

ME5411

ROBOT VISION AND AI

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Lecture 3 (Part1)

Image Preprocessing



Topics

- Definitions
- Gray Level Transformation
- Histogram Equalization
- Geometric Transformation
- Noise and Filtering



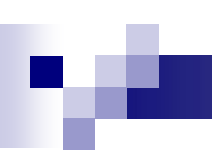
Key takeaways

- Gray level transforms: negative, thresholding, contrast stretch
- Histogram equalization: enhances contrast
- Geometric transforms: pixel mapping + interpolation
- Noise types: Gaussian, impulse, quantization
- Filtering: smoothing and edge detection



1. Definitions

- Image pre-processing refers to operations with images at the lowest level of abstraction—both input and output are intensity images.
- The aim of pre-processing is an improvement of the image data that suppresses unwelcome distortions or enhances some image features important for further processing.
- Hence, image pre-processing can be classified into image enhancement and image restoration.

- 
- After image enhancement, the result is more suitable than the original image for a specific application.
 - Problem oriented
 - Two categories of approaches
 - Spatial Domain Methods
 - Spatial domain refers to the image plane itself
 - Direct manipulation of pixels in the image
 - Get to see the results after processing
 - Easy to implement but can be computational intensive
 - Frequency Domain Methods
 - Processing techniques are based on modifying the Fourier Transform of an image
 - Effective, quite imaginative as you do not see the results immediately after processing
 - Image restoration takes a corrupt/noisy image and estimates the clean, original image.

Spatial Domain Methods

- Spatial domain methods operate directly on the pixels composing an image.
- Image processing function:

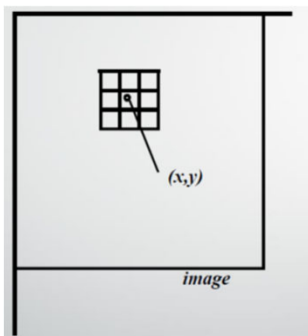
$$g(x, y) = T[f(x, y)]$$

$f(x, y)$: input image

$g(x, y)$: processed or output image

T : transfer function - an operator on $f(x, y)$ defined over neighborhood of (x, y)

T can also operate on a set of input images (e.g. pixel-by-pixel averaging for noise removal)



Neighborhood about (x, y) is usually a square or rectangular area centered at (x, y) . The centre of this sub-image moves from pixel to pixel during the operation. The operator is applied at each location to produce $g(x, y)$.

Point Processing

- Point processing is a type of image enhancement, also known as pixel brightness transformation.
- Neighborhood is 1×1 == the size of pixel neighbor is 0. Point processing operates only one pixel.
- Transformation is defined pixel by pixel.
- The transformation T has been referred to as gray-scale transformation or gray-level mapping function.

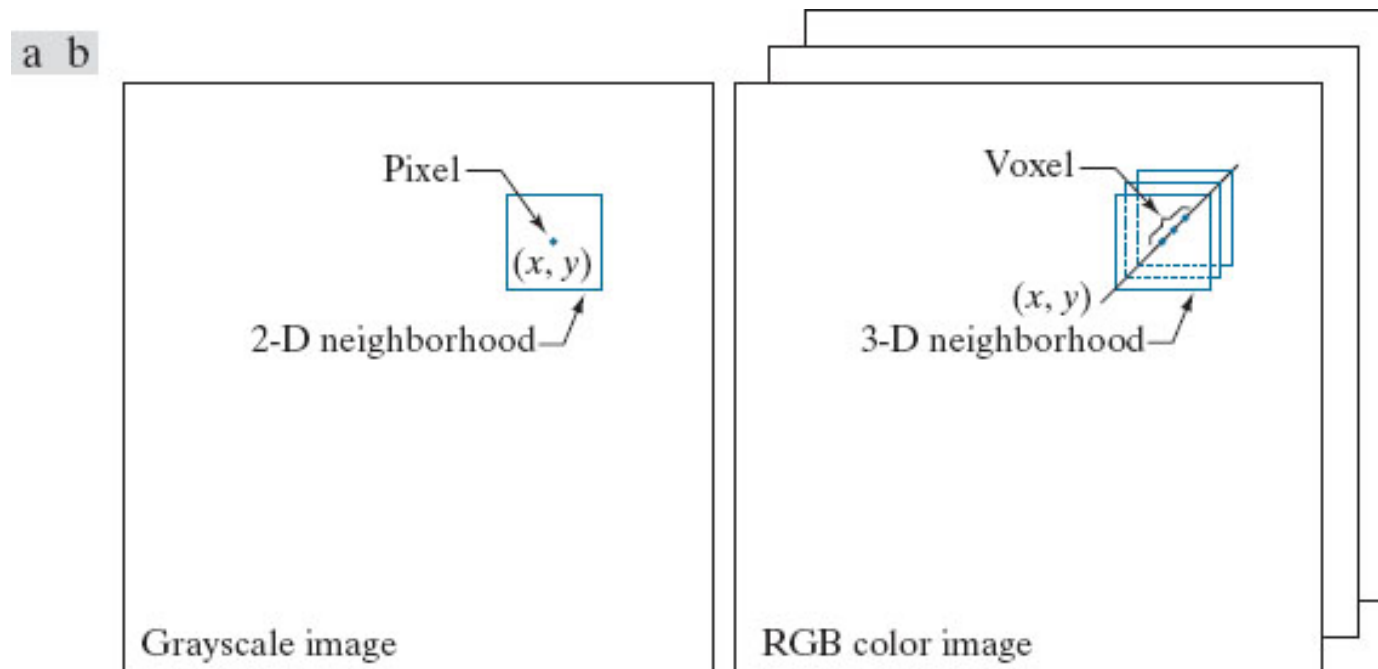
$$s = T(r)$$

r and s are the gray levels of $f(x,y)$ and $g(x,y)$ at (x,y) respectively.

- It changes the brightness of a pixel with no consideration of the pixel position.
- The mapping function can be specified in different ways, such as piecewise linear function, or based on the histogram of the input image.

Point Processing

- Color: $p(i,j) = [R(i,j) \ G(i,j) \ B(i,j)]$
 - RGB/8: $R(i,j) = 0..255$, $G(i,j) = 0..255$, $B(i,j) = 0..255$.
 - 8 bit of data per color channel.



Spatial neighborhoods for grayscale and RGB color images. Observe in (b) that a single pair of spatial coordinates, (x,y) , addresses the same spatial location in all three images.

Source: Digital Image Processing By Gonzalez and Woods, Pearson.



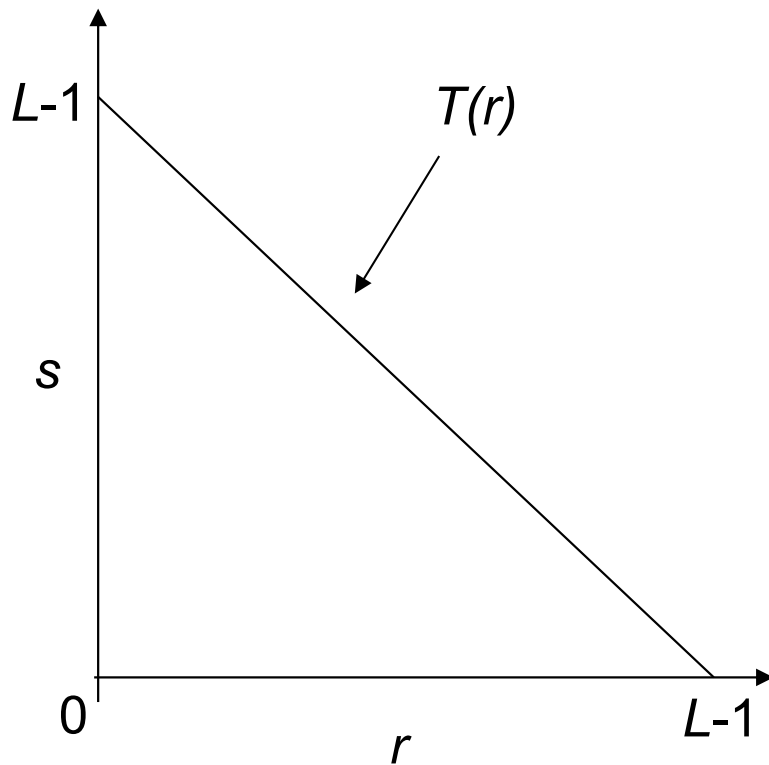
2. Gray-Scale Transformation

- The most common gray scale transformation are
 - Negative transformation
 - Brightness thresholding that results in a black and white image
 - Piecewise linear function that enhances the image contrast between two specific brightness values.

