

MODULE2

- **Spatial Domain: Some Basic Intensity Transformation Functions,**
- Histogram Processing, Fundamentals of Spatial Filtering, Smoothing
- Spatial Filters, Sharpening Spatial Filters
- [Text: Chapter 3: Sections 3.2 to 3.6]

- **Frequency Domain: Preliminary Concepts, The Discrete Fourier**
- Transform (DFT) of Two Variables, Properties of the 2-D DFT, Filtering
- in the Frequency Domain, Image Smoothing and Image Sharpening
- Using Frequency Domain Filters, Selective Filtering.
- [Text: Chapter 4: Sections 4.2, 4.5 to 4.10]

Image Enhancement- Spatial Domain

Introduction

- What is image enhancement?
 - A process of enhancing the visual quality of images due to non ideal image acquisition process (e.g., motion blurring, out-of-focus, poor illumination, coarse quantization etc.)
- Image visual quality assessment
 - Objective quality metrics (e.g., MSE) might not always match subjective quality scores
 - Human vision system (HVS) is the ultimate JUDGE.

Principle Objective of Enhancement

- Process an image so that the result will be more suitable than the original image for a specific application.
- The suitability is up to each application.
- A method which is quite useful for enhancing an image may not necessarily be the best approach for enhancing another images

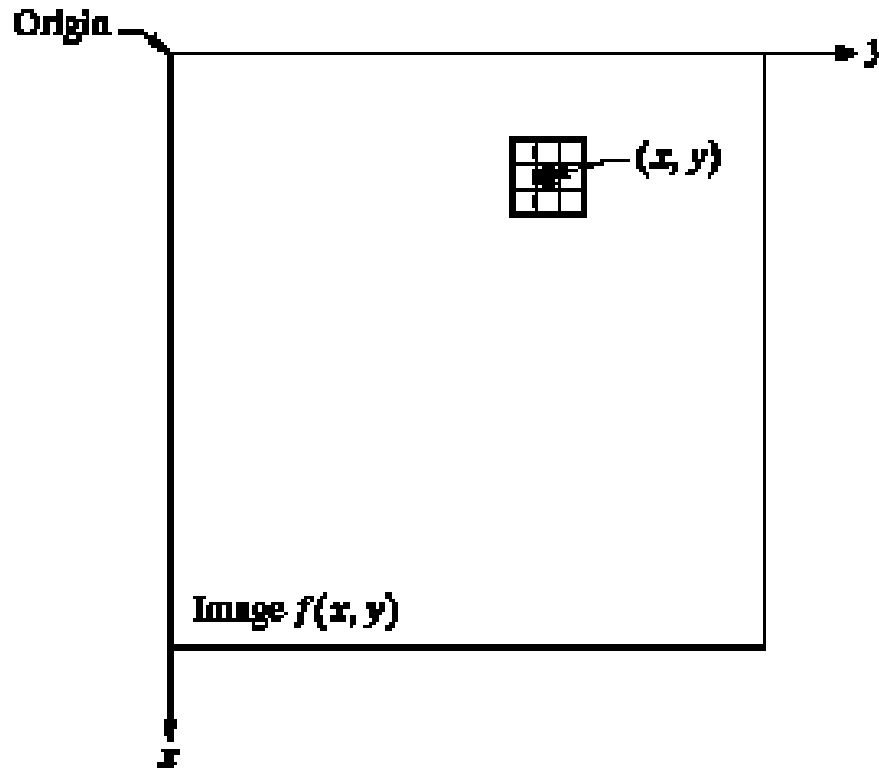
2 domains

- 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



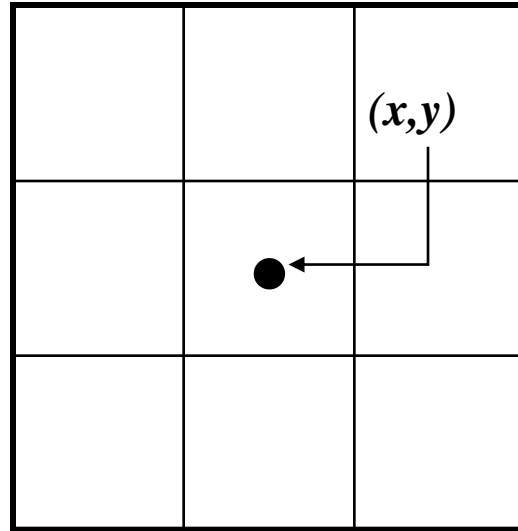
- Procedures that operate directly on pixels.

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

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

$$s = T(r)$$

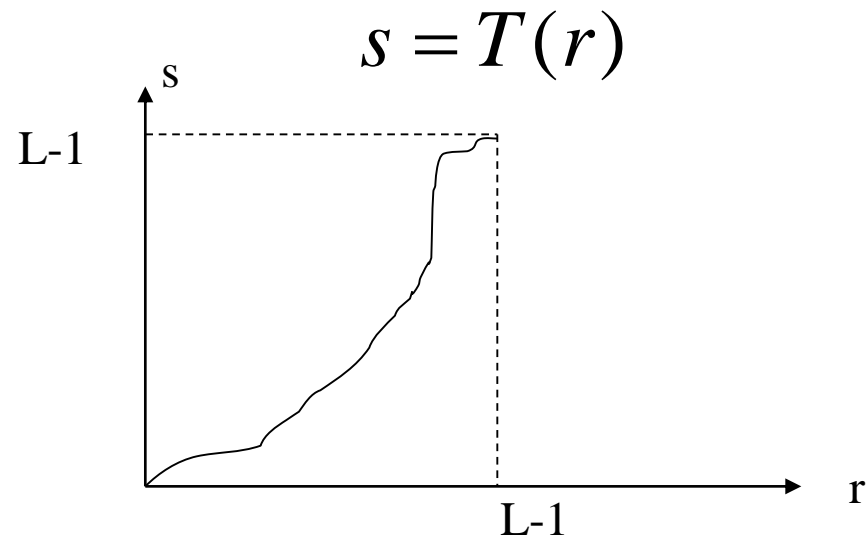
- Where
 - r = gray level of $f(x,y)$
 - s = gray level of $g(x,y)$

Enhancement by Point Processing

- These are methods based only on the intensity of single pixels.
 - r denotes the pixel intensity **before** processing.
 - s denotes the pixel intensity **after** processing.

