

20MCA24C – DIGITAL IMAGE PROCESSING

UNIT II

IMAGE ENHANCEMENT IN THE SPATIAL DOMAIN

FACULTY

Dr. K. ARTHI MCA, M.Phil., Ph.D.,

Assistant Professor,

Postgraduate Department of Computer Applications,

Government Arts College (Autonomous),

Coimbatore-641018.

Basic Gray level Transformations

- In a digital image, point = pixel
- Point Processing Operations –

Intensity transformation

–Histogram equalization

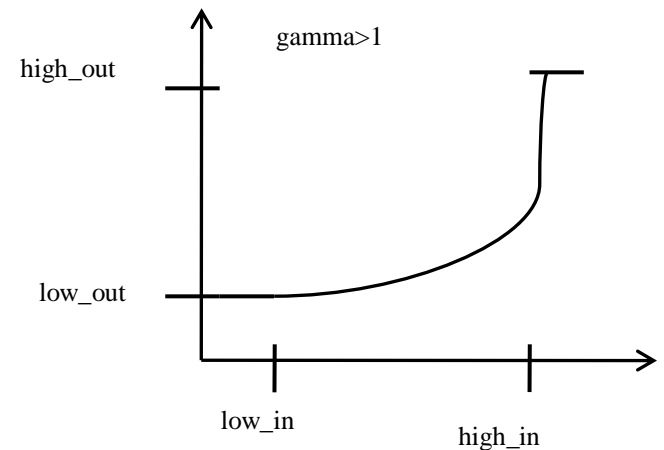
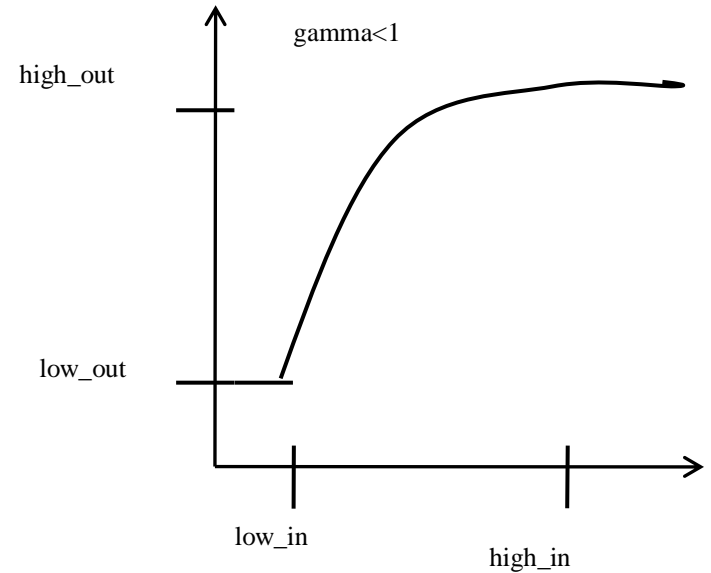
–Spatial filtering

Intensity Transformation Functions

- $s=T(r)$, where r denotes the intensity of f and s is the intensity of g , both at any (x, y) in the image

- **imadjust**

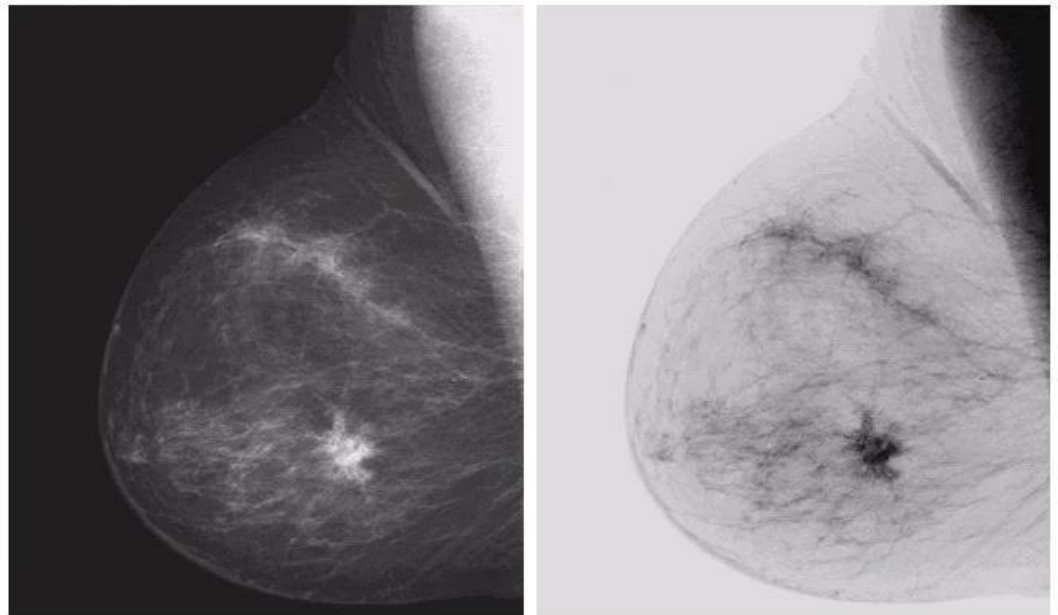
- $g=\text{imadjust}(f, [\text{low_in high_in}], [\text{low_out high_out}], \text{gamma})$