Negative Transformation

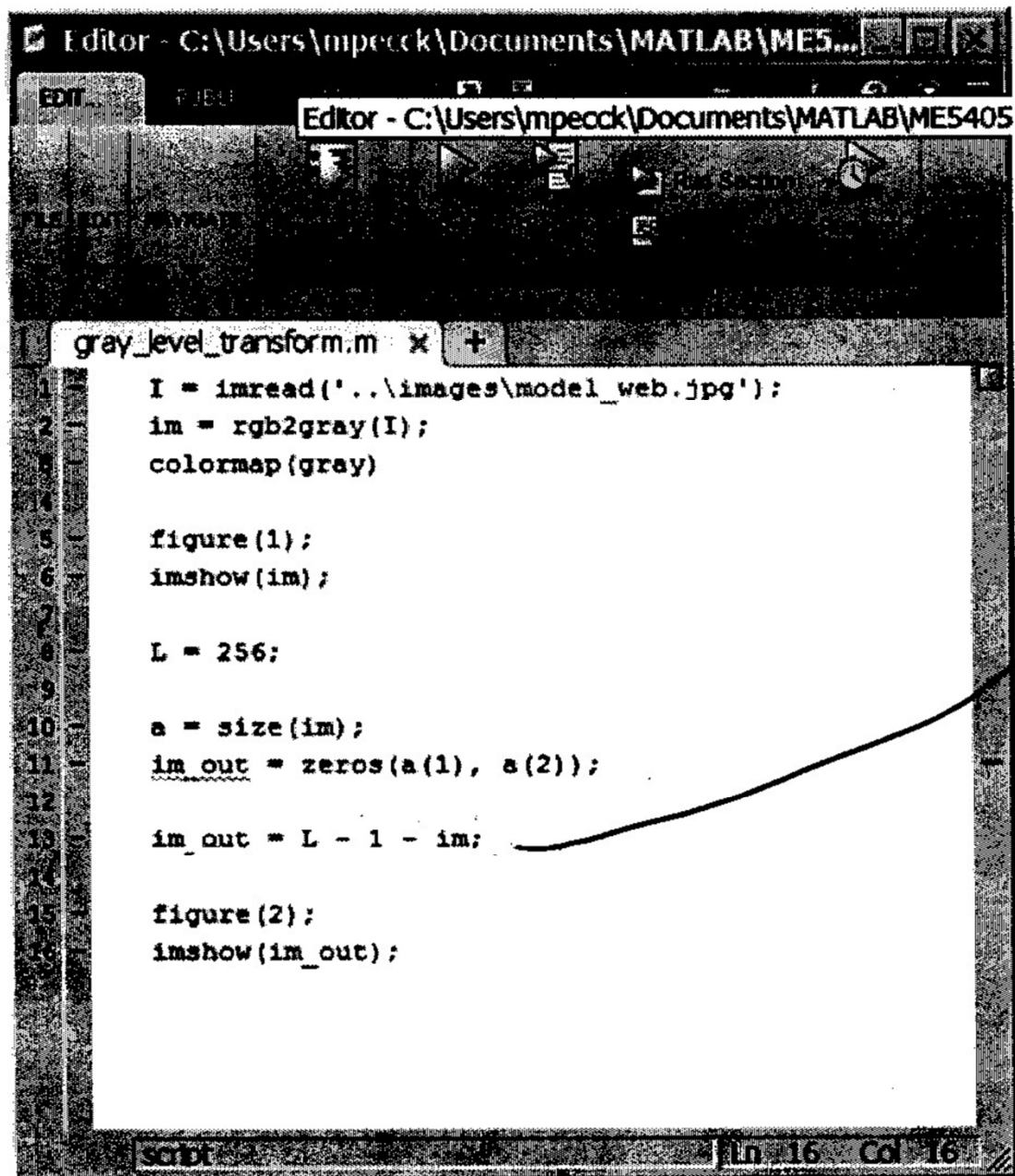
- Also known as negative image. It inverts the “color” of an image.

$$s = T(r) = L - 1 - r$$



$p(i,j)$

$T(p(i,j))$



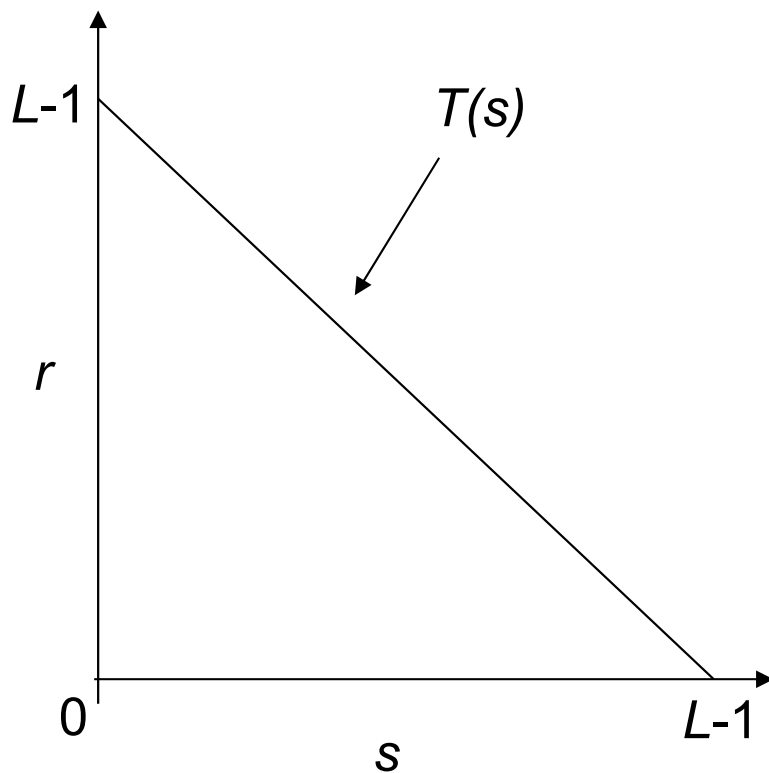
```
1 I = imread('..\images\model_web.jpg');
2 im = rgb2gray(I);
3 colormap(gray)
4
5 figure(1);
6 imshow(im);
7
8 L = 256;
9
10 a = size(im);
11 im_out = zeros(a(1), a(2));
12
13 im_out = L - 1 - im;
14
15 figure(2);
16 imshow(im_out);
```

The image shows a MATLAB Editor window with a script named `gray_level_transform.m`. The script performs the following steps: reads an image, converts it to grayscale, displays it in a figure, and then performs a level transformation. A line from the script points to a callout box.

```
for i = 1:a(1)
    for j = 1:a(2)
        im_out(i,j) = L-1-im(i,j)
    end
end
```

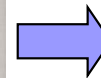
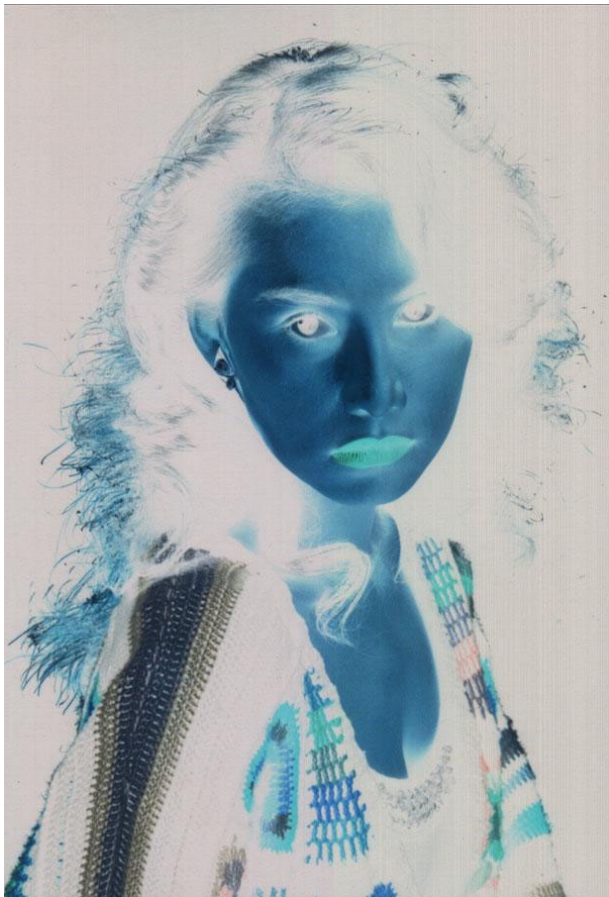
Negative Transformation

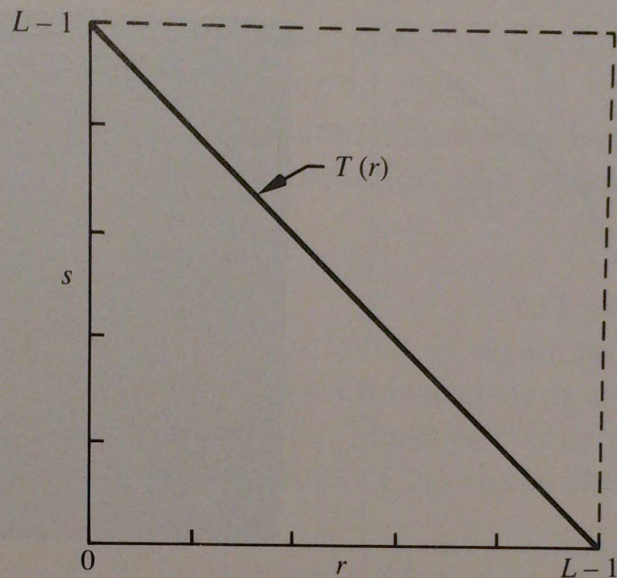
- Can we recover the original image by inverting the gray scale of a negative image?



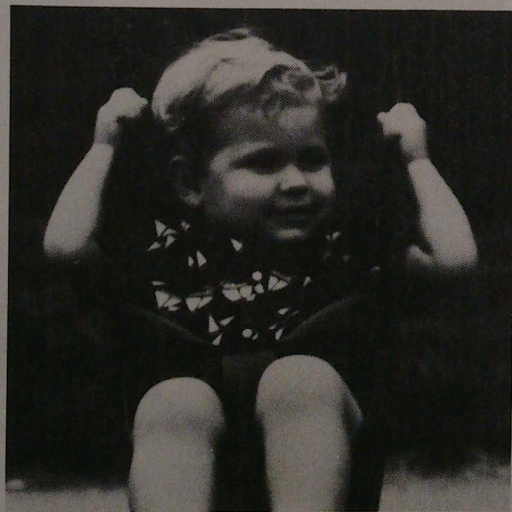
Negative Transformation

- Example – invert colors of a color image





(a)



(b)



(c)

Obtaining the negative of image:

- (a) gray-level transformation function, r and s denote the input and output gray levels respectively;
- (b) an image;
- and (c) negative of the image.

