# 영상처리

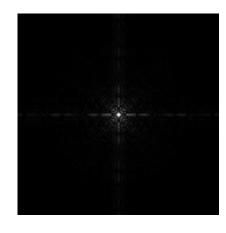
- Intensity Transformations and Spatial Filtering -

# Background

# The Basics of Intensity Transformations and Spatial Filtering (밝기값 변환과 공간 필터링의 기초)

Spatial domain vs Frequency domain





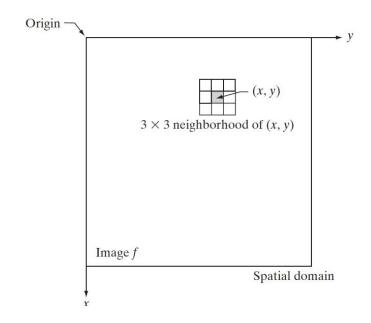
- 이번 chapter 에서는 spatial domain technique을 다룸
  - Image 자체에 가하는 processing
- Frequency domain → Fourier transform image에 processing

### The Basics of Intensity Transformations and Spatial Filtering

■ Spatial domain processes (공간도메인처리)

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

• g: processed image, f: input image, T: operator defined over a neighborhood of point (x, y)

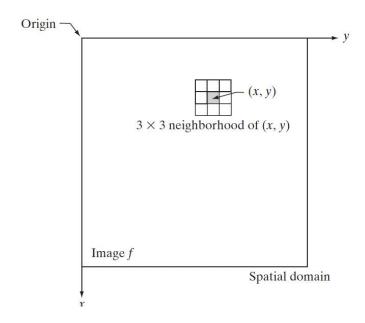


3x3 mask를 사용할 때 spatial domain operation (filtering)의 예시 → x, y의 intensity를 얻기 위해 주면 3x3의 값을 이 용

# Background

### The Basics of Intensity Transformations and Spatial Filtering

- Top left → bottom right (행렬)
- 경계는?



# Background

### The Basics of Intensity Transformations and Spatial Filtering

- Smallest possible neighborhood: 1x1
  - 이때, 결과 g 는 단일 점 (x, y) 에 있는 밝기값(intensity)과 변환 T에만 영향을 받음

$$s = T(r)$$

- s: intensity of g(x,y), r: intensity of f(x,y)
- intensity transformation function (also called gray-level or mapping function)

### The Basics of Intensity Transformations and Spatial Filtering

Smallest possible neighborhood: 1x1

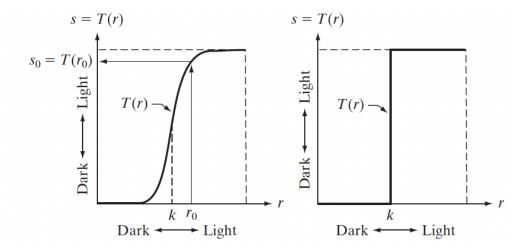
$$s = T(r)$$

a b

#### FIGURE 3.2

Intensity transformation functions. (a) Contraststretching function.

(b) Thresholding function.

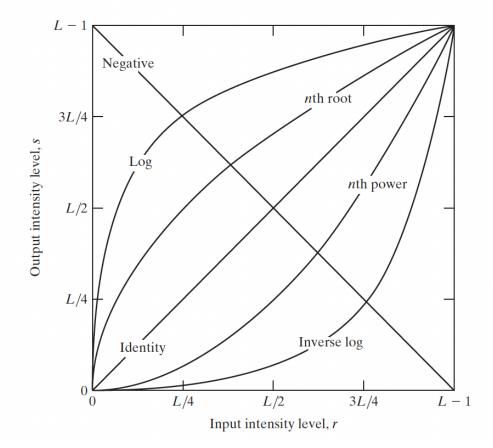


**FIGURE 3.3** Some basic intensity transformation functions. All curves were scaled to fit in the range shown.

Lookup table

X-axis: input

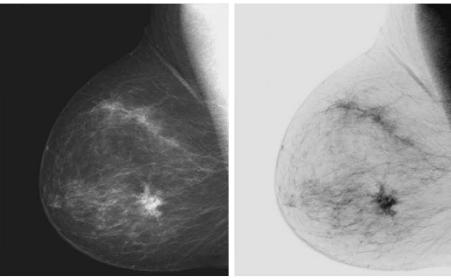
Y-axis: output



### **Image negatives**

■ Image with intensity levels in the range [0, L – 1]

$$s = L - 1 - r$$



a b

FIGURE 3.4

(a) Original digital mammogram.

(b) Negative image obtained using the negative transformation in Eq. (3.2-1).

