## Assignment 1

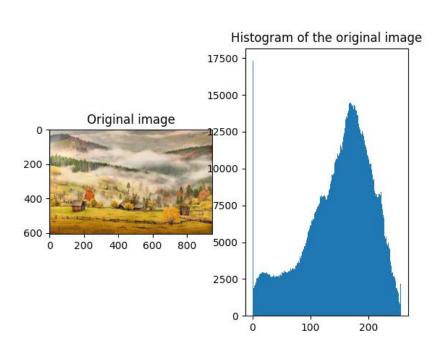
### Subsection 1:

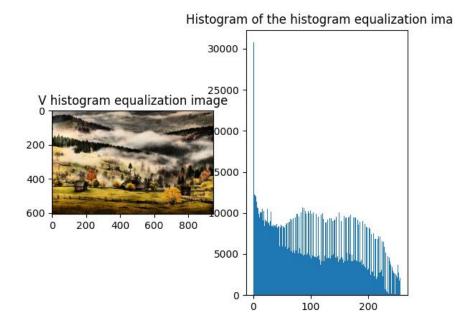
```
import numpy as np
import matplotlib.pyplot as plt
cv2.imread(r"C:\Users\Vladuts\Desktop\IPIVA\TEMA1\Background images\landscape
imgHSV = cv2.cvtColor(img, cv2.COLOR BGR2HSV)
plt.imshow(img[:, :, ::-1])
plt.show()
plt.subplot(1, 2, 1); plt.imshow(img[:, :, ::-1]); plt.title("Original
plt.subplot(1, 2, 2); plt.hist(img.ravel(), 256, [0, 256]);
plt.title("Histogram of the original image")
H, S, V = cv2.split(imgHSV)
plt.imshow(imgOutHE[:, :, ::-1])
plt.show()
plt.subplot(1, 2, 1); plt.imshow(imgOutHE[:, :, ::-1]); plt.title("\forall
```

```
plt.show()
#Contrast Limited Adaptive HE:
imgHSV_copy = np.copy(imgHSV)
clahe = cv2.createCLAHE(clipLimit=2, tileGridSize=(8, 8))
imgHSV_copy[:, :, 2] = clahe.apply(imgHSV_copy[:, :, 2])
img_CLAHE = cv2.cvtColor(imgHSV_copy, cv2.CoLOR_HSV2BGR)
# display the Contrast Limited Adaptive HE image:

plt.imshow(img_CLAHE[:, :, ::-1])
plt.show()
# display the Contrast Limited Adaptive HE image histogram:

plt.subplot(1, 2, 1); plt.imshow(img_CLAHE[:, :, ::-1]); plt.title("Contrast Limited Adaptive HE image")
plt.subplot(1, 2, 2); plt.hist(img_CLAHE.ravel(), 256, [0, 256]);
plt.title("Histogram of the Contrast Limited Adaptive HE image")
plt.show()
cv2.imwrite('back_eq.jpg', img_CLAHE)
#As we can see the CLAHE method gives as a softer contrast effect, the image is more clear.
#Also the histogram of the CLAHE image is more uniform then the orginal and HE image.
```







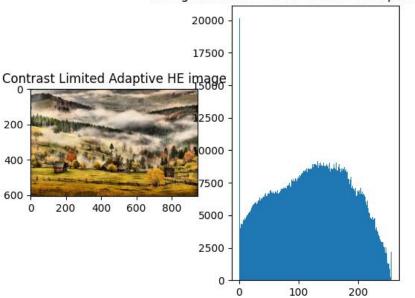




Figure 1 - back\_eq

Explanation: After analyzing the histograms, we can see that the CLAHE equalized image gives us the most uniform distribution. We want this so that the "luminance" effect is pleasant and each pixel is brightened in the necessary way.

For the histogram equalization equalized image we applied the equalization on the V channel of the image because the V channel represents the brightness of each pixel. The HE yields a more brutal contras effect and the resulted histogram is not so uniform. In this manner we are losing details in the darker areas, like in the distant trees.