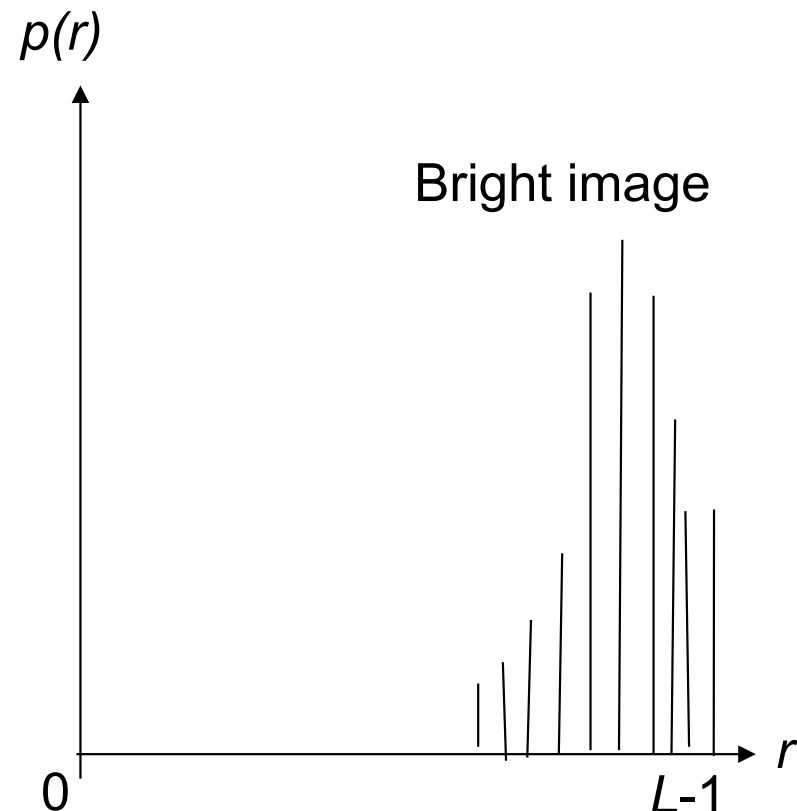
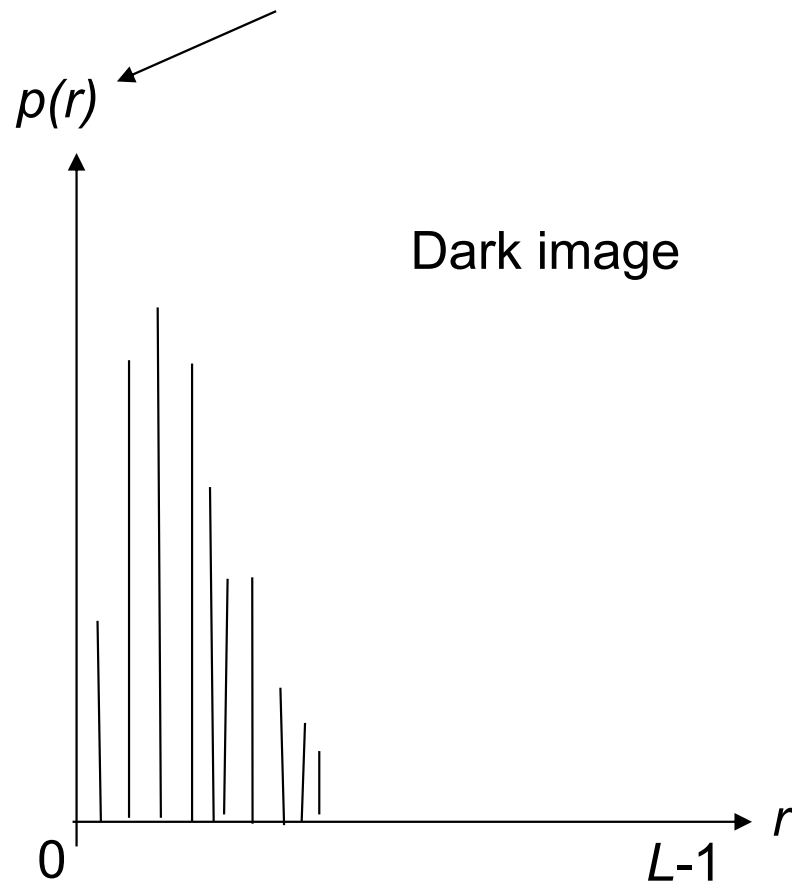
Histogram

- The histogram of an image shows the distribution of the pixel values in the image over the intensity (or dynamic) range.
- For a 8-bit gray scale image, the pixel values typically from 0 to 255.
- The i th item of the histogram is $p(i) = n_i/N$, $i = 0..255$.
- It represents the probability of the a randomly chosen pixel has the gray level i , where n_i is the number of pixels of gray level i , and N is the total number of pixels in the image. (probability density function).

Histogram: Dark and Bright Images

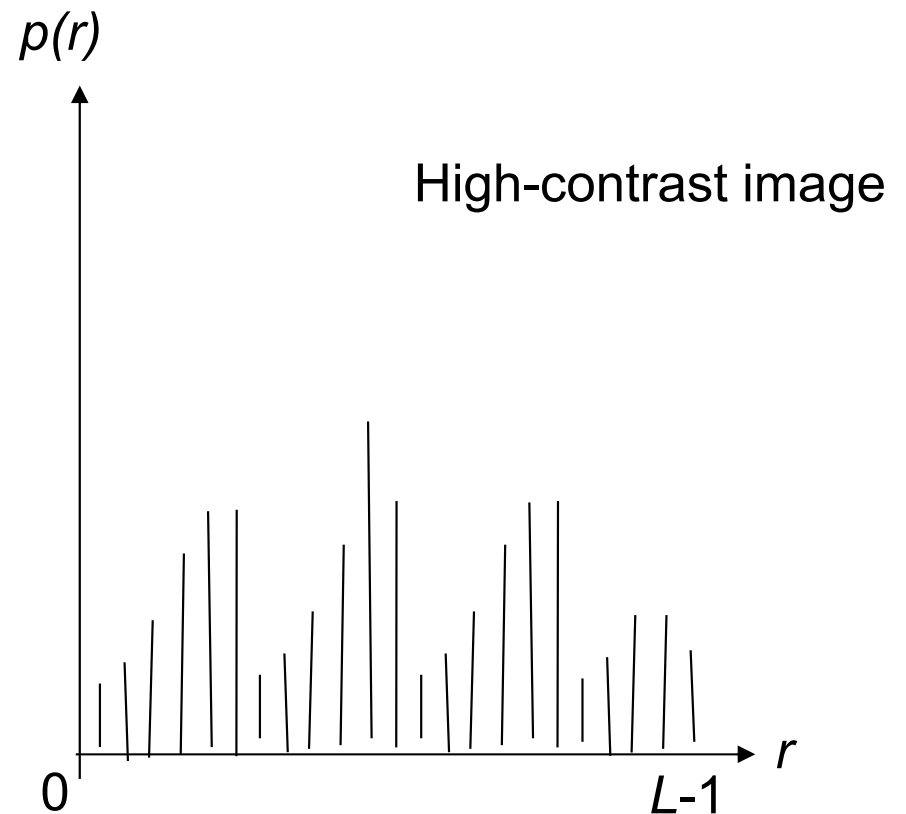
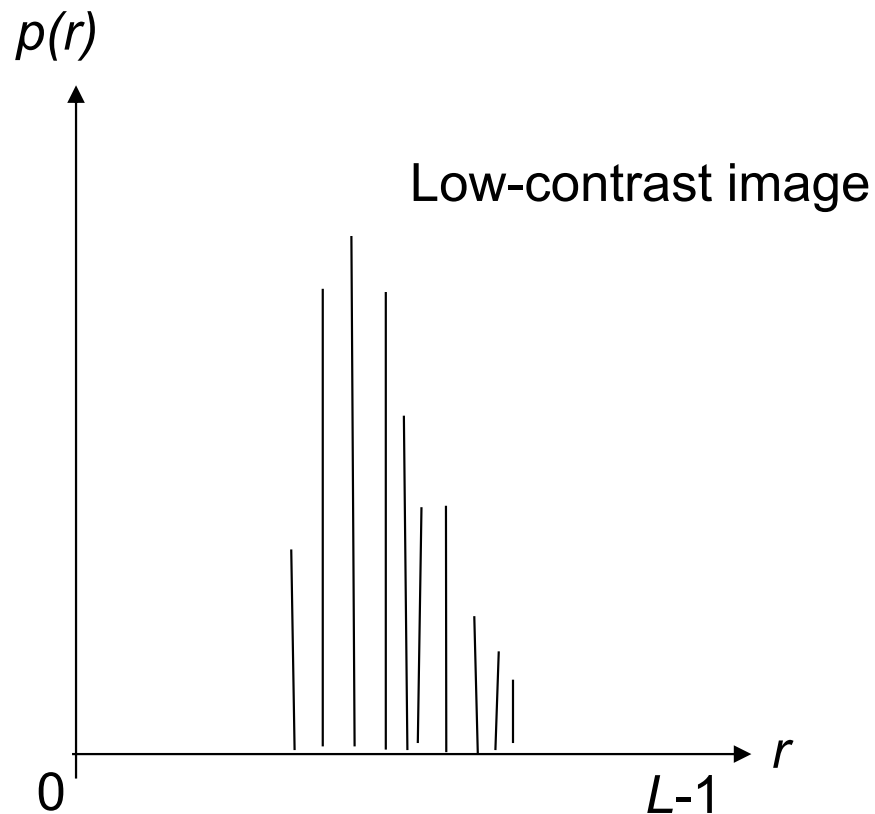
- Brightness refers to the overall lightness or darkness of an image.

