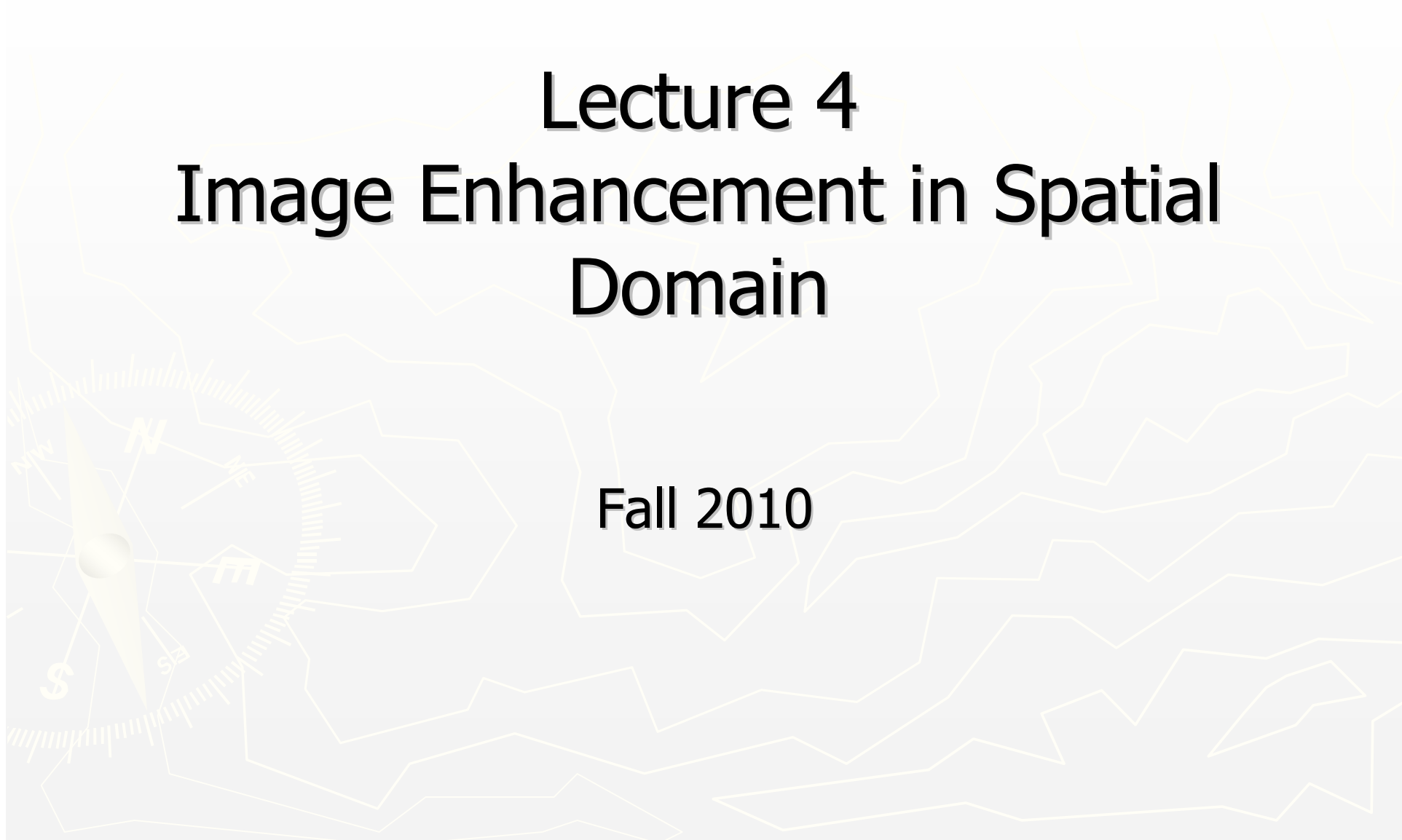


# Digital Image Processing

## Lecture 4 Image Enhancement in Spatial Domain

Fall 2010



# 2 domains

- ▶ **Spatial Domain : (image plane)**
  - Techniques are based on direct manipulation of pixels in an image
- ▶ **Frequency Domain :**
  - Techniques are based on modifying the Fourier transform of an image
- ▶ **There are some enhancement techniques based on various combinations of methods from these two categories.**

# Good images

- ▶ For human visual
  - The visual evaluation of image quality is a highly subjective process.
  - It is hard to standardize the definition of a good image.
- ▶ For machine perception
  - The evaluation task is easier.
  - A good image is one which gives the best machine recognition results.
- ▶ A certain amount of trial and error usually is required before a particular image enhancement approach is selected.

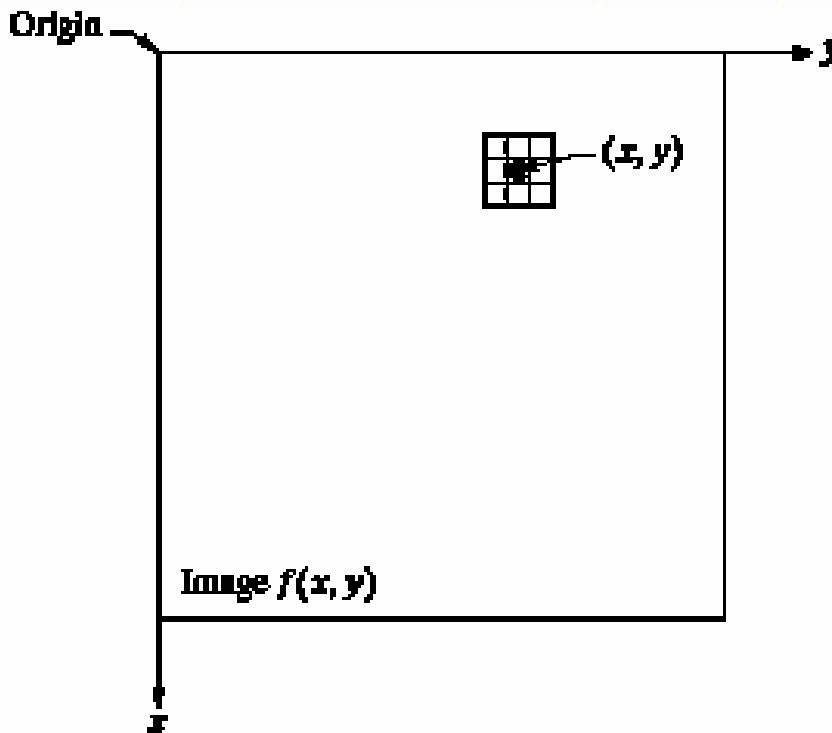
# Spatial Domain

- ▶ Procedures that operate directly on pixels.

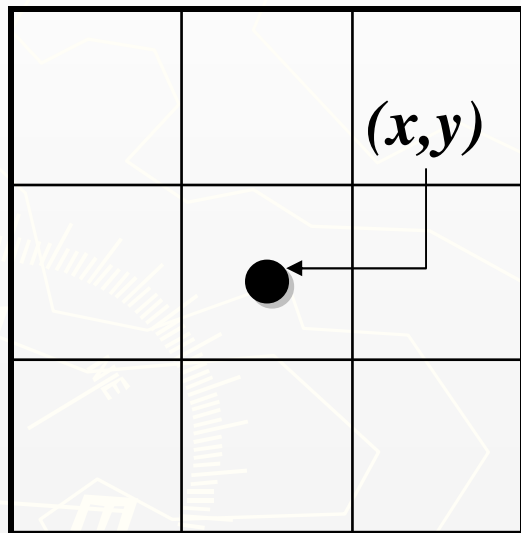
$$g(x,y) = T[f(x,y)]$$

where

- $f(x,y)$  is the input image
- $g(x,y)$  is the processed image
- $T$  is an operator on  $f$  defined over some neighborhood of  $(x,y)$



# Mask/Filter



- ▶ Neighborhood of a point  $(x,y)$  can be defined by using a square/rectangular (common used) or circular subimage area centered at  $(x,y)$
- ▶ The center of the subimage is moved from pixel to pixel starting at the top of the corner

