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# Lecture 2 Intensity Transformations

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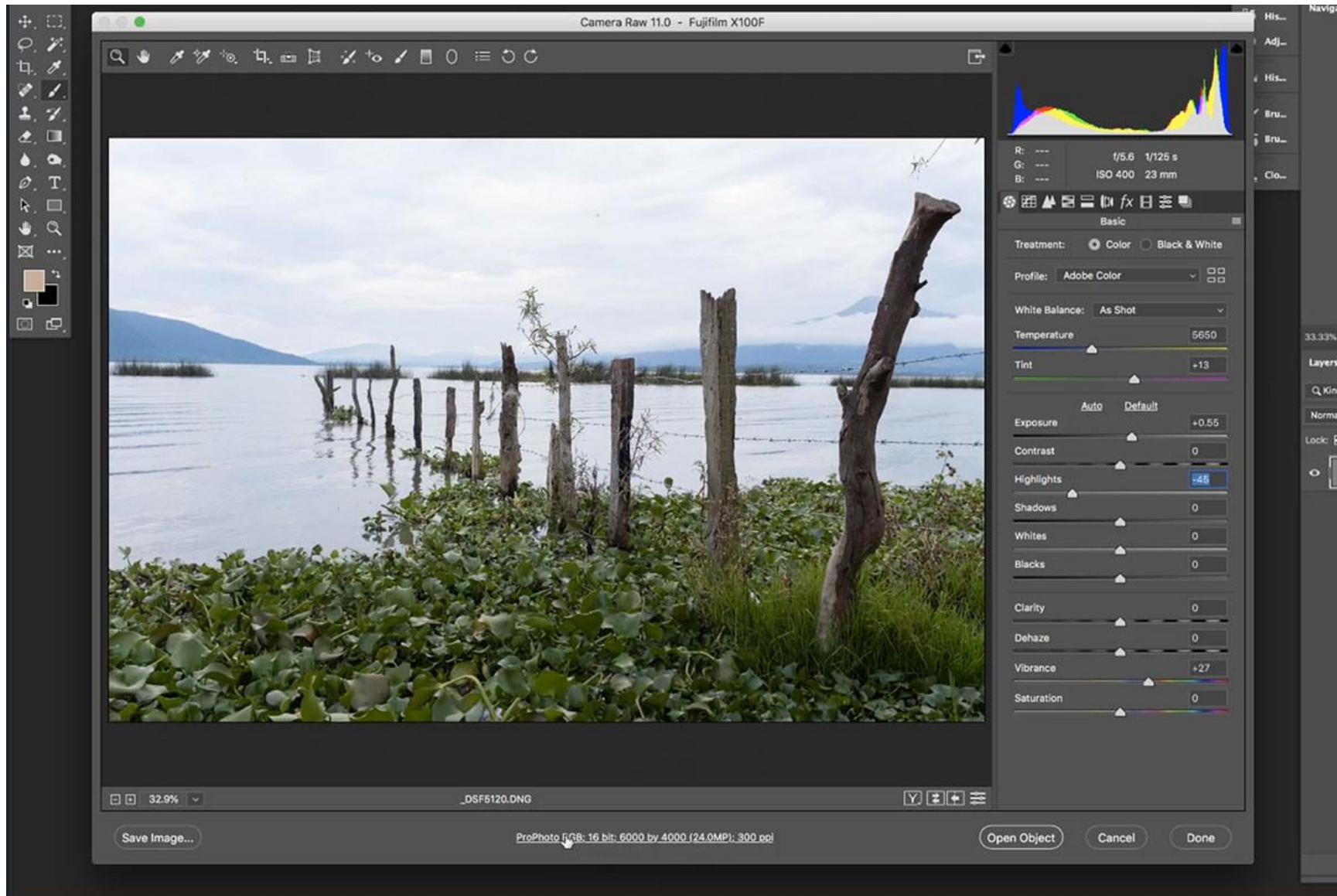
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October 2, 2020

# Photoshop



# Outline

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## □ Background

## □ Intensity Transformation

- Image Negatives
- Log Transformation
- Power-Law Transformations
- Piecewise-Linear Transformation

## □ Histogram Processing

- Histogram Equalization
- Histogram Specification

# Background

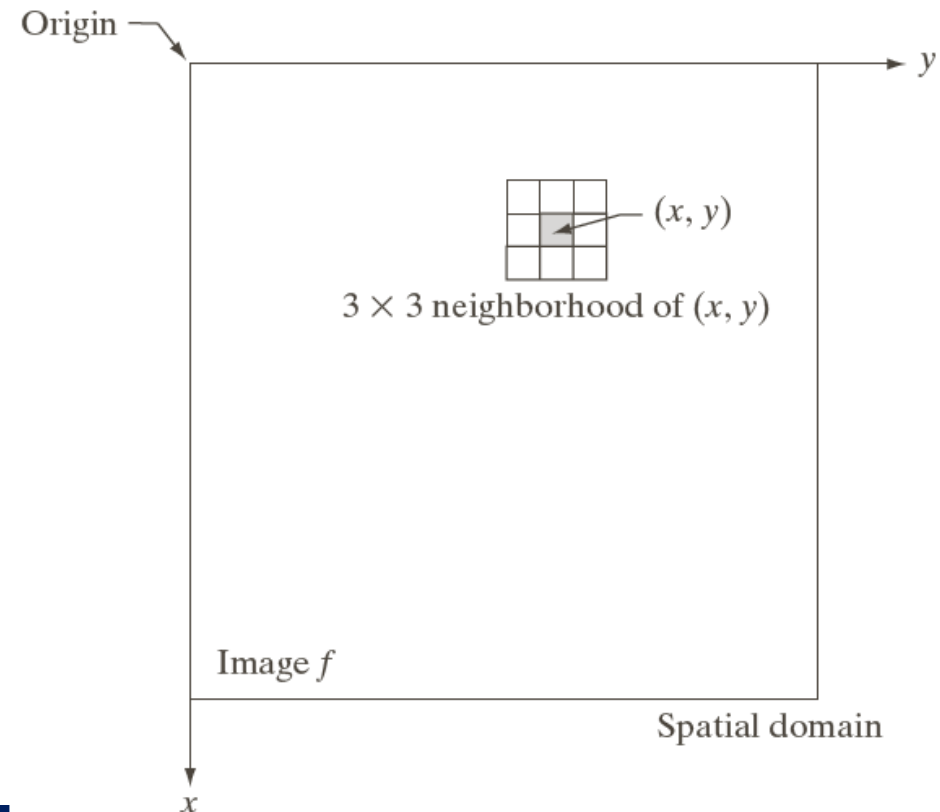
## □ Spatial domain

- Operate directly on pixels
- Contrast to transform domain, e.g. Fourier transform

## □ General expression

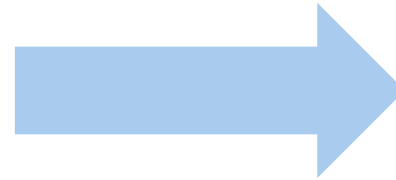
$$g(x, y) = T[f(x, y)]$$

- $f(x, y)$ : input image
- $g(x, y)$ : output image
- $T[\cdot]$ : operator on  $f$ , over a neighborhood of point  $(x, y)$



# Example

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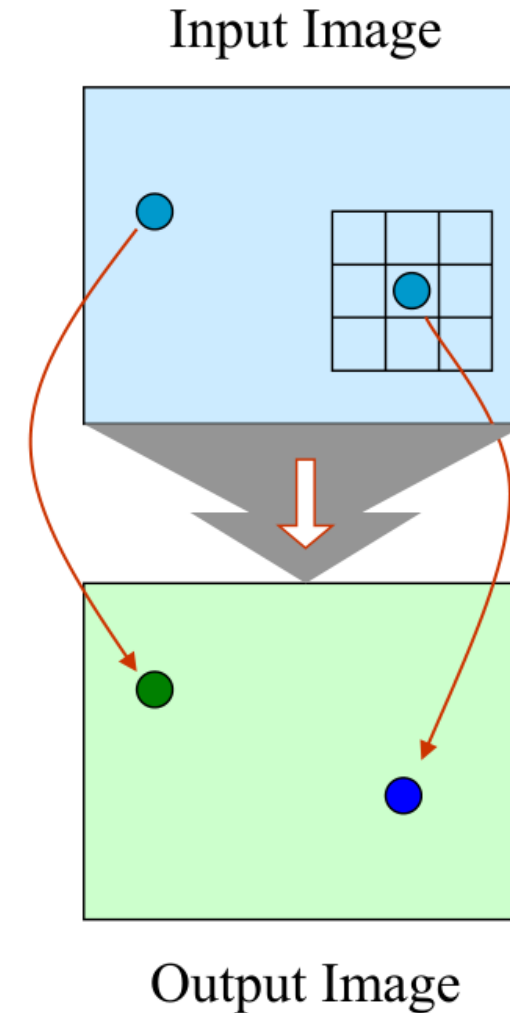


**Fig.** MRI image

# Operation Types

- ❑ **Point** Operation
  - Gray-level transformation
- ❑ **Local** Operation
  - Mask Processing or filtering
- ❑ **Global** Operation
  - Use values of all pixels
  - (e.g.) Fourier transform

Histogram equalization, etc



# Example 1

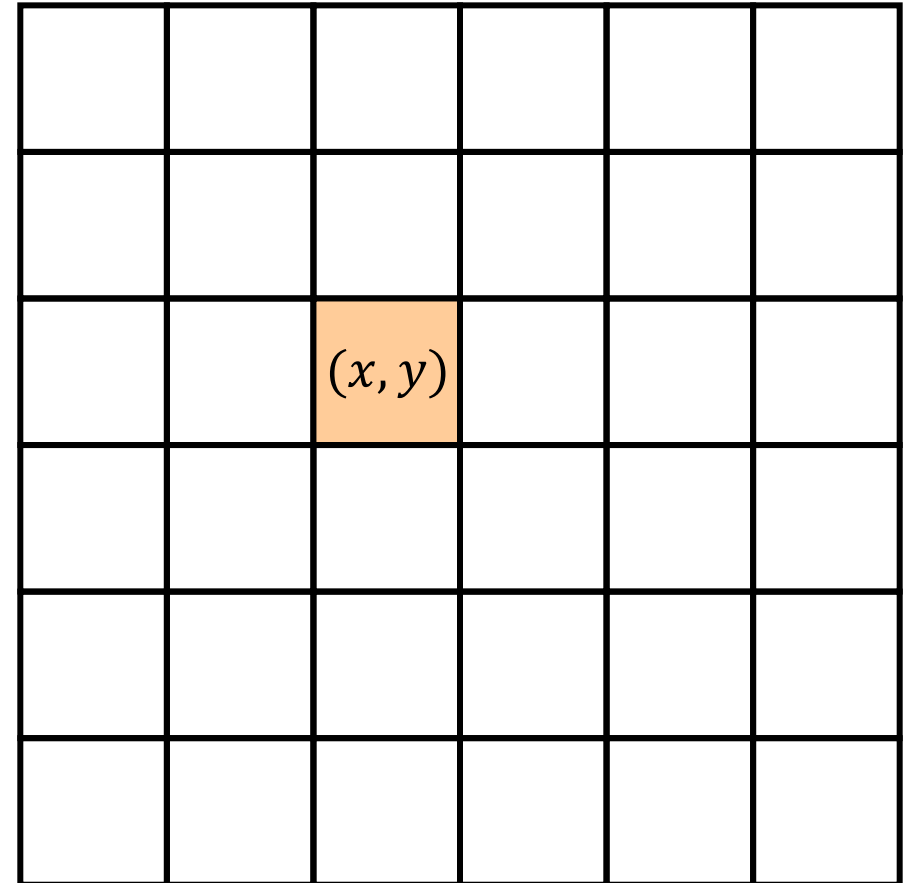
- $g(x, y) = T[f(x, y)]$ 
  - e.g. neighborhood is a  $3 \times 3$  square
  - $T$ : compute the average intensity of the neighborhood
  - then  $g(x, y) = 1/9$
  - This is called **spatial filtering**, and the  $3 \times 3$  neighborhood, along with the operation is called a **filter**

