

Basic Concepts and Terminology

Applications of Image Processing

Image Processing

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graph TD; A[Image Processing] --- B[Medical Application]; A --- C[Industrial Application]; A --- D[Consumer Electronics]; A --- E[Military Application]; A --- F[Security Application]
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Medical
Application

Industrial
Application

Consumer
Electronics

Military
Application

Security
Application

What is an Image?

Discrete representation of data
possessing both spatial(layout)
and intensity (color) information.

What is a digital image?

Representation of a two-dimensional image using a finite number of points usually called picture elements or *pixels*.

$I(0,0)$

$I(0,N)$



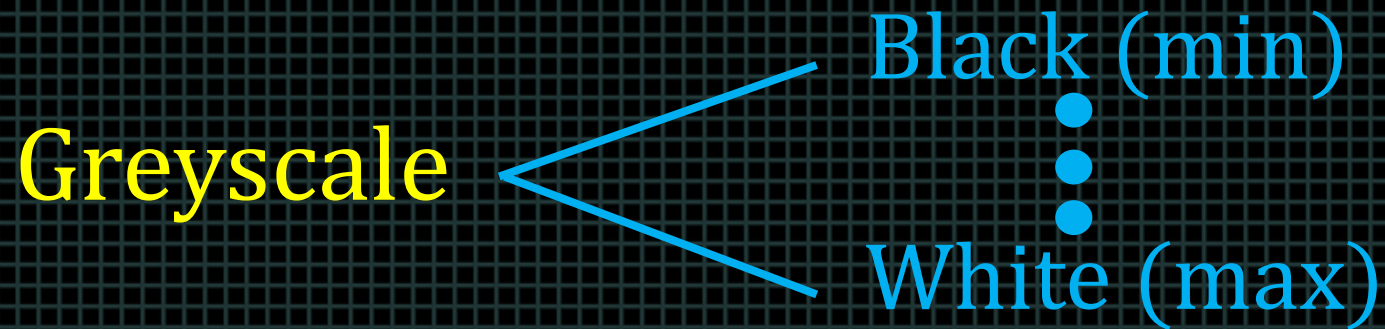
Image pixel location at
(m,n) denoted $I(m,n)$

$I(M,0)$

$I(M,N)$

Image Color

Black/white



RGB : Red, Green, Blue

HSV : Hue, Saturation, Value (Intensity)

Image Resolution and Quantization

Size of 2-D grid and data size stored for each individual image pixel.





Resolution of image source

- Spatial Resolution

$C \times R$ e.g. 640 x 480, 800 x 600

- Temporal Resolution

e.g. 25fps

- Bit Resolution

e.g. 24 bit

Image Formats

JPEG

Joint Photographic
Experts Group

Lossy compression

GIF

Graphics Interchange
Format

Lossless compression
Limited to 8bit color

BMP

Bit map picture

Basic format
Lossless compression

PNG

Portable network
graphics

Lossless compression

TIFF

Tagged Image file format

Very flexible
Compressed/ Uncompressed

Image Data Types

- Binary Image
- Intensity or greyscale Image
- RGB or true-color Image
- Floating – Point Image

Scope of Image Processing

Low Level

Primitive operations

e.g. noise reduction

Mid Level

Extraction of attributes

e.g. edges and contours

High Level

Analysis and interpretation

Example of some image processing
operations.

