



# **IMAGE PROCESSING**

## **Chapter 3: Image enhancement**

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# 1. What is image enhancement

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**Enhancement** is the process of manipulating an image so that the result is more suitable than the original for **a specific application**. Thus, for example, a method that is quite useful for enhancing X-ray images may not be the best approach for enhancing infrared images.

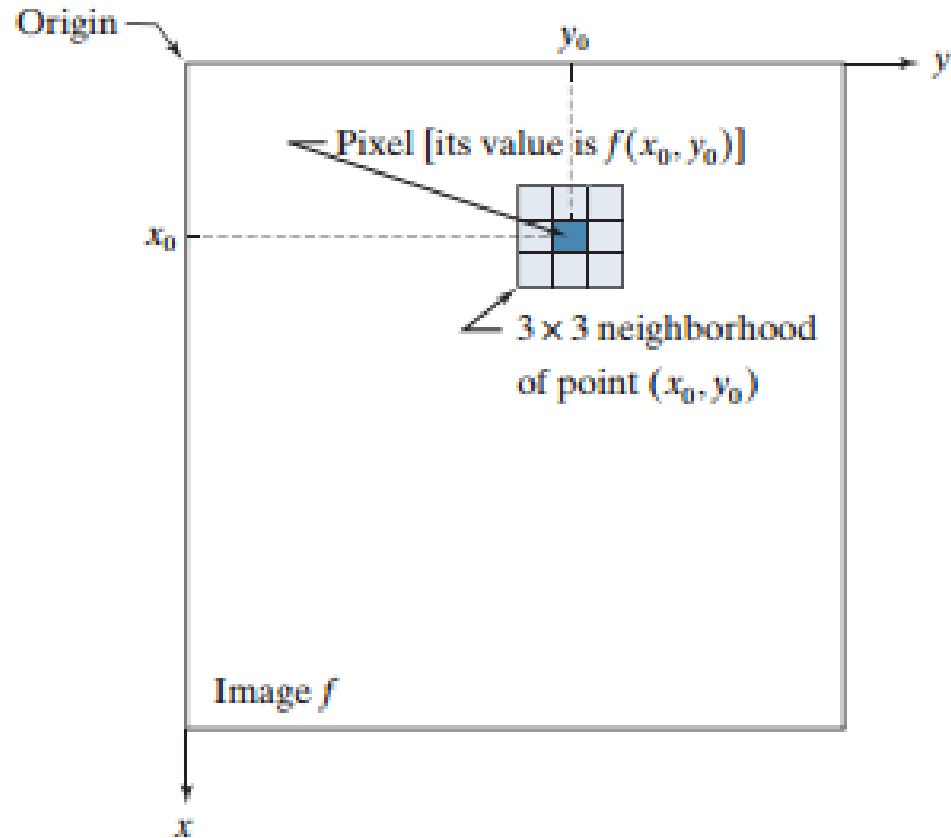
**The role of enhancement** is to adapt the image to better serve its intended use, whether for human interpretation or machine-based analysis: Improving Visual Interpretation, Facilitating Machine Perception, Pre-processing for Analysis, Aesthetic Improvement



# Spatial domain, some basic intensity transformation functions 4

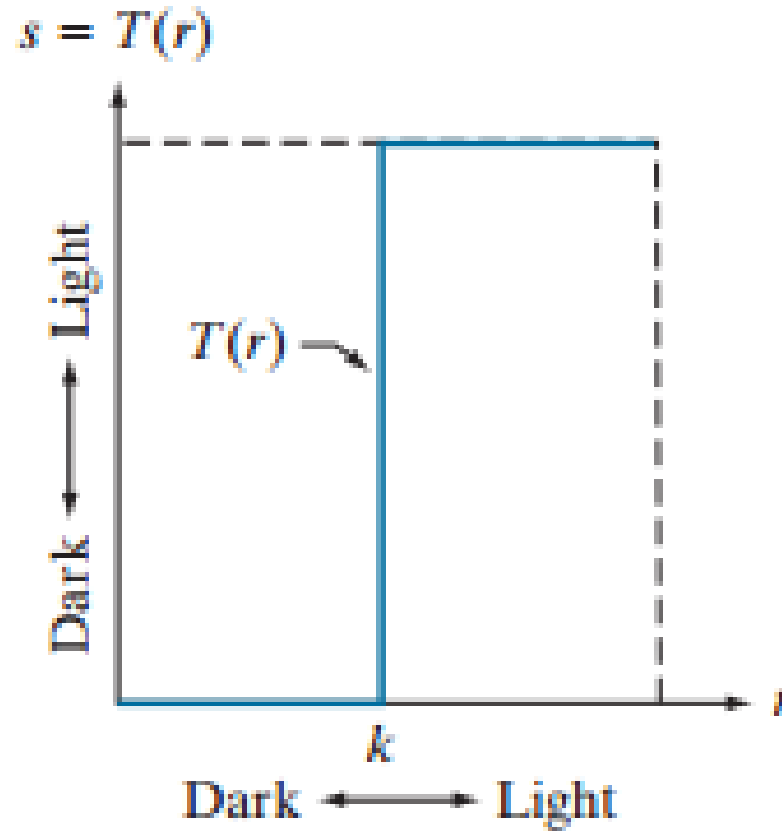
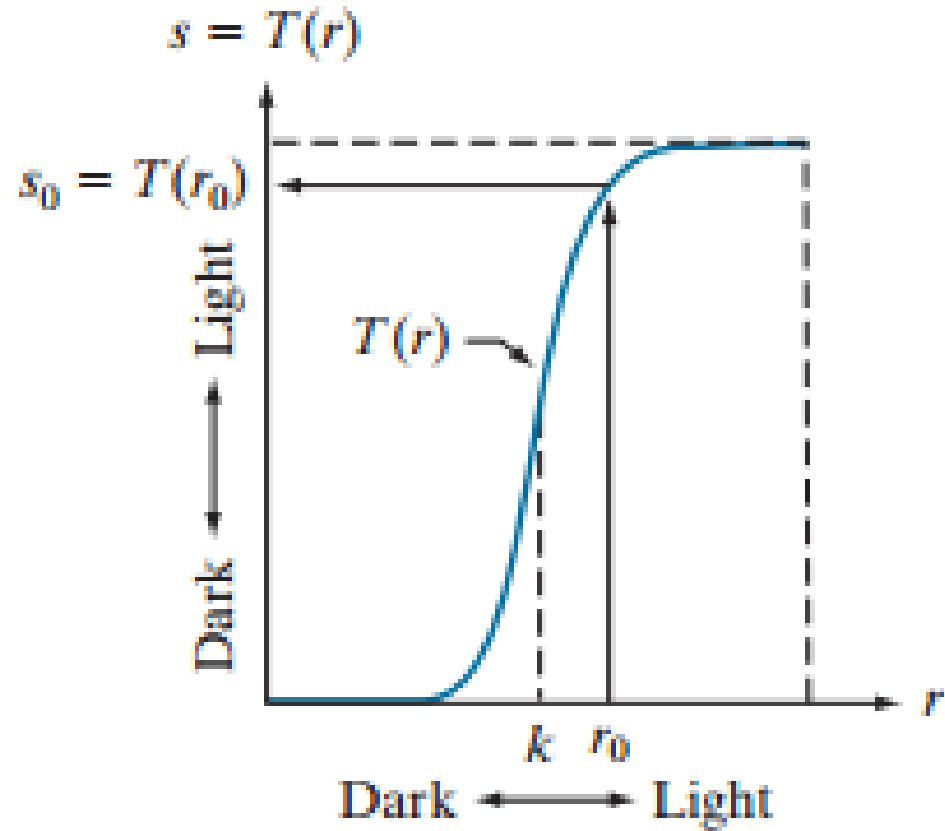
The spatial domain processes we discuss in this chapter are based on

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



$f(x, y)$  is an input image,  $g(x, y)$  is the output image, and  $T$  is an operator on  $f$  defined over a neighborhood of point  $(x, y)$

