# Assignment 1 - EN2550

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### Question 1

Created the transformation by horizontally stacking three numpy arrays

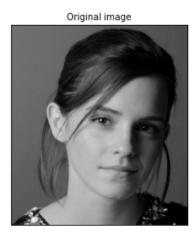
```
t1 = np.arange(0, 50, 1)

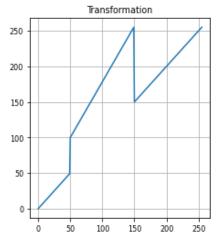
t2 = np.linspace(100, 255, 100)

t3 = np.arange(150, 256, 1)

t = np.hstack((t1, t2, t3))
```

Then used the CV2.LUT (Look up table) function to transform the image







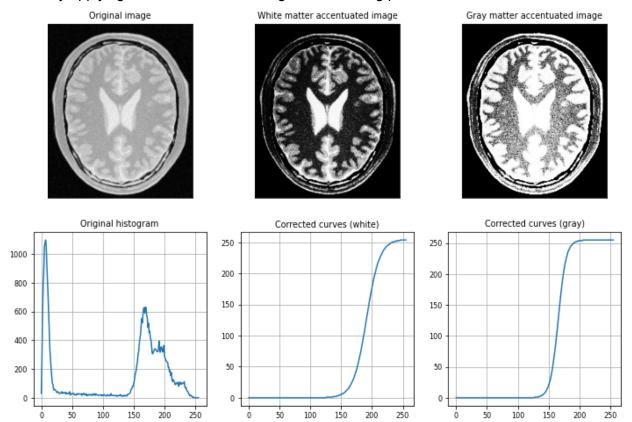
# Question 2

Used a transformed tanh function to do accentuate the required parts

```
def tanh_f(x):
    num = np.exp(x) - np.exp(-x)
    den = np.exp(x) + np.exp(-x)
    return num/den

def accentuating_transformation(center, width):
    center = center
    width = width
    x = np.linspace(center-width, center+width, 256)
    tanh = tanh_f(x)
    max_val = np.max([np.max(tanh), np.abs(np.min(tanh))])
    scaled_f = tanh/max_val*127.5
    shifted_f = scaled_f + 127.5
    quantized_f = np.round(shifted_f).astype(np.uint8)
    return quantized_f
```

By applying these transformations, got the following parts enhanced



# Question 3

- The L channel is what determines the luminosity of the image (also known as the Luma channel in graphic designing)
- Hence, by tweaking the Luma channel, the pixel values will be mapped to different ranges based on the **overall brightness** relevant to that pixel
- Used the following gamma transforms

```
gamma_light = 2
t_light = np.array([(p/255)**gamma_light*255 for p in range(256)]).astype(np.uint8)
gamma_dark = 0.8
t_dark = np.array([(p/255)**gamma_dark*255 for p in range(256)]).astype(np.uint8)
```

 Then converted the image to the LAB color space and applied the above transformations to the Luma channel

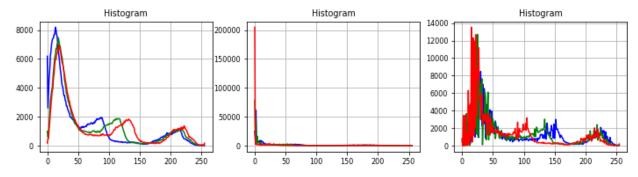
```
LAB_light_img = <u>cv</u>.cvtColor(inp_img, <u>cv</u>.COLOR_BGR2LAB)

LAB_light_img[:,:,0] = <u>cv</u>.LUT(LAB_light_img[:,:,0], t_light)
```