# Point Processing

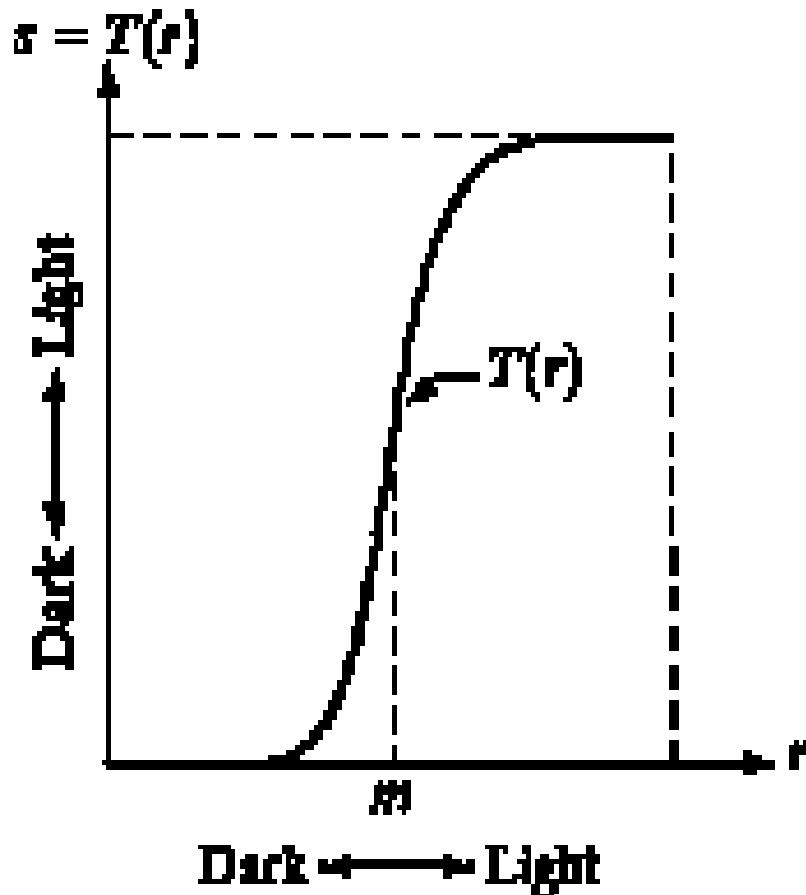
- ▶ Neighborhood = 1x1 pixel
- ▶  $g$  depends on only the value of  $f$  at  $(x,y)$
- ▶  $T$  = gray level (or intensity or mapping) transformation function

$$s = T(r)$$

- ▶ Where

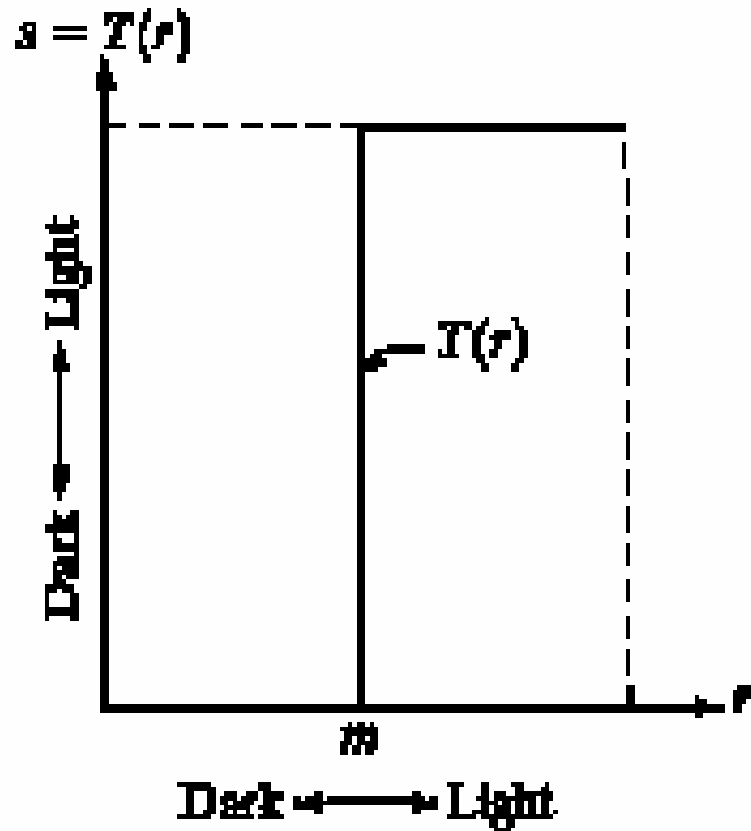
- $r$  = gray level of  $f(x,y)$
- $s$  = gray level of  $g(x,y)$

# Contrast Stretching



- ▶ Produce higher contrast than the original by
  - darkening the levels below  $m$  in the original image
  - Brightening the levels above  $m$  in the original image