1. Sharpening



original



sharper

2. Noise Removal



original



noisy

3. De-blurring



Blur image

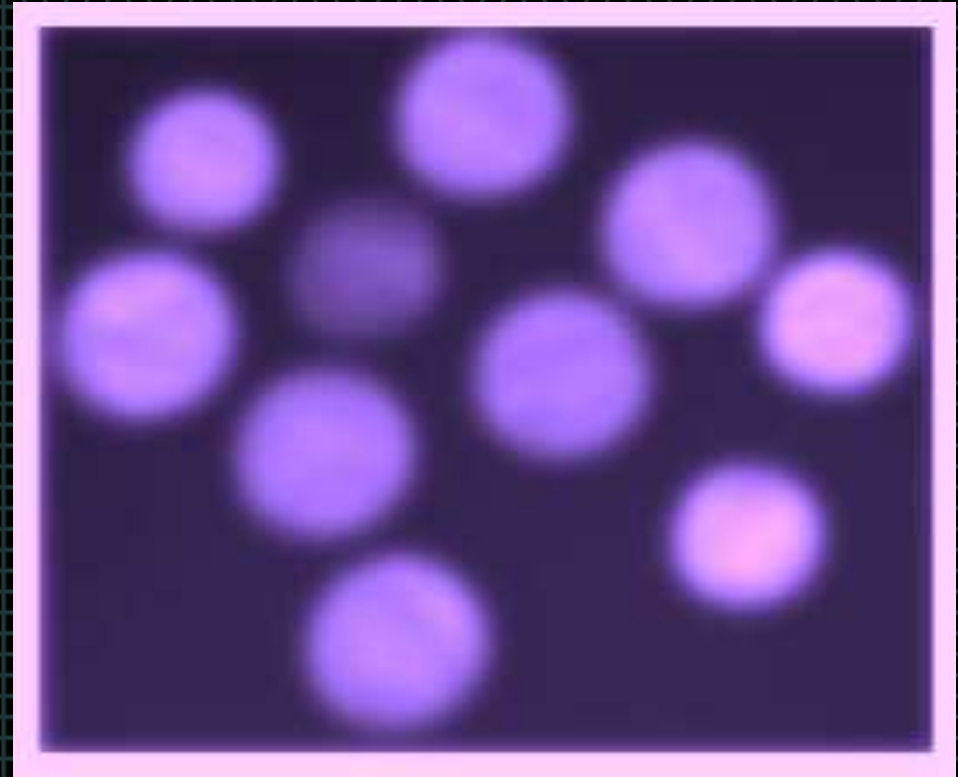


De-blurred image

4. Blurring



original image

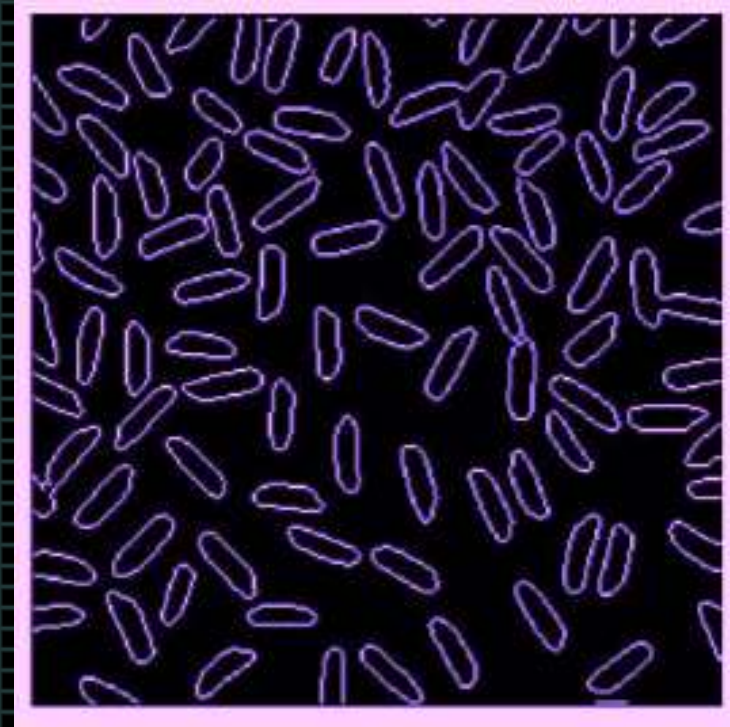


blurred image

4. Edge Extraction



original image

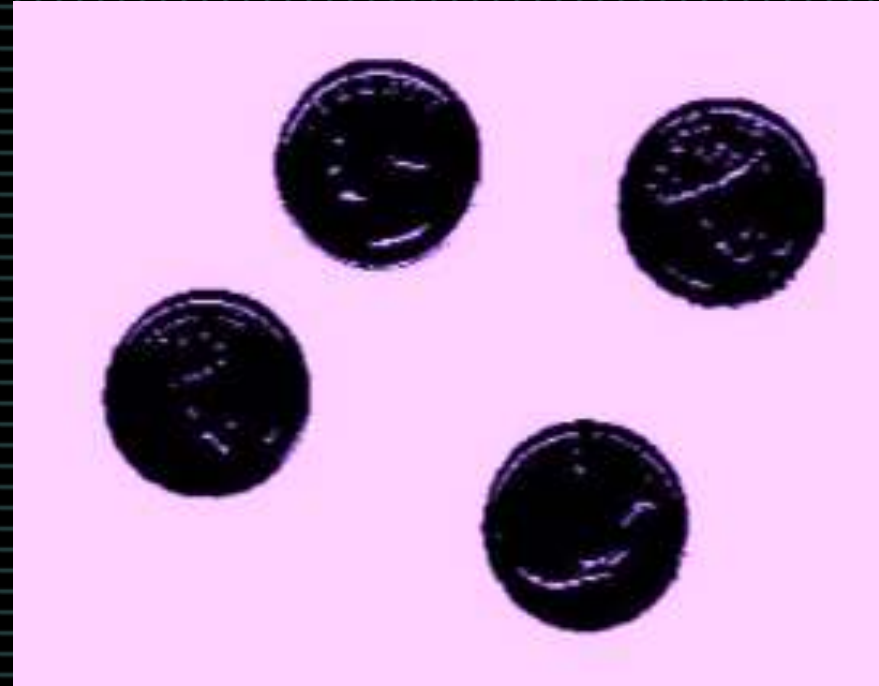


After edge extraction

5. Binarization



original image



After binarization

7. Contrast Enhancement

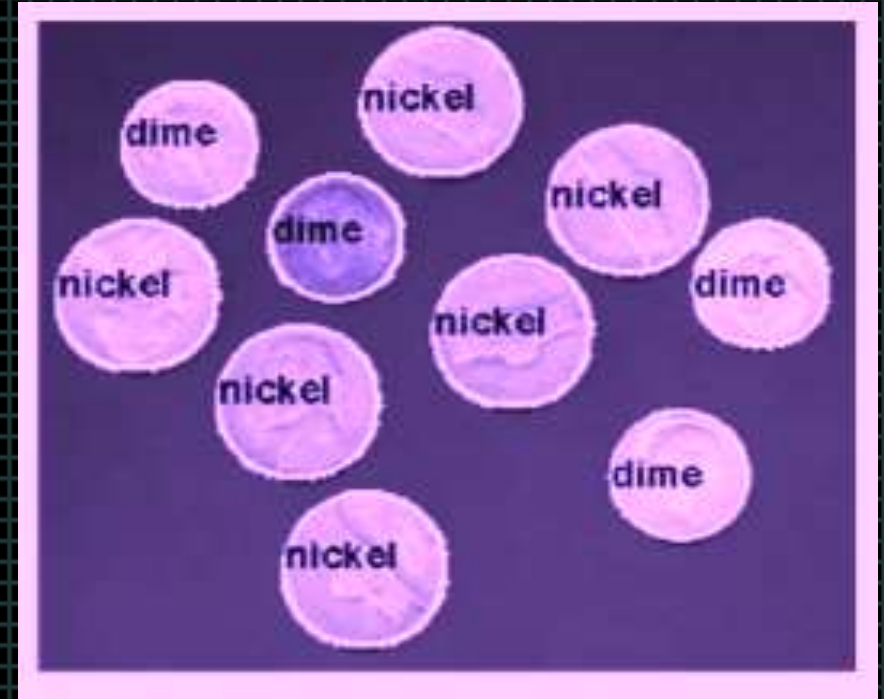


original image



After contrast enhancement

6. Object Segmentation and Labeling



Some Basic Terms

Image Topology

Investigation of fundamental image properties using morphological operators

Neighborhood