# Some basic intensity transformation functions



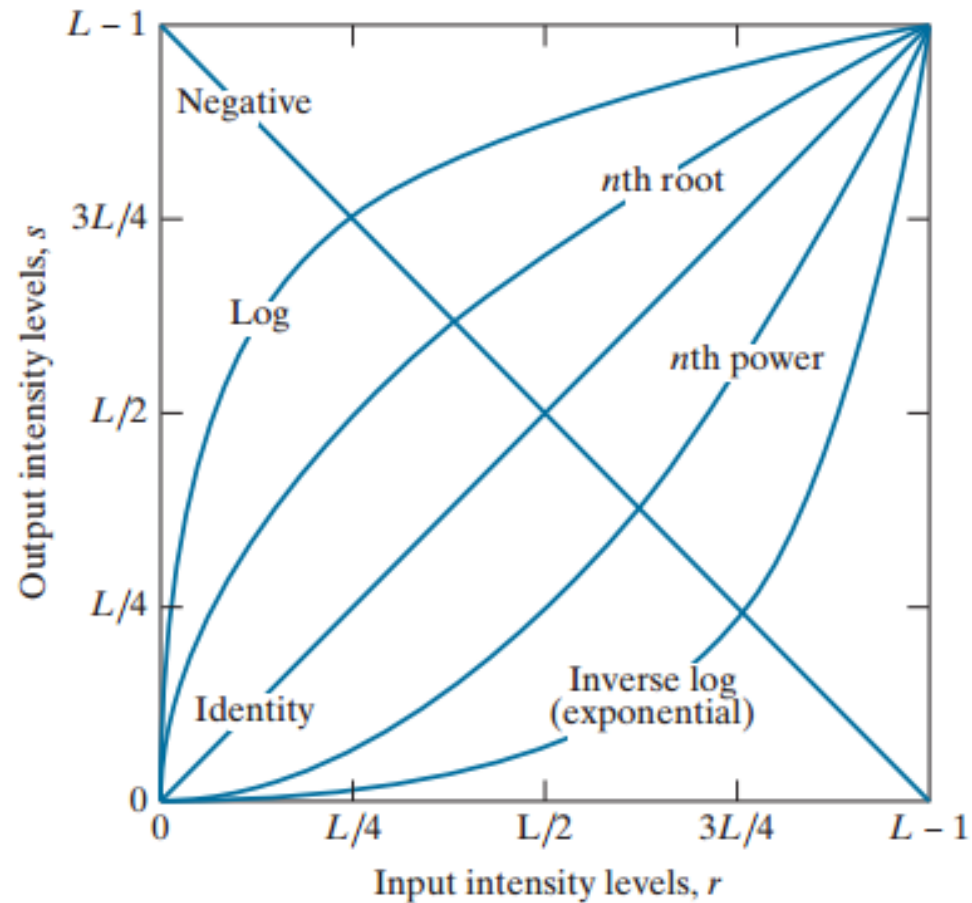
The smallest possible neighborhood is of size  $1 \times 1$ .  $s = T(r)$



# Some basic intensity transformation functions

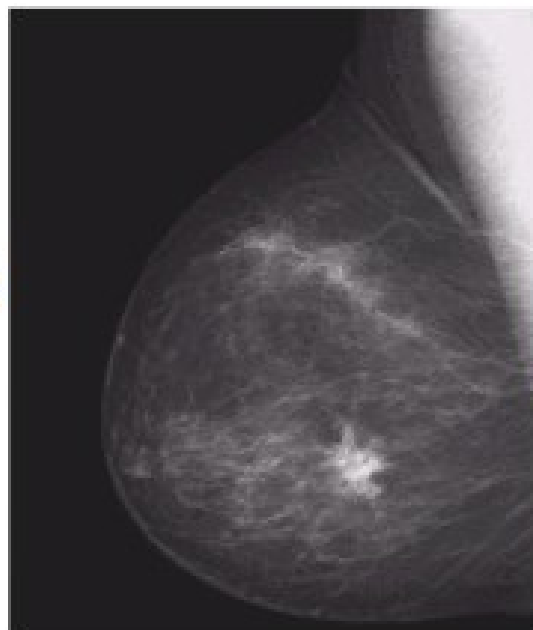
4

Image negatives  $S=L-1-r$ , intensity levels in the range  $[0, L-1]$

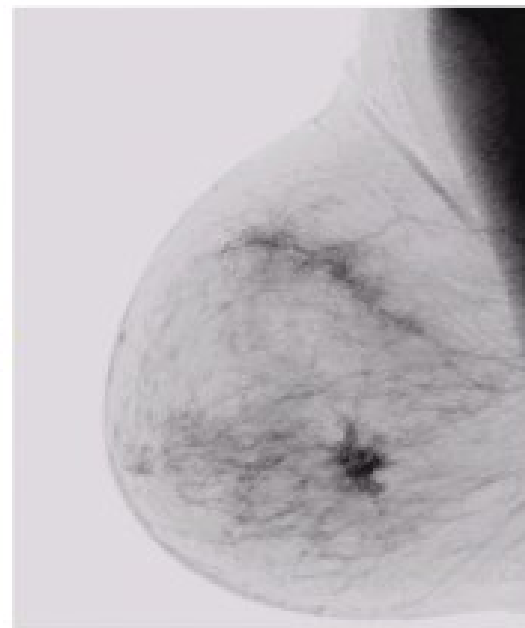


# Image negatives

Ảnh gốc



$$s = 1.0 - r$$



Ảnh âm bản



# Thresholding transformation

- Is used to highlight the objects of interest against the background of the image