- Values between `low_in` and `high_in` is mapped to values between `low_out` and `high_out`

Image Negatives

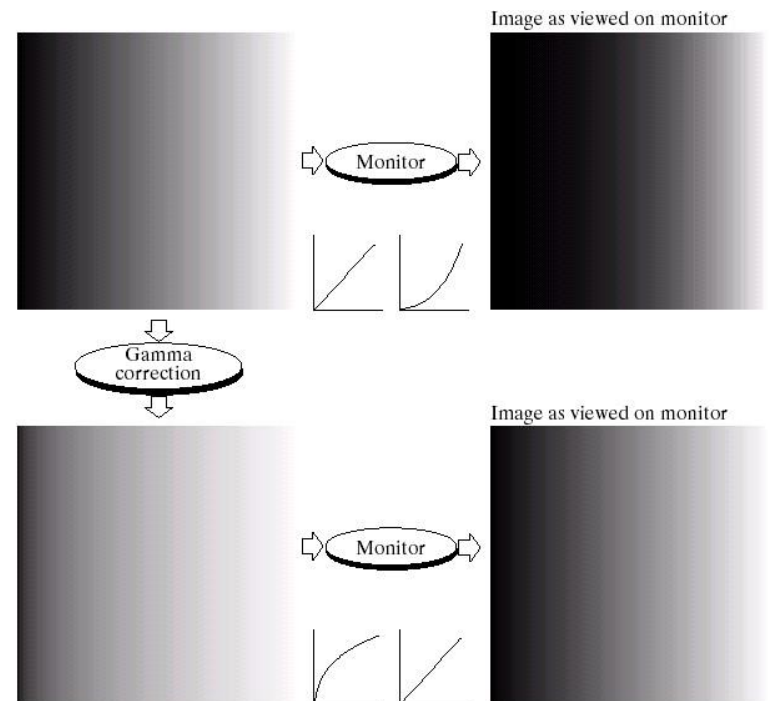
- $s = T(r) = L - 1 - r$
- Similar to photo negatives.
- Suitable for enhancing white or gray details in dark background.

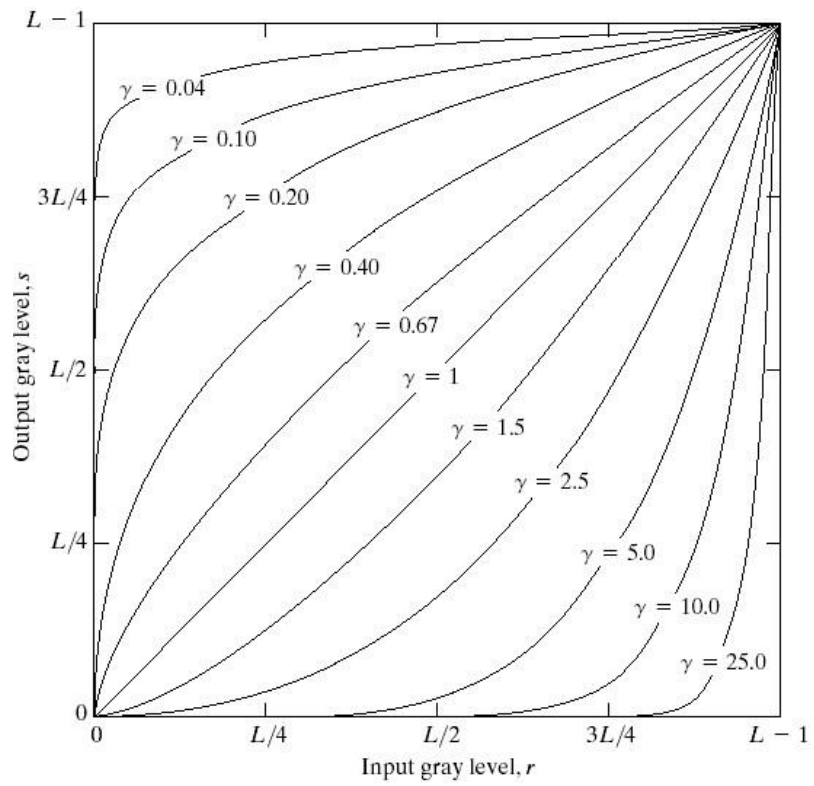


Power Law Gray-level Transform

$$s = T(r) = cr^g$$

- Gamma correction: to compensate the built-in power law compression due to display





characteristics.

Contrast Enhancement

- Piecewise linear Transformation
- Input : Poor illumination images
- Lack of dynamic range
- Increase the dynamic range of gray levels
- In raw imagery, data occupies only a small portion of the available range of digital values (commonly 8 bits or 256 levels).
- Contrast enhancement involves changing the original values so that more of the available range is used,
- Increases the contrast between targets and their backgrounds.

