
COMP 6771 Image Processing: Assignment 1

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1. Question 1

- (a) Explain why the discrete histogram equalization technique does not, in general, yield a flat histogram. If we want a flat histogram, we need to guarantee each component has the same pixels in

it. Let's set n be the total number of pixels, and L be the intensity levels. The flat histogram will require each component has n/L number of pixels. However, the Histogram Equalization process will re-mapping the pixels only based on their intensity level. So, we can not control the number of pixels in each component, it is difficult to yield a flat histogram.

- (b) Suppose that a digital image is subjected to histogram equalization. Please elaborate why a second pass of histogram equalization (on the histogram-equalized image) will produce exactly the same result as the first pass. Suppose an image has already pass a histogram equalization. Input this

image into the second second pass of histogram equalization. Let n is the total number of pixels, and L is the intensity level.

$$s_k = T(r_k) = (L - 1) \sum_{j=0}^k n_{rj} / n = \frac{L - 1}{n} \sum_{j=0}^k n_{rj} = (L - 1) \sum_{j=0}^k p_r(rj)$$

we will use the equation mentioned above to calculate the s_k . Since every pixel belong to r_k is distributed to s_k . So, the $n_{sk} = n_{rk}$

$$p_s(s_k) = n_{sk} / n = n_{rk} / n = p_r(r_k)$$

Also, no matter how many iterations,

$$new_k = (L - 1) \sum_{j=0}^k p_s(s_k) = (L - 1) \sum_{j=0}^k p_r(r_k) = s_k$$

Therefore, the second pass will get the same result as the first pass.

2. Given two arbitrary images $f(x,y)$ and $g(x,y)$ and two arbitrary constant a and b , H is said to be a linear operator if:

$$H[af(x, y) + bg(x, y)] = aH[f(x, y) + bH[g(x, y)]]$$

The median m of a set of numbers is such that half the values in the set are below m and the other half are above it (the mid-point value by population). Is an operator that computes the median of a set of pixels of a sub-image area linear or nonlinear? Explain your answer by giving examples.

In general, the median operator is not linear operator. Since we can not control the order of values in $f(x, y)$ or $g(x, y)$. We will re-order the values in $f(x, y)$ and $g(x, y)$ and the the result of $af(x, y) + bg(x, y)$ before we get the median value.

In some specific case, median operator will be linear, for example, if the the median value located in the center of $f(x, y)$ and $g(x, y)$ and the value in the left are smaller or equal then the median and the values in the right are greater or equal than the median in both $f(x, y)$ and $g(x, y)$, and after $af(x, y) + bg(x, y)$ the result still folloing the rules mentioned above, H will be a linear operator.

For example, suppose $f(x, y) = [1, -3, 5, 7, 9]$ and $g(x, y) = [2, 4, 6, 8, 10]$ and $a = 2, b = 3$.

$$\begin{aligned} H[af(x, y) + bg(x, y)] &= H[[2, -6, 10, 14, 18] + [6, 12, 18, 24, 30]] \\ &= H[8, 6, 28, 38, 48] \\ &= 28 \end{aligned} \quad (1)$$

$$\begin{aligned} aH[f(x, y)] + bH[g(x, y)] &= 2 * H[5] + 3 * H[6] \\ &= 10 + 18 \\ &= 28 \end{aligned} \quad (2)$$

Therefore, we can find that the Eq. 1 = Eq. 2.

On the other hand, If the values in $f(x, y)$ and $g(x, y)$ not following the rules mentioned about, the H is not a linear operator.

For example, Let us set $f(x, y) = [1, 3, -5, 7, 9]$ and $g(x, y) = [2, 4, 6, 8, 10]$, also, $a = 2, b = 3$.

$$\begin{aligned} H[af(x, y) + bg(x, y)] &= H[[2, 6, -10, 14, 18] + [6, 12, 18, 24, 30]] \\ &= H[8, 18, 8, 38, 48] \\ &= 18 \end{aligned} \quad (3)$$

$$\begin{aligned} aH[f(x, y)] + bH[g(x, y)] &= 2 * H[3] + 3 * H[6] \\ &= 6 + 18 \\ &= 24 \end{aligned} \quad (4)$$

Therefore, the Eq. 3 \neq Eq. 4.