# Thresholding

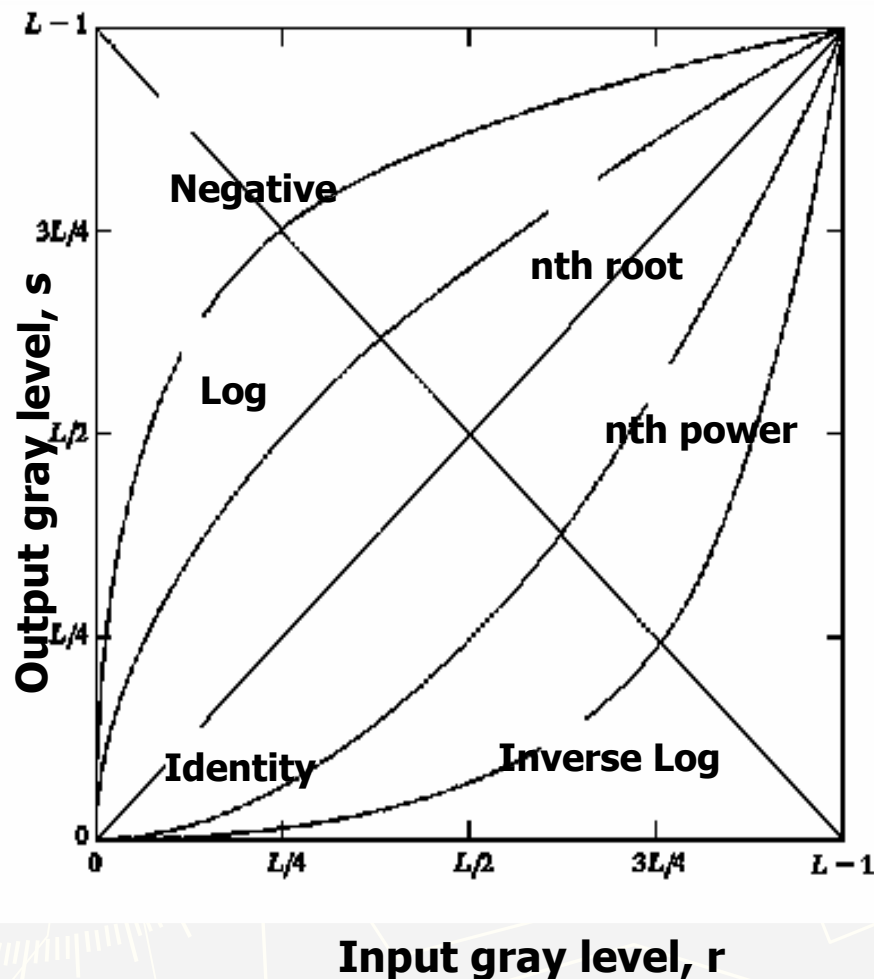


- ▶ Produce a two-level (binary) image

# Mask Processing or Filter

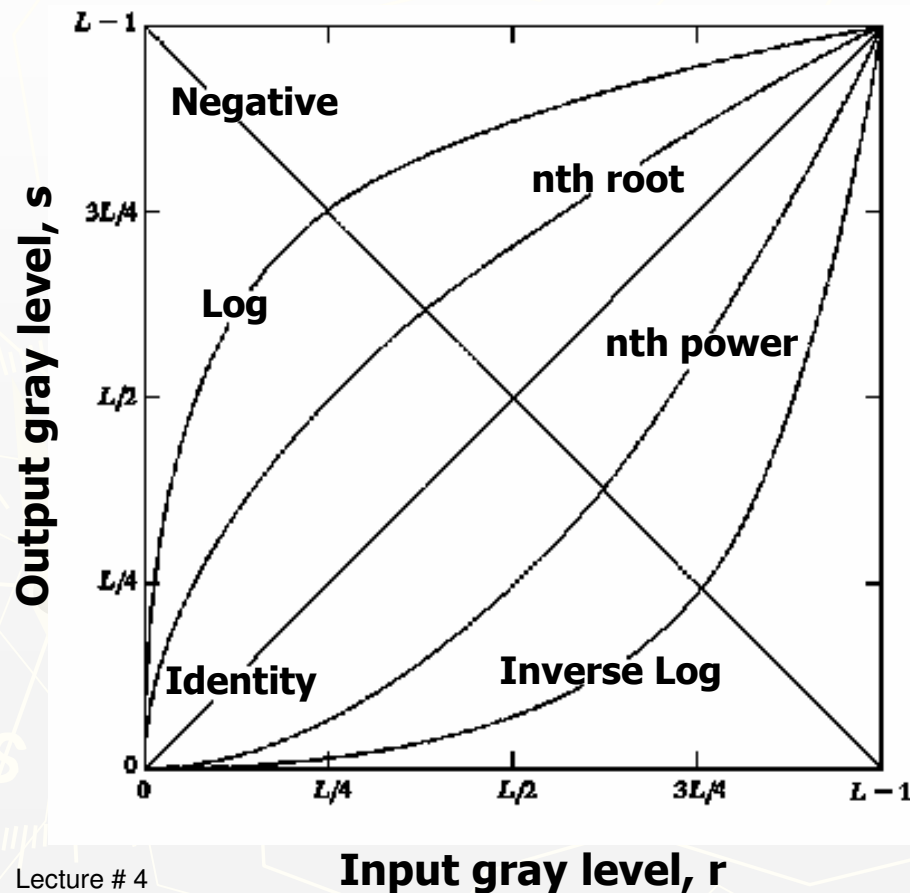
- ▶ Neighborhood is bigger than 1x1 pixel
- ▶ Use a function of the values of  $f$  in a predefined neighborhood of  $(x,y)$  to determine the value of  $g$  at  $(x,y)$
- ▶ The value of the mask coefficients determine the nature of the process
- ▶ Used in techniques
  - Image Sharpening
  - Image Smoothing

# 3 basic gray-level transformation functions



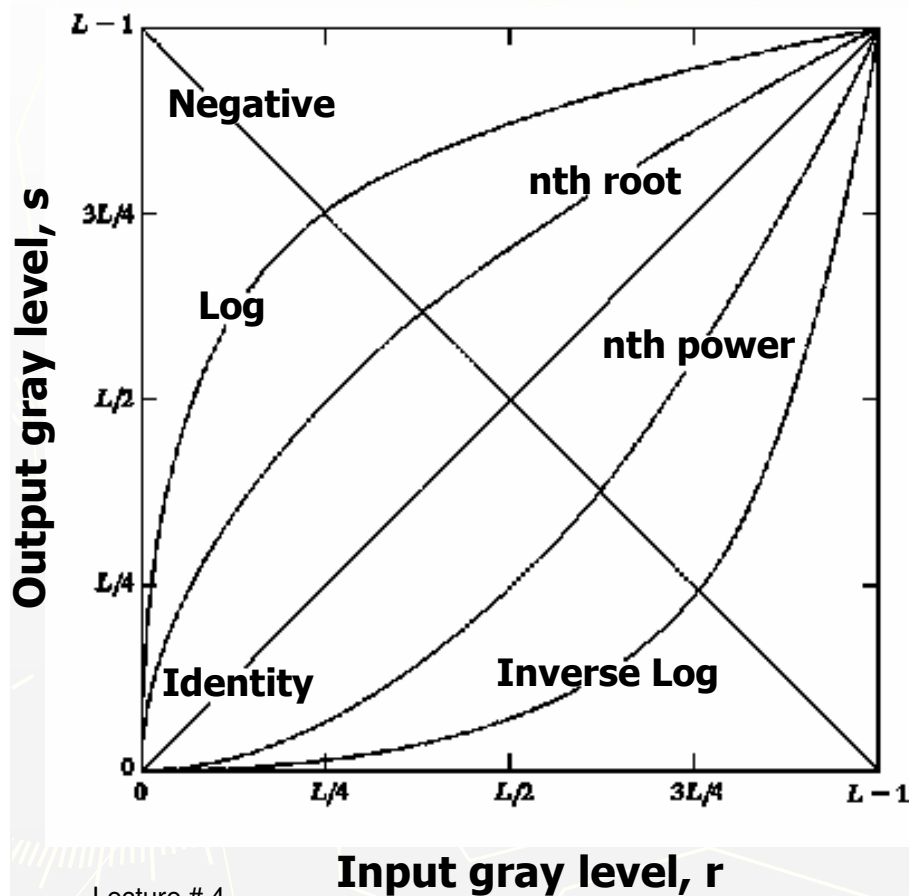
- ▶ Linear function
  - Negative and identity transformations
- ▶ Logarithm function
  - Log and inverse-log transformation
- ▶ Power-law function
  - $n^{\text{th}}$  power and  $n^{\text{th}}$  root transformations

# Identity function



- ▶ Output intensities are identical to input intensities.
- ▶ Is included in the graph only for completeness.

# Image Negatives

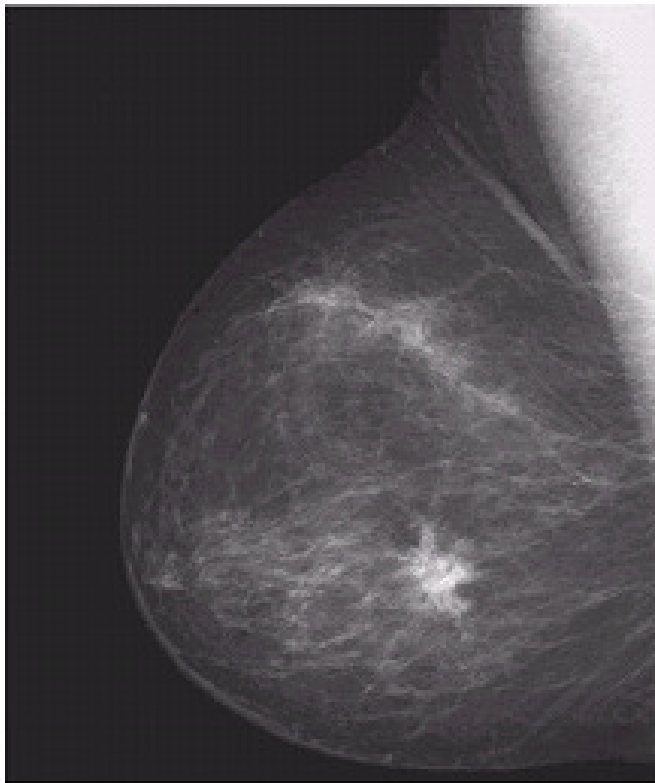


Lecture # 4

- ▶ An image with gray level in the range  $[0, L-1]$  where  $L = 2^n$ ;  $n = 1, 2, \dots$
- ▶ Negative transformation :  
$$s = L - 1 - r$$
- ▶ Reversing the intensity levels of an image.
- ▶ Suitable for enhancing white or gray detail embedded in dark regions of an image, especially when the black area dominant in size.