	1	1	1		
	1	1 (x, y)	1		
	1	1	1		

# Example 2

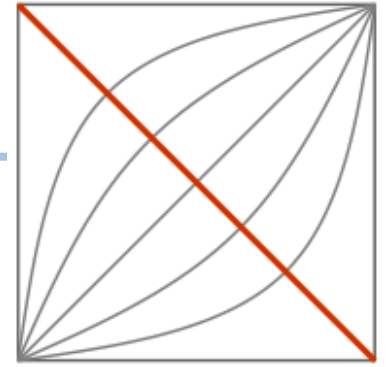
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- $g(x, y) = T[f(x, y)]$ 
  - e.g. neighborhood is a  $1 \times 1$
  - Called **intensity transformation**
  - $s = T[r]$
  - $r$ : intensity of input pixel
  - $s$ : intensity of output pixel

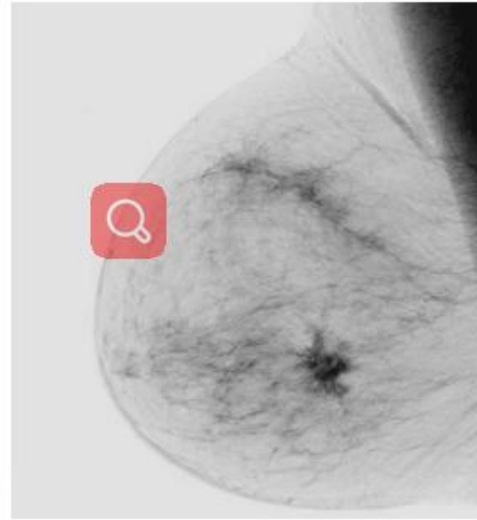
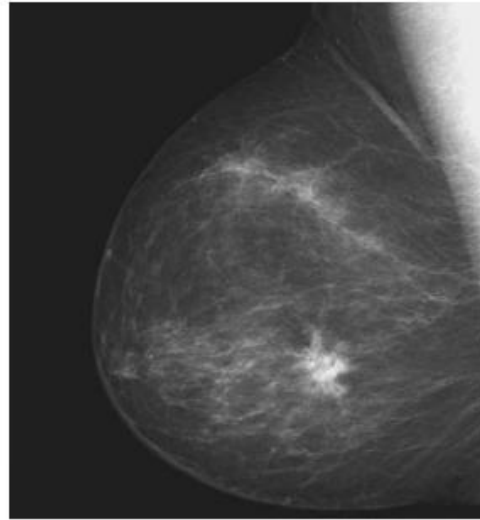




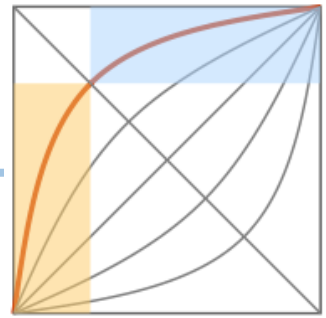
# Intensity Transformation



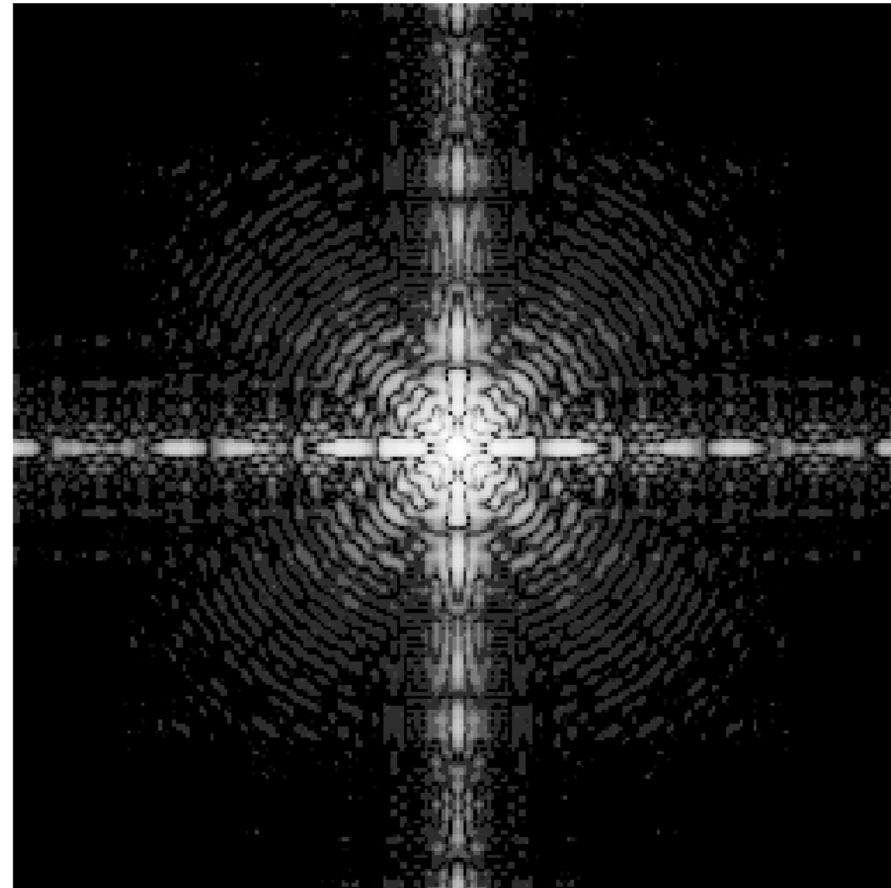
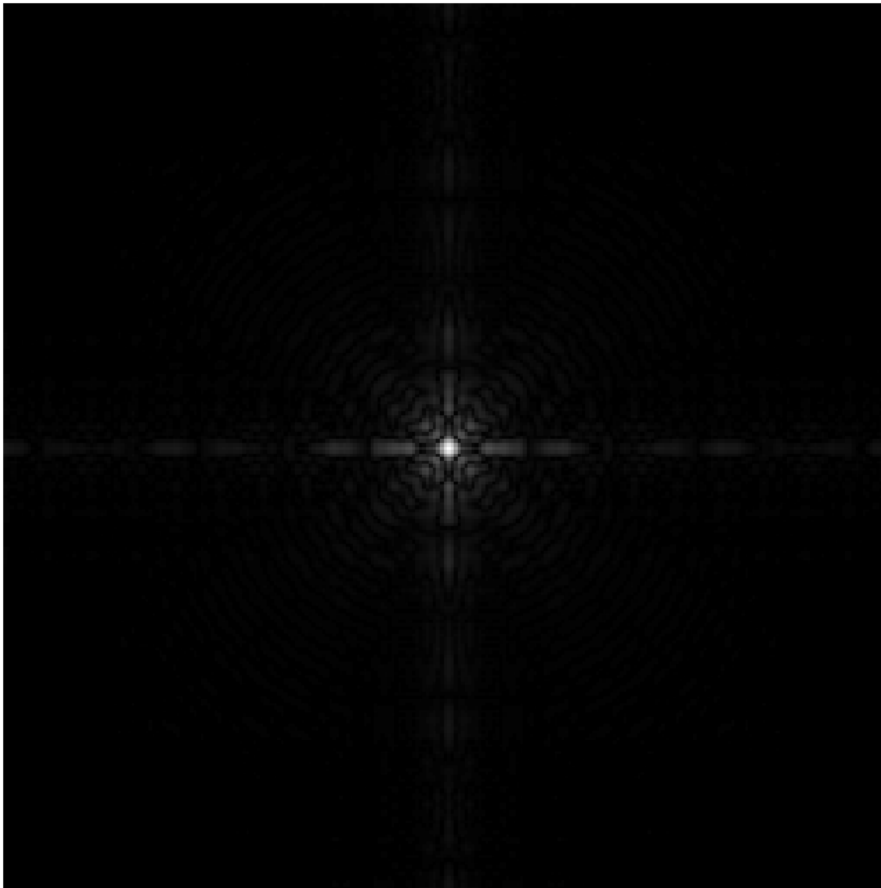
- Image Negatives:  $s = L - 1 - r$



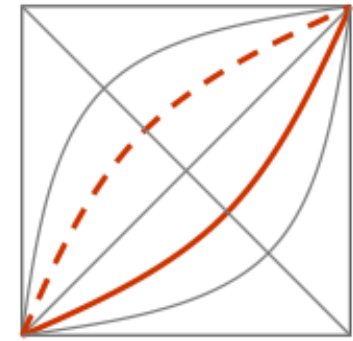
# Intensity Transformation



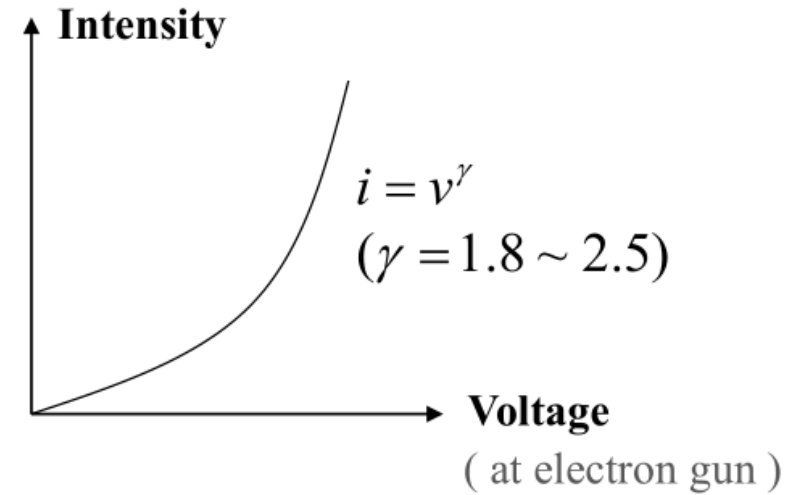
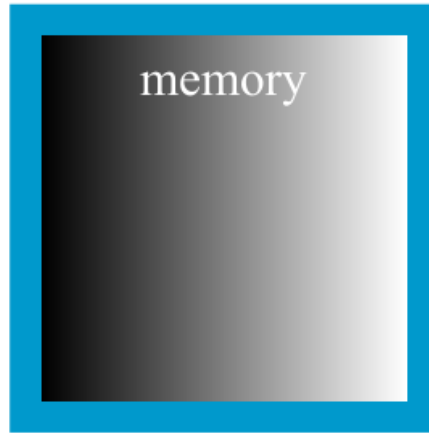
- Log Transformation:  $s = c \log(1 + r)$