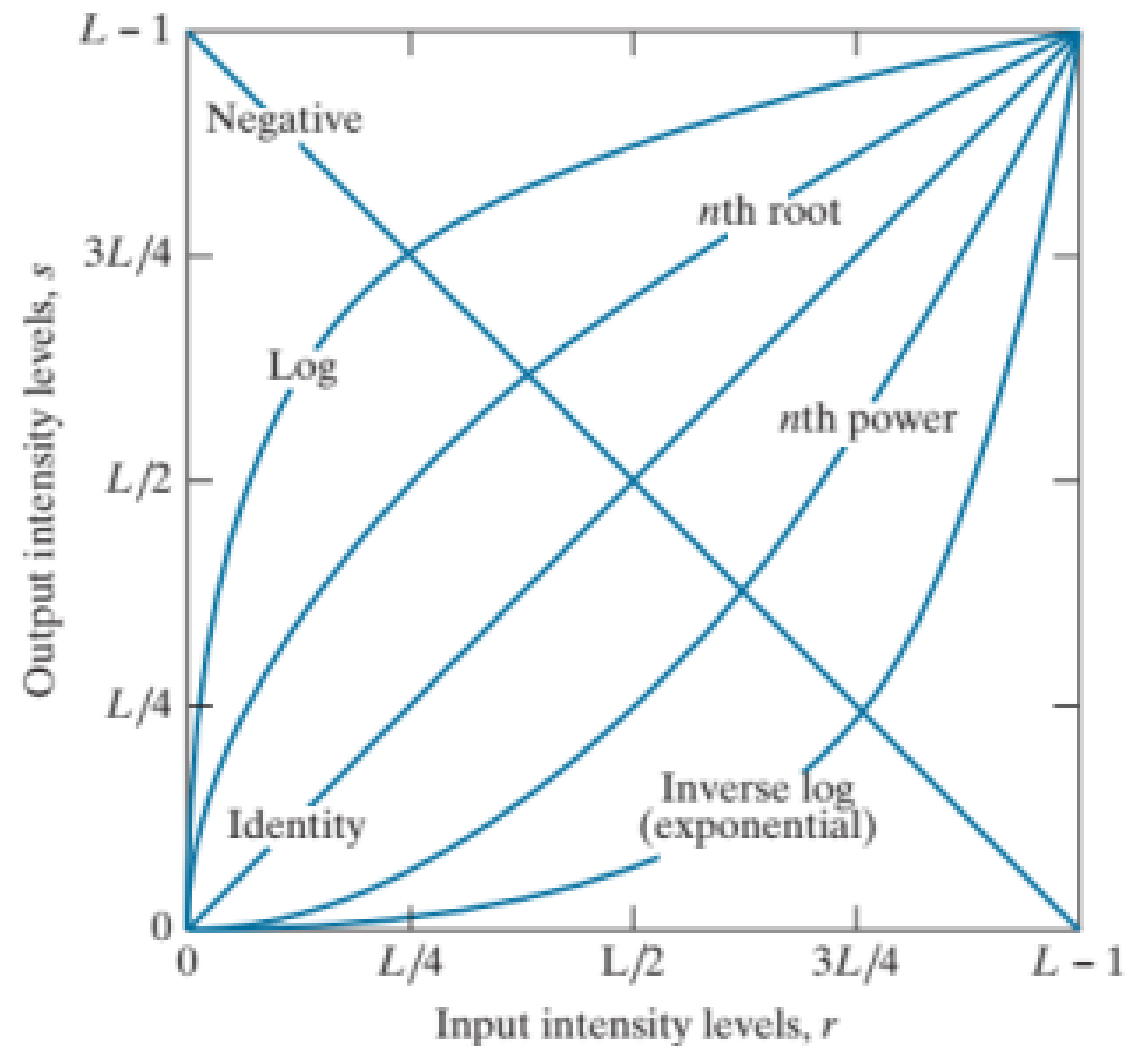


$$s = \begin{cases} 1.0 & r > \text{threshold} \\ 0.0 & r \leq \text{threshold} \end{cases}$$



# Some grayscale transformation functions

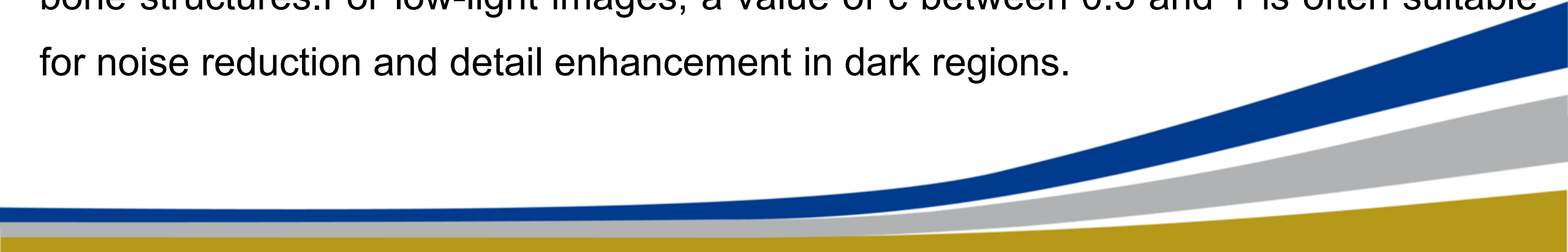
- Biến đổi tuyến tính  
Negative/Identity
- Biến đổi logarithm  
Log/Inverse log
- Biến đổi hàm mũ  
 $n^{\text{th}}$  power/ $n^{\text{th}}$  root



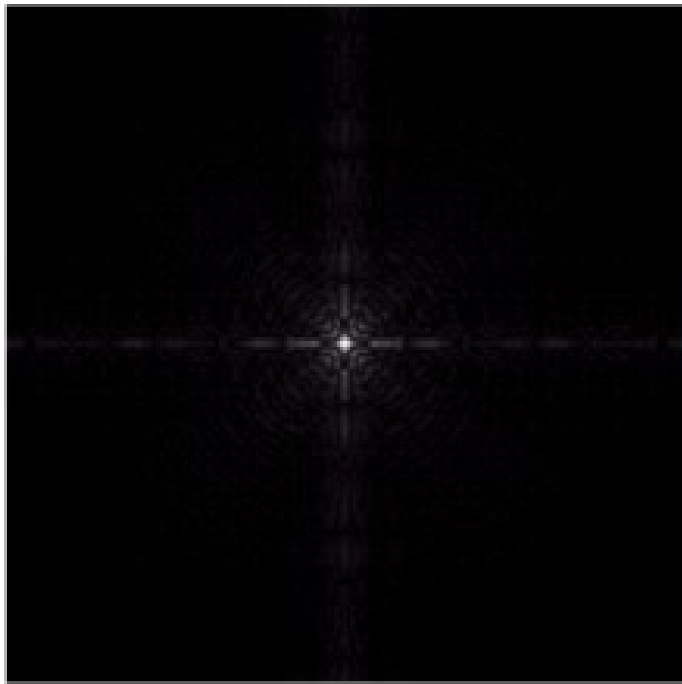
# Logarithm transformation

- A technique used to enhance the brightness of dark regions in an image while compressing the bright regions. This helps to highlight details in the dark areas, resulting in an image with higher
- $s = c * \log(1 + r)$ ;  $s$  is the output pixel value,  $r$  is the input pixel value,  $c$  is a scaling constant.

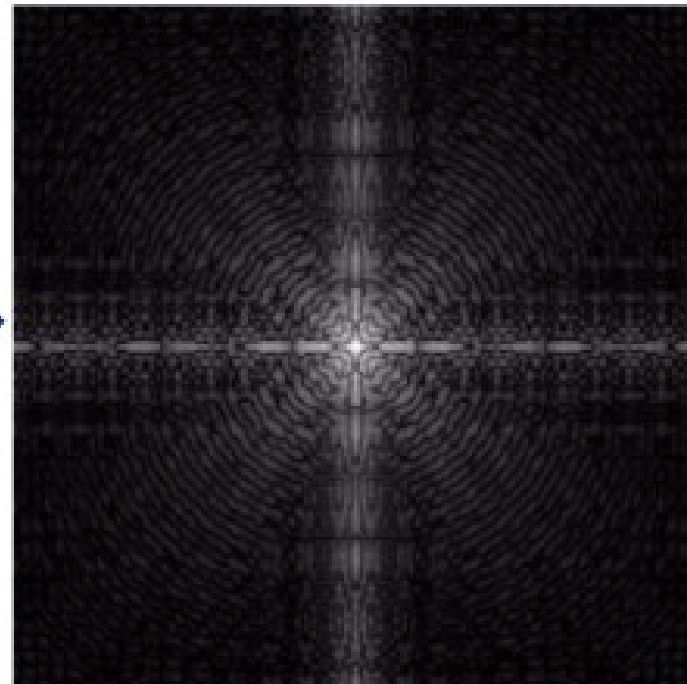
X-ray images, a larger value of  $c$ , e.g.,  $c = 2$ , can be used to enhance the visibility of bone structures. For low-light images, a value of  $c$  between 0.5 and 1 is often suitable for noise reduction and detail enhancement in dark regions.