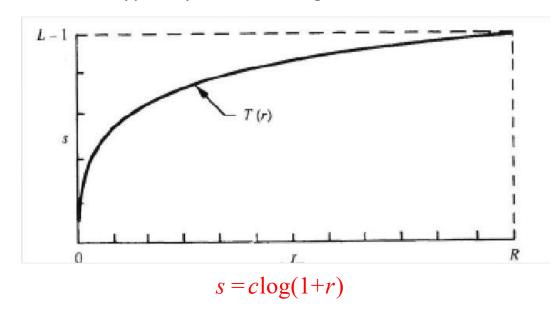
(Courtesy of G.E. Medical Systems.)

black area is dominant in size

Suitable for enhancing white or gray detail embedded in dark regions

### Log transformations

- When the dynamic range of the input gray values is very large (ex: Fourier transform magnitude), compared to that of the display,
  - we need to compress the Gray value range
  - Compression of Dynamic Range
    - Typically, we use a log scale.



Low range -> expanded High range -> compressed

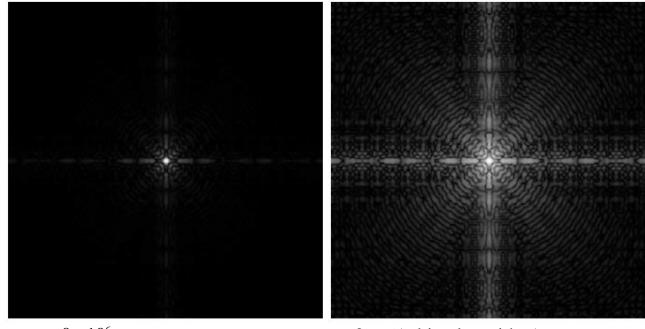
### Log transformations

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

a b

#### FIGURE 3.5

(a) Fourier
spectrum.
(b) Result of
applying the log
transformation in
Eq. (3.2-2) with
c = 1.

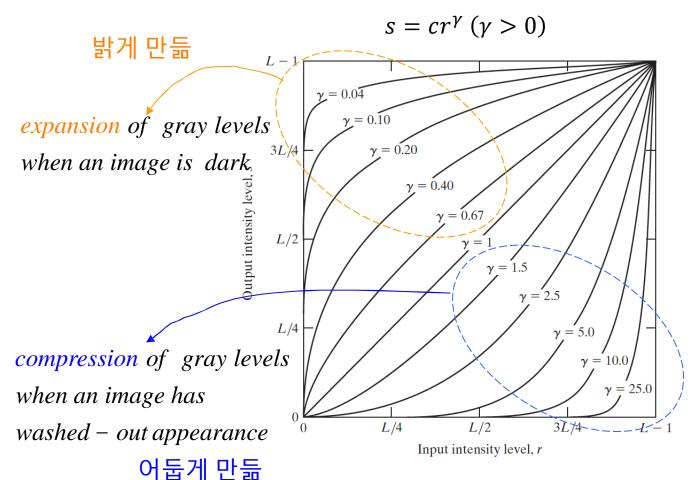


 $0 \sim 10^6$ 

0 ~ 6(taking logarithm)

 $\Rightarrow$  linearly scale down to 256 levels  $\Rightarrow$  scale up to 256 levels

### **Power-Law (Gamma) Transformations**



**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.

 $narrow \ range \ of \ dark \ input \ values \Rightarrow \ wider \ range \ of \ output \ value$  expansion

### **Power-Law (Gamma) Transformations**

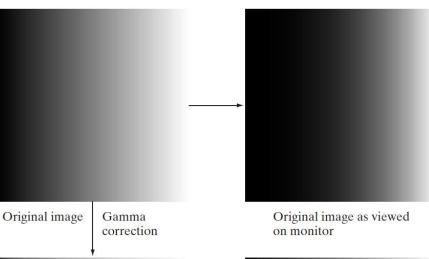
Example 1)

a b c d

#### FIGURE 3.7

(a) Intensity ramp image. (b) Image as viewed on a simulated monitor with a gamma of 2.5. (c) Gamma-corrected image. (d) Corrected image as viewed on the same monitor. Compare (d) and (a).

### Original image



Gamma-corrected image

Gamma correction으로 밝게 만든 이미지

Gamma-corrected image as

viewed on the same monitor Correction 후 모니터에 보이는 이미지

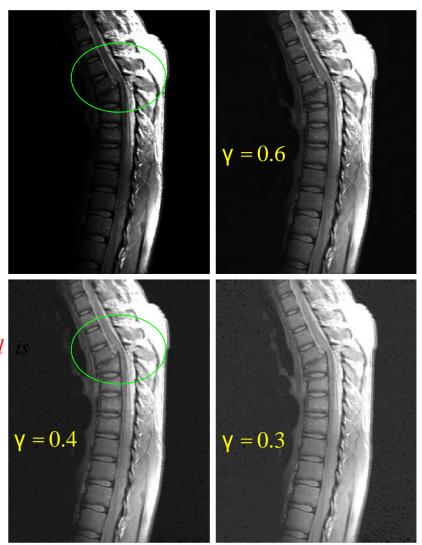
Gamma 2.5를 가진 모니터에서

보이는 이미지

**Power-Law (Gamma) Transformations** 

Example 2) Spine MRI

the best enhancement in terms of contrast and discernable detail achieved with  $\gamma=0.4$ 