Pixels surrounding a given pixel

Adjacency

Two pixels p and q are 4-adjacent if they are 4-neighbors of each other and 8-adjacent if they are 8-neighbors of one another.

Paths

A 4-path between two pixels p and q is a sequence of pixels starting with p and ending with q such that each pixel in the sequence is 4-adjacent to its predecessor in the sequence

Components

A set of pixels connected to each other

Connectivity

Existence path between two pixels

Overview of machine vision systems

Problem
domain



Acquisition



Preprocessing



Segmentation



Feature
Extraction



Classification



Result

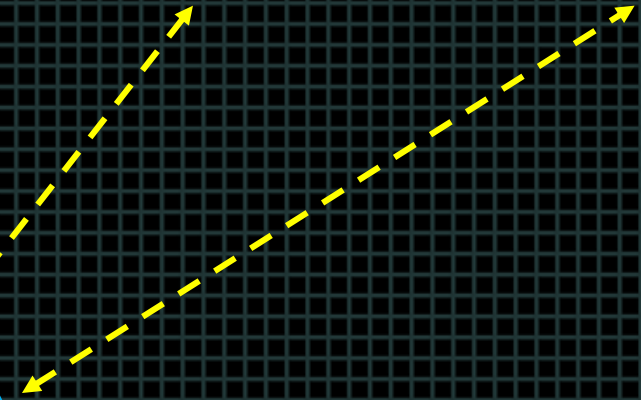
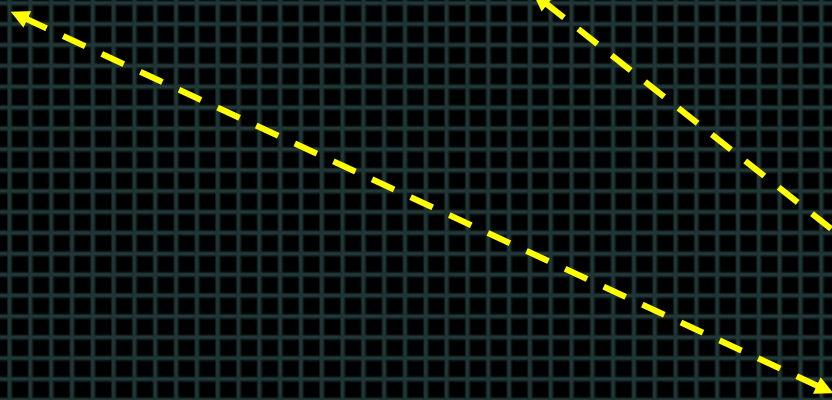