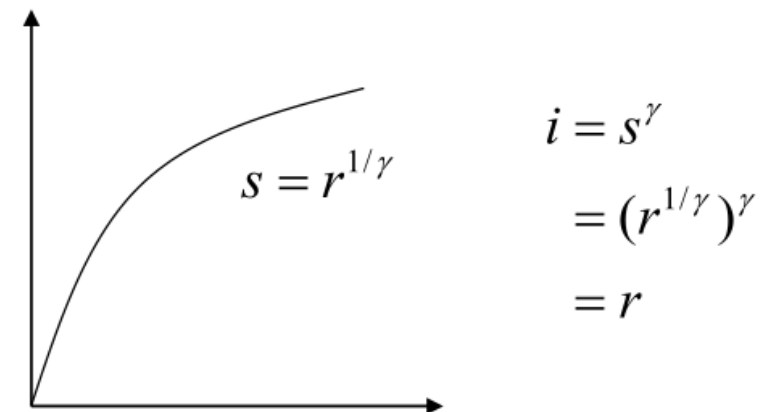
# Intensity Transformation



□ Power-Law Transformations:  $s = cr^\gamma$



$s = T(r)$  ↓



# Intensity Trans

□ Power-Law (G

Original MRI image



$$s = cr^\gamma$$

$$c = 1, \gamma = 0.6$$

$$c = 1, \gamma = 0.4$$



$$c = 1, \gamma = 0.3$$

# Intensity Transformations

## Power-Law

Original Aerial image



$$s = cr^\gamma$$
$$c = 1, \gamma = 3.0$$



$$c = 1, \gamma = 4.0$$

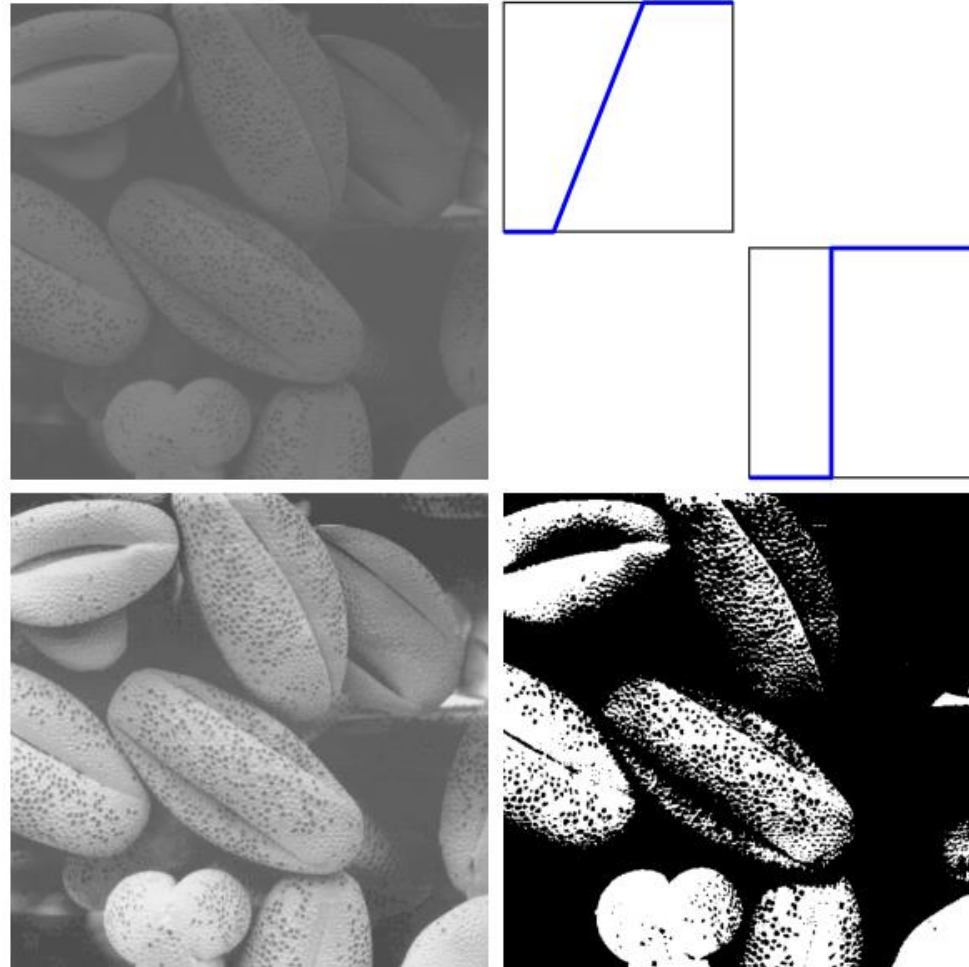
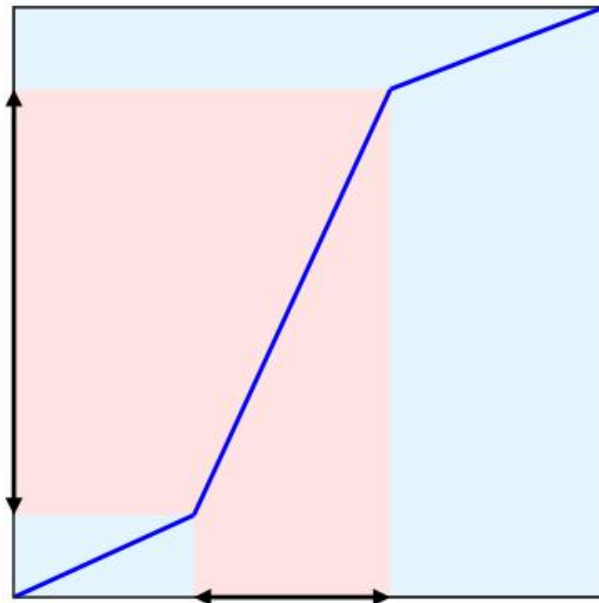


$$c = 1, \gamma = 5.0$$

# Intensity Transformation

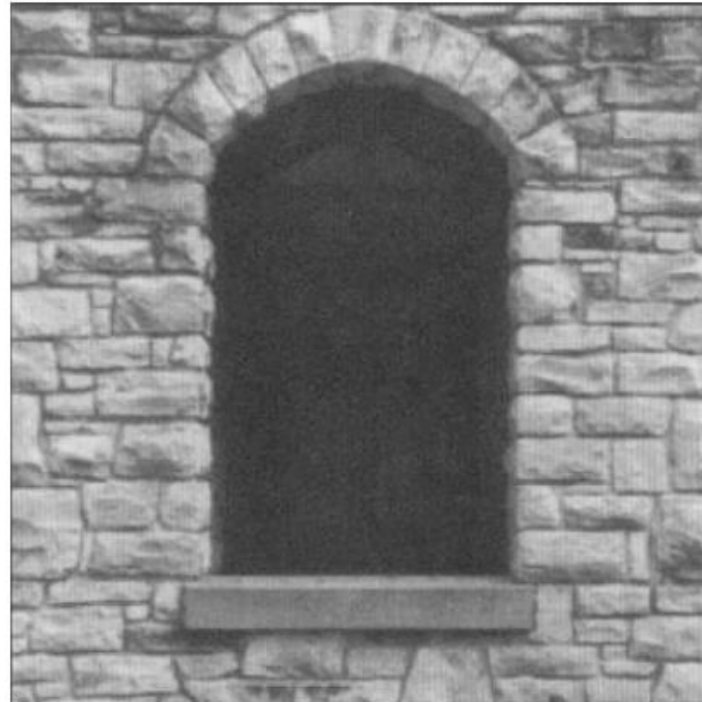
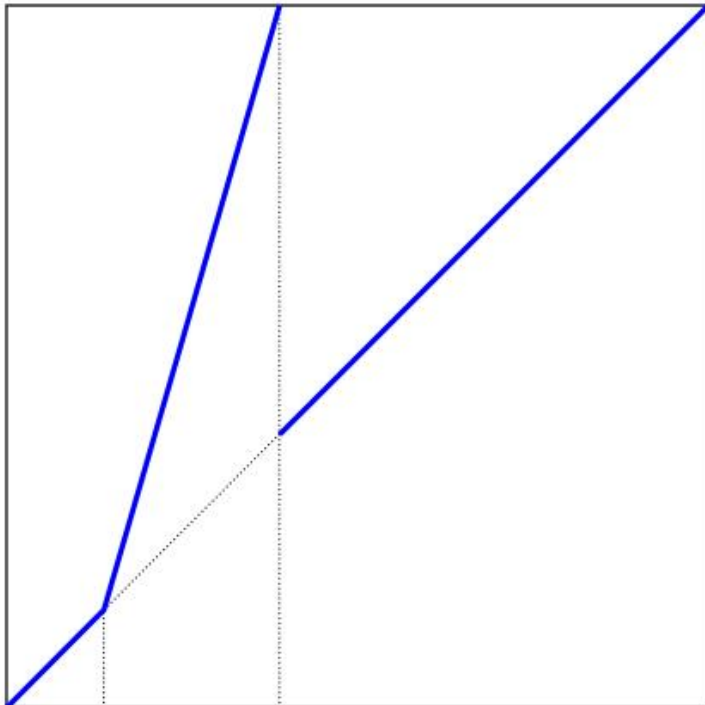
## □ Piecewise-Linear Transformation Functions

- Contrast Stretching



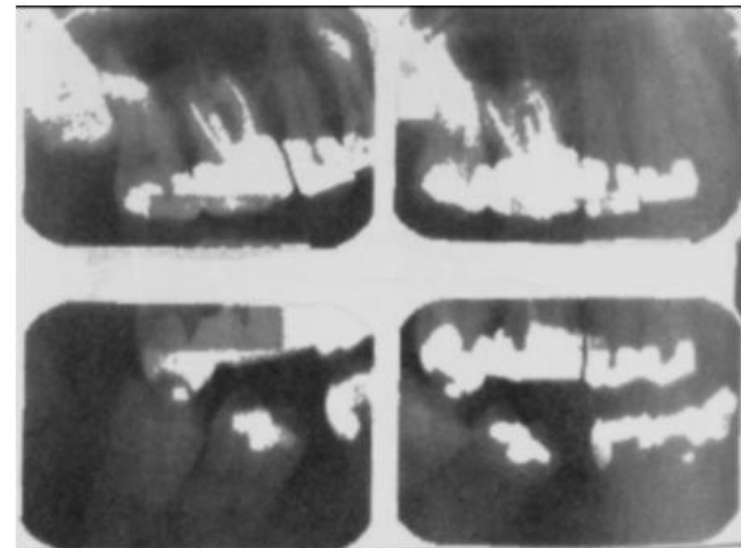
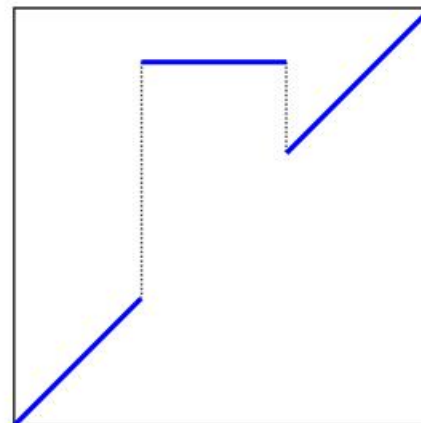
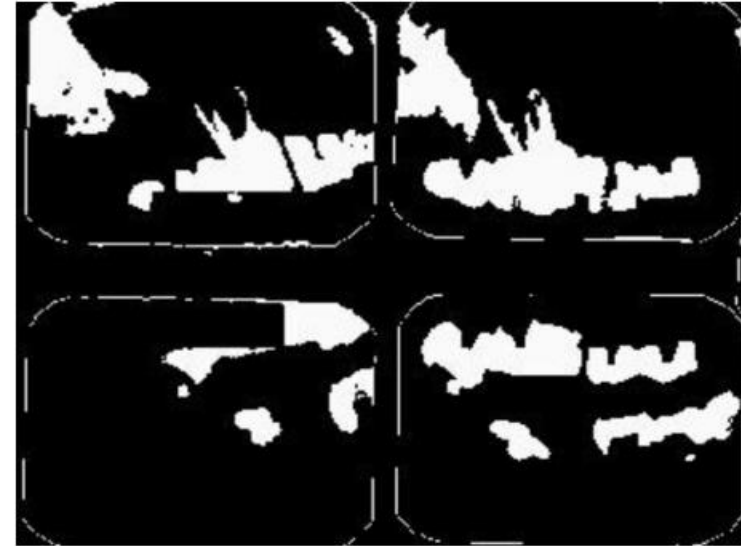
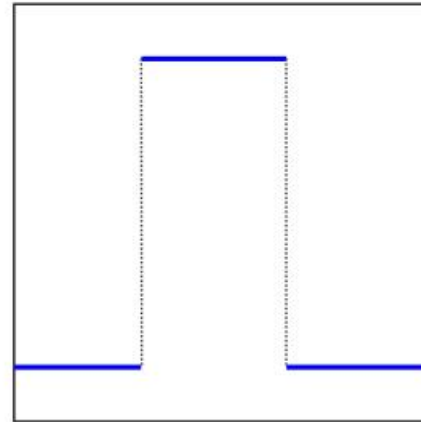
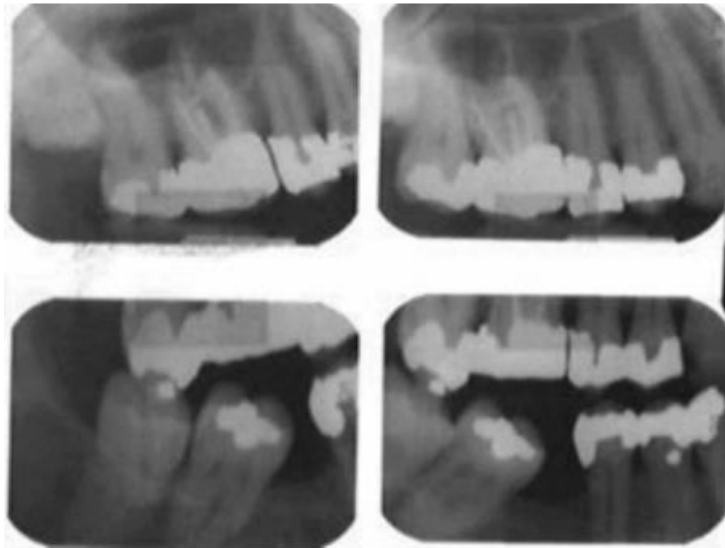
# Intensity Transformation

