

# BBM413 Fundamentals of Image Processing

## Image Enhancement

# Contents

- Point processing techniques
  - Negative images
  - Thresholding
  - Logarithmic transformation
  - Power law transforms
  - Grey level slicing
  - Bit plane slicing
- Histogram processing methods:
  - Histogram processing
  - Histogram Equalization
  - Histogram Matching

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  - Negative images
  - Thresholding
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  - Bit plane slicing
- Histogram processing methods:
  - Histogram processing
  - Histogram Equalization
  - Histogram Matching

# A Note About Grey Levels

- When we have spoken about image grey level values we have said they are in the range  $[0, 255]$ 
  - Where 0 is black and 255 is white
- There is no reason why we have to use this range
  - The range  $[0, 255]$  stems from display technologies.
- For many of the image processing operations in this lecture grey levels are assumed to be given in the range  $[0.0, 1.0]$

# What Is Image Enhancement?

Image enhancement is the process of making images more useful

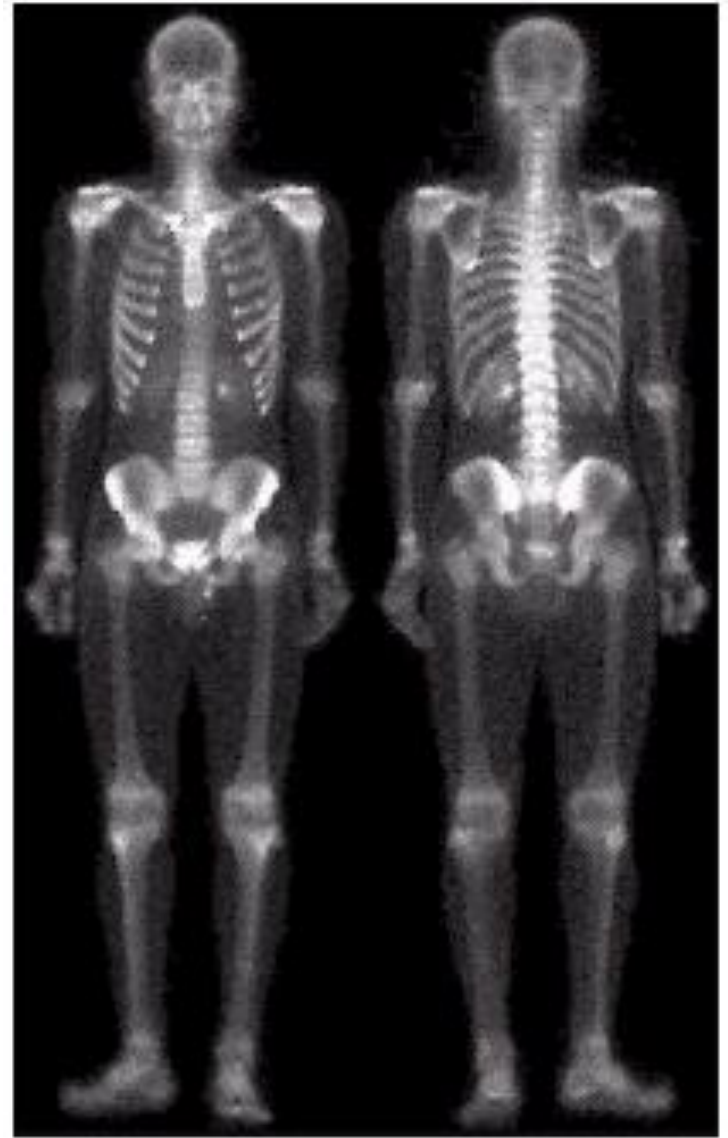
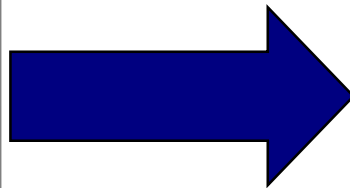
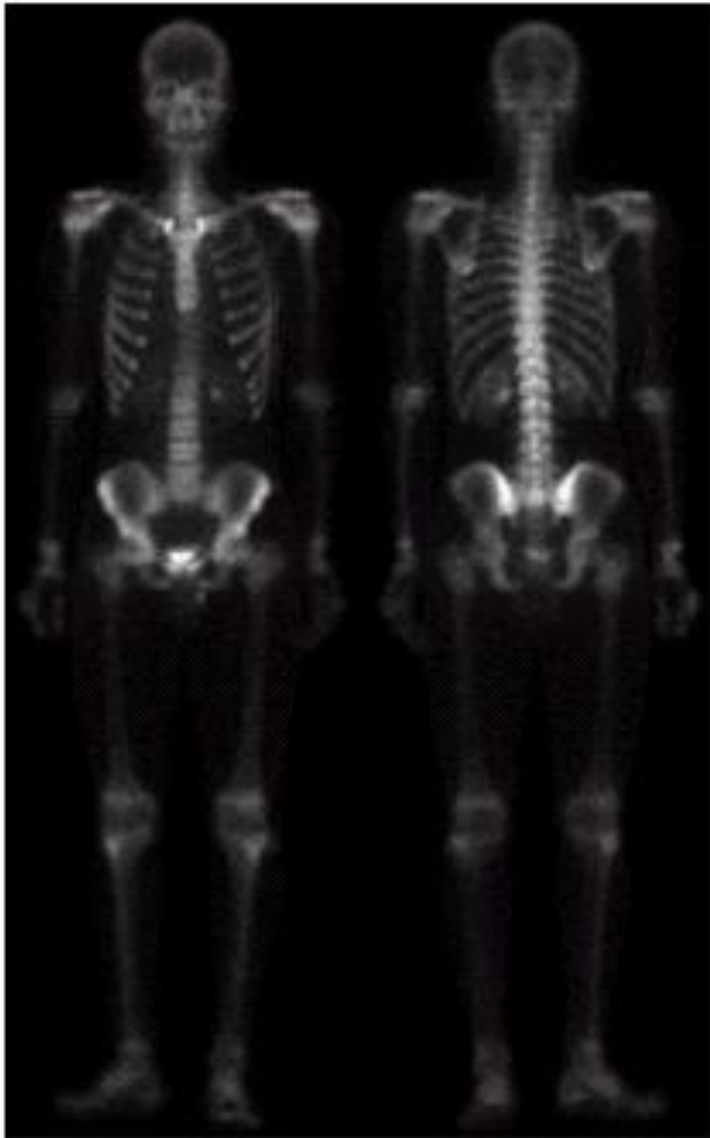
The reasons for doing this include:

- Highlighting interesting detail in images
- Removing noise from images
- Making images more visually appealing

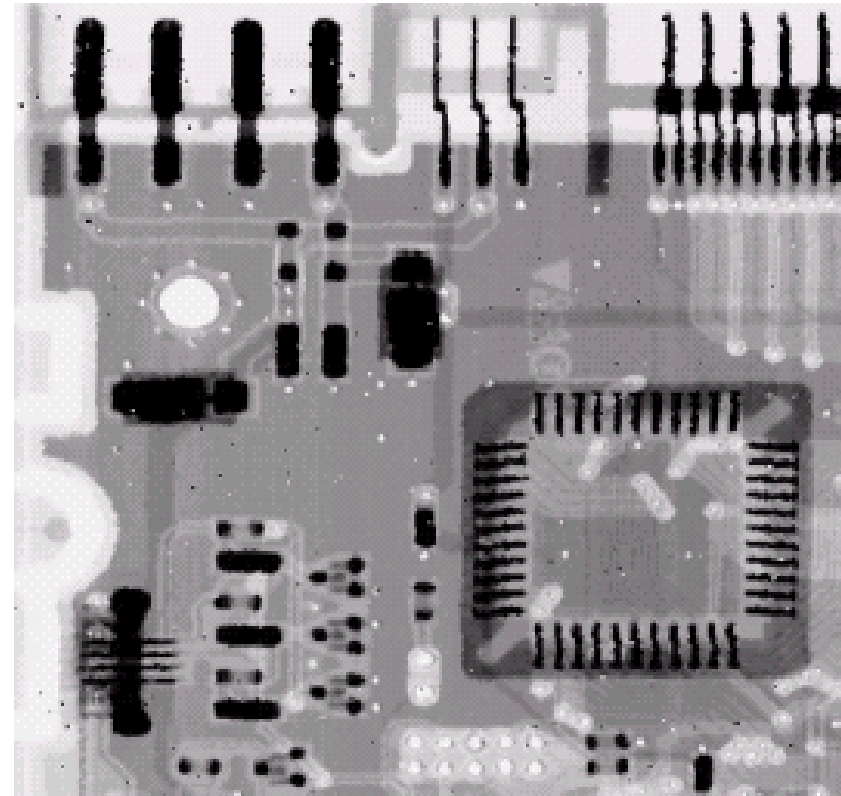
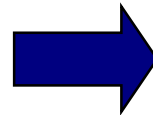
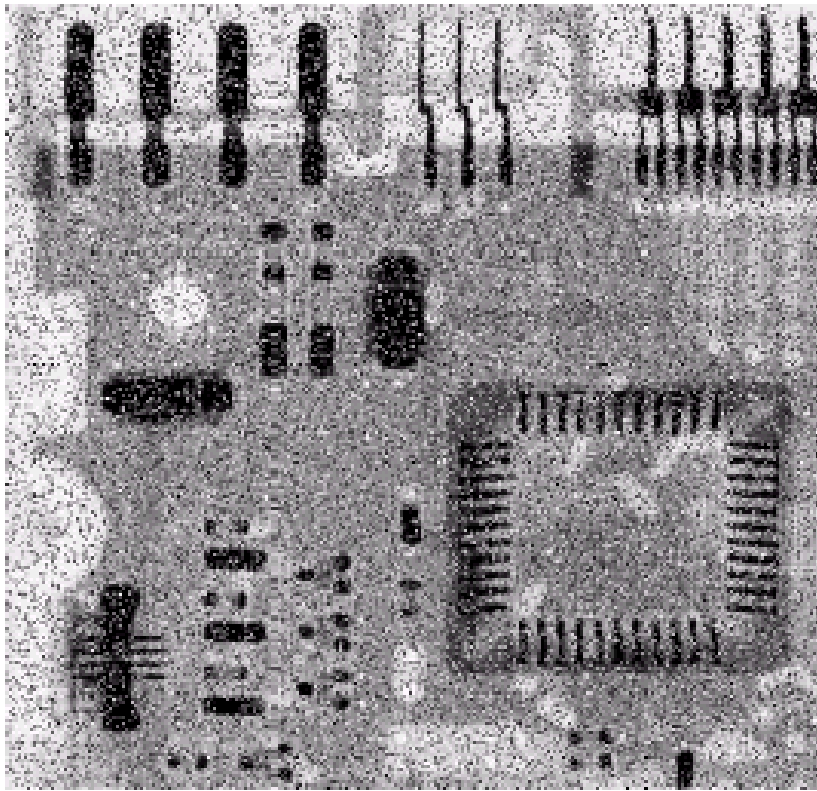
# Image Enhancement Examples



# Image Enhancement Examples (cont...)



# Image Enhancement Examples (cont...)





# Image Enhancement Examples (cont...)



# Spatial & Frequency Domains

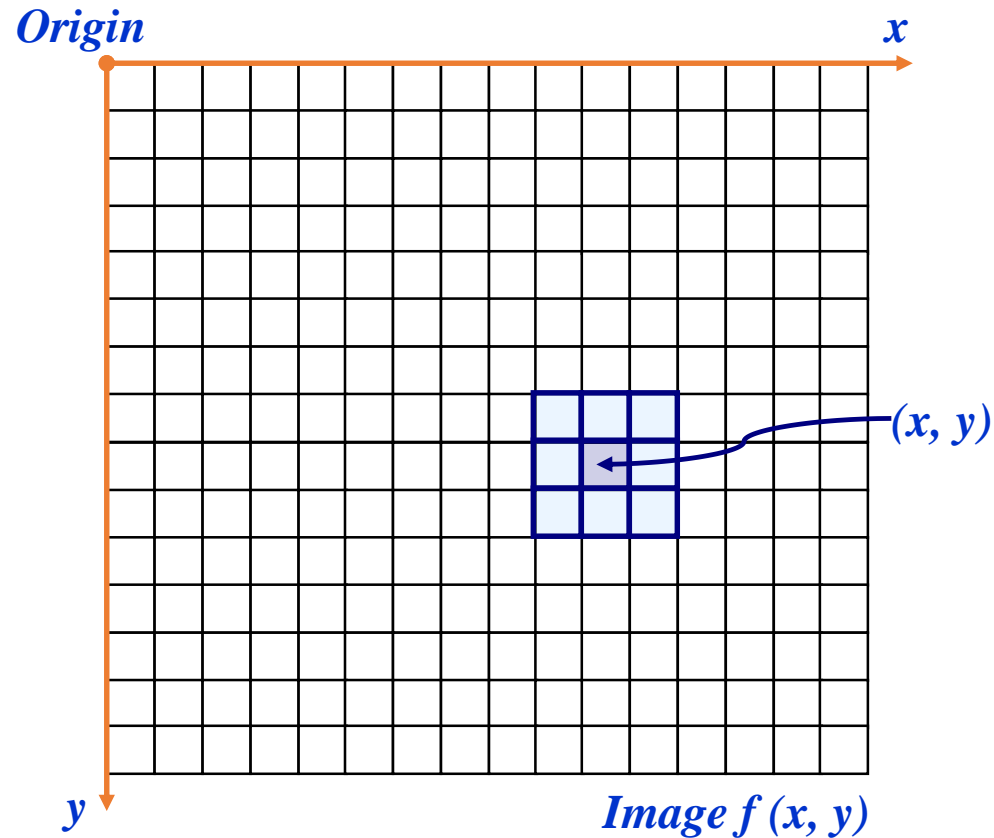
- There are two broad categories of image enhancement techniques
  - Spatial domain techniques
    - Direct manipulation of image pixels
  - Frequency domain techniques
    - Manipulation of Fourier transform or wavelet transform of an image
- For the moment we will concentrate on techniques that operate in the spatial domain

# Basic Spatial Domain Image Enhancement

- Most spatial domain enhancement operations can be reduced to the form

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

- where  $f(x, y)$  is the input image,  $g(x, y)$  is the processed image and  $T$  is some operator defined over some neighbourhood of  $(x, y)$



# Point Processing

- The simplest spatial domain operations occur when the neighbourhood is simply the pixel itself
- In this case  $T$  is referred to as a *grey level transformation function* or a *point processing operation*
- Point processing operations take the form

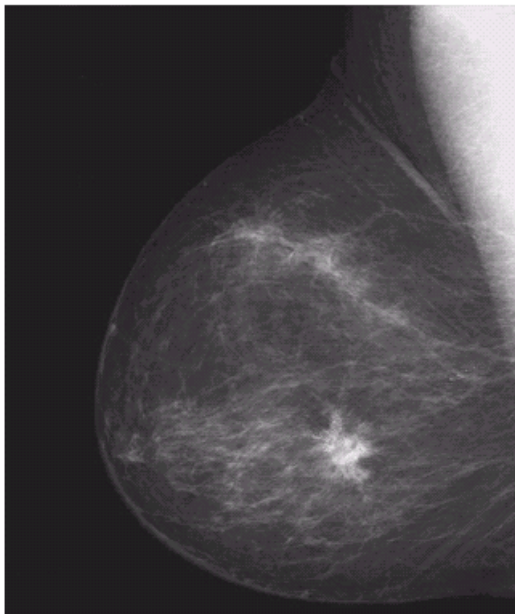
$$s = T( r )$$

- where  $S$  refers to the processed image pixel value and  $r$  refers to the original image pixel value

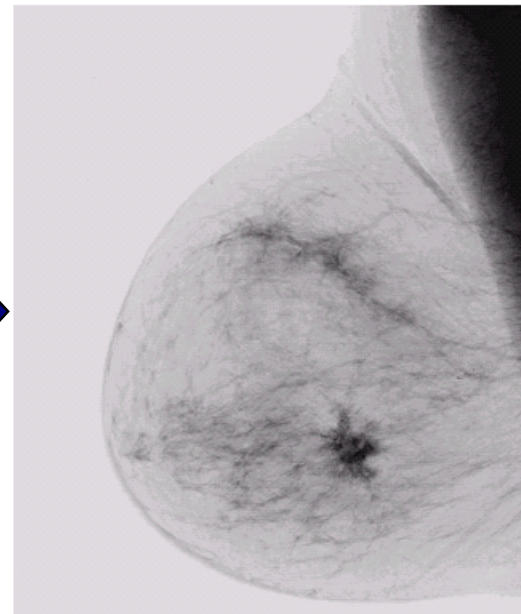
# Point Processing Example: Negative Images

- Negative images are useful for enhancing white or grey detail embedded in dark regions of an image
  - Note how much clearer the tissue is in the negative image of the mammogram below