Image Formation

Understanding the formation of an image

$$\text{Image} = \text{PSF} * \text{Object function} + \text{Noise}$$

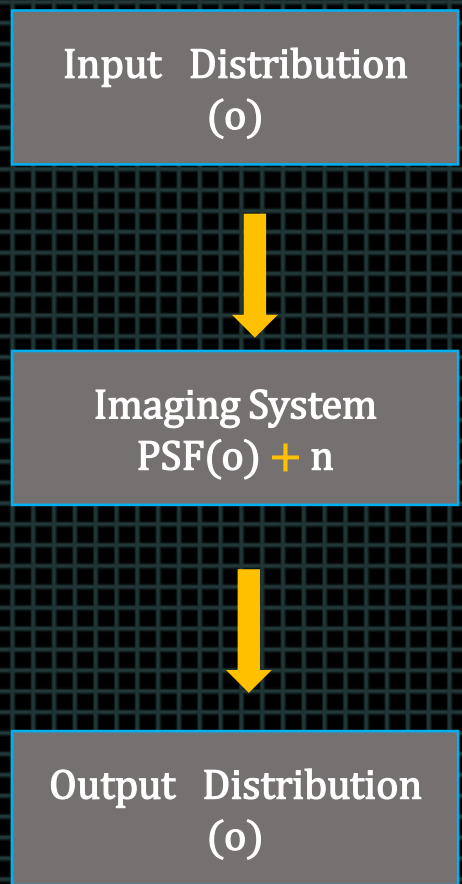
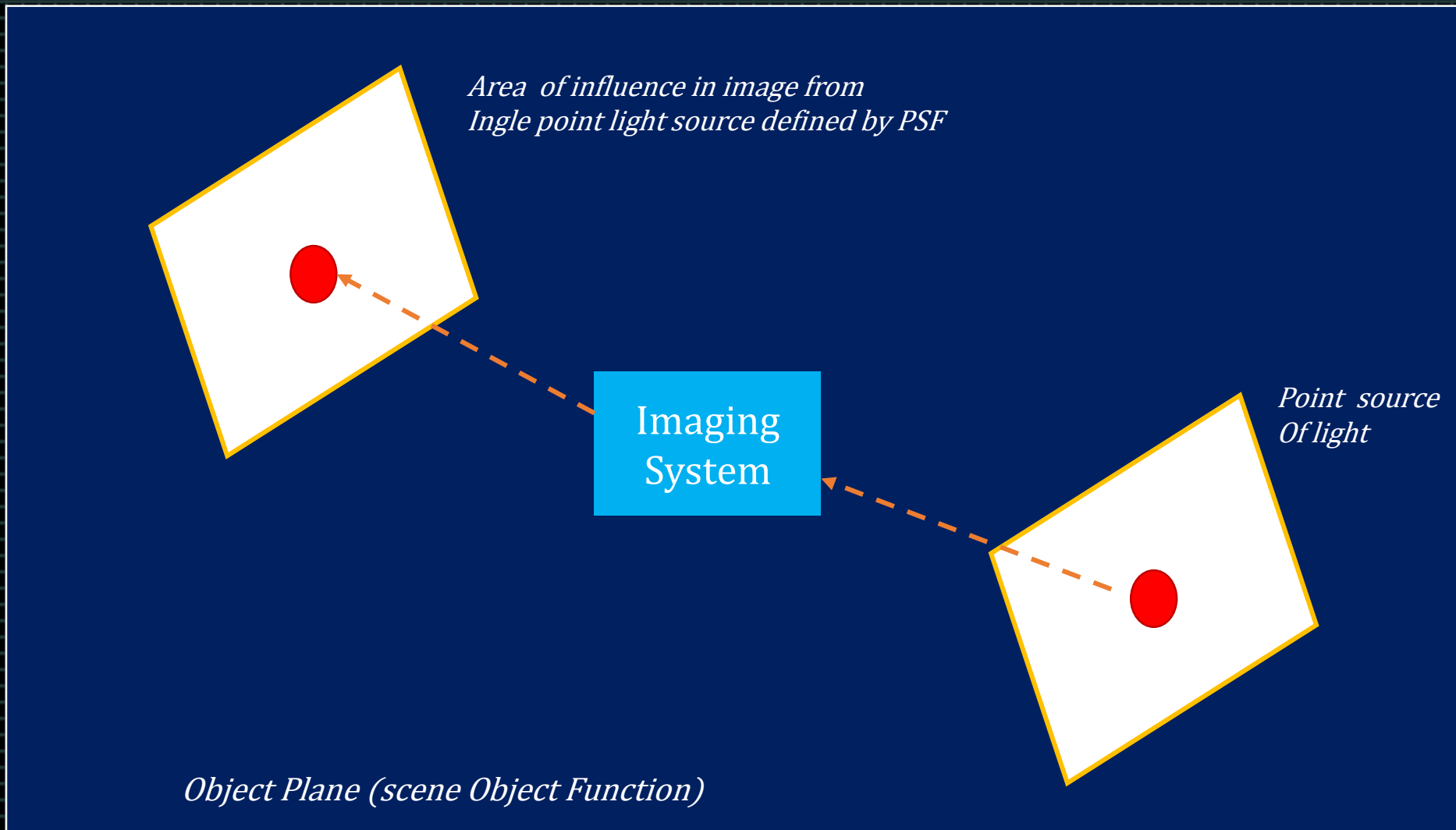
$$S = p * o + n$$