Probability density function (pdf)
= Number of pixels with intensity r / Total number of pixels



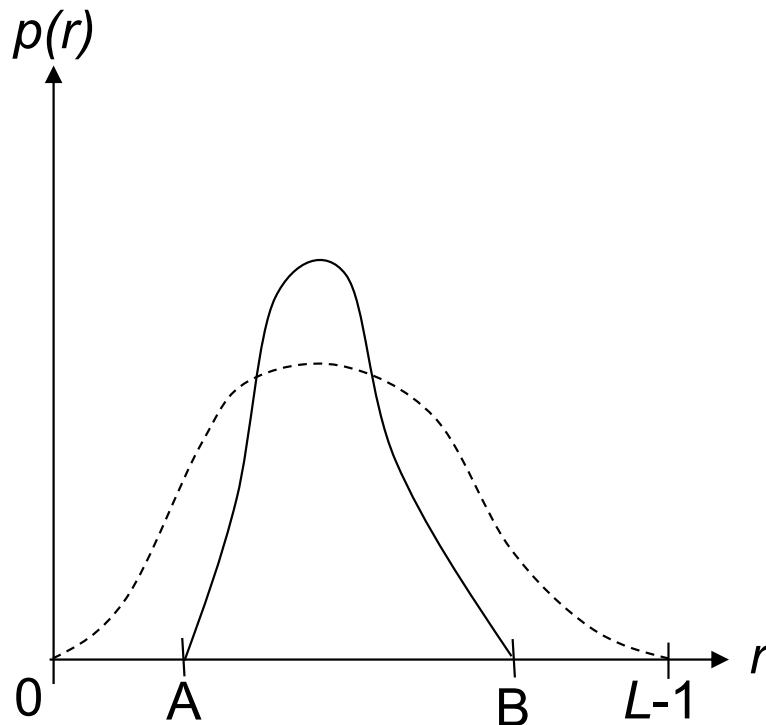
Histogram: Low-contrast and High-contrast Images

- When there are no sharp differences between black and white in an image, the image lacks contrast or does not have sufficient contrast.



Histogram Processing

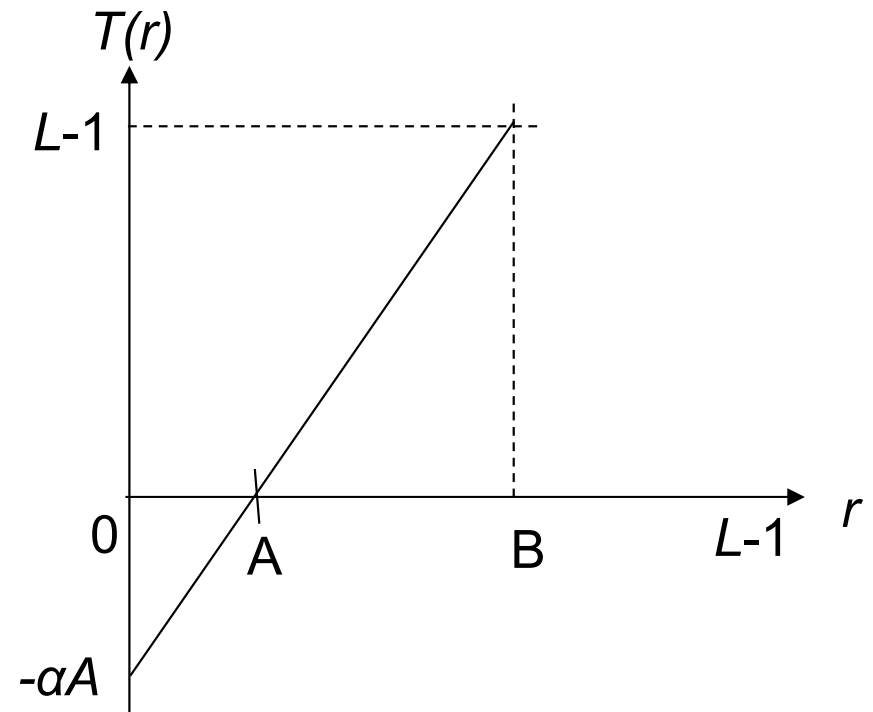
- Applying a linear function to transform the histogram.



$$f(r) = \alpha r + \beta$$

$$0 \equiv f(A) = \alpha A + \beta \Rightarrow \beta = -\alpha A$$

$$L-1 \equiv f(B) = \alpha B + \beta \Rightarrow \alpha = \frac{L-1}{B-A}$$

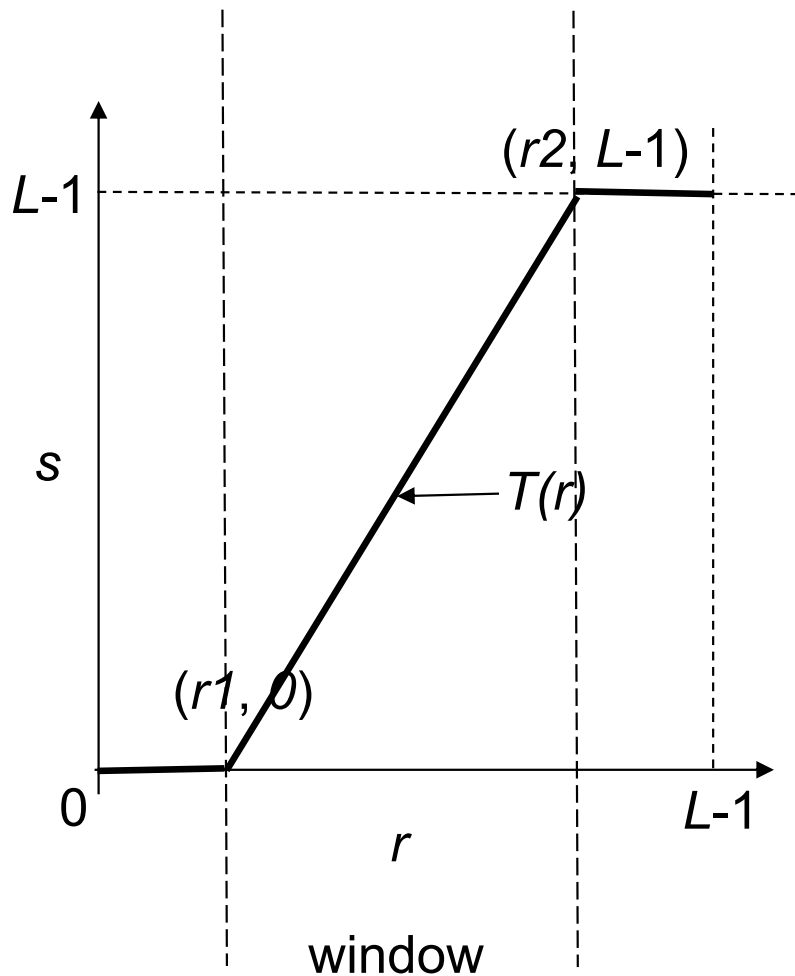




Contrast Stretching

- Contrast stretching is used to change the contrast or brightness of an image.
- In contrast stretching,
 - pixel values below a specified value are considered as black (or, have a value 0),
 - pixel values above another specified value are considered as white, and
 - pixel values in between these two values are considered as shades of gray.
- This is a linear mapping of a subset of pixel values to the entire range of grays from black to white.
- This will produce an image of higher contrast, but some details are lost.

Contrast Stretching



Original gray scale image

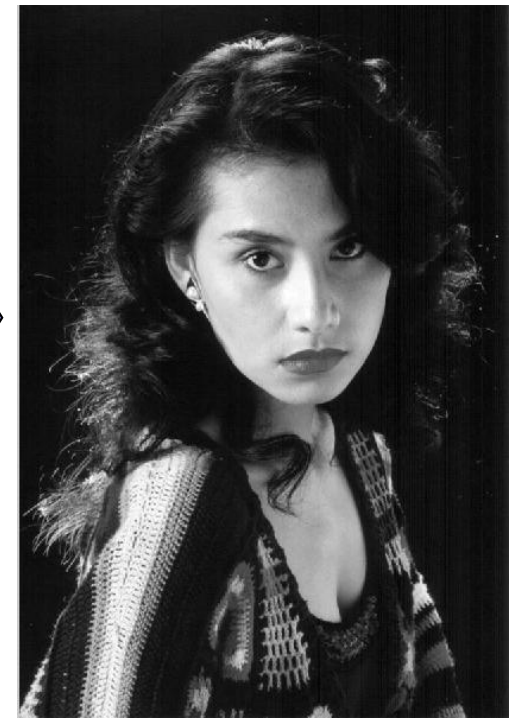
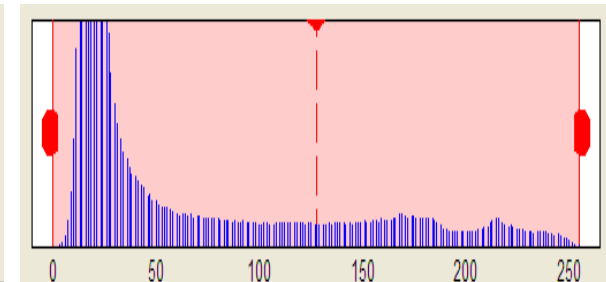
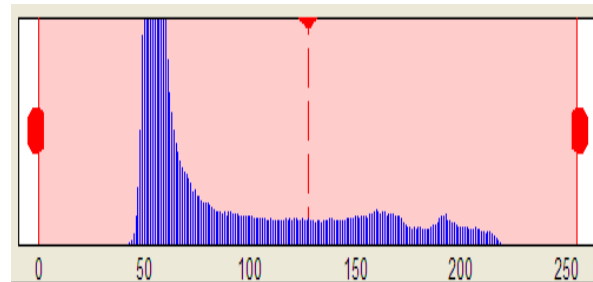