a b c d

#### FIGURE 3.8 (a) Magnetic resonance image (MRI) of a fractured human spine. (b)-(d) Results of applying the transformation in Eq. (3.2-3) with c = 1 and $\gamma = 0.6, 0.4, \text{ and}$ 0.3, respectively. (Original image courtesy of Dr. David R. Pickens, Department of Radiology and Radiological Sciences.

Vanderbilt University

Medical Center.)

### **Power-Law (Gamma) Transformations**

Example 3) Aerial image

Washed-out, so the compression of gray levels is required

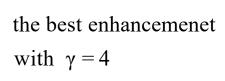
a b c d

#### FIGURE 3.9

(a) Aerial image. (b)–(d) Results of applying the transformation in Eq. (3.2-3) with c = 1 and  $\gamma = 3.0, 4.0$ , and 5.0, respectively. (Original image for this example courtesy of NASA.)











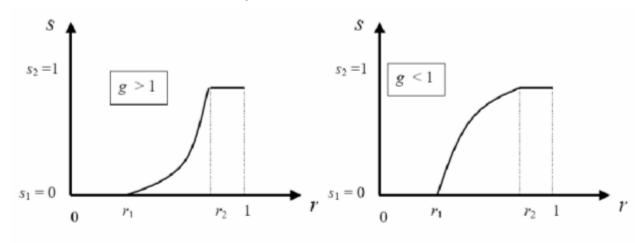
Too dark

### **Power-Law (Gamma) Transformations**

General form

$$s_1 = 0, s_2 = 1, and$$

$$T(r) = \begin{cases} 0, & r < r_1 \\ \left(\frac{r - r_1}{r_2 - r_1}\right)^{\gamma} & r_1 \le r \le r_2 \\ 1, & r > r_2 \end{cases}$$



Output Image is "darker"

Output Image is "brighter"

실습

DLIP\_practice1.IntensityTransformation.ipynb

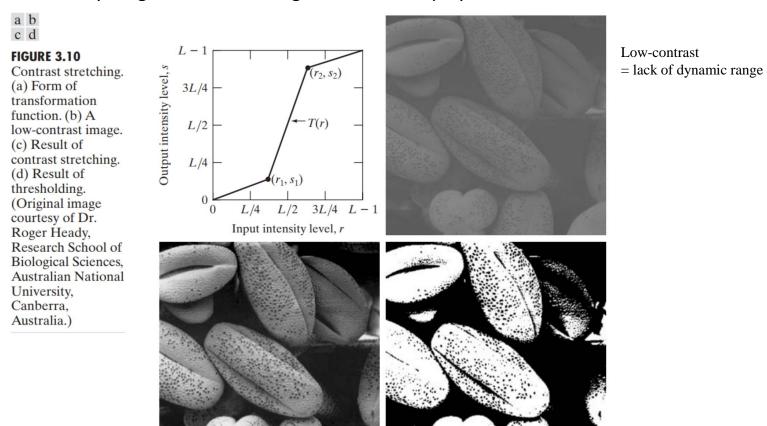
#### **Piecewise-Linear Transformation Functions**

- Intensity 구간별로 서로 다른 linear transformation function 적용
  - Advantage: The form of functions can be arbitrary complex
  - Disadvantage: Specification requires considerably more user input

#### **Piecewise-Linear Transformation Functions**

### Contrast stretching

 process that expands the range of intensity levels in an image so that it spans the full intensity range of the recording medium or display device

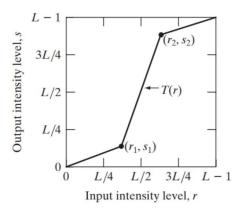


Full range linear stretching

Thresholding

#### **Piecewise-Linear Transformation Functions**

- Contrast stretching
  - Increase the dynamic range of gray values in the input image.
  - Suppose you are interested in stretching the input intensity values in the interval [r1, r2]:

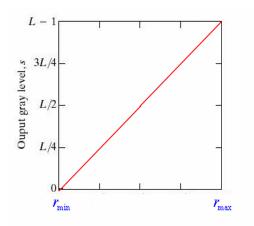


- Note that (r1- r2) < (s1- s2).</li>
- The gray values in the range [r1, r2] is stretched into the range [s1, s2].