Original  
Image



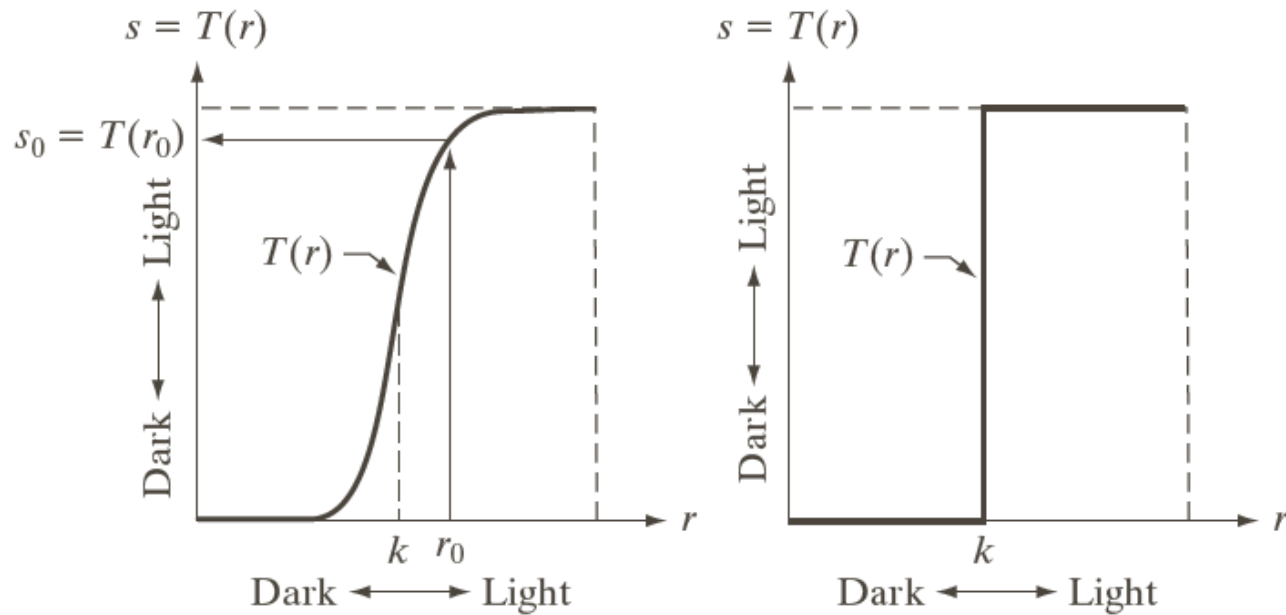
$$s = 1.0 - r$$



Negative  
Image

$$s = intensity_{max} - r$$

# Intensity Transformations



a b

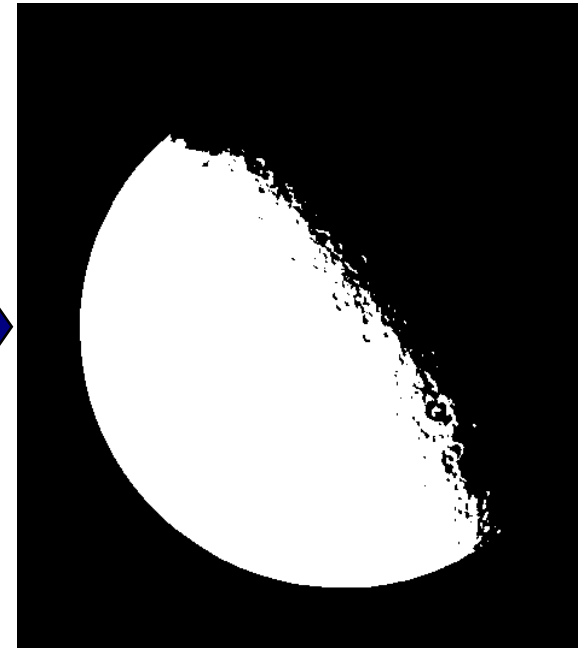
**FIGURE 3.2**  
Intensity transformation functions.  
(a) Contrast-stretching function.  
(b) Thresholding function.

# Point Processing Example: Thresholding

- Thresholding transformations are particularly useful for segmentation in which we want to isolate an object of interest from a background



$$s = \begin{cases} 1.0 & r > \text{threshold} \\ 0.0 & r \leq \text{threshold} \end{cases}$$

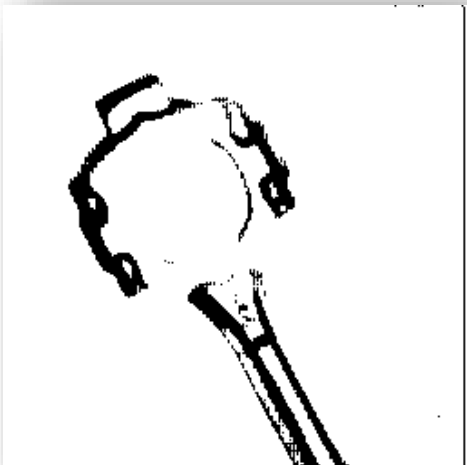
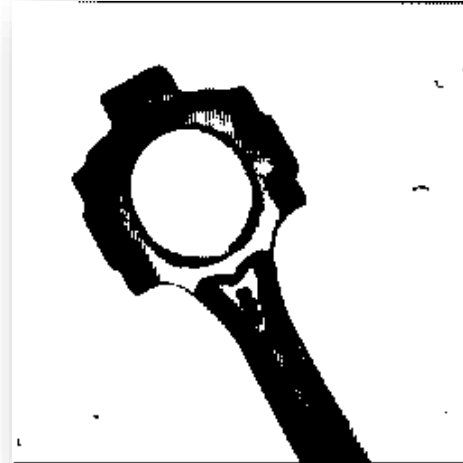


# Threshold Selection

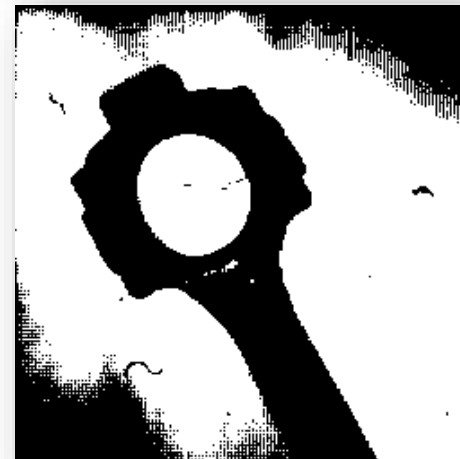
Original Image



Binary Image



Threshold too low



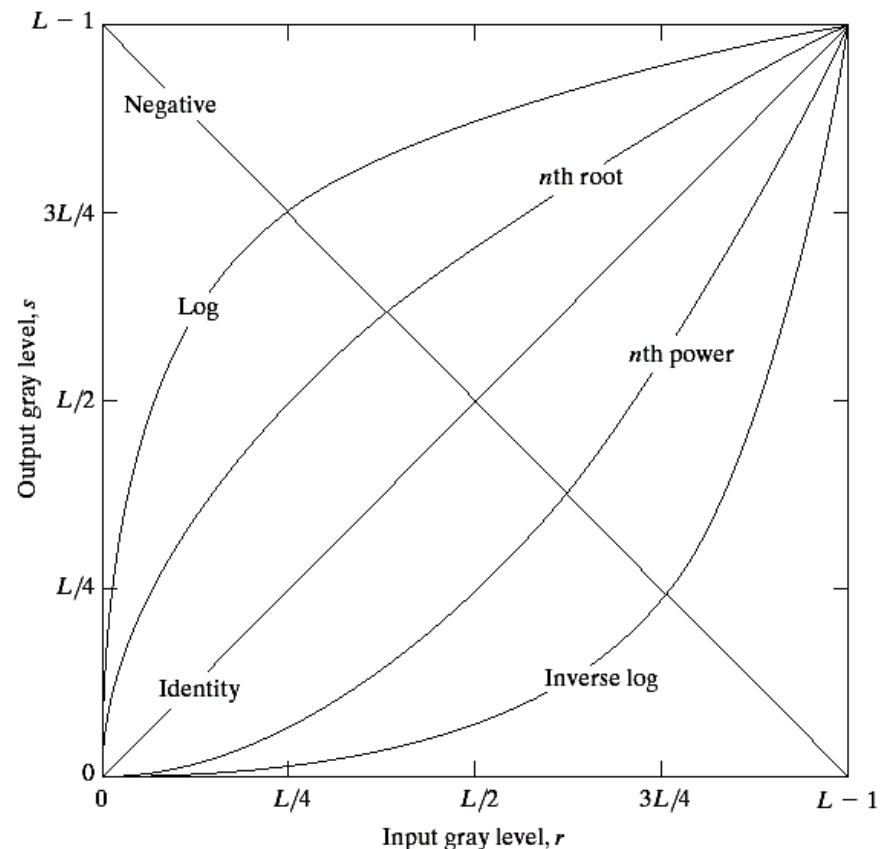
Threshold too high



# Basic Grey Level Transformations

- There are many different kinds of grey level transformations
- Three of the most common are shown here

- Linear
  - Negative/Identity
- Logarithmic
  - Log/Inverse log
- Power law
  - $n^{\text{th}}$  power/ $n^{\text{th}}$  root



# Logarithmic Transformations

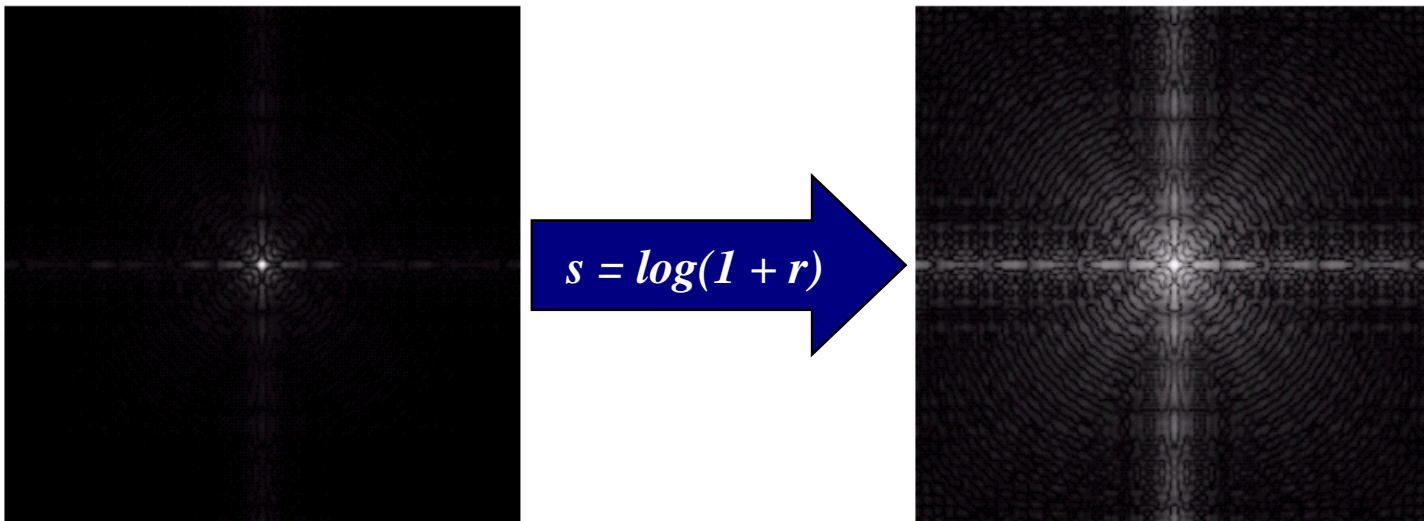
- The general form of the log transformation is

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

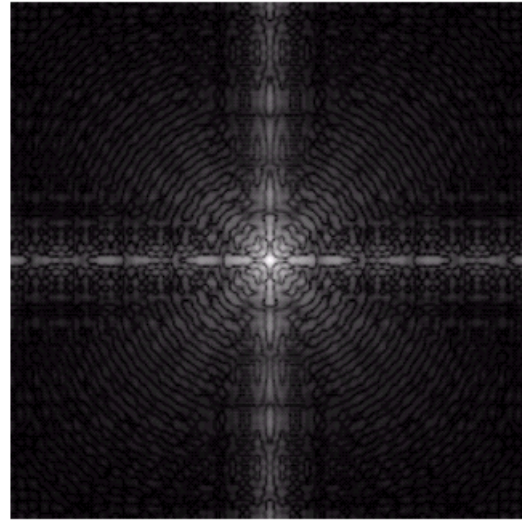
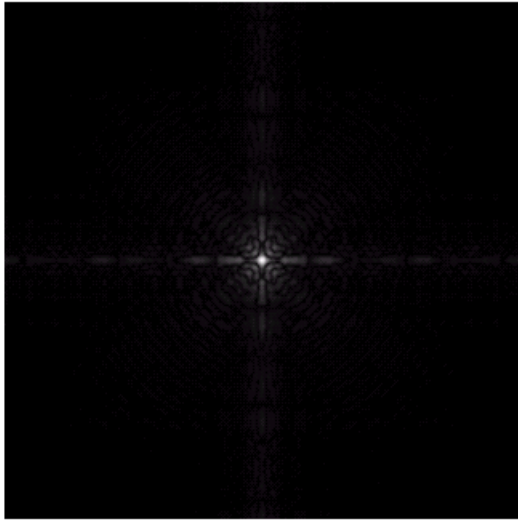
- The log transformation maps a narrow range of low input grey level values into a wider range of output values
- The inverse log transformation performs the opposite transformation

# Logarithmic Transformations (cont)

- Log functions are particularly useful when the input grey level values may have an extremely large range of values
- In the following example the Fourier transform of an image is put through a log transform to reveal more detail



# Logarithmic Transformations (cont...)



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

We usually set  $c$  to 1

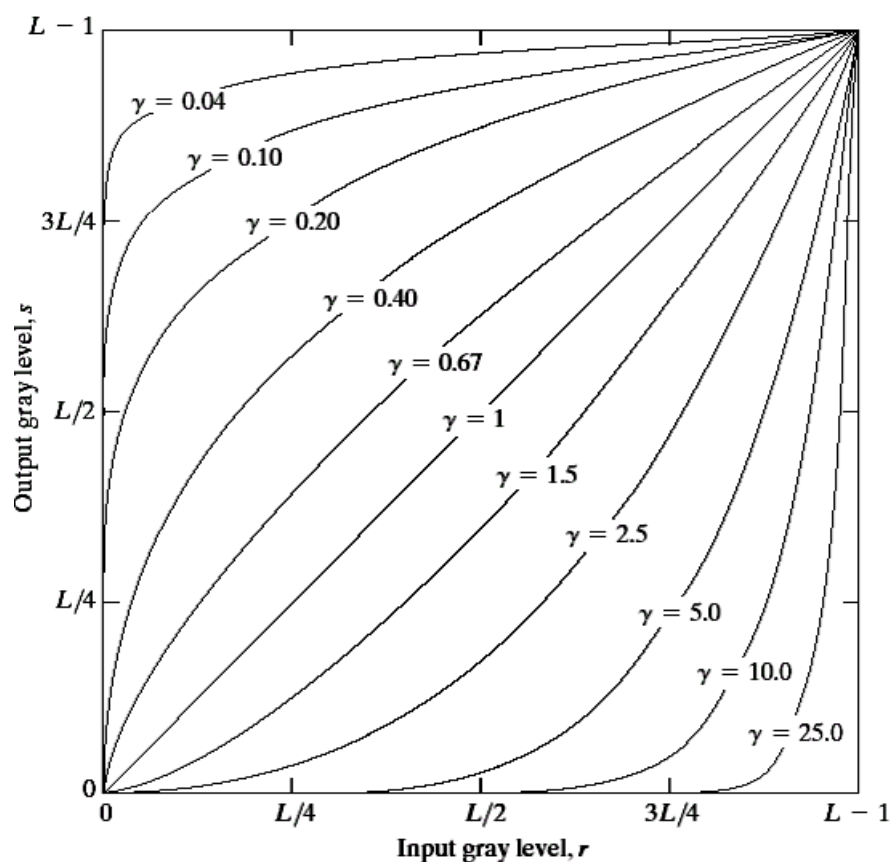
Grey levels must be in the range  $[0.0, 1.0]$

# Power Law (Gamma Correction) Transformations

- Power law transformations have the following form

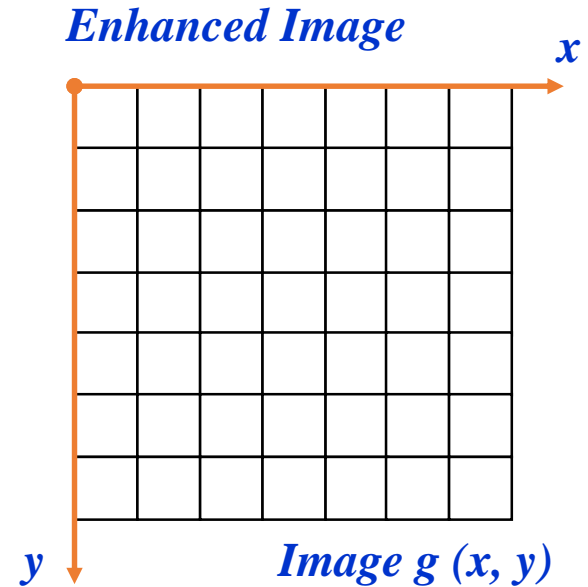
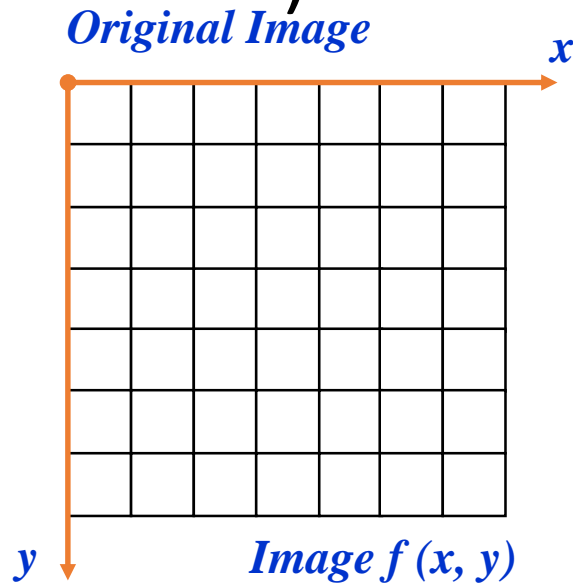
$$s = c * r^\gamma$$

- Map a narrow range of dark input values into a wider range of output values or vice versa
- Varying  $\gamma$  gives a whole family of curves



**FIGURE 3.6** Plots of the equation  $s = cr^\gamma$  for various values of  $\gamma$  ( $c = 1$  in all cases). All curves were scaled to fit in the range shown.