Image after contrast stretching



Contrast Stretching

- Using Matlab

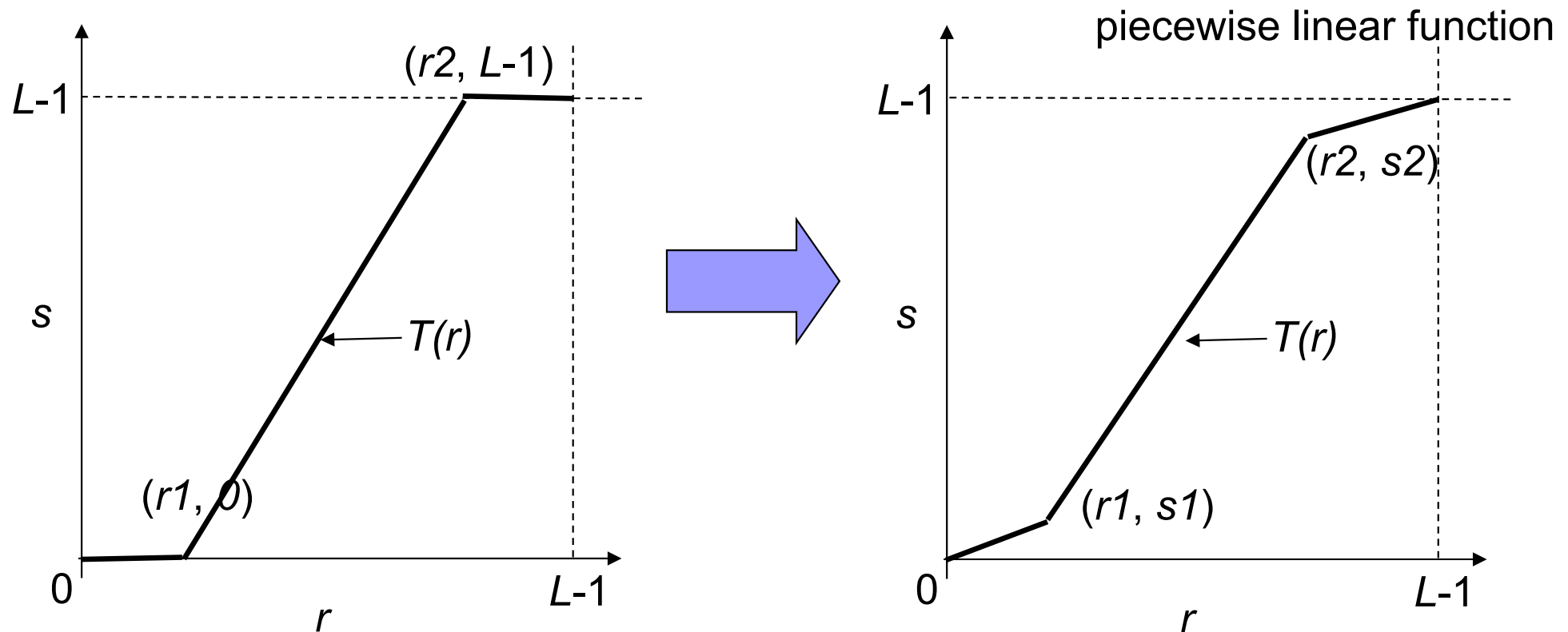
```
% read the input image from disk  
model = imread('model_web.jpg');
```

```
% convert the input color image to 256 gray level image  
modelGray = rgb2gray(model);  
colormap(gray(256));
```

```
% using image tool to adjust image contrast  
imtool(modelGray);
```

Contrast Stretching

- To reduce the loss in details of image.



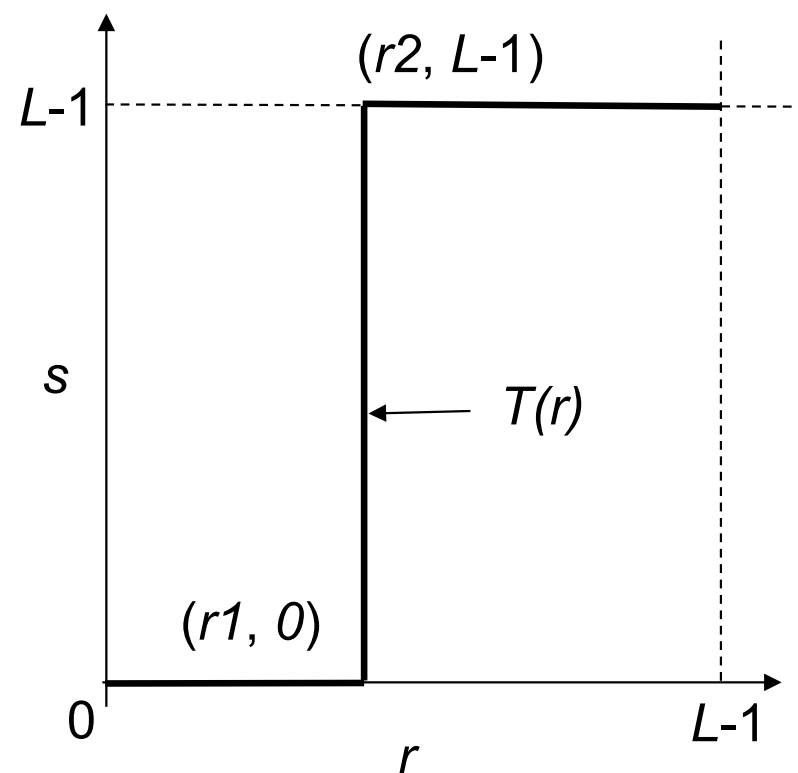


Piecewise Linear Function

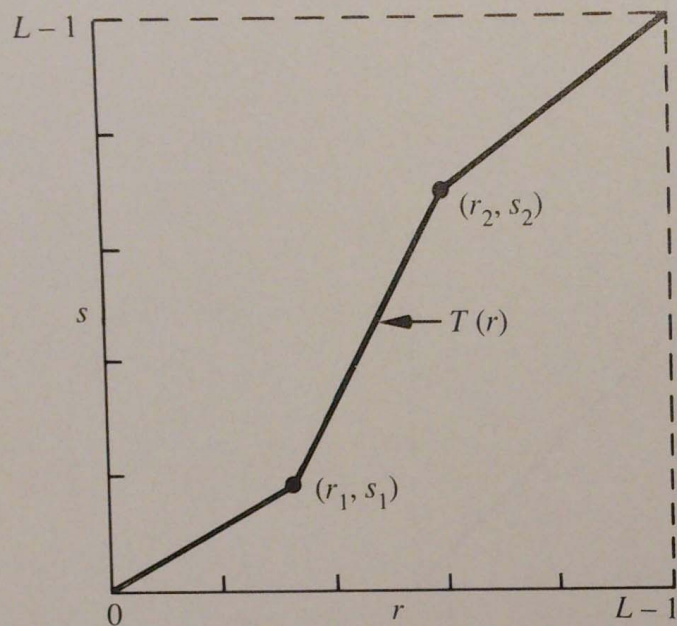
- A piecewise linear function is a function composed of straight line segments, defined by different linear expressions over specific intervals of its domain.
- It is a generalization of typical contrast stretching.

Brightness Thresholding

- It is a special case of piecewise linear function, and a simple way for image segmentation.
- A thresholding function will map all pixel values below a specified threshold to zero and all above to 255 for a 8-bit gray scale image.
- Brightness thresholding will result in a black-and-white image.



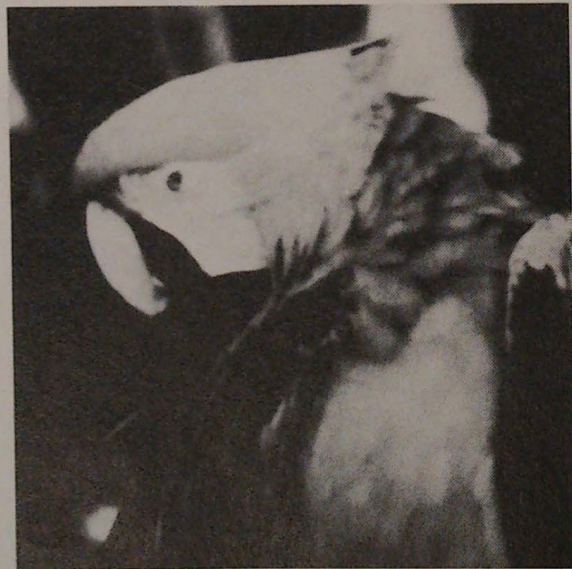
$$r_1 = r_2$$
$$s = T(r) = \begin{cases} 0 & r < r_1 \\ L-1 & r \geq r_2 \end{cases} \text{ or } s = T(r) = \begin{cases} 0 & r \leq r_1 \\ L-1 & r > r_2 \end{cases}$$



(a)



(b)



(c)