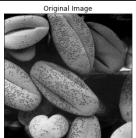


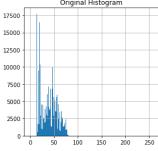
# Question 4

• Used the following function

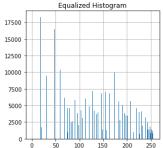
```
def getHistBalImg(img):
    '''
    Takes any input image channel from the user and
    return the channel by equalizing the histogram
    '''
    L = 256
    MN = img.size
    return_img = np.zeros(img.shape, dtype='uint8')
    intensity_bins = np.zeros(L)
    for i in range(L):
        intensity_bins[i] = img[img==i].size
        cum_intensity_bins = intensity_bins.cumsum()
        scaled_cum_intensity_bins = (L-1)/MN*cum_intensity_bins
        return_img = scaled_cum_intensity_bins[img]
    return return_img

Original Image
Original Histogram
Histogram Equalized Image
Equalized Histogram
```







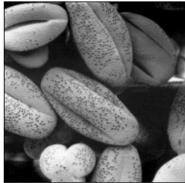


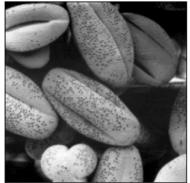
### Question 5

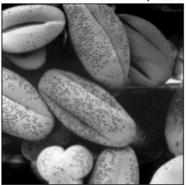
 Used the following function which handles both Nearest Neighbour Method and Bilinear Interpolation Method, and all worked fine

```
zoom type may take the two values 0 and 1
       0: Will zoom with nearest neighbour method
       1: Will zoom with bilinear interpolation
   rows = inp_img.shape[0]*scale
   cols = inp img.shape[1]*scale
      out_img = np.zeros((rows, cols))
       for i in range(rows):
            for j in range(cols):
                out img[i][j]
inp img[np.round(i/scale).astype(int)-1][np.round(j/scale).astype(int)-1]
       for i in range(rows):
           for j in range(cols):
                p1 = inp img[int(i/scale)][int(j/scale)]
                p2 = inp_img[int(i/scale)][int(j/scale)]
                if int((i/scale)+1) < rows/scale:</pre>
                    p2 = inp_img[int(i/scale)+1][int(j/scale)]
                p3 = inp_img[int(i/scale)][int(j/scale)]
                if int((j/scale)+1) < cols/scale:</pre>
                    p3 = inp img[int(i/scale)][int(j/scale)+1]
                p4 = inp_img[int(i/scale)][int(j/scale)]
                if int((j/scale)+1) < cols/scale and int((i/scale)+1) < rows/scale:</pre>
                    p4 = inp_img[int(i/scale)+1][int(j/scale)+1]
                x1 = p1*(int(j/scale)+1-j/scale)+p3*(j/scale-int(j/scale))
                x2 = p2*(\underline{int}(j/scale)+1-j/scale)+p4*(j/scale-\underline{int}(j/scale))
                x = x2*(i/scale - int(i/scale)) + x1*(int(i/scale) + 1 - i/scale)
                out img[i][j] = x
       out_img = np.round(out_img).astype(int)
   return out_img
```

Original image Zoomed with Nearest Neighbour Methadomed with Bilinear Interpolation







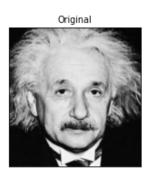
# Question 6

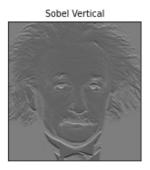
(a)

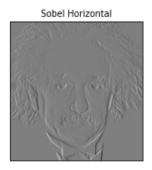
• The casual method,

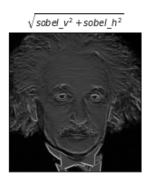
```
sobel_v_kernel = np.array([[-1, -2, -2], [0, 0, 0], [1, 2, 1]], dtype=np.float32)
img_sobel_v = cv.filter2D(img, -1, sobel_v_kernel)
sobel_h_kernel = np.array([[-1, 0, 1], [-2, 0, 2], [-1, 0, 1]], dtype=np.float32)
img_sobel_h = cv.filter2D(img, -1, sobel_h_kernel)
img_grad = np.sqrt(img_sobel_v**2 + img_sobel_h**2)
```

Resulted following









(b)

