EN2550 - Fundamentals of Image Processing and Machine Vision

Assignment 1



Submitted by

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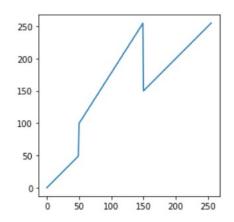
GitHub Repository link

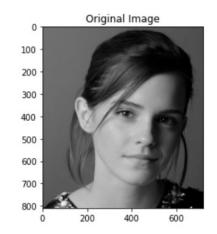
https://github.com/dulmi-19/Image-Processing-and-Machine-Vision

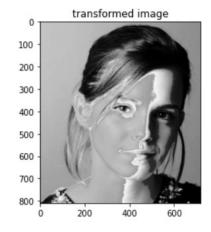
Question 1

```
# linearly spaced values
t1=np.linspace(0,50,50)
t2=np.linspace(100,255,100)
t3=np.linspace(150,255,106)

transform = np.concatenate((t1,t2,t3), axis = 0).astype(np.uint8) #connect all 3 transformations
assert len(transform) == 256
emma_transformed= cv.LUT(emma,transform)
```







The intensity transformation is implemented using a lookup table. In this transformation the lower and higher grey values of input image are unchanged, but the middle values in (50-100) range are changed to (100-255) range

Question 2 (a) white matter

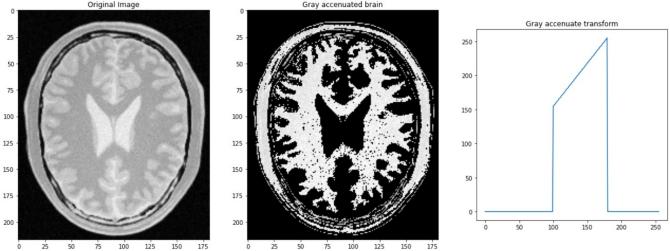
```
# accenuate white
  p1=np.linspace(0,0,190)
  p2=np.linspace(200,255,66)
  accenuate white = np.concatenate((p1,p2), axis = 0).astype(np.uint8) #connect all transformations
  assert len(accenuate white) == 256
 white_accenuated_brain = cv.LUT(brain,accenuate_white)
               Original Image
                                                                                            White accenuate transform
25
                                                                                 250
50
                                                                                 200
75
                                                                                 150
100
                                        100
                                        125
125
                                                                                 100
150
                                        150
175
200
                                                                                                       150
                                                                                                             200
                                                                                                                   250
```

To accentuate white matter, the intensity of color values near white color(255) are increased. Because of the gradient, we can compare the intensity variation as well.

Question 2 (b) gray matter

```
# accenuate Grey
t1=np.linspace(0,0,100)
t2=np.linspace(155,255,80)
t3=np.linspace(0,0,76)
accenuate_grey = np.concatenate((t1,t2,t3), axis = 0).astype(np.uint8) #connect all transformations
assert len(accenuate_grey) == 256
grey_accenuated_brain = cv.LUT(brain,accenuate_grey)
Original Image

Gray accenuated brain
```

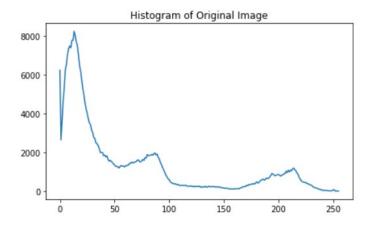


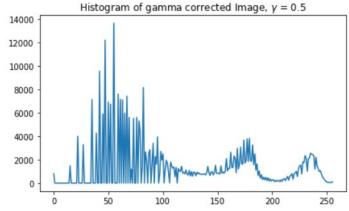
To accentuate gray matter, the intensity of middle range color values are increased while keeping the other color values at 0. Because of the gradient, we can compare the intensity variation as well.

Question 3

```
im_lab = cv.cvtColor(im, cv.COLOR_BGR2Lab)
  (l,a,b)= cv.split(im_lab) # split the l,a,b components
  gamma = 0.5
  g transform = np.array([(p/255)**gamma*255 for p in range(0,256)]).astype(np.uint8)
  g transformed = cv.LUT(l, g transform) # gamma correction for lightness
  im_lab[:,:,0]=g_transformed
  im hist = cv.calcHist([im],[0],None,[256],[0,256])
  im_lab_hist = cv.calcHist([im_lab],[0],None,[256],[0,256])
                Original Image
                                                                        y = 0.5
100
                                                    100
200
                                                    200
300
                                                    300
400
                                                    400
       100
            200
                       400
                            500
                                                            100
                                                                                 500
```

The gamma transform is applied to L plane of CIELAB color space. Therefore, the details in darker areas are more visible in the gamma transformed image.