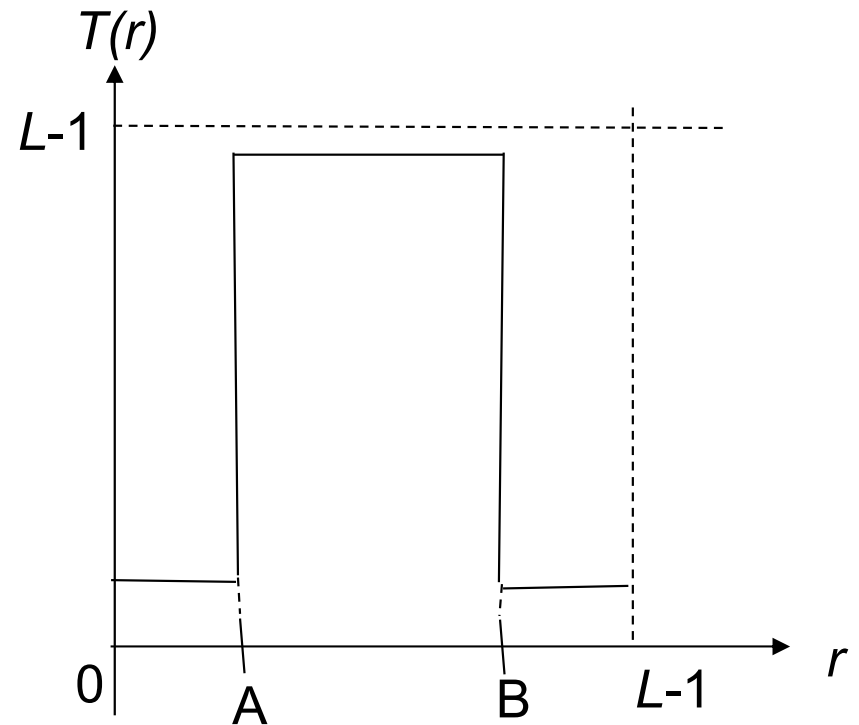
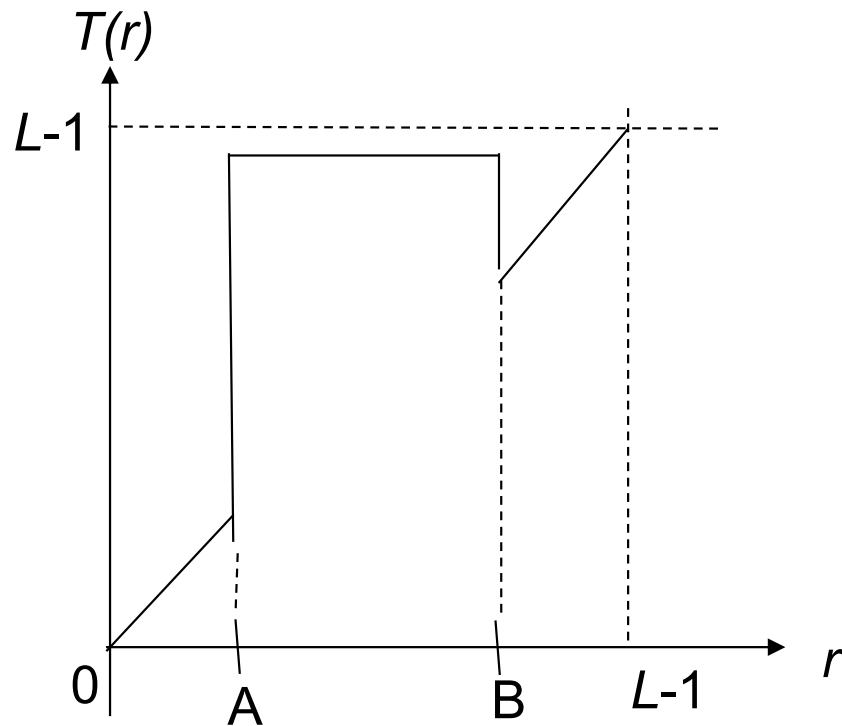
(d)

Contrast stretching:
(a) gray-level transformation function;
(b) a low-contrast image;
(c) result of contrast stretching;
(d) result of thresholding.

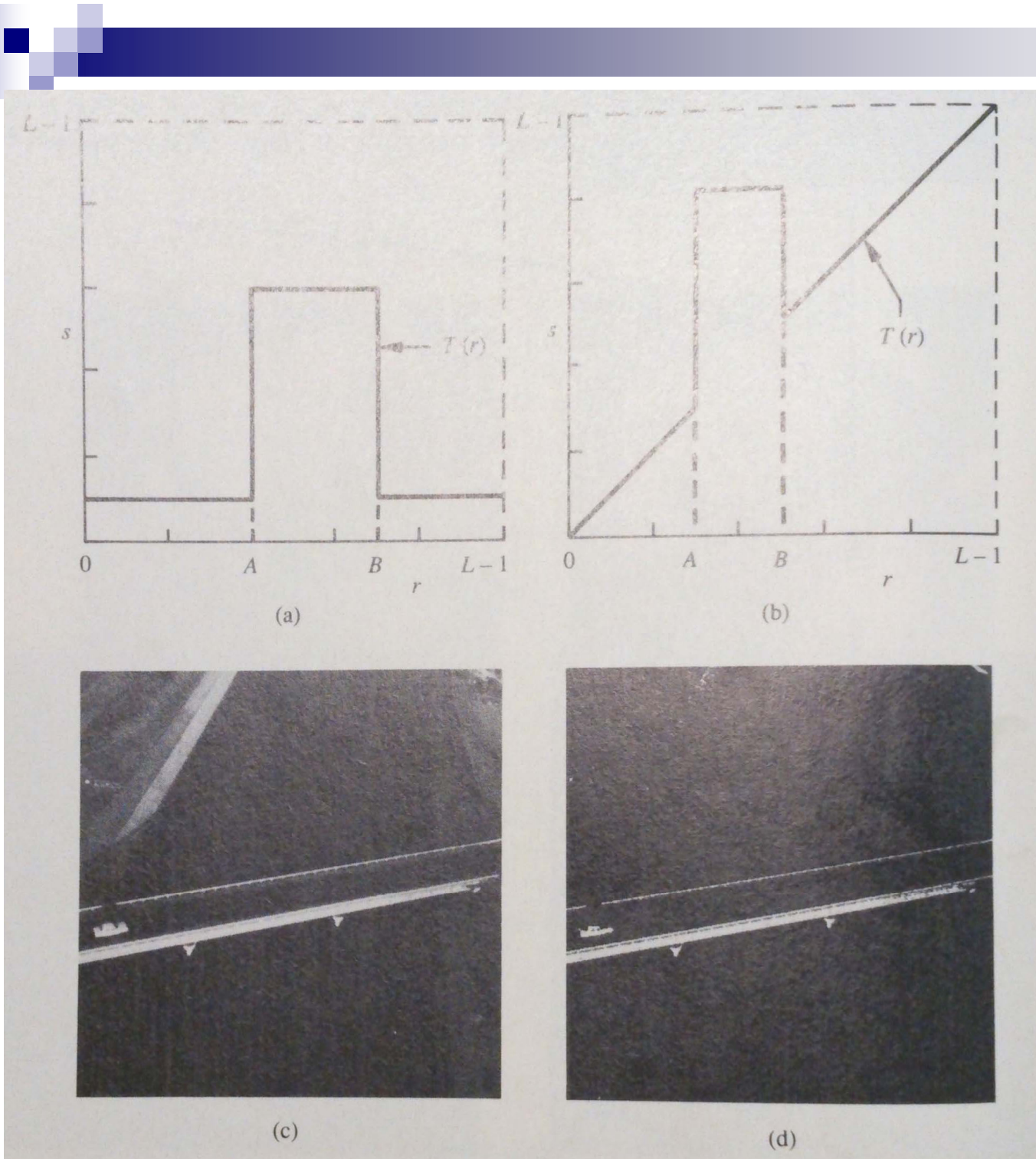
The threshold is set at $r = L/2$ with output set at $L-1$ (white) for any gray level in the input image of $L/2$ or higher and at 0 (black) for all other values.

Gray Level Slicing

- To highlight the gray scales between $[A, B]$.

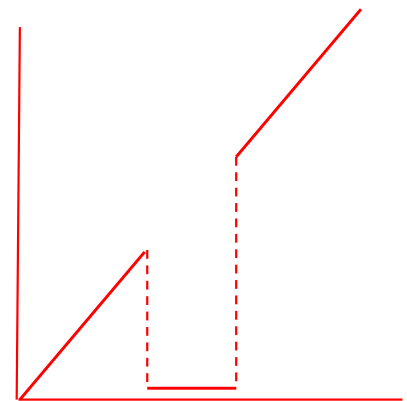


Note that gray scales outside $[A, B]$ are diminished



Gray level slicing (or intensity-level slicing):

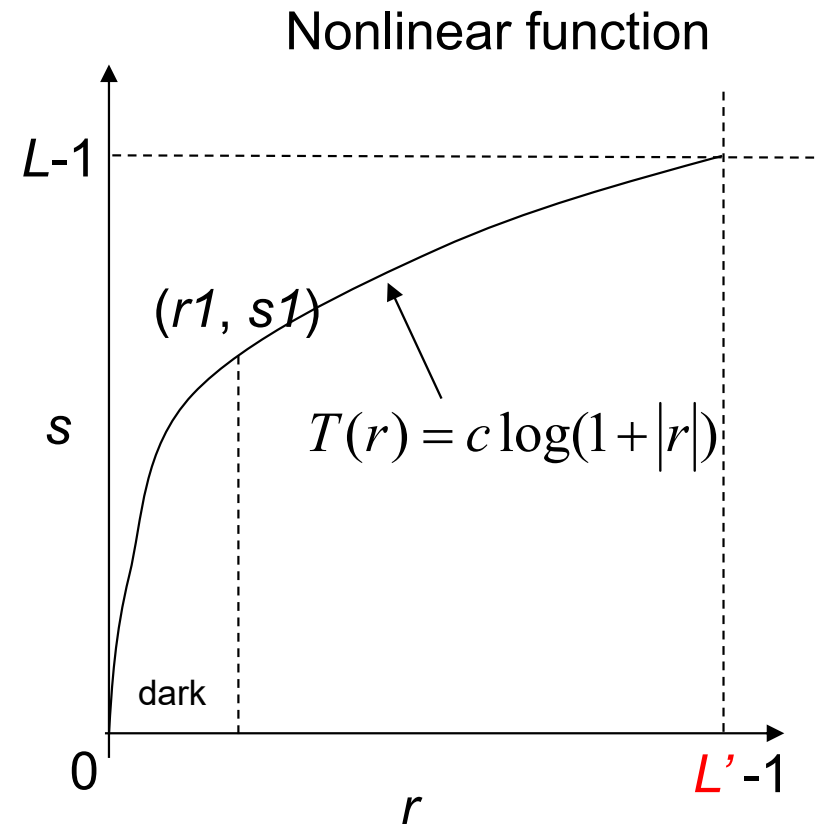
(a) transformation function that highlights a range $[A,B]$ of intensities while diminishing all others to a constant, low level; (b) transformation function that highlights a range $[A,B]$ of intensities but preserves all others; (c) an image; (d) result of using the transformation below with the intensities between A and B darkened.