- Gray-Level Stretching



# Intensity Transformation

- Gray-Level Slicing



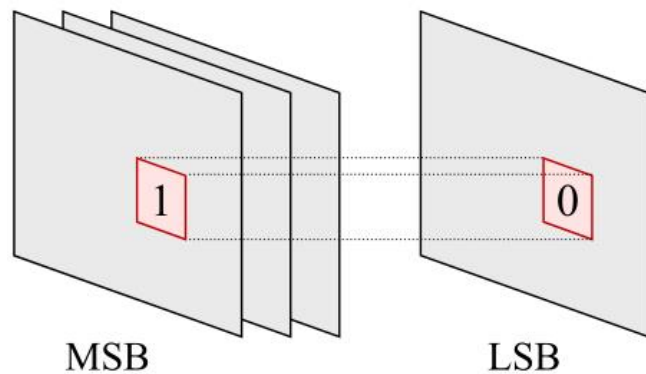


# Intensity Transformation

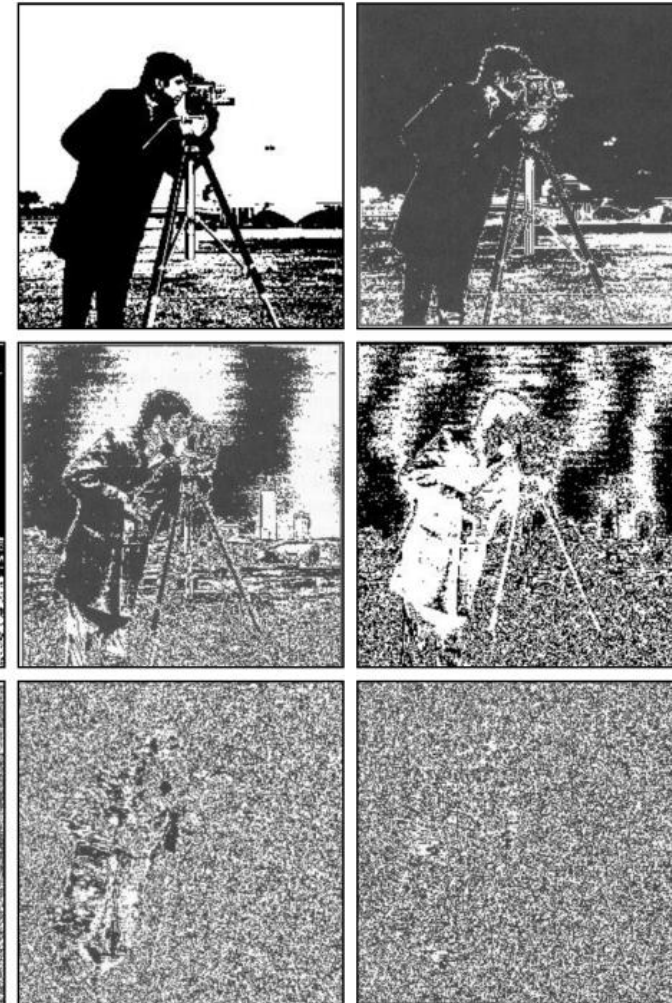
- Bit-Plane Slicing



1 0 1 1 0 1 0 0



bit	7	6
5	4	3
2	1	0



# Histogram Processing

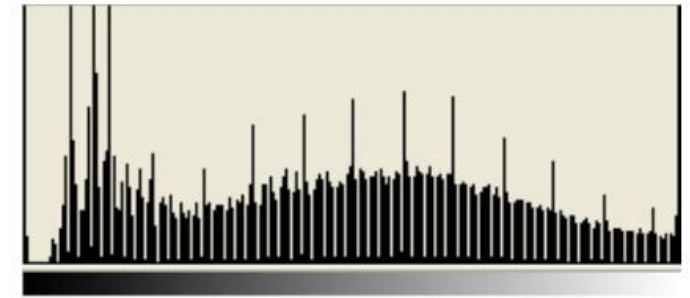
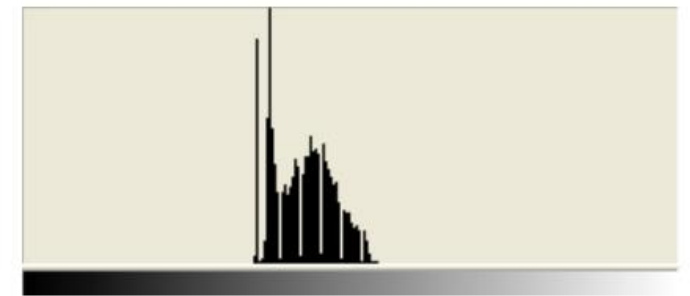
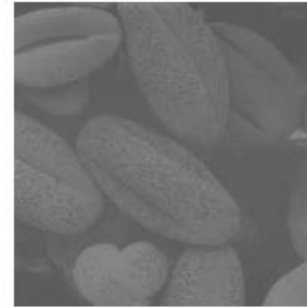
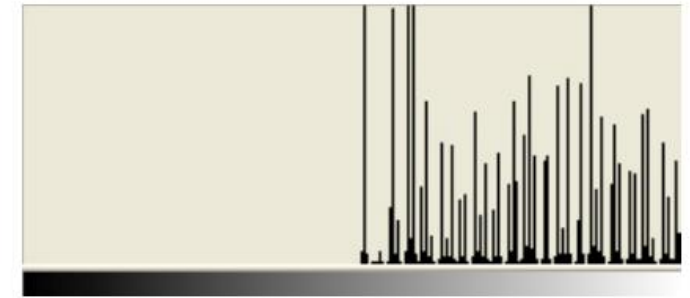
## □ Histogram

$$h(r_k) = n_k$$

$$k = 0, 1, \dots, L - 1$$

$$p(r_k) = \frac{n_k}{\sum_k n_k}$$

$$\sum_k p(r_k) = 1$$



## □ Background (option)

- **Cumulative Distribution Function (CDF),  $F(x)$**  for a continuous random variable  $X$

$$F(X) = P(X \leq x)$$

1.  $F(-\infty) = 0$
2.  $F(\infty) = 1$
3.  $0 \leq F(x) \leq 1$
4.  $F(x_1) \leq F(x_2)$  if  $x_1 \leq x_2$
5.  $P(x_1 < x \leq x_2) = F(x_2) - F(x_1)$

- **Probability Distribution Function (PDF),  $p(x)$**  for a continuous random variable  $x$

$$p(x) = \frac{dF(x)}{dx}$$

1.  $p(x) \geq 0$  for all  $x$
2.  $\int_{-\infty}^{\infty} p(x)dx = 1$
3.  $F(x) = \int_{-\infty}^x p(\alpha)d\alpha$ , where  $\alpha$  is a dummy variable
4.  $P(x_1 < x \leq x_2) = \int_{x_1}^{x_2} p(x)dx$

- Transform of PDF using a monotonic function  $T(x)$

$$y = T(x), 0 \leq x \leq 1$$

We assume that

(a)  $T(x)$  is a single-valued and monotonically increasing in the interval

$$0 \leq x \leq 1; \text{ and}$$

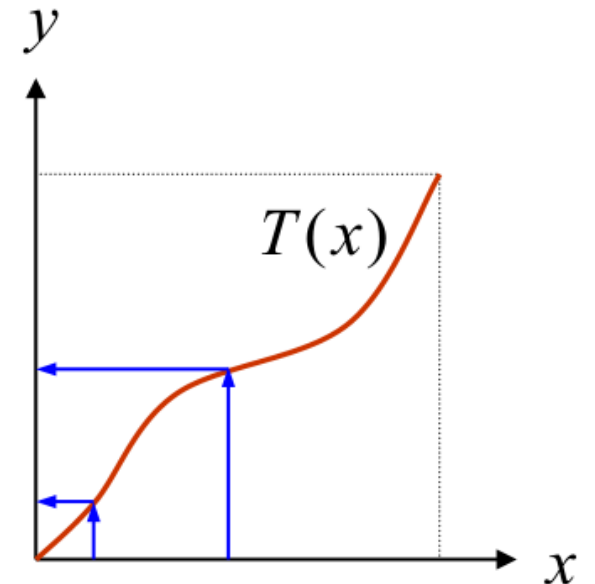
(b)  $0 \leq T(x) \leq 1$  for  $0 \leq x \leq 1$

$$f(x) \leq f(y) \text{ for } x \leq y$$

*non-decreasing*

$$f(x) < f(y) \text{ for } x < y$$

*strictly increasing  $\Rightarrow$  one-to-one*



□ The inverse transformation is denoted

$$x = T^{-1}(y), 0 \leq y \leq 1$$

If  $T^{-1}(y)$  is single-valued and non-decreasing, then

If  $y = T(x) = \int_0^x p_o(\alpha) d\alpha,$

$$\frac{dy}{dx} = \frac{d}{dx} \left[ \int_0^x p_o(\alpha) d\alpha \right] = p_o(x). \quad \text{Leibniz's rule}$$

$$p_t(y) = p_o(x) \left| \frac{dx}{dy} \right|$$

$$\therefore p_t(y) = p_o(x) \left| \frac{dx}{dy} \right| = p_o(x) \left| \frac{1}{p_o(x)} \right| = 1$$