3. The purpose of this question is to perform histogram equalization to a given histogram and plot the resulting histogram. Given the following histogram where GL is Gray level, and NP is Number of pixels:

GL	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
NP	0	5	13	57	100	39	21	12	7	2	0	0	0	0	0	0

- (a) Plot the histogram of the image given in the table above

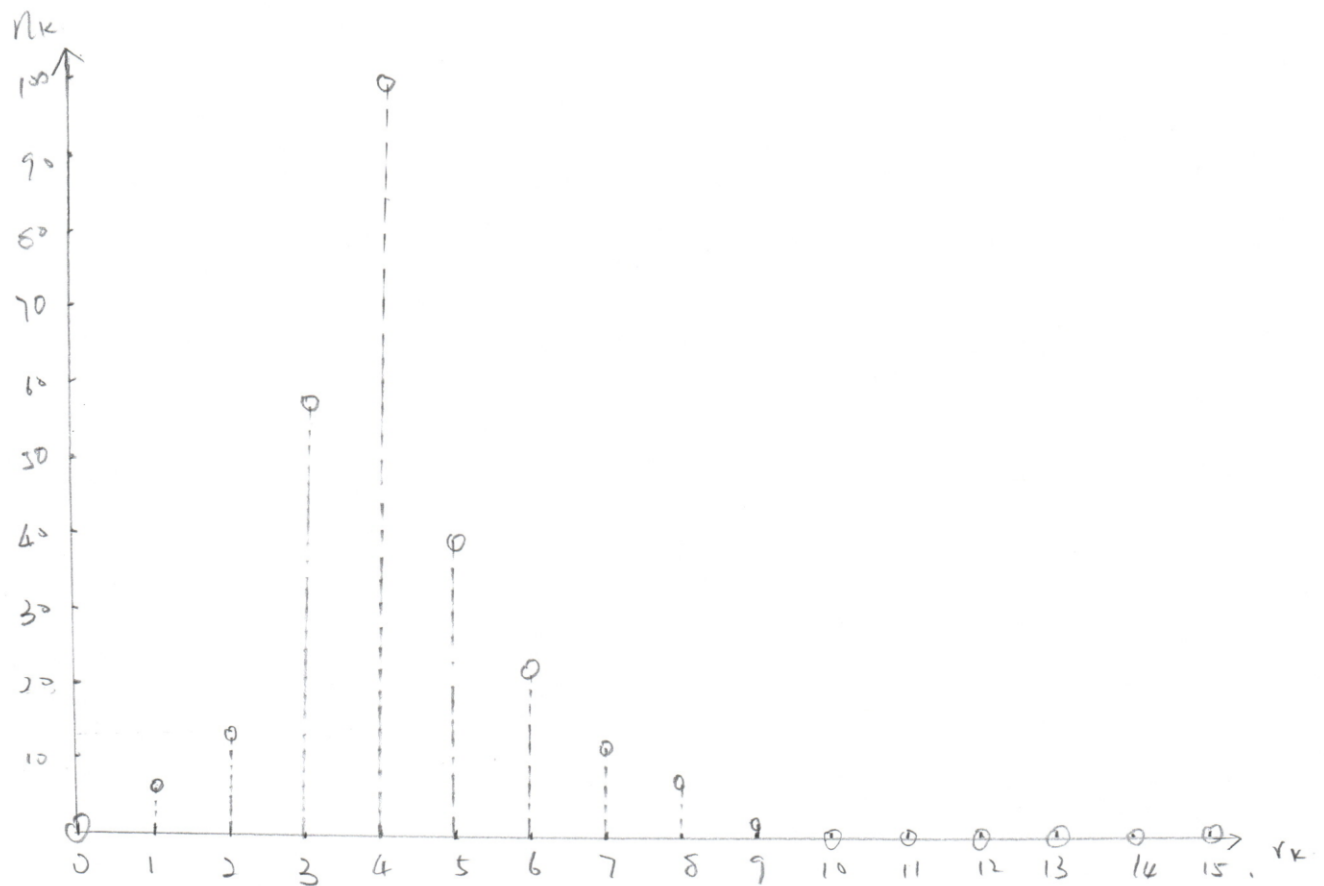


Figure 1: Histogram image

The Fig. 1 presents the histogram based on the given table.

