

# **Pixels and Histogram**

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# Pixel

- Each pixel is made up of a red, green and blue subpixel that lights up at different intensities to create different colors.

## What do pixels look like?

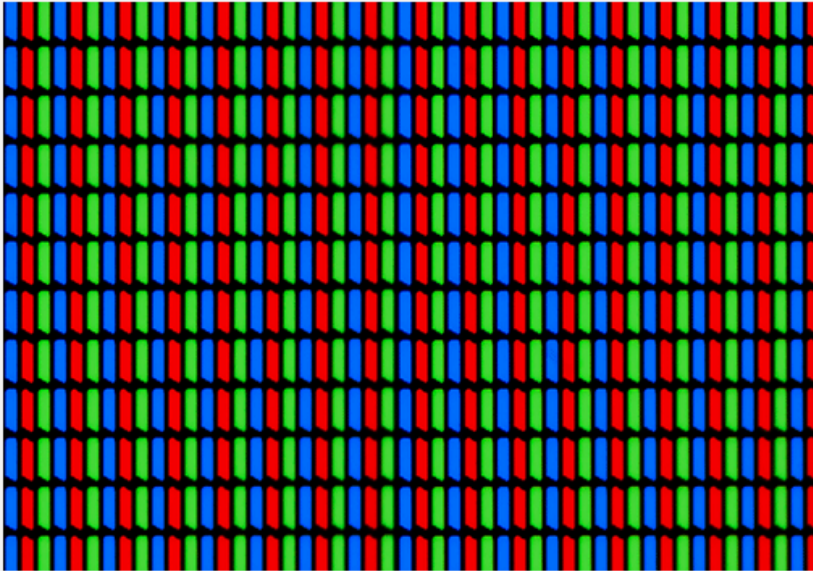
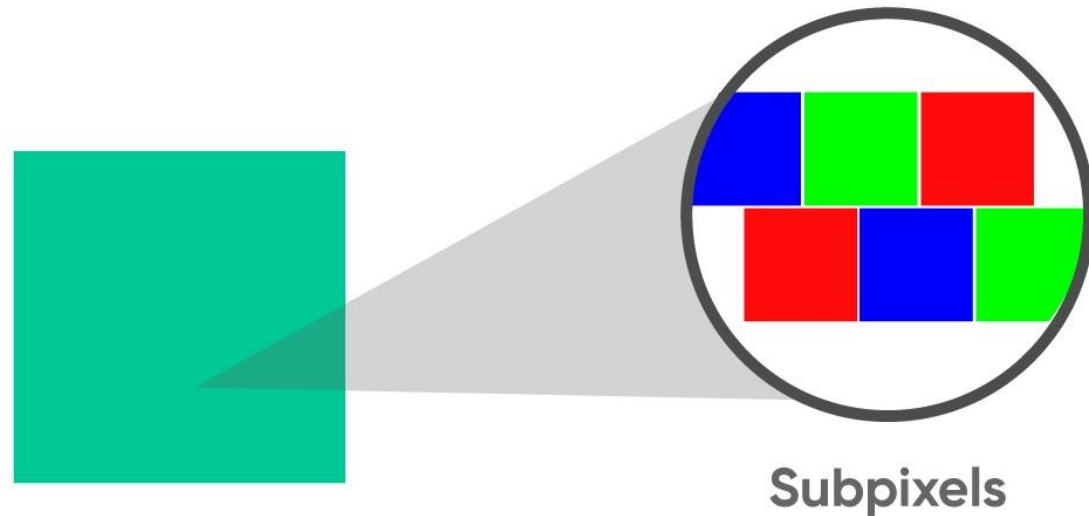


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Pixels

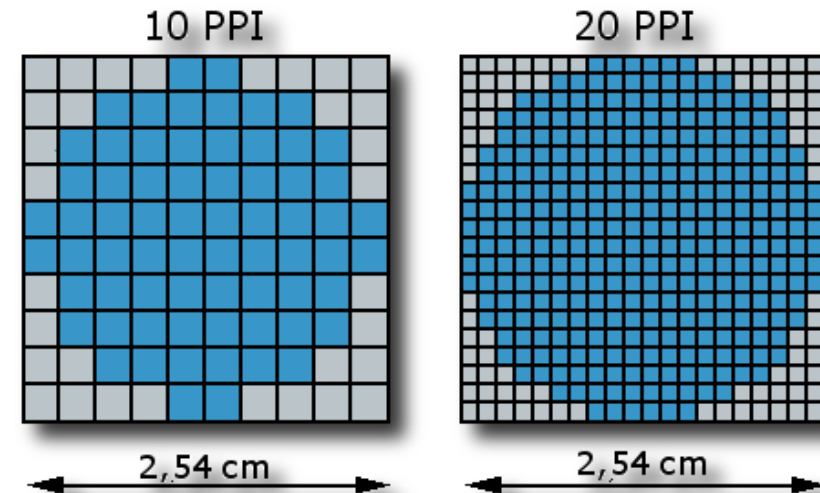
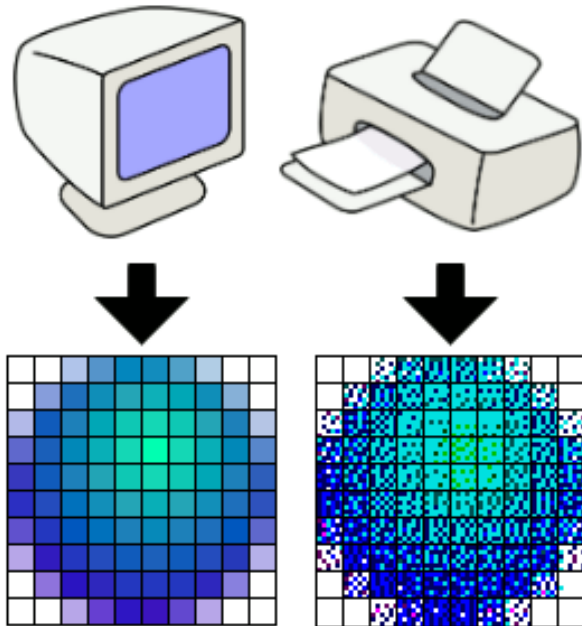
Subpixels

# Pixel: DPI vs PPI

- DPI (Dots Per Inch) is a measure of spatial printing, video or image scanner dot density
- PPI (Pixels Per Inch) is a measure of the pixel density of an electronic image/display device
- Number of dots/pixels that can be placed in a line within the span of 1 inch (2.54 cm)

<https://photographycourse.net/dpi-vs-ppi/>

$$\text{PPI (monitor)} = \frac{\text{Number of Pixels}}{\text{Size in Inches}} = \frac{1920}{20} = 96 \text{ ppi}$$



# PPI vs DPI: Do they affect each other?

- Imagine you want to print a 300 PPI image at 600 DPI
  - Simply divide 600 DPI/ 300 PPI and you have your answer 2 (DPI/PPI)

## PPI vs DPI

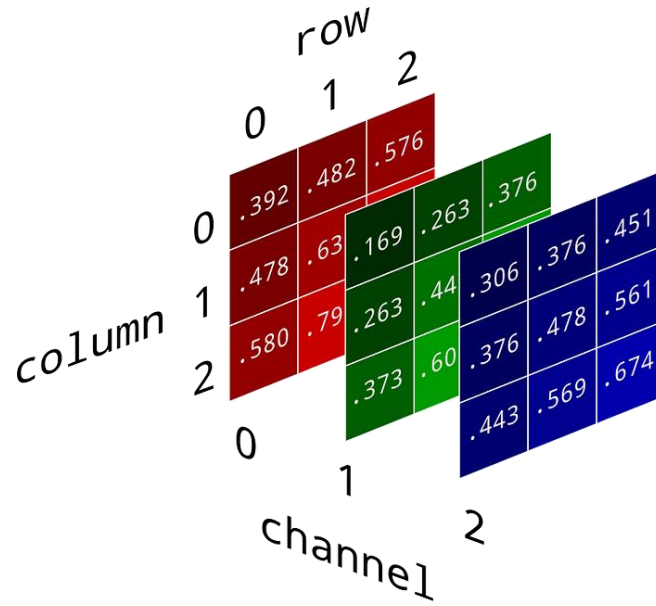
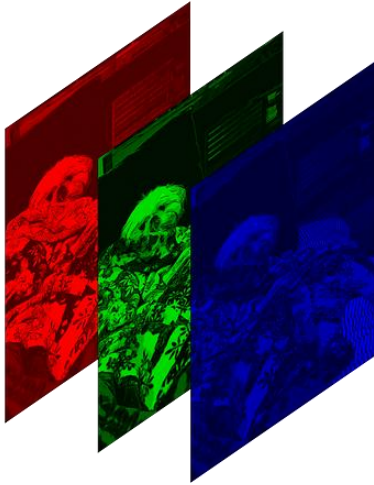