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# Example of Negative Image

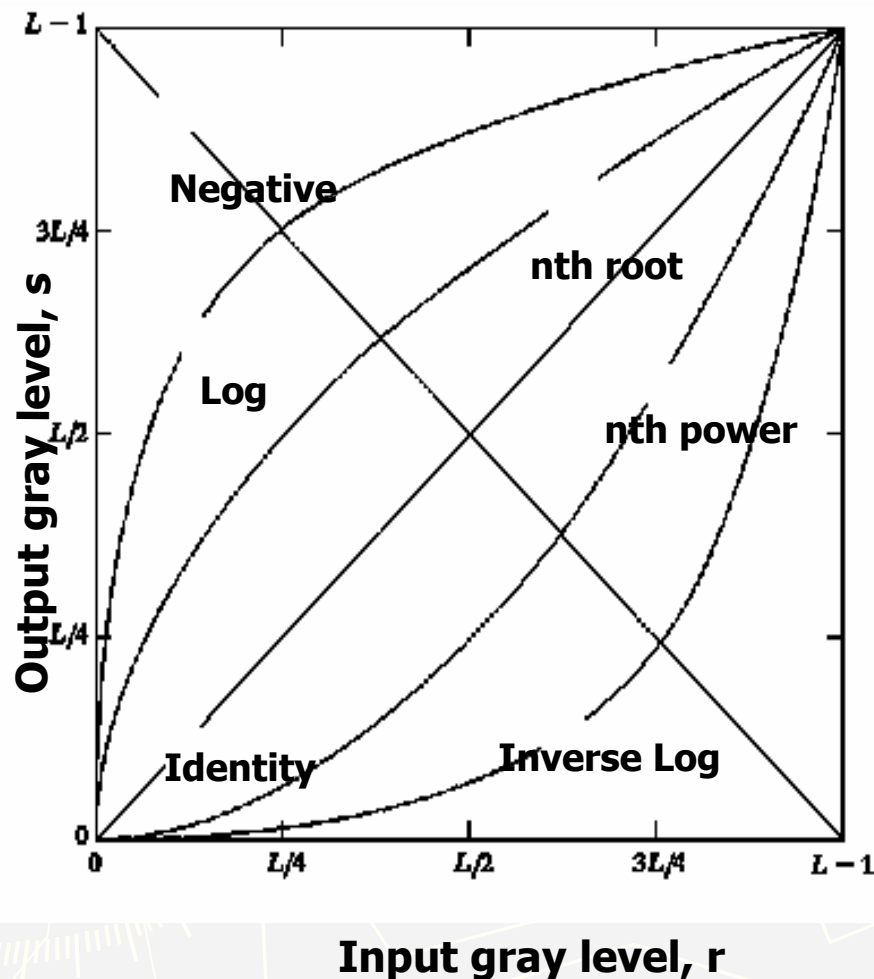


**Input Image**



**Negative Image : gives a better vision to analyze the image**

# Log Transformations



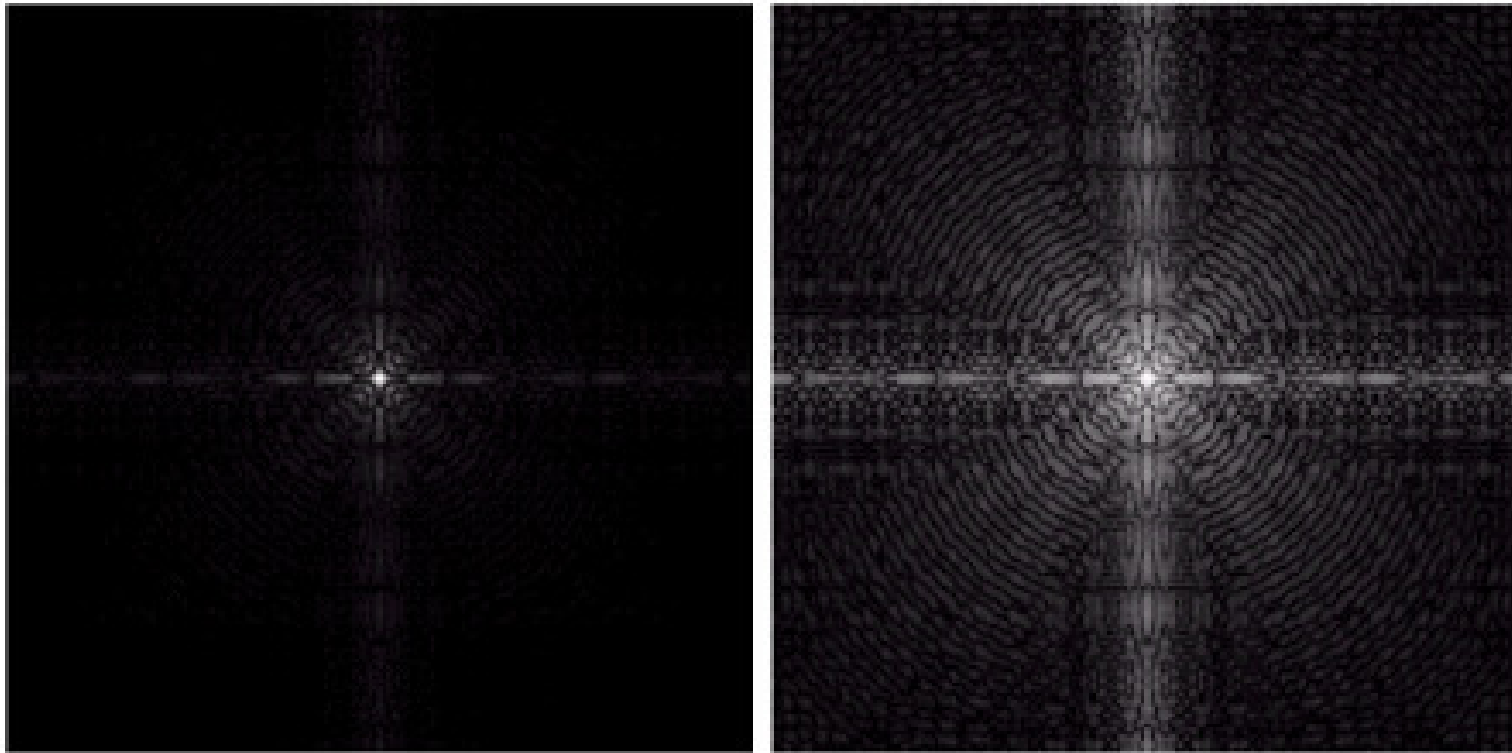
$$s = c \log(1+r)$$

- ▶  $c$  is a constant and  $r \geq 0$
- ▶ Log curve maps a narrow range of low gray-level values in the input image into a wider range of output levels.
- ▶ Used to expand the values of dark pixels in an image while compressing the higher-level values.

# Log Transformations

- ▶ It compresses the dynamic range of images with large variations in pixel values
- ▶ Example of image with dynamic range: Fourier spectrum image
- ▶ It can have intensity range from 0 to  $10^6$  or higher.
- ▶ We can't see the significant degree of detail as it will be lost in the display.

