

Image Processing And Computer Vision

Algorithms and Applications

Motivation

- Editing pictures, movies



+



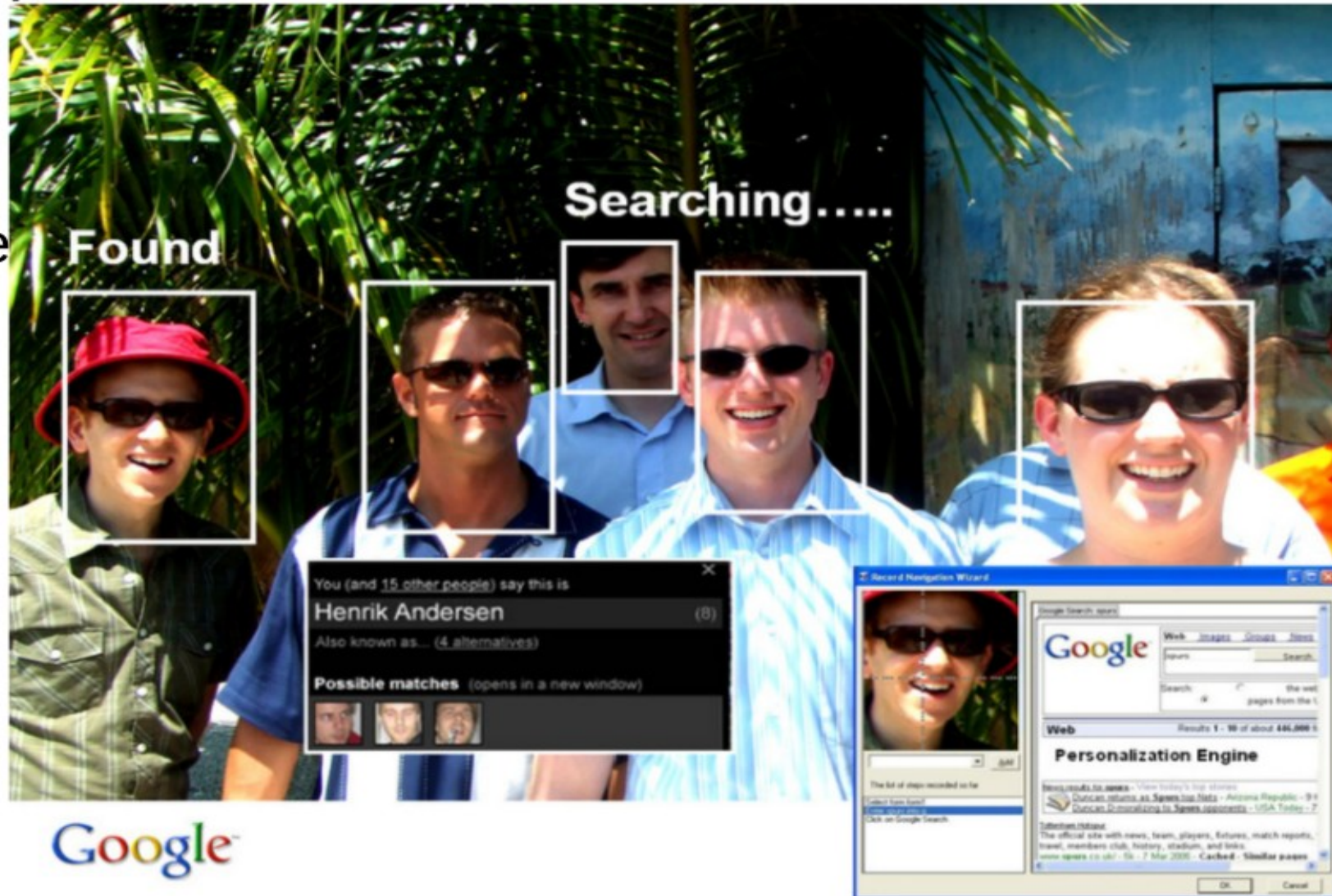
+



Motivation

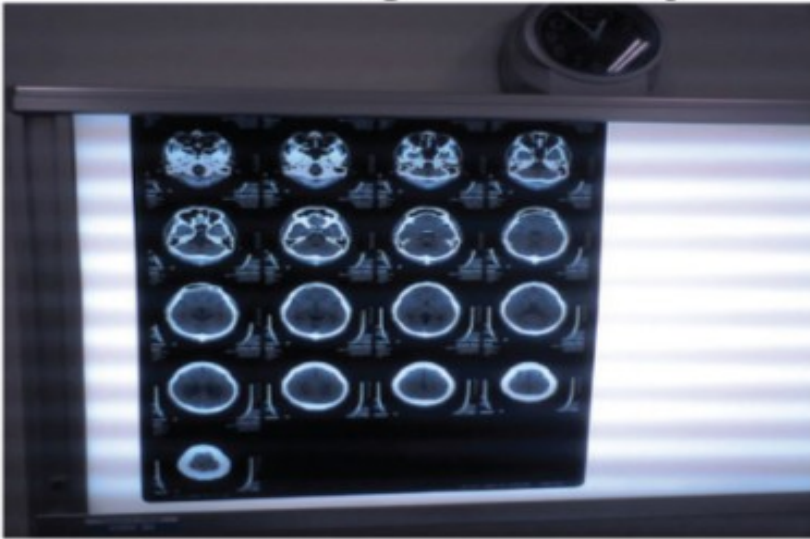
- Internet applications

- Tagging:
face
detect,
recognize



Motivation

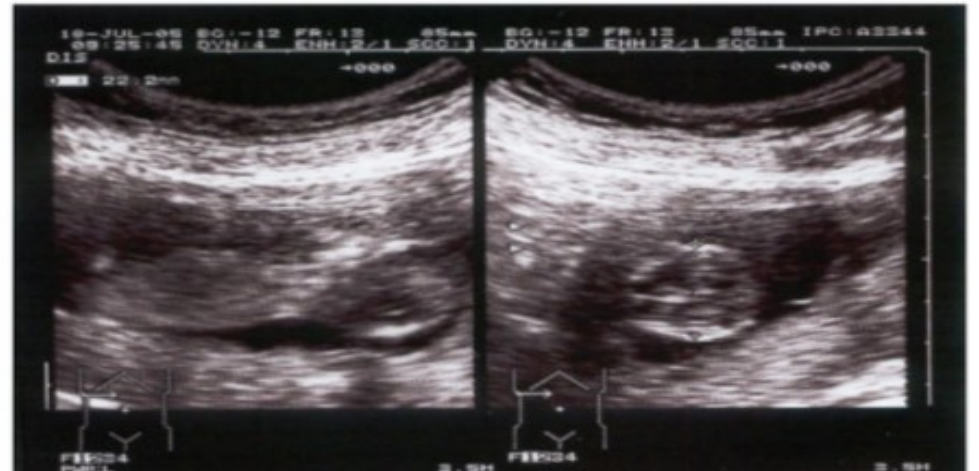
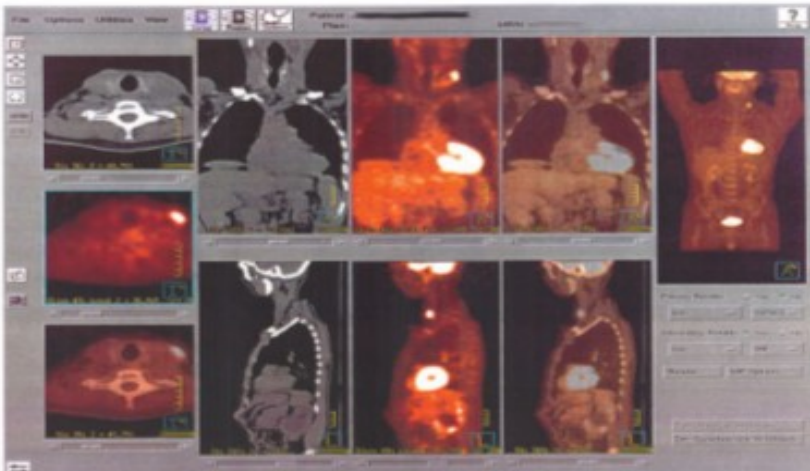
- Medical image analysis



(a)



(b)