- Display resolution
- Number of pixels displayed in one inch of a digital image

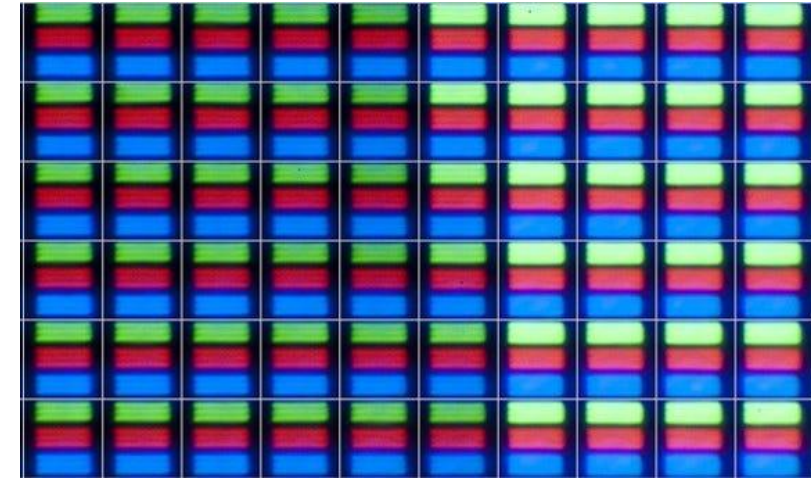


- Printer resolution
- Number of dots of ink on a printed image

# Pixel Values



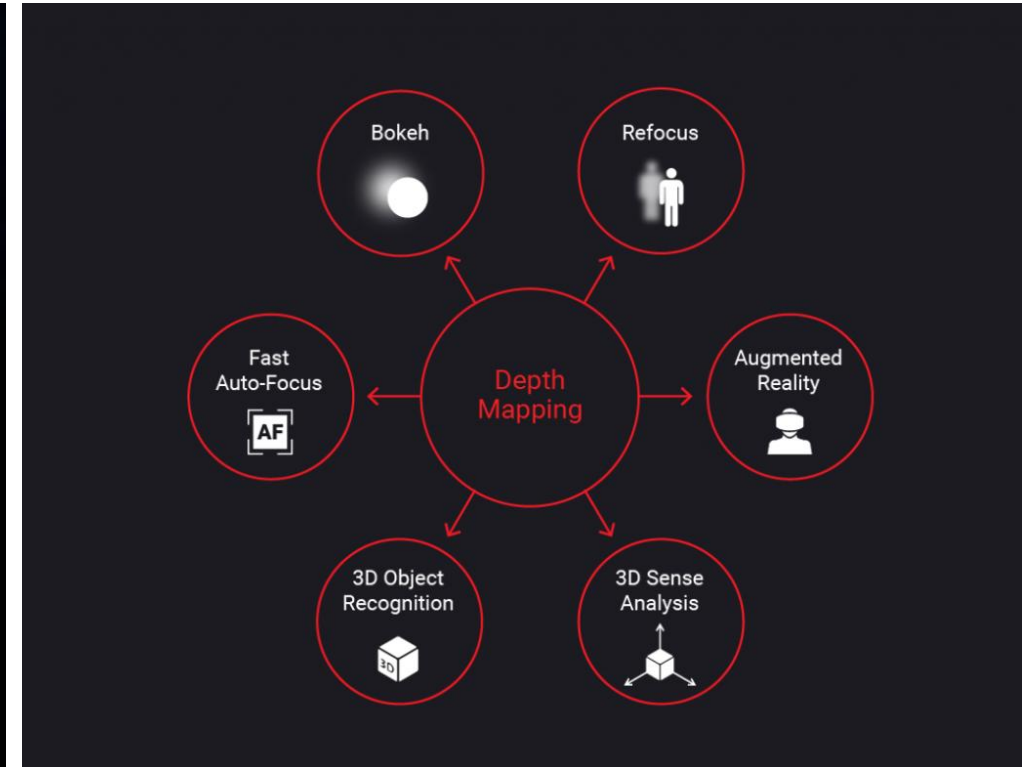
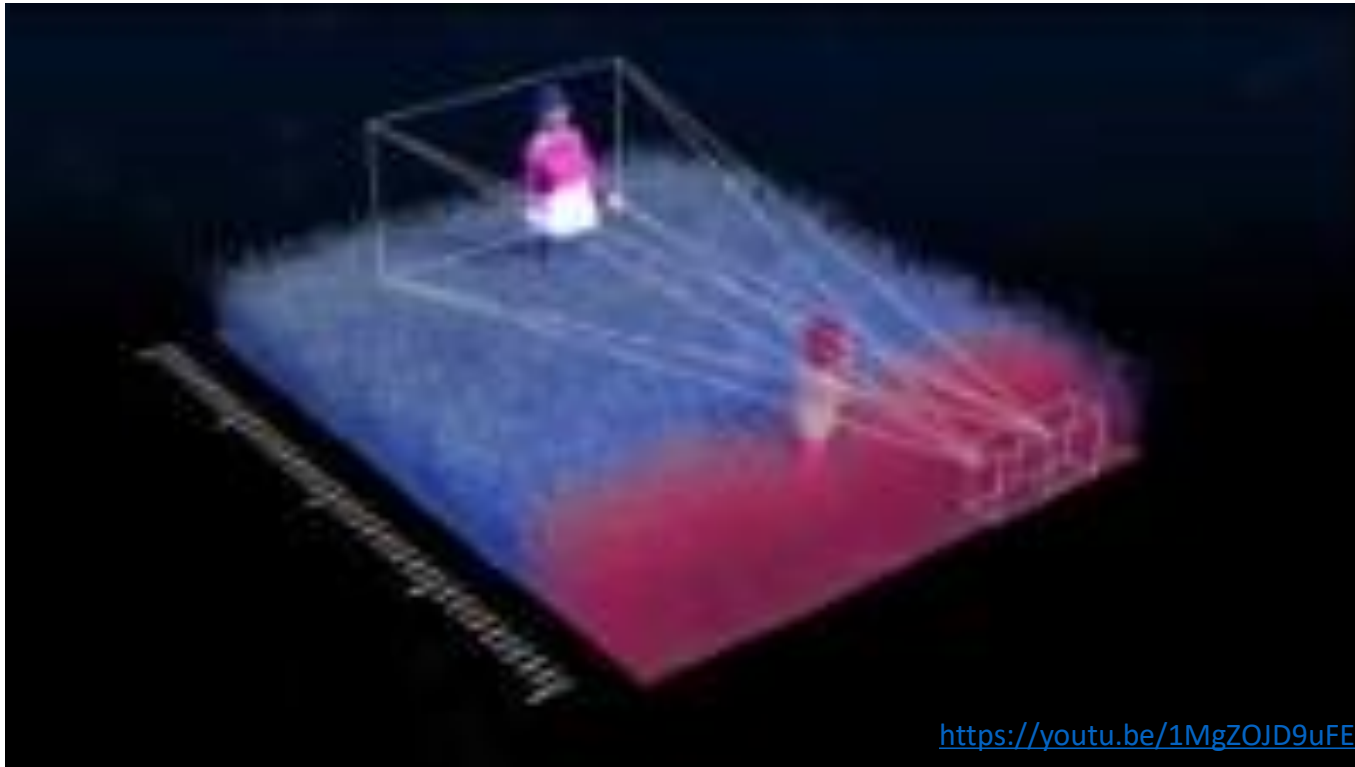
Pixel array on an Image



Pixel array on a display monitor

# Image vs (Image) Map

- Image represents a pixel array what we see, or a picture
- (Image or 2D) Map represents an array where each element is mapped to each pixel of the target image



**Now we have pixels and their values  
Let's analyze them!  
for further processing**

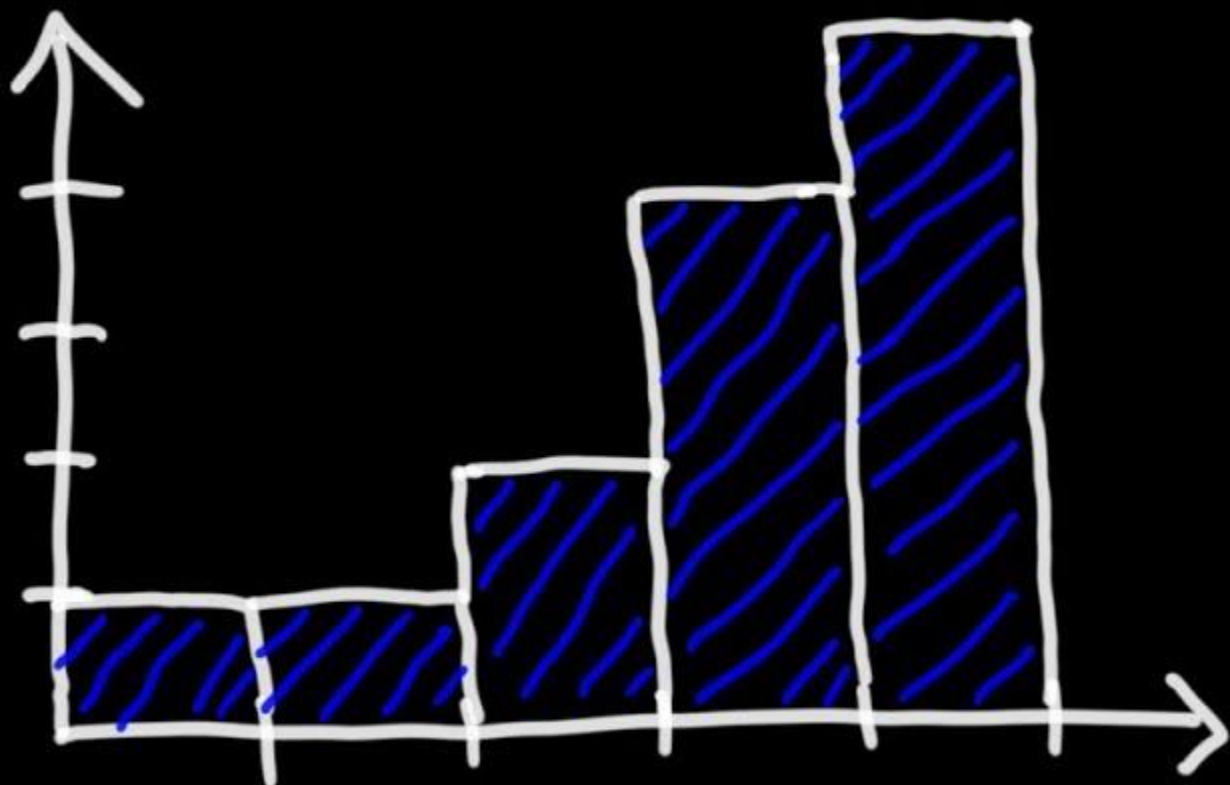
**Where do we start?** 



# Histograms

1. How many students received, at most, a score of 69 on the exam?
2. How many students received a score of at least 80 on the exam?
3. How many students received a score between 60 and 90?

Grades vs freq.