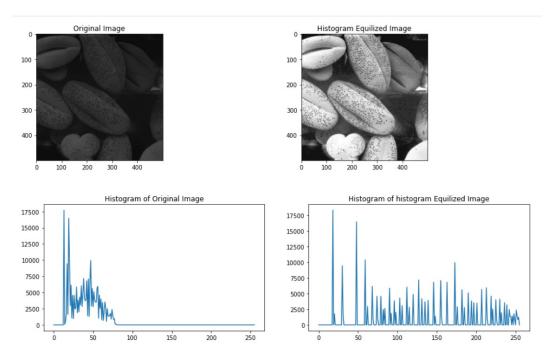
Question 4

```
shells = cv.imread("shells.png",cv.IMREAD_GRAYSCALE)
assert shells is not None
shells_hist = cv.calcHist([shells],[0],None,[256],[0,256])

#normalize
n_pixels = np.sum(shells_hist)
shells_histo = shells_hist/n_pixels

#normalized cumulative histogram
cdf_hist = np.cumsum(shells_histo)

transform = np.floor(255 * cdf_hist).astype(np.uint8)
eq_shells = cv.LUT(shells,transform)
equi_shells_histo = cv.calcHist([eq_shells],[0],None,[256],[0,256])
```



According to the histogram of original image, we can see that the color values are concentrated in the lower region of values. After the histogram equalization, the intensity values are distributed all over the region, therefore we can see the details of the image clearly. In here, we are using discrete intensity values instead of continuous values.

Question 5

```
def zoom_im(image, method, scaling_fact):
   rows, cols =scaling_fact* image.shape[0], scaling_fact* image.shape[1]
   zoomed_img = np.zeros((rows,cols,3),dtype=image.dtype)
if (method == 0):
        #nearest-neighbor method
        for r in range(rows):
            for c in range(cols):
                im_r, im_c= int(np.round(r/scaling_fact)), int(np.round(c/scaling_fact))
                if im_r >= image.shape[0]:
                    im_r -=1
                if im_c >= image.shape[1]:
                    im_c -=1
                zoomed_img[r][c]= image[im_r][im_c]
        return zoomed_img
        #bilinear interpolation method
        for r in range(rows-1):
            for c in range(cols-1):
                im_r, im_c= r/scaling_fact, c/scaling_fact
                left_bottom = [math.floor(im_r), math.floor(im_c)]
                right_bottom = [math.floor(im_r), math.ceil(im_c)]
                left_top = [math.ceil(im_r), math.floor(im_c)]
                right_top = [math.ceil(im_r),math.ceil(im_c)]
                if left_top[0] >= image.shape[0]:
                    left_top[0] =image.shape[0] - 1
                    right_top[0] = image.shape[0] - 1
                if right_bottom[1] >= image.shape[1]:
                    right_bottom[1]=image.shape[1] - 1
                    right_top[1] = image.shape[1] - 1
                ratio_vertical = im_r-left_bottom[0]
                ratio_horizontal = im_c-left_bottom[1]
                horizontal\_1 = ratio\_vertical*image[left\_top[0]][left\_top[1]] + (1-ratio\_vertical)*image[left\_bottom[0]][left\_bottom[1]]
                horizontal\_2 = ratio\_vertical*image[right\_top[0]][right\_top[1]] + (1-ratio\_vertical)*image[right\_bottom[0]][right\_bottom[1]]
                {\tt zoomed\_img[r][c]= np.rint(horizontal\_1*(1-ratio\_horizontal) + horizontal\_2*ratio\_horizontal)}
        return zoomed_img
```



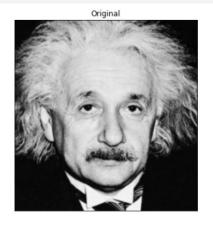
```
nn_ssd=np.sum((org_img_2[:,:,0:3]-nn_2[:,:,0:3])**2)/org_img_2.size
bi_ssd=np.sum((org_img_2[:,:,0:3]-bi_2[:,:,0:3])**2)/org_img_2.size
print("SSD for nearest-neighbor", nn_ssd)
print("SSD for bilinear interpolation", bi_ssd)
```

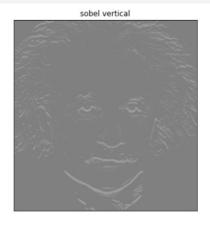
SSD for nearest-neighbor 16.792970920138888 SSD for bilinear interpolation 16.40395240162037