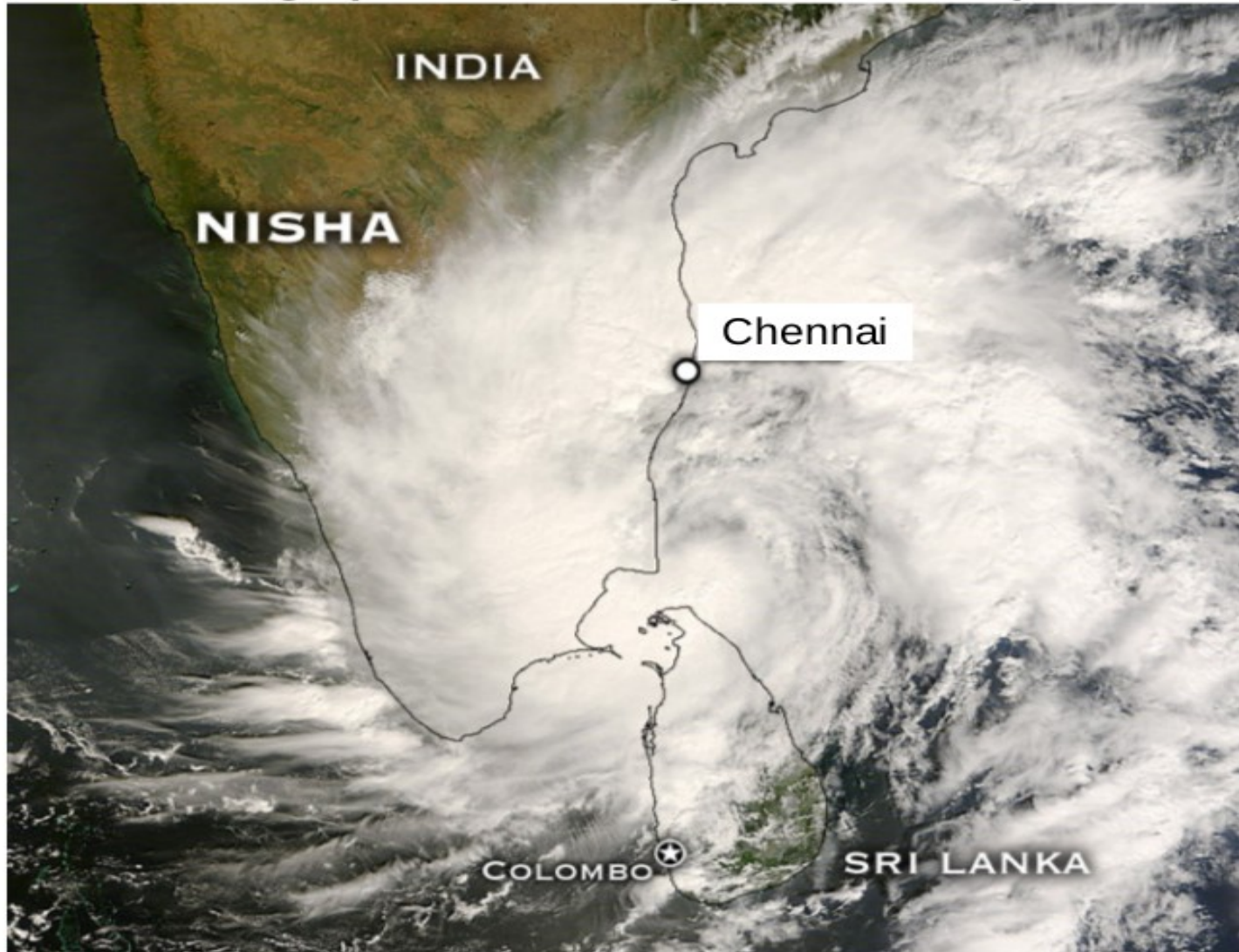
Motivation

- Remote sensing (classify: water, forested, urban, ...)



Motivation

- Remote sensing (weather prediction)



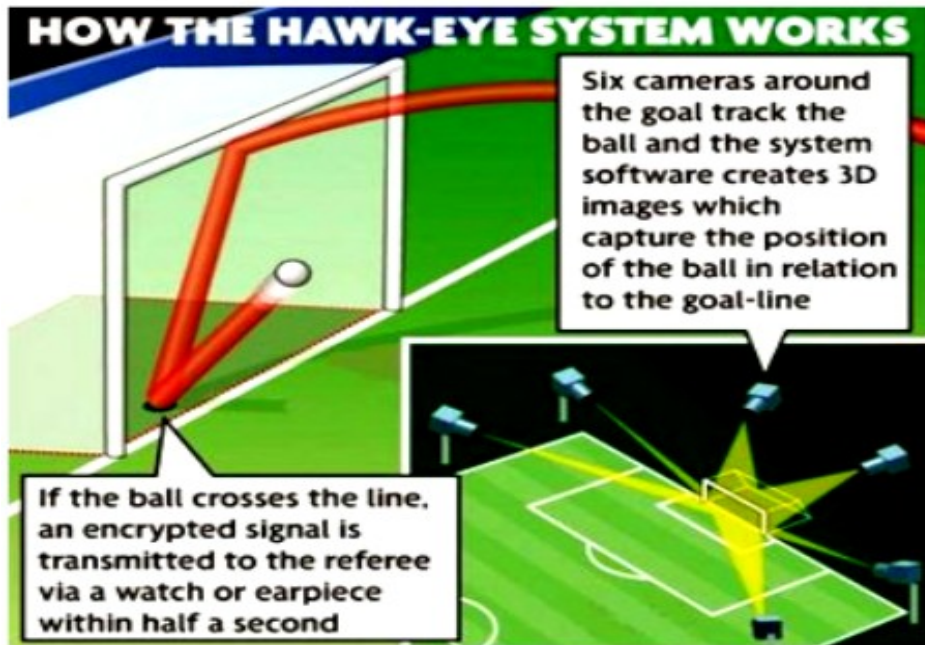
Motivation

- Surveillance



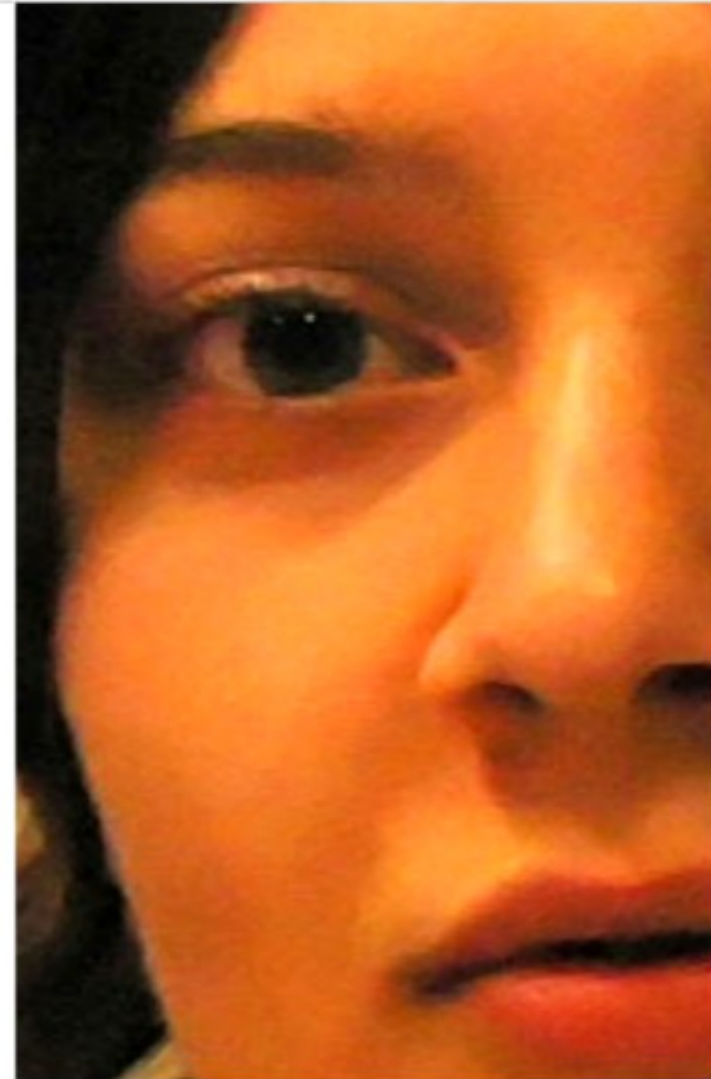
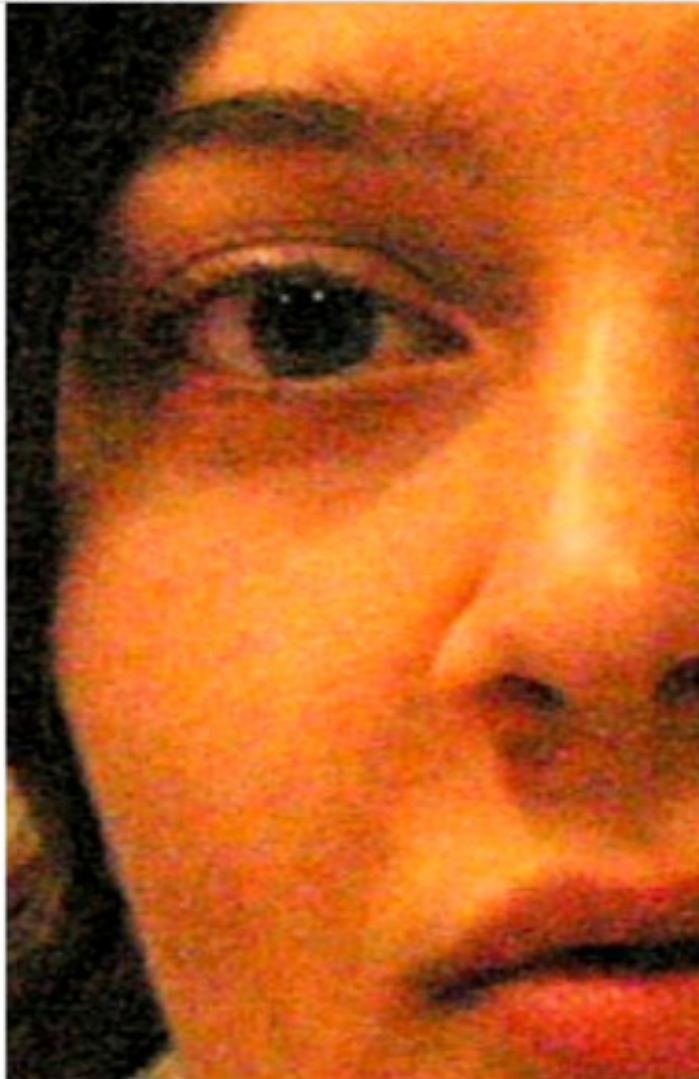
Motivation