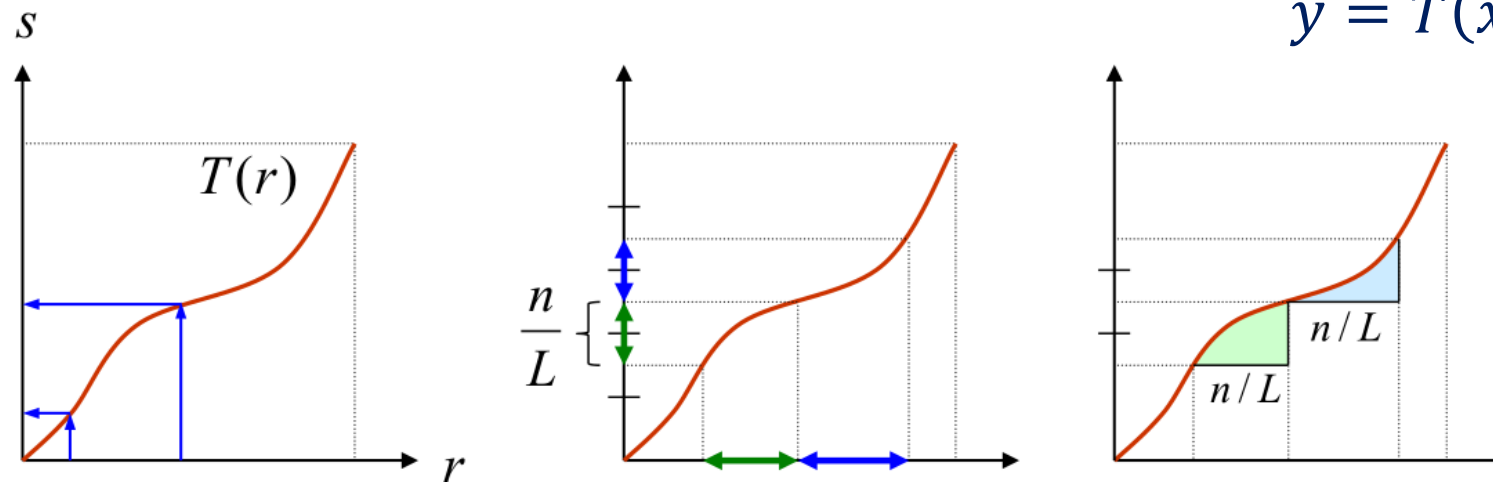
∴ Uniform PDF

## □ Histogram Equalization

- Discrete version of transform of PDF

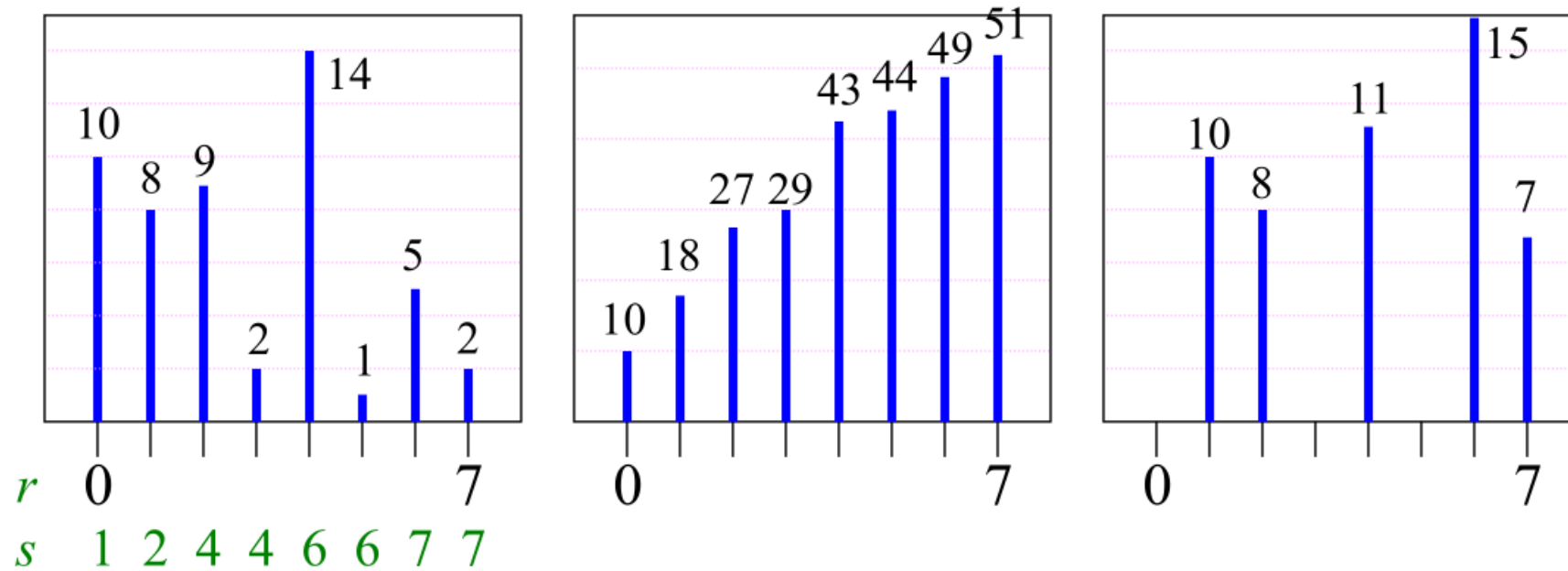
$$s_k = T(r_k) = \sum_{j=0}^k p_r(r_j) = \sum_{j=0}^k \frac{n_j}{n} \quad n = \sum_{i=0}^{L-1} n_i$$

- Implementation



$$y = T(x) = \int_0^x p_x(\alpha) d\alpha$$

# Histogram Processing



$$\frac{(10, 18, 27, 29, 43, 44, 49, 51)}{51} \times 7$$

$$\approx (1.37, 2.47, 3.71, 3.98, 5.90, 6.04, 6.73, 7.00)$$

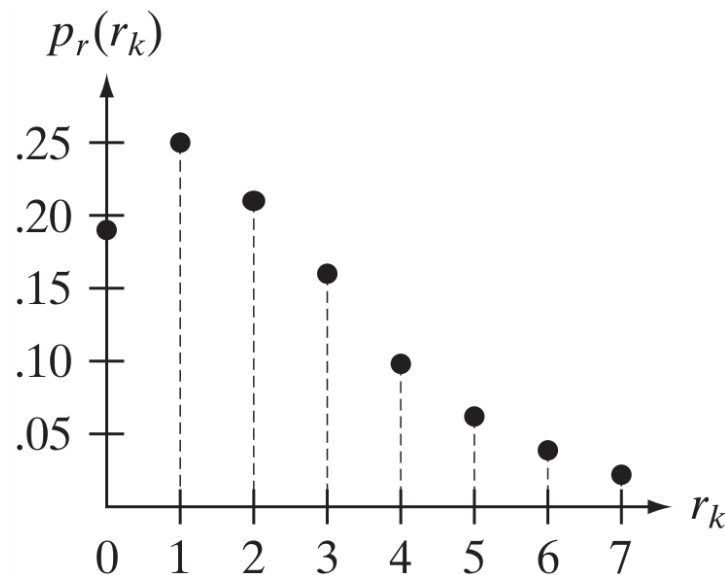
$$\approx (1, 2, 4, 4, 6, 6, 7, 7)$$



# Histogram Processing

## □ Example

Suppose that a 3-bit image of size  $MN$  pixels has the intensity distribution shown in the Table, where the intensity levels are integers in the range  $[0, L-1] = [0, 7]$ .



$r_k$	$n_k$	$p_r(r_k) = n_k/MN$
$r_0 = 0$	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

# Histogram Processing

## □ Example

Using

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

$r_k$	$n_k$	$p_r(r_k) = n_k/MN$
$r_0 = 0$	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

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

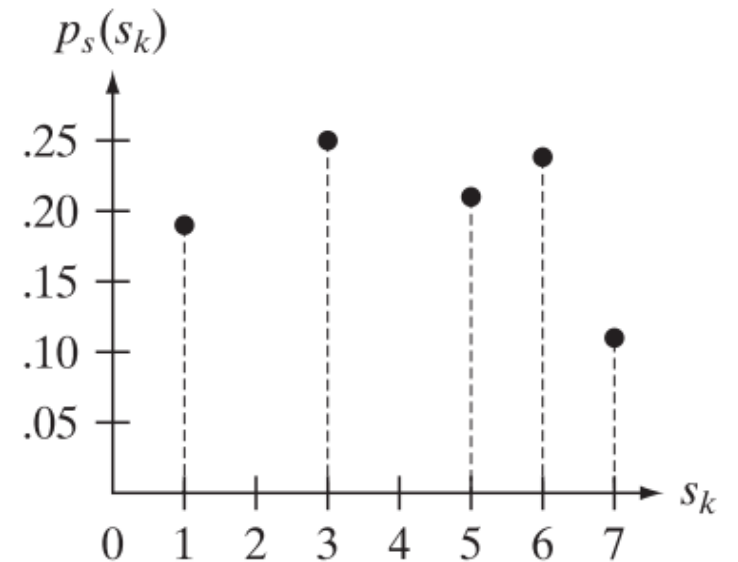
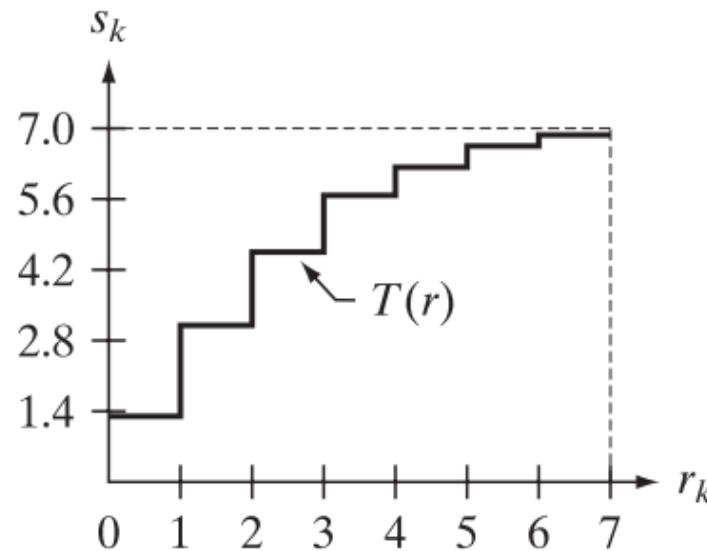
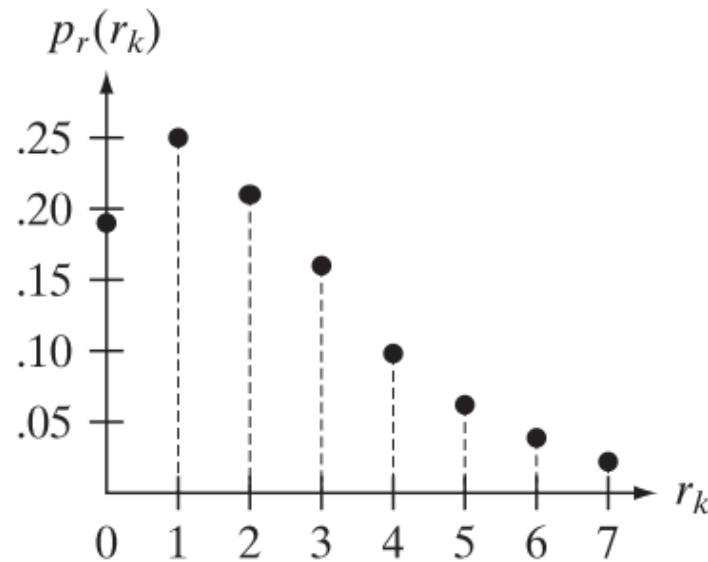
$$s_1 = T(r_1) = 7 \sum_{j=0}^1 p_r(r_j) = 7p_r(r_0) + 7p_r(r_1) = 3.08$$