- i. Calculate s_k from the table

GL	NP	$p_r(r_k)$	s_k	Round
0	0	0	0	0
1	5	0.0195	0.2929	0
2	13	0.0507	1.0546	1
3	57	0.2226	4.3945	4
4	100	0.3906	10.253	10
5	39	0.1523	12.539	13
6	21	0.0820	13.769	14
7	12	0.0468	14.472	14
8	7	0.0273	14.882	15
9	2	0.0078	15	15
10	0	0	15	15
11	0	0	15	15
12	0	0	15	15
13	0	0	15	15
14	0	0	15	15
15	0	0	15	15

Figure 2: Calculation result

The Fig. 2 shows the calculation result of s_k . In This Figure, Let GL be the Gray Level r_k , and NP is the number of pixels with value r_k , we will use n_k later. Based on the given table, the total pixel $n = 256$, and $L = 15$

$$p_r(r_k) = \frac{n_k}{n} \quad (5)$$

$$s_k = \frac{L-1}{n} \sum_{j=0}^k n_j \quad (6)$$

And in the Fig. 2, we utilize the Eq. 5 and Eq. 6 to calculation the $p_r(r_k)$ and s_k respectively.

(b) Plot the probability density functions $p_r(r_k)$ and $p_s(s_k)$

$GL(s_k)$	Mapped $NP(n_{sk})$	$p_s(s_k)$
0	5	0.0195
1	13	0.0507
2	0	0
3	0	0
4	57	0.2226
5	0	0
6	0	0
7	0	0
8	0	0
9	0	0
10	100	0.3906
11	0	0
12	0	0
13	39	0.1523
14	33	0.1289
15	9	0.0351

Figure 3: Calculation of $p_s(s_k)$

We following the Histogram Equalization process to calculate those values in Fig. 3.