- Sports



Motivation

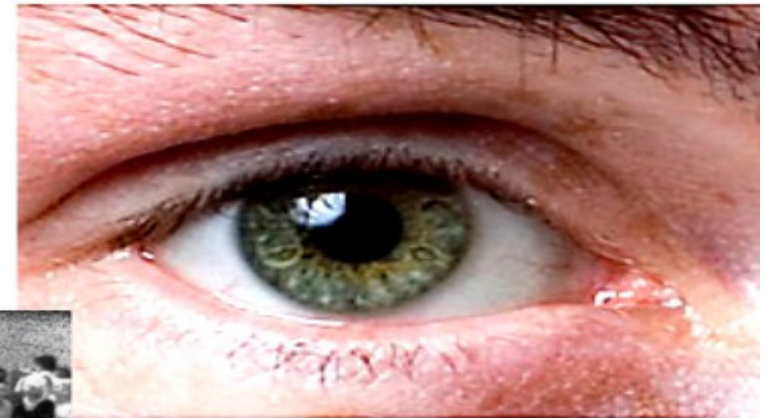
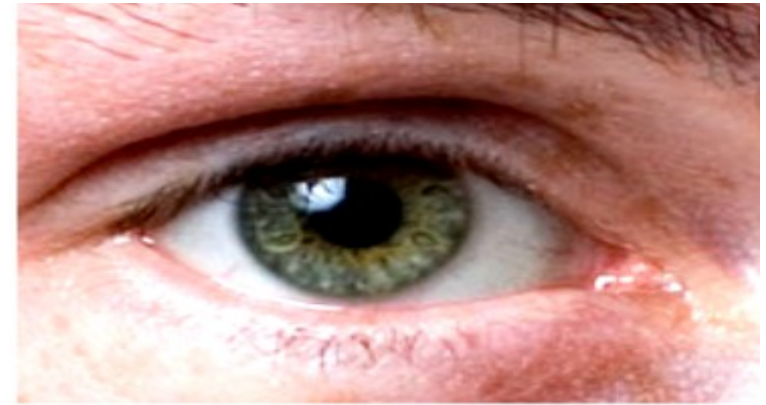
- Image restoration
 - Denoising



Motivation

Applications

- Image enhancement
 - Contrast increase
 - Sharpen



Let's Get Started

- We will talk about digital images here.
- We will be using OpenCV as our toolkit. So along with the concepts I will start referring the OpenCV functions and terminologies as well.
- We will have a demo at the end of the session.

Digital Images

- Digital (Discrete) Image

- **Array / grid** of numbers

- Integers, Real
- Signed, Unsigned (non negative)
- Many dimensions

- What makes data discrete ?

- Acquisition

- e.g., charge-coupled device (CCD) array in camera

- Representation

- e.g., digitization : analog print → scanned digital copy

- *Mat* is a data structure to represent such discrete 2D arrays

8 rows	52	55	61	66	70	61	64	73
	63	59	55	90	109	85	69	72
	62	59	68	113	144	104	66	73
	63	58	71	122	154	106	70	69
	67	61	68	104	126	88	68	70
	79	65	60	70	77	68	58	75
	85	71	64	59	55	61	65	83
	87	79	69	68	65	76	78	94
8 columns								

Contrast and Brightness

Two commonly used point processes are *multiplication* and *addition* with a constant

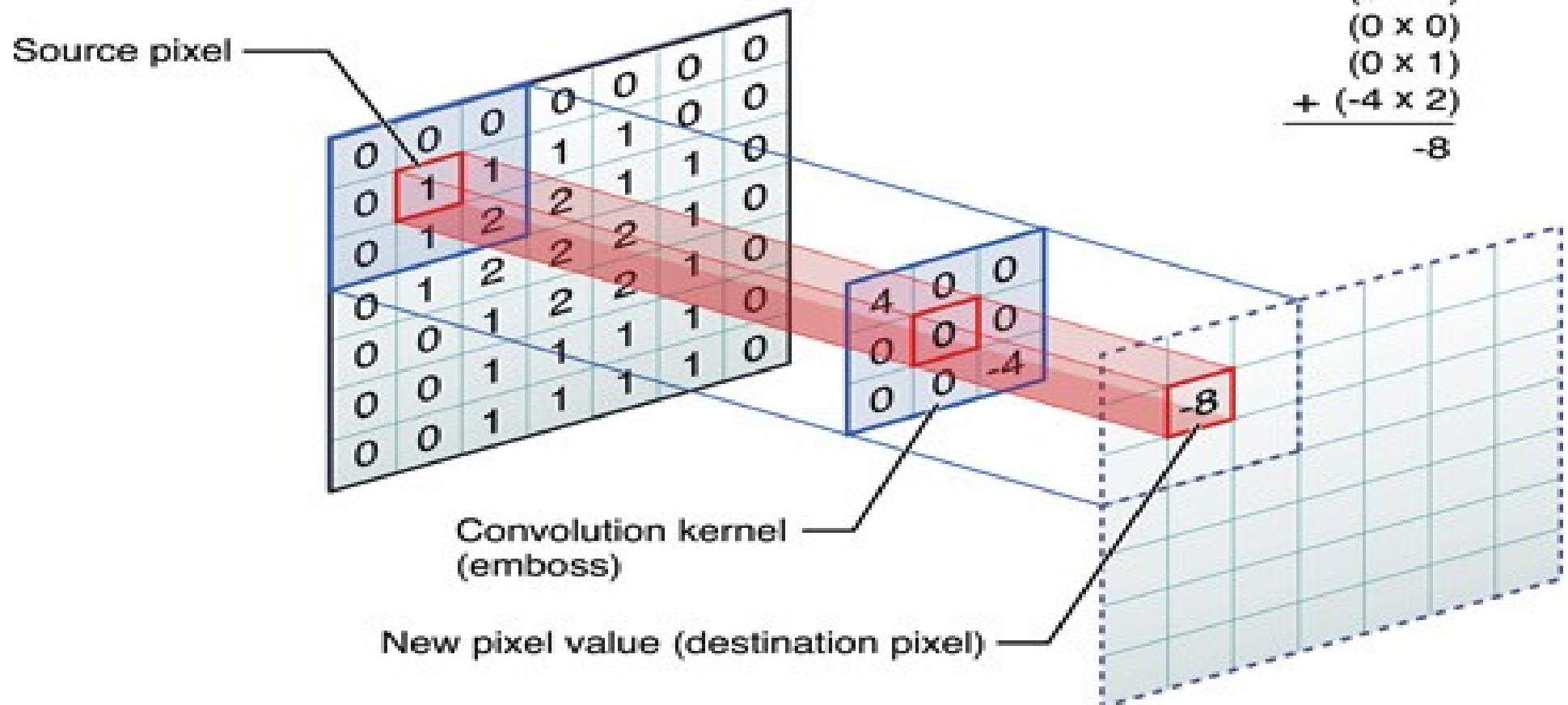
$$G(x,y) = AF(x,y) + B$$

The parameters $A > 0$ and B control the *contrast* and *brightness* parameters.