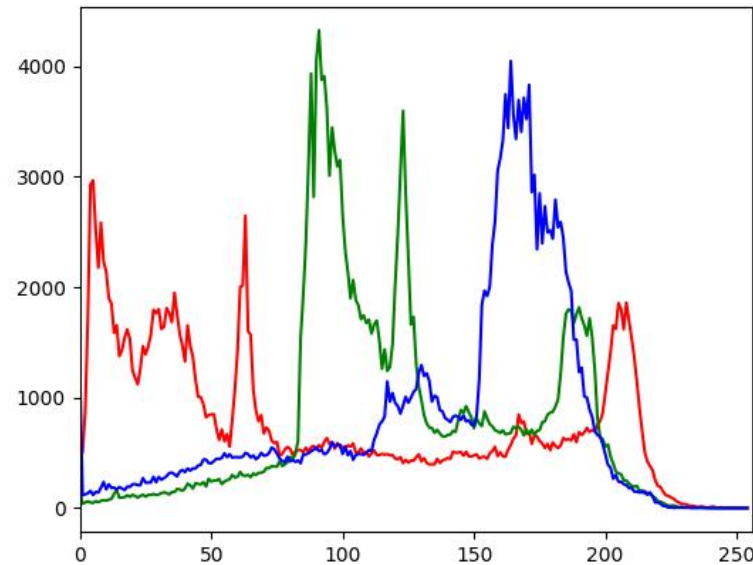
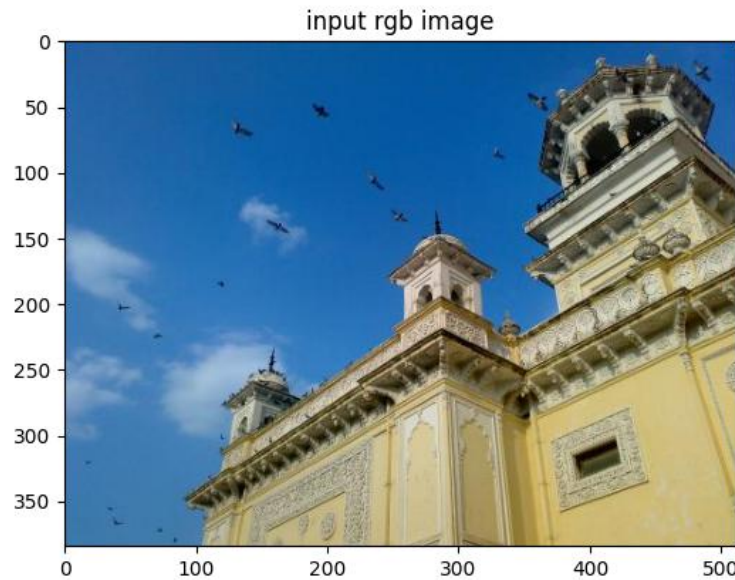


Test Scores: 74, 83, 69, 95, 78, 85, 42, 98, 73, 68, 90, 85, 84, 71, 88, 52, 94

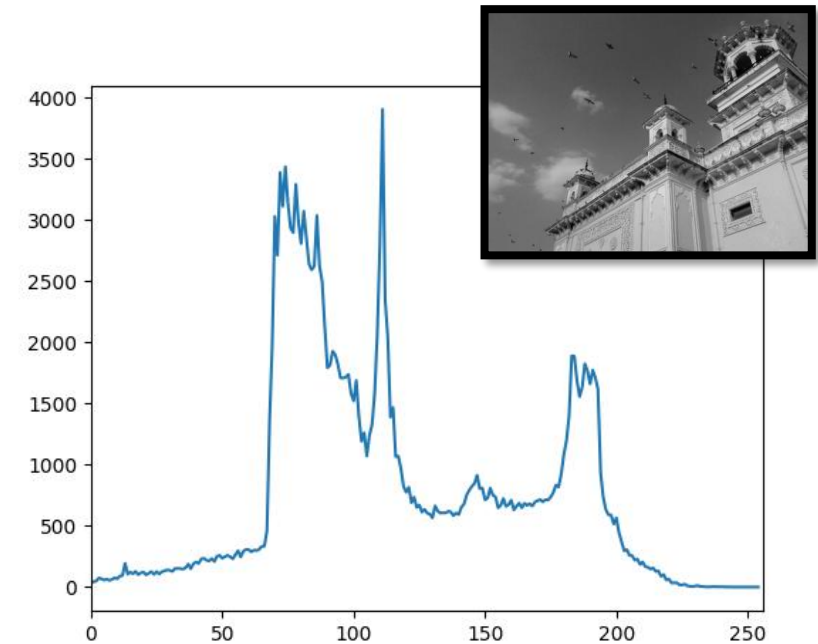


# (Image) Histogram

- A plot with pixel values in x-axis and corresponding number of pixels in the image on y-axis
- For a RGB image, there are three histograms
- A representative value per pixel is normally used, i.e., gray scale image



histograms for each RGB channel



1-channel histogram for gray image

# Thresholding

- Do something if a metric value is larger/less than a criterion value, i.e., ***threshold value***
- Specific function is applied to pixels whose values are larger/less than a threshold value
- Simple thresholding (global thresholding) uses **a single threshold value**
- Adaptive thresholding uses **'per-pixel-dynamic' threshold values**
  - Locally determined by considering a target pixel's neighborhood pixel values
  - This kind of approach is called **'adaptive'** method



# Simple (Global) Thresholding

```
import cv2 as cv
import numpy as np
from matplotlib import pyplot as plt
```

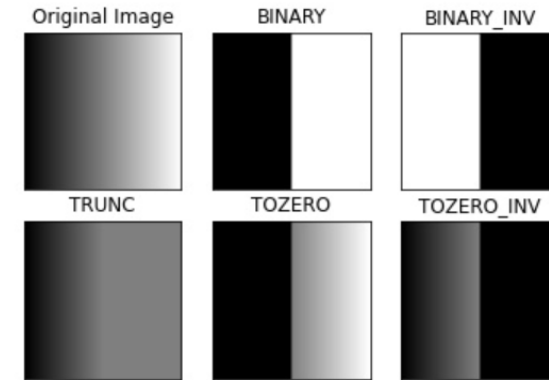
*“matplotlib is a very useful plot GUI in python!!”*

```
img = cv.imread('gradient.png',0)
ret,thresh1 = cv.threshold(img,127,255,cv.THRESH_BINARY)
ret,thresh2 = cv.threshold(img,127,255,cv.THRESH_BINARY_INV)
ret,thresh3 = cv.threshold(img,127,255,cv.THRESH_TRUNC)
ret,thresh4 = cv.threshold(img,127,255,cv.THRESH_TOZERO)
ret,thresh5 = cv.threshold(img,127,255,cv.THRESH_TOZERO_INV)

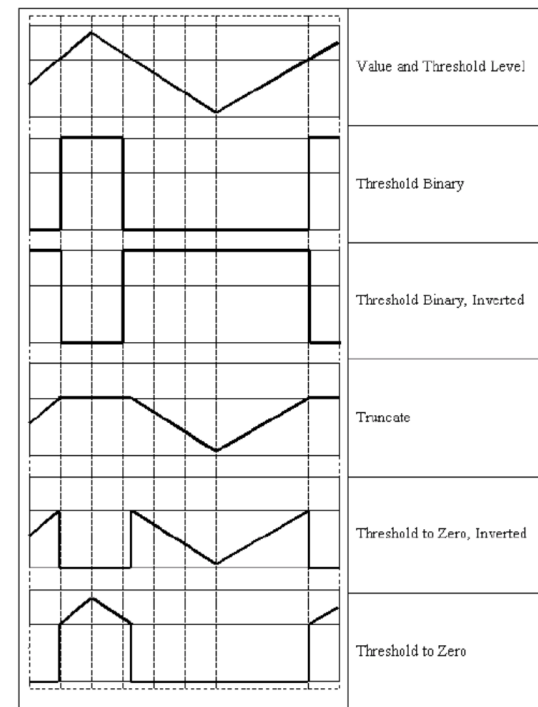
titles = ['Original Image','BINARY','BINARY_INV','TRUNC','TOZERO','TOZERO_INV']
images = [img, thresh1, thresh2, thresh3, thresh4, thresh5]
```

```
for i in range(6):
    plt.subplot(2,3,i+1),plt.imshow(images[i],'gray',vmin=0,vmax=255)
    plt.title(titles[i])
    plt.xticks([],plt.yticks([]))

plt.show()
```



image



threshold types

**OpenCV**
4.5.2-pre

Open Source Computer Vision

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[Examples](#)
[Java documentation](#)

OpenCV-Python Tutorials
Image Processing in OpenCV

THRESH\_TRUNC cv  
thresh\_trunc\_func cv:cudev

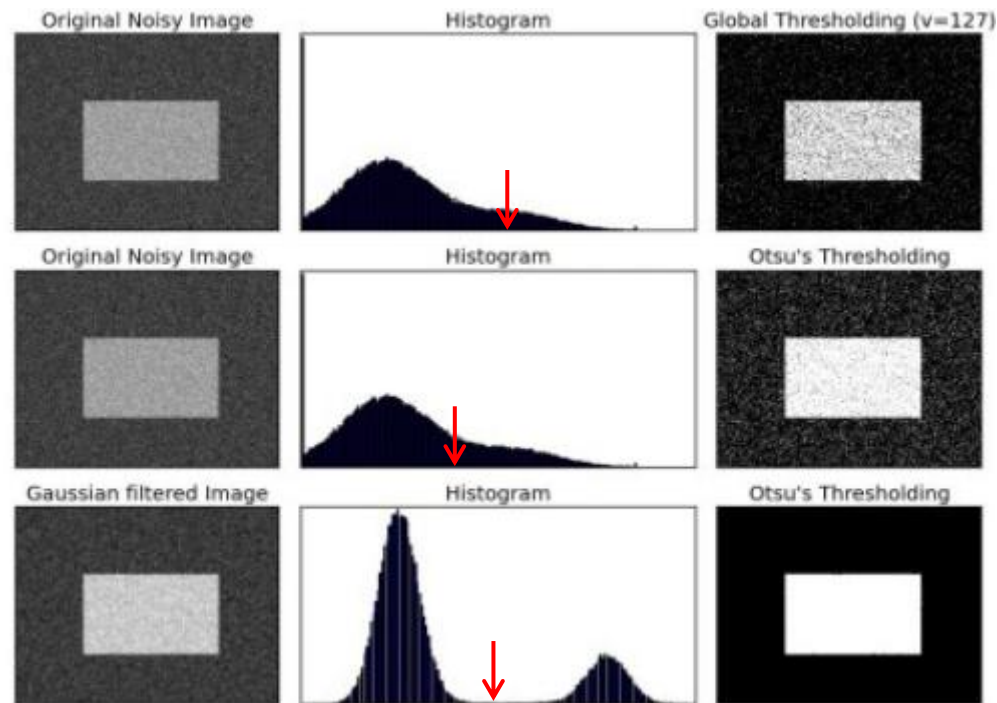
## Image Thresholding

### Goal

- In this tutorial, you will learn simple thresholding, and how to use the functions `cv.threshold` and `cv.adaptiveThreshold`.
- You will learn the functions `cv.threshold` and `cv.adaptiveThreshold`.