# Example of Logarithm Image

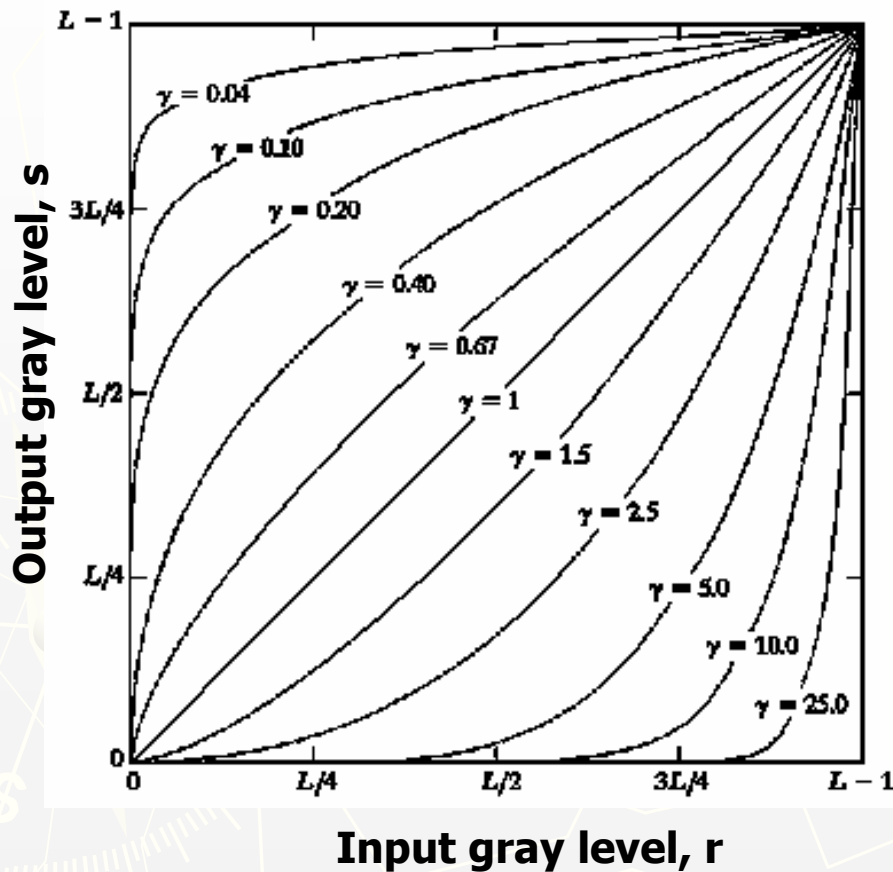


# Inverse Logarithm Transformations

- ▶ Do opposite to the Log Transformations
- ▶ Used to expand the values of high pixels in an image while compressing the darker-level values.

# Power-Law Transformations

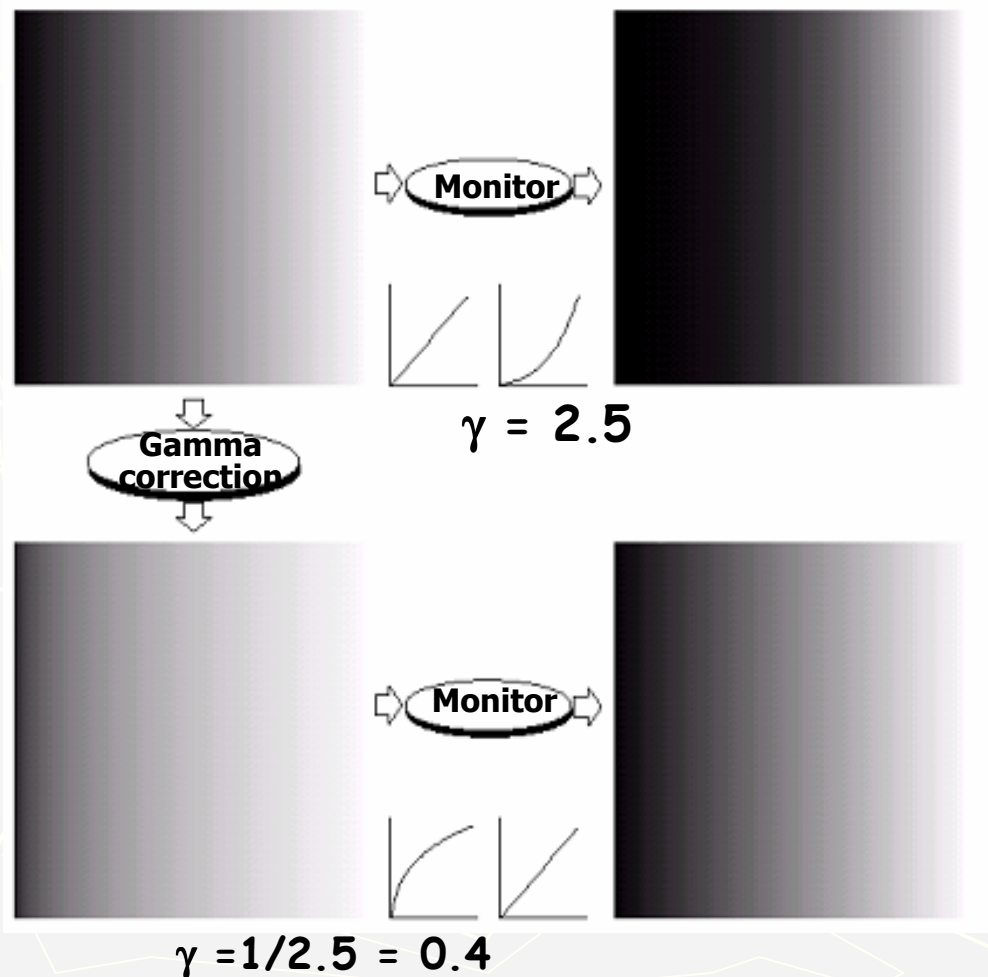
$$s = cr^\gamma$$



- ▶  $c$  and  $\gamma$  are positive constants
- ▶ Power-law curves with fractional values of  $\gamma$  map a narrow range of dark input values into a wider range of output values, with the opposite being true for higher values of input levels.
- ▶  $c = \gamma = 1 \Rightarrow$  Identity function

Lecture 7 Plots of  $s = cr^\gamma$  for various values of  $\gamma$   
( $c = 1$  in all cases)

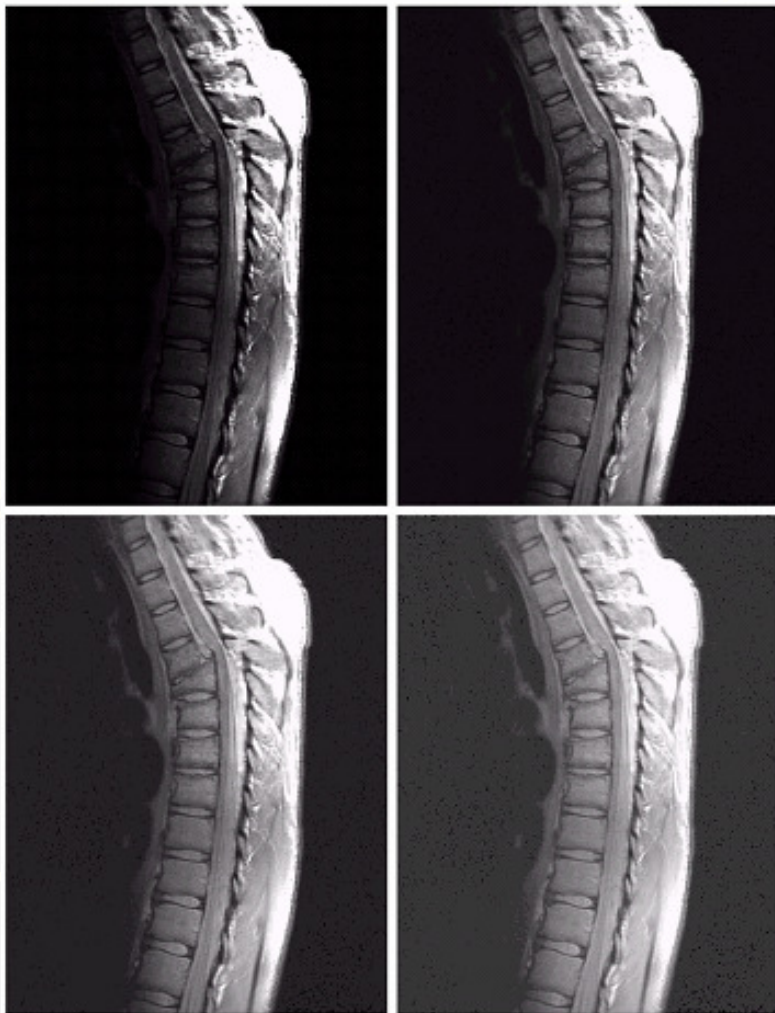
# Gamma correction



- ▶ Cathode ray tube (CRT) devices have an intensity-to-voltage response that is a power function, with  $\gamma$  varying from 1.8 to 2.5
- ▶ The picture will become darker.
- ▶ Gamma correction is done by preprocessing the image before inputting it to the monitor with  $s = cr^{1/\gamma}$