Point Operations Overview

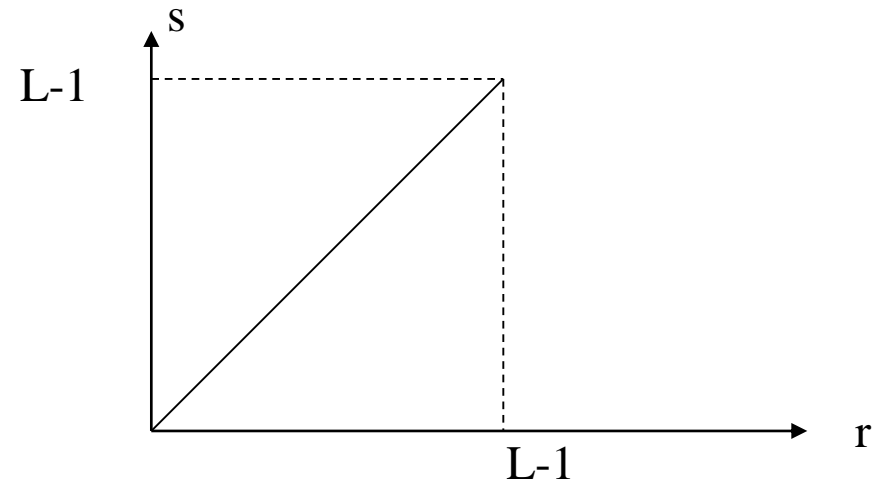
Point operations are **zero-memory** operations where a given gray level $r \in [0, L-1]$ is mapped to another gray level $s \in [0, L-1]$ according to a transformation



$L=256$: for grayscale images

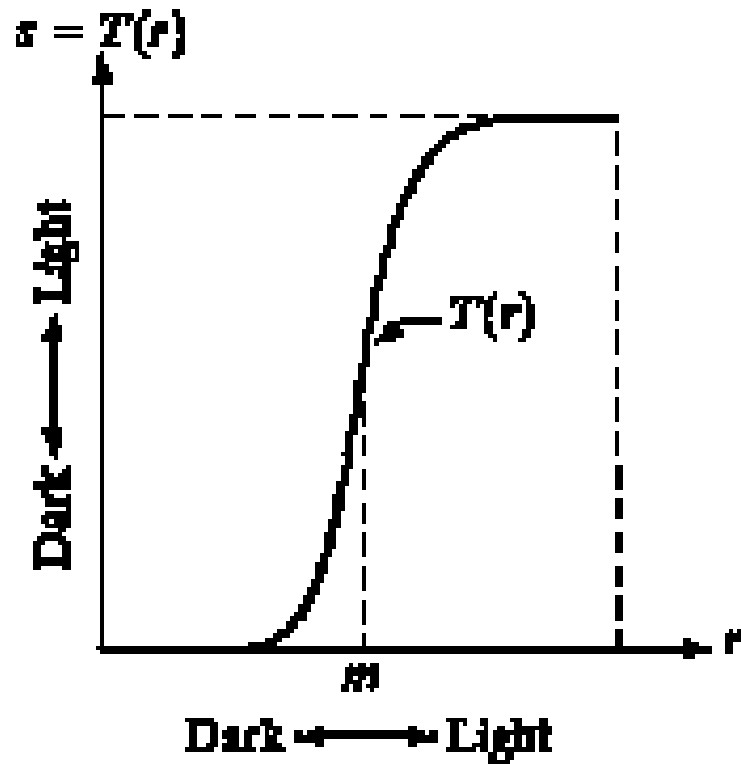
Lazy Man Operation

$$s = r$$



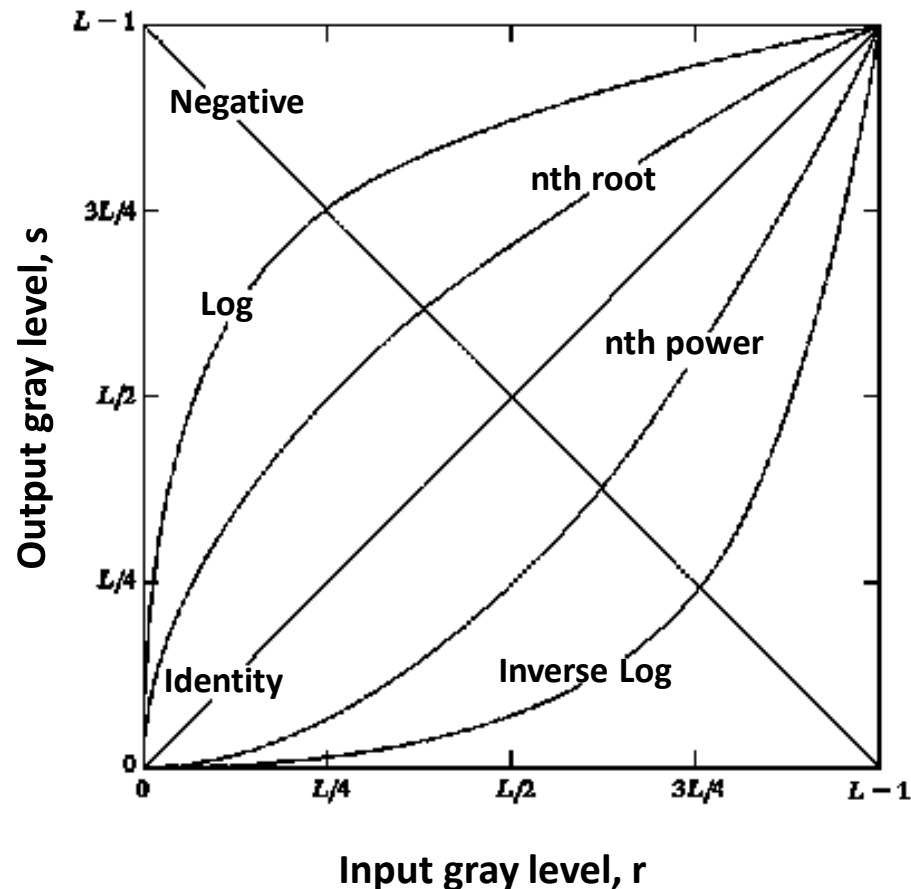
No influence on visual quality at all

Contrast Stretching



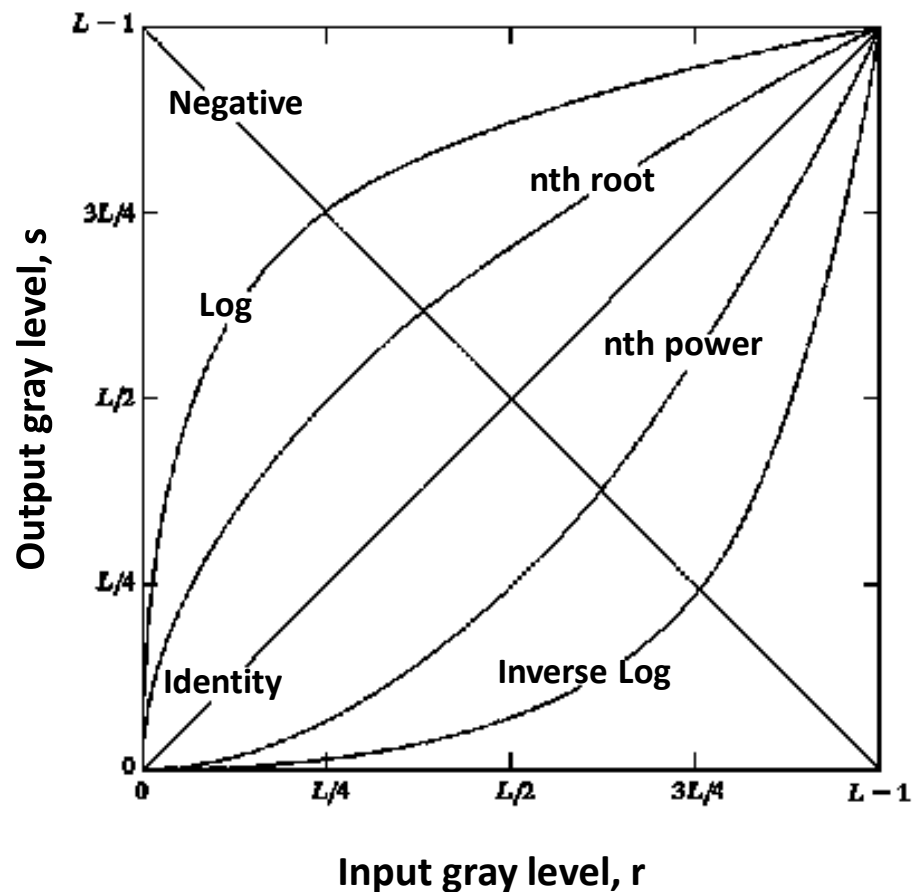
- Produce higher contrast than the original by
 - darkening the levels below m in the original image
 - Brightening the levels above m in the original image