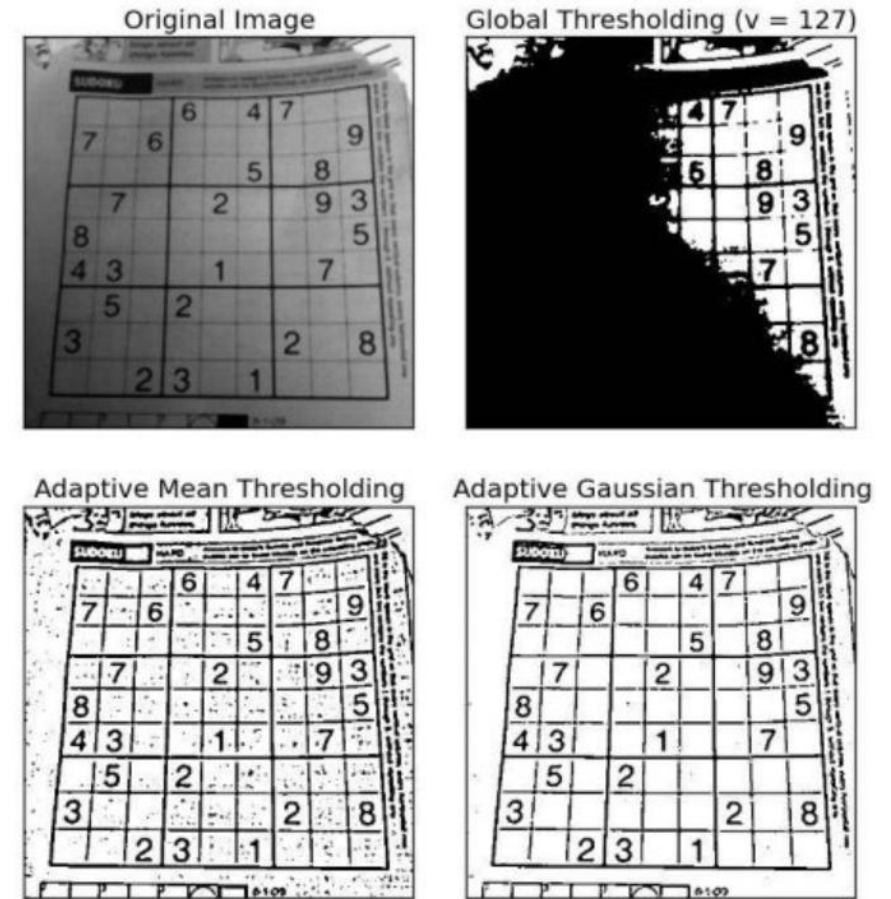
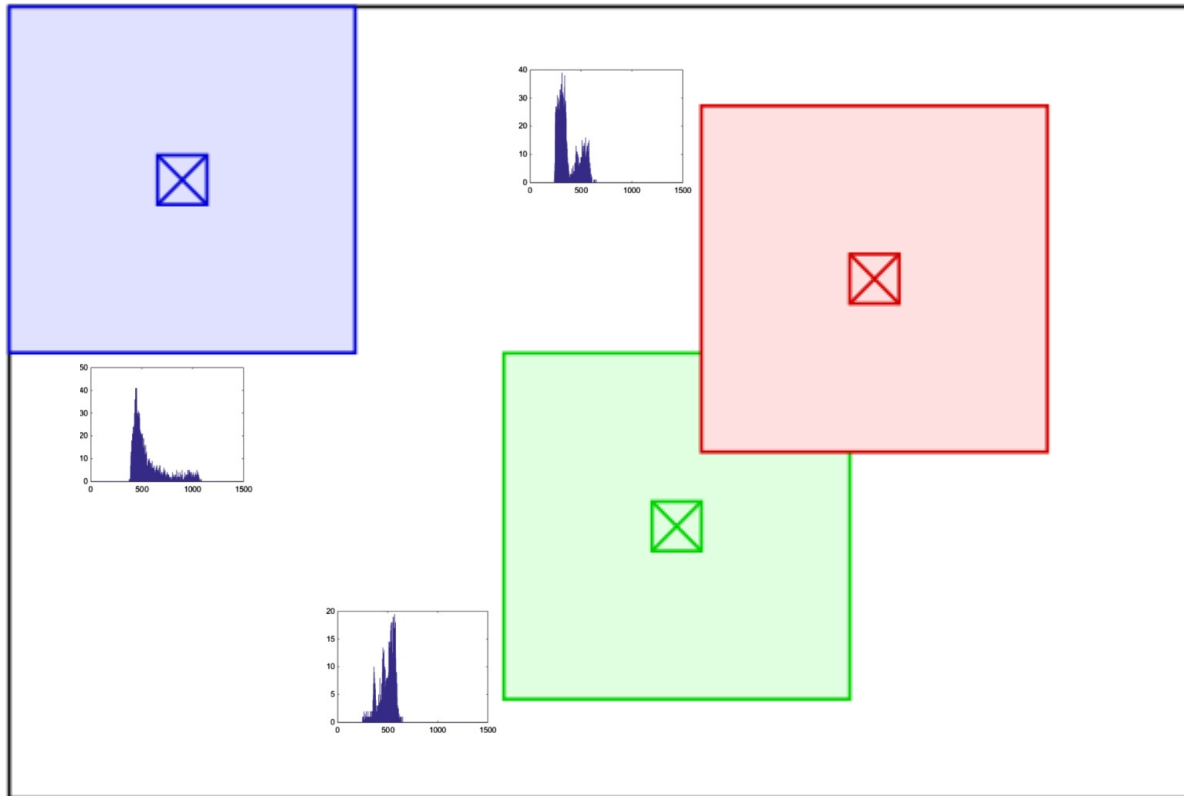
# Threshold Value

- It is important to determine the appropriate threshold value
- Otsu method (or, Otsu thresholding) determines the threshold value automatically
  - Otsu threshold value minimizes the within-cluster(or class) variance of the histogram
  - useful for a simple binary classification case



# Adaptive Thresholding

- Locally-determined threshold for a pixel based on a small region around it
  - How do we decide the threshold value? 😊





# Which image looks better?



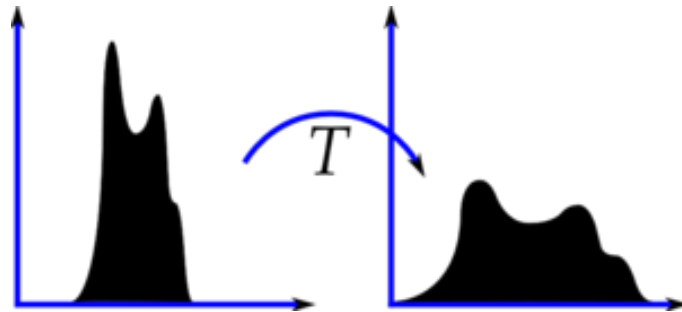


# Histogram Equalization (HE)

- Generally, consider 1-channel histogram (e.g., gray image histogram)
- To enhance the image's contrast, we can modify the individual pixel values
  - Use the histogram processing, **Histogram Equalization**



the image uses a small range of intensity values



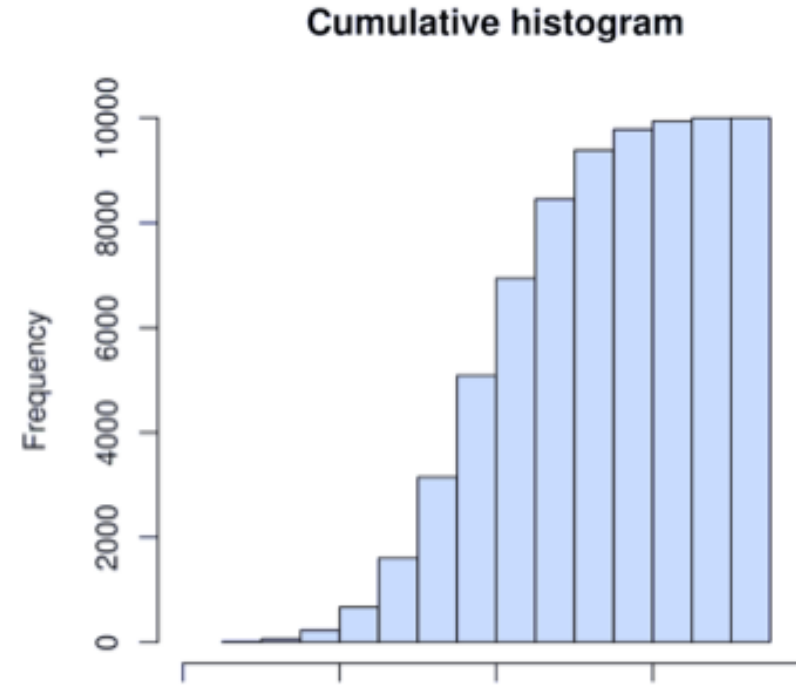
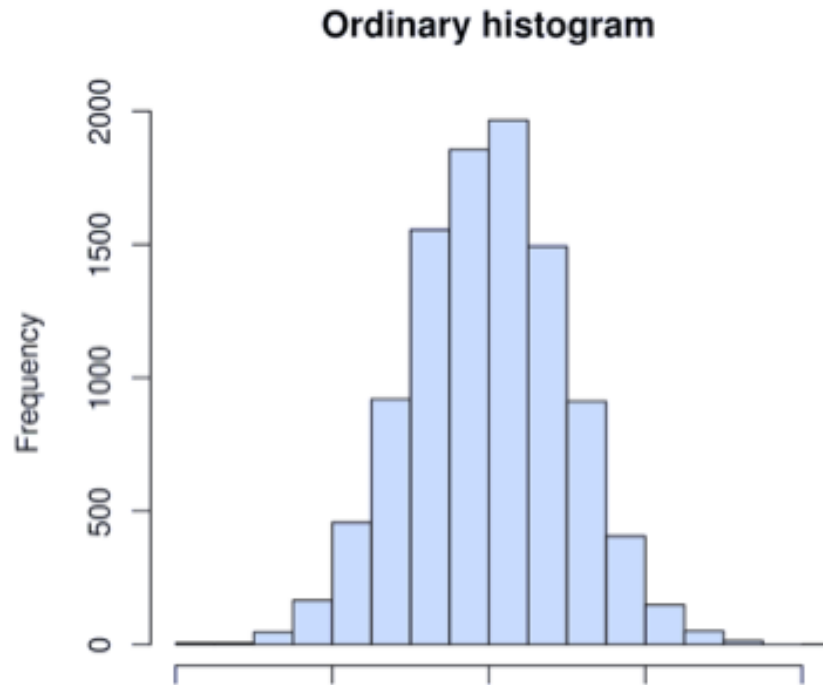
the small range pixel values are stretched out along the valid pixel values

looks better!!