Convolution

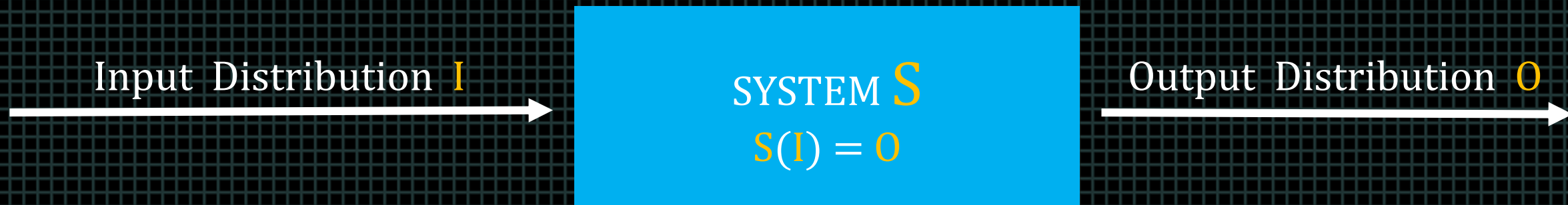
- The way information on the object function is spread .
- Characteristic of imaging device

The way light is reflected from object to imaging instrument

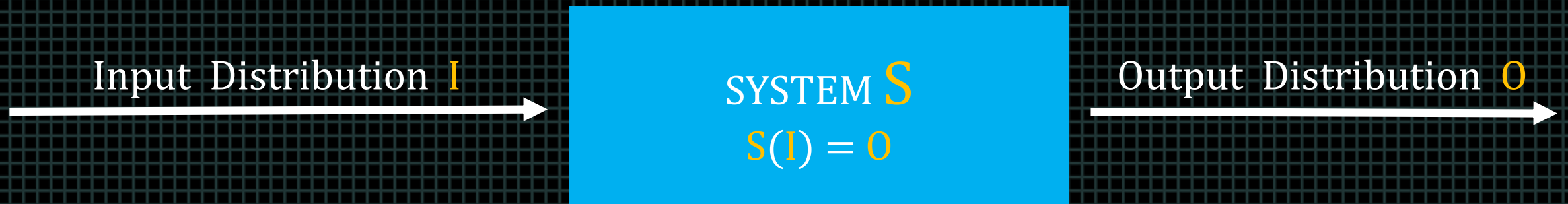
Unwanted external disturbances



Understanding the mathematics of image formation







Linear imaging systems

$$S\{aX + bY\} = aS\{X\} + bS\{Y\}$$

A



B



$A + B$



$S\{A + B\}$



$S\{A\}$