Here $F(x,y)$ is the input image and $G(x,y)$ is the output image.

Kernal Operations: Convolution

Center element of the kernel is placed over the source pixel. The source pixel is then replaced with a weighted sum of itself and nearby pixels.



$$\begin{array}{r} (4 \times 0) \\ (0 \times 0) \\ (0 \times 0) \\ (0 \times 0) \\ (0 \times 1) \\ (0 \times 1) \\ (0 \times 0) \\ (0 \times 1) \\ + (-4 \times 2) \\ \hline -8 \end{array}$$

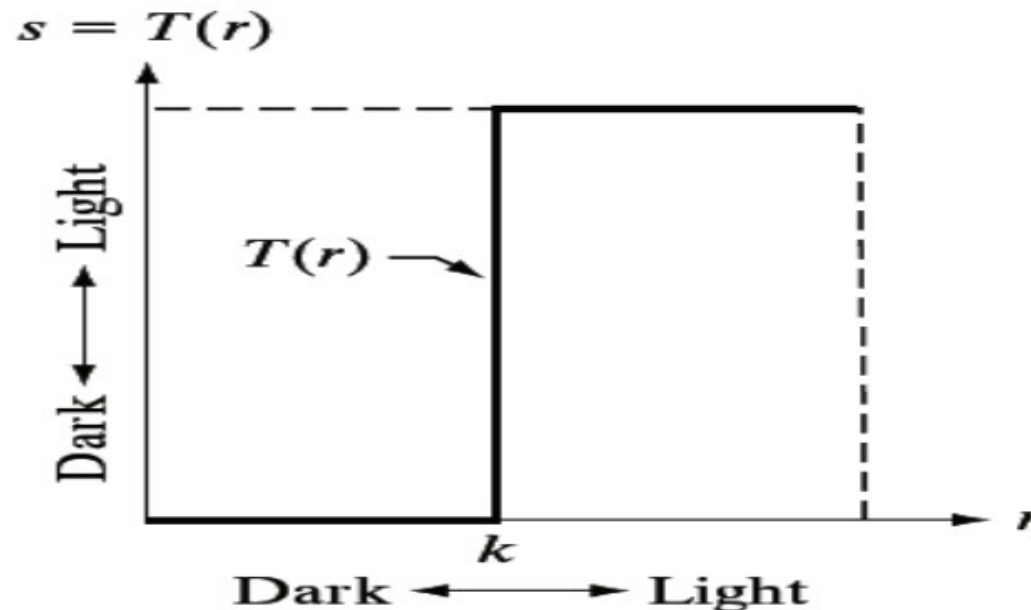
Operation1: Blurring

- Blurring or smoothing is one of the most common operations on image using kernel convolution.
- Average blur, Gaussian Blur, Median Blur are few commonly used blur kernels.

$$K = \frac{1}{K_{\text{width}} \cdot K_{\text{height}}} \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & 1 & 1 & \dots & 1 \\ \cdot & \cdot & \cdot & \dots & 1 \\ \cdot & \cdot & \cdot & \dots & 1 \\ 1 & 1 & 1 & \dots & 1 \end{bmatrix}$$

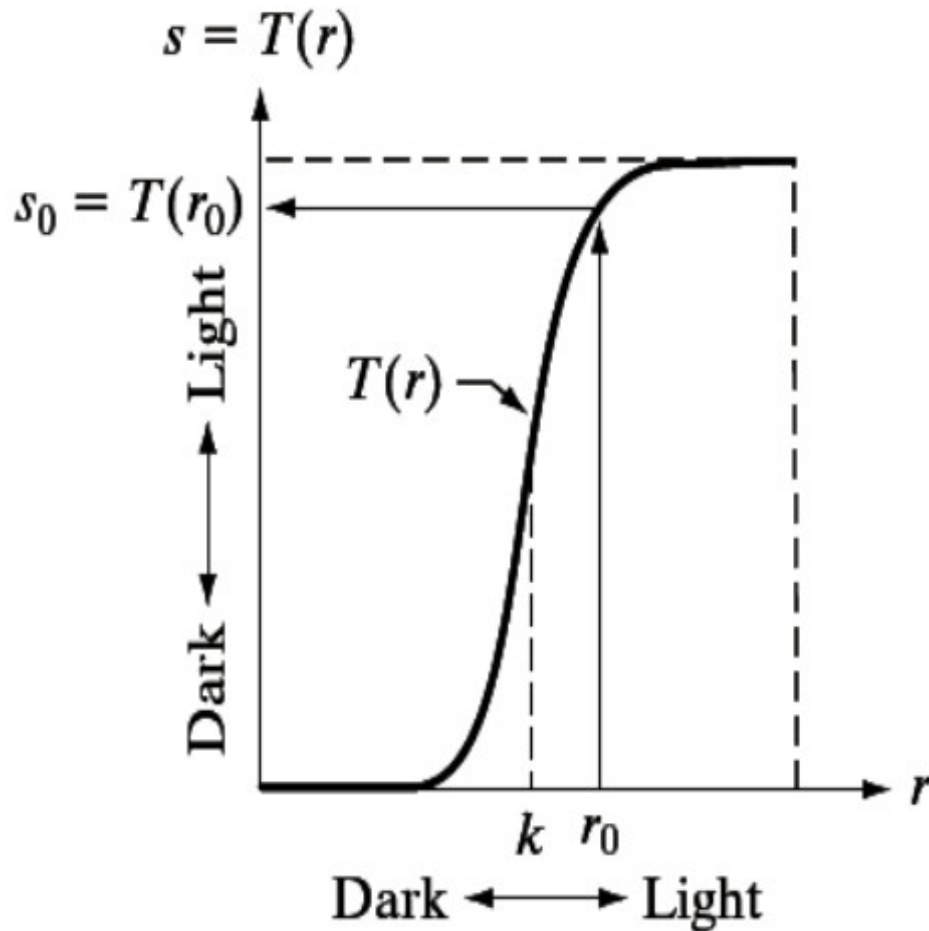
Operation2: Thresholding

- Binary and Inverted Binary are two common thresholding available in OpenCV, there is a function named threshold. Thresholding can be understood by this graph.



Operation3: Contrast Stretching

- Contrast stretching



Original



Pixel range: 79-136

Contrast Stretched



Pixel range: 0-255

Operation4: Edge Detection

- There are many edge detection algorithms, most popular are Sobel, Laplace and Canny, these three have their functions predefined in OpenCV. The basic concept is the same.
- When we take the derivative of the image we can find spikes at edges as over there we get a sharp change in intensity.

$$G_x = \begin{bmatrix} -1 & 0 & +1 \\ -2 & 0 & +2 \\ -1 & 0 & +1 \end{bmatrix}$$
$$G_y = \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ +1 & +2 & +1 \end{bmatrix}$$

$$G = |G_x| + |G_y|$$

Edge Detection



Canny



Roberts



Prewitt



Sobel

Any Questions??