# Power Law Transformations (cont...)



$$s = r^\gamma$$

- We usually set  $C$  to 1
- Grey levels must be in the range  $[0.0, 1.0]$

# Power Law Example



# Power Law Example (cont...)

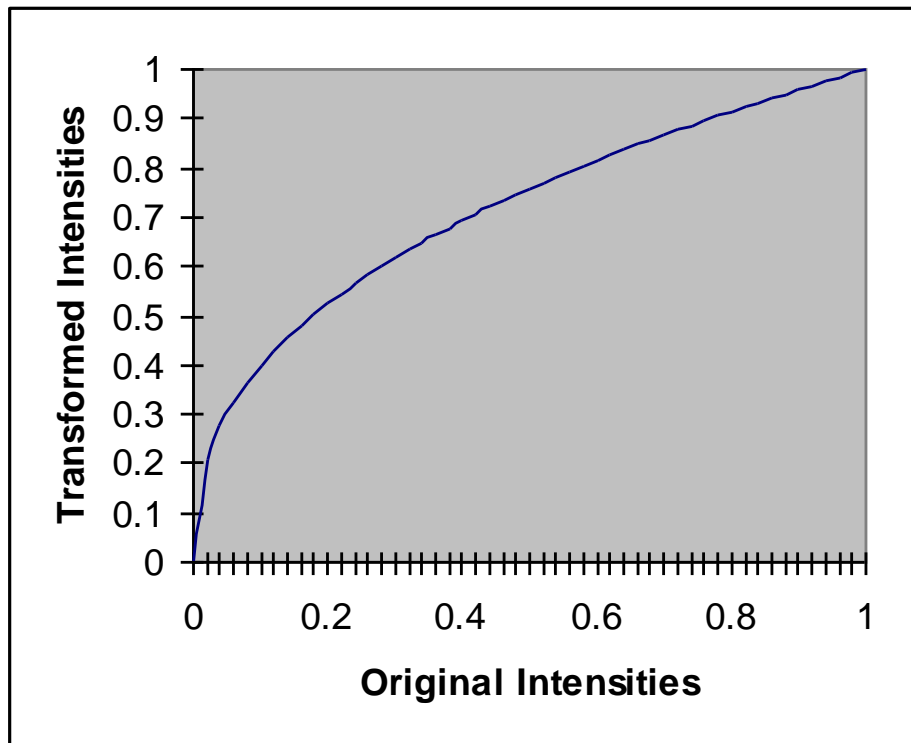
$$\gamma = 0.6$$





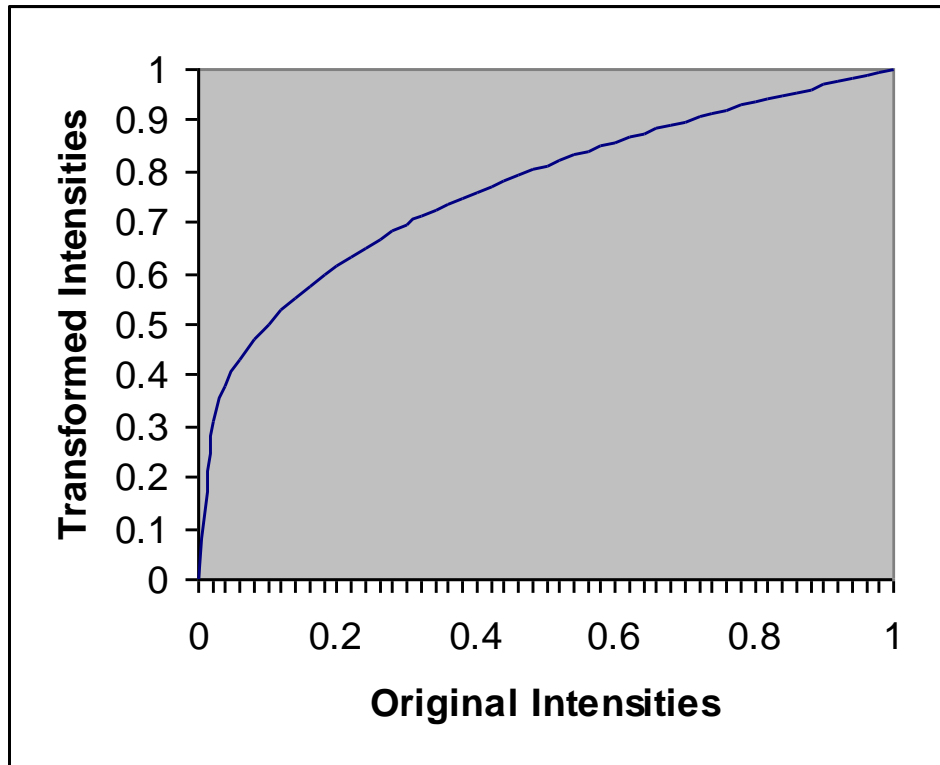
# Power Law Example (cont...)

$$\gamma = 0.4$$



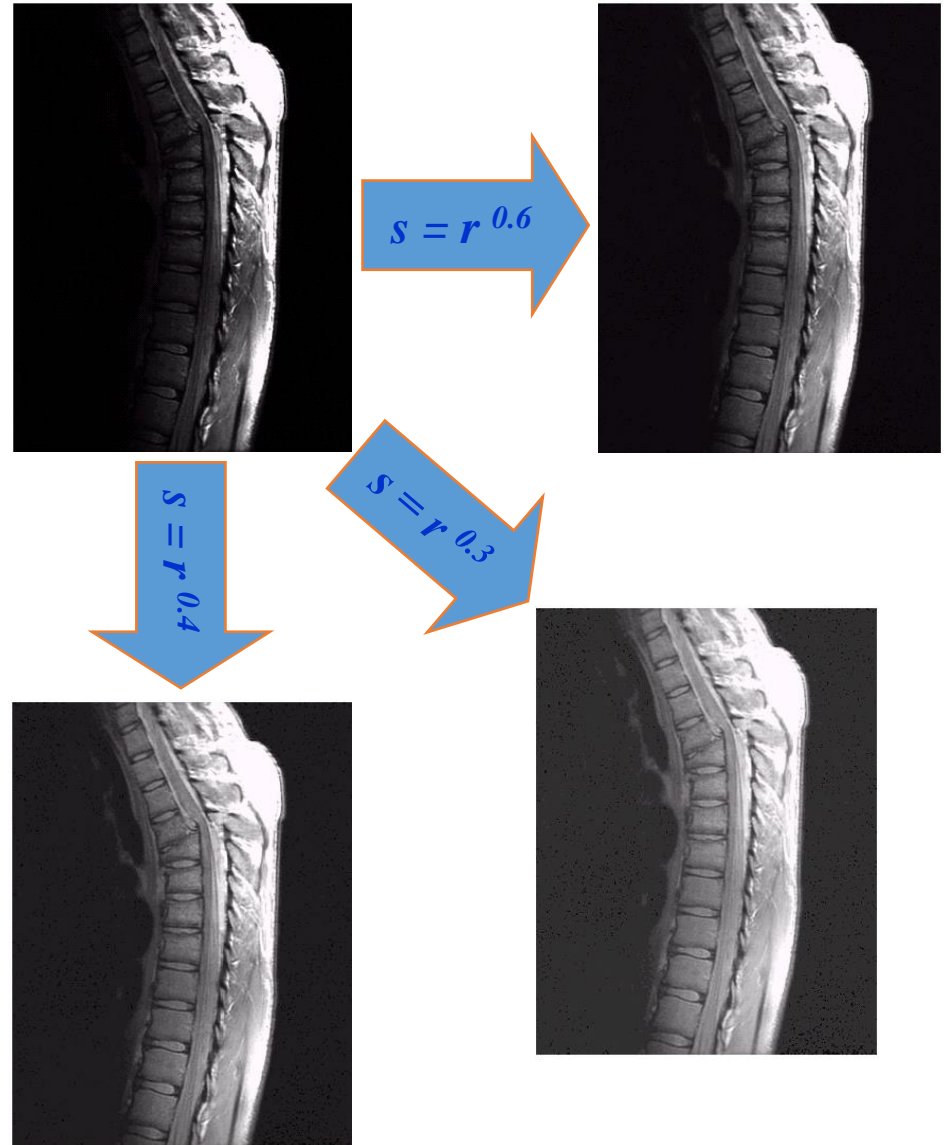
# Power Law Example (cont...)

$$\gamma = 0.3$$



# Power Law Example (cont...)

- The images to the right show a magnetic resonance (MR) image of a fractured human spine
- Different curves highlight different detail

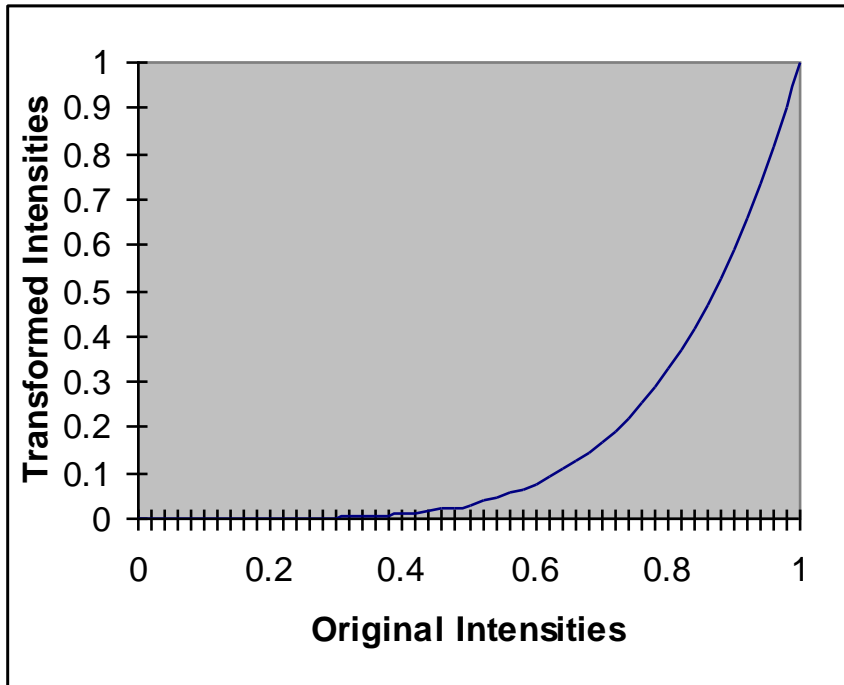


# Power Law Example



# Power Law Example (cont...)

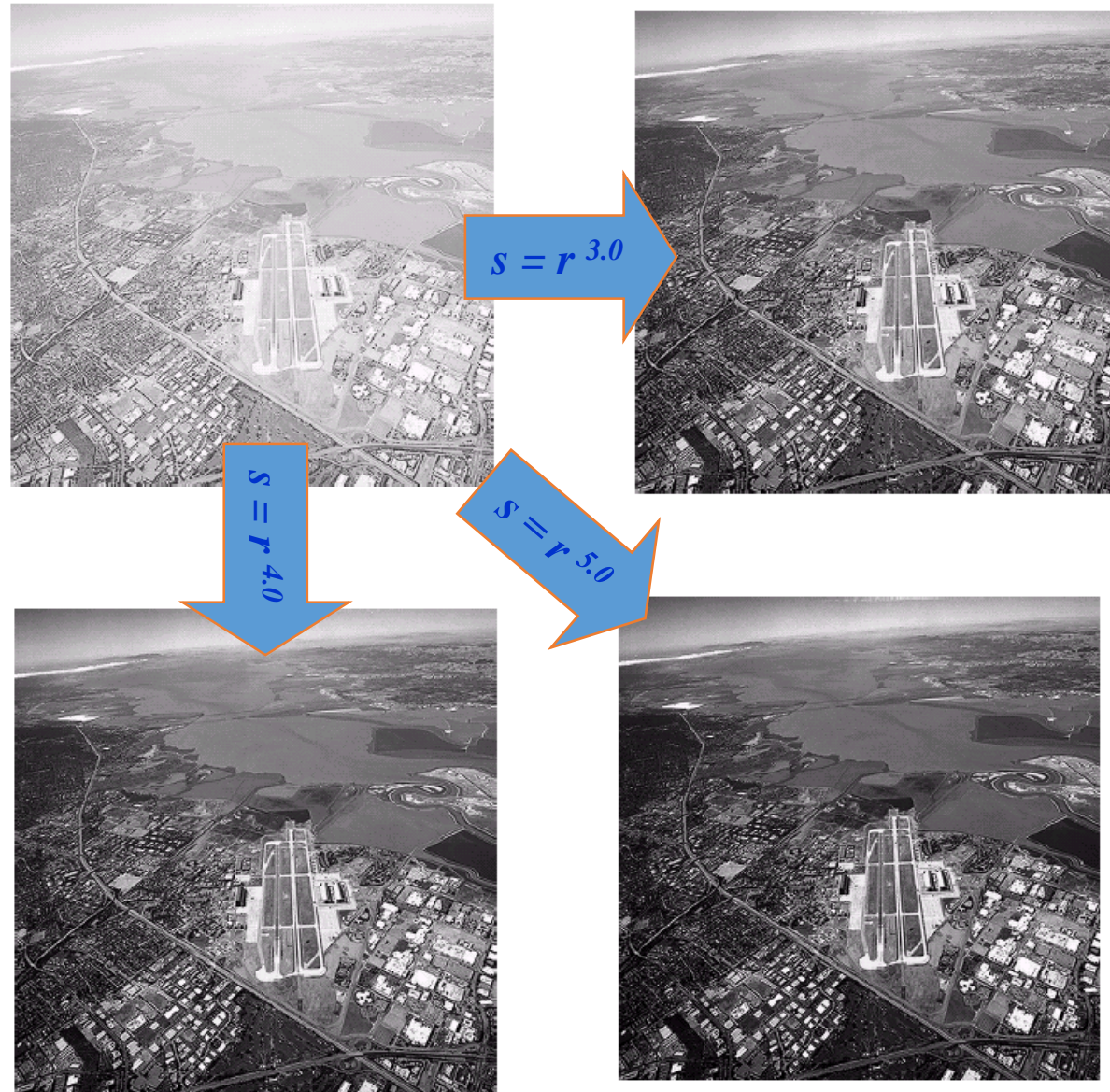
$$\gamma = 5.0$$





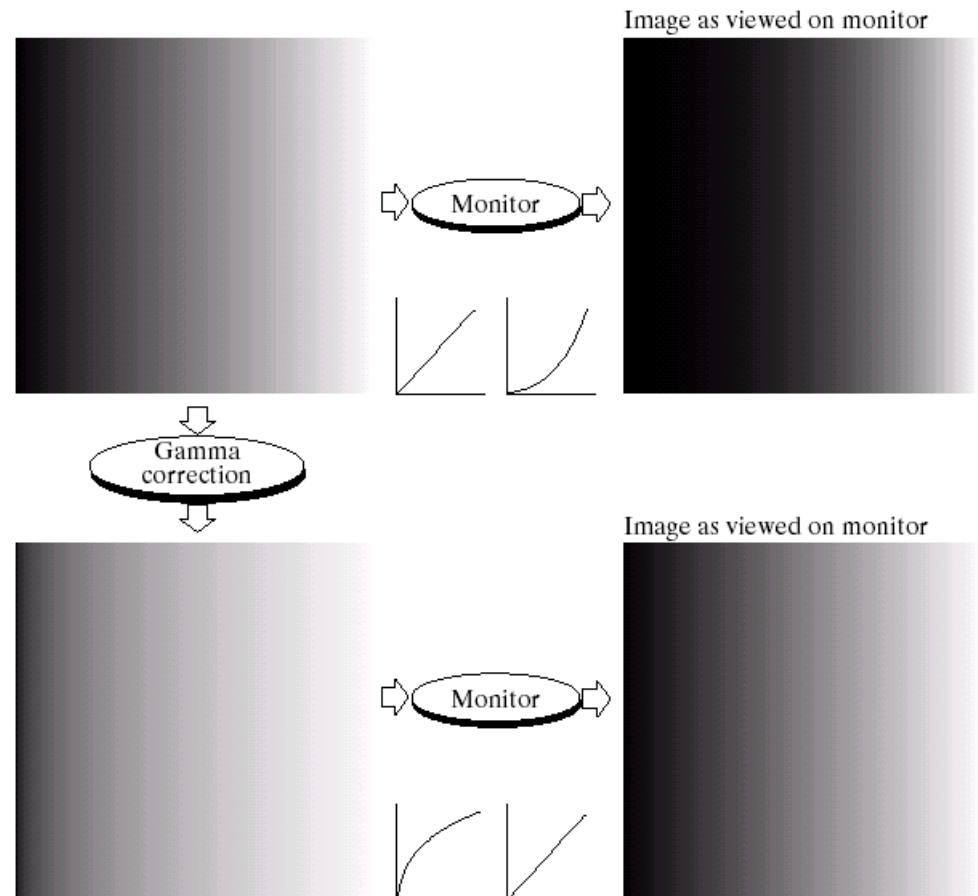
# Power Law Transformations (cont...)

