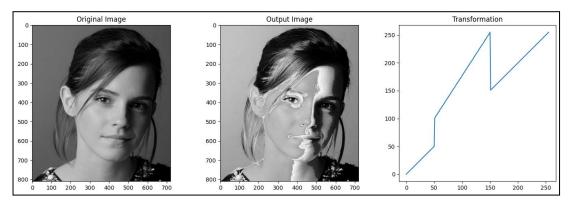
Only the necessary parts of the codes are included. (No imread,imshow functions are showed) Question 01

import cv2 as cv
import numpy as np
import matplotlib.pyplot as plt

```
coordinates = np.array([(0,0),(50,50),(50,100),(150,250),(150,150),(255,255)]) \\ trans = np.array([0]) \\ \textbf{for i in } range(int(len(coordinates)/2)): \\ line = np.linspace(coordinates[0+2*i,1], coordinates[1+2*i,1], coordinates[1+2*i,0] - coordinates[0+2*i,0] + 1) \\ trans = np.concatenate((trans, line)).astype(np.uint8) \\ transformed = trans[image1]
```



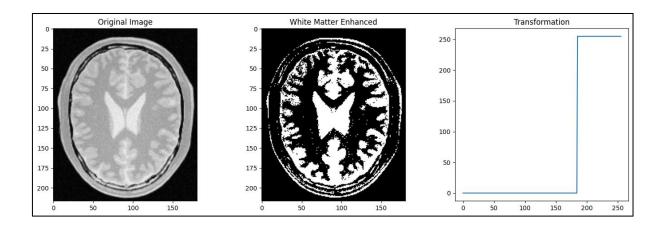
Question 02

a) White Matter Enhanced

```
x = np.arange(0,256).astype(np.uint8)
divert_point = 184
coordinates = np.array([(1,0),(divert_point,0),(divert_point+1,255),(255,255)])
trans = np.array([0])
```

for i in range(int(len(coordinates)/2)):

line = np.linspace(coordinates[0+2*i,1], coordinates[1+2*i,1], coordinates[1+2*i,0] - coordinates[0+2*i,0]+1) trans = np.concatenate((trans, line)).astype(np.uint8) transformed = trans[image2]



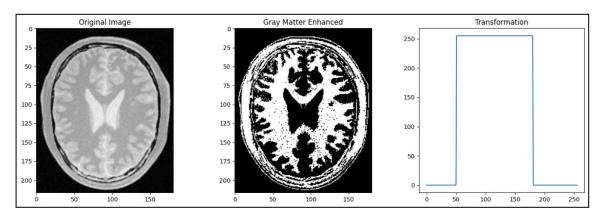
b) Gray Matter Enhanced

```
coordinates = np.array([(1,0),(50,0),(50+1,255),(180,255),(180+1,0),(255,0)]) trans = np.array([0])
```

for i in range(int(len(coordinates)/2)):

 $line = np.linspace(coordinates[0+2*i ,1], coordinates[1+2*i ,1], coordinates[1+2*i ,0] - coordinates[0+2*i ,0] + 1) \\ trans = np.concatenate((trans, line)).astype(np.uint8)$

transformed = trans[image2]



Question 03

image3_copy = image3.copy()

image3_copy = cv.cvtColor(image3_copy,cv.COLOR_BGR2RGB)

image3 = cv.cvtColor(image3, cv.COLOR_BGR2LAB)

l_img , a_img , b_img = cv.split(image3)

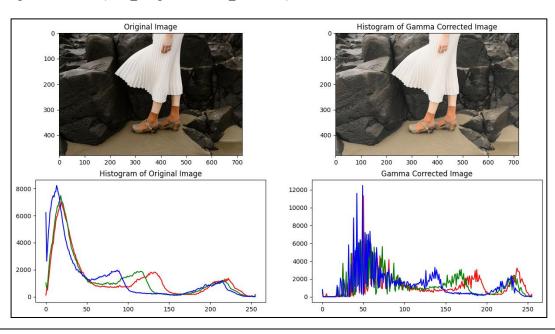
gamma = 0.5

table = np.array([(i/255.0)**(gamma)*255.0 for i in np.arange(0,256)]).astype(np.uint8)

img_gamma = cv.LUT(l_img, table)

final_image = cv.merge((img_gamma, a_img, b_img))

final_image = cv.cvtColor(final_image, cv.COLOR_LAB2RGB)