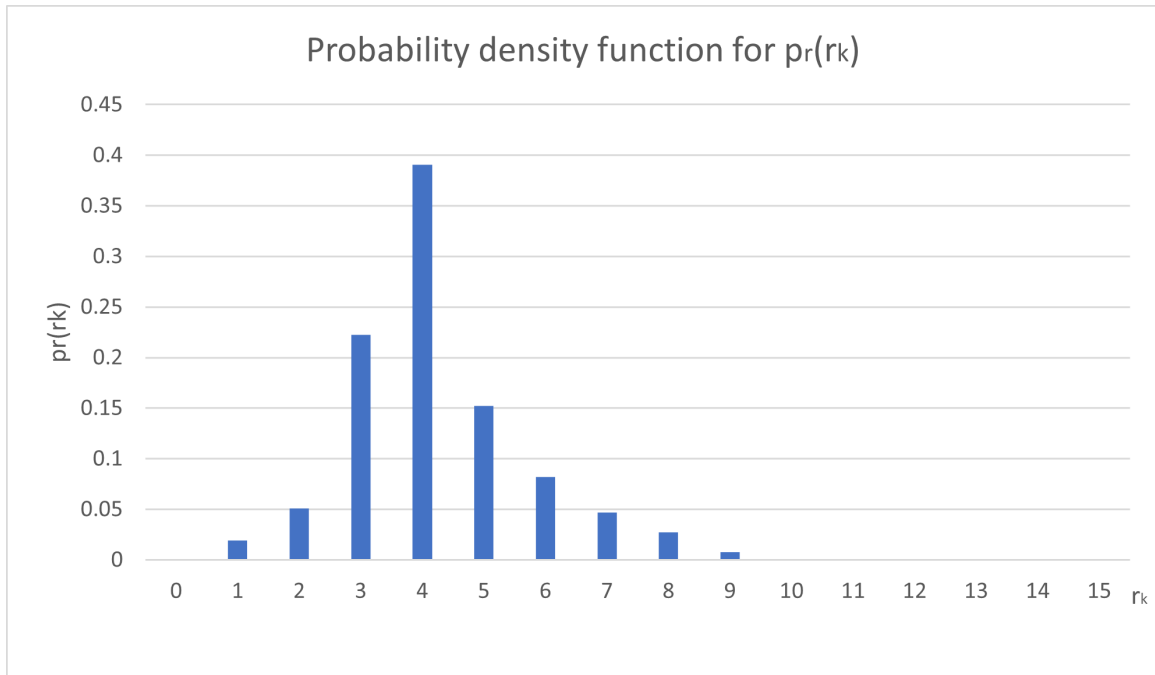


Figure 4: Probability densi function for $pr(r_k)$

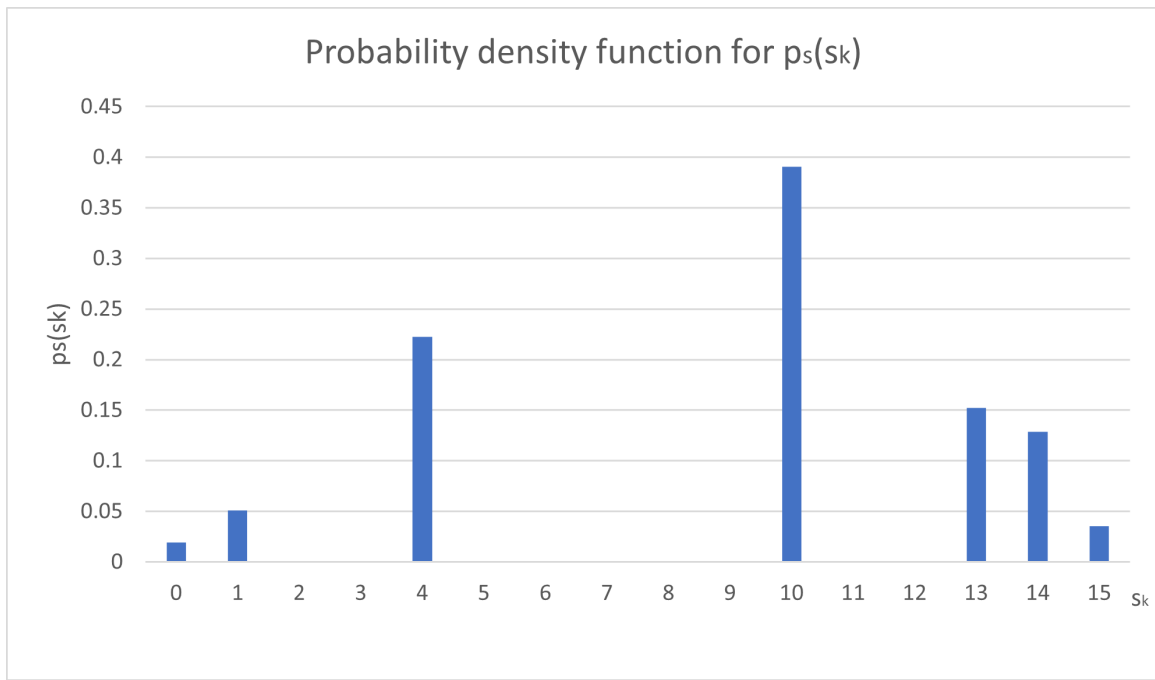


Figure 5: Probability density function for $p_s(s_k)$

Finally, we will get two probability density function present in Fig. 4 and Fig. 5.

(c) Plot the new histogram after performing the histogram equalization.