$$s_2 = 4.55, s_3 = 5.67, s_4 = 6.23, s_5 = 6.65, s_6 = 6.86, s_7 = 7.00.$$

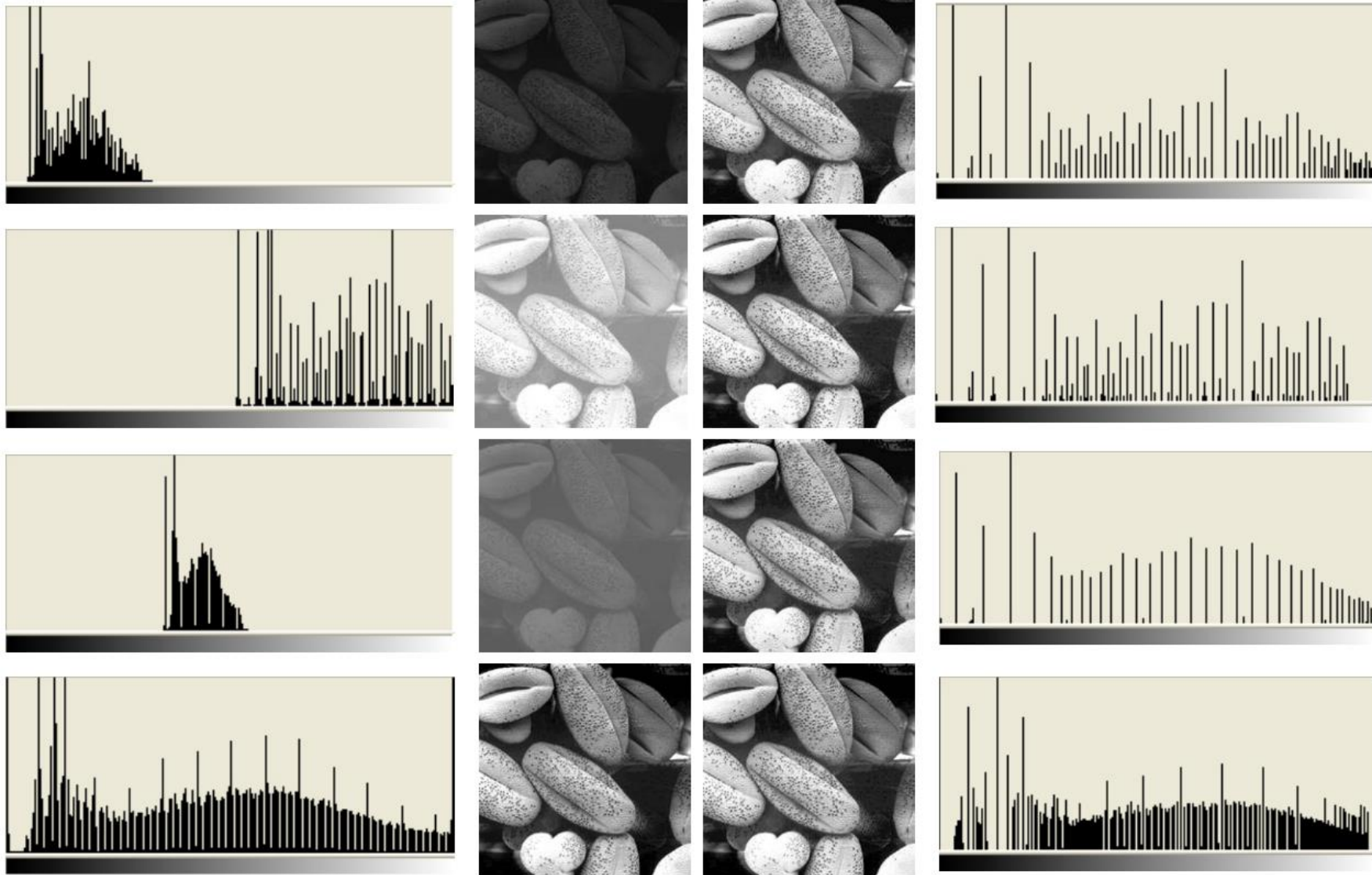
# Histogram Processing

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

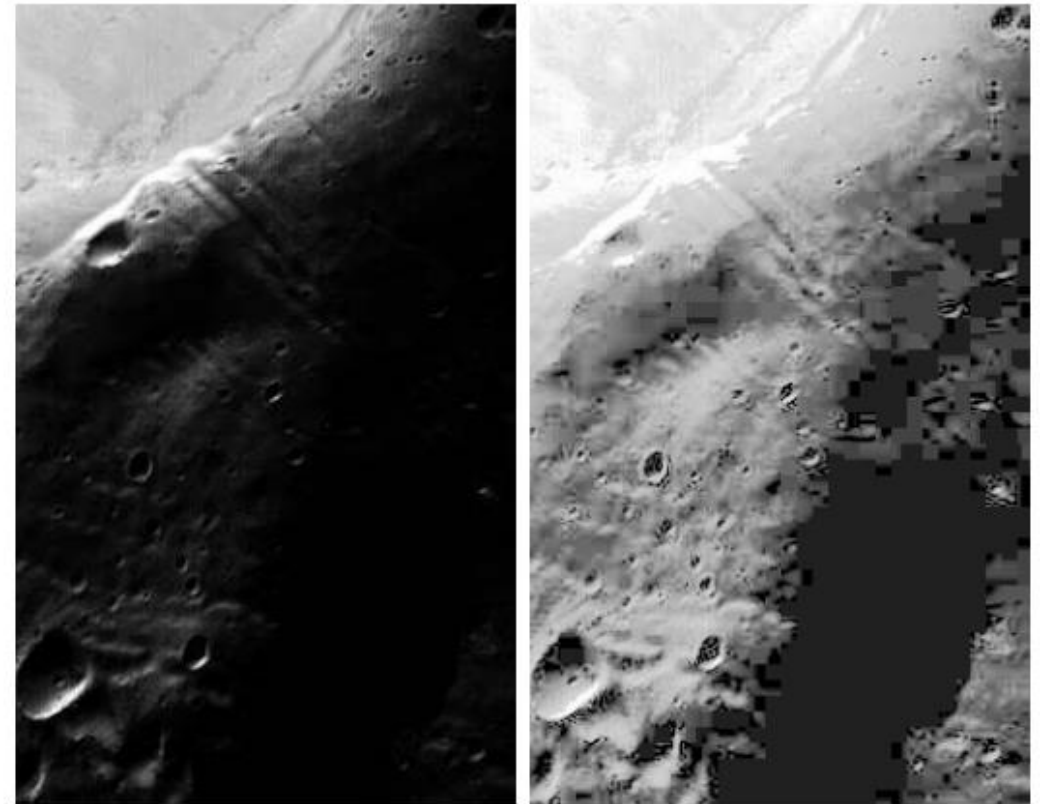
$s_0 = 1.33 \rightarrow 1$	$s_4 = 6.23 \rightarrow 6$
$s_1 = 3.08 \rightarrow 3$	$s_5 = 6.65 \rightarrow 7$
$s_2 = 4.55 \rightarrow 5$	$s_6 = 6.86 \rightarrow 7$
$s_3 = 5.67 \rightarrow 6$	$s_7 = 7.00 \rightarrow 7$



# Histogram Processing

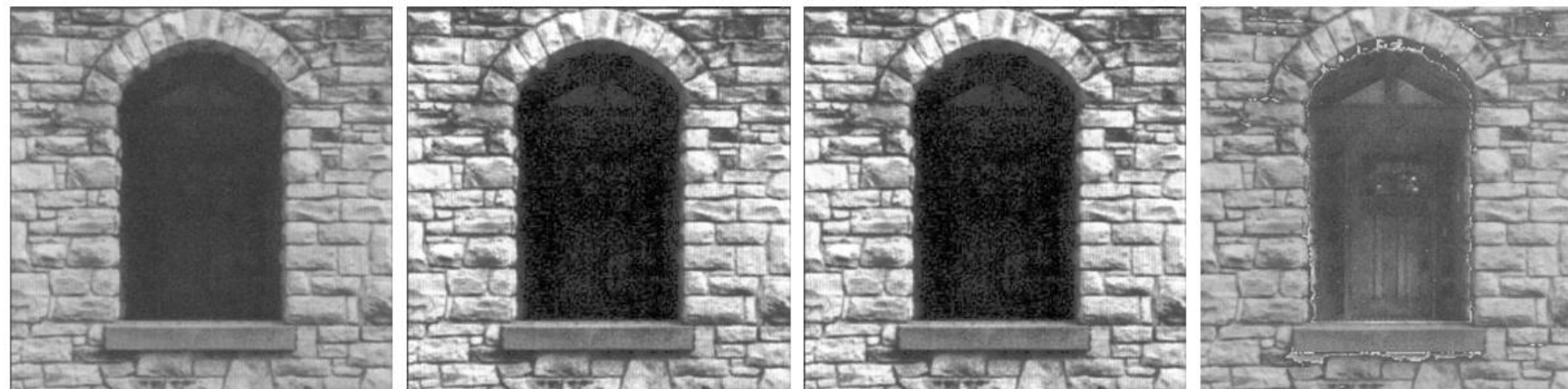


# Histogram Processing



## □ Histogram Specification

- Histogram equalization is an automatic process
- Repeat appliance of H.E. is useless



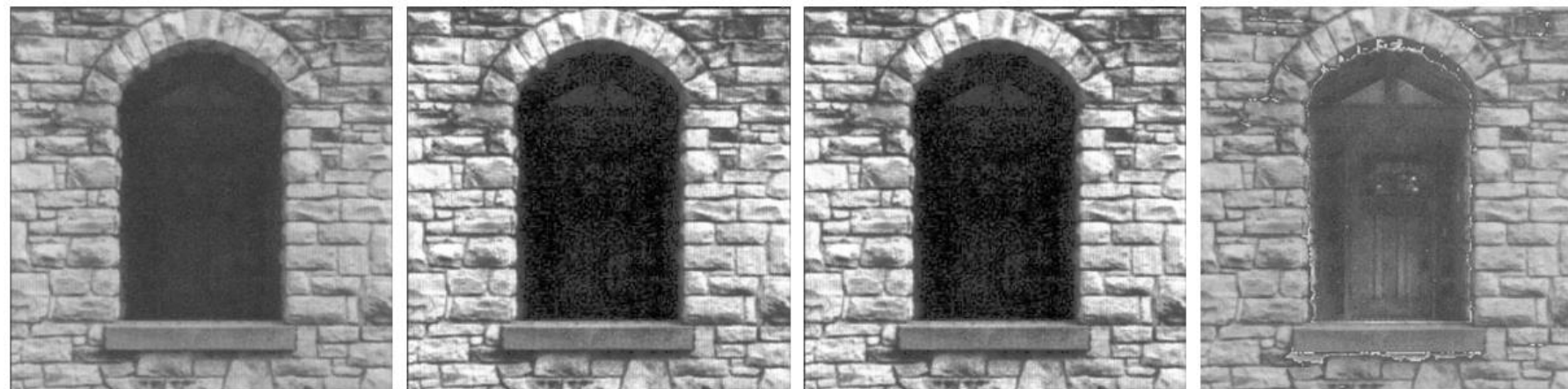
H.E.

H.E.

Histogram Specification ( manual )

## □ Histogram Specification

- Histogram equalization is an automatic process
- Repeat appliance of H.E. is useless



H.E.

H.E.

Histogram Specification ( manual )