3 basic gray-level transformation functions



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

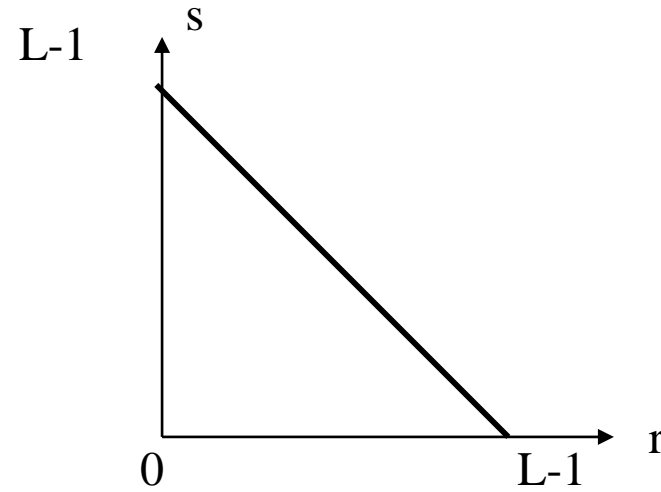
Image Negatives



- 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.

Digital Negative

$$s = L - 1 - r$$



Some Examples



$$I'(x, y) = I(x, y) + b$$

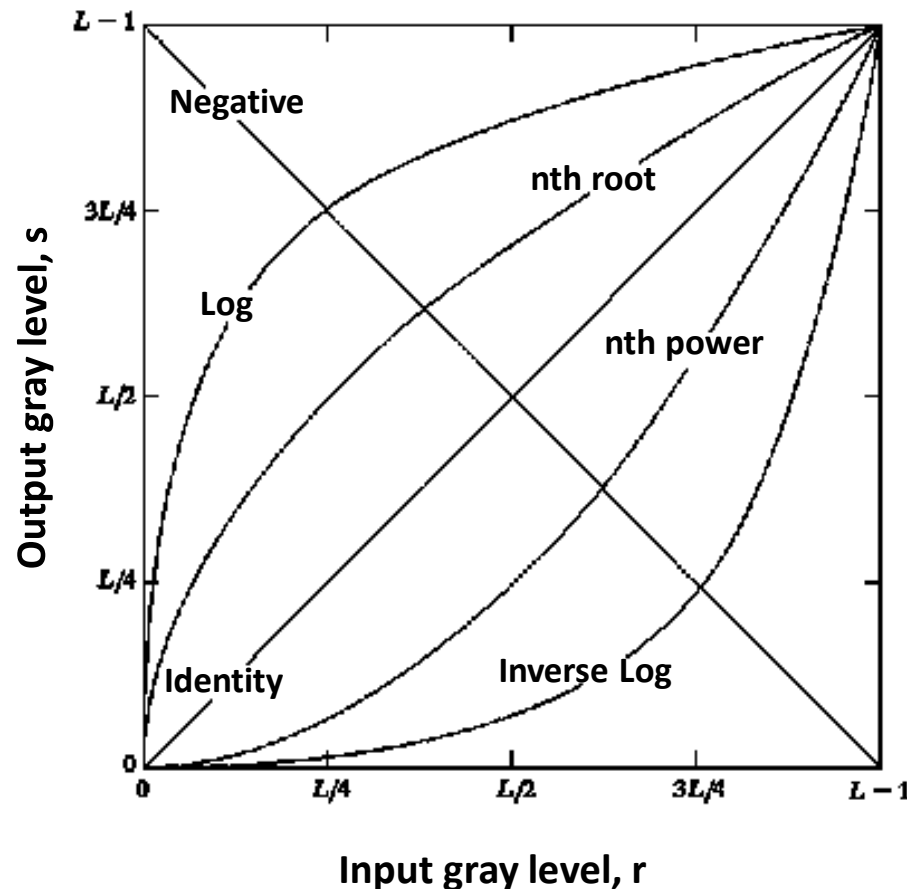


$$I'(x, y) = a \cdot I(x, y)$$

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Log Transformations

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

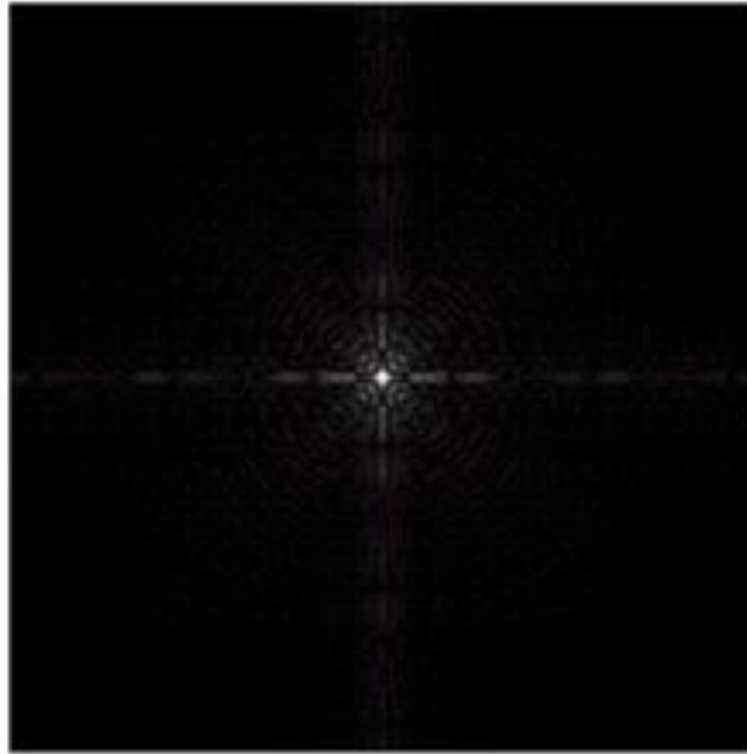


- c is a constant and $r \geq 0$
- Log curve maps a narrow range of low gray-level values in the input image into a wider range of output levels.
- Opposite is true of higher values of input values.
- Used to expand the values of dark pixels in an image while compressing the higher-level values.