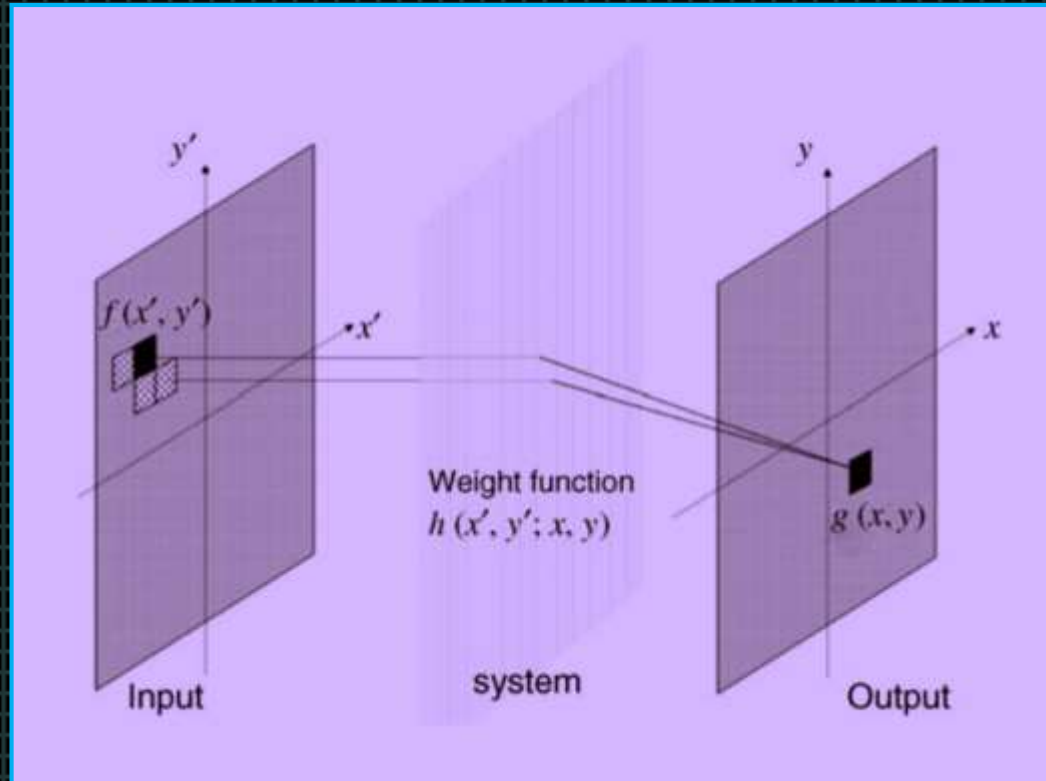
$S\{B\}$



$S\{A\} + S\{B\}$

Linear superposition integral

$$g(x, y) = \iint f(x', y') h(x, y; x', y') dx' dy'$$



The Dirac delta or impulse function

- Represents a bright intensity source
- Occupies infinitesimal region in space

1-D rectangle function :

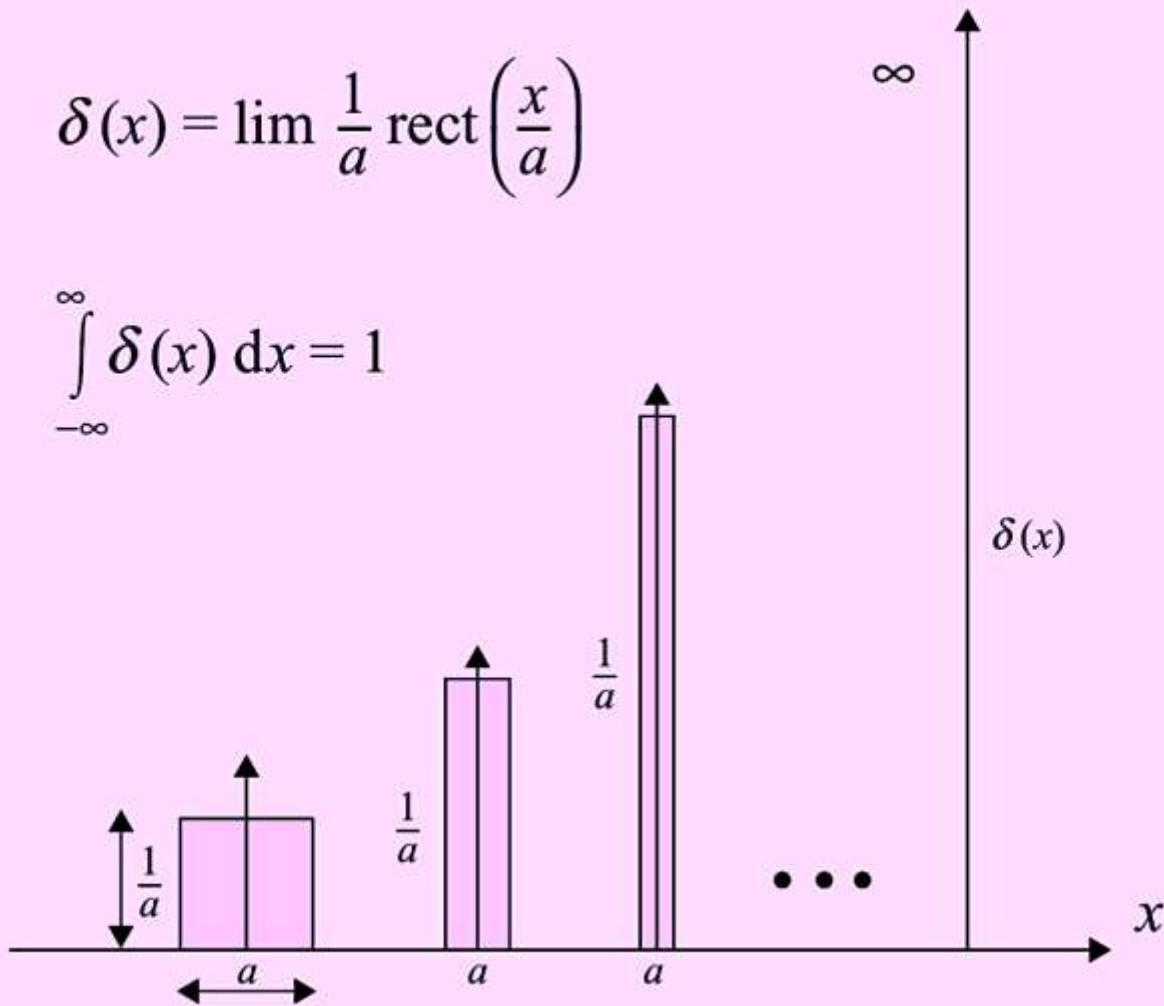
$$\begin{aligned} \text{rect}\left(\frac{x}{a}\right) &= 1 \quad |x| < a/2 \\ &= 0 \quad \text{otherwise} \end{aligned}$$