## □ Histogram Specification

$$s = T(r) = (L - 1) \int_0^r p_r(w) dw$$

$$G(z) = (L - 1) \int_0^z p_z(t) dt = s$$

Histogram equalization

$r$ : input image intensity

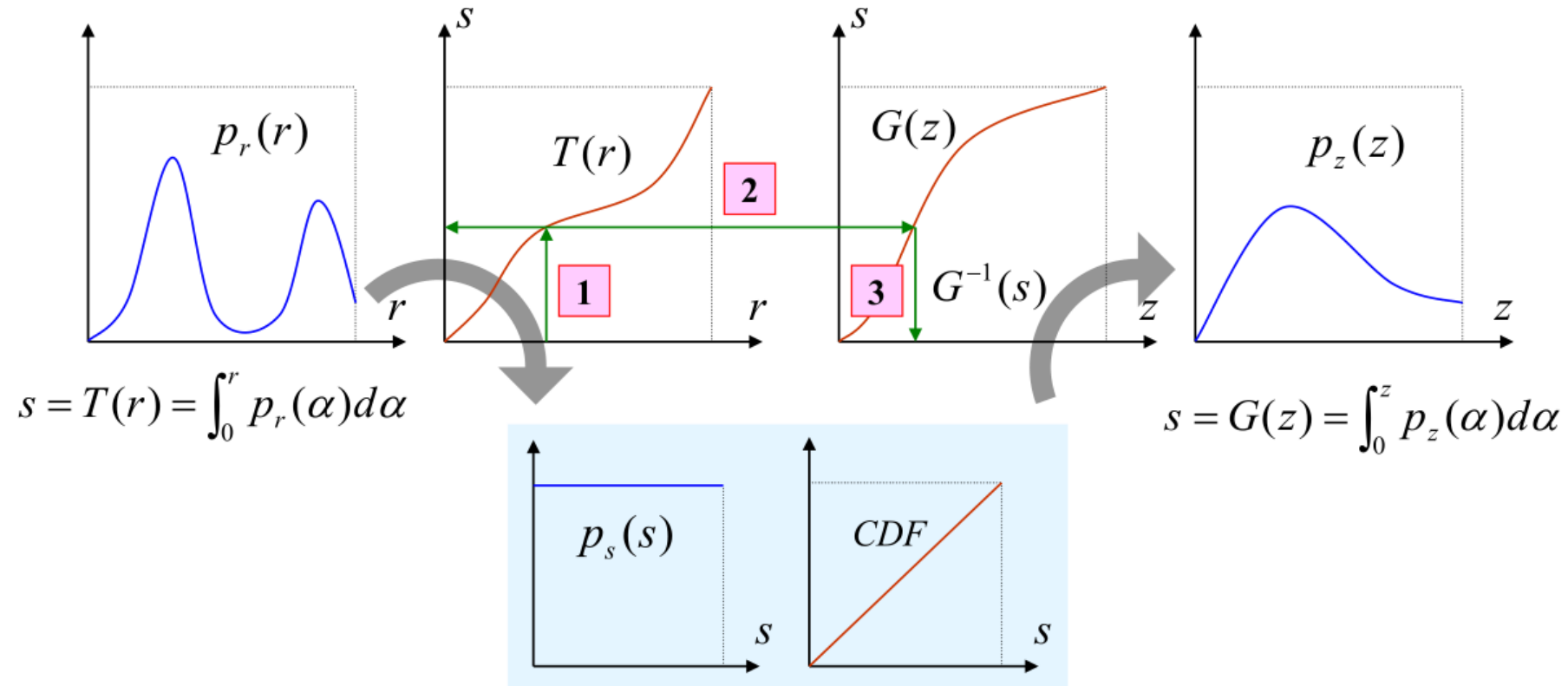
$z$ : desired image intensity

$$z = G^{-1}[T(r)] = G^{-1}(s)$$



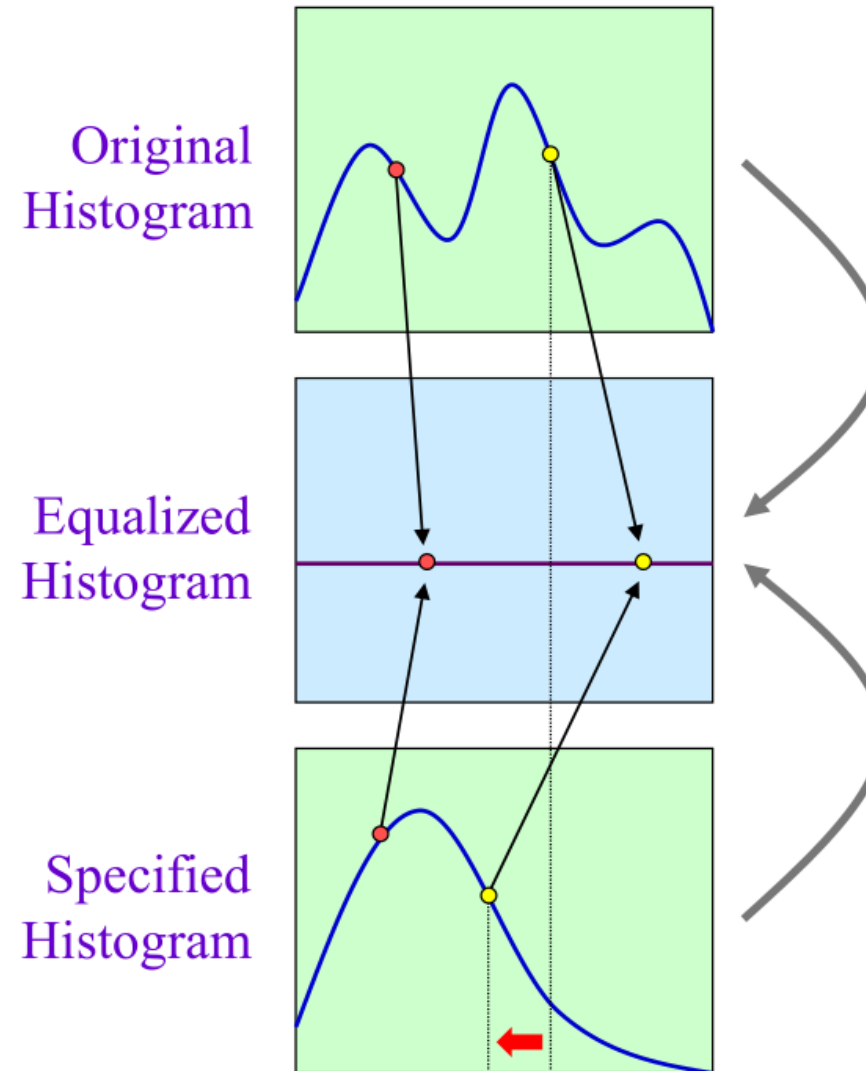
## □ Histogram Specification

$$z = G^{-1}(s) = G^{-1}[T(r)]$$



# Histogram Processing

## □ Histogram Specification

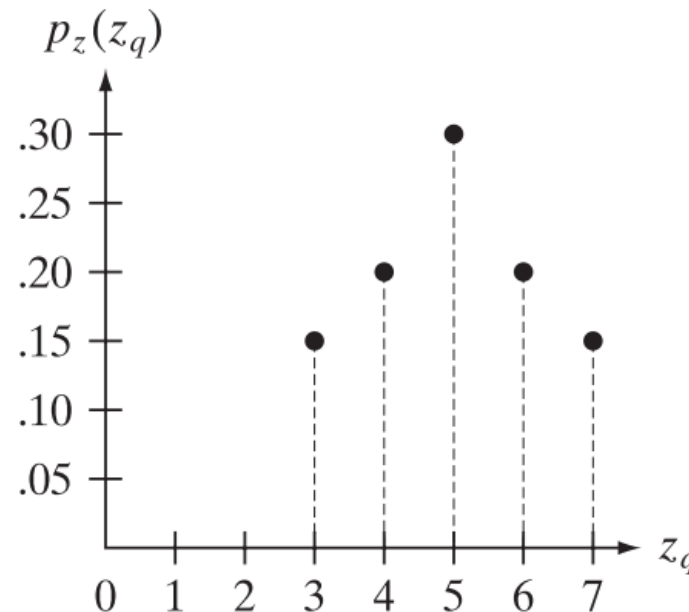
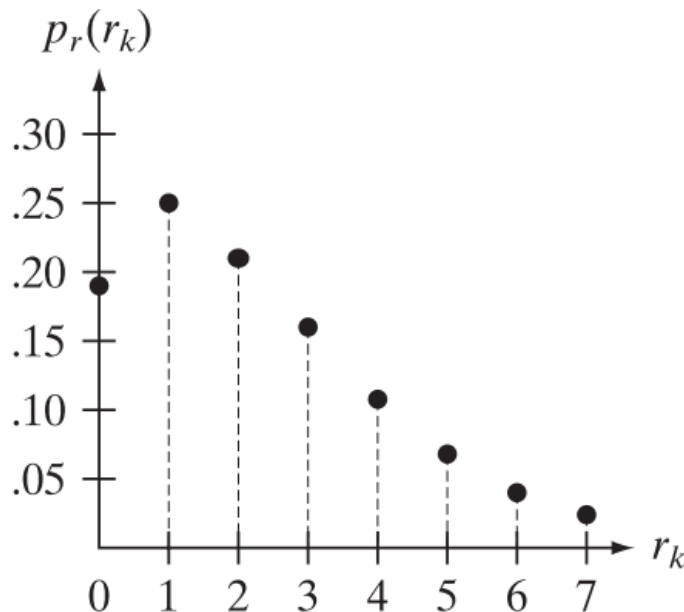


# Histogram Processing

## Example

$r_k$	$n_k$	$p_r(r_k) = n_k/MN$
$r_0 = 0$	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

Consider a  $64 * 64$  hypothetical image, whose histogram is shown in the Figure (a). It is desired to transform this histogram so that it will have the values specified in the second column of the table. Figure (b) shows a sketch of this histogram.



$z_q$	Specified $p_z(z_q)$
$z_0 = 0$	0.00
$z_1 = 1$	0.00
$z_2 = 2$	0.00
$z_3 = 3$	0.15
$z_4 = 4$	0.20
$z_5 = 5$	0.30
$z_6 = 6$	0.20
$z_7 = 7$	0.15

## □ Example (cont'd)

### a) Performing histogram equalization as in the previous example

$$s_0 = T(r_0) = 7 \sum_{i=0}^0 p_r(r_i) = 7p_r(r_0) = 1.33$$
$$s_1 = T(r_1) = 7 \sum_{j=0}^1 p_r(r_j) = 7p_r(r_0) + 7p_r(r_1) = 3.08$$
$$s_2 = 4.55, s_3 = 5.67, s_4 = 6.23, s_5 = 6.65, s_6 = 6.86, s_7 = 7.00.$$

After rounding up,

$$s_0 = 1 \quad s_2 = 5 \quad s_4 = 7 \quad s_6 = 7$$

$$s_1 = 3 \quad s_3 = 6 \quad s_5 = 7 \quad s_7 = 7$$

## □ Example (cont'd)

### b) Computing all the values of the transformation function, $G$ , using

$$G(z_q) = (L - 1) \sum_{i=0}^q p_z(z_i)$$

we could get

$$G(z_0) = 7 \sum_{j=0}^0 p_z(z_j) = 0.00$$

$$G(z_1) = 7 \sum_{j=0}^1 p_z(z_j) = 7[p(z_0) + p(z_1)] = 0.00$$

and

$$G(z_2) = 0.00 \quad G(z_4) = 2.45 \quad G(z_6) = 5.95$$

$$G(z_3) = 1.05 \quad G(z_5) = 4.55 \quad G(z_7) = 7.00$$

# Histogram Processing

## □ Example (cont'd)

After rounding-up:

$$G(z_0) = 0.00 \rightarrow 0$$

$$G(z_1) = 0.00 \rightarrow 0$$

$$G(z_2) = 0.00 \rightarrow 0$$

$$G(z_3) = 1.05 \rightarrow 1$$

$$G(z_4) = 2.45 \rightarrow 2$$

$$G(z_5) = 4.55 \rightarrow 5$$

$$G(z_6) = 5.95 \rightarrow 6$$

$$G(z_7) = 7.00 \rightarrow 7$$

$$G(z_0) = 7 \sum_{j=0}^0 p_z(z_j) = 0.00$$

$$G(z_1) = 7 \sum_{j=0}^1 p_z(z_j) = 7[p(z_0) + p(z_1)] = 0.00$$

$$G(z_2) = 0.00 \quad G(z_4) = 2.45 \quad G(z_6) = 5.95$$

$$G(z_3) = 1.05 \quad G(z_5) = 4.55 \quad G(z_7) = 7.00$$

## □ Example (cont'd)

c) Finding the smallest value of  $z_q$  so that the value  $G(z_q)$  is the closest to  $s_k$ .

We do this for every value of  $s_k$  to create the required mappings from  $s$  to  $z$ . For example,  $s_0 = 1$ , and we see that  $G(z_3) = 1$ , which is a perfect match in this case, so we have the correspondence  $s_0 \rightarrow z_3$ .

$s_k$	$G(z_q)$	$z_q$
$s_0 = 1$	0	$z_0 = 0$
$s_1 = 3$	0	$z_1 = 1$
$s_2 = 5$	0	$z_2 = 2$
$s_3 = 6$	1	$z_3 = 3$
$s_4 = 7$	2	$z_4 = 4$
$s_5 = 7$	5	$z_5 = 5$
$s_6 = 7$	6	$z_6 = 6$
$s_7 = 7$	7	$z_7 = 7$

# Histogram Processing

## Example (cont'd)

$s_k$	$G(z_q)$	$z_q$
$s_0 = 1$	0	$z_0 = 0$
$s_1 = 3$	0	$z_1 = 1$
$s_2 = 5$	0	$z_2 = 2$
$s_3 = 6$	1	$z_3 = 3$
$s_4 = 7$	2	$z_4 = 4$
$s_5 = 7$	5	$z_5 = 5$
$s_6 = 7$	6	$z_6 = 6$
$s_7 = 7$	7	$z_7 = 7$