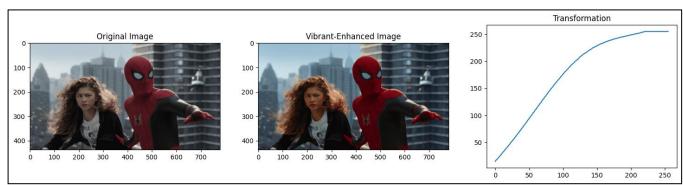
Discussion

- L channel: This plane represent the Lightness / Brightness of image. When adding gamma correction with gamma=0.5, it will increase the brightness of the L plane and the final output image.
- A plane: This plane represent the color from green to red. Negative value correspond to green while positive values correspond to red.
- B plane: This plane represent the color from blue to yellow. Negative value correspond to blue while positive values correspond to yellow.

Question 04

```
hsv_image = cv.cvtColor(image4, cv.COLOR_BGR2HSV)
hue, saturation, value = cv.split(hsv_image)
a , alpha = 0.65, 70
x = np.arange(0,256,1).astype(np.uint8)
table = np.array([min(255, (x + (a * 128) * np.exp(-((x - 128) ** 2) / (2 * alpha ** 2))))) for x in np.arange(0, 256)]).astype('uint8')
sat_update = cv.LUT(saturation, table)
merge_img = cv.merge((hue, sat_update , value))
merge_img = cv.cvtColor(merge_img, cv.COLOR_HSV2RGB)
```

'a' value that output a visually pleasing image: 0.65

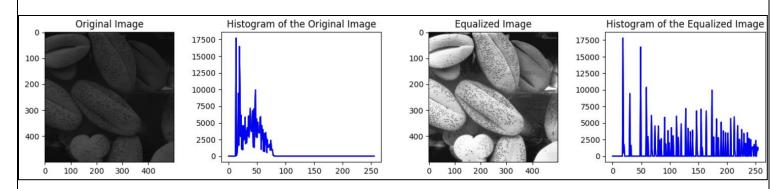


Question 05

```
height, width = image5.shape[0], image5.shape[1]
r_plane, g_plane, b_plane = cv.split(image5)
def intensity_count(plane) :
  count list = np.zeros(256).astype(np.uint16)
  for i in range(height):
    for j in range(width):
      count_list[plane[i][j]] += 1
  return count list
def histogram_eq(plane, intensity_count) :
  equalized intensities = np.zeros(256).astype(np.uint16)
  count = 0
  for i in range(len(intensity_count)) :
    count += intensity count[i]
    new val = (255/(height*width)) * count
    new val = round(new val)
    equalized_intensities[i] = new_val
```

```
for h in range(height) :
    for w in range(width) :
        plane[h][w] = equalized_intensities[plane[h][w]]
    return plane

r_plane_count = intensity_count(r_plane)
eq_r_plane = histogram_eq(r_plane, r_plane_count)
g_plane_count = intensity_count(g_plane)
eq_g_plane = histogram_eq(g_plane, g_plane_count)
b_plane_count = intensity_count(b_plane)
eq_b_plane = histogram_eq(b_plane, b_plane_count)
new_image = cv.merge((eq_r_plane,eq_g_plane,eq_b_plane))
```



Question 06 Part a,b,c

h_plane, s_plane, v_plane = cv.split(image6_hsv)
th_val, binary_image = cv.threshold(s_plane, 11, 255, cv.THRESH_BINARY)
result = cv.bitwise_and(image6, image6, mask = binary_image)

Part d

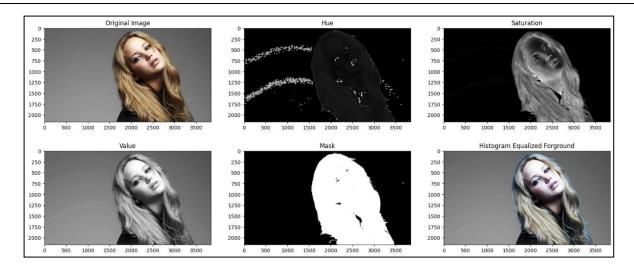
hist, bins = np.histogram(result.ravel(),256,[0,256]) cdf = hist.cumsum() cdf_normalized = cdf * hist.max() / cdf.max()

Part e

 $result_r \ , \ result_g, \ result_b = cv.split(result) \\ equ_r, \ equ_g, \ equ_b = cv.equalizeHist(result_r), \ cv.equalizeHist(result_g), \ cv.equalizeHist(result_b) \\$