- An aerial photo of a runway is shown
- This time power law transforms are used to darken the image
- Different curves highlight different detail



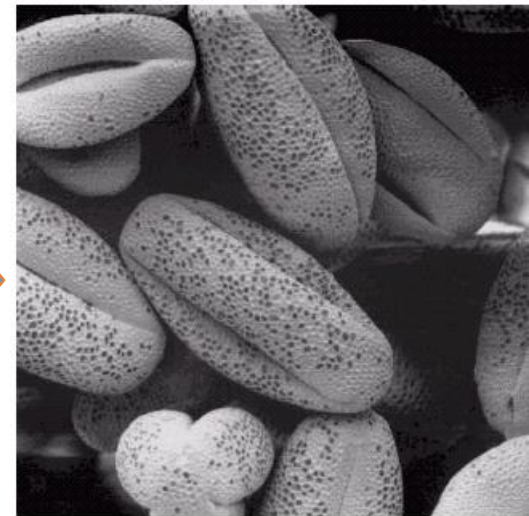
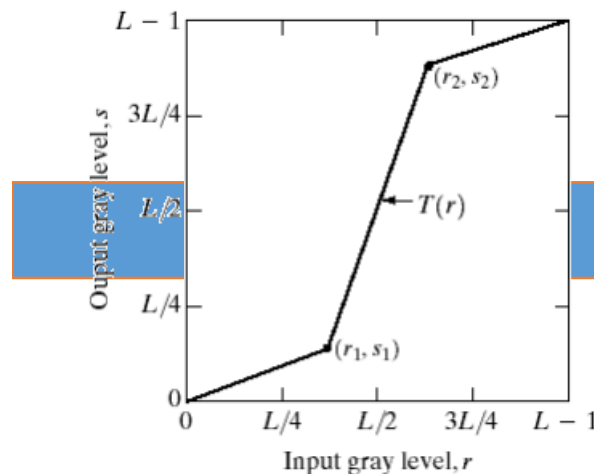
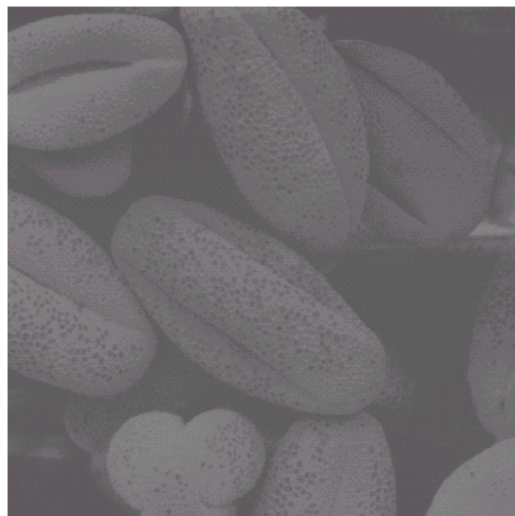
# Gamma Correction

- Many of you might be familiar with gamma correction of computer monitors
- Problem is that display devices do not respond linearly to different intensities



# Piecewise Linear Transformation Functions

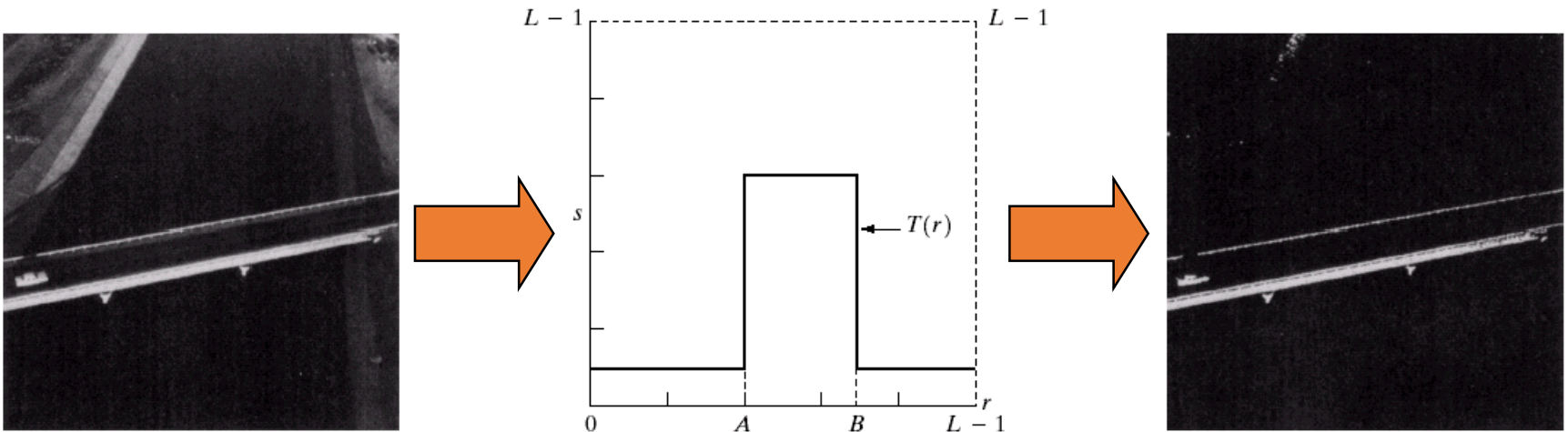
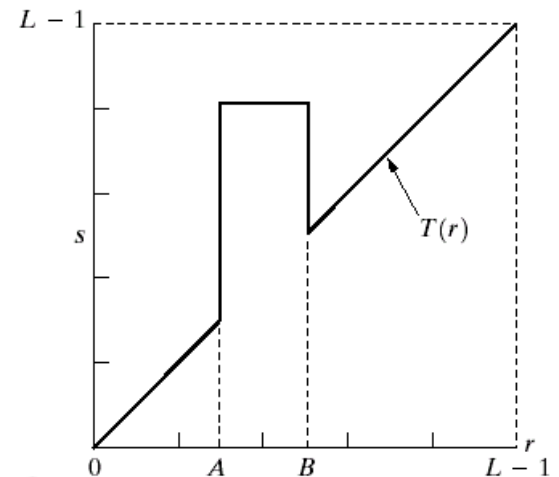
- Rather than using a well defined mathematical function we can use arbitrary user-defined transforms
- The images below show a contrast stretching linear transform to add contrast to a poor quality image





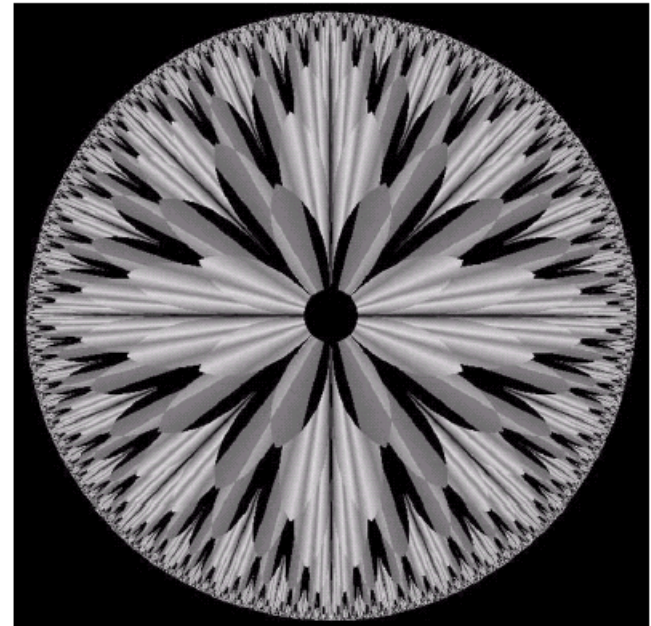
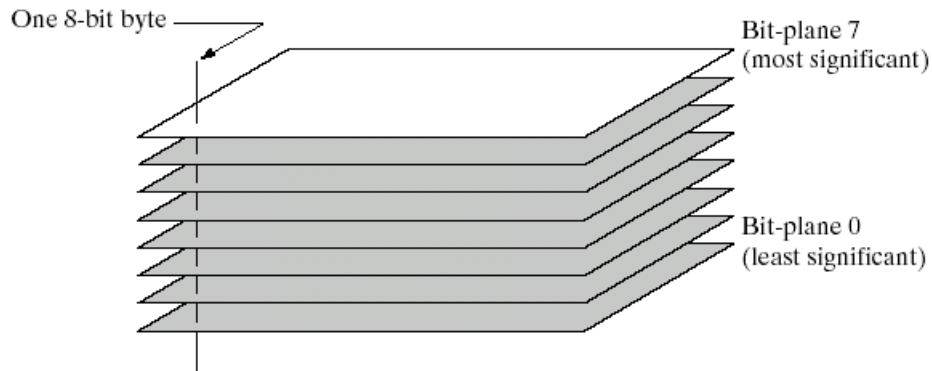
# Gray Level Slicing

- Highlights a specific range of grey levels
  - Similar to thresholding
  - Other levels can be suppressed or maintained
  - Useful for highlighting features in an image



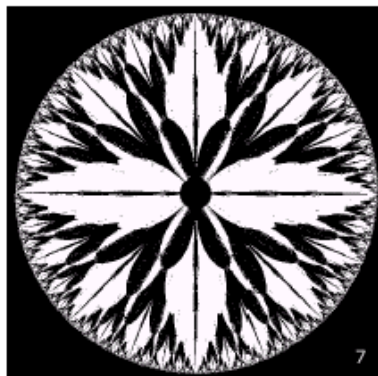
# Bit Plane Slicing

- By isolating particular bits of the pixel values in an image we can highlight interesting aspects of that image
  - Higher-order bits usually contain most of the significant visual information
  - Lower-order bits contain subtle details

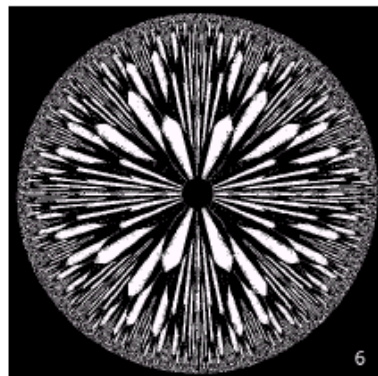


# Bit Plane Slicing (cont...)

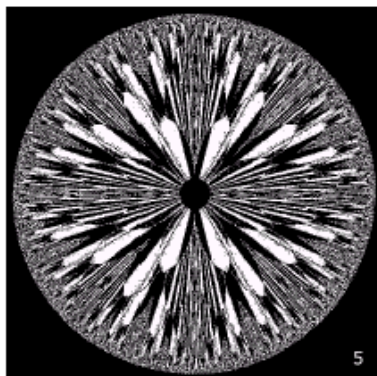
[10000000]



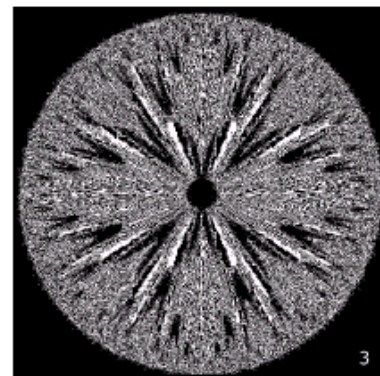
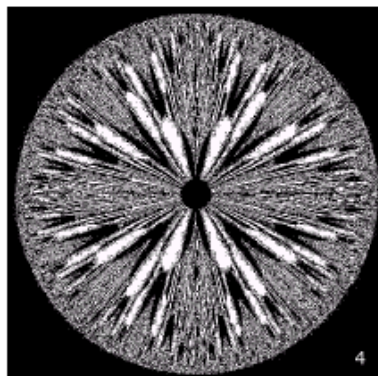
[01000000]



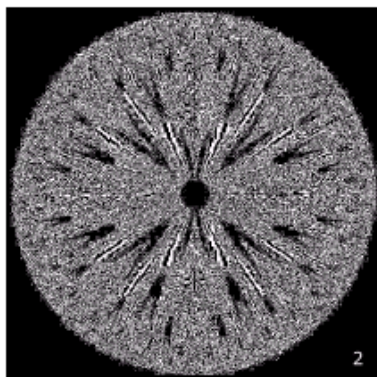
[00100000]



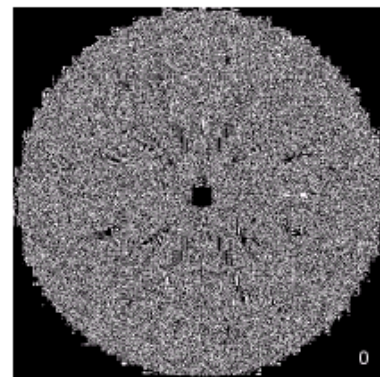
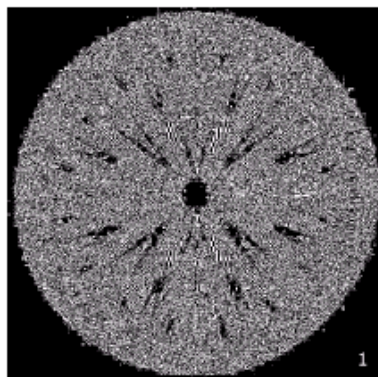
[00001000]



[00000100]



[00000001]





# Bit Plane Slicing (cont...)



a	b	c
d	e	f
g	h	i

**FIGURE 3.14** (a) An 8-bit gray-scale image of size  $500 \times 1192$  pixels. (b) through (i) Bit planes 1 through 8, with bit plane 1 corresponding to the least significant bit. Each bit plane is a binary image.

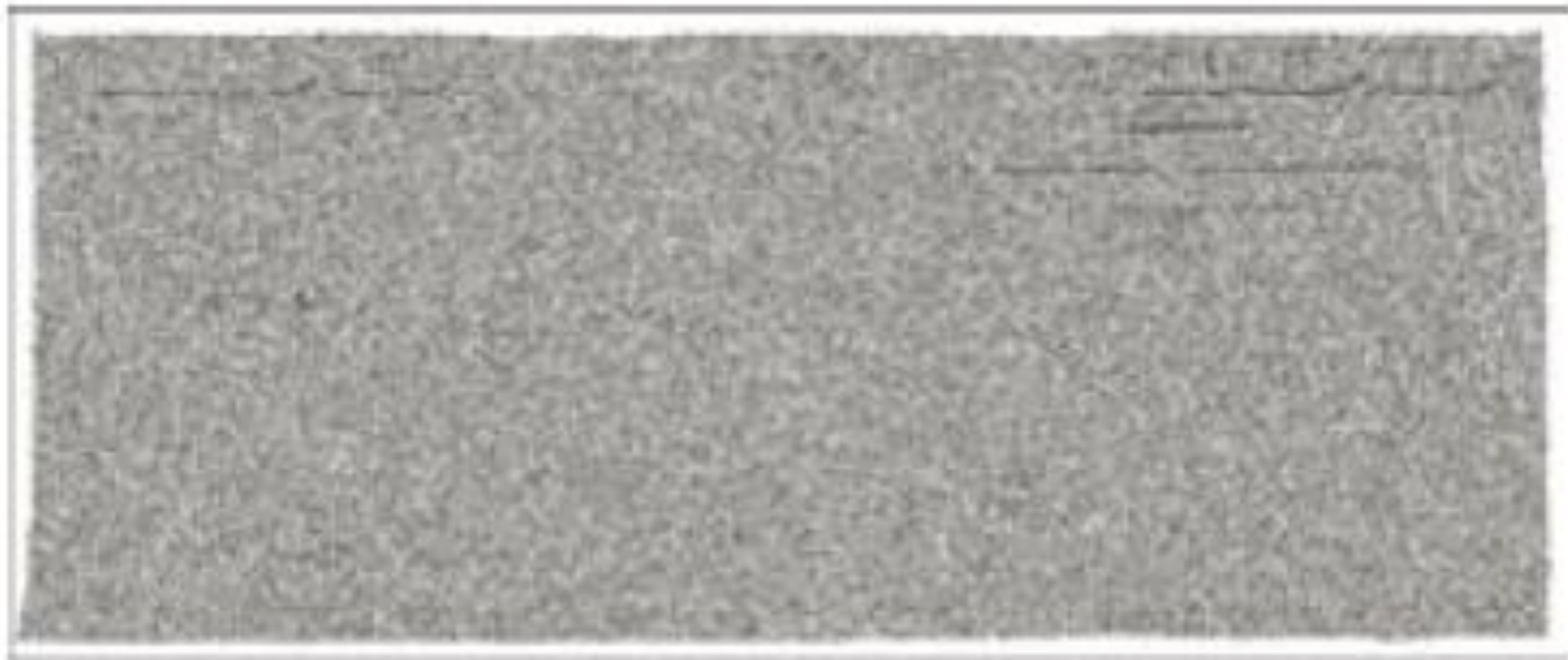
# Bit Plane Slicing (cont...)