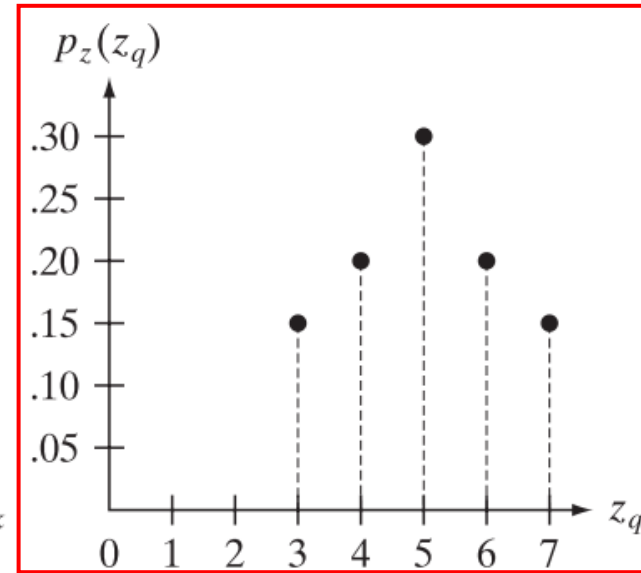
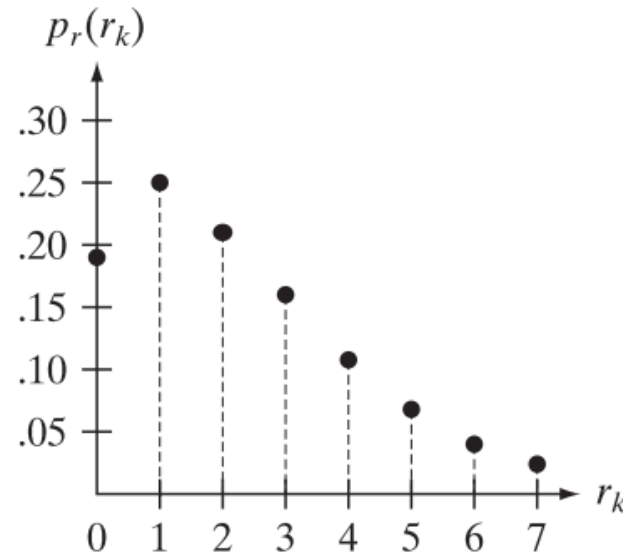
$s_k$	$\rightarrow$	$z_q$
1	$\rightarrow$	3
3	$\rightarrow$	4
5	$\rightarrow$	5
6	$\rightarrow$	6
7	$\rightarrow$	7



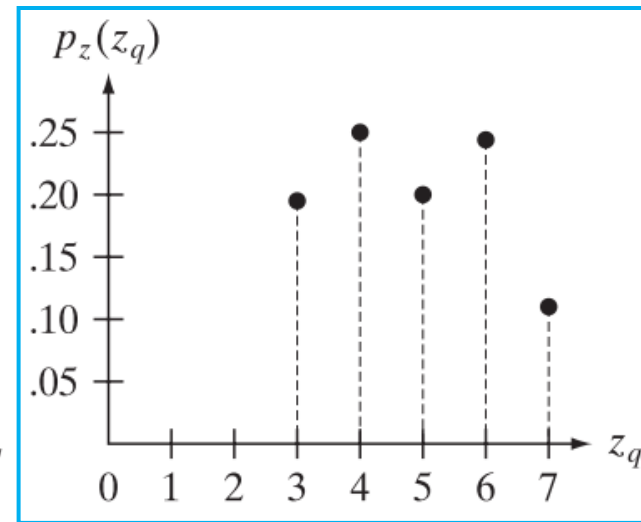
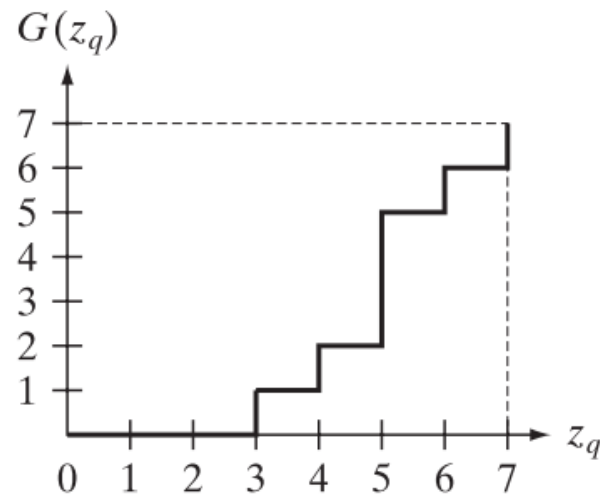
# Histogram Processing

## Example (cont'd)

Finally,



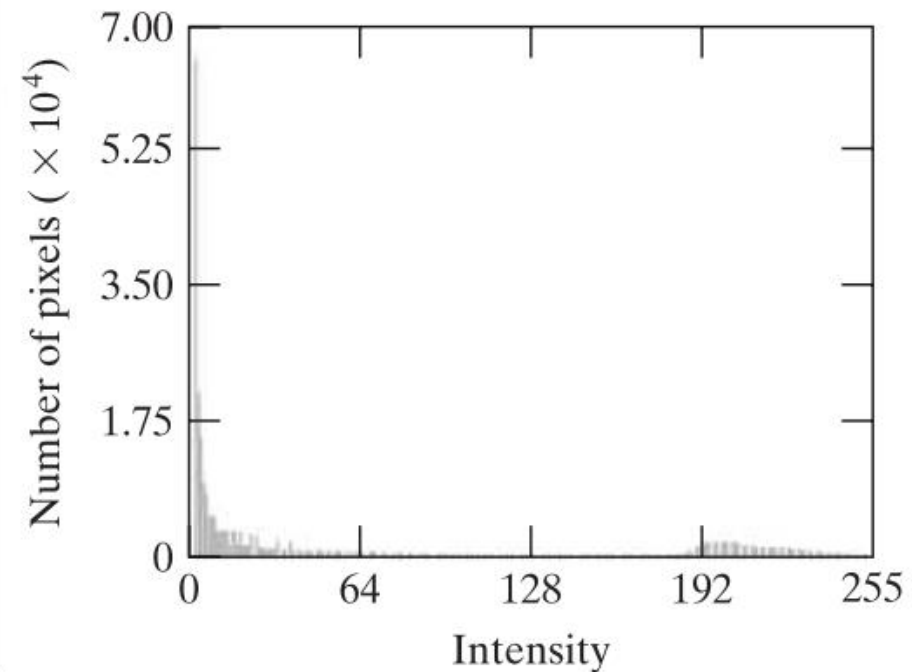
Specified histogram



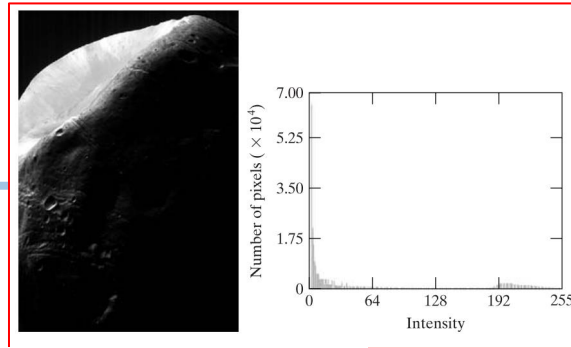
After performing histogram specification

# Histogram Processing

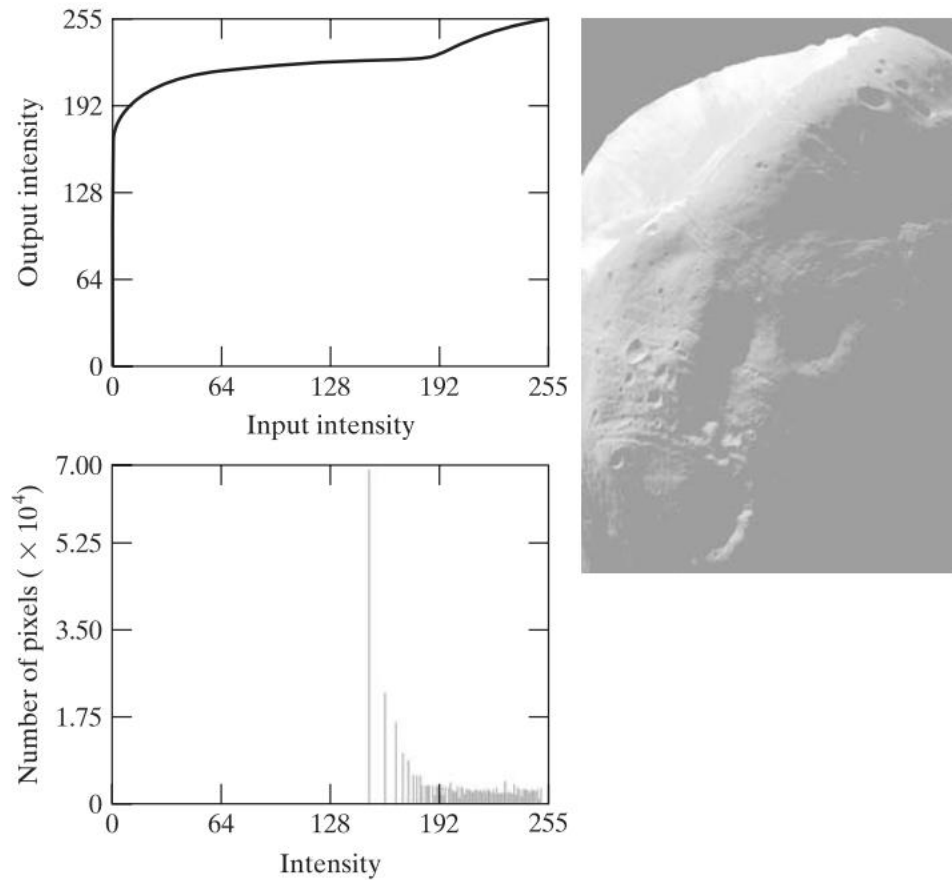
- Histogram equalization v.s. histogram specification



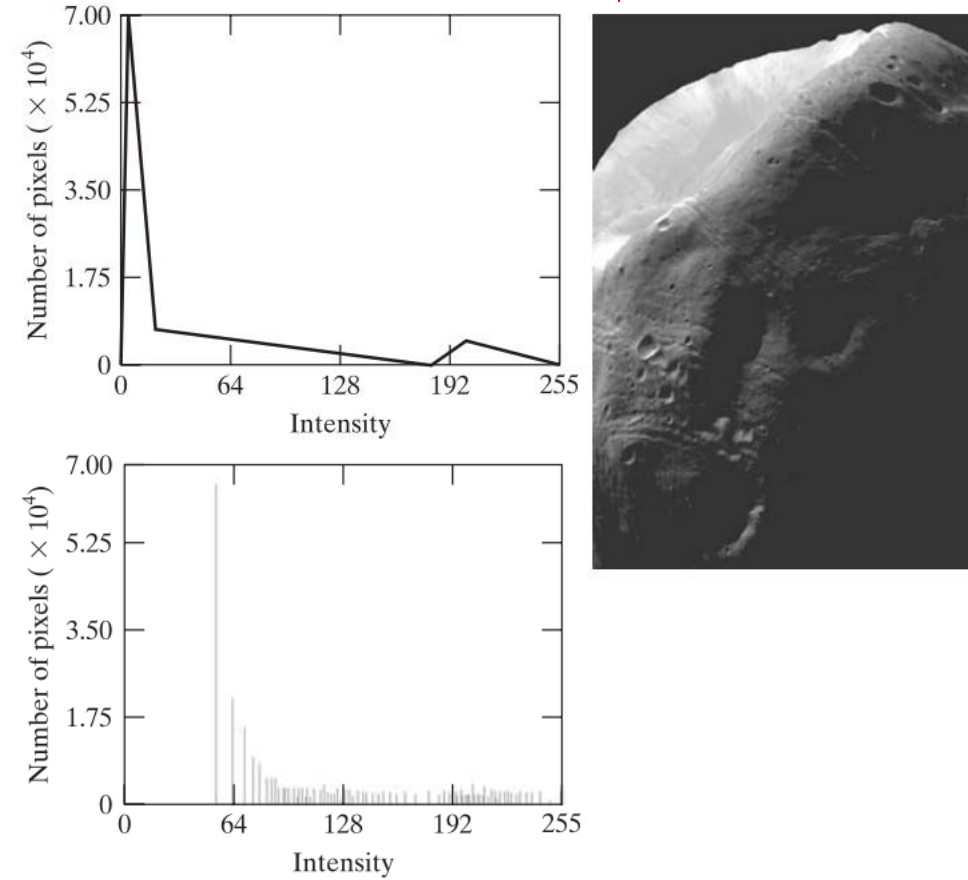
# Histogram



## □ Histogram equalization v.s. histogram specification



H.E.



H.S.

# Summary

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- Background
- Intensity Transformation
- Histogram Processing

Next

Lecture 3: Get Hand Dirty by Coding



# Thank You!