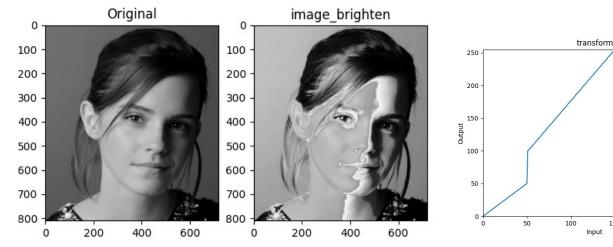
EN3160 - Assignment 1 on Intensity Transformations and Neighborhood Filtering

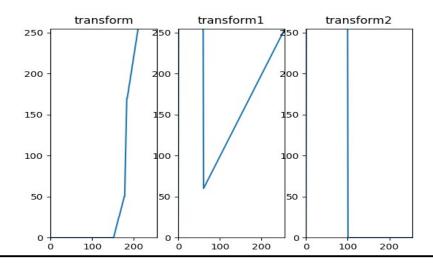
200310E – Tharaka Kodithuwakku

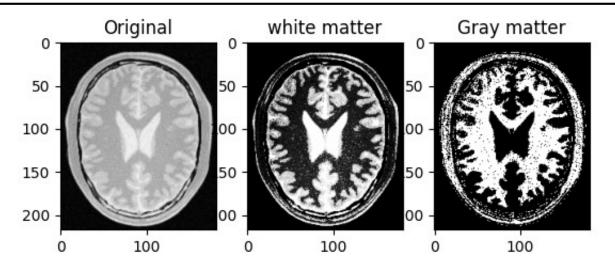
01) Implement the intensity transformation



02) Accentuate white matter gray matter in the brain proton density image

```
image_transformed = cv.LUT(img_original,transform)
image_transformed21 = cv.LUT(img_original,transform1)
image_transformed22 = cv.LUT(image_transformed21,transform)
image_transformed2 = cv.LUT(image_transformed22,transform2)
```



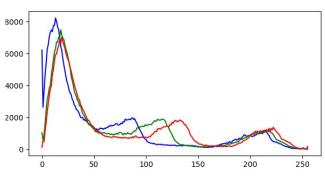


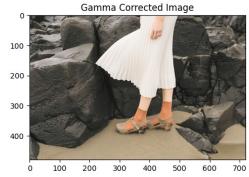
03) Apply gamma correction to the L plane in the L*a*b* color space

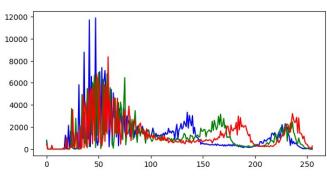
```
# Gamma correction factor
gamma = 0.51
# Convert BGR image to Lab color space
img_lab = cv.cvtColor(img_org, cv.COLOR_BGR2Lab)
# Split the Lab image into L, a, and b channels
L, a, b = cv.split(img_lab)
# Calculate the lookup table for gamma correction
table = np.array([(i / 255.0) *** (gamma) * 255.0 for i in np.arange(0, 256)]).astype('uint8')
# Apply gamma correction to the L channel using the lookup table
L_corrected = cv.LUT(L, table)
```

```
color = ('b','g','r')
for i, c in enumerate(color):
    hist_orig = cv.calcHist([img_org], [i], None, [256], [0, 256])
    axarr[1, 0].plot(hist_orig, color=c)
    hist_gamma = cv.calcHist([img_gamma_corrected], [i], None, [256], [0, 256])
    axarr[1, 1].plot(hist_gamma, color=c)
```