# Cumulative Histogram

- The histogram can be thought of a distribution function



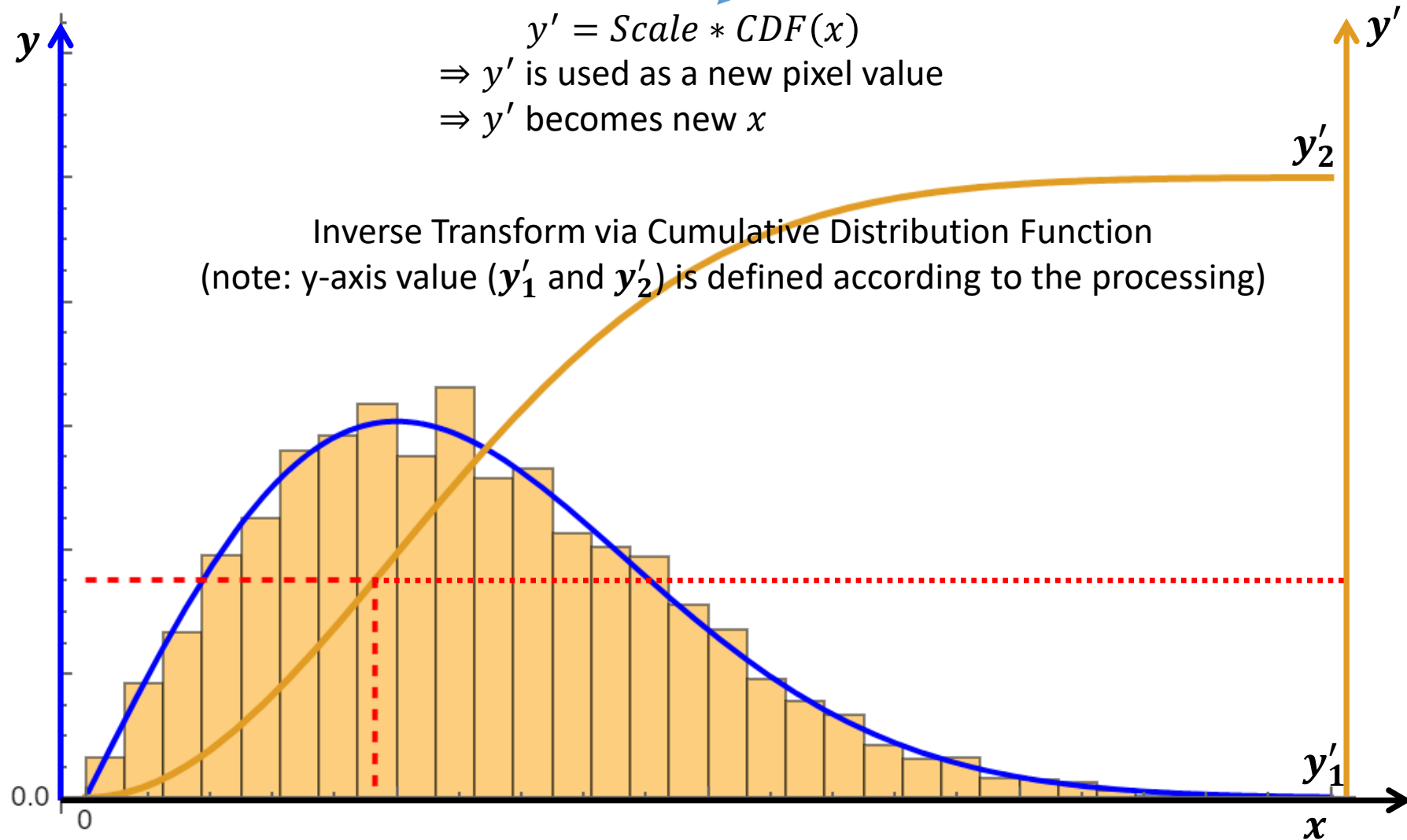
# Inverse Transform of a Histogram

$$y'_1 = \text{Histo}(x) \longrightarrow y'_2 = \text{CDF}(x)$$

$$y' = \text{Scale} * \text{CDF}(x)$$

$\Rightarrow y'$  is used as a new pixel value

$\Rightarrow y'$  becomes new  $x$

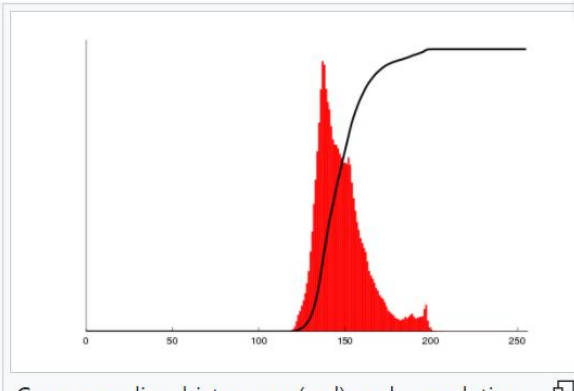


# Histogram Equalization (HE)

- HE is accomplished by the following steps
  1. Making a histogram of an image
  2. Making a cumulative histogram of the histogram
  3. Mapping min/max (user-defined) pixel values to 0/255 satisfying **linearized CDF!!**
- HE allows to **gain a higher contrast!**



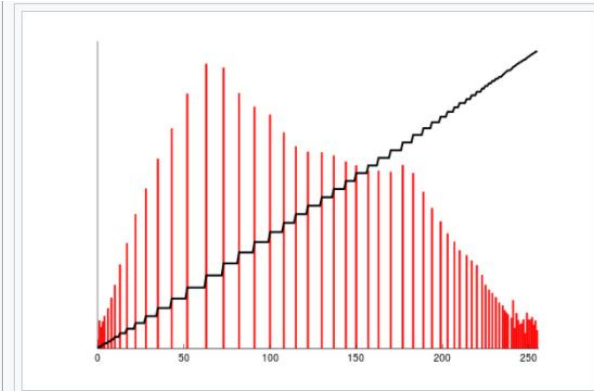
Before Histogram Equalization



Corresponding histogram (red) and cumulative histogram (black)

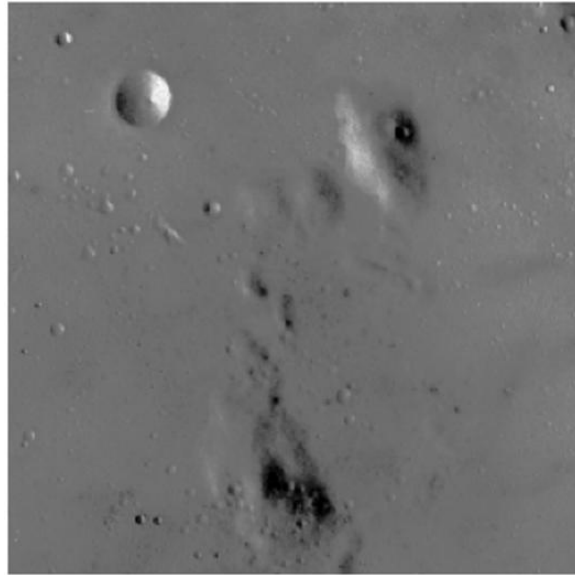


After Histogram Equalization

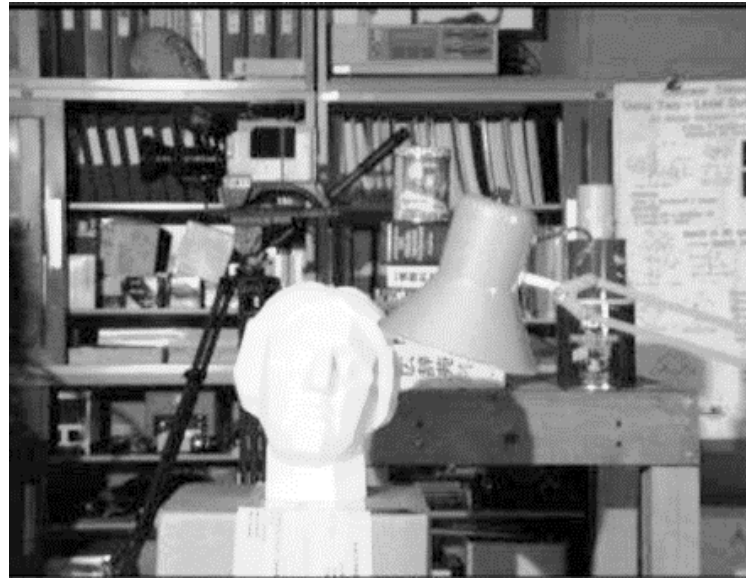


Corresponding histogram (red) and cumulative histogram (black)