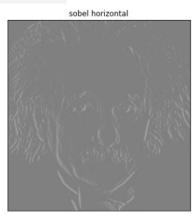
Question 6 (a)

```
einstein = cv.imread("einstein.png",cv.IMREAD_GRAYSCALE)
assert einstein is not None

sobel_v = np.array([(-1,-2,-1),(0,0,0),(1,2,1)],dtype=np.float32)
einstein_x = cv.filter2D(einstein,-1,sobel_v)
sobel_h = np.array([(-1,0,1),(-2,0,2),(-1,0,1)],dtype=np.float32)
einstein_y = cv.filter2D(einstein,-1,sobel_h)
```





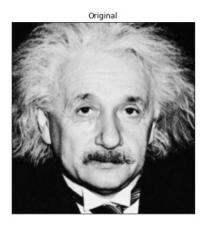


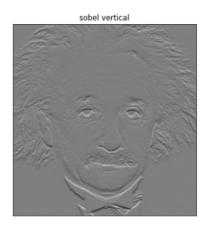
Question 6 (b)

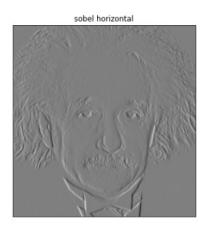
```
def sobel_filter(image,kernal):
    [r,c]= image.shape
    filtered_image = np.zeros((r,c),dtype=np.float32)
    padded_array = np.zeros((r+2,c+2),dtype=np.float32)
    padded_array = np.zeros((r+2,c+2),dtype=np.float32)
    padded_array[1:r+1,1:c+1] = image
    n=1
    m=1
    height_kernal=math.floor(kernal.shape[0]/2)
    if len(kernal.shape)==1:
        width_kernal=0
        m=2
    else:
        width_kernal=math.floor(kernal.shape[1]/2)
        if kernal.shape[1]==1:
            | m=2
            | fernal.shape[0]==1:
            | n=2

            | for row in range(height_kernal,r+2-height_kernal-n):
            | for col in range(width_kernal,c+2-width_kernal-m):
            | filtered_image[row,col]=np.dot(padded_array[row-height_kernal:row+height_kernal+1,col-width_kernal:col+width_kernal+1].flatten(),kernal.flatten())
    return filtered_image

image_v = sobel_filter(einstein,sobel_v)
    image_h = sobel_filter(einstein,sobel_v)
    image_h = sobel_filter(einstein,sobel_v)
```



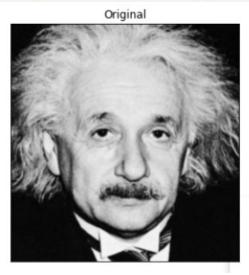


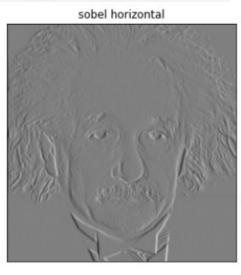


The outputs of filter 2D filter and the user defined code are similar. In the user defined code, some depth details can be seen.

Question 6 (c)

```
sobel_h1=np.array([(1),(2),(1)] ,dtype=np.float32)
sobel_h2=np.array([(1,0,-1)] ,dtype=np.float32)
filltered_1=sobel_filter(einstein,sobel_h1)
Filtered_2=sobel_filter(filltered_1,sobel_h2)
```





The output is same, but computational complexity decreases in this method.

Question 7 (a)

```
mask = np.zeros(daisy.shape[:2],dtype=daisy.dtype)
rectangle= (40,40,540,800)
background_mask = np.zeros((1, 65), np.float64)
foreground_mask = np.zeros((1, 65), np.float64)

cv.grabCut(daisy, mask, rectangle, background_mask, foreground_mask, 15, cv.GC_INIT_WITH_RECT)

grab_cut_mask = np.where((mask == 2) | (mask == 0), 0, 1).astype('uint8')
daisy_fg = daisy*grab_cut_mask[:,:,np.newaxis]
daisy_bg=np.subtract(daisy,daisy_fg)
```



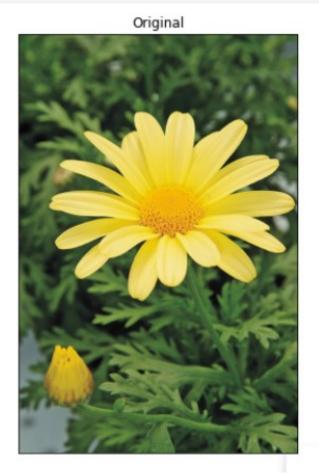






Question 7 (b)

```
k_size = 9
sigma = 4
blurred_bg =cv.GaussianBlur(daisy_bg,(k_size,k_size),sigma)
enhanced_daisy=np.add(daisy_fg,blurred_bg)
```





Question 7 (c)

When Gaussian blur is applied to the background image, the green and black pixels near the edge of the flower get mixed.