Histogram Processing: 1

- For image enhancement, histogram based methods are really good.
- We will look at a really common and simple method called histogram equalization. In OpenCV we have a function as *equalizeHist(src, dst);*
- What Histogram Equalization does is to stretch out this range. It takes the values from populated areas and give it to unpopulated ones in the spatial(intensity) domain.
- For working:

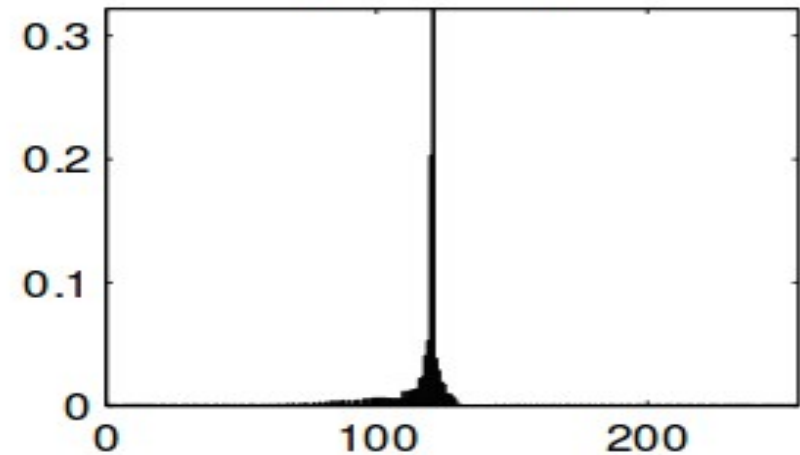
http://en.wikipedia.org/wiki/Histogram_equalization

Histogram Processing: 2

original image



original histogram



transformed image



transformed histogram

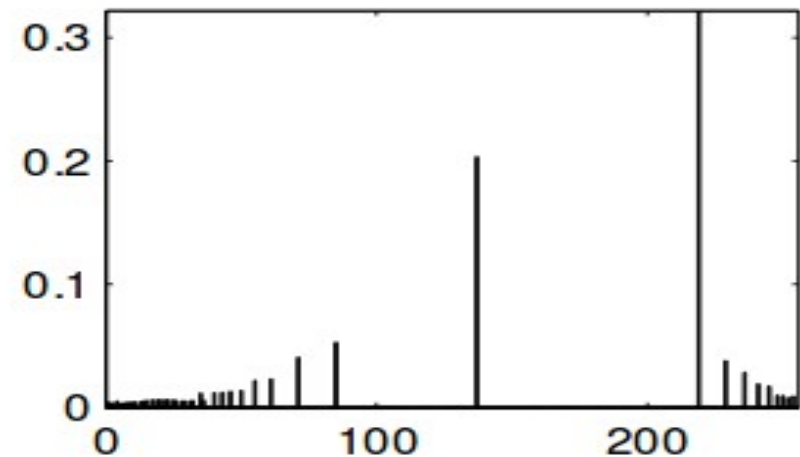


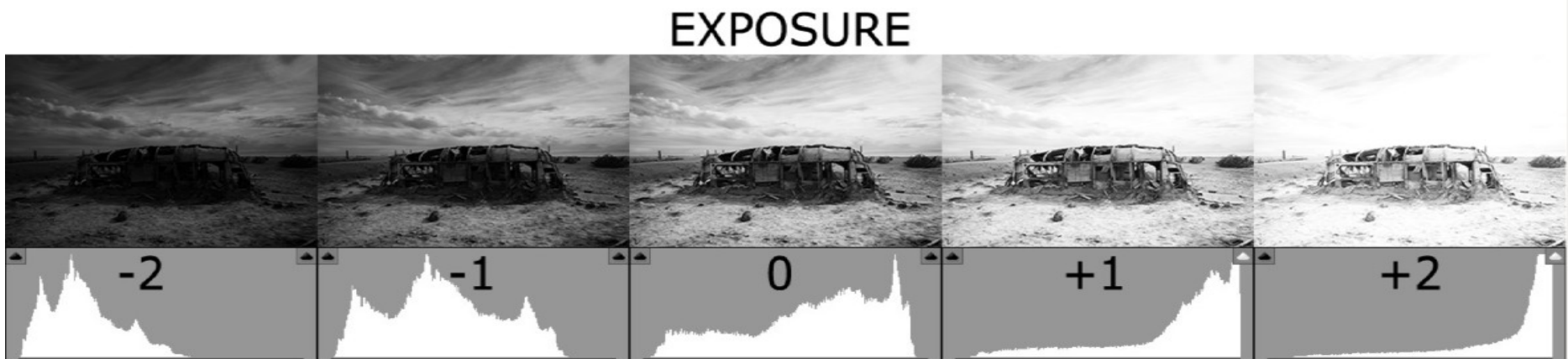
Figure 1: Histogram equalization applied to low contrast image

Histogram Processing: 3

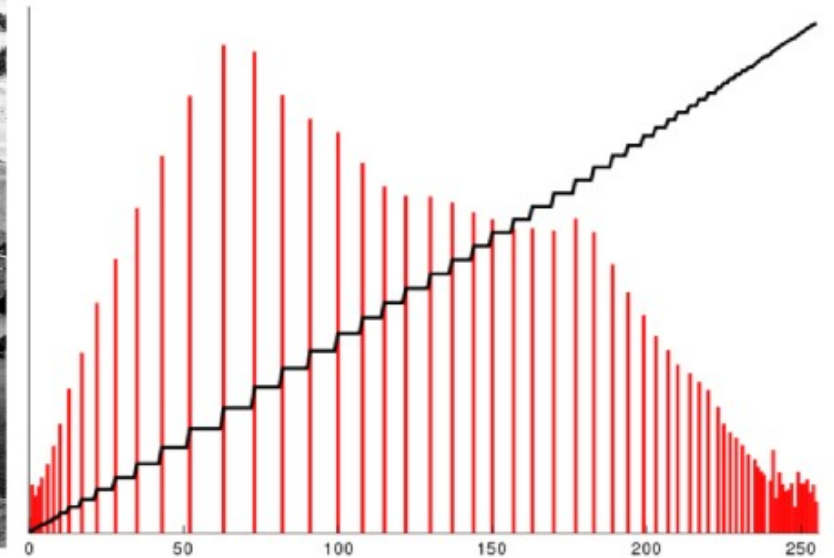
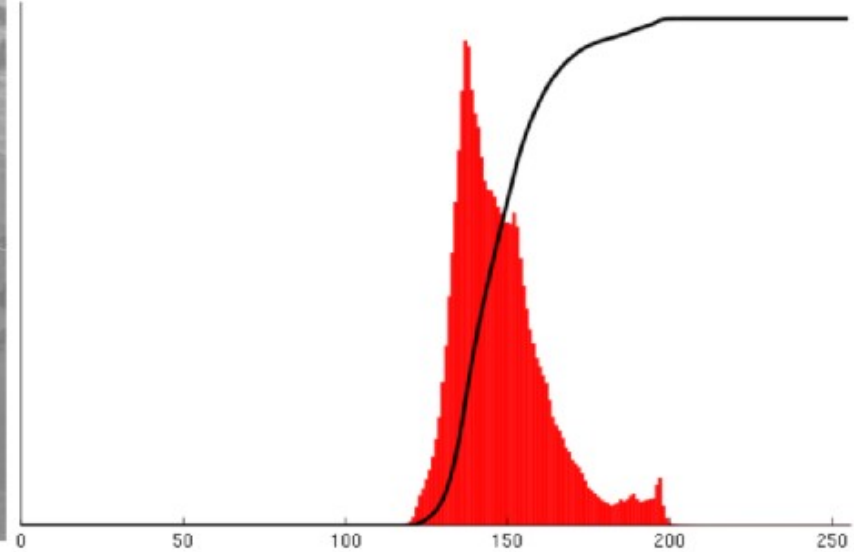
- Adaptive Histogram Equalization (AHE) and Contrast Limited Adaptive Histogram Equalization (CLAHE) are better versions of this algorithm. If interested you can have a look here.