# Bit Plane Slicing (cont...)

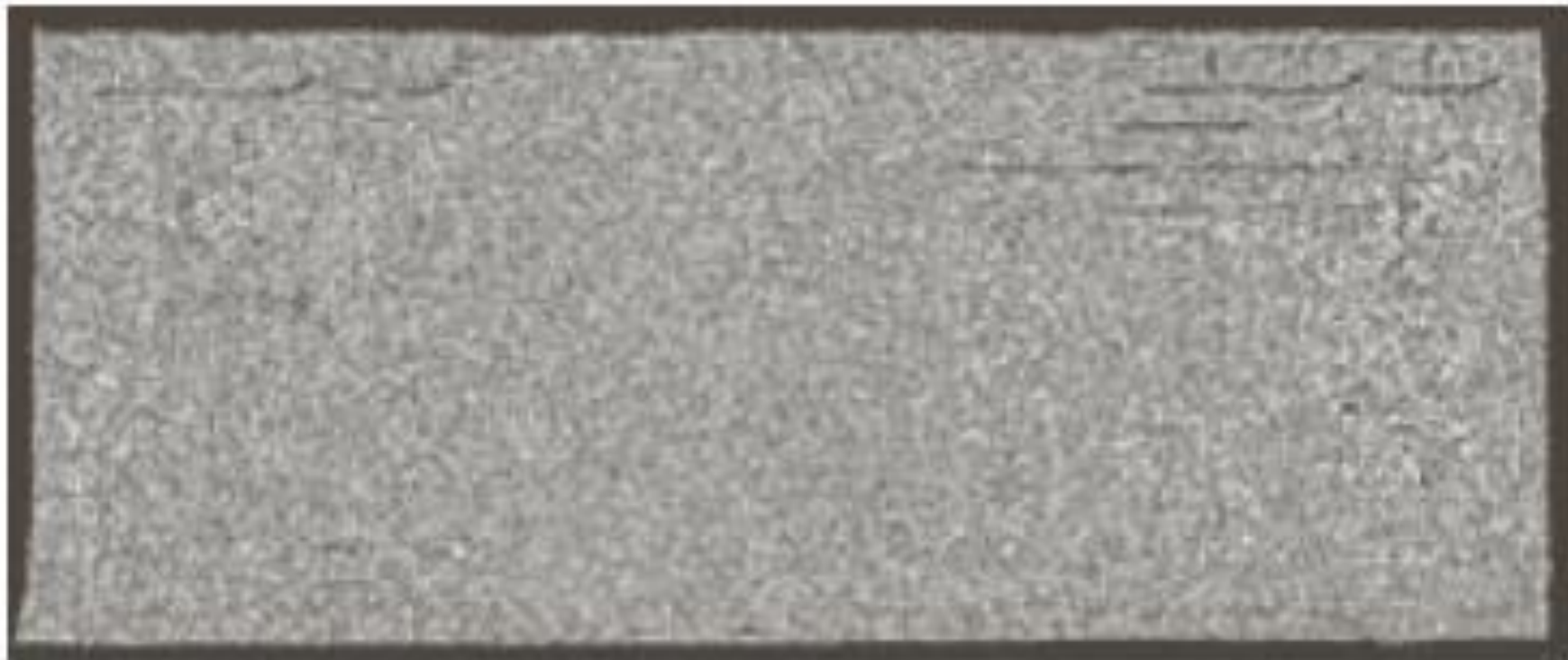


# Bit Plane Slicing (cont...)



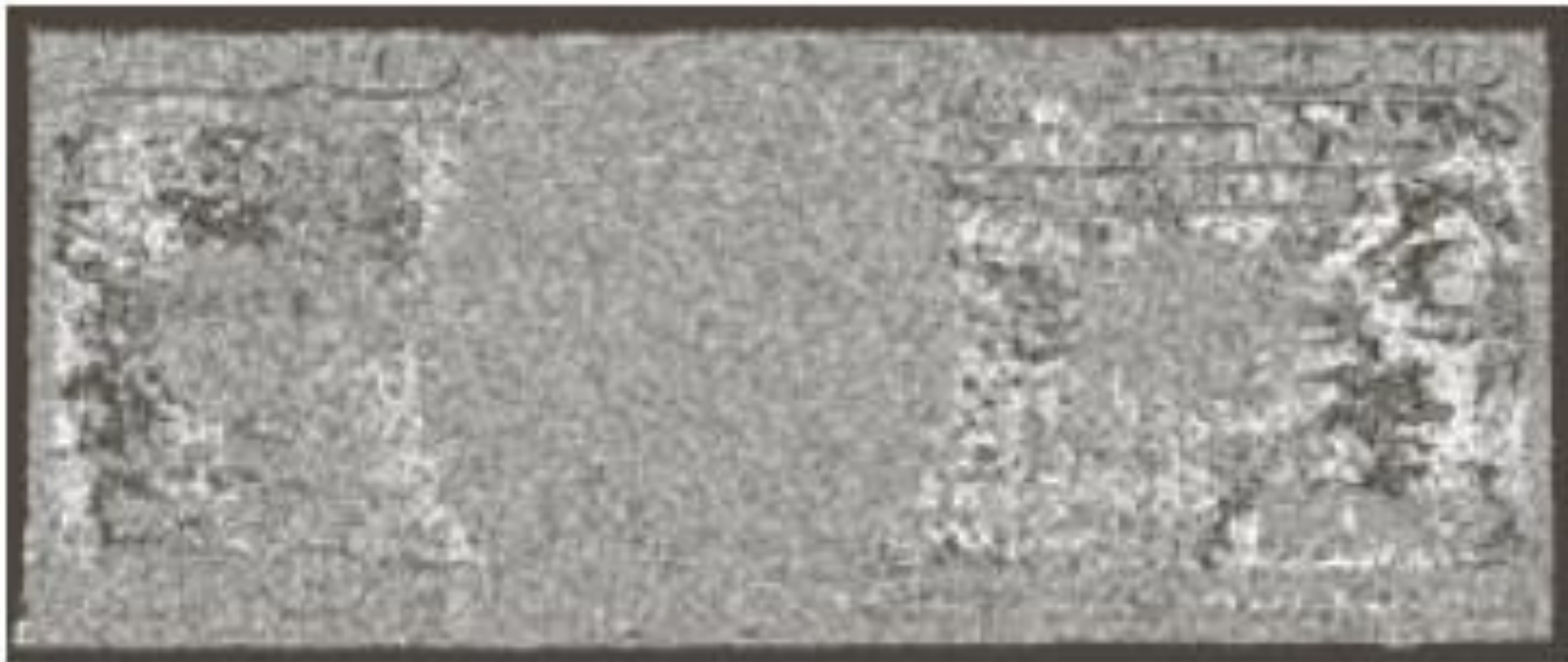


# Bit Plane Slicing (cont...)





# Bit Plane Slicing (cont...)



# Bit Plane Slicing (cont...)



# Bit Plane Slicing (cont...)



# Bit Plane Slicing (cont...)



## Bit Plane Slicing (cont...)





# Bit Plane Slicing (cont...)



Reconstructed image  
using only bit planes 8  
and 7

# Bit Plane Slicing (cont...)



Reconstructed image  
using only bit planes 8, 7  
and 6

# Bit Plane Slicing (cont...)



Reconstructed image  
using only bit planes 7, 6  
and 5

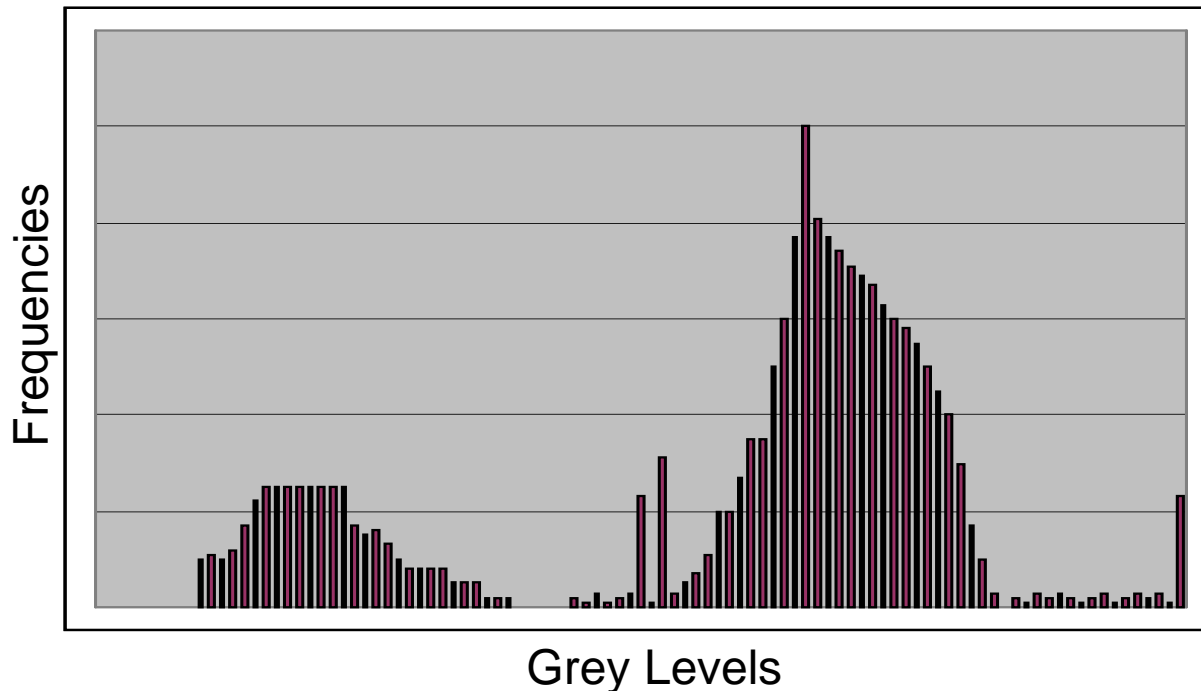


# Contents

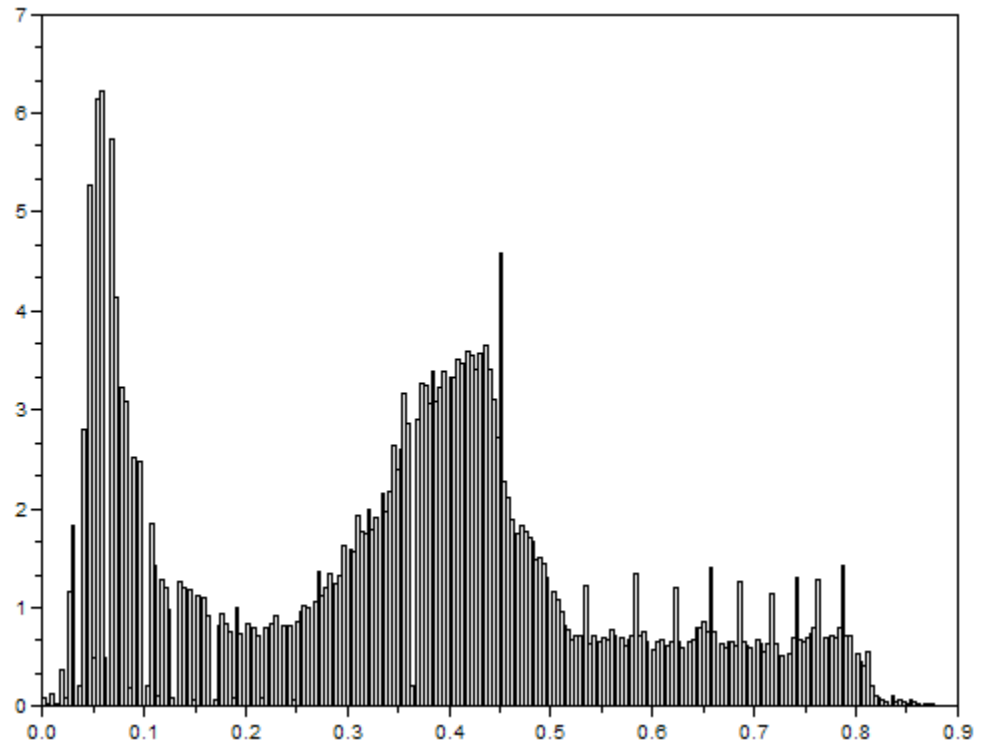
- Point processing techniques
  - Negative images
  - Thresholding
  - Logarithmic transformation
  - Power law transforms
  - Grey level slicing
  - Bit plane slicing
- Image enhancement techniques working in the spatial domain:
  - Histogram processing
  - Histogram Equalization
  - Histogram Matching

# Image Histograms

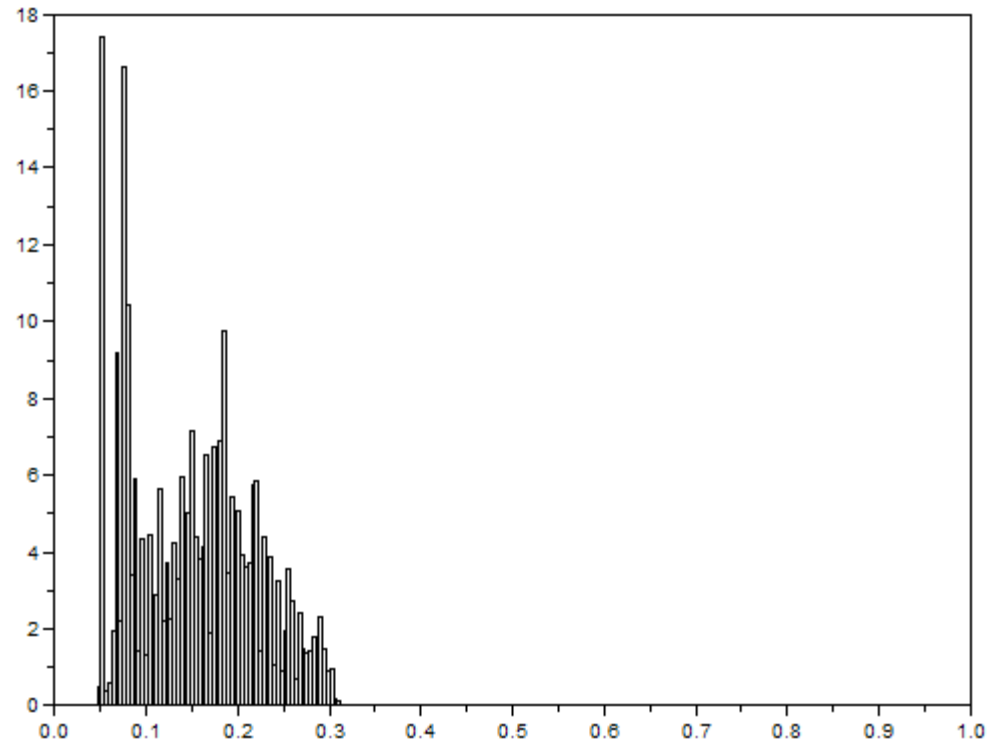
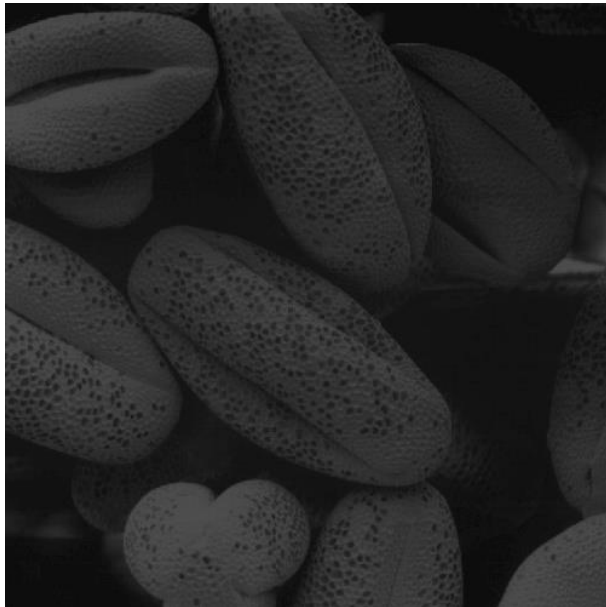
- The histogram of an image shows us the distribution of grey levels in the image
- Massively useful in image processing, especially in segmentation and enhancement



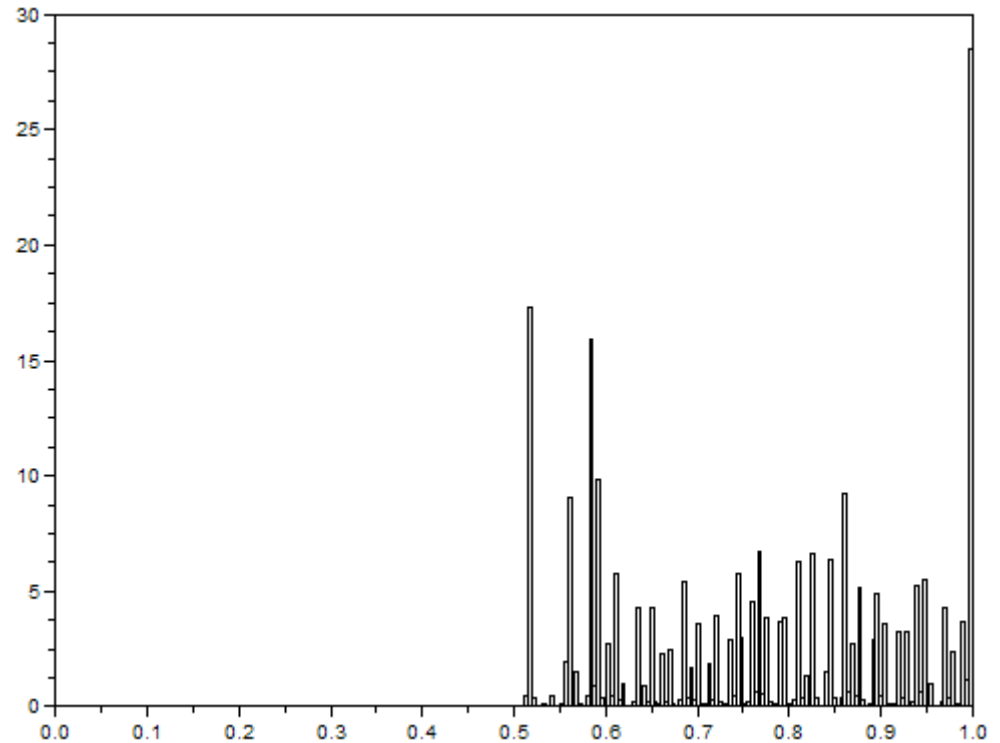
# Histogram Examples



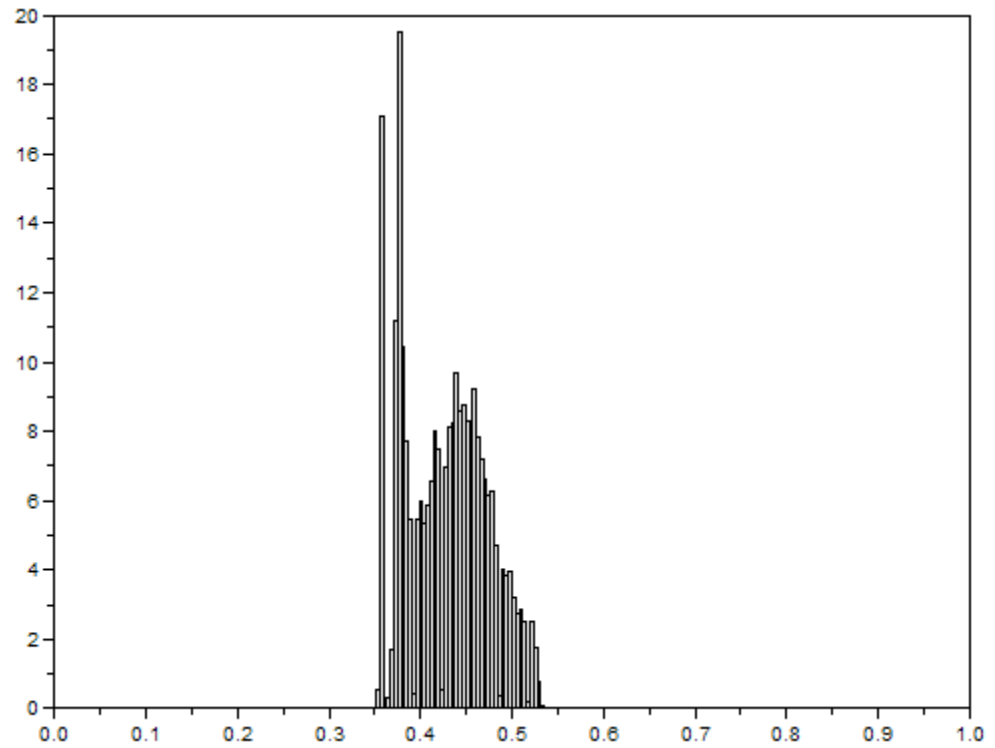
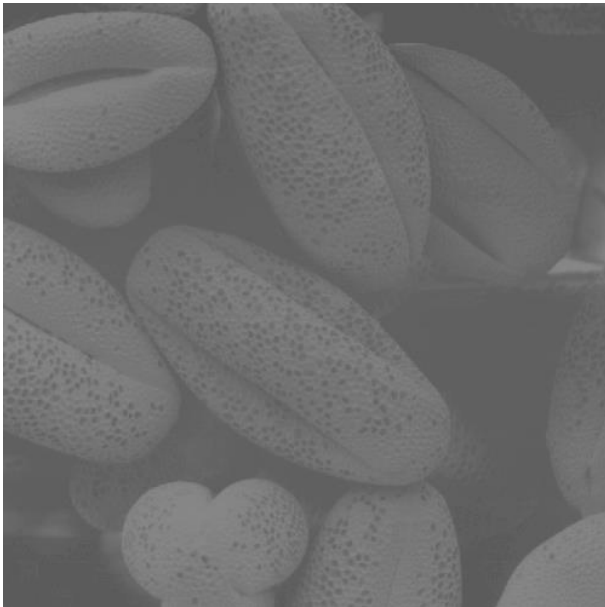
# Histogram Examples (cont...)