# Another example : MRI

a	b
c	d



(a) a magnetic resonance image of an upper thoracic human spine with a fracture dislocation and spinal cord impingement

- The picture is predominately dark
- An expansion of gray levels are desirable  $\Rightarrow$  needs  $\gamma < 1$

(b) result after power-law transformation with  $\gamma = 0.6$ ,  $c=1$

(c) transformation with  $\gamma = 0.4$   
(best result)

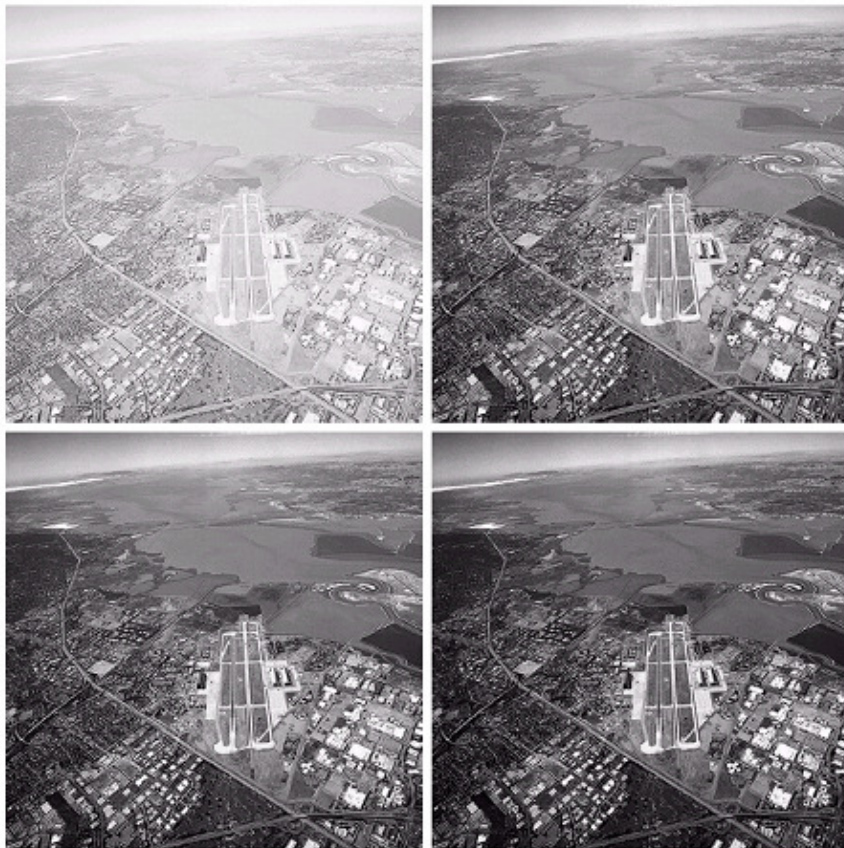
(d) transformation with  $\gamma = 0.3$   
(under acceptable level)

# Effect of decreasing gamma

- ▶ When the  $\gamma$  is reduced too much, the image begins to reduce contrast to the point where the image started to have very slight “wash-out” look, especially in the background

# Another example

a	b
c	d



- (a) image has a washed-out appearance, it needs a compression of gray levels  $\Rightarrow$  needs  $\gamma > 1$
- (b) result after power-law transformation with  $\gamma = 3.0$  (suitable)
- (c) transformation with  $\gamma = 4.0$  (suitable)
- (d) transformation with  $\gamma = 5.0$  (high contrast, the image has areas that are too dark, some detail is lost)

# Piecewise-Linear Transformation Functions

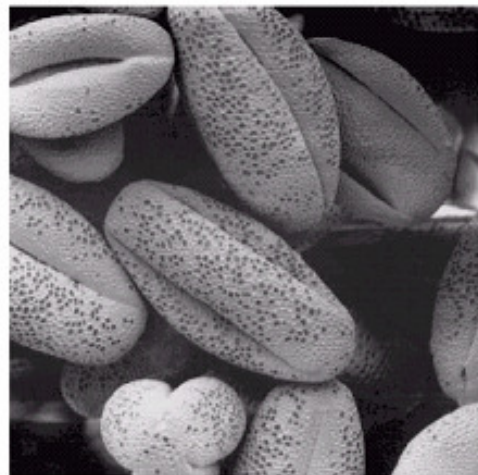
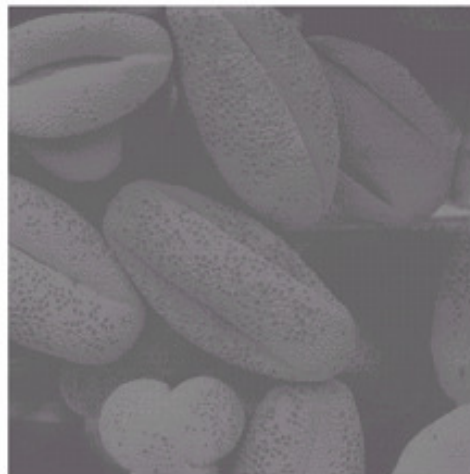
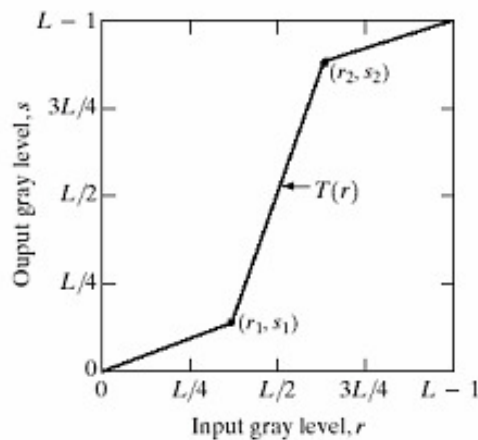
## ► Advantage:

- The form of piecewise functions can be arbitrarily complex

## ► Disadvantage:

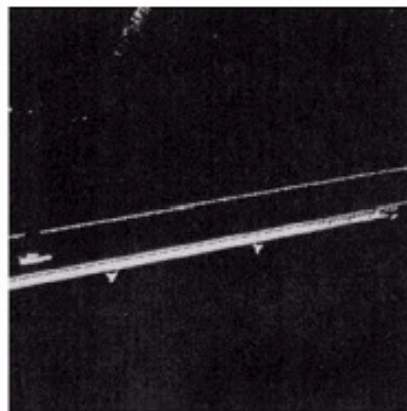
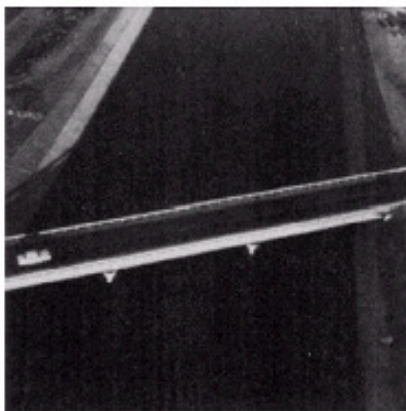
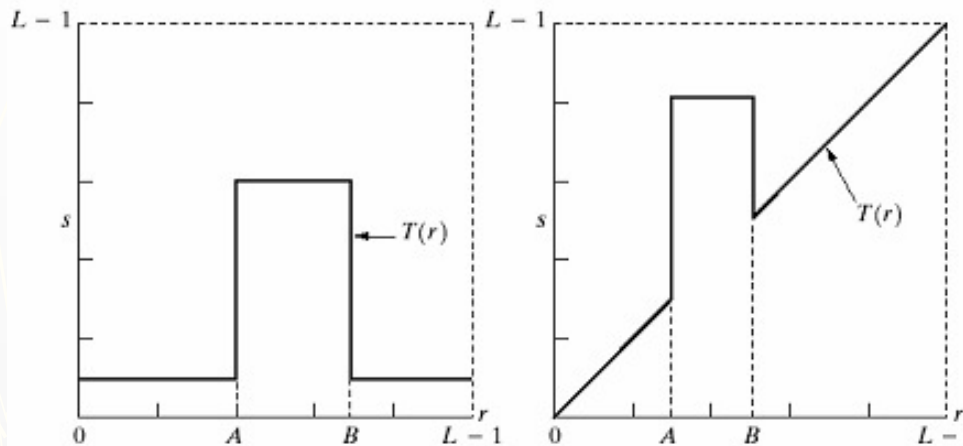
- Their specification requires considerably more user input