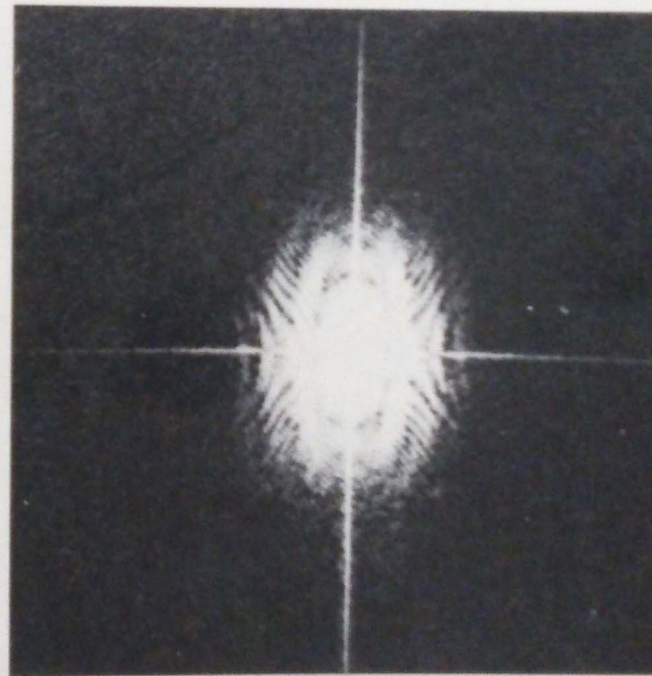
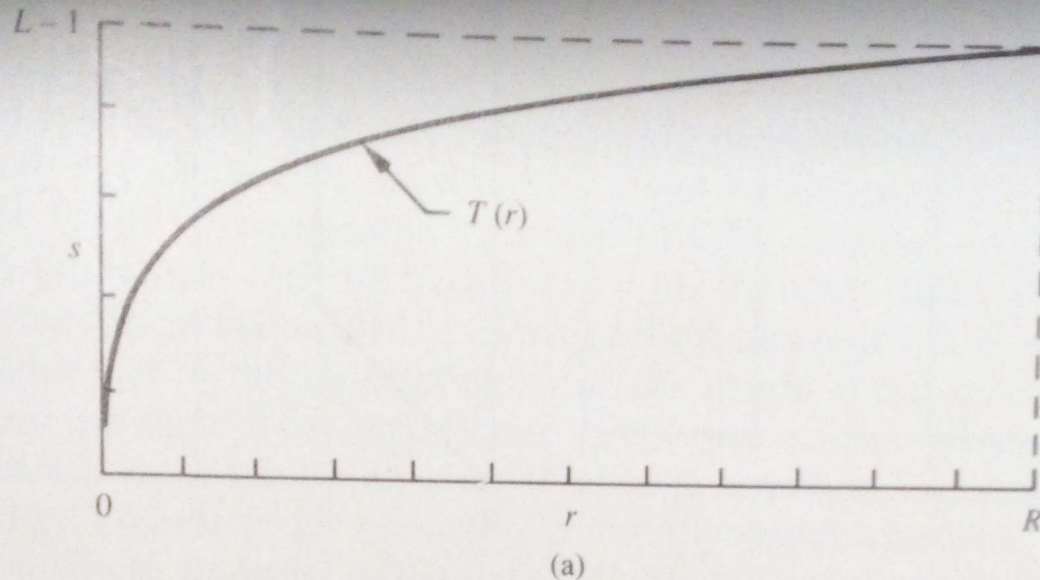
Dynamic Range Compression

- Application of nonlinear functions
 - Compression of dynamic range
 - The dynamic range of a processed image may exceed the capability of display device – only the brightness parts of image are visible.
 - c : scaling constant; logarithm function performs desired compression.

Dynamic range is the range of the different between the lightest light and darkest dark of an image.



If $\log(1 + |r|)$ yields 0 to 6.4, setting $c = 255/6.4$ will scale the range to 0 to 255 gray levels.

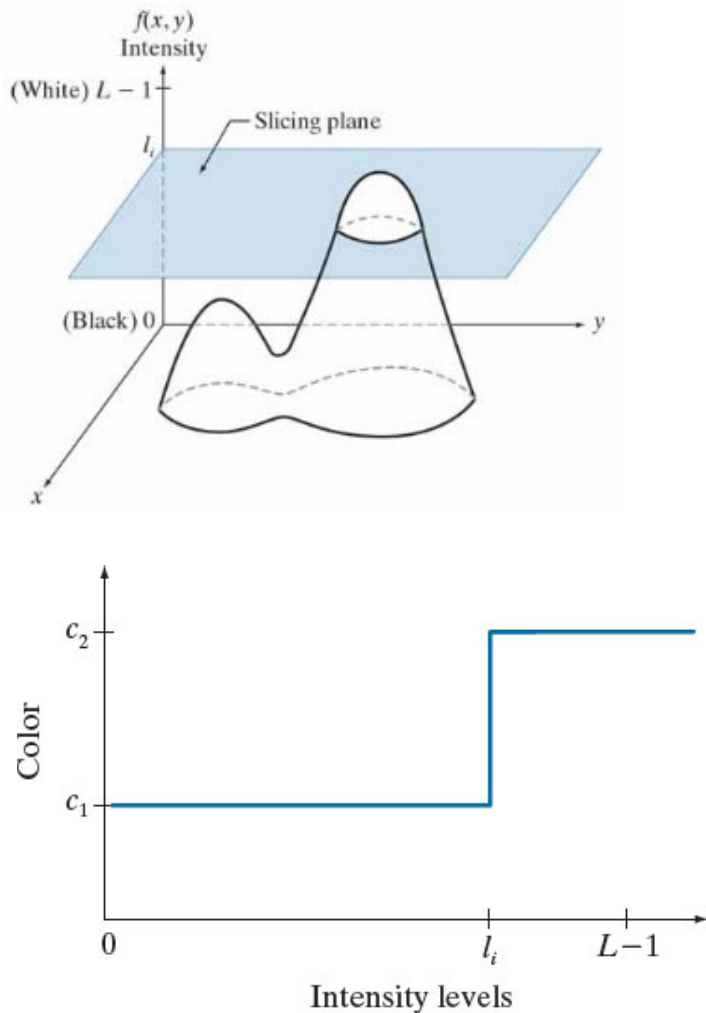


Compression of dynamic range:
(a) logarithm gray-level transformation function;
(b) image with large dynamic range (pixel values ranging from 0 to 2.5×10^6);
(c) result after transformation.

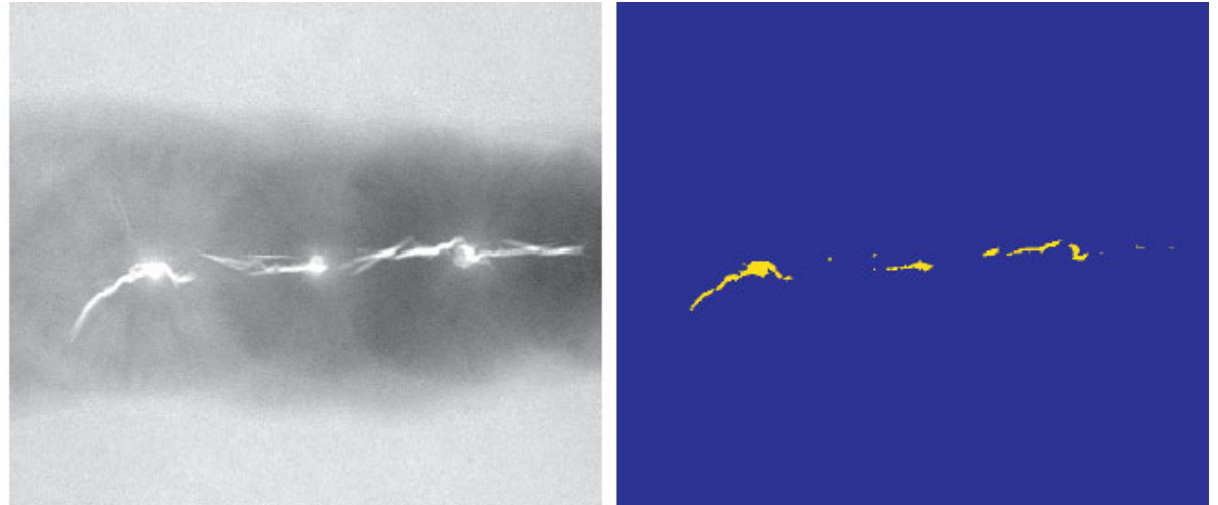
HDR –
High-Dynamic-Range
photography

Pseudocolor Image Processing – intensity (or density) slicing

Pseudo color Image Processing is a digital image processing technique where colors are assigned to grayscale images (intensity images) to enhance their visualization and interpretation.



a b



(a) X-ray image of a weld. (b) Result of color coding.
(Original image courtesy of X-T E K Systems, Ltd.)

Source: Digital Image Processing By Gonzalez and Woods, Pearson.



3. Histogram Equalization

- Histogram equalization is a widely used image technique in image enhancement for increasing contrast.
- It is a method of contrast adjustment using the image's histogram.
- It involves finding a gray level transformation function that creates an output image with a uniform or nearly uniform histogram.
- The transformation replaces each intensity in the input image by a new one.
- Transformation matrix is applied to the whole image at once.

Histogram Equalization

- The aim is to create an image with equally distributed brightness levels over the whole brightness scale
- An ideal equalized image has an equal number of pixels at all brightness levels, resulting in a straight horizontal line on the histogram graph.