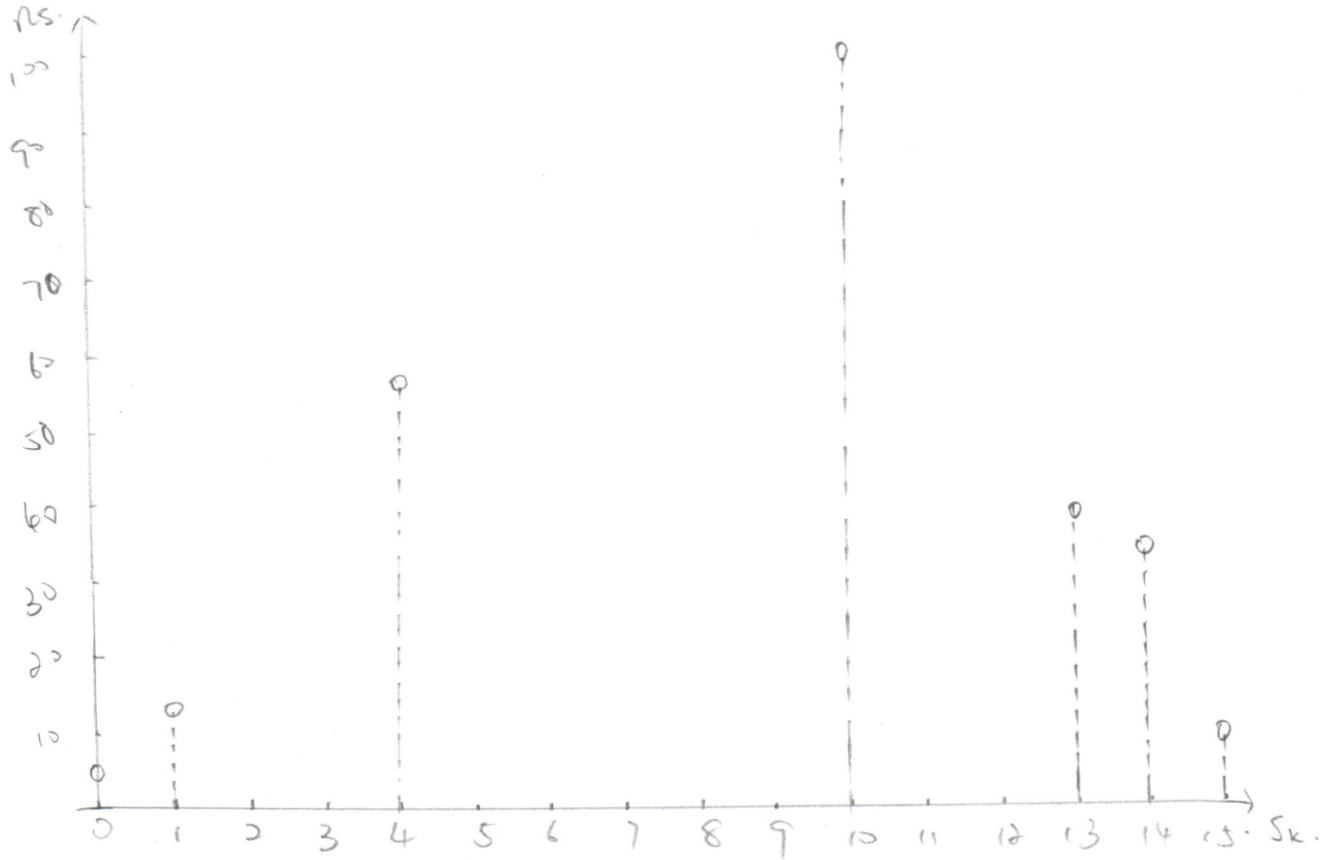


Figure 6: New histogram after histogram equalization

The new histogram is presented in Fig. 6.

4. The Given two images $f(x,y)$ and $g(x,y)$ with histogram h_f and h_g . Assuming that all the pixels of image $g(x,y)$ have the same intensity equal to a non-zero positive constant value c . The gray levels of the pixels of both images have positive values. Please explain the relationship between the histograms of $f(x,y)$ and the new images formed by $f(x,y) + g(x,y)$ and $f(x,y)*g(x,y)$
 - (a) Since all the pixels of image $g(x,y)$ have the same intensity equal to a non-zero positive constant value c . Set, $n(r_k)$ be the number of pixels belong to the intensity level r_k , then, the $f(x,y) + g(x,y) = f(x,y) + c$, So, the $n(r_k)$ will not change, but $r'_k = r_k + c$. So, the histogram of new image $f(x,y) + g(x,y)$ will have the same hight as the histogram of $f(x,y)$ but all components will move to right with c .
 - (b) As mentioned above, $f(x,y) * g(x,y) = f(x,y) * c$, So the hight will not change after the product operator, but the distance between each component will be enlarged by c , as $r'_k = r_k * c$. For example, $r_k = 1$ will go to $r_k = c$ and $r_k = 2$ will go to $r_k = 2c$, $r_k = 3$ will go to $r_k = 3c$...
5. Download the image from the assignment folder, and perform the following operations using MATLAB or any software packages you are familiar with. Please show your steps in the report.


```
1 import numpy as np
2 import cv2
3 from matplotlib import pyplot as plt
```

(a) Write a program to read the grayscale of the image.

```
1 def read_grayscale(img):
2     img = cv2.imread(img,0)
3     return img
4 img_name = 'HawkesBay.jpeg'
5 img = read_grayscale(img_name)
```

(b) Write a program to calculate the histogram of the image and display the histogram chart.

```
1 def calculate_histogram(img, L):
2     image_flatten = img.flatten()
3     new_list = np.zeros(L)
4     for each_pixel in image_flatten:
5         new_list[each_pixel] += 1
6     return new_list
7
8 def histogram_show(hist):
9     plt.bar(range(len(hist)), hist)
10    plt.xlabel("rk")
11    plt.ylabel("nk")
12    plt.title("Histogram of input image")
13    plt.savefig("Histogram_input")
14
15 histogram_statis = calculate_histogram(img, 256)
16 histogram_show(histogram_statis)
```

The histogram is presented in Fig.7.