Other Piece-wise Transformation

□ Gray level Slicing □

Bit plane slice

Bit-Slicing

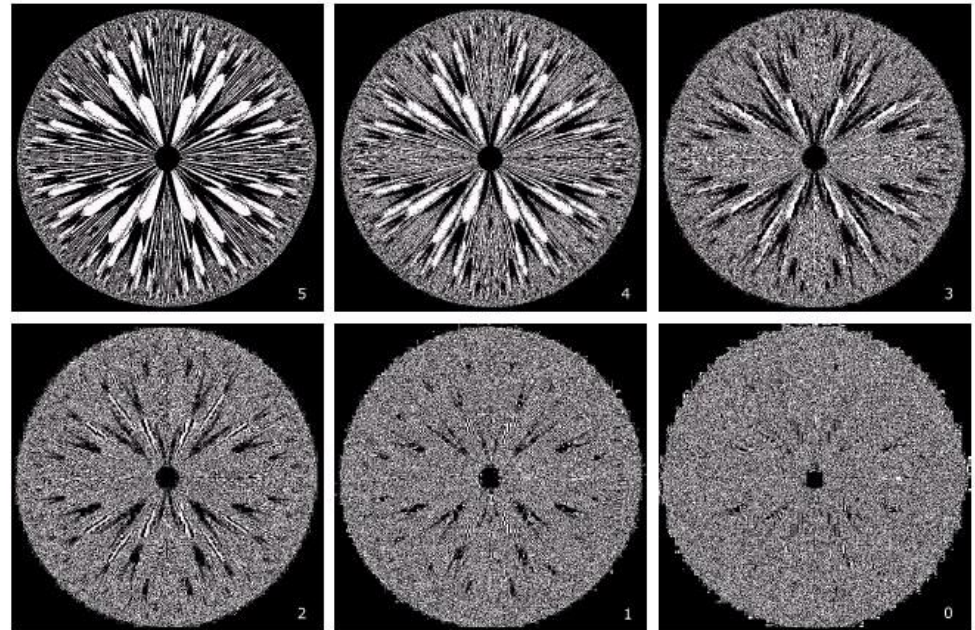
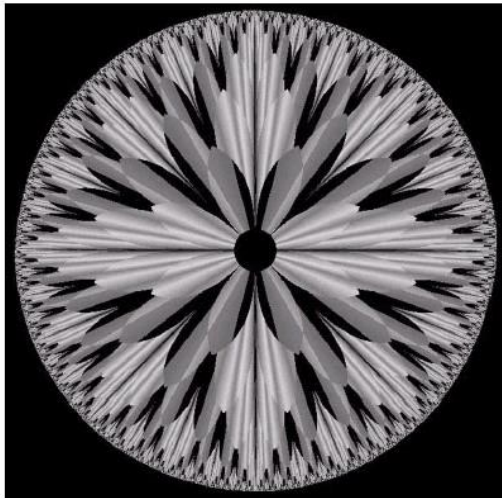
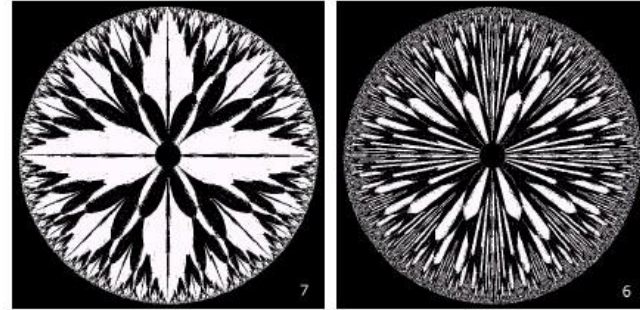
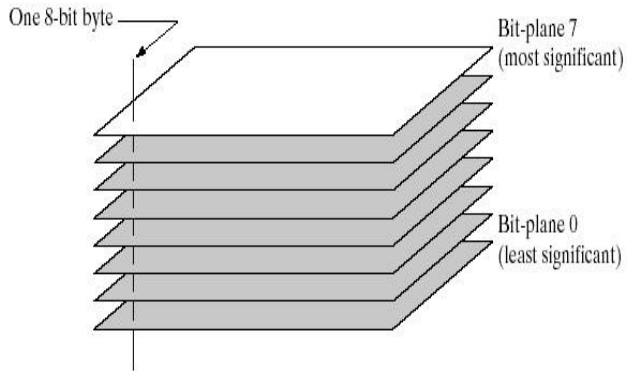


