.title.set_text(img_n)
        # ax.set_title(i)
        img = result[img_n]
        if len(img.shape) == 2:
            plt.imshow(img, cmap='gray')
        else:
            img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
            plt.imshow(img)
    #plt.tight_layout()
    plt.show()

```