# Logarithm transformation

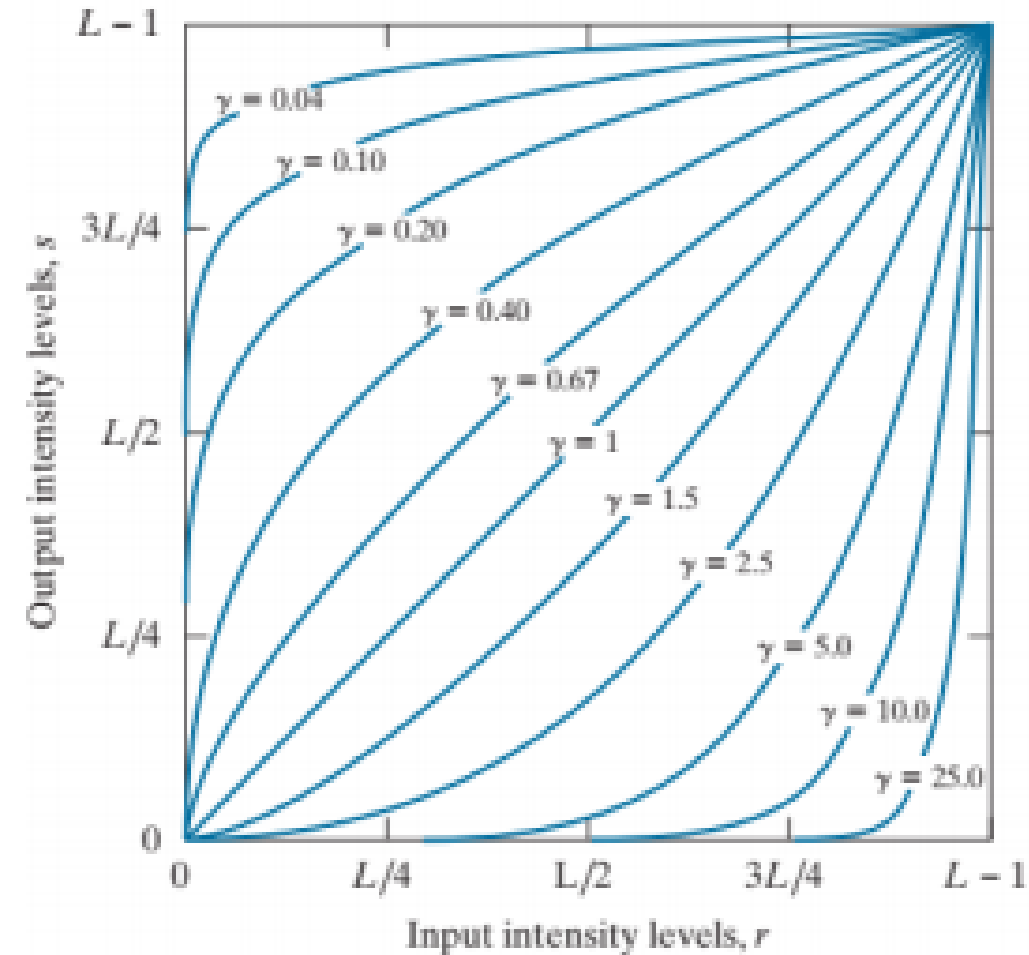


$$s = \log(1 + r)$$

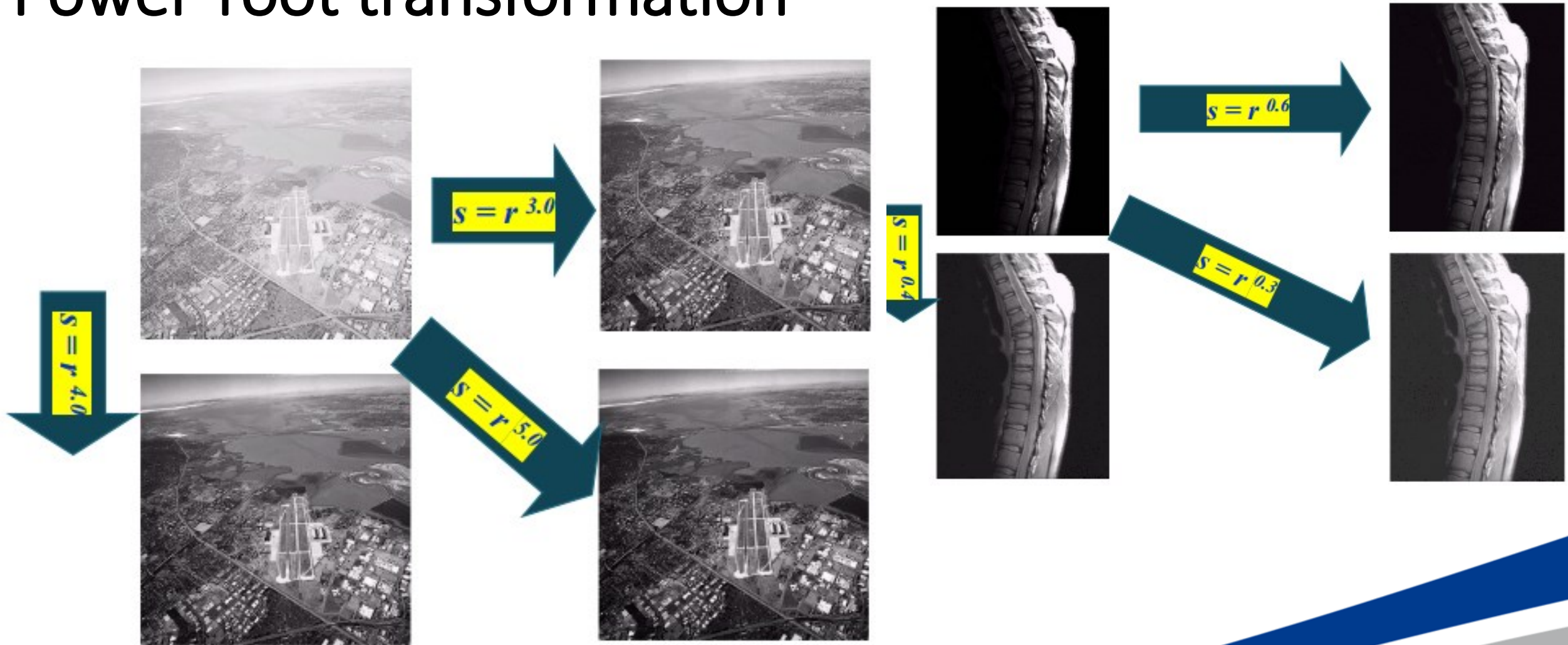


# Power-root transformation

- $s = c * r^\gamma$ ;  $s$  -the output pixel value,  $r$ -the input pixel value,  $c$ -a scaling factor, and  $\gamma$  -the gamma value.
- The effect of the transformation depends on the value of  $\gamma$ . For  $\gamma < 1$ , the transformation compresses the higher intensity values and expands the lower ones, enhancing the visibility of details in dark regions. Conversely, for  $\gamma > 1$ , the transformation compresses the lower intensity values and expands the higher ones, reducing the risk of clipping in bright regions.

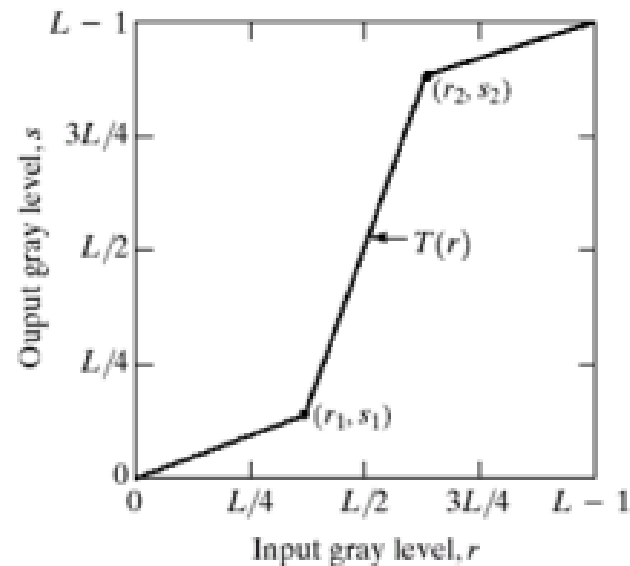
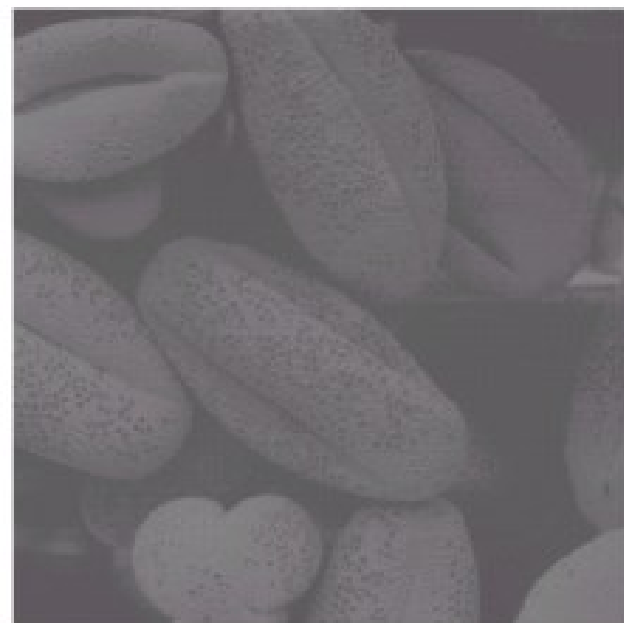