## Subsection 2:

# Assignment 1 Tudorache Vlad Adrian 442C

```
#(607/2) - (352/2):(607/2) + (352/2), (950/2) - (280/2):(950/2) + (280/2)]
img_cropped = back_eq[128:480, 335:615]
print(img_cropped.shape)
plt.imshow(img_cropped[:, :, ::-1])
plt.show()
cv2.imwrite('back_eq_crop.jpg', img_cropped)
```

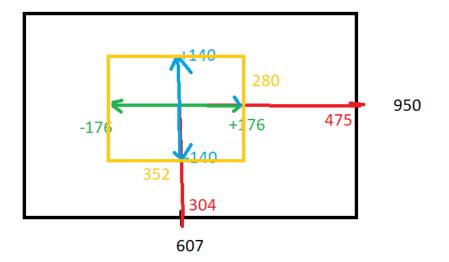




Figure 2 - back\_eq\_crop

### Subsection 3:

Explanation: After analyzing the noisy image, I determined that noise is the salt-and-pepper type. As discussed in the second laboratory, the best way to tackle this noise is by using a median blur. After multiple tries, the best results came from a kernel size equal to 5.

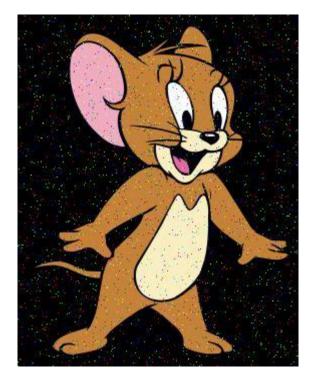


Figure 3 - im\_4\_noisy

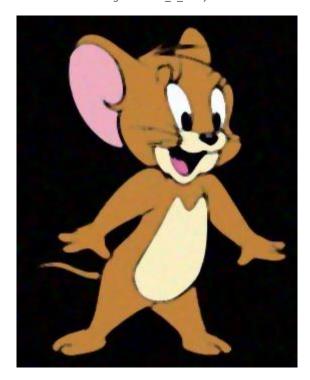


Figure 4 - im\_4\_filt

### Subsection 4:

```
SUBSECTION 4 ----
x = list(range(0, 256))
y init = [0, 50, 100, 150, 200, 250]
#interpolate this values:
interpol = np.interp(x, y_init, y_final)
plt.figure()
plt.plot(x, interpol,'-r', x init,y init,'--k'), plt.grid();plt.title("Tone 3
plt.show()
table = np.array([round(i) for i in interpol])
B_r G_r R = cv2.split(im 4 filt)
im 4 tone = cv2.merge((im 4 tone B, im 4 tone G, im 4 tone R))
plt.imshow(im 4 tone[:, :, ::-1])
plt.show()
```

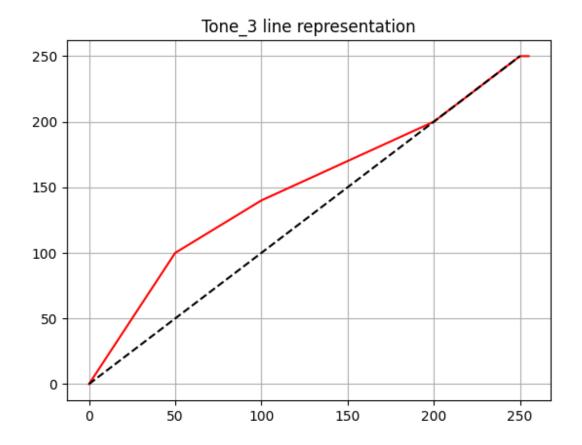




Figure 5 - im\_4\_tone

### Subsection 5:

```
SUBSECTION 5 ----
BKG = img cropped.copy()
maskedROI = cv2.multiply(roi,(1- jerrMask))
maskedJerry = cv2.multiply(jerrBGR, jerrMask)
BKG_with_Jerry = cv2.add(maskedROI, maskedJerry)
big image = img CLAHE.copy()
big image[128:480, 335:615] = BKG with Jerry
plt.imshow(big image[:, :, ::-1])
plt.show()
cv2.imwrite('im 4 final.jpg', big image)
```



Figure 6- im\_4\_final