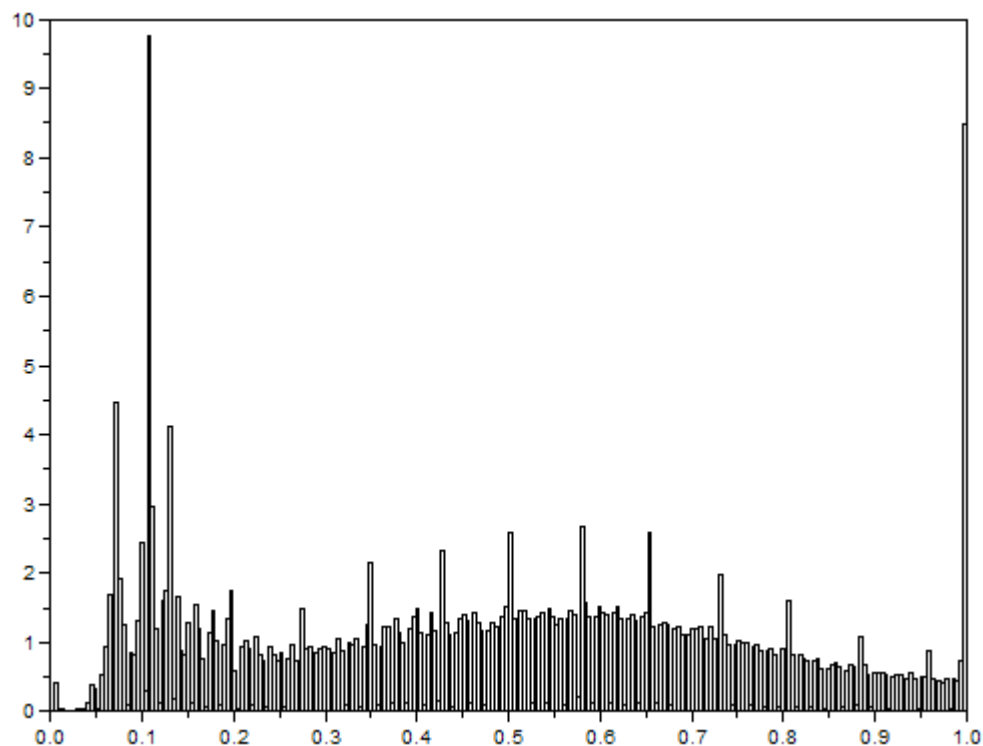
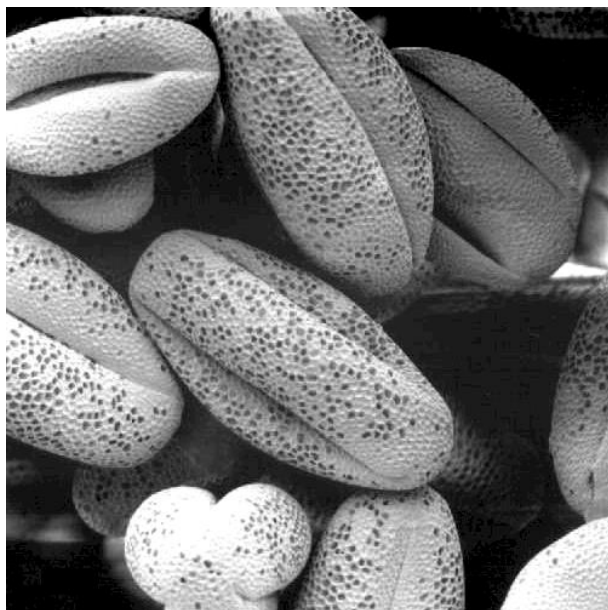
# Histogram Examples (cont...)



# Histogram Examples (cont...)s

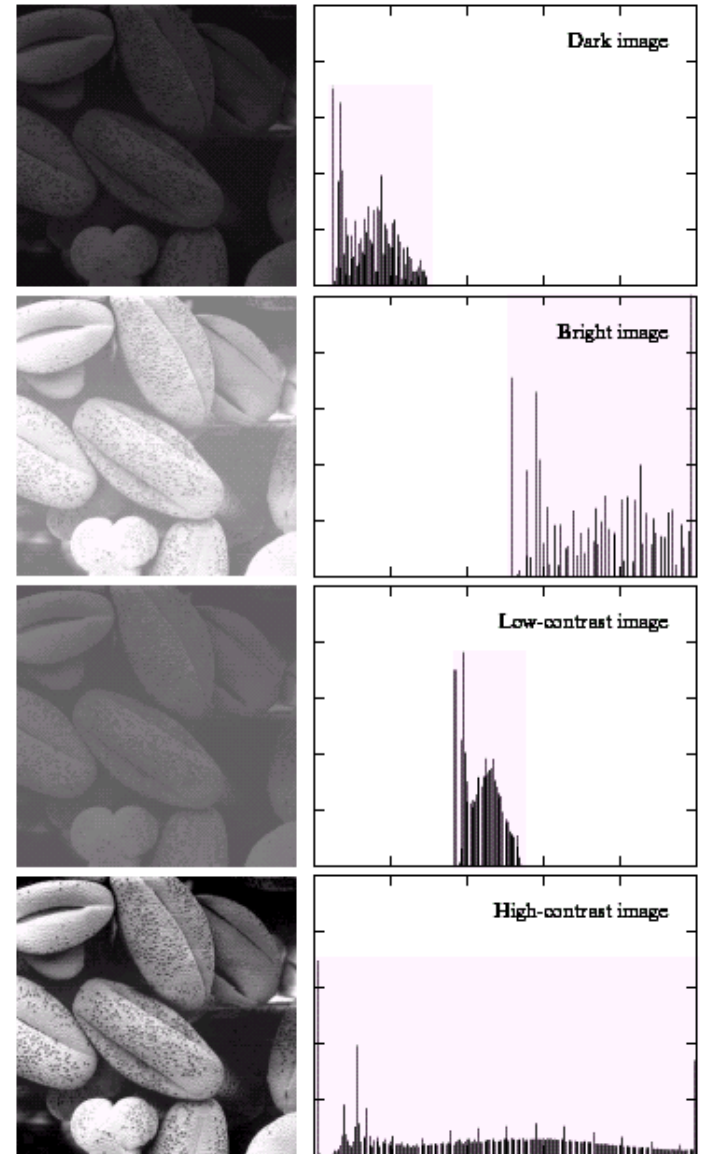


# Histogram Examples (cont...)



# Histogram Examples (cont...)

- A selection of images and their histograms
- Notice the relationships between the images and their histograms
- Note that the high contrast image has the most evenly spaced histogram





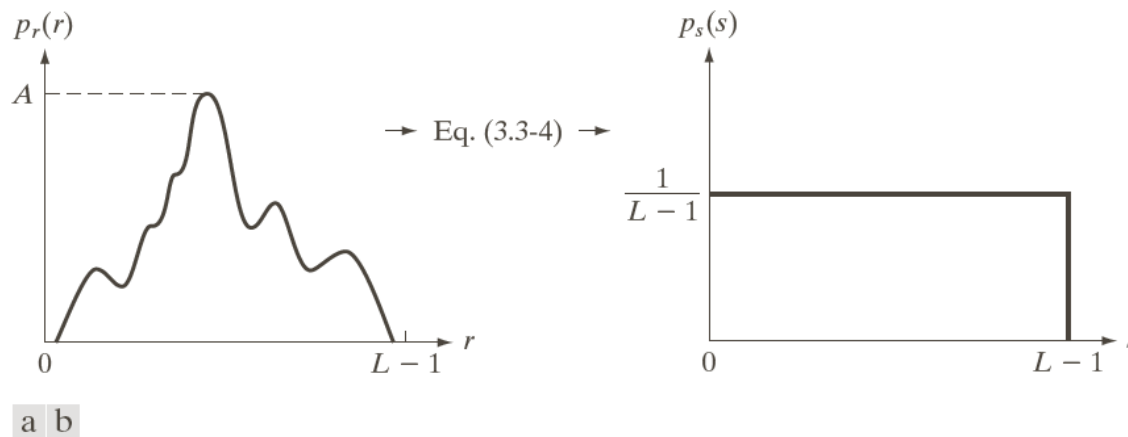
# Contrast Stretching

- We can fix images that have poor contrast by applying a pretty simple contrast specification
- The interesting part is how do we decide on this transformation function?



# Histogram Equalisation

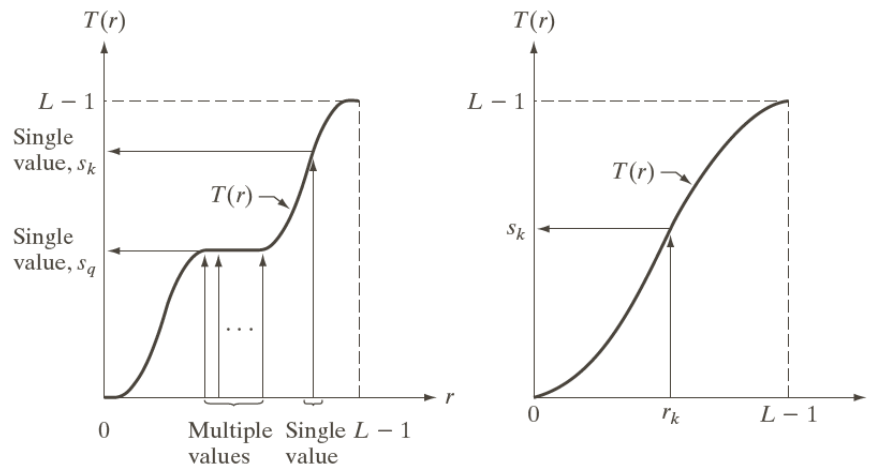
- Spreading out the frequencies in an image (or equalising the image) is a simple way to improve dark or washed out images



**FIGURE 3.18** (a) An arbitrary PDF. (b) Result of applying the transformation in Eq. (3.3-4) to all intensity levels,  $r$ . The resulting intensities,  $s$ , have a uniform PDF, independently of the form of the PDF of the  $r$ 's.

# Histogram Equalisation

- Spreading out the frequencies in an image (or equalising the image) is a simple way to improve dark or washed out images



a b

**FIGURE 3.17**

(a) Monotonically increasing function, showing how multiple values can map to a single value. (b) Strictly monotonically increasing function. This is a one-to-one mapping, both ways.

# Histogram Equalisation

- The formula for histogram equalisation is given where

- $r_k$ : input intensity
- $s_k$ : processed intensity
- $k$ : the intensity range (0 ... L-1)  
(e.g 0 – 255)
- $n_j$ : the frequency of intensity  $j$
- $n$ : the sum of all frequencies

$$\begin{aligned} s_k &= T(r_k) \\ &= (L - 1) \sum_{j=0}^k p_r(r_j) \\ &= (L - 1) \sum_{j=0}^k \frac{n_j}{n} \end{aligned}$$

# Example (3.5 G&W)

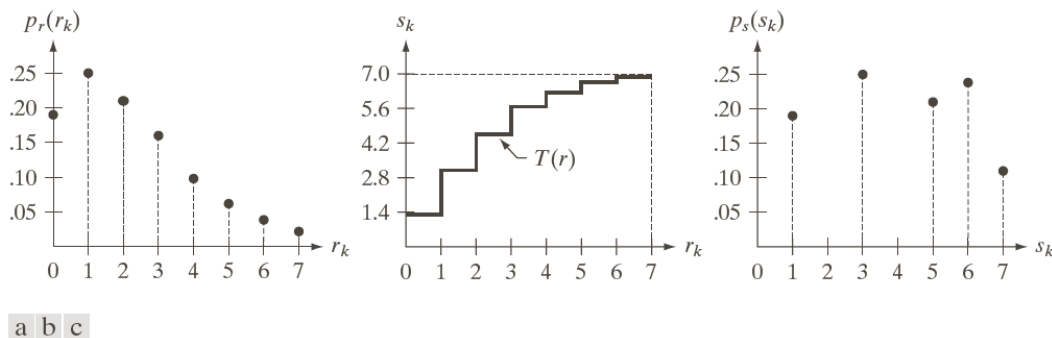
$r_k$	$n_k$	$p_r(r_k) = n_k/MN$
$r_0 = 0$	790	0.19
$r_1 = 1$	1023	0.25
$r_2 = 2$	850	0.21
$r_3 = 3$	656	0.16
$r_4 = 4$	329	0.08
$r_5 = 5$	245	0.06
$r_6 = 6$	122	0.03
$r_7 = 7$	81	0.02

**TABLE 3.1**  
Intensity  
distribution and  
histogram values  
for a 3-bit,  
 $64 \times 64$  digital  
image.

$$s_k = T(r_k) = (L-1) \sum_{j=0}^k p_r(r_j) \quad k = 0, 1, 2, \dots, L-1$$

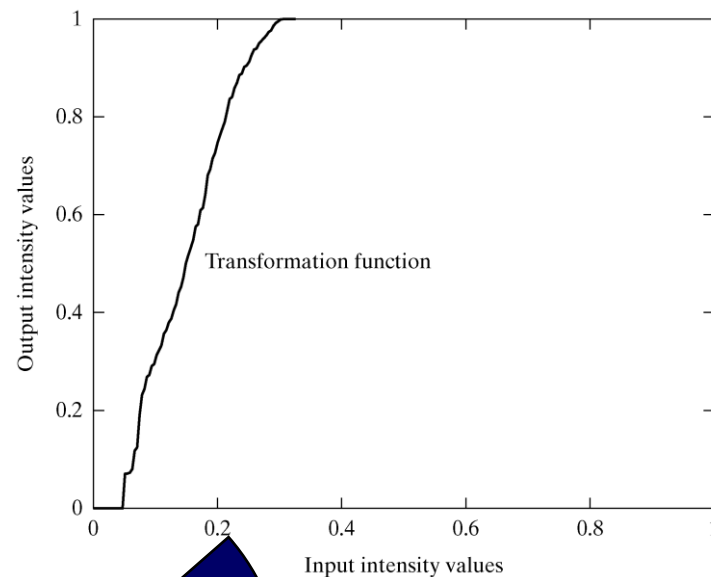
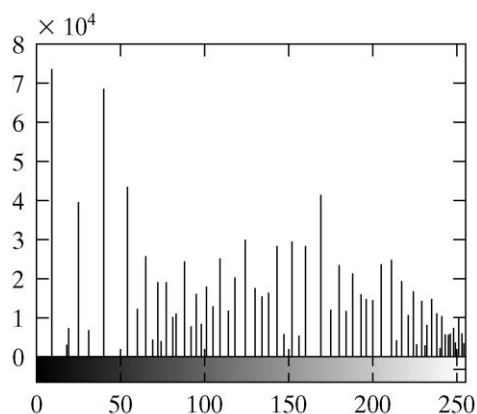
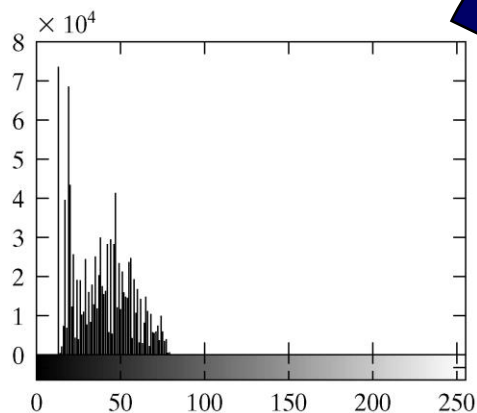
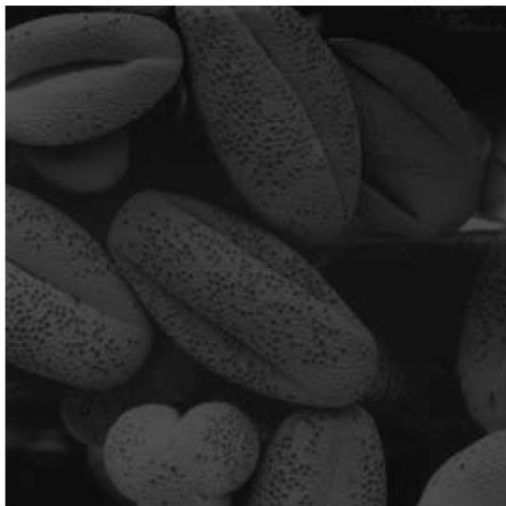
$$s_0 = T(r_0) = 7 \sum_{j=0}^0 p_r(r_j) = 7 p_r(r_0) = 1.33$$

$$\begin{array}{llll} s_0 = 1.33 \rightarrow 1 & s_2 = 4.55 \rightarrow 5 & s_4 = 6.23 \rightarrow 6 & s_6 = 6.86 \rightarrow 7 \\ s_1 = 3.08 \rightarrow 3 & s_3 = 5.67 \rightarrow 6 & s_5 = 6.65 \rightarrow 7 & s_7 = 7.00 \rightarrow 7 \end{array}$$