# Power-root transformation

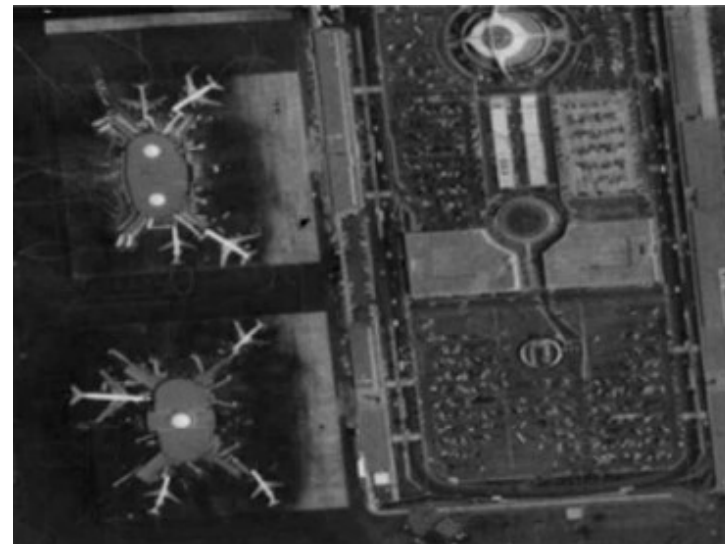
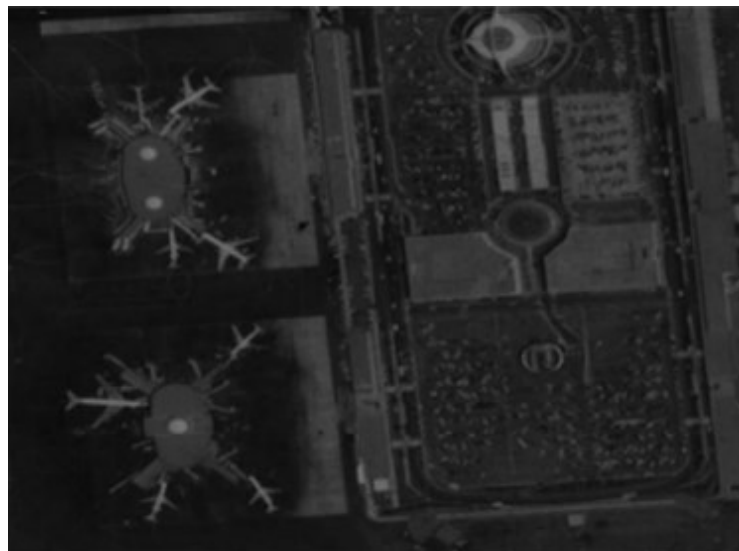


# Comments

- By segmenting the image and applying different mathematical functions to each segment, we can achieve a more flexible and tailored intensity transformation.