# Is only HE enough?

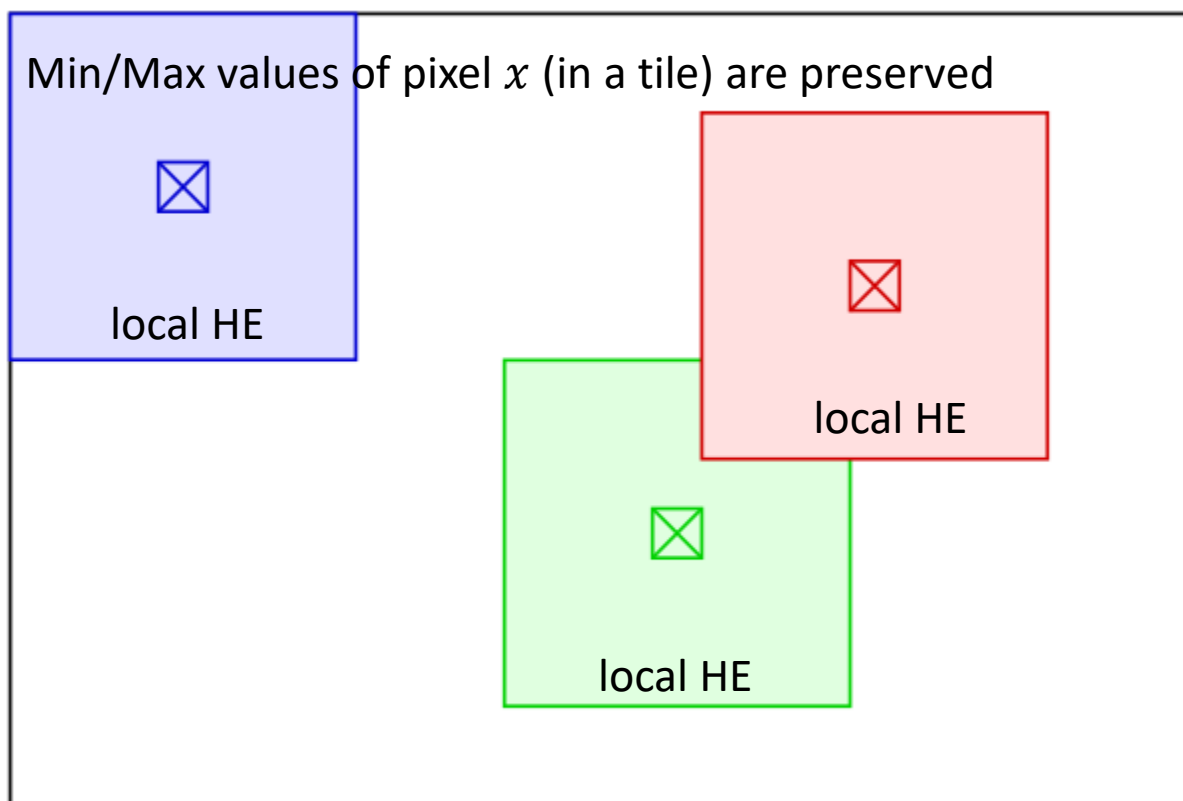


☹️?!



# Adaptive Histogram Equalization (AHE)

- In the previous, only considered the global contrast of the image
  - In many cases, over- or under- brightness occurs
- Obviously, **Adaptive** method exists based on **local analysis** of the image histogram
  - At a pixel, HE is performed on a **tile**, or **block**, centered at the pixel



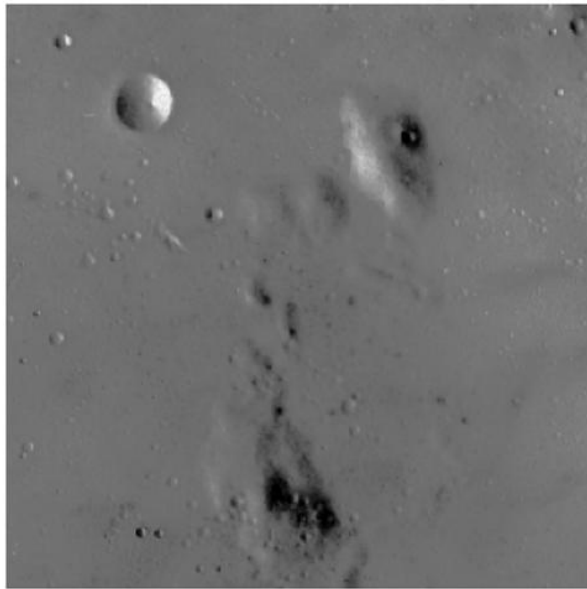
OpenCV uses 8x8 tile size by default



# Adaptive Histogram Equalization (AHE)

- In the previous, only considered the global contrast of the image
  - In many cases, over- or under- brightness occurs
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Low contrast image

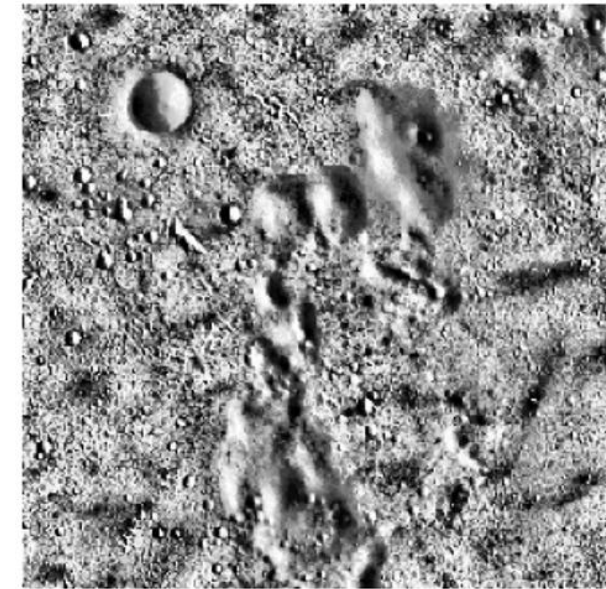


Global equalization



over- or under- brightness

Local equalization



better local details!!

with more consistent contrast enhancement!

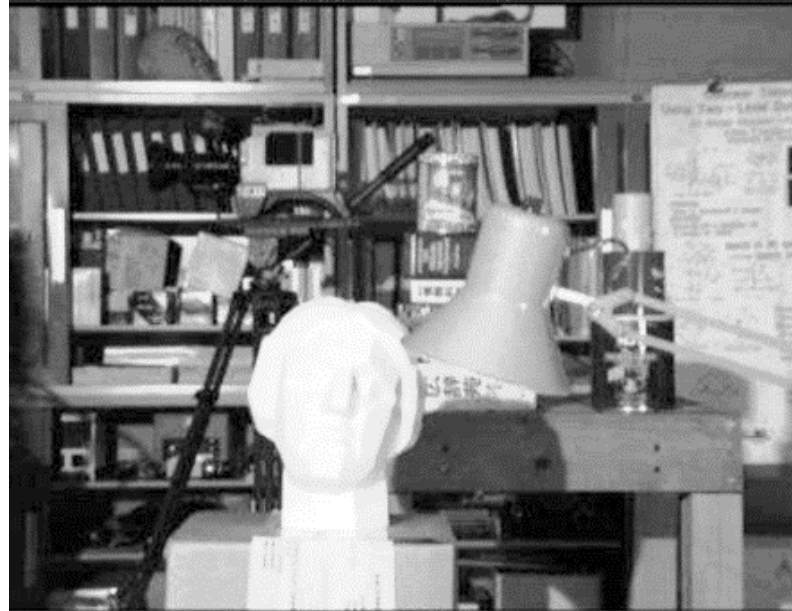
# Contrast Limited AHE (CLAHE)

- AHE often suffers from overamplification problem
  - Due to the pixels whose values are biased at a narrow range of the local histogram