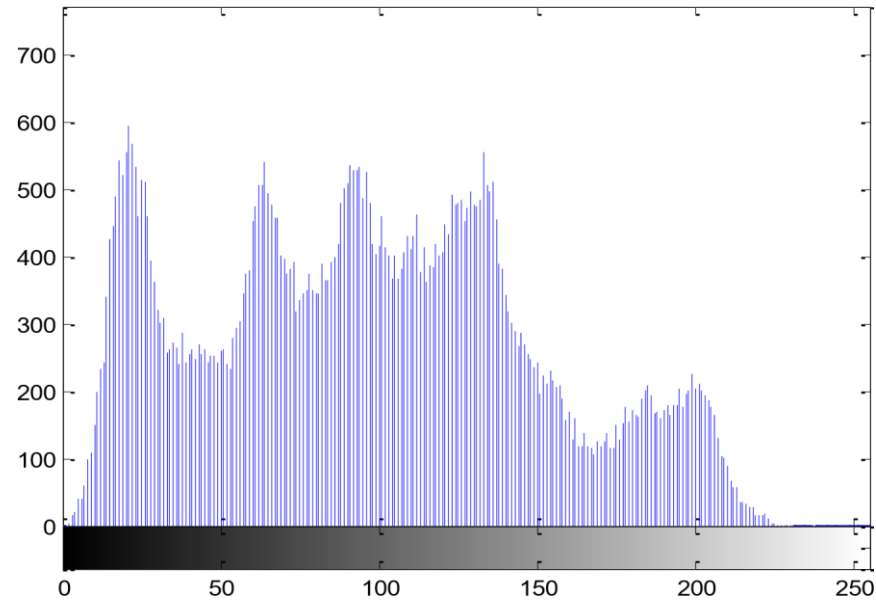
FIGURE 3.14 The eight bit planes of the image in Fig. 3.13. The number at the bottom, right of each image identifies the bit plane.

Histogram

- Let I be a grayscale image.
- $I(r,c)$ is an 8-bit integer between 0 and 255.
- Histogram, h_I , of I :
 - a 256-element array, h_I
 - $h_I(g)$, for $g = 1, 2, 3, \dots, 256$, is an integer
 - $h_I(g) =$ number of pixels in I that have value $g-1$.

Histogram of a Grayscale image

```
f=imread('lena.bmp');  
Imhist(f);
```



Histogram of a Color image

□ If I is a 3-band image (truecolor, 24-bit) □

Either I has 3 histograms:

- $h_R(g) = \#$ of pixels in $I(:, :, 1)$ with intensity value $g-1$
- $h_G(g) = \#$ of pixels in $I(:, :, 2)$ with intensity value $g-1$

- $h_B(g) = \#$ of pixels in $I(:,:,3)$ with intensity value $g-1$ or 1 vector-valued histogram, $h(g, 1, b)$ where
 - $h(g, 1, 1) = \#$ of pixels in I with red intensity value $g-1$
 - $h(g, 1, 2) = \#$ of pixels in I with green intensity value $g-1$
 - $h(g, 1, 3) = \#$ of pixels in I with blue intensity value $g-1$

Histogram Equalization

- Remap image I (*Mapping*)
- The transformation function is

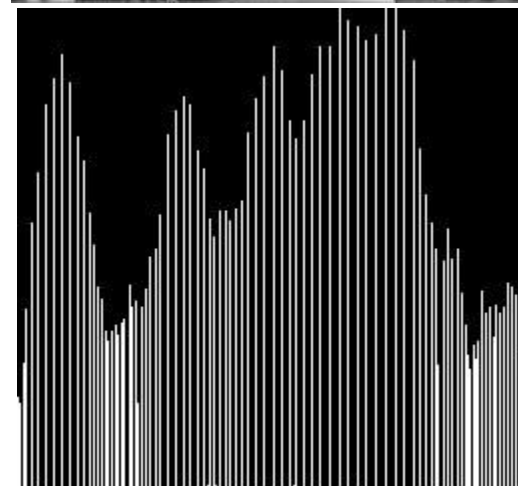
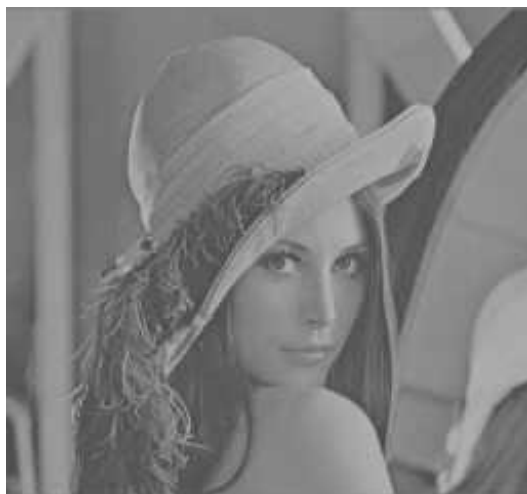
$$s_k = \sum_{j=0}^k \frac{n_j}{n}$$

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

Histogram Equalization

- `>>g=histeq(f, nlev)`
- `f` is the input image and `nlev` is the number of intensity levels specified for the output image
- If `nlev` is equal to `L` (total number of possible levels in the input image) then equalization is similar to input
- If `nlev` is less than `L`, then `histeq` attempts to distribute the levels so that they will approximate a flat histogram
- Default value for `nlev` is 64