# Contrast Stretching Giãn độ tương phản

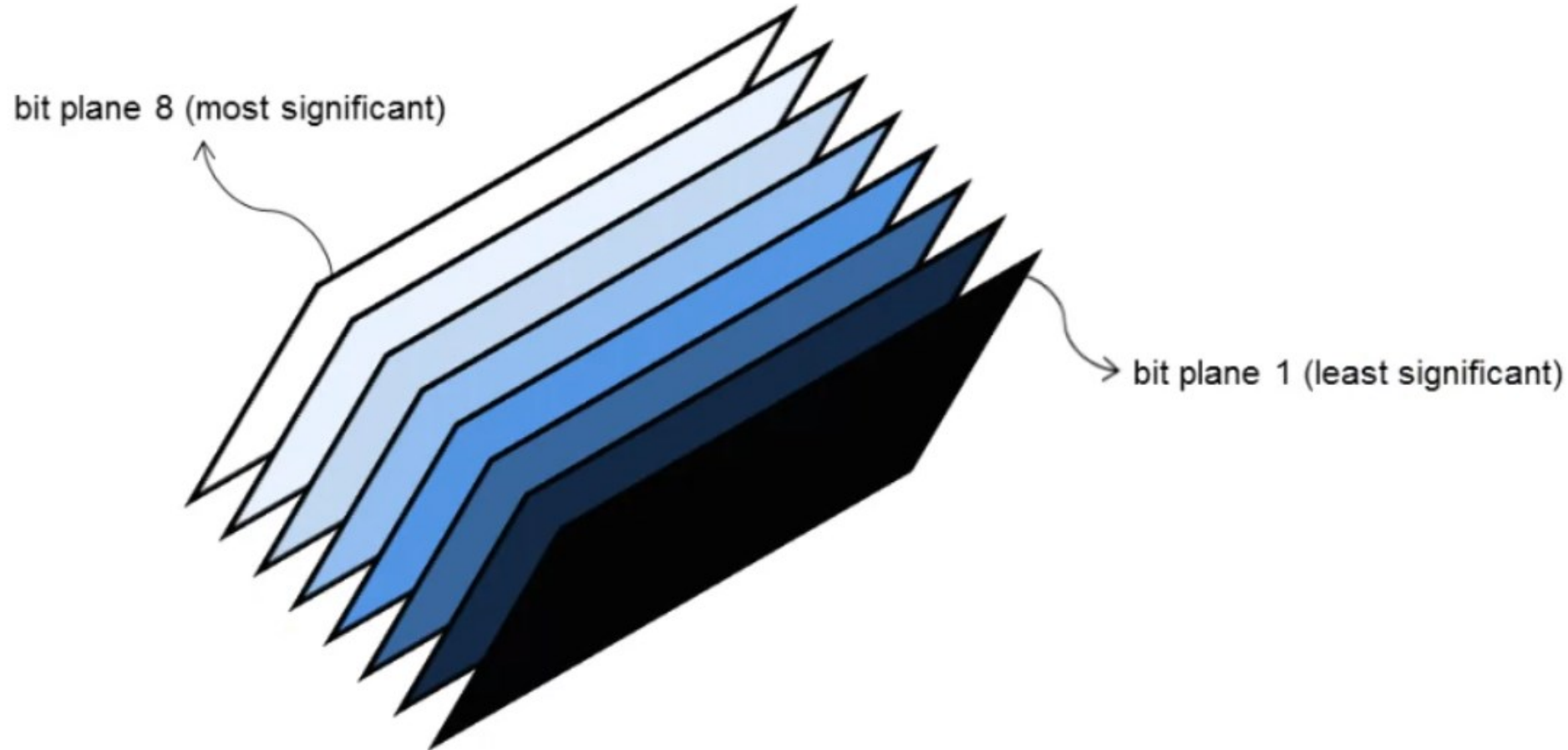


$$y(\mathbf{n}) = \frac{2^B - 1}{x_{max} - x_{min}} (x(\mathbf{n}) - x_{min})$$

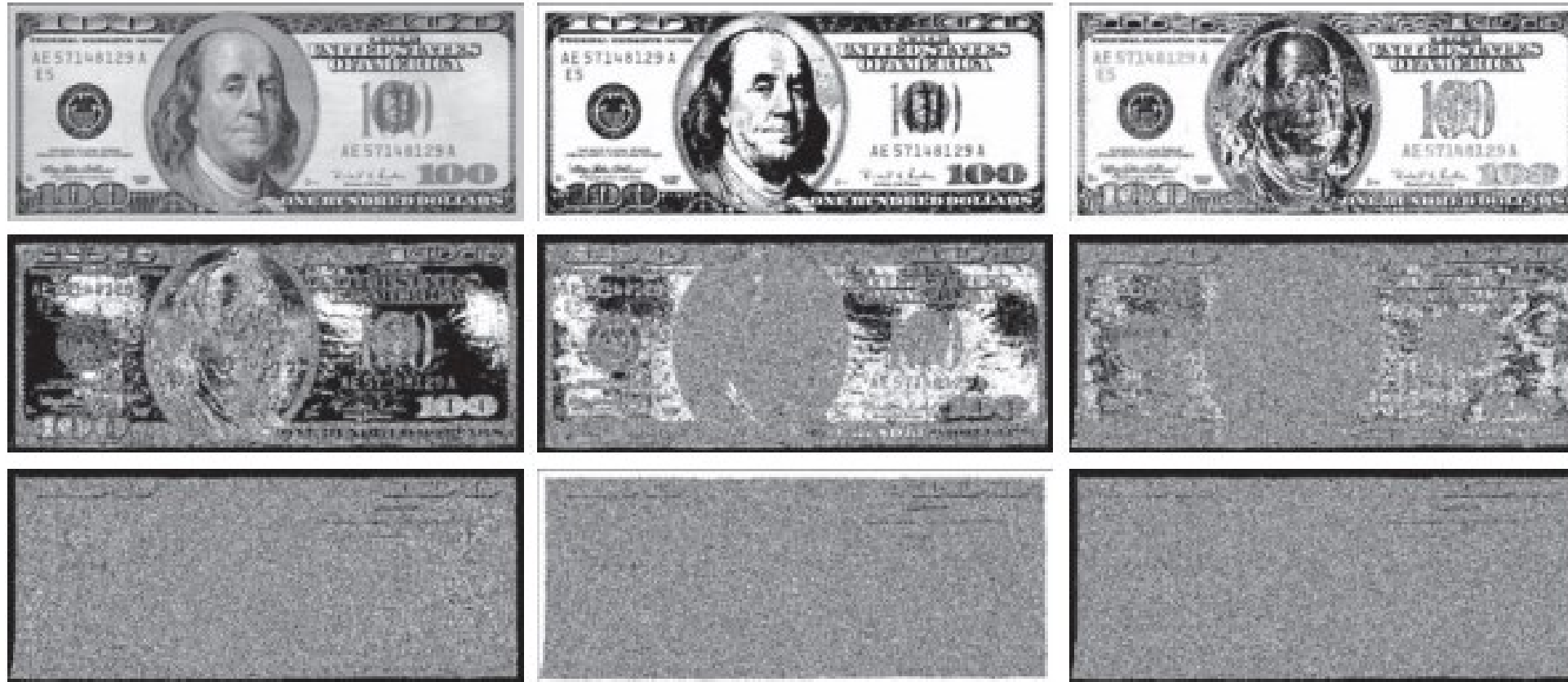
$$x_{max} = \max_{\mathbf{n}} x(\mathbf{n}) \quad x_{min} = \min_{\mathbf{n}} x(\mathbf{n})$$



# Bit-Plane slicing- Cắt theo mặt phẳng bít



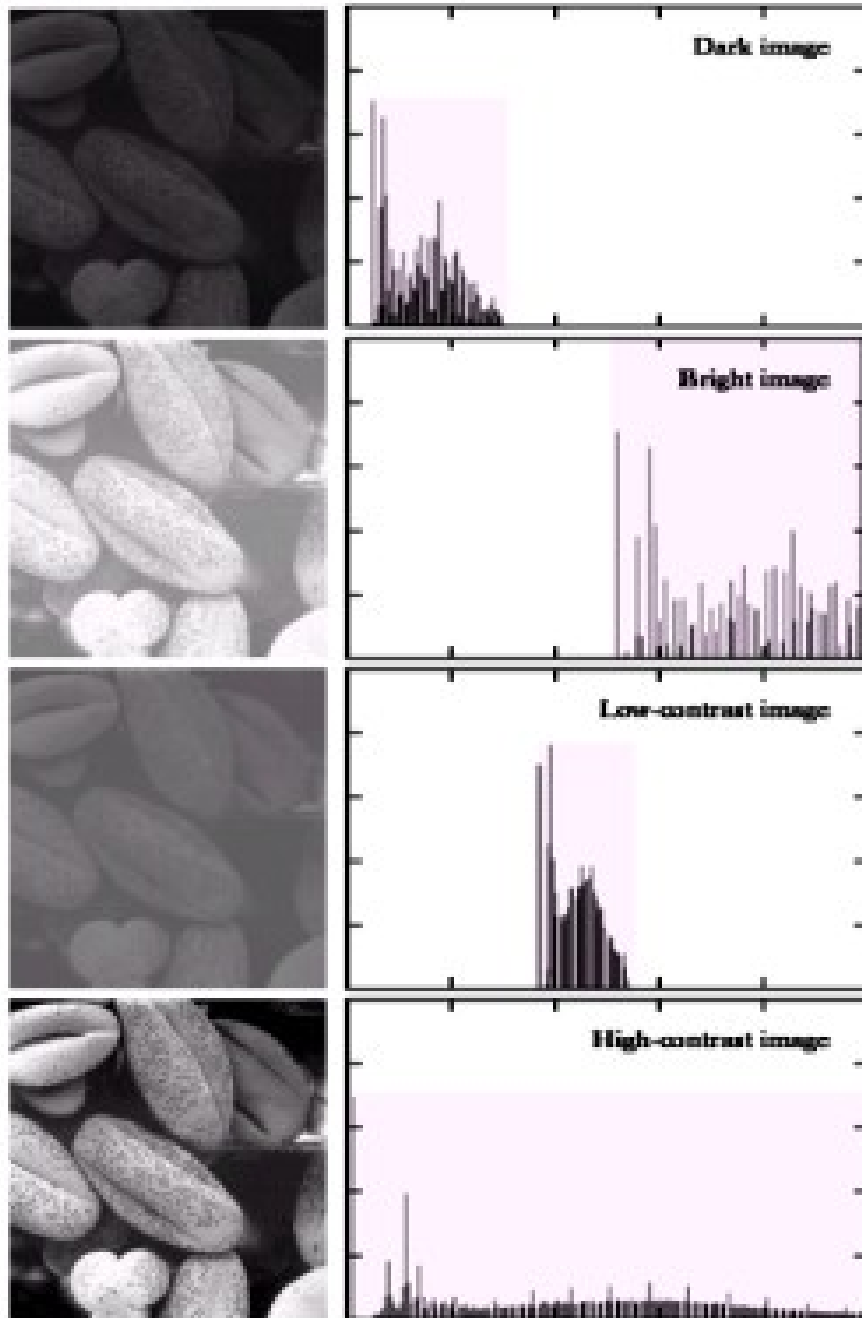
# Bit-Plane slicing- Cắt theo mặt phẳng bít