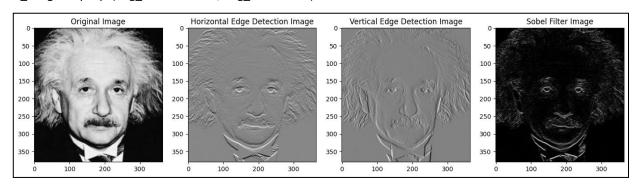
Part f

equalized_result = cv.merge((equ_r, equ_g, equ_b))
backgnd_mask = cv.bitwise_not(binary_image)
backgnd = cv.bitwise_and(image6, image6, mask = backgnd_mask)
final = cv.add(equalized_result, backgnd)



Question 07 Part a

image7 = cv.imread('einstein.png', cv.IMREAD_GRAYSCALE).astype(np.float32) vertical_sobel = np.array([[-1,-2,-1],[0,0,0],[1,2,1]], dtype = np.float32) horizontel_sobel = np.array([[-1,0,1],[-2,0,2],[-1,0,1]], dtype = np.float32) img_horizontal = cv.filter2D(image7, -1, vertical_sobel) img_vertical = cv.filter2D(image7, -1, horizontel_sobel) final_image = np.sqrt(img_horizontal**2, img_vertical**2)



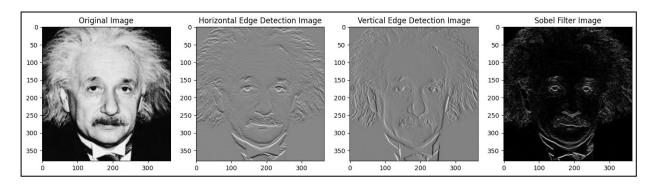
Part b

def sobel fiter(image, kernel):

```
h_im, w_im = image.shape[0] , image.shape[1]
h_ke, w_ke = kernel.shape[0] , kernel.shape[1]
h_ke_half, w_ke_half = h_ke // 2, w_ke // 2
result_h=np.zeros(image.shape,dtype=np.float32)
image = cv.normalize(image.astype('float'), None, 0.0, 1.0, cv.NORM_MINMAX)

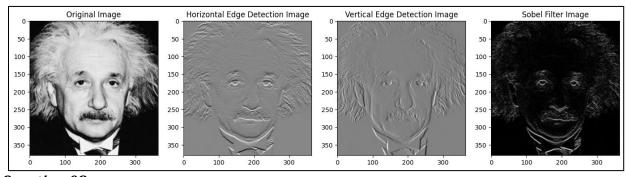
for i in range(h_ke, h_im - h_ke):
    for j in range(w_ke, w_im - w_ke):
# indentify the image window to get the 2D convolution with kernal, then convolution
    image_window = image[i - h_ke_half : i + h_ke_half + 1, j - w_ke_half : j + w_ke_half + 1]
    result_h[i, j] = np.sum(image_window * kernel)
result_h = (result_h - np.min(result_h)) / (np.max(result_h) - np.min(result_h)) * 255
return result_h
```

```
img_h = sobel_fiter(image7, vertical_sobel)
img_v = sobel_fiter(image7, horizontel_sobel)
final_image = np.sqrt(img_horizontal**2, img_vertical**2)
```



Part C

```
kernal_vertical = np.array([-1,0,1],dtype=np.float32)
kernal_horizontal = np.array([1,2,1],dtype=np.float32)
img_vertical = cv.sepFilter2D(image7, -1, kernal_vertical, kernal_horizontal)
img_horizontal = cv.sepFilter2D(image7, -1, kernal_horizontal, kernal_vertical)
final_image = np.sqrt(img_horizontal**2, img_vertical**2)
```