http://en.wikipedia.org/wiki/Adaptive_histogram_equalization

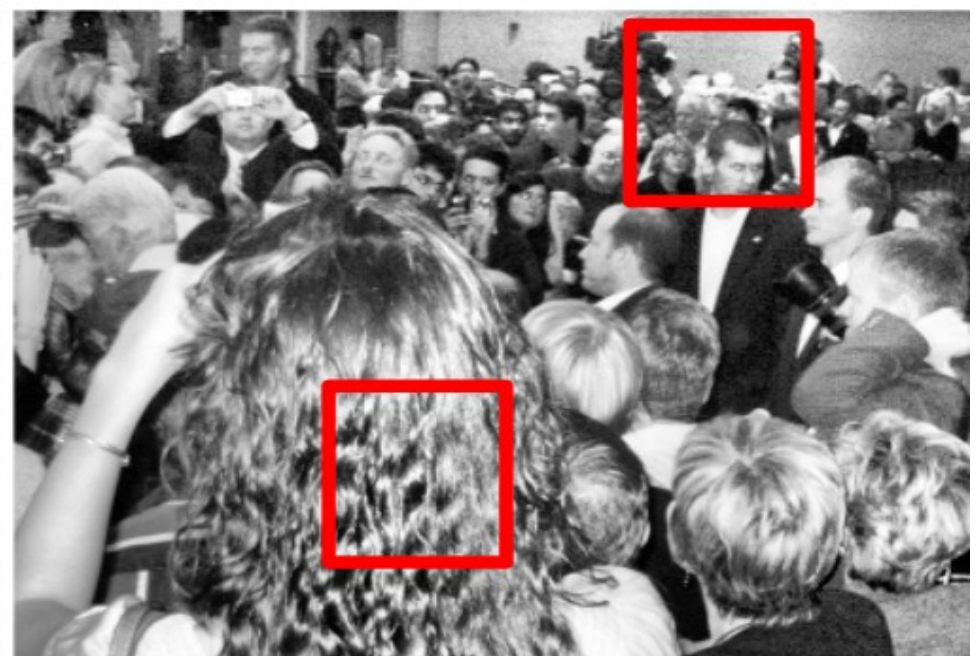
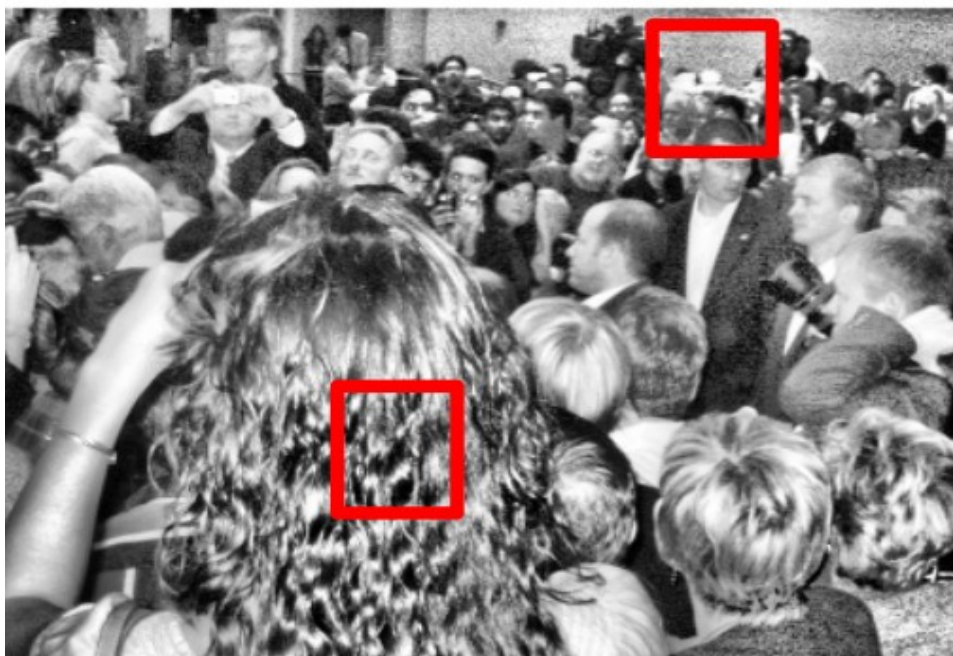
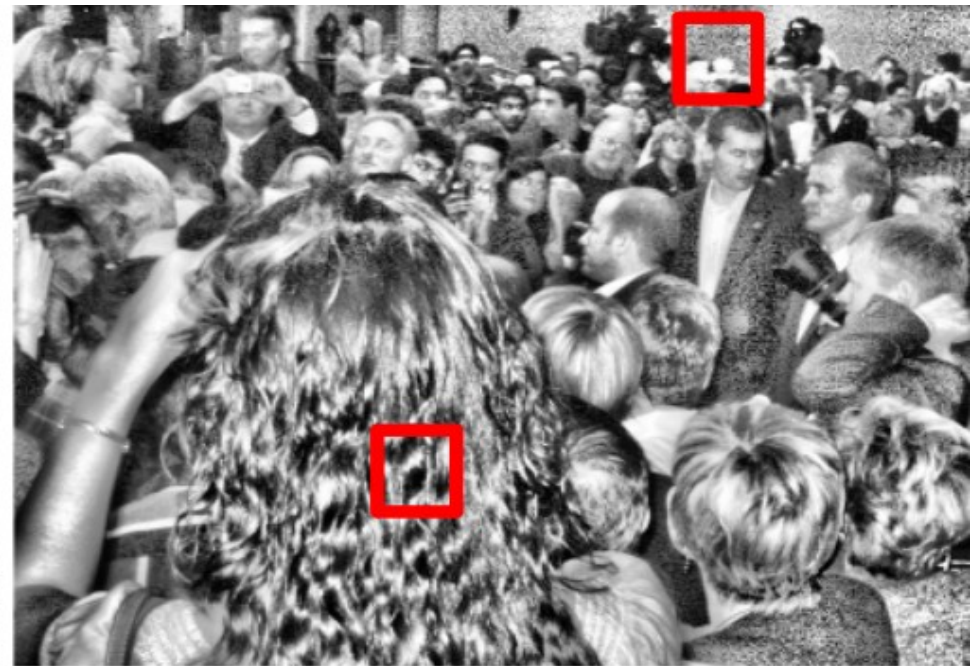
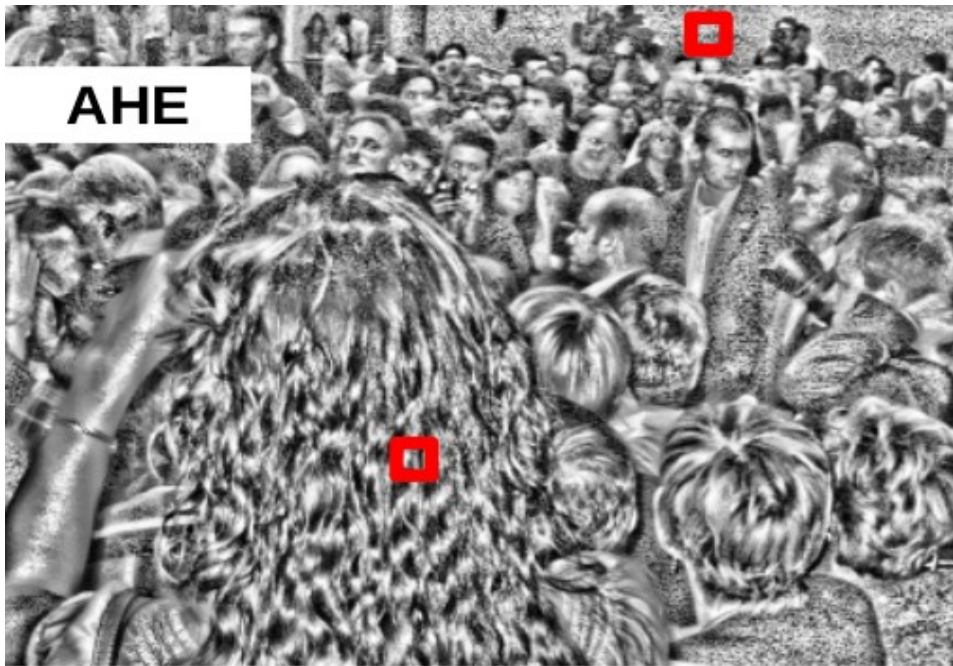
- Histograms are related to photograph exposure
 - Aperture size (amount of light per unit time)
 - Shutter speed (time of exposure)



