

# Image processing concepts

Part 2

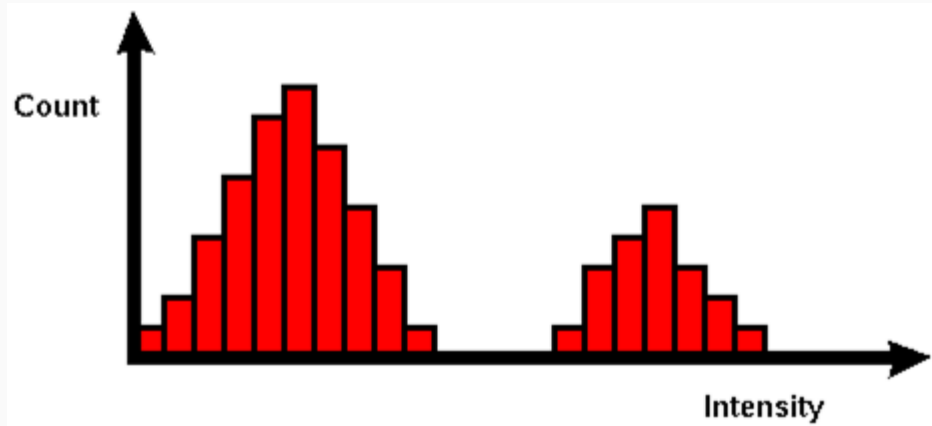


# Histogram

It is a graph showing the number of pixels in an image at each different intensity value found in that image, in other words it shows the **predominant intensities** of an image.

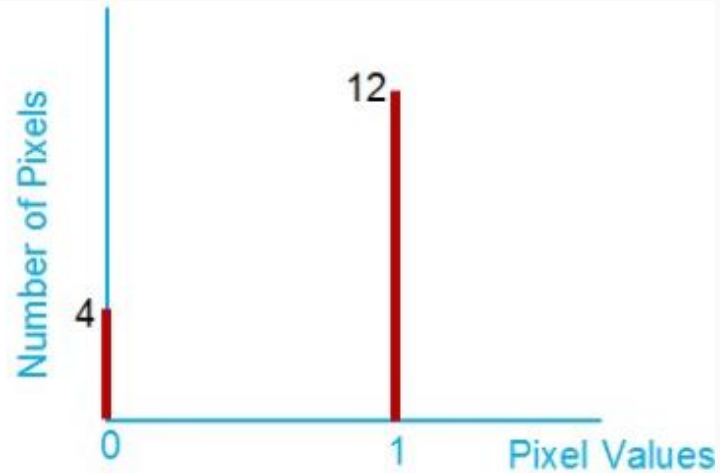
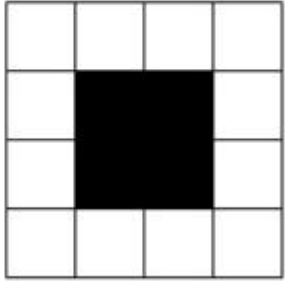
**x-axis** is the **intensity value** from 0 to 255

**y-axis** is the **number of pixels** with that intensity value, these values varies depending on the **number of the pixels** in the image and **how their intensities are distributed**