$$\delta(x) = \lim_{a \rightarrow 0} \frac{1}{a} \text{rect}\left(\frac{x}{a}\right) \quad \text{1-D}$$

$$\delta(x, y) = \lim_{a \rightarrow 0} \frac{1}{a} \text{rect}\left(\frac{x}{a}\right) \text{rect}\left(\frac{y}{a}\right) \quad \text{2-D}$$

$$\delta(x) = \lim_{a \rightarrow 0} \frac{1}{a} \text{rect}\left(\frac{x}{a}\right)$$

$$\int_{-\infty}^{\infty} \delta(x) dx = 1$$



$$\begin{aligned} \delta(x) &= \infty & x &= 0 \\ &= 0 & x &\neq 0 \end{aligned}$$

$$\int_{-\infty}^{\infty} \delta(x) dx = 1$$

$$\begin{aligned} \delta(x - x_0) &= \infty & x &= x_0 \\ &= 0 & x &\neq x_0 \end{aligned}$$

$$\begin{aligned} \delta(x, y) &= \infty & x &= 0, y = 0 \\ &= 0 & \text{otherwise} \end{aligned}$$

$$\iint_{-\infty}^{\infty} \delta(x, y) dx dy = 1$$

$$\int_{-\infty}^{\infty} f(x) \delta(x - x_0) dx = f(x_0) \quad \text{1-D}$$

$$\iint_{-\infty}^{\infty} f(x, y) \delta(x - x_0, y - y_0) dx dy = f(x_0, y_0) \quad \text{2-D}$$

The Sifting theorem

- *Singularity*

$$\begin{aligned}\delta(x) &= \infty & x &= 0 \\ &= 0 & x &\neq 0\end{aligned}$$

- *Unit area*

$$\int_{-\infty}^{\infty} \delta(x) dx = 1$$

- *Shift property*

$$\int_{-\infty}^{\infty} f(x) \delta(x - x_0) dx = f(x_0)$$

Properties of the delta function

The Point – Spread Function (PSF)