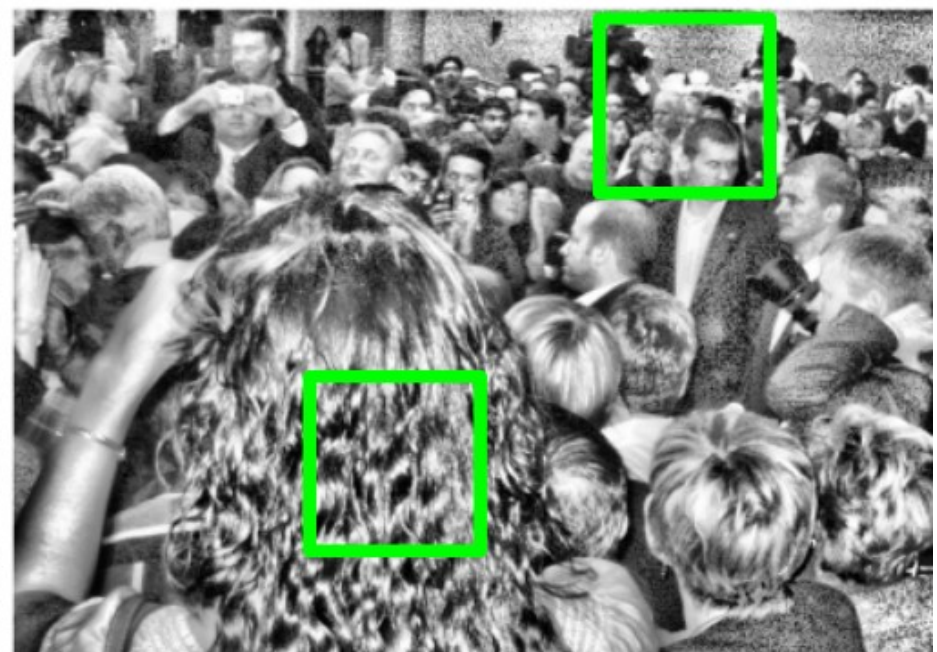
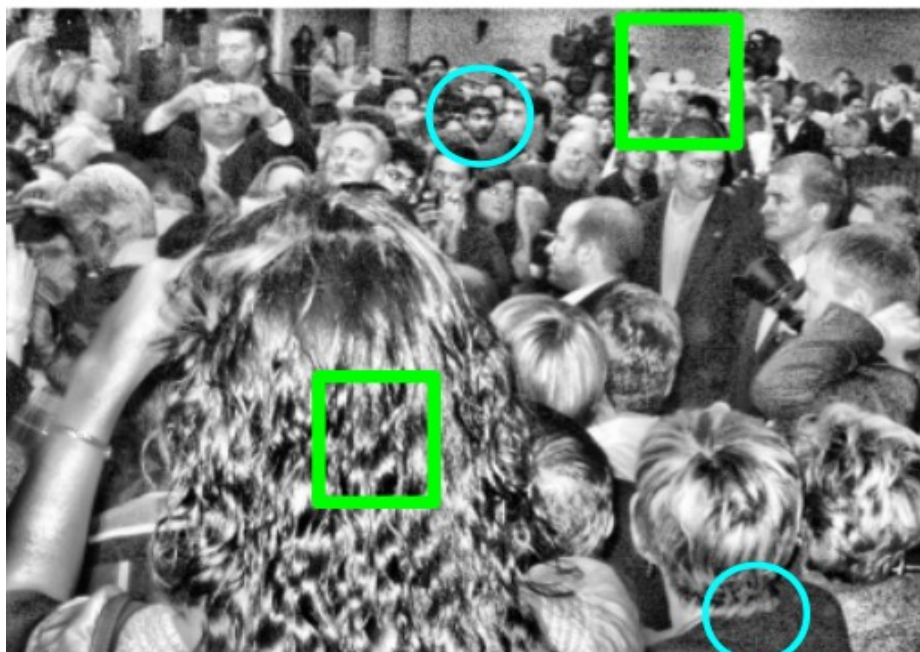
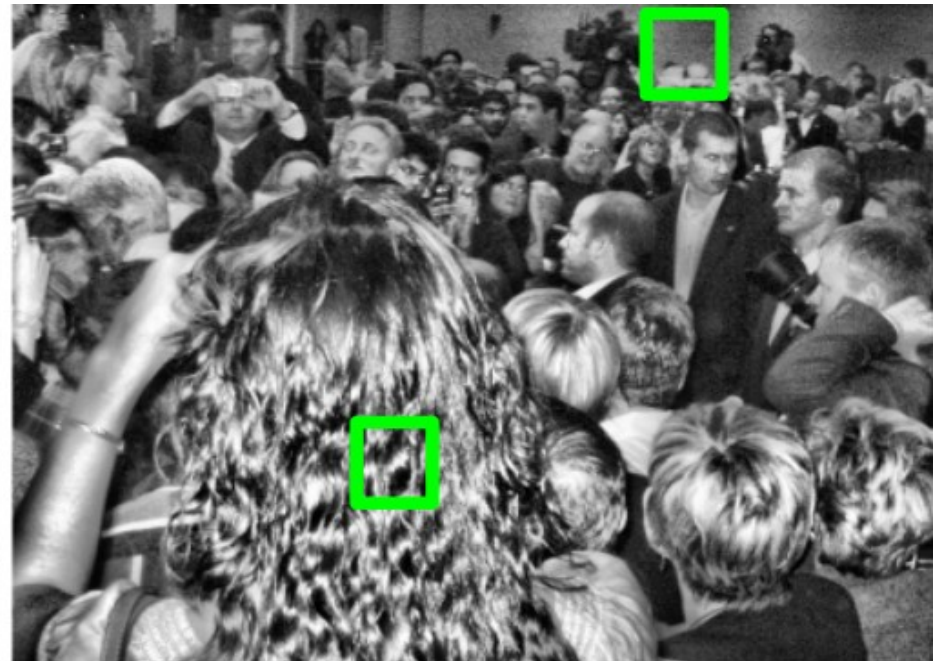
Histogram Processing: 4



Histogram Processing: 5



Histogram Processing: 6



Any Questions??

Fuzzy!!

- I like to cover fuzzy based image enhancement, so if everyone is willing to do this we will do it.
- If we are doing it we can do it after the demo of before it.

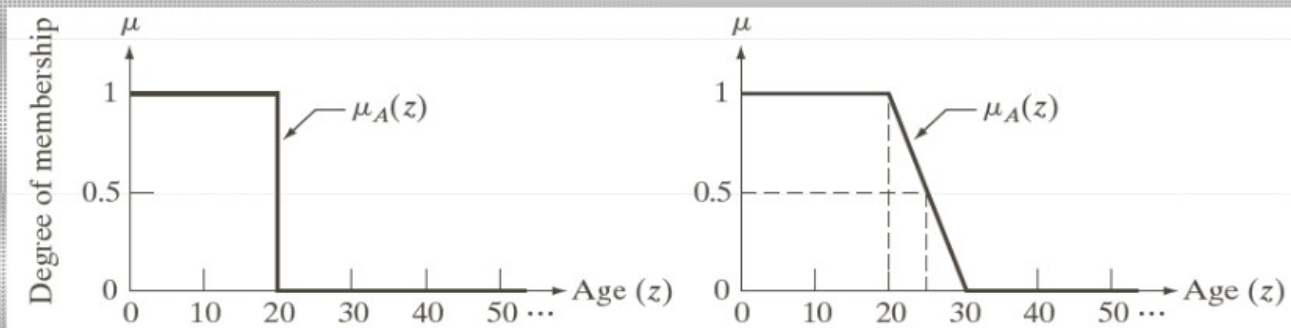
Fuzzy Logic

- We will perform contrast enhancement using this method.
- It's a set theory method for expressing data more reliably.

A fuzzy set is defined in terms of a membership function $\varphi_A : U \rightarrow [0,1]$.

A *characteristic function* is a special case of a membership function and a regular set is a special case of a fuzzy set.