a	b	c
d	e	f
g	h	i

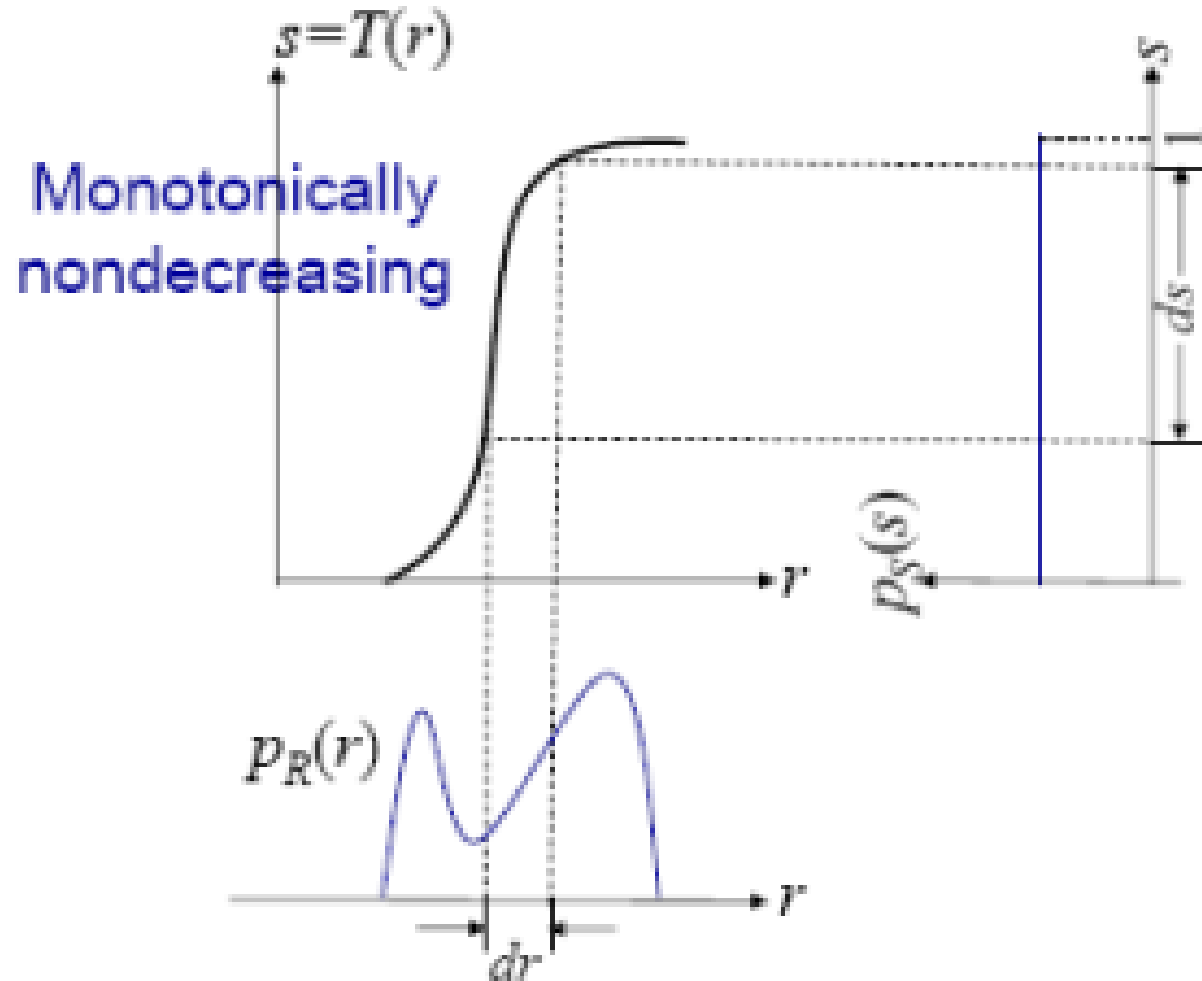
**FIGURE 3.14** (a) An 8-bit gray-scale image of size  $550 \times 1192$  pixels. (b) through (i) Bit planes 8 through 1, with bit plane 1 corresponding to the least significant bit. Each bit plane is a binary image..

# Histogram processing



In a high-contrast image, the grayscale values tend to cluster at the extremes, exhibits a stretched histogram

# Histogram Equalization



$T$  ?

$p_R(r) \Rightarrow T \Rightarrow \text{uniform } p_S(s)$

# Histogram process equalization

- Biểu đồ của ảnh có mức xám trong khoảng  $[0, L - 1]$  là hàm rời rạc có dạng như sau:
- $p(r_k) = \frac{n_k}{n}$  với  $\begin{cases} n_k \text{ là số lượng pixel có giá trị xám } r_k \\ n \text{ là tổng số pixel của ảnh} \end{cases}$
- Hàm  $p(r_k)$  mô tả tỷ lệ giữa tổng số pixel có mức xám  $r_k$  và tổng số pixel của ảnh
- Histogram mô tả tổng thể phân bố mức xám của ảnh

# Algorithm histogram equalization

$r_k$ : mức xám đầu vào

$s_k$ : mức xám đầu ra

$k$ : dải cường độ sáng

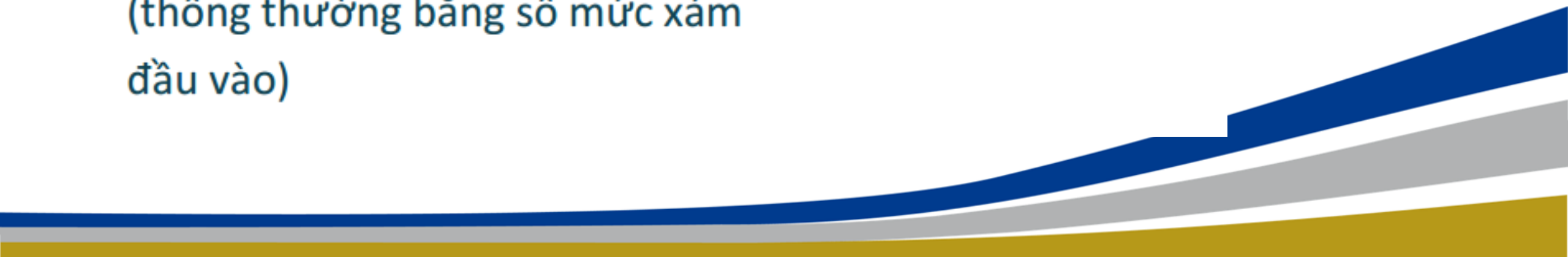
$n_j$ : tần suất của cường độ  $j$

$n$ : tổng số điểm ảnh

$L'$ : số mức xám ở ảnh đầu ra

(thông thường bằng số mức xám đầu vào)

$$\begin{aligned} s_k &= T(r_k) \\ &= (L'-1) \sum_{j=0}^k p_r(r_j) \\ &= (L'-1) \sum_{j=0}^k \frac{n_j}{n} \end{aligned}$$



# Example

Gray Levels ( $r_k$ )	No. of pixels ( $n_k$ )	$P(r_k) = n_k/n$ (PDF)	$S_k$ (CDF)	$S_k \times 7$	Histogram Equalization level
0	9	0.141	0.141	0.987	1
1	8	0.125	0.266	1.862	2
2	11	0.172	0.438	3.066	3
3	4	0.0625	0.5005	3.5035	4
4	10	0.156	0.6565	4.5955	5
5	15	0.234	0.8905	6.2335	6
6	4	0.0625	0.953	6.671	7
7	3	0.047	1	7	7
<u><math>n = 64</math></u>					