Histogram Equalization: Example



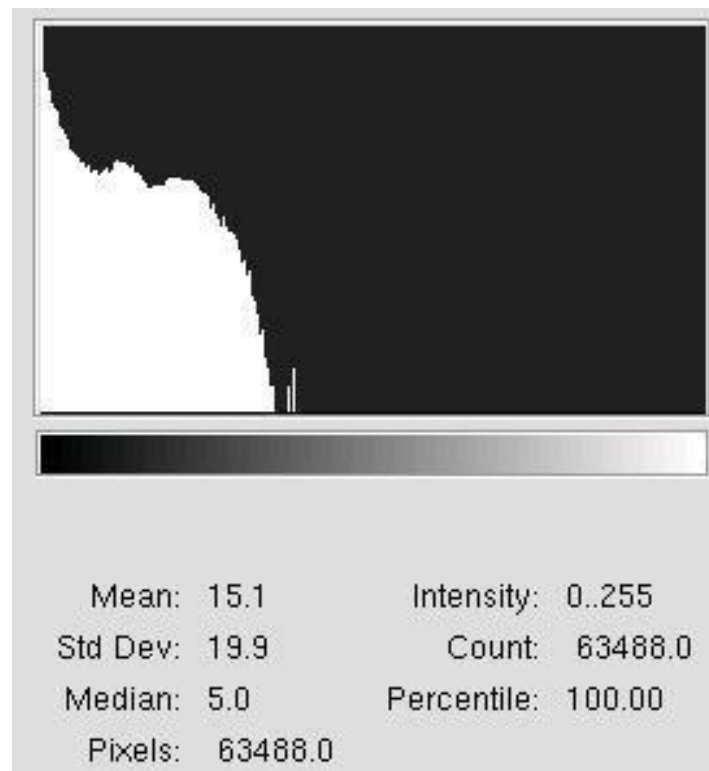
Original

Equalized

HE -examples

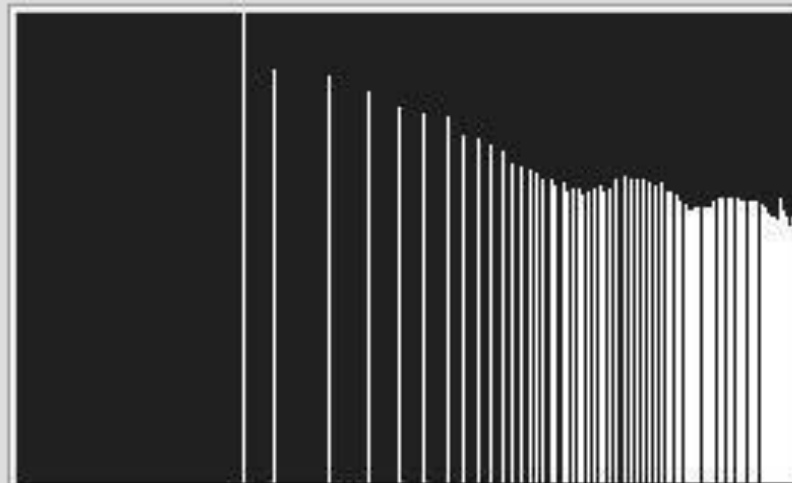
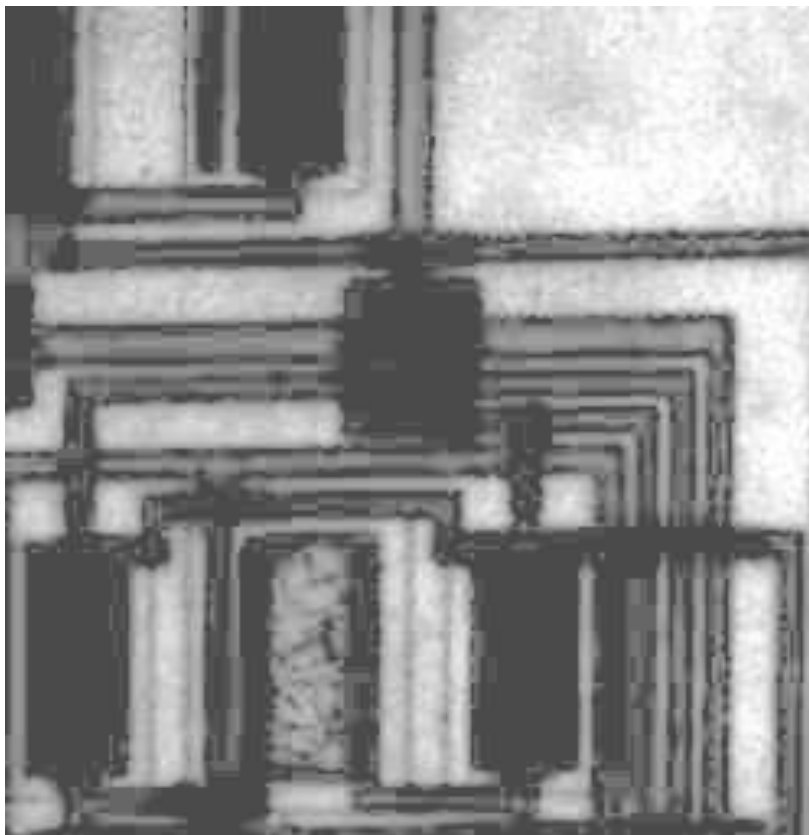