original

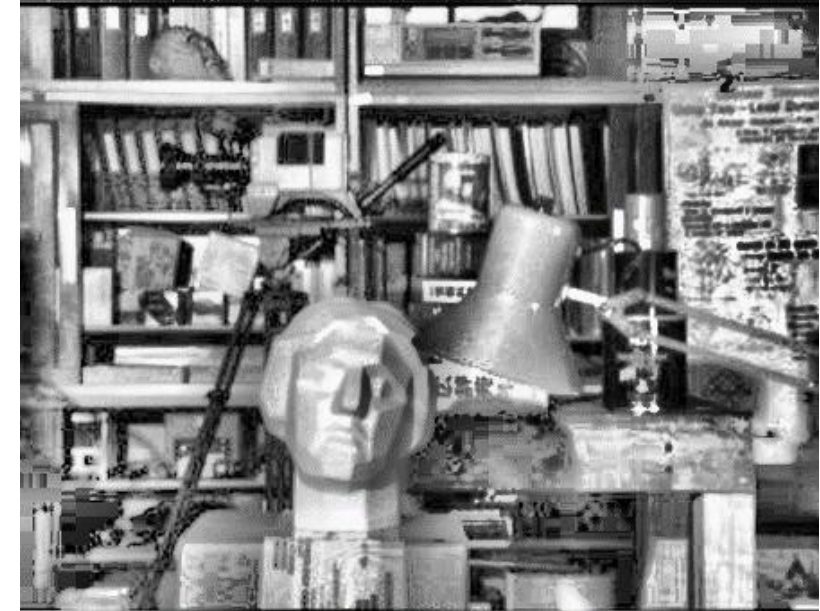


global HE



over-brightness

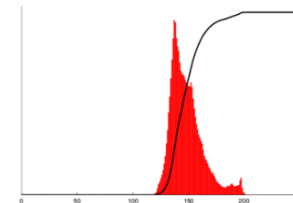
adaptive HE



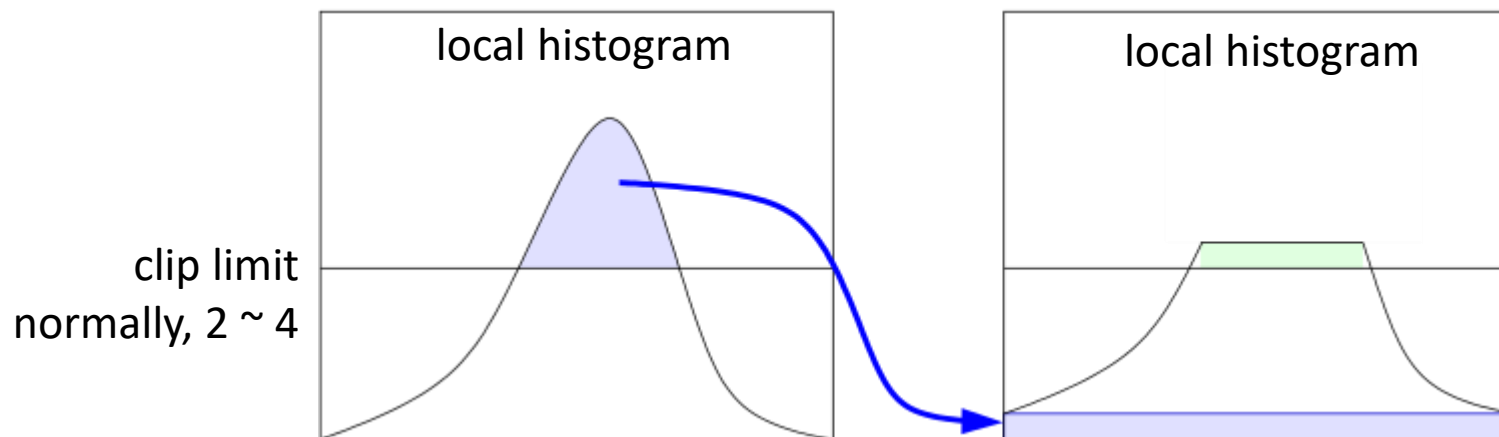
noise-like areas  
(by overamplified)

`cv.createCLAHE(clipLimit= $\infty$ ,...).apply(...)`

# Contrast Limited AHE (CLAHE)



- AHE often suffers from overamplification problem
  - Due to the pixels whose values are biased at a narrow range of the local histogram
- By limiting contrast (or, clipping contrast), avoid an **over-sloped CDF** of the local histogram



Imagine the CDF of each histogram (before/after clipping)!

Note that the linear-like slope of CDF provides better contrast!

Min/Max values of pixel  $x$  (in a tile) are preserved



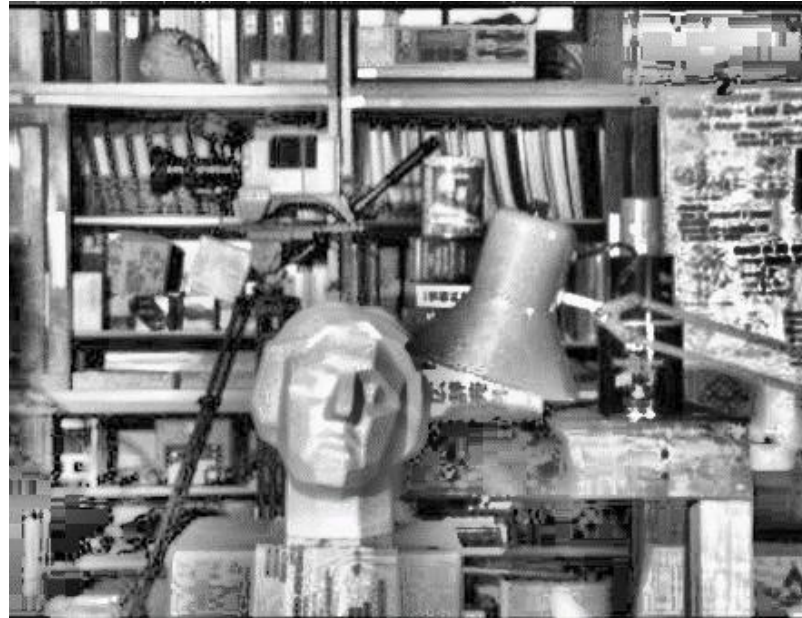
# Contrast Limited AHE (CLAHE)