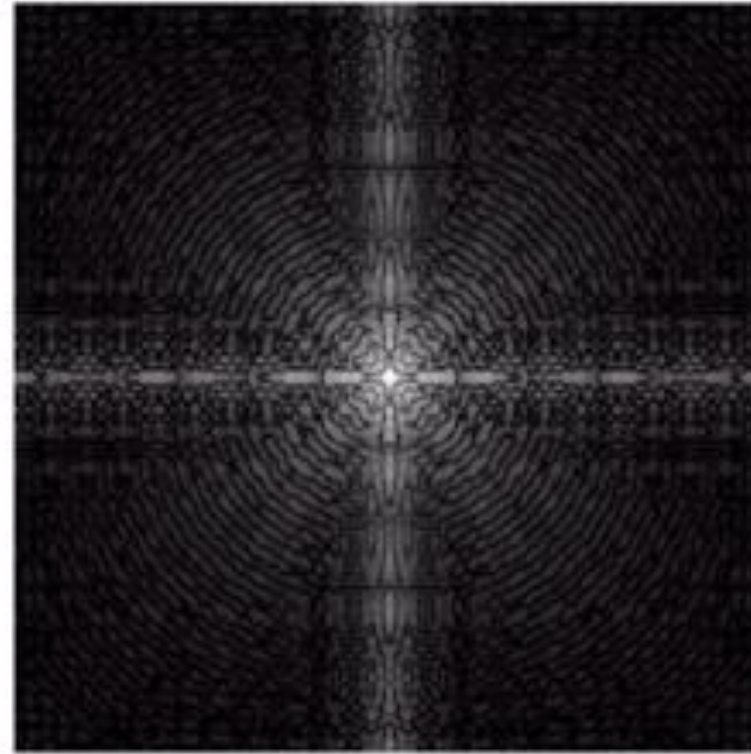
Log Transformations

- 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



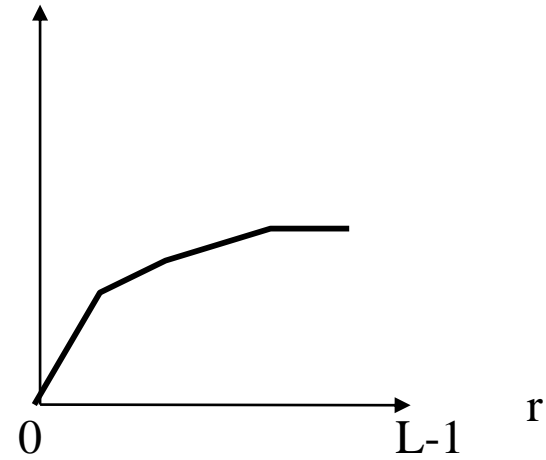
**Fourier Spectrum with range = 0
to 1.5×10^6**



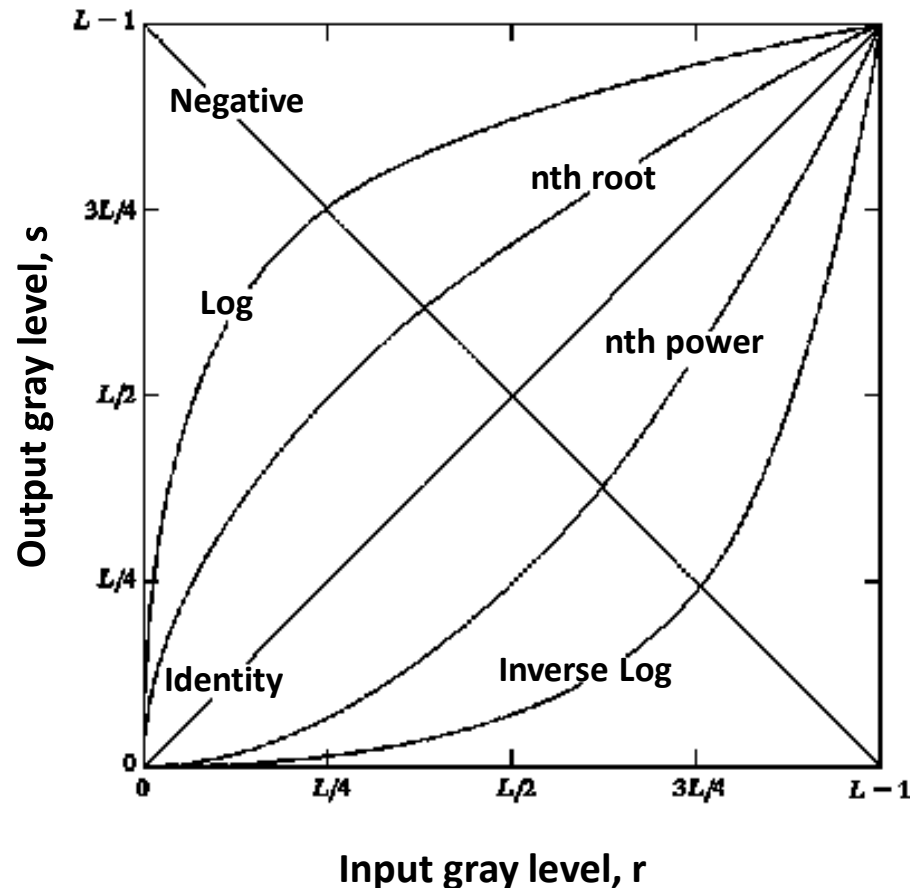
**Result after apply the log
transformation with $c = 1$, range =
0 to 6.2**