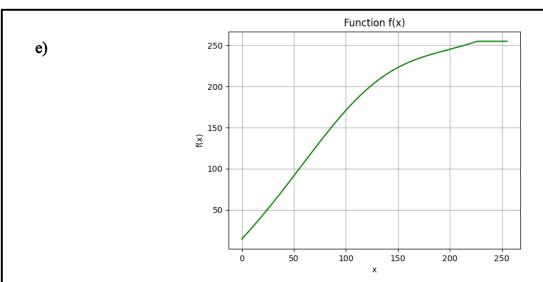




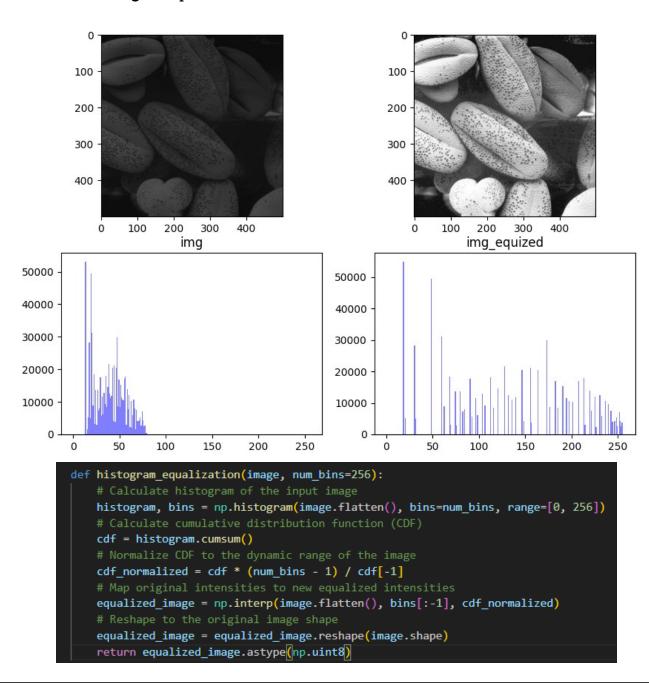


04) Increasing the vibrance of a photograph img_n = cv.cvtColor(img,cv.COLOR_BGR2RGB) a) img hsv = cv.cvtColor(img n, cv.COLOR BGR2HSV) img hsv n = cv.cvtColor(img hsv,cv.COLOR BGR2RGB) # Split the HSV image into H, S, and V channels H, S, V = cv.split(img_hsv) # Apply the saturation adjustment formula **b**) a = 0.6 # Adjust this parameter as needed sigma = 70 # Adjust this parameter as needed S_adjustment = a * 128 * np.exp(-(S - 128) ** 2 / (2 * sigma ** 2)) ## s + s_ad; S_corrected = np.minimum(np.maximum(S + S_adjustment, 0), 255).astype('uint8') d) # Merge the corrected S channel with the original H and V channels img_hsv_corrected = cv.merge((H, S_corrected, V)) Original Image Vibrance-Enhanced Image e) Original S Channel Histogram Corrected S Channel Histogram

As the parameter "a" gets closer to 1, it enhances the vibrancy or saturation of the image. When "a" is set to 0.6, it strikes a balance that maintains a more natural appearance without excessive vibrancy. Adjusting the "a" value allows you to control the level of saturation enhancement in the image, with lower values producing a more subdued effect and higher values making the colors appear more vibrant.



05) A function of histogram equalization



06) Apply histogram equalization only to the foreground of an image to produce a image with a histogram equalized foreground

```
a) # Split the HSV image into H, S, and V components
H, S, V = cv.split(hsv_img)
```

```
b) threshold_value = 13 # determines the point at which the pixels will be classif
foreground_mask = cv.threshold(S, threshold_value, 255, cv.THRESH_BINARY)[1]
```