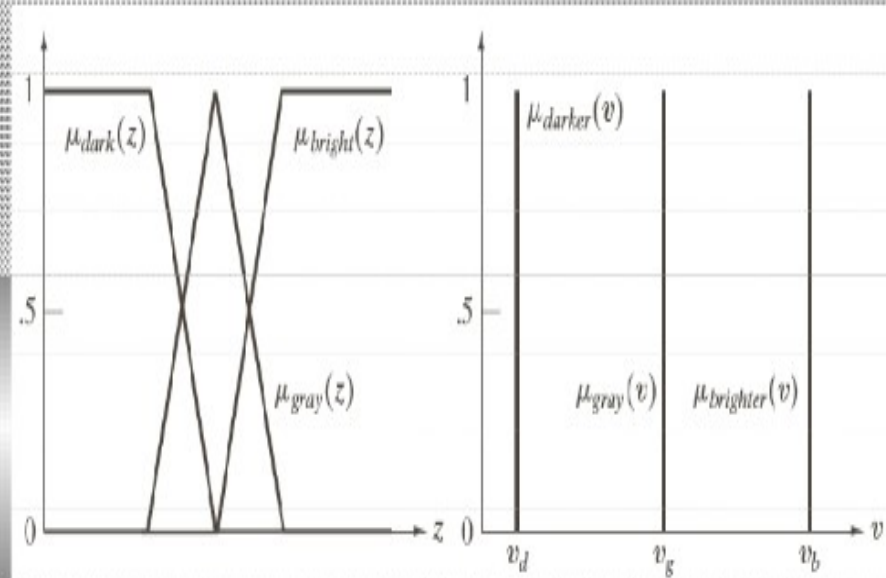
Illustrating The membership functions of regular and fuzzy set



Fuzzy: Contrast Enhancement 1

Using fuzzy sets for intensity transformation

1. Define a set of rules to change pixel intensity.
2. Transfer the rules into fuzzy set
3. Use the rules to change intensity



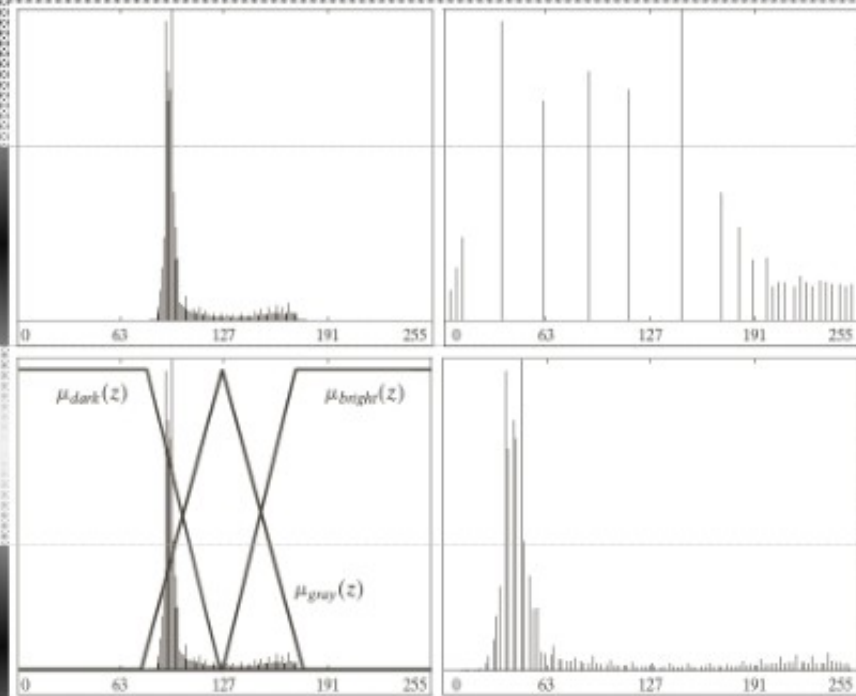
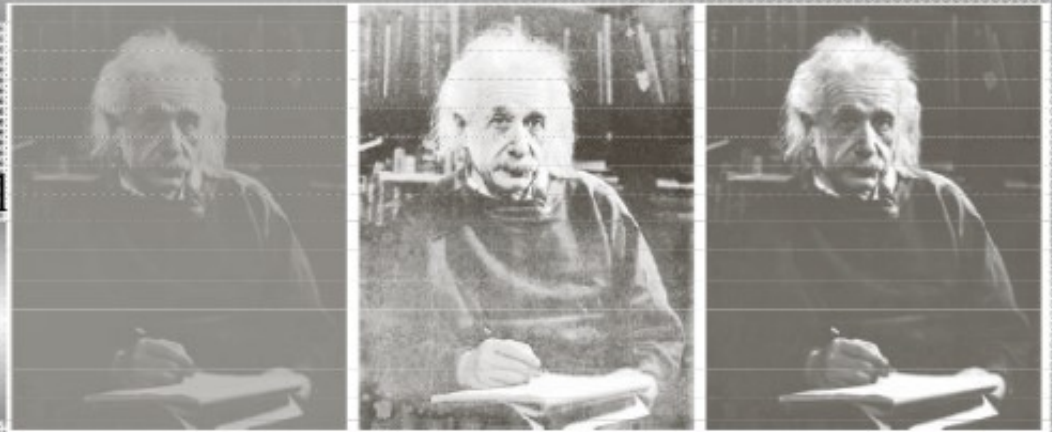
Example:

1. If a pixel is dark, then make it darker
2. If a pixel is gray, then make it gray
3. If a pixel is bright, then make it brighter

Fuzzy: Contrast Enhancement 2

Using fuzzy sets for intensity transformation

1. Define a set of rules to change pixel intensity.
2. Use fuzzy set to apply this rules.



Things For Demo : 1

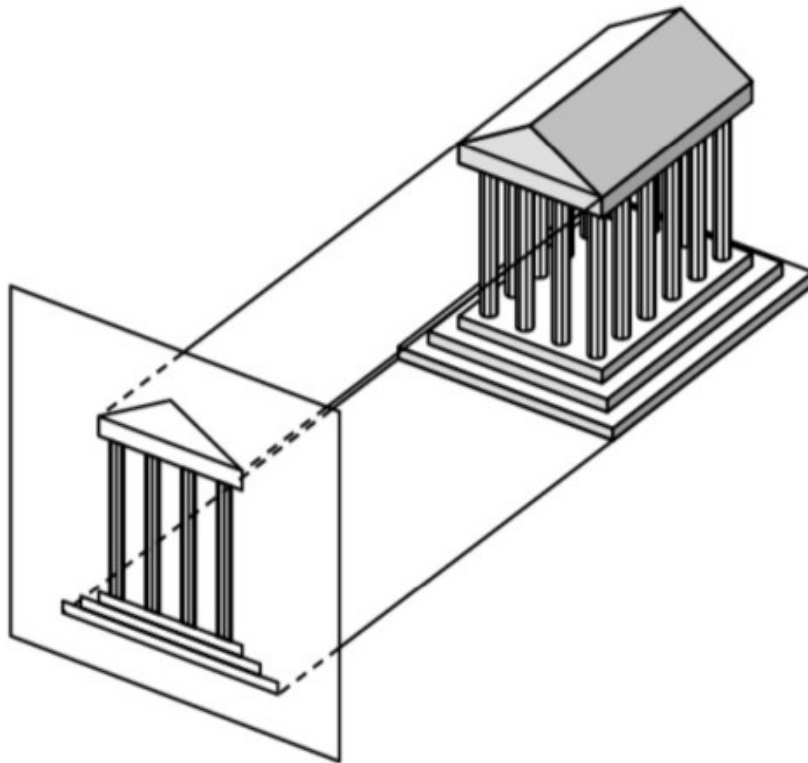
- Basic definition is given by the function

$$M_{ij} = \sum_x \sum_y x^i y^j I(x, y)$$

- Area (for binary images) or sum of grey level (for greytone images): M_{00}
- Centroid: $\{ x, y \} = \{ M_{10}/M_{00}, M_{01}/M_{00} \}$
- Useful in tracking applications, how to use moment functions will be explained in demo.

Things For Demo : 2

Orthographic or Parallel Projection



<http://www.cs.cmu.edu/afs/cs/academic/class/15462-s09/www/lec/06/lec06.pdf>

Pendulum

- The folder given to you will have a video of a makeshift pendulum that I recorded yesterday in orthographic projection. We will find the length of this pendulum and hence the time period.
- Given:
 - > 4 screenshots of the video instances.
 - > The breadth of the image is 480 in pixels and 11 inches in world (see the scale in the side of the video).

Pendulum

- Flow of the code:
 - > We read the 4 screenshots.
 - > We process each of them sequentially with techniques we have learned in a logical fashion.
 - > We get the center of the hanging ball in each image say a_1, a_2, a_3, a_4 .
 - > use SOT to find r for each 4C3 combinations and average to get r_{avg} in pixels.
 - > then use $11\text{inches}/480\text{ pixels} = \text{length}/r_{avg}$.
 - >Voila!

Thank You