#### **Piecewise-Linear Transformation Functions**

### Contrast stretching

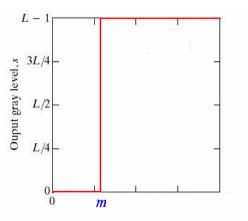
### full range linear stretching



 $r_{\min}$  : minimum gray level in the image

 $r_{\text{max}}$ : maximum gray level in the image

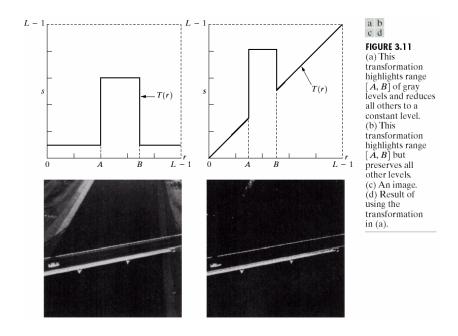
### Thresholding function



m = mean[graylevel in the image]

#### **Piecewise-Linear Transformation Functions**

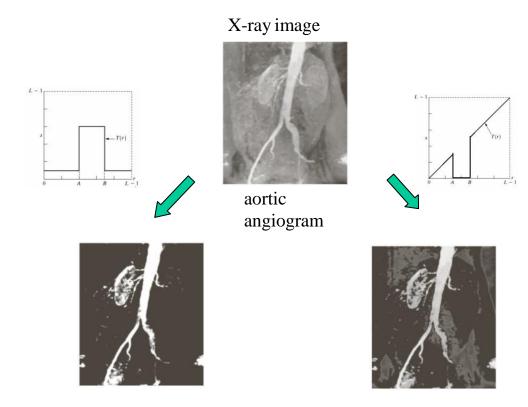
- Intensity-level slicing
  - Highlighting a specific range of gray levels



#### **Piecewise-Linear Transformation Functions**

Intensity-level slicing

How can we preserve kidney and blood vessel?



B/W representation

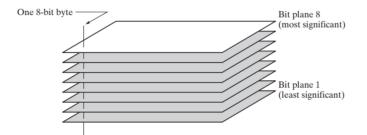
#### **Piecewise-Linear Transformation Functions**

- Bit-Plane slicing
  - mainly used for data compression and progressive transmission

 highlight the contribution made to total image appearance by specific bits

Pivels most significant bit = 1

FIGURE 3.13 Bit-plane representation of an 8-bit image.



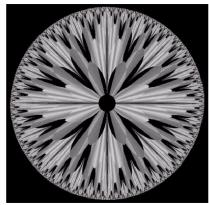
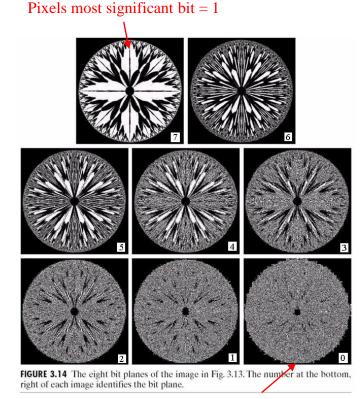


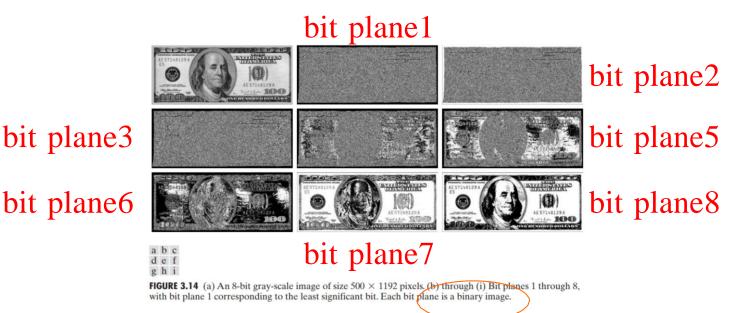
FIGURE 3.13 An 8-bit fractal image. (A fractal is an image generated from mathematical expressions). (Courtesy of Ms. Melissa D. Binde, Swarthmore College, Swarthmore, PA.)



Pixels least significant bit = 1 (detail or noise?)

#### **Piecewise-Linear Transformation Functions**

- Bit-Plane slicing
  - mainly used for data compression and progressive transmission



#### **Piecewise-Linear Transformation Functions**

- Bit-Plane slicing
  - mainly used for data compression and progressive transmission







a b c

**FIGURE 3.15** Images reconstructed using (a) bit planes 8 and 7; (b) bit planes 8, 7, and 6; and (c) bit planes 8, 7, 6, and 5. Compare (c) with Fig. 3.14(a).

실습

DLIP\_practice2.IntensityTransformation.ipynb