A dark image



Histogram

HE-examples



Mean: 140.8	Intensity: 0.255
Std Dev: 59.4	Count: 63488.0
Median: 133.0	Percentile: 100.00
Pixels: 63488.0	

Image Subtraction

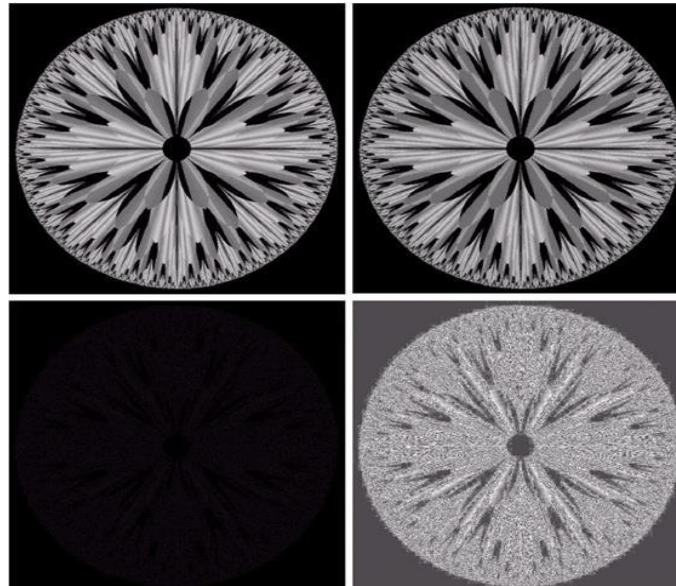
- A more interesting arithmetic operation is pixel-wise subtraction of two images.

$$g(x, y) = f(x, y) - h(x, y)$$

- Refer to image

the fractal again.