Range Compression

$$s = c \log_{10}(1 + r)$$



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, γ : positive constants

- Gamma correction

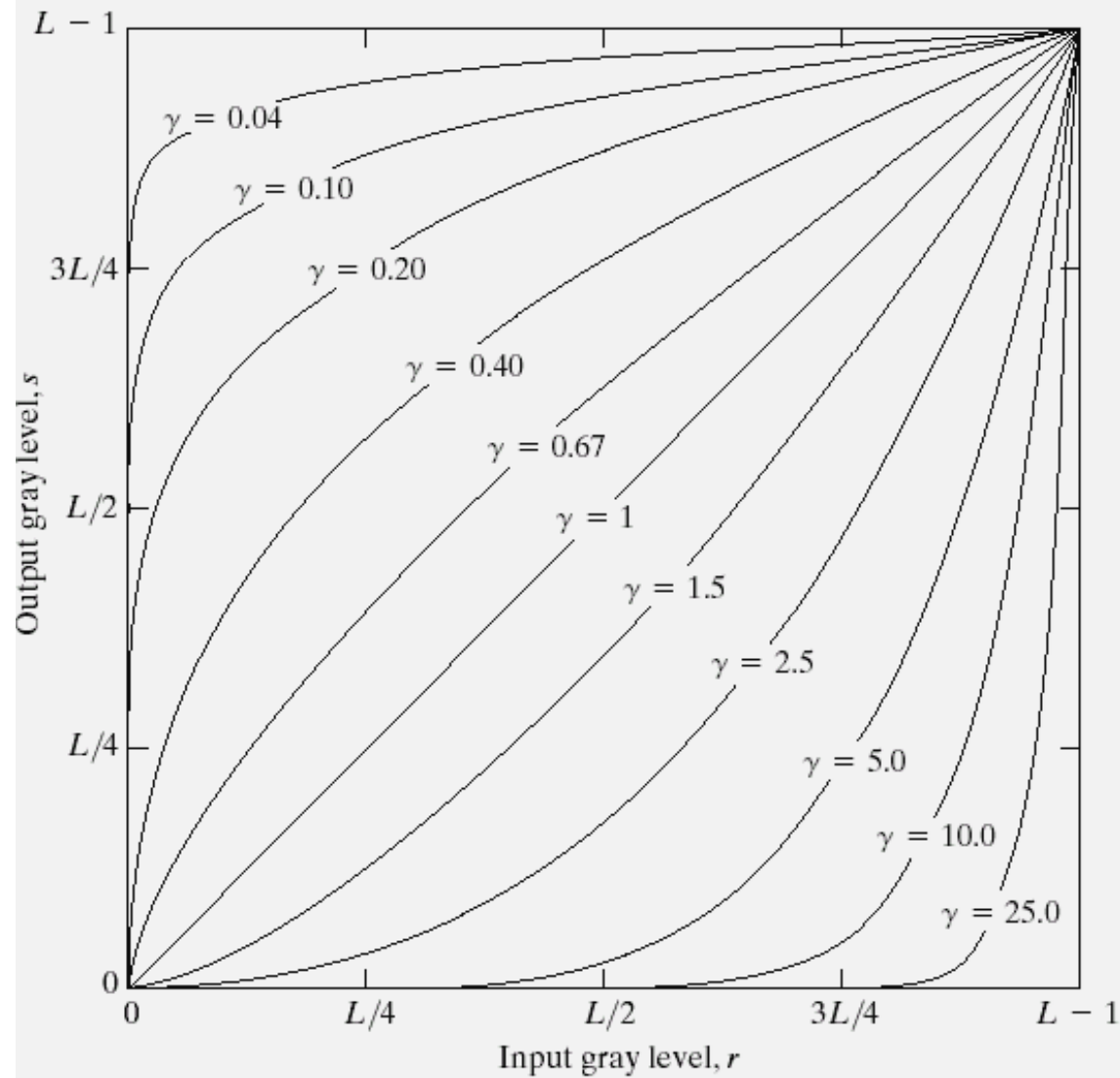


FIGURE 3.6 Plots of the equation $s = cr^\gamma$ for various values of γ ($c = 1$ in all cases).

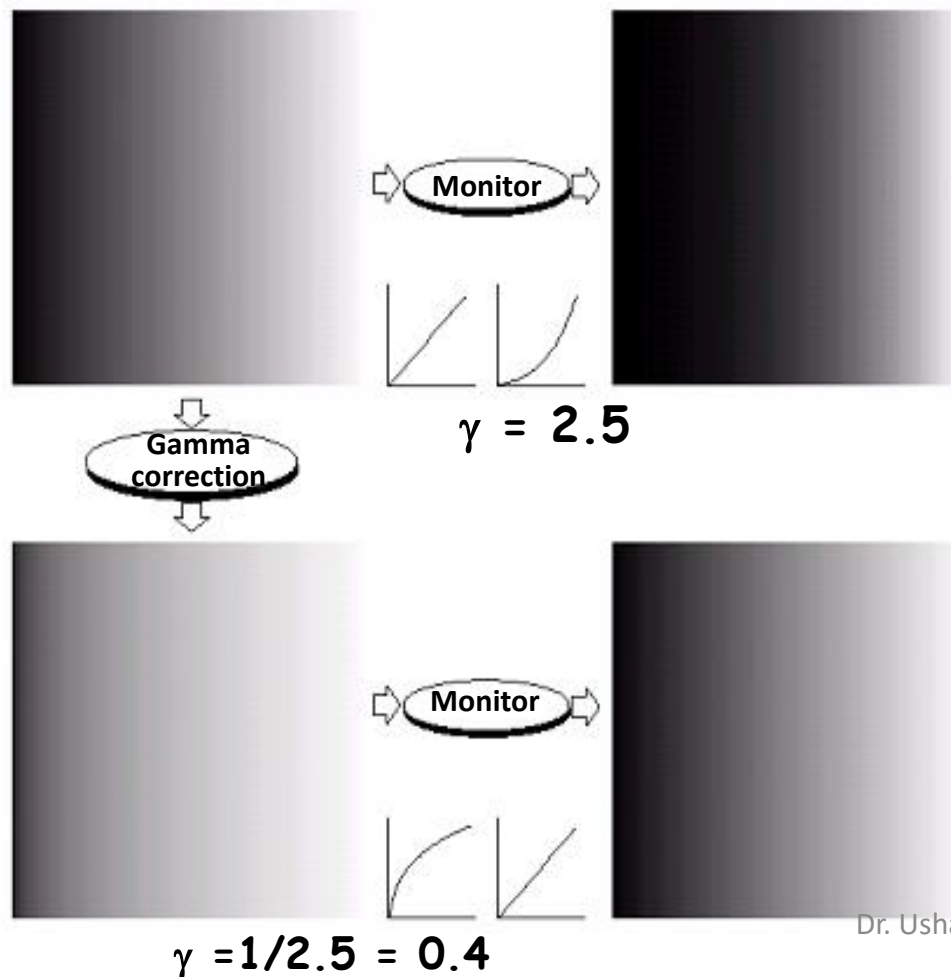
$\gamma=c=1$: identity

Power-Law Transformations

$$s = cr^\gamma$$

- c and γ are positive constants
- Power-law curves with fractional values of γ 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