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original

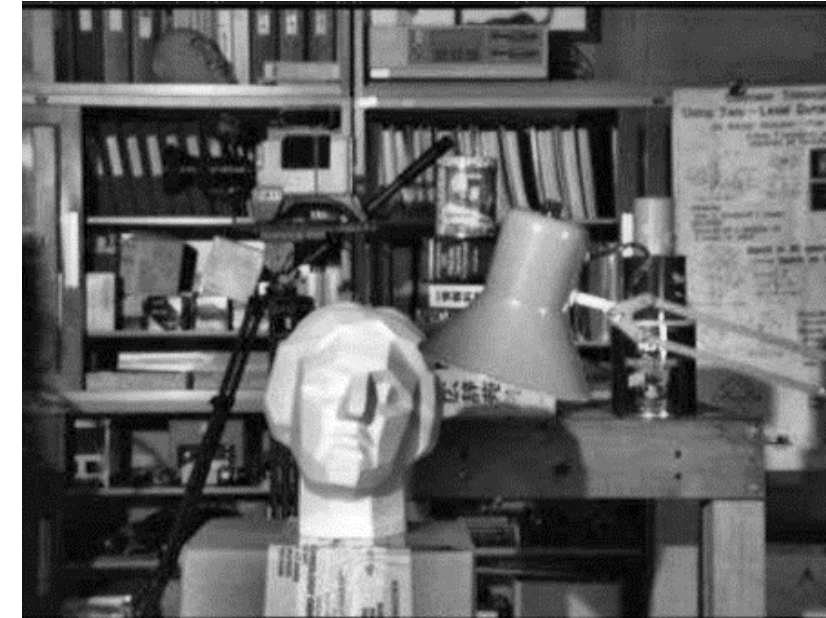


adaptive HE



`cv.createCLAHE(clipLimit=  $\infty$ ,...).apply(...)`

CLAHE



`cv.createCLAHE(clipLimit=2,...).apply(...)`

Op

trast (CLA

- 

overam

hose val

- 

clipping

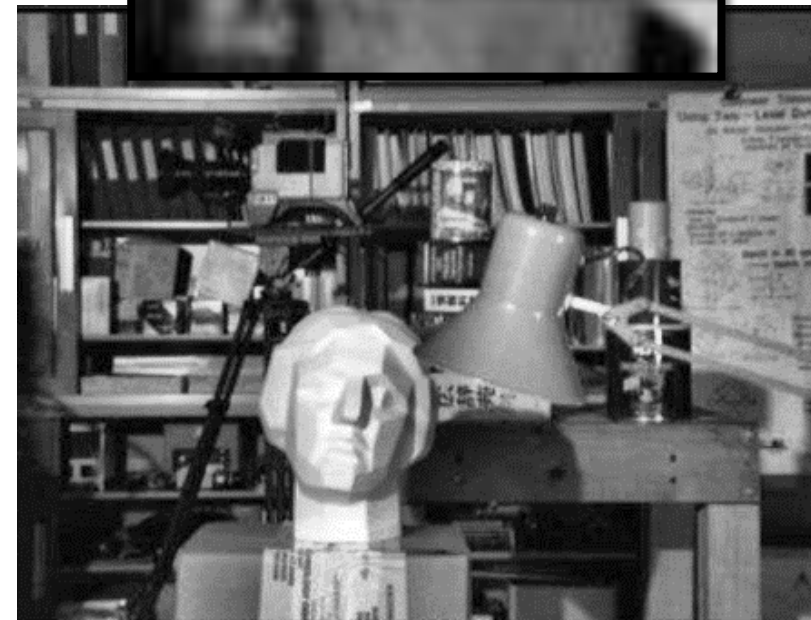
low range

er-sloped

am



`cv.createCLAHE(clipLimit=  $\infty$ ,...).apply(...)`

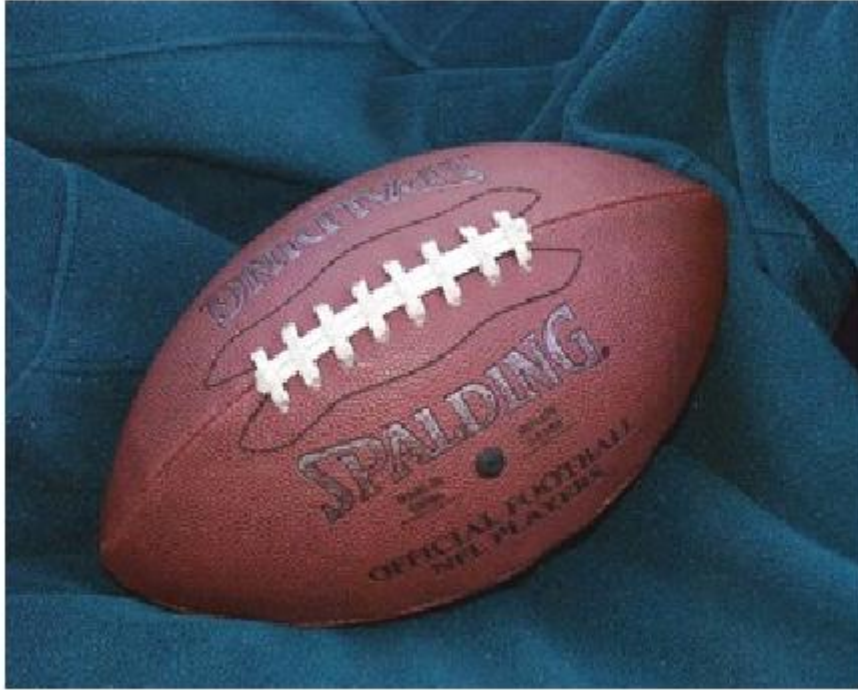


`cv.createCLAHE(clipLimit=2,...).apply(...)`

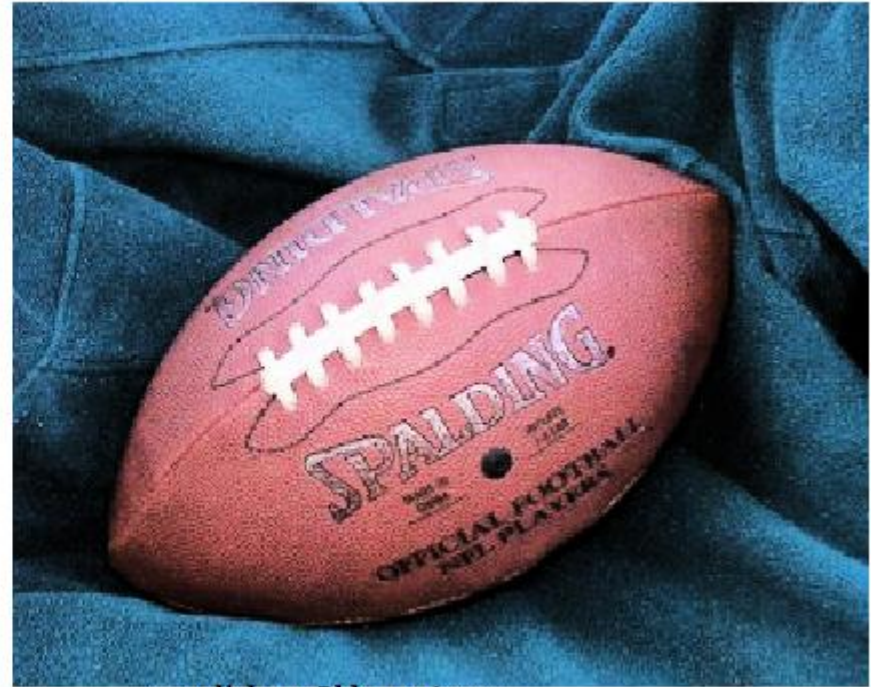


# Can you apply HE to Color Image?

Before Histogram Equalization



After Histogram Equalization



angeljohnsy@blogspot.com

Hint : remember color space you already learned!





# Mathematical Approaches for Image Processing

# Variance and Standard Deviation in Statistics

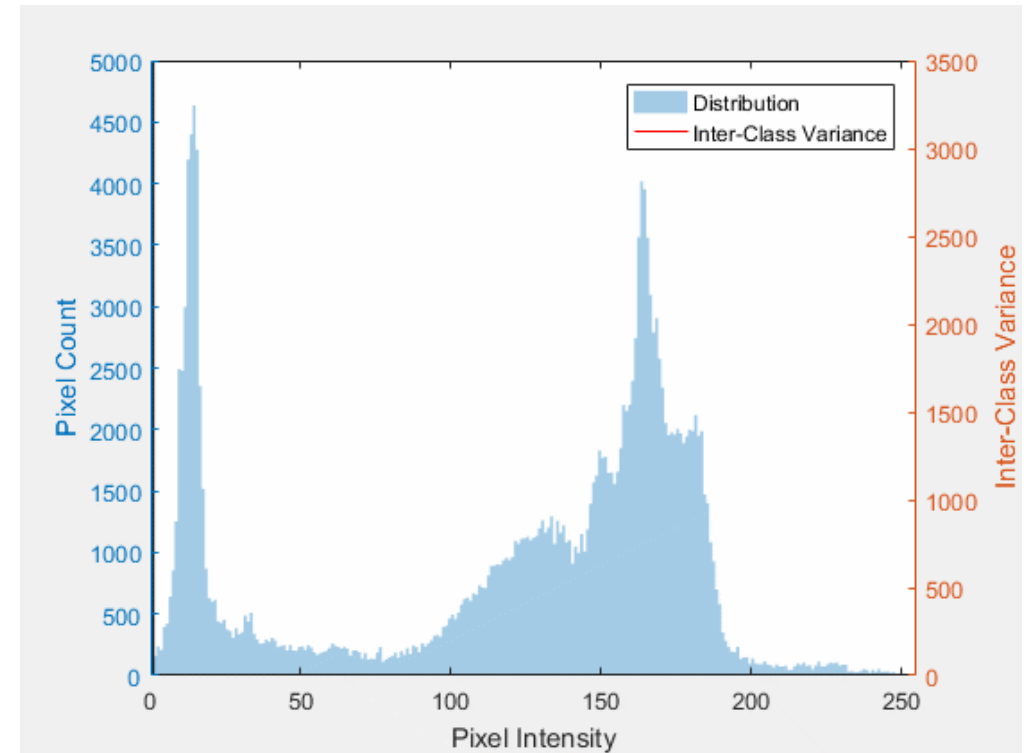
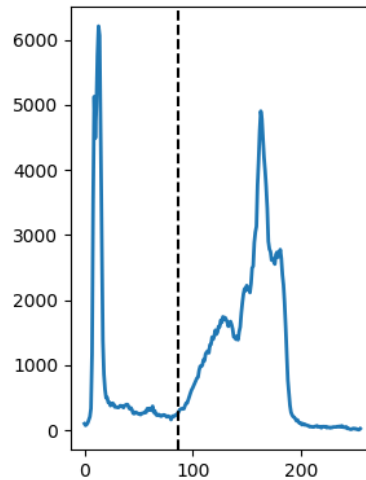


Although they are same concept,  
Why Standard Deviation? instead of Variance?



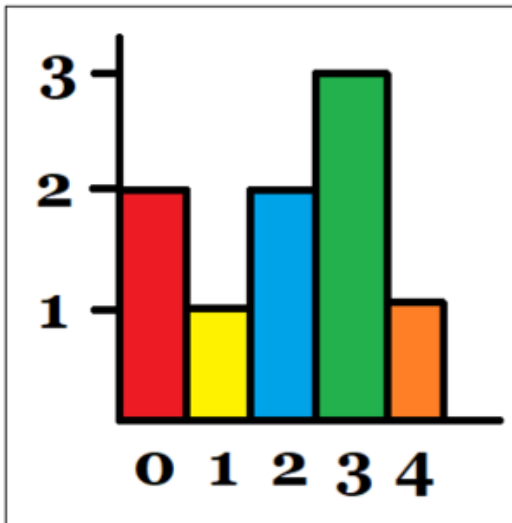
# Otsu Thresholding

- Considering the histogram variance enables to detect a global threshold value for binary classification automatically
- Otsu threshold value **maximizes the inter-class variance** of the histogram (or **minimizes the within-class variance**)

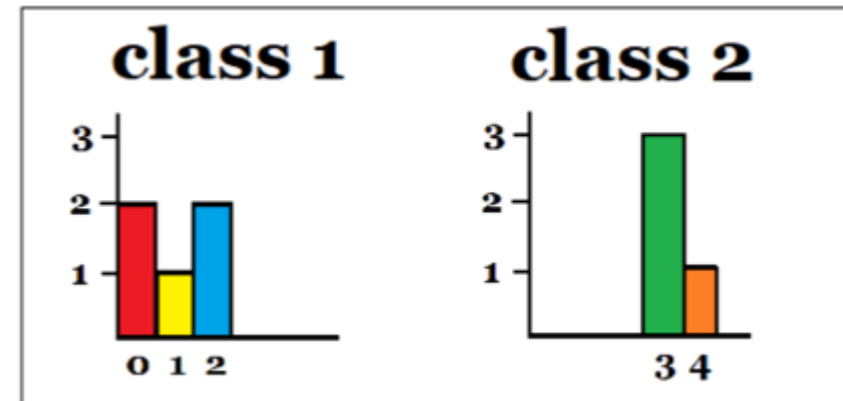


# Otsu Thresholding

- We know 'variance'  $\sigma^2 = \frac{\sum_{i=0}^N (X_i - \mu)^2}{N}$   $X_i$ : i-th pixel value (histogram),  
 $\mu$ : mean,  $N$ : # of pixels on an image
- Let's assume that pixels are classified into **2 classes**
- The within-class variance  $(V_w) = \sum_{i=0}^N (W_i * \sigma_i^2)$   $W_i$ : # of pixels in class  $i$  /  $N$



divide into  
two classes



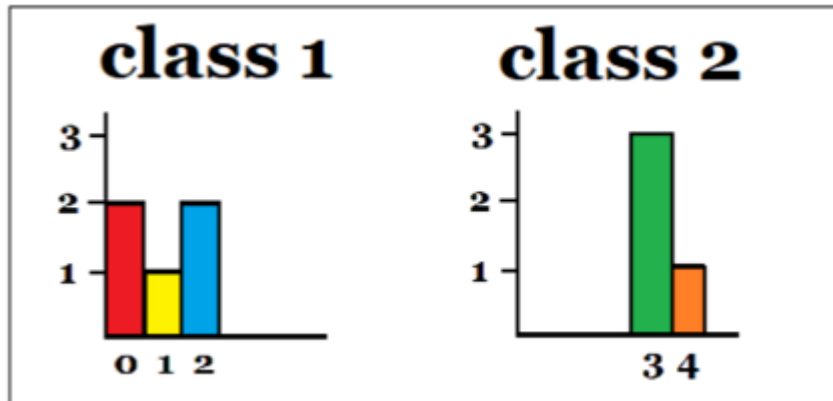
$$W_1 = 5/9 \quad W_2 = 4/9$$
$$\sigma_1^2 = 4/5 \quad \sigma_2^2 = 3/16$$

$$V_w = W_1 * \sigma_1^2 + W_2 * \sigma_2^2 = 0.52777$$

compute  $V_w$  along the pixel values in histogram,  
and find the value that minimizes the  $V_w$

# Otsu Thresholding

- We know 'variance'  $\sigma^2 = \frac{\sum_{i=0}^N (X_i - \mu)^2}{N}$   $X_i$ : i-th pixel value (histogram),  
 $\mu$ : mean,  $N$ : # of pixels on an image
- Let's assume that pixels are classified into **2 classes**
- The within-class variance  $(V_w) = \sum_{i=0}^N (W_i * \sigma_i^2)$   $W_i$ : # of pixels in class  $i$  /  $N$
- The inter-class variance  $(V_b) = V_T - V_w$   $V_T$ : total variance



compute  $V_w$  along the pixel values in histogram,  
and find the value that **minimizes the  $V_w$**



compute  $V_b$  along the pixel values in histogram,  
and find the value that **maximizes the  $V_b$**

here, the value is the **Otsu threshold value**

# Adaptive Thresholding

- Locally-determined threshold for a pixel based on a small region around it
  - How do we decide the local threshold value?

How about Local-Otsu Thresholding? 😊