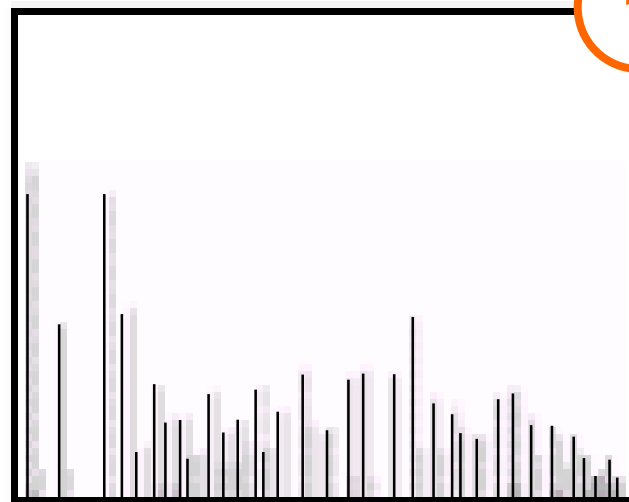
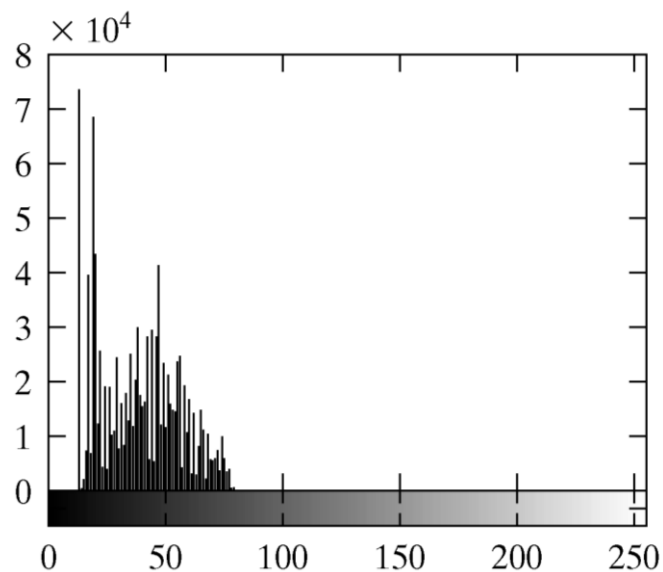


**FIGURE 3.19** Illustration of histogram equalization of a 3-bit (8 intensity levels) image. (a) Original histogram. (b) Transformation function. (c) Equalized histogram.

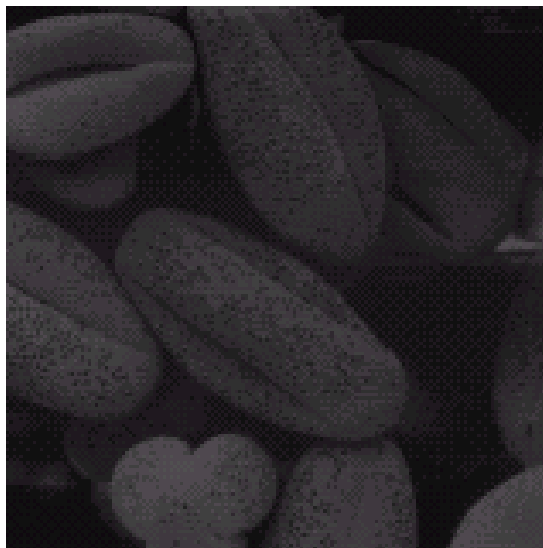
# Equalisation Transformation Function



# Equalisation Examples



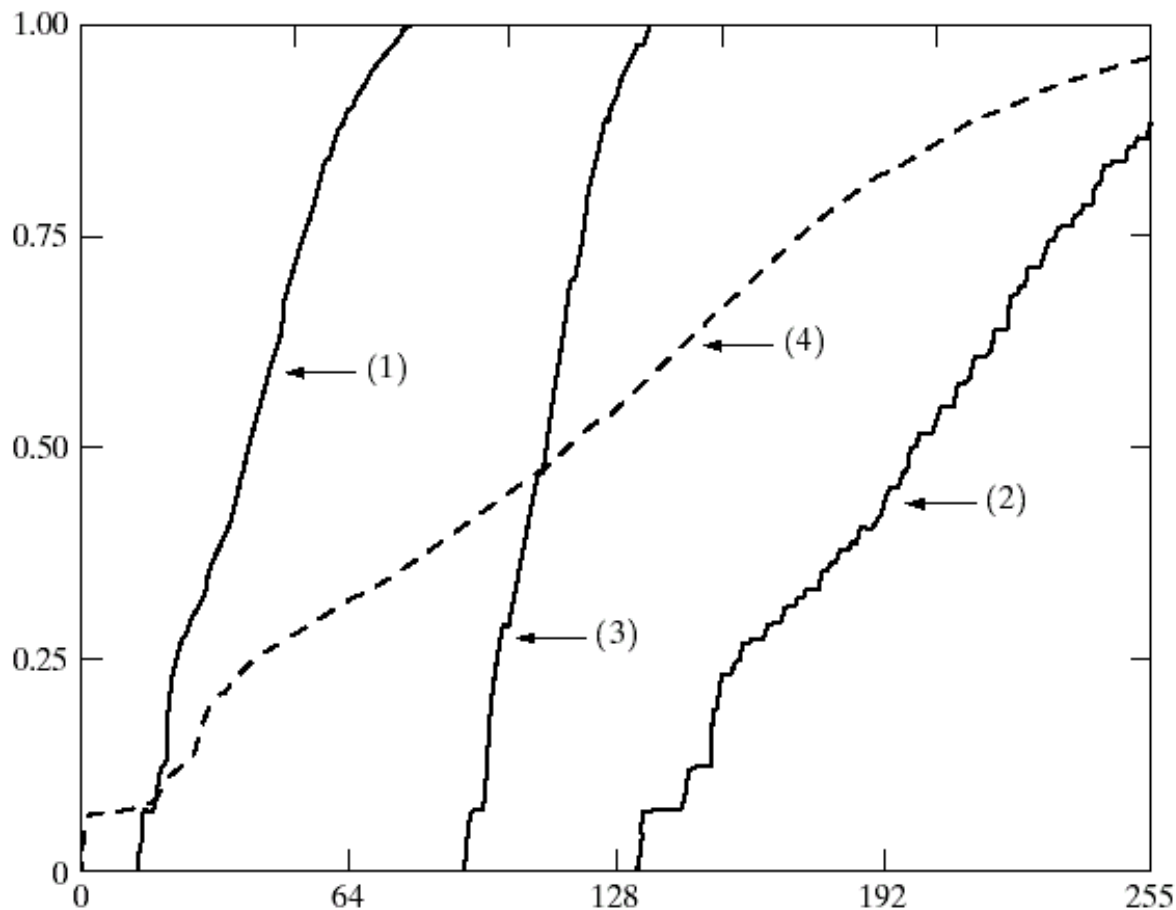
1



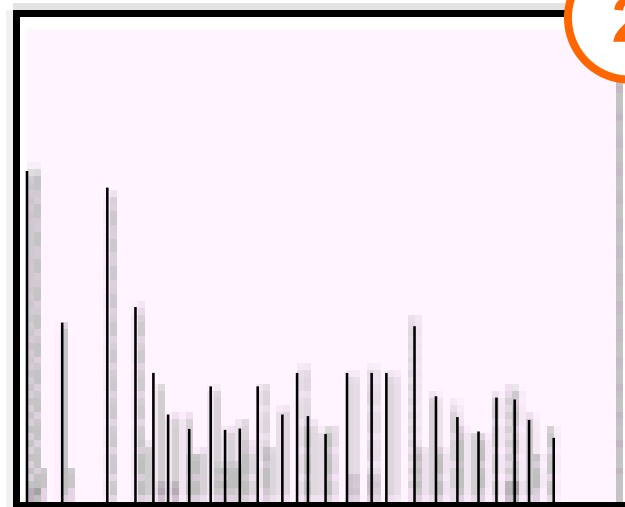
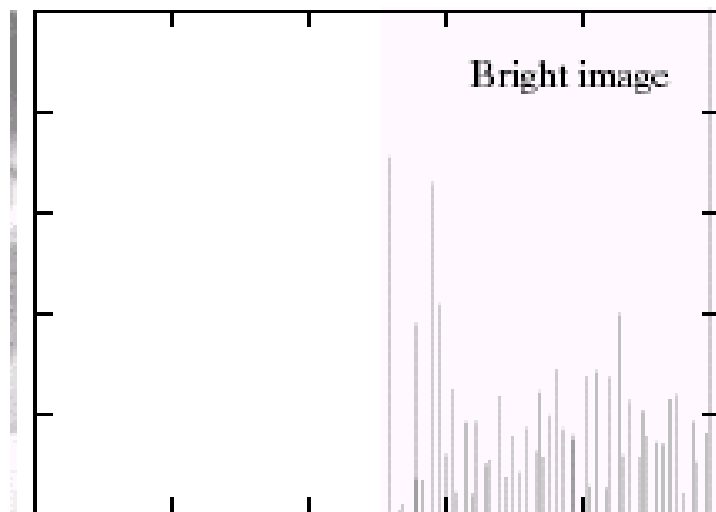


# Equalisation Transformation Functions

The functions used to equalise the images in the previous example

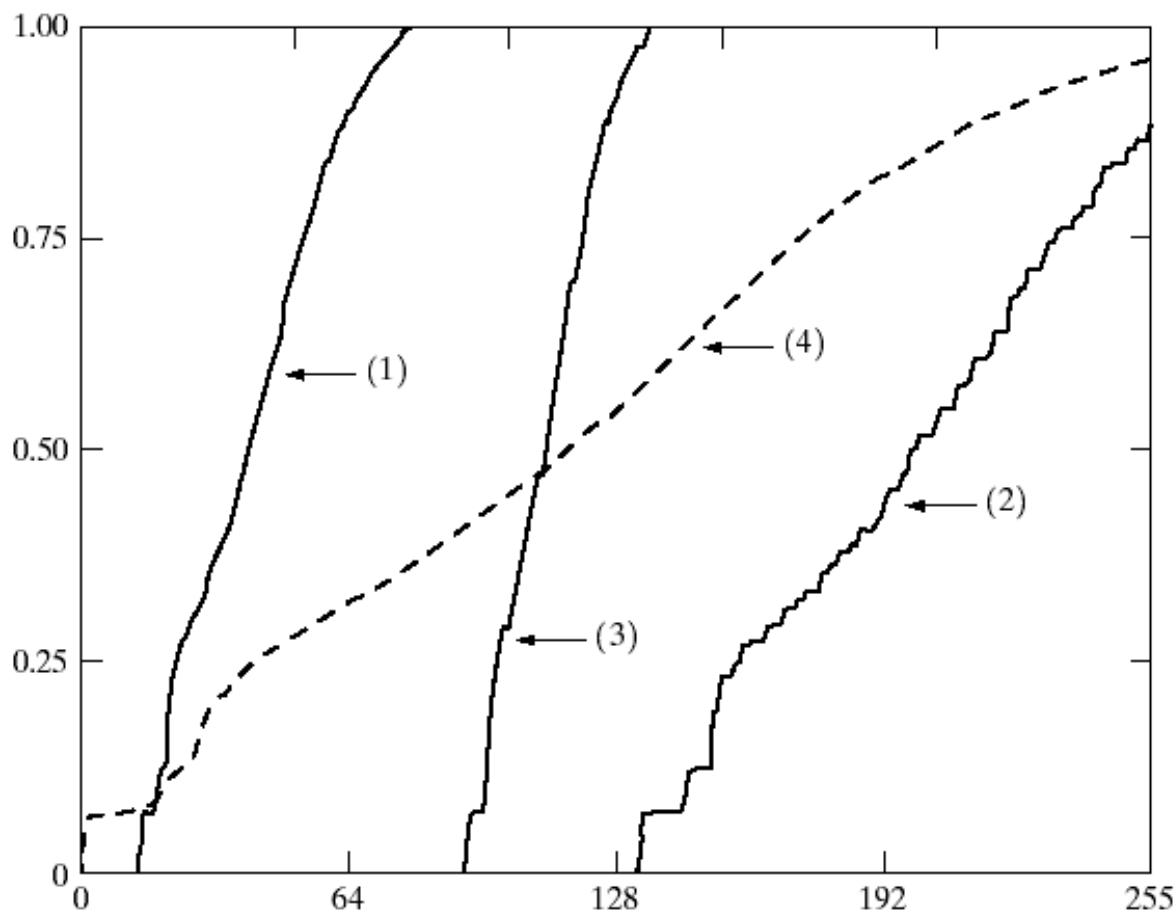


# Equalisation Examples

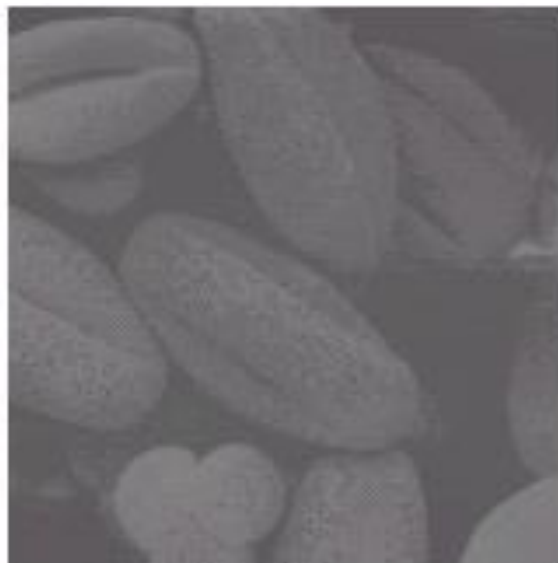


# Equalisation Transformation Functions

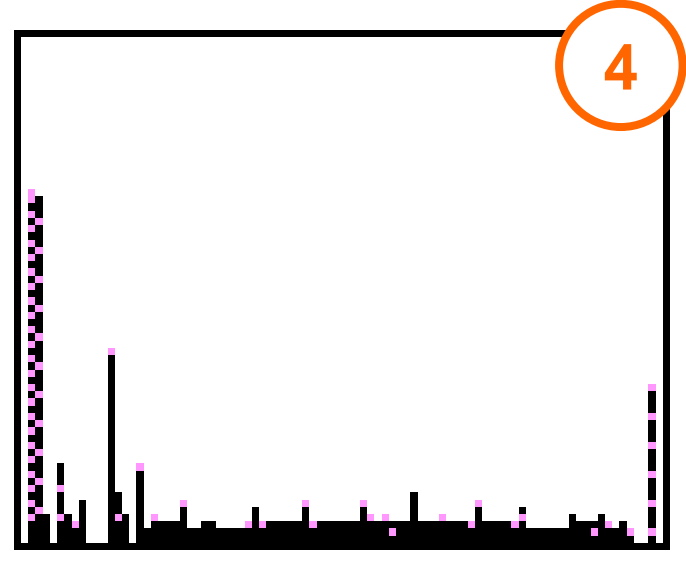
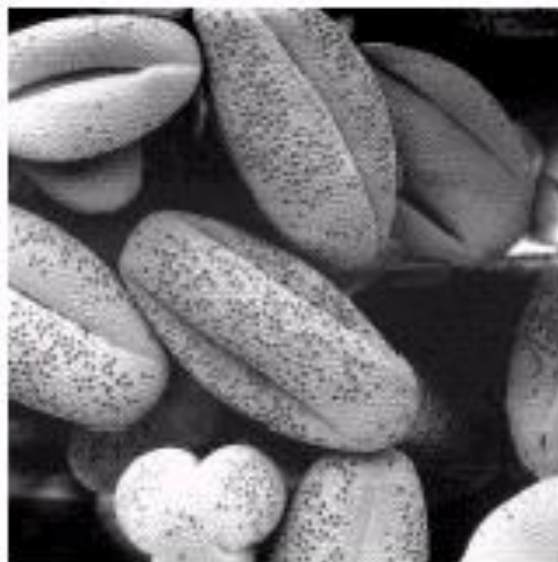
The functions used to equalise the images in the previous example