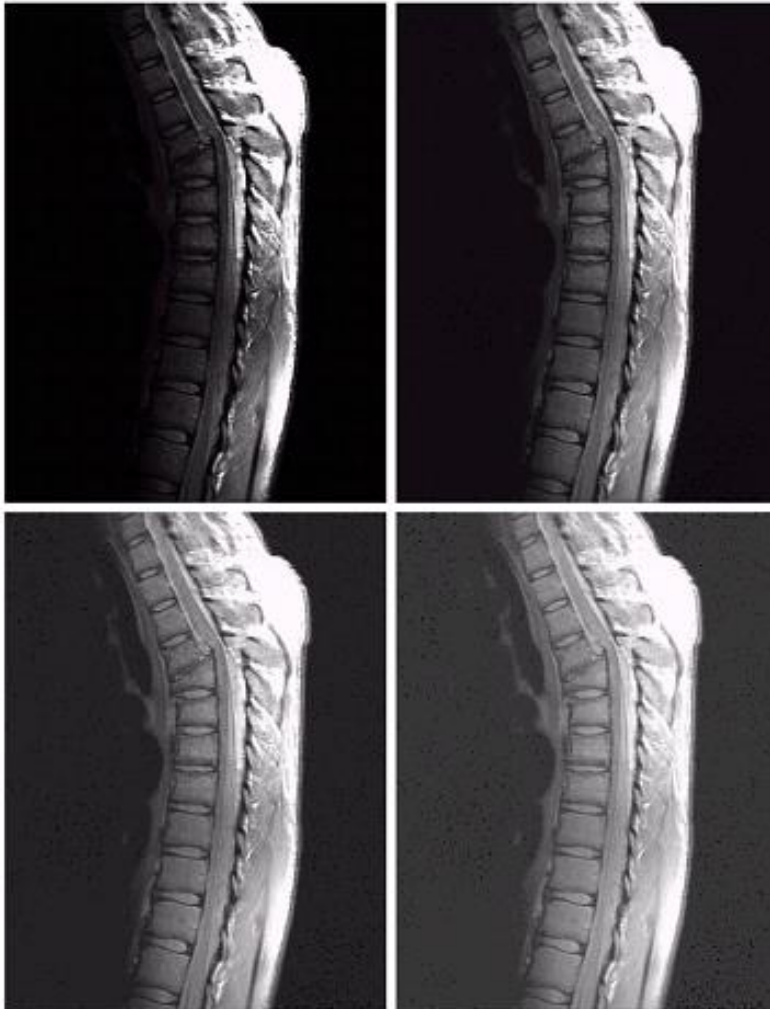
Gamma correction



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

a	b
c	d

Another example : MRI



(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 γ 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

a	b
c	d

Another example



(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)

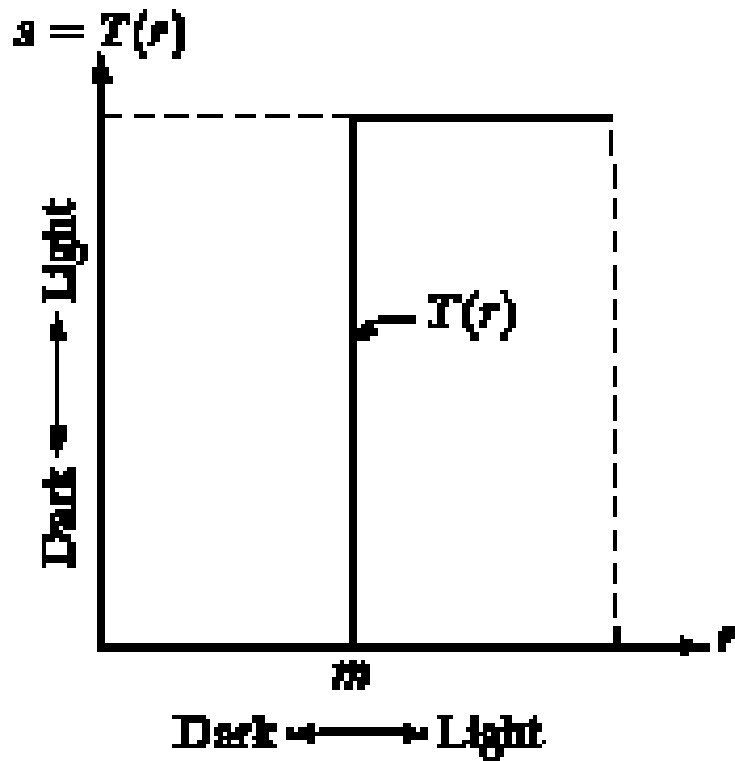
(d) transformation with $\gamma = 5.0$

(high contrast, the image has areas that are too dark, some detail is lost)

Some Simple Intensity Transformations

- Piecewise-Linear Transformation Functions:
 - Contrast stretching
 - Gray-level slicing
 - Bit-plane slicing

Thresholding

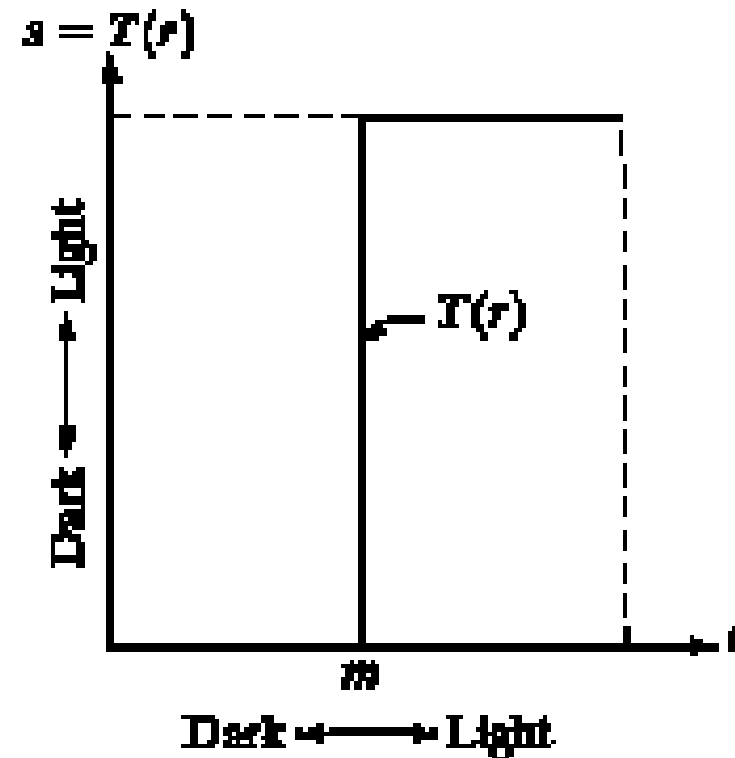


- Produces a two-level image

For the given sub-image, if $m=140$

Find the new image after thresholding

0	10	50	100
5	95	150	200
110	150	190	210
175	210	255	100



??



$$I'(x,y) = \begin{cases} 255 & \text{if } I(x,y) > 150 \\ 0 & \text{otherwise} \end{cases}$$

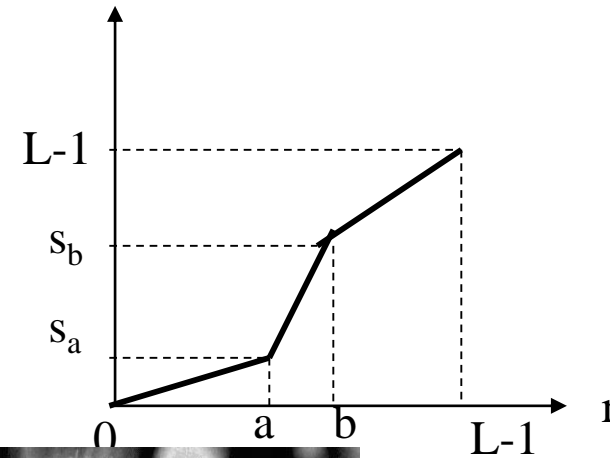
Piecewise-Linear Transformation Functions