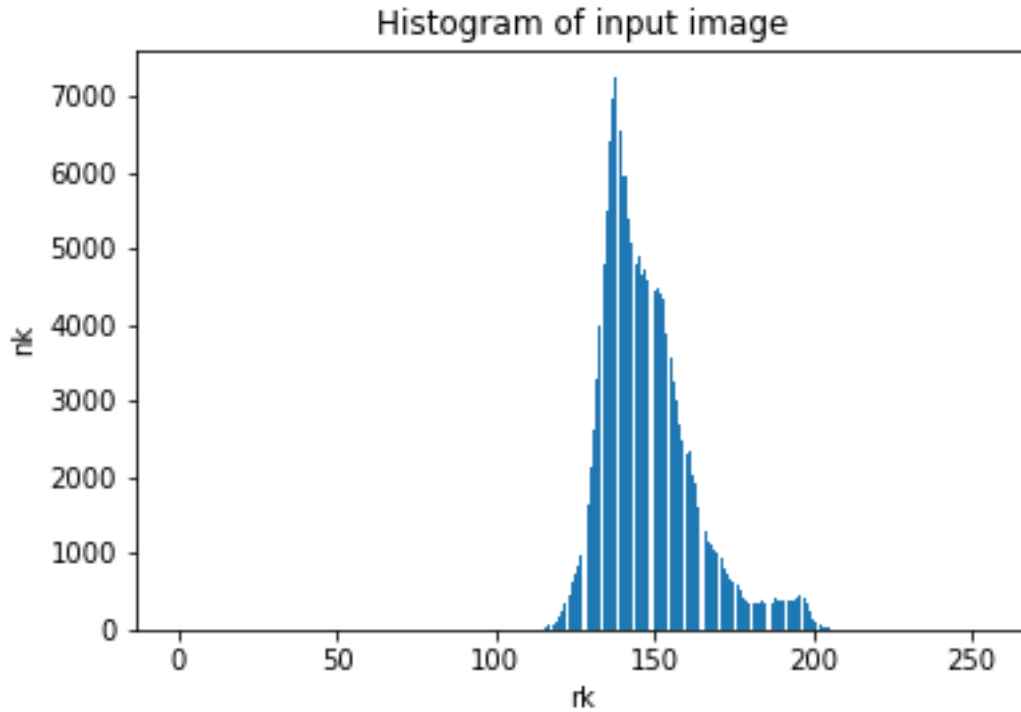


Figure 7: Histogram of input image

- (c) Compare the calculated histogram obtained by using your own program with the one using the `imhist` function in MATLAB.

```
1 img = imread("HawkesBay.jpeg")
2 imhist(img,256)
```