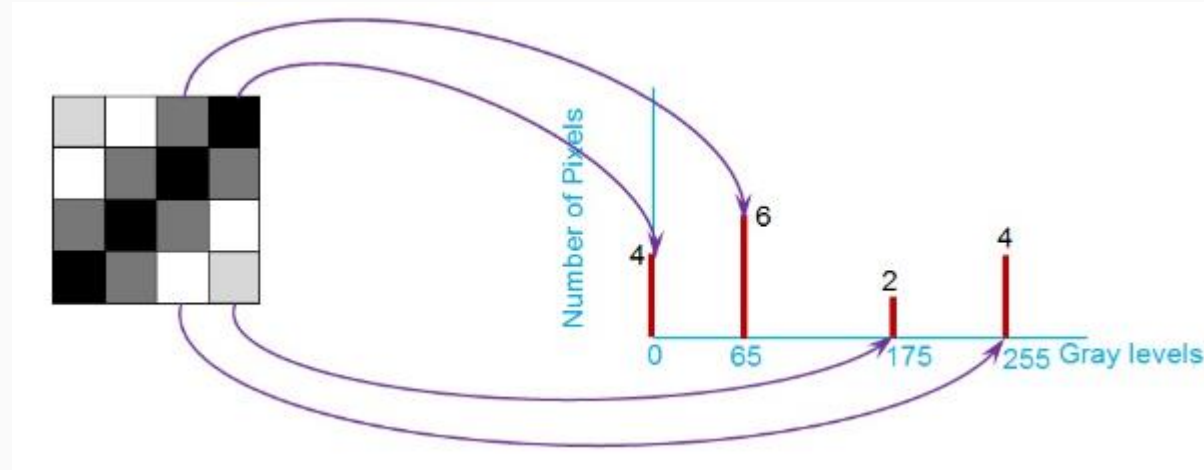
# Example 1

Histogram of simple  $4 \times 4$  **black-and-white** image



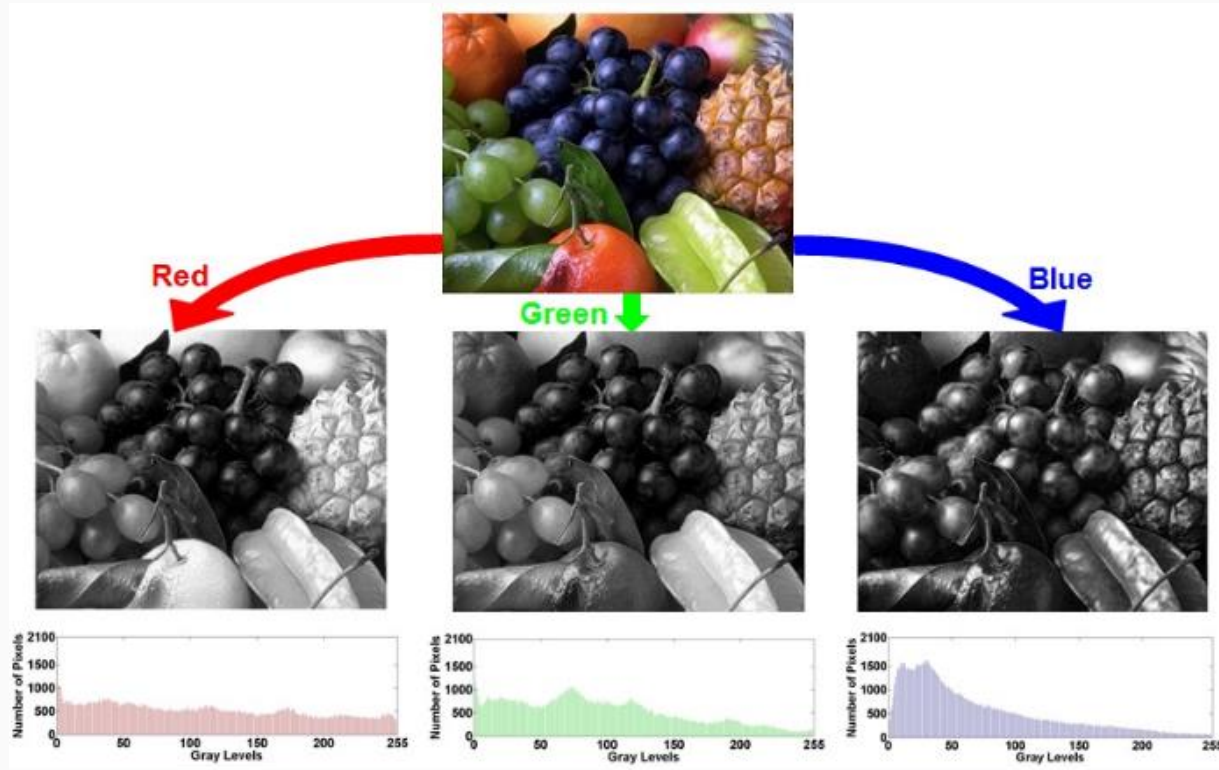
## Example 2

Histogram of **Grayscale** image



# Example 3

Histogram of a **color** image

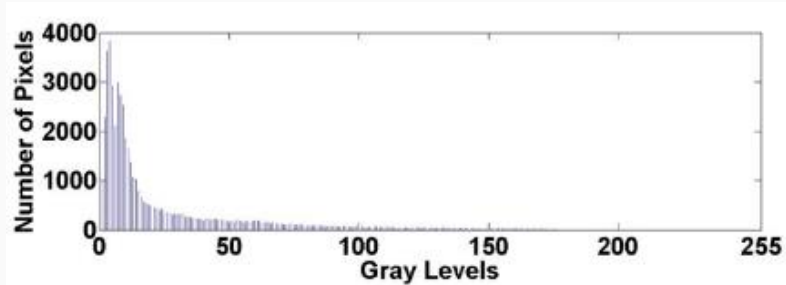


# Histogram Advantages

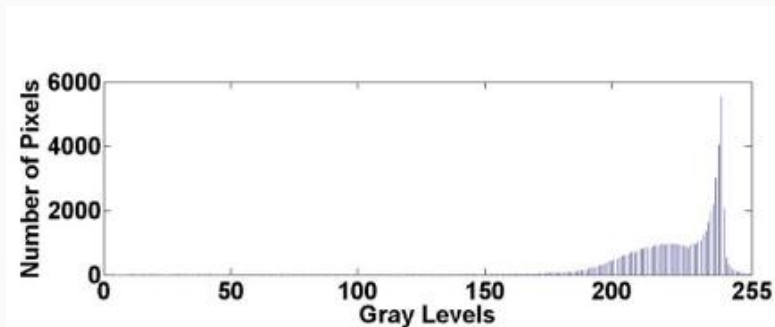
A mere look at the histogram reveals important facts