- Advantage:
 - The form of piecewise functions can be arbitrarily complex
- Disadvantage:
 - Their specification requires considerably more user input

- Low contrast images can occur due to
 - poor or non uniform lighting conditions
 - due to non linearity in imaging sensors
 - due to small dynamic range of imaging sensors
- The slope of transformation is chosen greater than unity in the region of stretch.
- The parameters a & b are obtained by examining histogram

Contrast Stretching

$$s = \begin{cases} \alpha r & 0 \leq r < a \\ \beta(r - a) + s_a & a \leq r < b \\ \gamma(r - b) + s_b & b \leq r < L \end{cases}$$



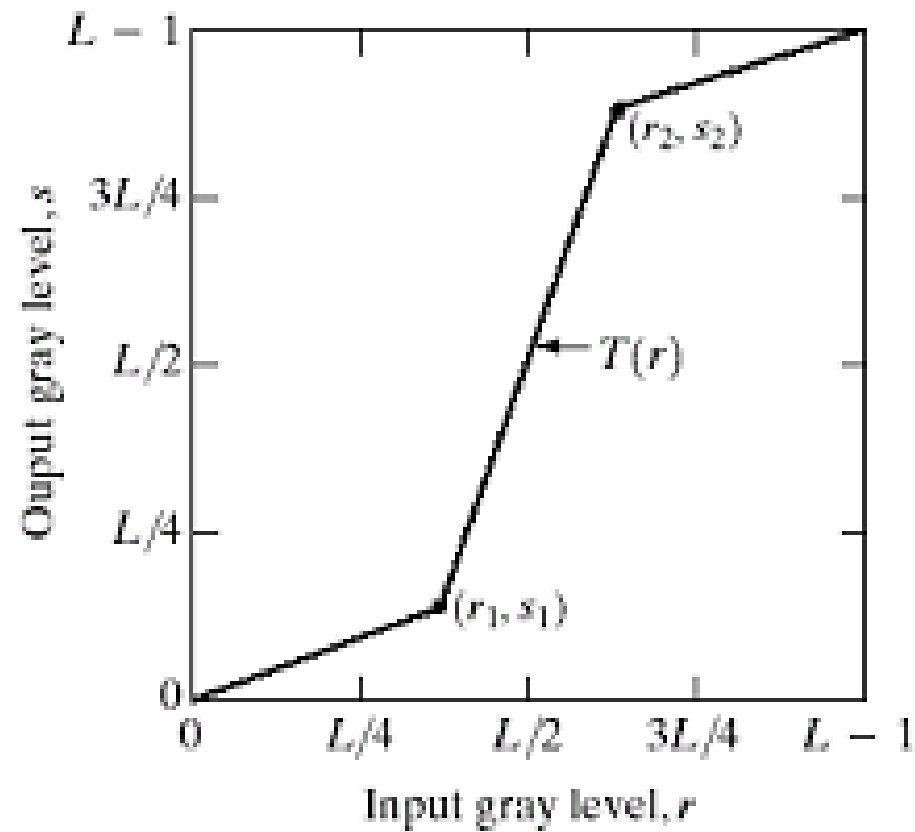
$$a = 50, b = 150, \alpha = 0.2, \beta = 2, \gamma = 1, s_a = 30, s_b = 200$$

For the given sub-image, if

$$a = 50, b = 150, \alpha = 0.2, \beta = 2, \gamma = 1, s_a = 30, s_b = 200$$

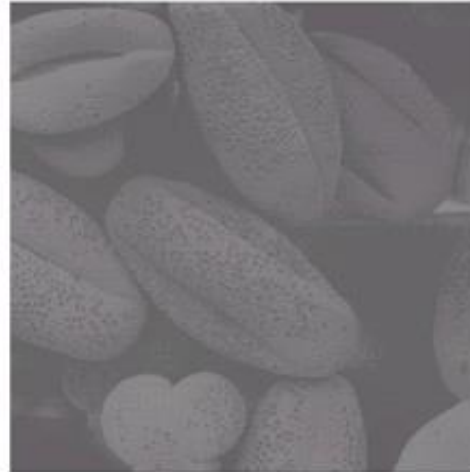
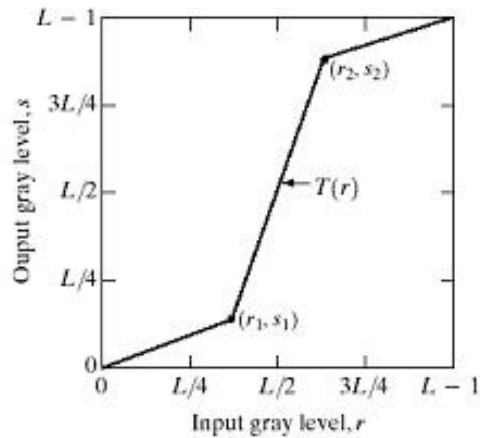
Find the new image after contrast stretching

0	10	50	100
5	95	150	200
110	150	190	210
175	210	255	100



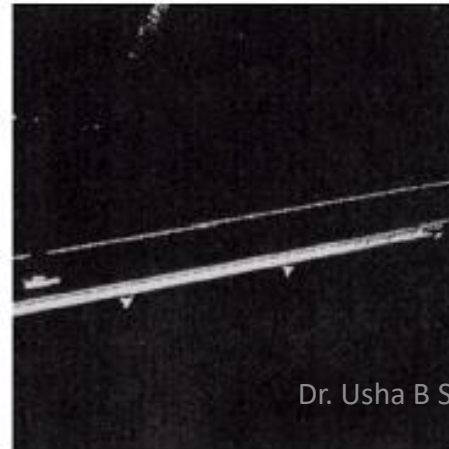
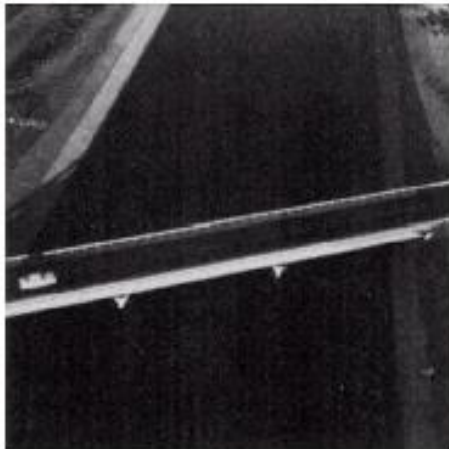
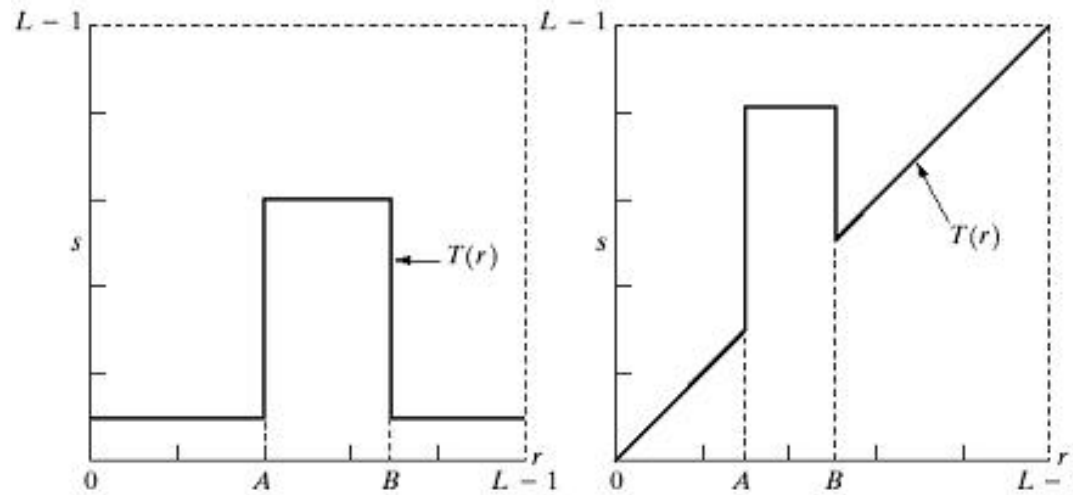
- If $s_1=r_1, s_2=r_2$
- no change in the gray level values
- If $r_1=r_2, s_1=0, s_2=L-1$
- it is Thresholding function
- $r_1 \leq r_2, s_1 \leq s_2$