original4 lower-order



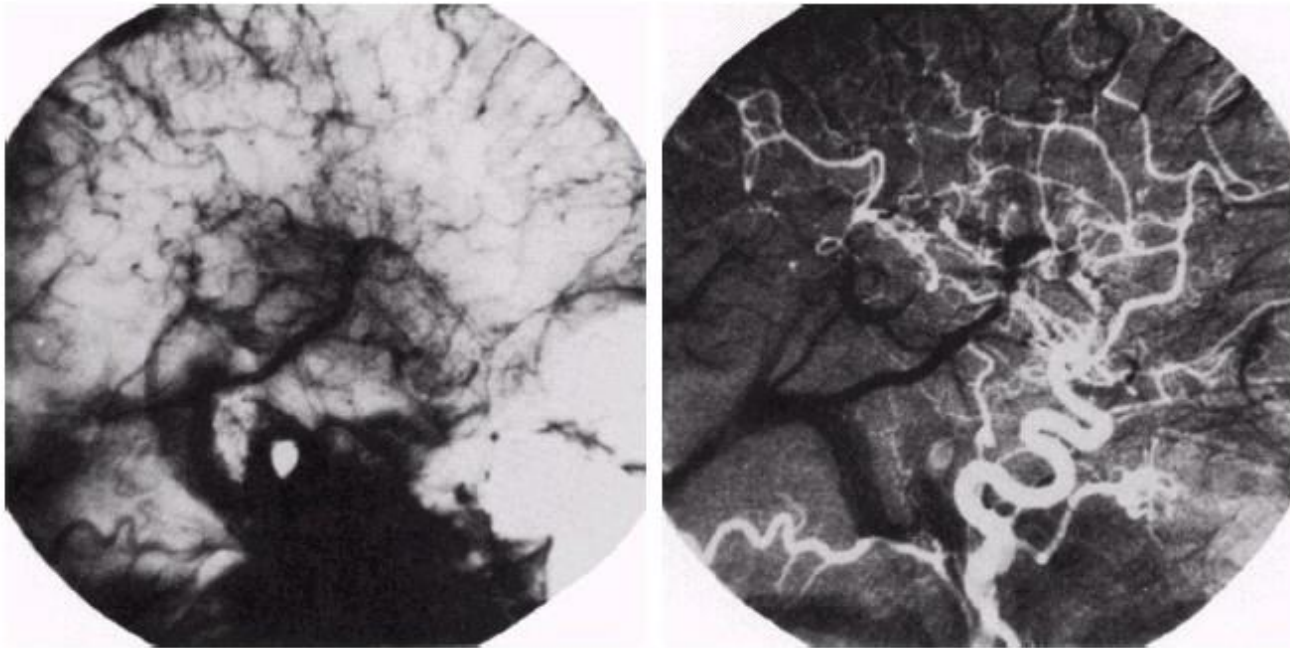
bit

planes zeroed out

differencecontrast

enhanced

Image Subtraction



a b

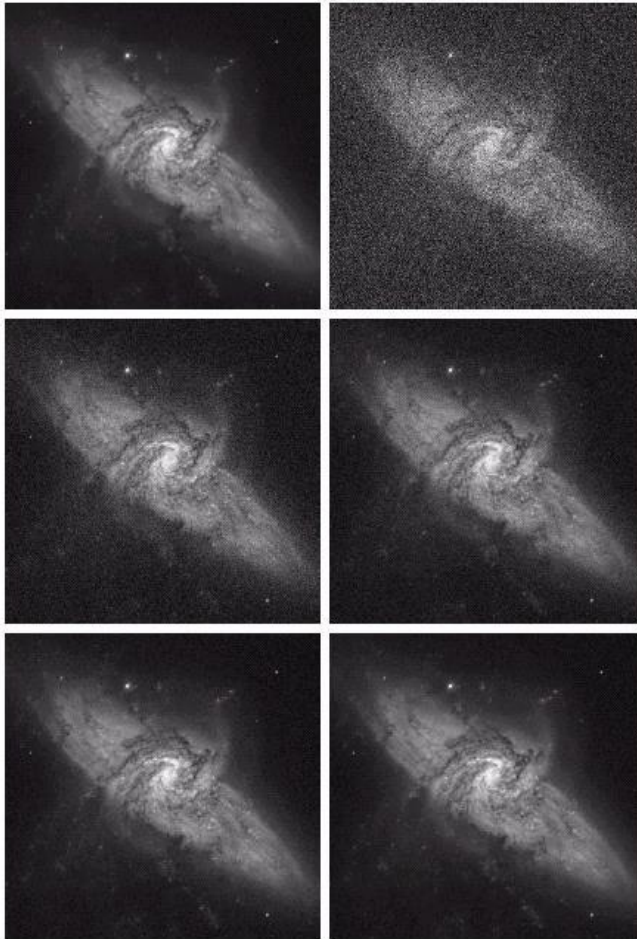
FIGURE 3.29

Enhancement by image subtraction. (a) Mask image. (b) An image (taken after injection of a contrast medium into the bloodstream) with mask subtracted out.

Mask mode radiography

Image

Averaging



- Same signal, but different noise realization.
- Averaging of many such images will enhance SNR.

Basics of Spatial Filtering

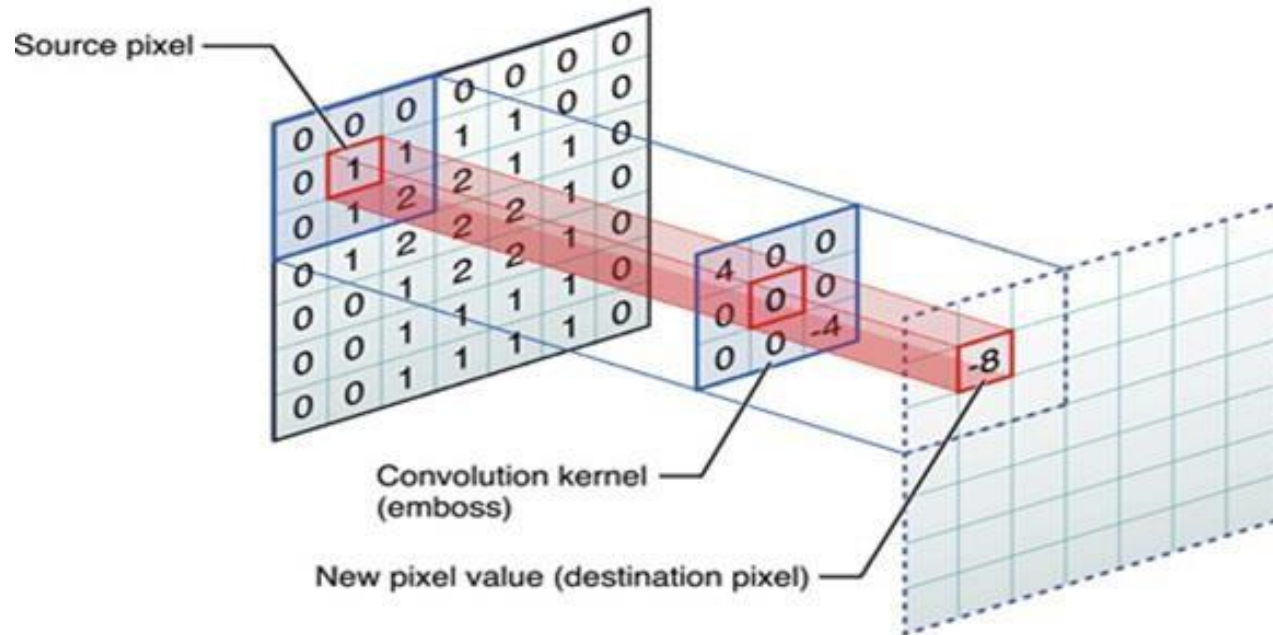
- Defining a center point (x, y)
- Performing an operation that involves only the pixels in a predefined neighborhood about that center point and sub image.
- Sub image has the same dimensions as neighborhood.
- The sub image is called a filter, mask, kernel, template or window.
- Letting the result of that operation be the response of the process at that point
- Repeating the process for every point in the image