# Contrast Stretching



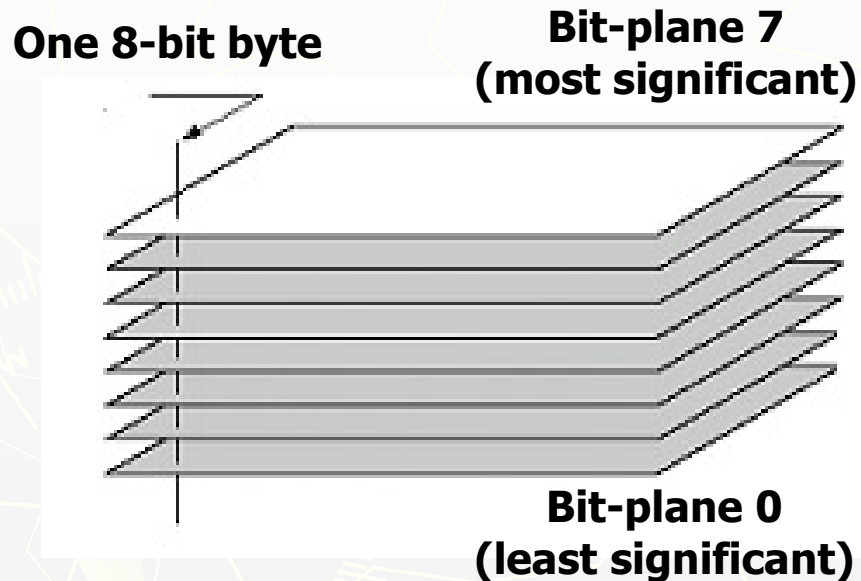
- ▶ increase the dynamic range of the gray levels in the image
- ▶ (b) a low-contrast image : result from poor illumination, lack of dynamic range in the imaging sensor, or even wrong setting of a lens aperture of image acquisition
- ▶ (c) result of contrast stretching:  $(r_1, s_1) = (r_{\min}, 0)$  and  $(r_2, s_2) = (r_{\max}, L-1)$
- ▶ (d) result of thresholding

# Gray-level slicing



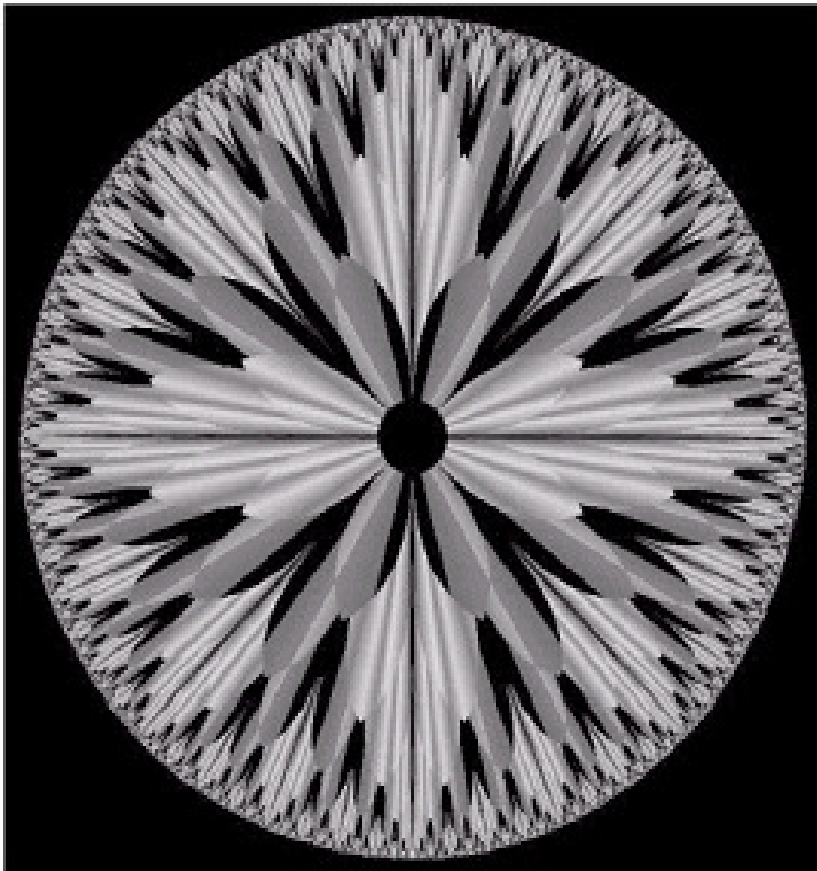
- ▶ Highlighting a specific range of gray levels in an image
  - Display a high value of all gray levels in the range of interest and a low value for all other gray levels
- ▶ (a) transformation highlights range  $[A, B]$  of gray level and reduces all others to a constant level
- ▶ (b) transformation highlights range  $[A, B]$  but preserves all other levels

# Bit-plane slicing



- ▶ Highlighting the contribution made to total image appearance by specific bits
- ▶ Suppose each pixel is represented by 8 bits
  - Higher-order bits contain the majority of the visually significant data
  - Useful for analyzing the relative importance played by each bit of the image

