

## I. HISTOGRAM EQUALIZATION

In this report, I will first explain what is Histogram Equalization; Then give the implement in Python; Finally discuss the relationship between Histogram Equalization and Histogram Matching.

### A. What is Histogram Equalization

Histogram equalization is a method in image processing of contrast adjustment using the image's histogram. Since digital images are represent by pixel array, for a grayscale image it can be described by a 2D array, each pixel's value range from 0 to 255, the brightness will grow over the value increase. For an unprocessed Image, If we count the each value's occurrence and make a histogram, it would be like the left picture in Fig.1

Our goal is to let the occurrence of each value distributed more equally on the axis 0-255, like the right picture in Fig.1; For the intuition this will make the image shows more details in where used to be too dark or bright.

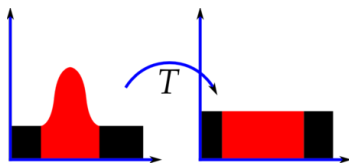


Fig. 1. shows the basic concept of Histogram Equalization

### B. Implement in Python

The shortcut is to use the Opencv package, which include the Function `equalizeHist()`. To use this function just input your source image then it will return the processed image:

```
result = cv2.equalizeHist(source).
```

Also, there is an easy way to implement it step by step using Numpy:

First, we flatten the image array and count the occurrence of each value using Numpy `bincount()`:

```
occurrence=np.bincount(flatten,minlength=256)
```

Second, calculate the PDF and CDF of the pixels, getting a projection list `sk[]`:

```
nk=occurrence/(MN), MN is the total size of pixels.
```

```
sk[k]=255*nk[0:k].sum(), calculate CDF and get the sk[].
```

Finally, apply the projection to original image:

```
img[img==pix]=sk[pix]
```

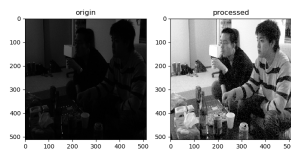


Fig. 2. shows left is original image, right is processed

### C. relationship with Histogram Matching

In short, Histogram equalization is a special case of Histogram Matching; In Histogram equalization we supposed the after processed histogram is uniformly distributed. But in Histogram Matching, first we need another image as a target distribution, second we need to find the distribution projection between the original and target histogram; finally we can apply this projection on the original image.