The histogram from Matlab `imhist` function is present as Fig. 8

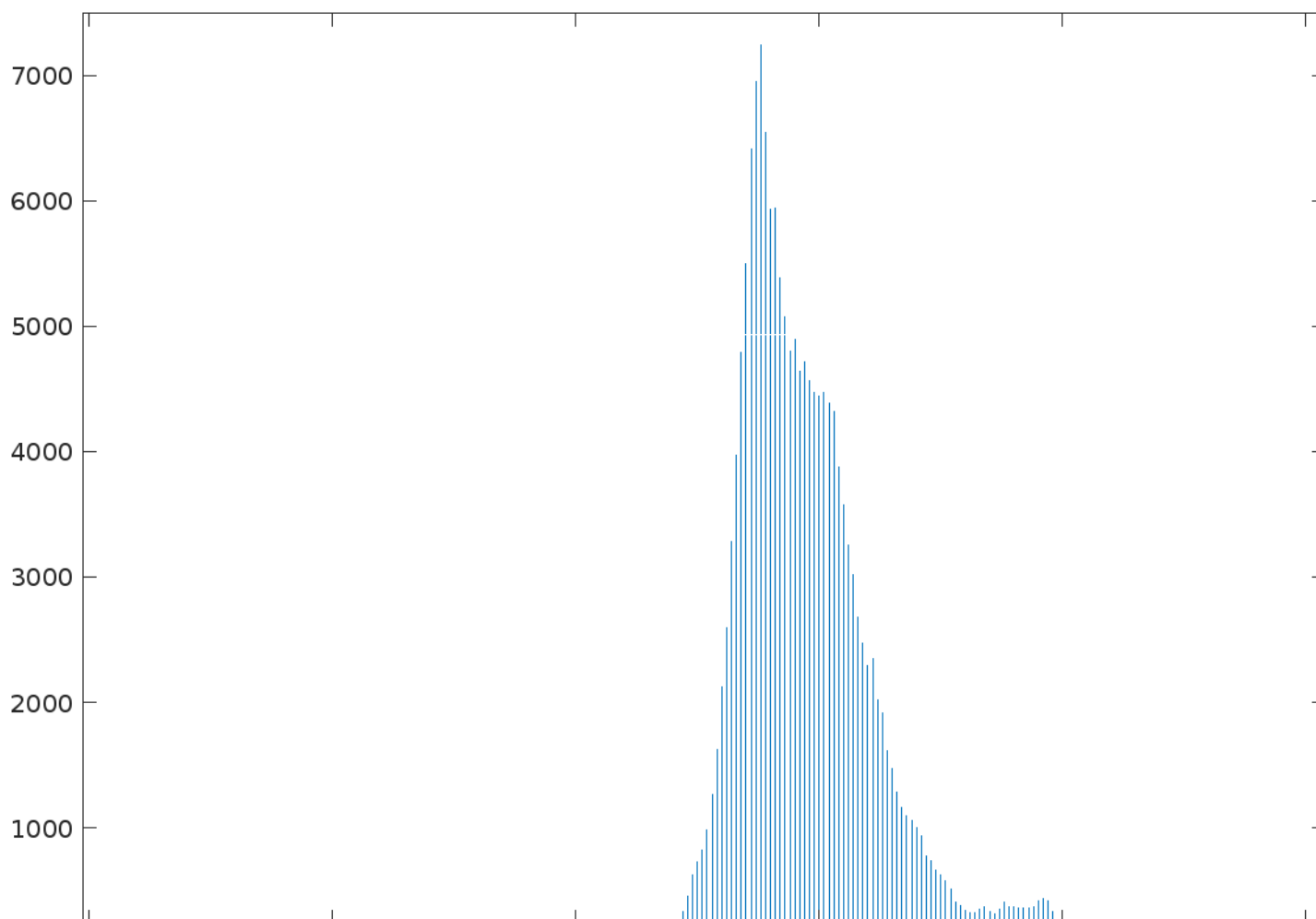


Figure 8: Histogram of Matlab imhist function

(d) Write a program to do histogram equalization on this image.

```

1  def calculate_prob(hist , total_pixels):
2      return [values_each_bin/total_pixels for values_each_bin in hist]
3
4  def calculate_sk(cal_pr , L):
5      new_list = np.zeros(L)
6      for index , each_pr in enumerate(cal_pr):
7          cal_result = (L-1) * sum(cal_pr[0: index+1])
8          new_list[index] = round(cal_result)
9      return new_list
10
11 def new_image(img , sk):
12     height , width = img.shape
13     new_img = np.zeros(img.shape , dtype = 'uint8 ')
14     for i in range(img.shape[0]):
15         for j in range(img.shape[1]):

```

```

16         new_img[i, j] = sk[img[i, j]]
17     return new_img
18 tp = img.shape[0]*img.shape[1]
19 pr = calculate_prob(histogram_statis, tp)
20 sk = calculate_sk(pr, 256)
21 new_img = new_image(img, sk)

```

- (e) Compare the histogram-equalized image obtained by using your own program with the one by using histeq function in MATLAB.

```

1 J = histeq(I,256)
2 imshow(J)

```

The new image from matlab histeq function is presented in Fig. 9



Figure 9: Histogram-equalized image obtained by Matlab histeq

```

1 cv2.imwrite("he_newimage.png", new_img)

```

The new image from the histogram equalization problem is shown in Fig. 10



Figure 10: "histogram-equalized image obtained by my own program"

This two histogram-equalized images are very similar.