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

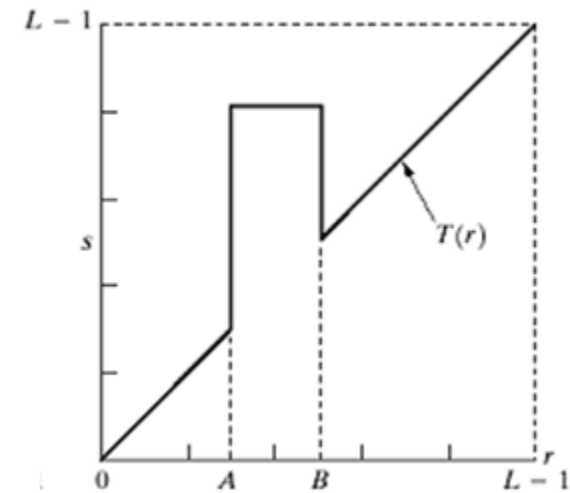


- Highlighting a specific range of gray levels in an image
 - Display a high value of all gray levels in the range of interest and a low value for all other gray levels
- (a) transformation highlights range $[A, B]$ of gray level and reduces all others to a constant level
- (b) transformation highlights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

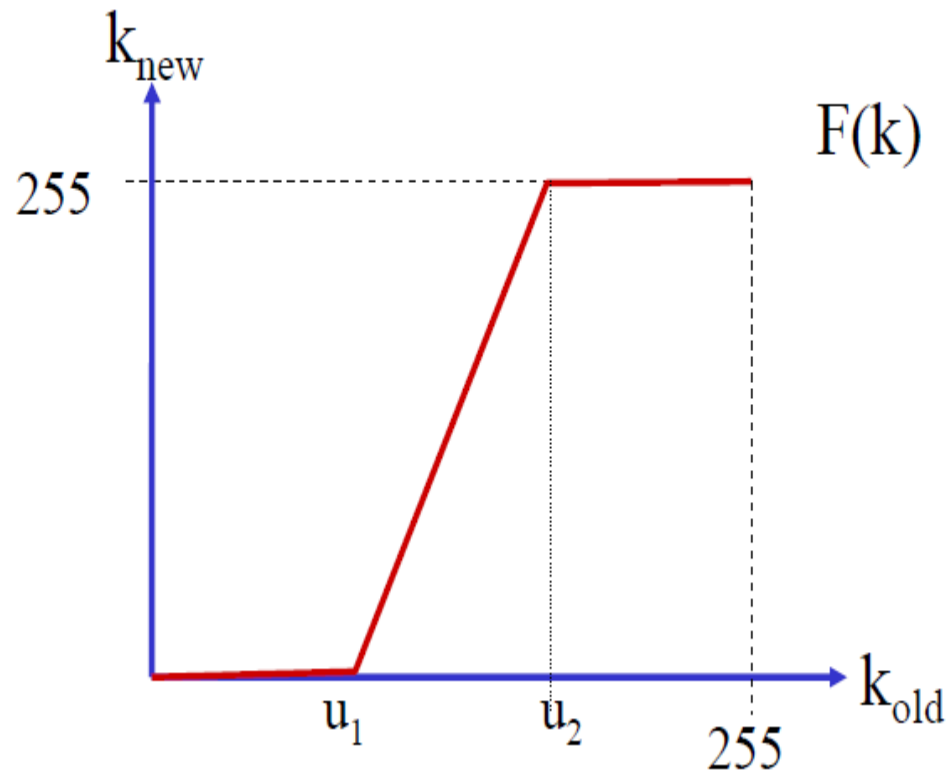
For the given sub-image, if $A=50$ and $B=170$, $M=200$

Find the new image after gray level slicing

0	10	50	100
5	95	150	200
110	150	190	210
175	210	255	100



Clipping Function

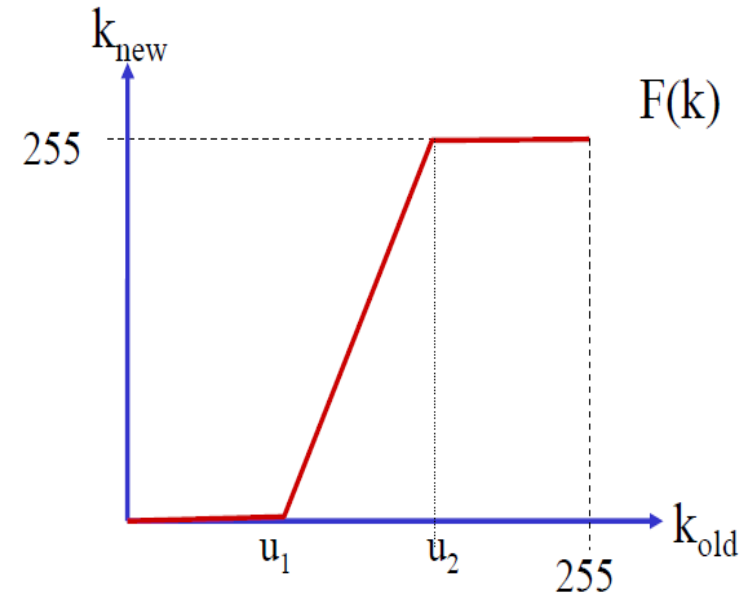


If most of the gray-levels in the image are in $[u_1 \ u_2]$, the following mapping increases the image contrast.

For the given sub-image, if $u_1=50$ and $u_2=170$, $\beta=2$

Find the new image after gray level slicing

0	10	50	100
5	95	150	200
110	150	190	210
175	210	255	100



Clipping Function



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