Gray levels	1	2	3	4	5	6	7
No. of pixels	9	8	11	4	10	15	7

# Example

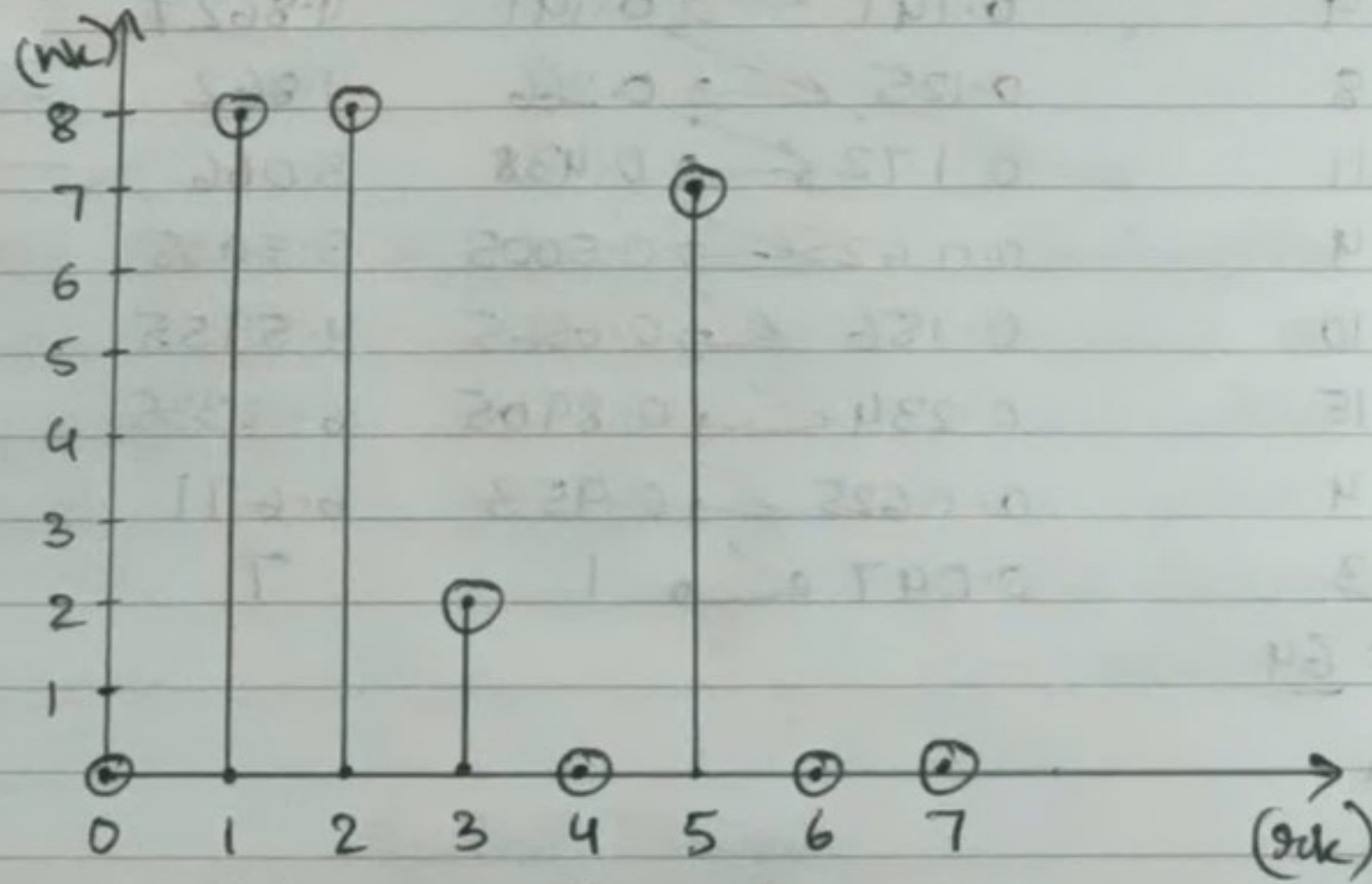
1	2	1	1	1
2	5	3	<u>5</u>	2
2	5	5	5	2
2	5	3	5	2
1	1	1	2	1

Input image



## Example

Gray levels (rk)	0	1	2	3	4	5	6	7
No. of pixels (nk)	0	8	8	2	0	7	0	0



Histogram of input image

# Example

Gray Levels( $r_k$ )	No. of pixels( $n_k$ )	$P(r_k) = n_k/n$ (PDF)	$S_k$ (CDF)	$S_k \times T$	Histogram Equalization level
0	0	0	0	0	0
1	8	0.32	0.32	2.24	2
2	8	0.32	0.64	4.48	4
3	2	0.08	0.72	5.04	5
4	0	0	0.72	5.04	5
5	7	0.28	1	7	7
6	0	0	1	7	7
7	0	0	1	7	7

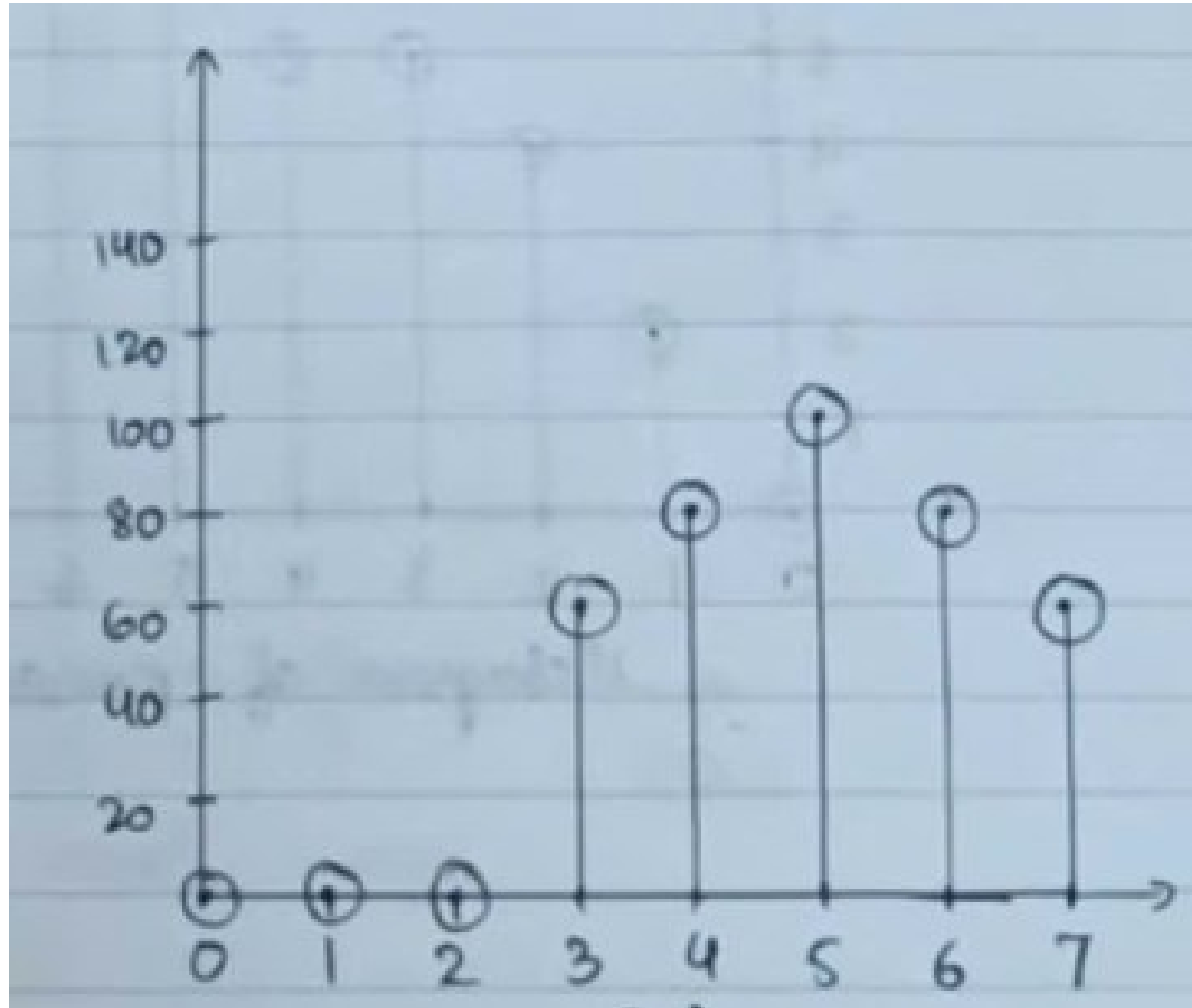
# Example

1	2	1	1	1
2	5	3	<u>5</u>	2
2	5	5	5	2
2	5	3	5	2
1	1	1	2	1

Input image

Output image?

# Example



Histogram of Output image

# Example

Gray level	nr
0	80
1	100
2	90
3	60
4	30
5	20
6	10
7	0



Gray level	$n_k$	$p(g_k) = n_k/n$ (PDF)	$S_k$ (CDF)	$S_k \times 7$	Histogram equalization	New
0	80	0.20	0.20	1.4	1	80
1	100	0.25	0.45	3.15	3	100
2	90	0.23	0.68	4.76	5	90
3	60	0.15	0.83	5.81	6	60+30 90
4	30	0.07	0.9	6.3	6	
5	20	0.05	0.95	6.65	7	20+10 30
6	10	0.02	0.97	6.79	7	
7	0	0	0.97	6.79	7	
	<u><math>n = 390</math></u>					

- Given the image  $I$ , draw the grayscale histogram and perform histogram equalization. Find the image  $I'$  after the histogram equalization.

$$I = \begin{pmatrix} 1 & 2 & 0 & 4 \\ 1 & 0 & 0 & 7 \\ 2 & 2 & 1 & 0 \\ 4 & 1 & 2 & 1 \\ 2 & 0 & 1 & 1 \end{pmatrix}$$