## 1. Image brightness

Histogram of a **dark** image: values are concentrated toward the **left**



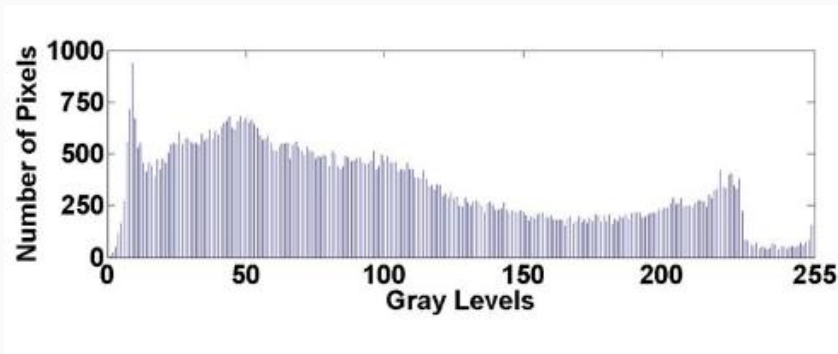
Histogram of a **bright** image: values are concentrated toward the **right**



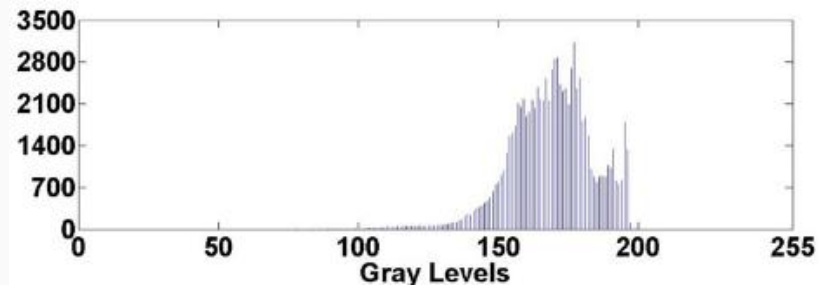
# Histogram Advantages

## 2. Contrast of the image

Histogram of a **high-contrast** image: pixel counts evenly **cover a broad range** of grayscale levels.



Histogram of a **low-contrast** image: Pixel counts that are **restricted** to a smaller range indicate low contrast



# Image Entropy

Entropy is a statistical measure of **randomness** that can be used to characterize the input image.

Entropy formula:

$$e = - \sum_{i=0}^{n-1} p_i \log p_i$$

**n** is the number of gray levels (256 for 8-bit images),

**p<sub>i</sub>** is the probability of a pixel having gray level intensity

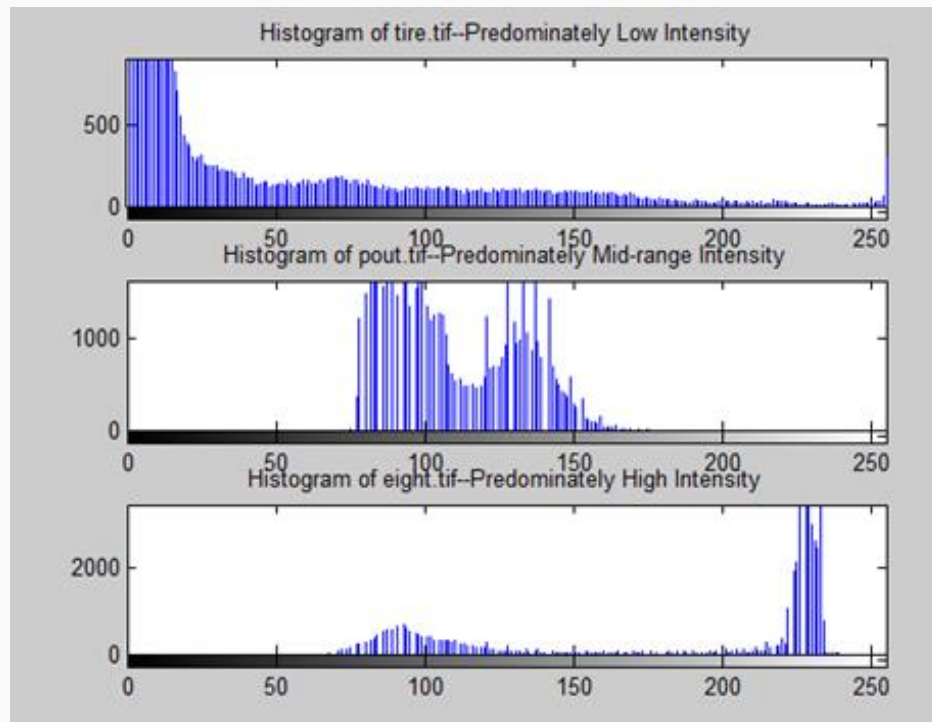


# Image Entropy

Suppose we have these 3 images and their related histograms,

We calculate entropy for each image and get three values : **5.75** , **4.44** , **6.95**

Which value relate to each of these images?? And Why ??



# Image Entropy

Write java code to calculate histogram, draw it and check the previous values.  
Here is Snippet of code

```
count = np.zeros(shape: 256, np.uint64)
for i in range(row):
    for j in range(col):
        count[img[i,j]] += 1
```

```
entropy = 0
for i in range(len(prob)):
    if prob[i]:
        entropy -= prob[i]*np.log2(prob[i])
```