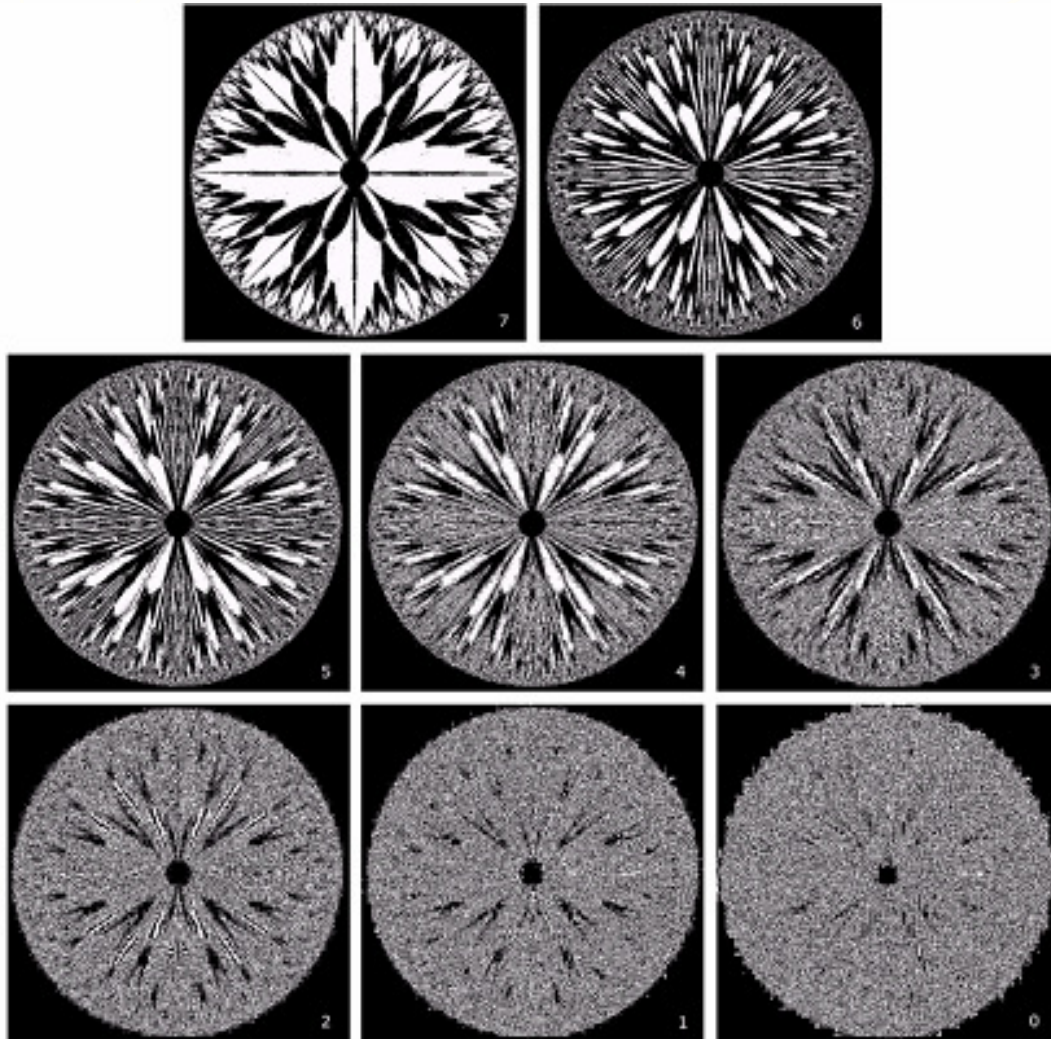
# Example



**An 8-bit fractal image**

- ▶ The (binary) image for bit-plane 7 can be obtained by processing the input image with a thresholding gray-level transformation.
  - Map all levels between 0 and 127 to 0
  - Map all levels between 129 and 255 to 255

# 8 bit planes



Bit-plane 7		Bit-plane 6	
Bit-plane 5	Bit-plane 4	Bit-plane 3	
Bit-plane 2	Bit-plane 1	Bit-plane 0	

# Histogram Processing

- ▶ Histogram of a digital image with gray levels in the range  $[0, L-1]$  is a discrete function

$$h(r_k) = n_k$$

- ▶ Where

- $r_k$  : the  $k^{\text{th}}$  gray level
- $n_k$  : the number of pixels in the image having gray level  $r_k$
- $h(r_k)$  : histogram of a digital image with gray levels  $r_k$

# Normalized Histogram

- ▶ dividing each of histogram at gray level  $r_k$  by the total number of pixels in the image,  $n$

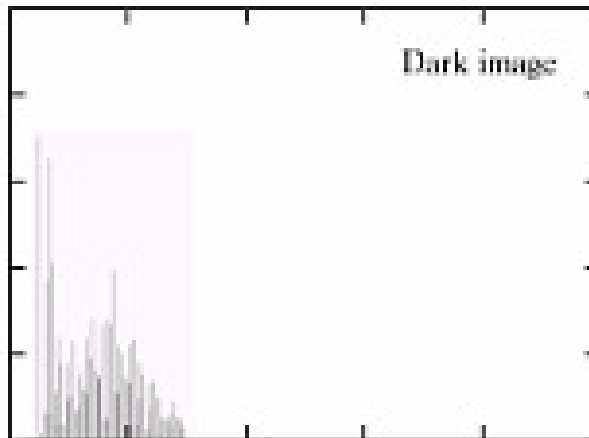
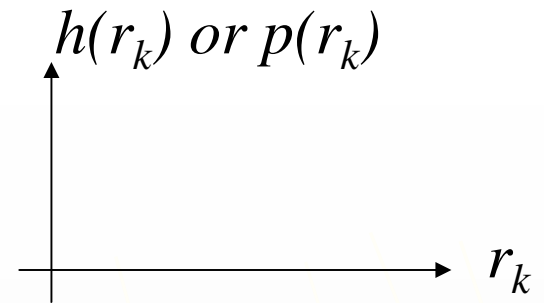
$$p(r_k) = n_k / n$$

- ▶ For  $k = 0, 1, \dots, L-1$
- ▶  $p(r_k)$  gives an estimate of the probability of occurrence of gray level  $r_k$
- ▶ The sum of all components of a normalized histogram is equal to 1

# Histogram Processing

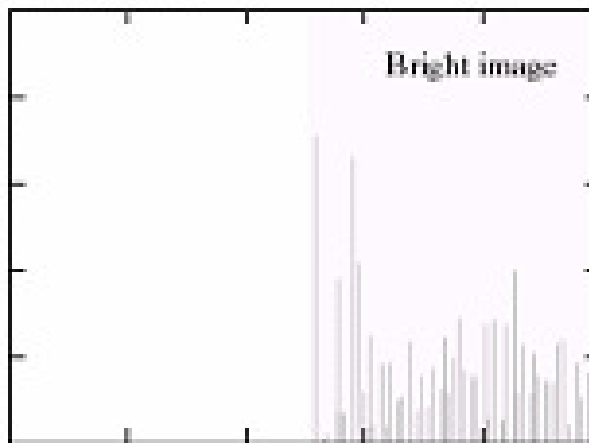
- ▶ Basic for numerous spatial domain processing techniques
- ▶ Used effectively for image enhancement
- ▶ Information inherent in histograms also is useful in image compression and segmentation

# Example



## Dark image

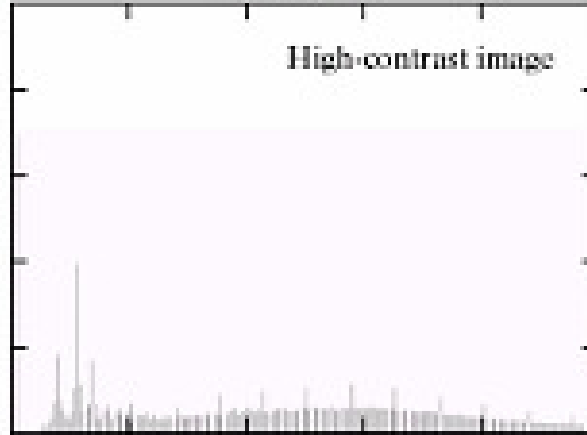
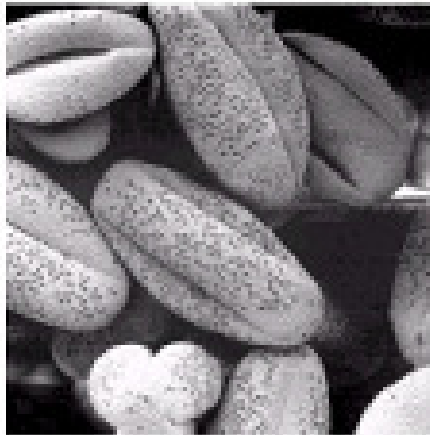
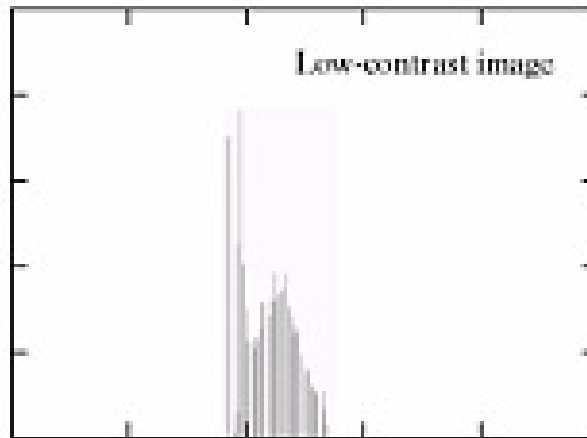
Components of histogram are concentrated on the low side of the gray scale.



## Bright image

Components of histogram are concentrated on the high side of the gray scale.

# Example



## Low-contrast image

histogram is narrow and centered toward the middle of the gray scale

## High-contrast image

histogram covers broad range of the gray scale and the distribution of pixels is not too far from uniform, with very few vertical lines being much higher than the others