Process of Spatial Filtering

The process consists of,

- Moving the filter mask from point to point in an image.
- At each point (x, y) , the response of the filter at that point is calculated.
- The response is sum of products of the filter coefficients and the corresponding image pixels in the area spanned by the filter mask.
- It is similar to frequency domain concept called convolution.
- So, the linear filtering process is often referred to as

“Convoluting a mask with an image”



<i>a</i>	<i>b</i>	<i>c</i>
<i>d</i>	<i>e</i>	<i>e</i>
<i>f</i>	<i>g</i>	<i>h</i>

**Source
Pixels**

<i>r</i>	<i>s</i>	<i>t</i>
<i>u</i>	<i>v</i>	<i>w</i>
<i>x</i>	<i>y</i>	<i>z</i>

Kernel

$$\begin{aligned} \text{New pixel} = & v * e + z * a + y * b + x * c + w * d \\ & + u * e + t * f + s * g + r * h \end{aligned}$$

Convolution

- Response R of an $m \times n$ mask at any point (x, y) , is expressed as follows:

$$\begin{aligned} R &= w_1 z_1 + w_2 z_2 + \dots + w_{mn} z_{mn} \\ &= \sum_{i=1}^{mn} w_i z_i \end{aligned}$$

- Example : For the 3×3 general mask, the response at any point (x, y) in the image is given by,

$$R = w_1 z_1 + w_2 z_2 + \dots + w_9 z_9$$

$$= \sum_{i=1}^9 w_i z_i$$

Correlation and Convolution

- Correlation is the process of passing the mask w by the image array f
- Convolution is the same process, except that w is rotated by 180° prior to passing it by f

	f	w
Correlation	0 0 0 1 0 0 0 0	1 2 3 2 0

Convolution 0 0 0 1 0 0 0 0

0 2 3 2 1

Smoothing Filter

- Smoothing filters are used for blurring and for noise reduction.
- It is preprocessing step such as removal of small details, bridging of small gaps.
- Noise reduction can be done by blurring with linear and non-linear filtering.
- Pixel averaging in the spatial domain:
 - Each pixel in the output is a weighted average of its neighbors.
 - Is a convolution whose weight matrix sums to 1.

8-Neighbor Mean filter

1/9	1/9	1/9
-----	-----	-----

4-Neighbor Mean filter

0	1/5	0
---	-----	---

1/5	0	1/5
-----	---	-----

1/9	1/9	1/9
1/9	1/9	1/9

1/5	1/5	1/5
0	1/5	0

0	1/5	0
1/5	0	1/5

Smoothing Filter

- These filters are also called averaging filter/low-pass filter.
- Idea behind Smoothing filters are straightforward.
- i.e. Replacing the value of every pixel in an image by the average of the gray levels in the neighborhood defined by the filter mask.
- The result is reduced “sharp” transitions in gray levels.
- Order statistics filter- Mean, median
- Reduces Salt & pepper noise

$$1/9 \times$$

1	1	1
1	1	1
1	1	1

1	2	1
2	4	2
1	2	1

Sharpening Filter

- These filters are also called Differencing/High-pass/Laplacian filter.
- Based on first and second order derivatives.
- Highlight fine details in an image.
- Applications are electronic printing, medical imaging, industrial inspections, and autonomous guidance in military systems.
- Pixel-differenced in the spatial domain:
 - Each pixel in the output is a difference between itself and a weighted average of its neighbors.

- Is a convolution whose weight matrix sums to 0.

Blurring vs. Sharpening

- Blurring/smooth is done in spatial domain by pixel averaging in a neighbors, it is a process of integration.
- Sharpening is an inverse process, to find the difference by the neighborhood, done by spatial differentiation.

Derivative operator

- The strength of the response of a derivative operator is proportional to the degree of discontinuity of the image at the point at which the operator is applied.
- Image differentiation
 - enhances edges and other discontinuities (noise)
 - deemphasizes area with slowly varying gray-level values.

High-Pass Filters

Laplacian Operator

0	1	0
1	-4	1

0	-1	0
-1	4	-1

-1	-1	-1
-1	8	-1

1	1	1
1	-8	1

0	1	0
---	---	---

0	-1	0
---	----	---

-1	-1	-1
----	----	----

1	1	1
---	---	---

Sobel Operator

-1	0	1
-2	0	2
-1	0	1

1	2	1
0	0	0
-1	-2	-1

0	1	2
-1	0	1
-2	-1	0

-2	-1	0
-1	0	1
0	1	2

THANK YOU

THIS CONTENT IS TAKEN FROM THE TEXT BOOKS AND REFERENCE BOOKS
PRESCRIBED IN THE SYLLABUS