# Equalisation Examples (cont...)



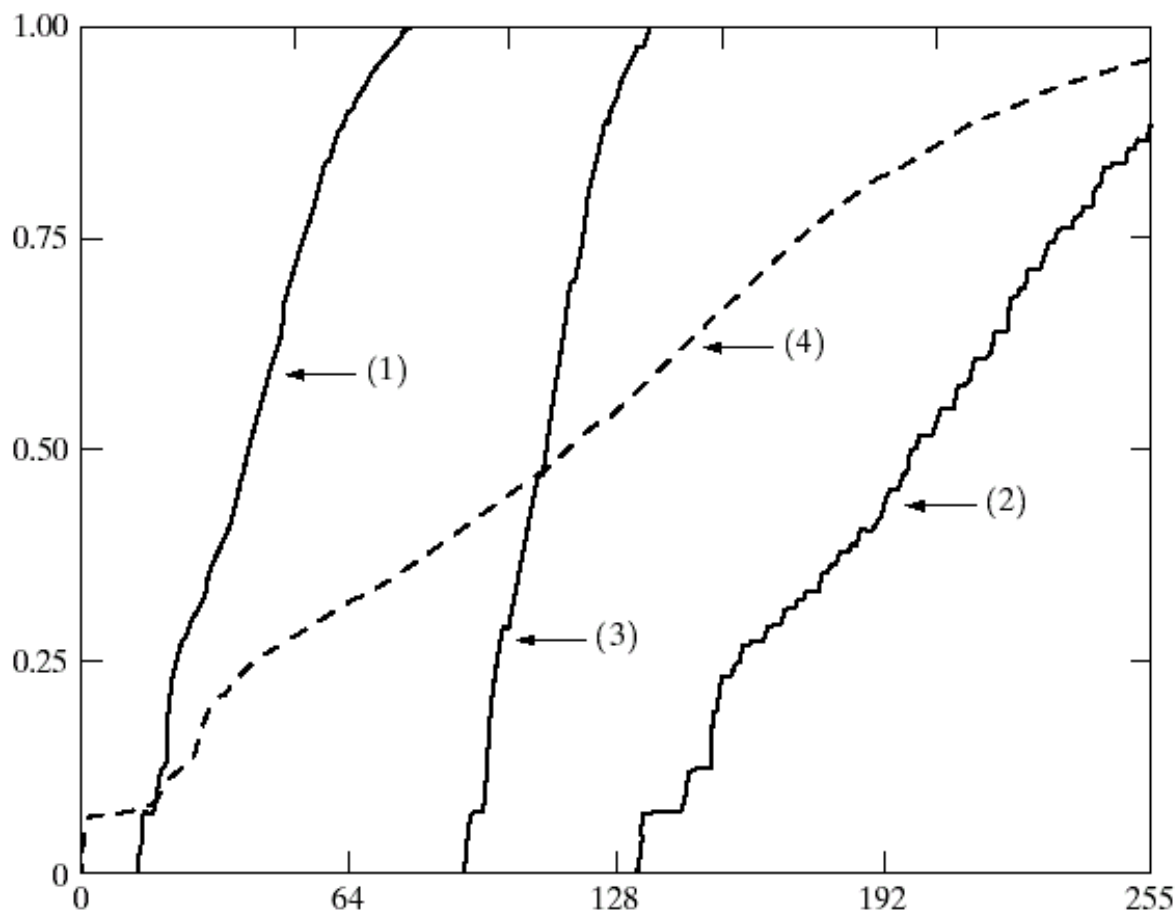
3



4

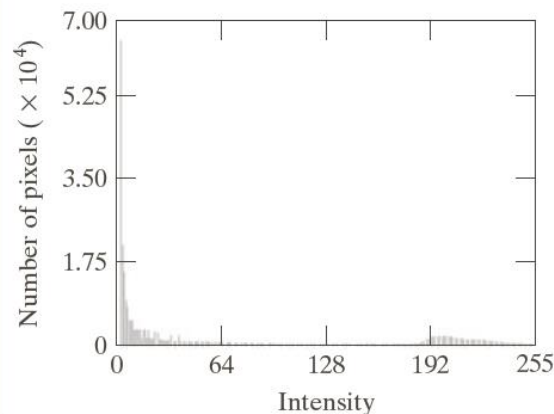
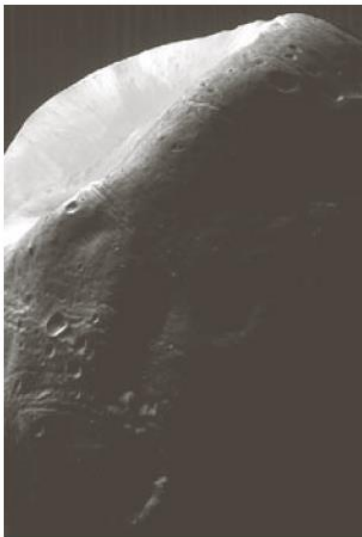
# Equalisation Transformation Functions

The functions used to equalise the images in the previous examples



# Histogram Matching

- There are applications in which histogram equalization is not suitable.
- It is useful sometimes to be able to specify the shape of the histogram that we wish the processed image to have.

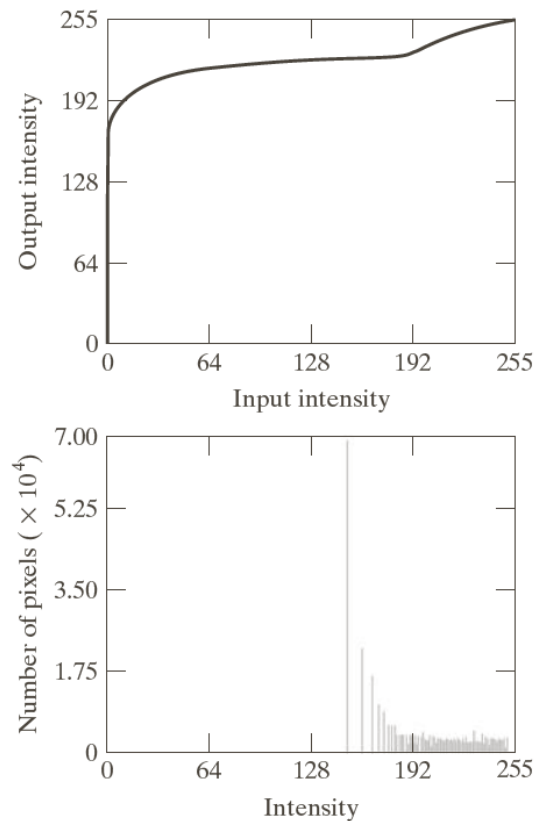


a b

**FIGURE 3.23**

(a) Image of the Mars moon Phobos taken by NASA's *Mars Global Surveyor*.  
(b) Histogram.  
(Original image courtesy of NASA.)

# Histogram Matching

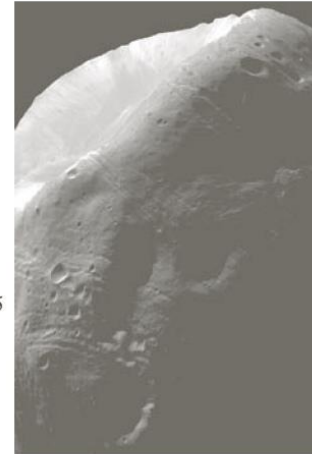
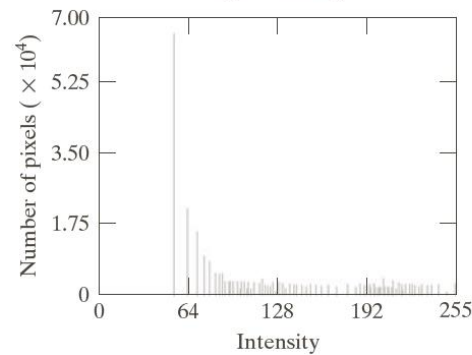
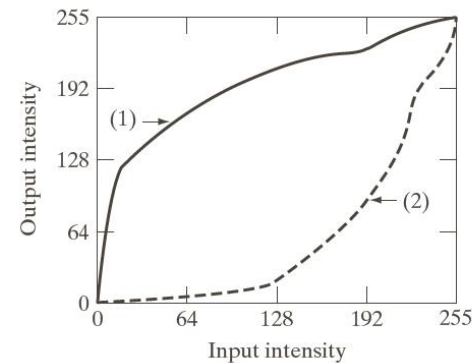
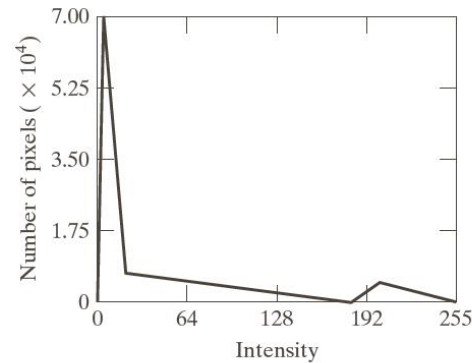


a b  
c

**FIGURE 3.24**  
(a) Transformation function for histogram equalization.  
(b) Histogram-equalized image (note the washed-out appearance).  
(c) Histogram of (b).



# Histogram Matching



a c  
b  
d

**FIGURE 3.25**

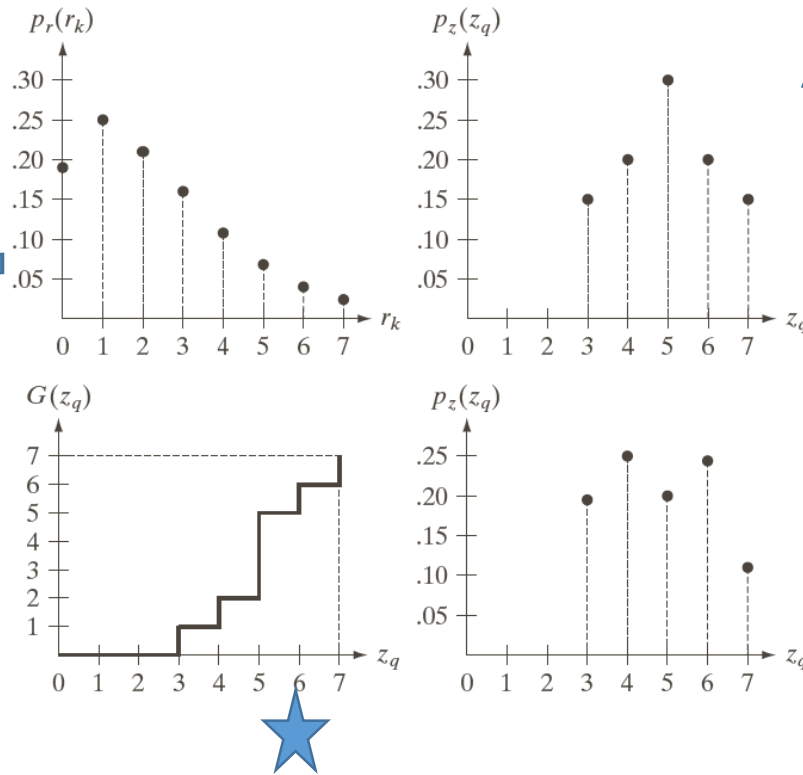
(a) Specified histogram.

(b) Transformations.

(c) Enhanced image using mappings from curve (2).

(d) Histogram of (c).

# Example (3.7 G&W)



**FIGURE 3.22**

(a) Histogram of a 3-bit image. (b) Specified histogram. (c) Transformation function obtained from the specified histogram. (d) Result of performing histogram specification. Compare (b) and (d).

$$\begin{aligned}
 G(z_0) &= 0.00 \rightarrow 0 & G(z_4) &= 2.45 \rightarrow 2 \\
 G(z_1) &= 0.00 \rightarrow 0 & G(z_5) &= 4.55 \rightarrow 5 \\
 G(z_2) &= 0.00 \rightarrow 0 & G(z_6) &= 5.95 \rightarrow 6 \\
 G(z_3) &= 1.05 \rightarrow 1 & G(z_7) &= 7.00 \rightarrow 7
 \end{aligned}$$

$$\begin{aligned}
 s_0 &= 1.33 \rightarrow 1 & s_2 &= 4.55 \rightarrow 5 & s_4 &= 6.23 \rightarrow 6 & s_6 &= 6.86 \rightarrow 7 \\
 s_1 &= 3.08 \rightarrow 3 & s_3 &= 5.67 \rightarrow 6 & s_5 &= 6.65 \rightarrow 7 & s_7 &= 7.00 \rightarrow 7
 \end{aligned}$$

$z_q$	$G(z_q)$
$z_0 = 0$	0
$z_1 = 1$	0
$z_2 = 2$	0
$z_3 = 3$	1
$z_4 = 4$	2
$z_5 = 5$	5
$z_6 = 6$	6
$z_7 = 7$	7

# Example (3.7 G&W)

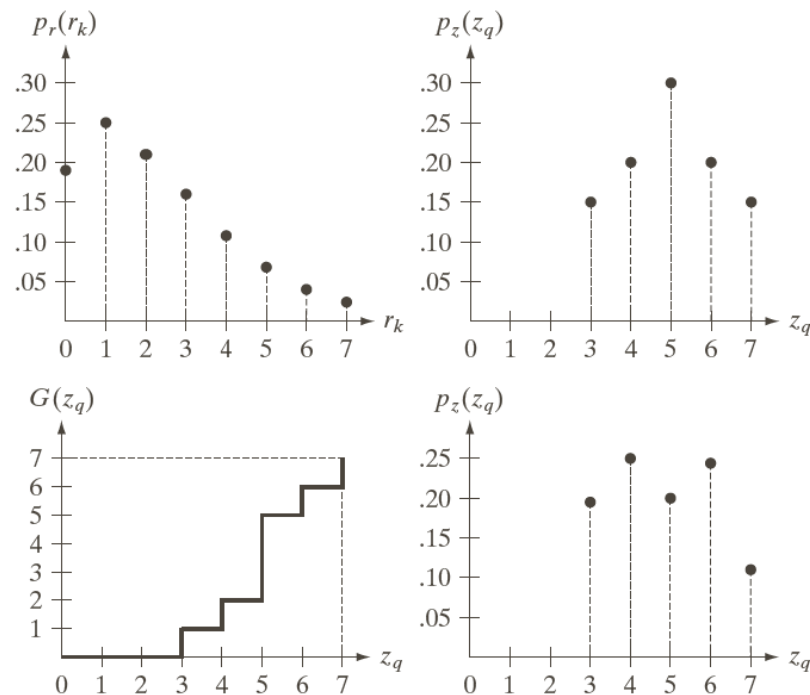
$r_k$	$n_k$	$p_r(r_k) = n_k/MN$
$r_0 = 0$	790	0.19
$r_1 = 1$	1023	0.25
$r_2 = 2$	850	0.21
$r_3 = 3$	656	0.16
$r_4 = 4$	329	0.08
$r_5 = 5$	245	0.06
$r_6 = 6$	122	0.03
$r_7 = 7$	81	0.02

$z_q$	$G(z_q)$
$z_0 = 0$	0
$z_1 = 1$	0
$z_2 = 2$	0
$z_3 = 3$	1
$z_4 = 4$	2
$z_5 = 5$	5
$z_6 = 6$	6
$z_7 = 7$	7

$$\begin{array}{llll}
 s_0 = 1.33 \rightarrow 1 & s_2 = 4.55 \rightarrow 5 & s_4 = 6.23 \rightarrow 6 & s_6 = 6.86 \rightarrow 7 \\
 s_1 = 3.08 \rightarrow 3 & s_3 = 5.67 \rightarrow 6 & s_5 = 6.65 \rightarrow 7 & s_7 = 7.00 \rightarrow 7
 \end{array}$$

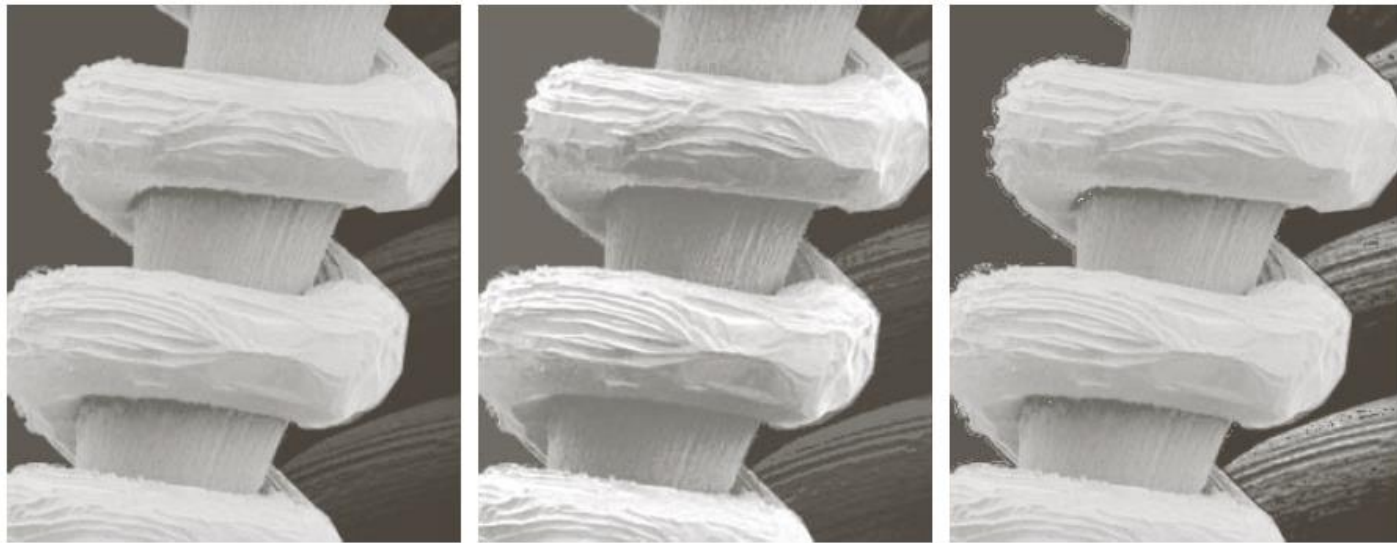
Mapping

# Example (3.7 G&W)



$z_q$	Specified $p_z(z_q)$	Actual $p_z(z_k)$
$z_0 = 0$	0.00	0.00
$z_1 = 1$	0.00	0.00
$z_2 = 2$	0.00	0.00
$z_3 = 3$	0.15	0.19
$z_4 = 4$	0.20	0.25
$z_5 = 5$	0.30	0.21
$z_6 = 6$	0.20	0.24
$z_7 = 7$	0.15	0.11

# Partial Transformations



a b c

**FIGURE 3.27** (a) SEM image of a tungsten filament magnified approximately 130 $\times$ . (b) Result of global histogram equalization. (c) Image enhanced using local histogram statistics. (Original image courtesy of Mr. Michael Shaffer, Department of Geological Sciences, University of Oregon, Eugene.)

# Summary

- We have looked at:
  - Different kinds of image enhancement
  - Histograms
  - Histogram equalisation, matching.
- We have looked at different kinds of point processing image enhancement
- Next time we will start to look at neighbourhood operations – in particular *filtering* and *convolution*