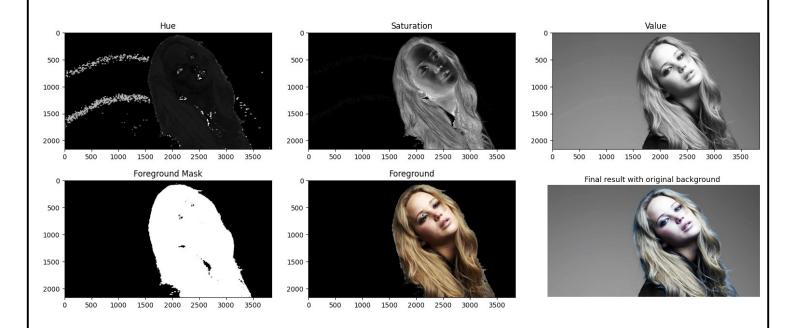
```
B, G, R = cv.split(img)

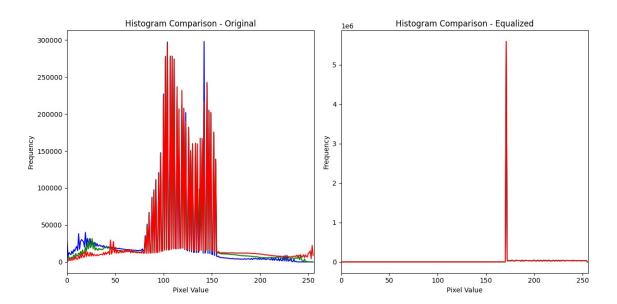
# Extract the foreground for each color channel
foreground_B = cv.bitwise_and(B, B, mask=foreground_mask)
foreground_G = cv.bitwise_and(G, G, mask=foreground_mask)
foreground_R = cv.bitwise_and(R, R, mask=foreground_mask)

# Merge the foreground channels back into a single image
foreground = cv.merge([foreground_R, foreground_G, foreground_B])
```

```
d) cumulative_sum = np.cumsum(hist_foreground)
40 print(cumulative_sum)
```

```
background_mask_3d = 255 - foreground_mask
background_hsv = np.bitwise_and(hsv_img, background_mask_3d) # Extract
background_rgb = cv.cvtColor(background_hsv, cv.COLOR_HSV2RGB)
final_image = background_rgb + equalized_result # Add with foreground_rgb
```





07) Filtering with the Sobel operator

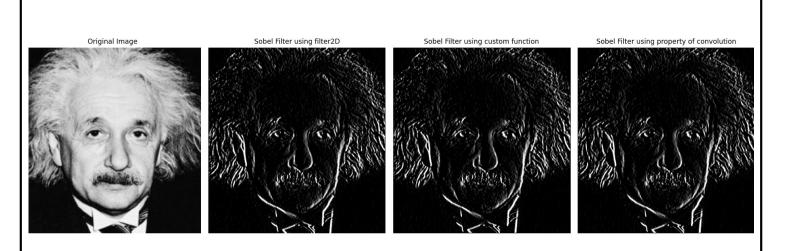
```
def filter(image , kernel):
    assert kernel.shape[0]%2 == 1 and kernel.shape[1]%2 == 1
    k_hh, k_hw = kernel.shape[0] // 2, kernel.shape[1] // 2
    h, w = image.shape
    image_float = cv.normalize(image.astype('float'), None, 0, 1, cv.NORM_MINMAX)
    result = np.zeros(image.shape, 'float')

for m in range(k_hh, h - k_hh):
    for n in range(k_hw, w - k_hw):
        result[m, n] = np.dot(image_float[m-k_hh: m+k_hh+1, n-k_hw: n+k_hw+1].flatten(), kernel.flatten())

result = result * 255  # Undo normalization
    result = np.minimum(255, np.maximum(0, result)).astype(np.uint8) # Limit between 0 and 255
    return result
```

```
def filter_in_steps(image, kernel1, kernel2):
    image_float = cv.normalize(image.astype('float'), None, 0, 1, cv.NORM_MINMAX)
    result = filter_step(filter_step(image_float, kernel1), kernel2)
    result = result * 255
    result = np.minimum(255, np.maximum(0, result)).astype(np.uint8) # Limit betwee
    return result
```