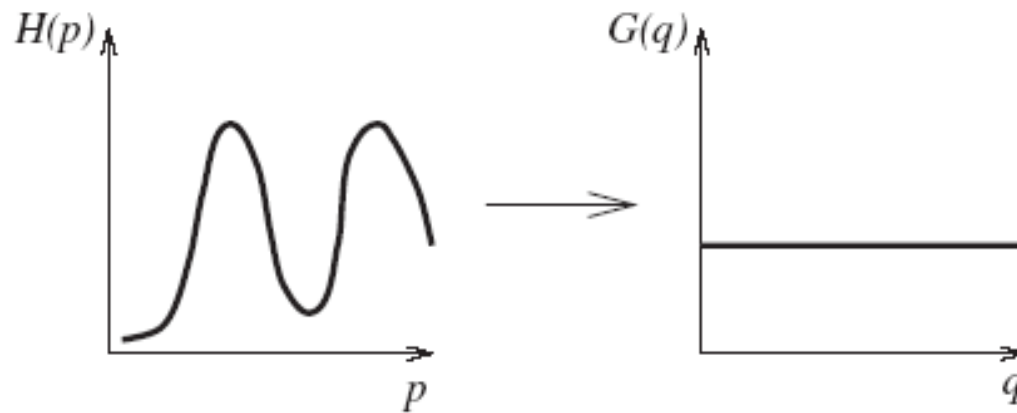


Figure 5.2: Histogram equalization.

Histogram Equalization

- $H(p)$ = input histogram, input gray-scale range = $[p_0, p_k]$
- To find a monotonic transformation $q=T(p)$ such that $G(q)$ is uniform over the output range of $[q_0, q_k]$.
- The histograms can be treated as a discrete probability density function. (Equation 1)
- Suppose that the image has N rows and columns, the equalized histogram $G(q)$ corresponds to the uniform probability density function f . (Equation 2)
- The equalized histogram can be obtained for the continuous probability density function (Equation 3).
- The desired pixel brightness transformation T is the cumulative histogram (Equation 4).
- The discrete approximation to the continuous $T(p)$ is given in Equation 5.

$$\sum_{i=0}^k G(q_i) = \sum_{i=0}^k H(p_i) \quad (1)$$

$$f = \frac{1}{N^2(q_k - q_0)} \quad (2)$$

$$N^2 \int_{q_0}^q \frac{1}{q_k - q_0} ds = \frac{N^2(q - q_0)}{q_k - q_0} = \int_{p_0}^p H(s) ds \quad (3)$$

$$q = T(p) = \frac{q_k - q_0}{N^2} \int_{p_0}^p H(s) ds + q_0 \quad (4)$$

$$q = T(p) = \frac{q_k - q_0}{N^2} \sum_{i=p_0}^p H(i) + q_0 \quad (5)$$

Algorithm 5.1: Histogram equalization

1. For an $N \times M$ image of G gray-levels (often 256), create an array H of length G initialized with 0 values.
2. Form the image histogram: Scan every pixel and increment the relevant member of H —if pixel p has intensity g_p , perform

$$H[g_p] = H[g_p] + 1 .$$

3. Form the cumulative image histogram H_c :

$$\begin{aligned} H_c[0] &= H[0] , \\ H_c[p] &= H_c[p-1] + H[p] , \quad p = 1, 2, \dots, G-1 . \end{aligned}$$

4. Set

$$T[p] = \text{round} \left(\frac{G-1}{NM} H_c[p] \right) .$$

(This step obviously lends itself to more efficient implementation by constructing a look-up table of the multiples of $(G-1)/NM$, and making comparisons with the values in H_c , which are monotonically increasing.)

5. Rescan the image and write an output image with gray-levels g_q , setting

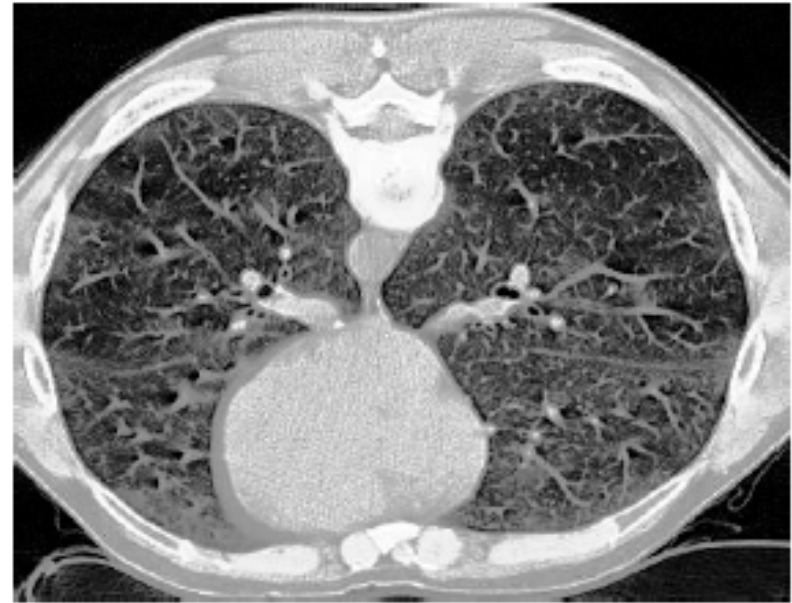
$$g_q = T[g_p] .$$

Histogram Equalization

- Example



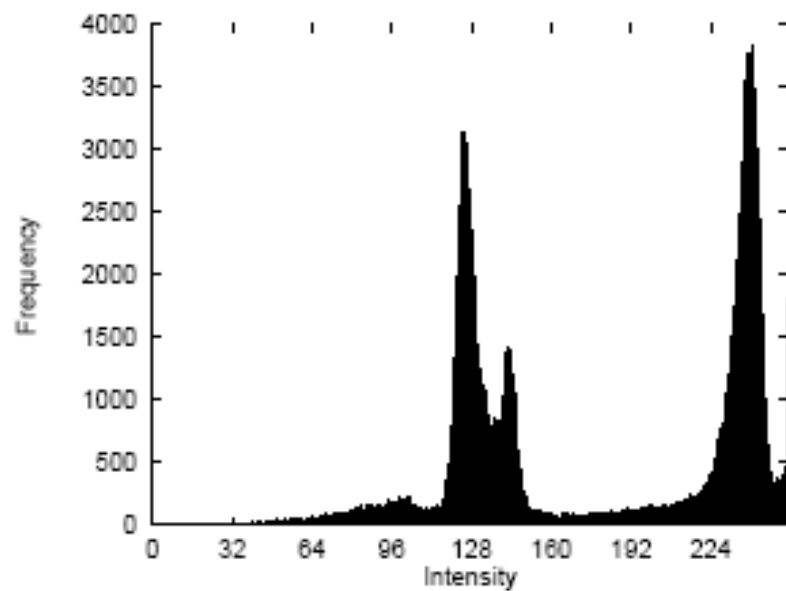
(a)



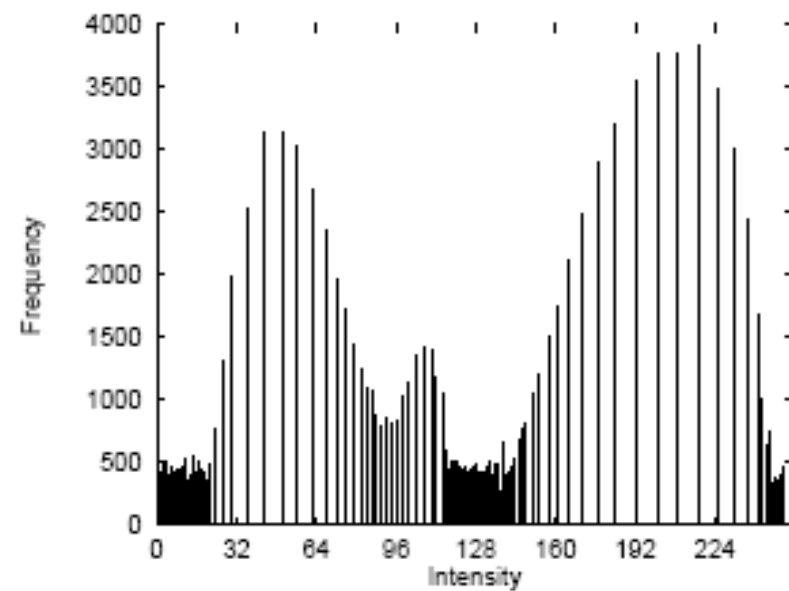
(b)

Figure 5.3: Histogram equalization. (a) Original image. (b) Equalized image.

Histogram Equalization



(a)



(b)

Figure 5.4: Histogram equalization: Original and equalized histograms corresponding to Figure 5.3a,b.



Matlab Implementation of Histogram Equalization

function [im_out, H, Hc, T] = hist_eq(im)

input:

im [m x n] input image

output:

im_out [m x n] equalized image

H [1x256] histogram of the input image

Hc [1x256] cumulative histogram of the input function

T [1x256] transformation function of the intensity



Matlab Implementation of Histogram Equalization

<https://www.mathworks.com/help/images/histogram-equalization.html>



Histogram Equalization

- Useful in images with both background and foreground are dark or bright.
 - It can reveal good detail in over or under-exposed photographs.
 - The method will provide good view of hard tissue (bone) in x-ray image.
- Disadvantage: global application leads to indiscriminate modification of image
 - The method may increase contrast of background noise and decrease usable signal.
 - It could not support the need to highlight certain gray level range.



Histogram Specification

- Histogram Specification is a generalized version of histogram equalization
- A “target” histogram that actually define the desired shape of the image histogram is specified.
- A nonlinear stretch operation is applied to force the image histogram to have that shape.



Summary

- Image most frequently represented as multidimensional array can be added, subtracted and made equal to another image.
- Logical operation can also be perform on image.
- Gray-scale transformations change brightness without regard to position in the image. The common gray-scale transformations are piecewise linear function, negative transformation and brightness thresholding.
- The goal of histogram equalization is to create an image with equally distributed brightness levels over the whole brightness scale.