· Defined the following function, which resulted in the same

```
def sobelFilt(inp_img, v_h):
    v_h can take two values
        0: vertical sobel kernel
        1: horizontal sobel kernel
        ...
```

```
sobel_kernel = np.zeros((3,3))
       sobel_kernel = np.array([[-1, -2, -2], [0, 0, 0], [1, 2, 1]],
dtype=np.float32)
       sobel kernel = np.array([[-1, 0, 1], [-2, 0, 2], [-1, 0, 1]],
dtype=np.float32)
       raise Exception("Invalid v_h parameter")
   zero_row = np.zeros(arr.shape[1]+2)
   zero col = \underline{np}.zeros((arr.shape[0], 1))
   zero padded = np.vstack((zero row, np.hstack((zero col, arr, zero col)),
zero row))
   out_img = np.zeros(inp img.shape)
   for i in range(inp_img.shape[0]):
        for j in range(inp img.shape[1]):
            out_img[i][j] = np.sum(np.multiply(zero_padded[i:i+3,
j:j+3],sobel kernel))
   return out img
```

### (c)

 Defined the following function which reduced the computational time considerably. Which also resulted in the same

```
def zeroPad(inp_arr):
    arr = inp_arr
    zero_row = np.zeros(arr.shape[1]+2)
    zero_col = np.zeros((arr.shape[0], 1))
    zero_padded = np.vstack((zero_row, np.hstack((zero_col, arr, zero_col)),
zero_row))
    return zero_padded

def sobelFiltOptim(inp_img, v_h):
    v_h can take two values
        0: vertical sobel kernel
        1: horizontal sobel kernel
        1: horizontal sobel kernel
        1: sobel_1 = np.zeros((3,1))
        sobel_2 = np.zeros((1,3))
        if v_h == 0:
            sobel_1 = np.array((1,0,-1)).reshape((3,1))
            sobel_2 = np.array((1,2,1)).reshape((1,3))
        elif v_h == 1:
```

```
sobel_1 = np.array((1,2,1)).reshape((3,1))
sobel_2 = np.array((1,0,-1)).reshape((1,3))
else:
    raise Exception("Invalid v_h parameter")
zero_padded = zeroPad(inp_img)
out_img = np.zeros(inp_img.shape)
for i in range(inp_img.shape[0]):
    for j in range(inp_img.shape[1]):
        out_img[i][j] = np.sum(np.multiply(zero_padded[i, j:j+3],sobel_1))
zero_padded = zeroPad(out_img)
for i in range(inp_img.shape[0]):
    for j in range(inp_img.shape[1]):
        out_img[i][j] = np.sum(np.multiply(zero_padded[i:i+3, j],sobel_2))
return out_img
```

#### Question 7

(a)

• Used the following code

```
mask = np.zeros(inp_img.shape[:2],np.uint8)
bgdModel = np.zeros((1, 65), np.float64)
fgdModel = np.zeros((1, 65), np.float64)
rect = (56, 154, 500, 384)
cv.grabCut(inp_img, mask, rect, bgdModel, fgdModel, 5, cv.GC_INIT_WITH_RECT)
mask2 = np.where((mask == 2) | (mask == 0), 0, 1).astype('uint8') # foreground
mask3 = np.zeros(mask2.shape) # background
mask3[mask2 == 0] = 1
inp_img = cv.cvtColor(inp_img, cv.COLOR_BGR2RGB)
img_foreground = inp_img*mask2[:,:,np.newaxis].astype(np.uint8)
img_backgound = inp_img*mask3[:,:,np.newaxis].astype(np.uint8)
```

• 5 iterations we enough to get a decent output







(b)

• Blurred the background, applied the background mask again, added to the foreground

```
blured_background = cv.GaussianBlur(img_backgound, (15,15), 10)
enhanced_img = blured_background*mask3[:,:,np.newaxis].astype(np.uint8)
enhanced_img += img_foreground
```

Yielded the following





(c)

• Because when blurring, the black pixels in the background gets also added to the neighboring pixels (in this case, the boundary pixels of the background and foreground)