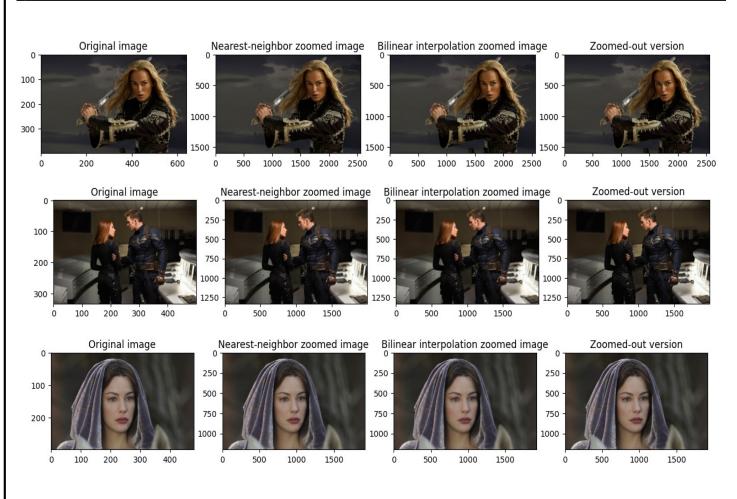
```
kernel = np.array([(-1,-2,-1),(0,0,0),(1,2,1)],dtype='float')
imgc = cv.filter2D(img,-1,kernel)
```



08) A program to zoom images by a given factor $s \in (0,10)$

```
def images_set():
    for j in range(4):
        image = cv.imread(original_images[j])
        image_zoom_out = cv.imread(zoom_outs[j])

    image_bilinear = cv.resize(image, None, fx=4, fy=4, interpolation=cv.INTER_LINEAR)
        image_near = cv.resize(image, None, fx=4, fy=4, interpolation=cv.INTER_NEAREST)
```



09) A flower image problem

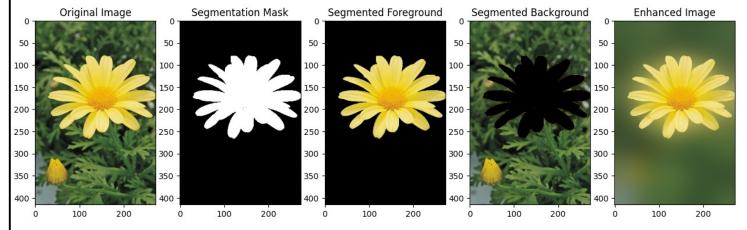
```
# (a) Perform image segmentation
segmentation_mask = np.zeros(original_image.shape[:2], dtype=np.uint8)

# Define a region of interest (ROI) within the image
roi_rect = (30, 30, original_image.shape[1] - 30, original_image.shape[0] - 150)
background_model = np.zeros((1, 65), dtype=np.float64)
foreground_model = np.zeros((1, 65), dtype=np.float64)
cv.grabCut(original_image, segmentation_mask, roi_rect, background_model, foreground_model, 5, cv.GC_INIT_WITH_RECT)

# Create a binary mask where 1 represents the foreground and 0 represents the background
binary_mask = np.where((segmentation_mask == 2) | (segmentation_mask == 0), 0, 1).astype('uint8')

# Apply the mask to the original image to extract the segmented foreground
segmented_foreground = original_image * binary_mask[:, :, np.newaxis]
```

```
# (b) Enhance the image
blurred_background = cv.GaussianBlur(original_image, (0, 0), 30)
enhanced_image = np.where(binary_mask[:, :, np.newaxis] == 1, original_image, blurred_background)
|
```



The observed dark edges in the enhanced image are a result of the blending process during Gaussian blur, where the black pixels within the mask boundary are combined with the background pixels along the mask's border. This blending can lead to an undesirable visual effect, causing the edges to appear darker than intended.

Github Link - https://github.com/baymax06in19/EN-3160