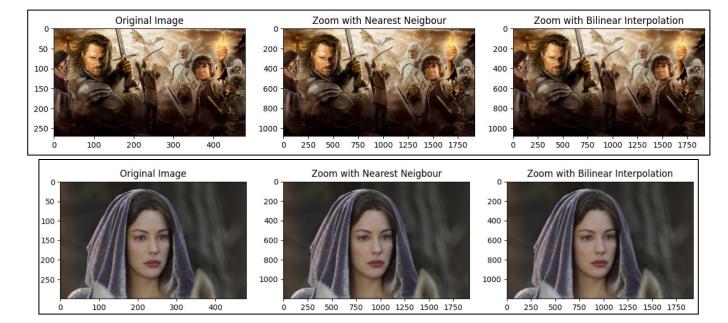
Question 08

```
def zoom(image, scale, method):
    h_image , w_image , channels = image.shape
    h_zoomed = h_image*scale
    w_zoomed = w_image*scale
zoomed_image = np.zeros((h_zoomed, w_zoomed, channels), dtype = np.float32)
if method == 'nearest neighbour':
    for i in range(h_zoomed):
        zoomed_image[i,j] = image[i // scale, j // scale]

if method == 'bilinear interpolation':
    for i in range(h_zoomed):
        for j in range(w_zoomed):
        x_image , y_image = (i / scale) , (j / scale)
        x1 , y1 = int(x_image) , int(y_image)
        x2 , y2 = x1+1, y1+1
```

```
if x2 >= h_image : x2 = x1
if y2 >= w_image : y2 = y1
dx, dy = x_image - x1 , y_image - y1
w1,w2,w3,w4 = (1 - dx)*(1 - dy),dx*(1 - dy),(1 - dx)*dy, dx*dy
zoomed_image[i,j] = w1*image[x1,y1] + w2*image[x1,y2] + w3*image[x2,y1] + w4*image[x2,y2]
return zoomed_image.astype(np.uint8)
```

zoomed_nn = zoom(image8, 4, 'nearest neighbour')
zoomed_bi = zoom(image8, 4, 'bilinear interpolation')



Discussion

Nearest neighbour method gives a sharp zoomed images while bilinear interpolation gives a smooth zoomed images. It happens because, interpolation method is taking the weighted sum of the four nearest pixel values. Therefore sharp edges will smooth when zooming images using bilinear interpolation.

Question 08 algorithm testing (for image 01)

```
zoomed_image = cv.imread('zooming/im01.png')
resized = cv.resize(image8, (1920, 1080), interpolation=cv.INTER_LINEAR)
squared_diff1 = np.sum(np.square(zoomed_image - zoomed_nn))
squared_diff2 = np.sum(np.square(zoomed_image - zoomed_bi))
squared_diff3 = np.sum(np.square(zoomed_image - resized))
```

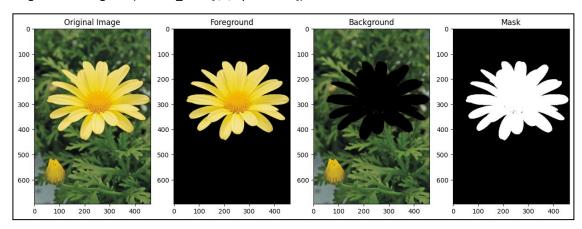
Squared Difference of zoomed image by Nearest Neighbour: 522633462 Squared Difference of zoomed image by the Bilinear interpolation: 543747100 Squared Difference of zoomed image by the resize Function: 522553013

Question 09

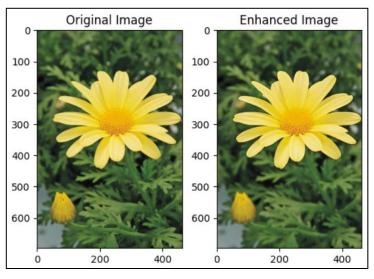
```
h, w, _ = image9.shape
mask = np.zeros((h, w), dtype = np.uint8)
bgdModel = np.zeros((1,65),np.float64)
```

```
fgdModel = np.zeros((1,65),np.float64)
rectangle = (0,0,h,w)
mask, bg_model, fg_model = cv.grabCut(image9, mask, rectangle, bgdModel, fgdModel, 5,
cv.GC_INIT_WITH_RECT)

new_mask = np.where((mask == 0) | (mask == 2), 0, 1).astype(np.uint8)
foreground = image9 * new_mask[:, :, np.newaxis]
background = image9 * (1 - new_mask[:, :, np.newaxis])
```



Part B
blur_bg = cv.GaussianBlur(background, (11,11), 0,0)
final_image = cv.add(foreground, blur_bg)



Part C

- When background is blurred, it will reduce the contrast and details at the edge where the background and foreground get separated. Therefore, there can be some deformations around the edge and it could result some dark appearences around the edge of the foreground.
- When Grabcut segmentation is used to segment the foreground and background, it may
 missclassified some pixel values around the edge between the foreground and background.
 Therefore, there could be some dark appearence closer to the edge of the foreground.

Github Link: https://github.com/SadeepRathnayaka/EN3160 Assignment1