**The response of a system to
an input distribution consisting of a very
small intensity point.**

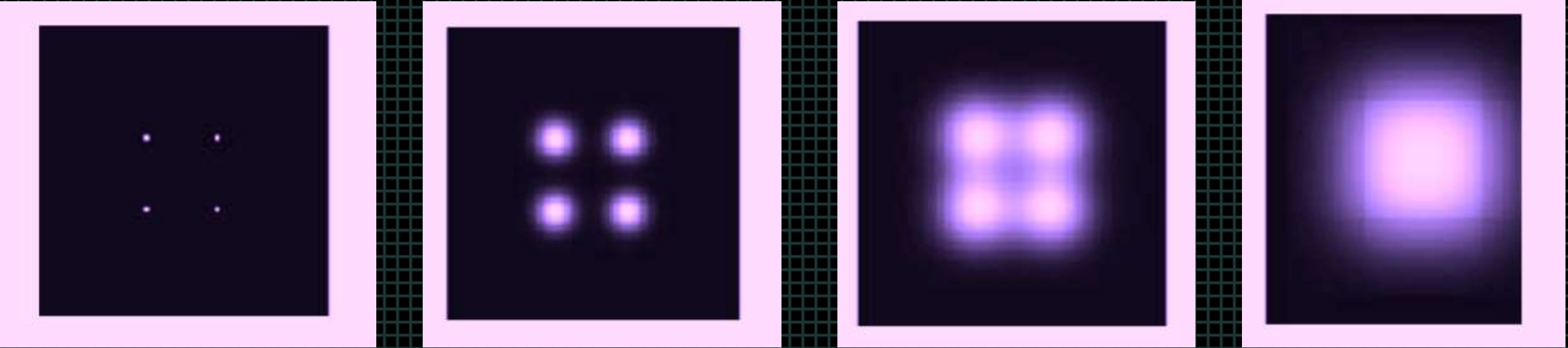
(Dirac delta function)

Input function : $f(x', y') = \delta(x' - x_0, y' - y_0)$

(Linear superposition integral)

$$g(x, y) = \iint \delta(x' - x_0, y' - y_0) h(x, y; x', y') dx' dy'$$

$$g(x, y) = h(x, y; x_0, y_0)$$



Effects of system PSF

Arithmetic and Logical Operations

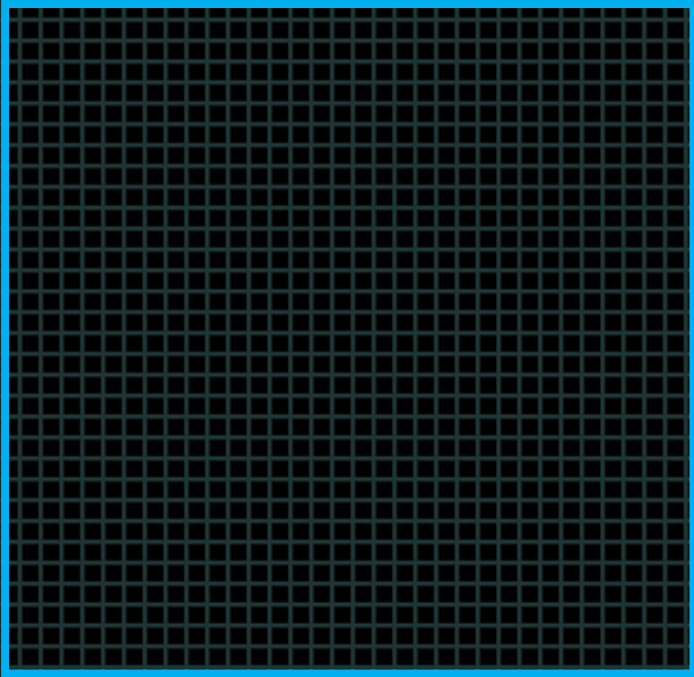
Arithmetic Operations

$$X \text{ } opn \text{ } Y = Z$$

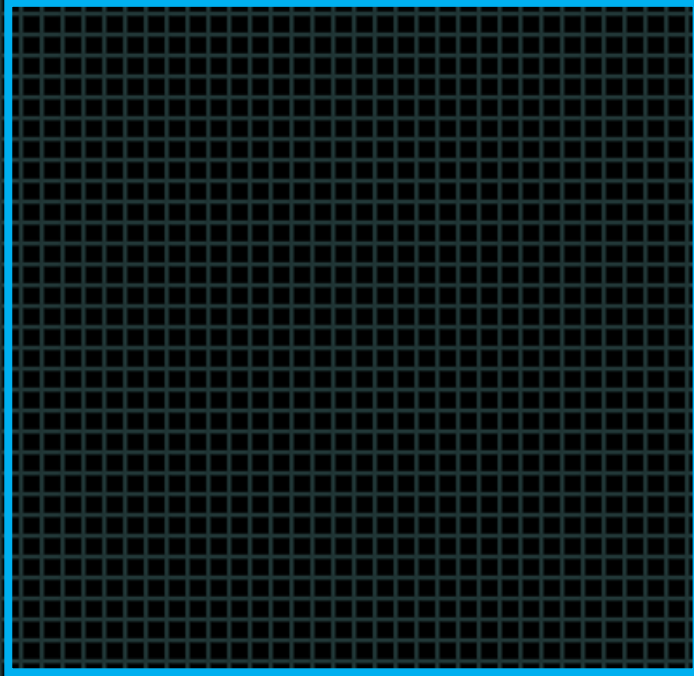
opn : (+, −, ×, /)

X : 2D array

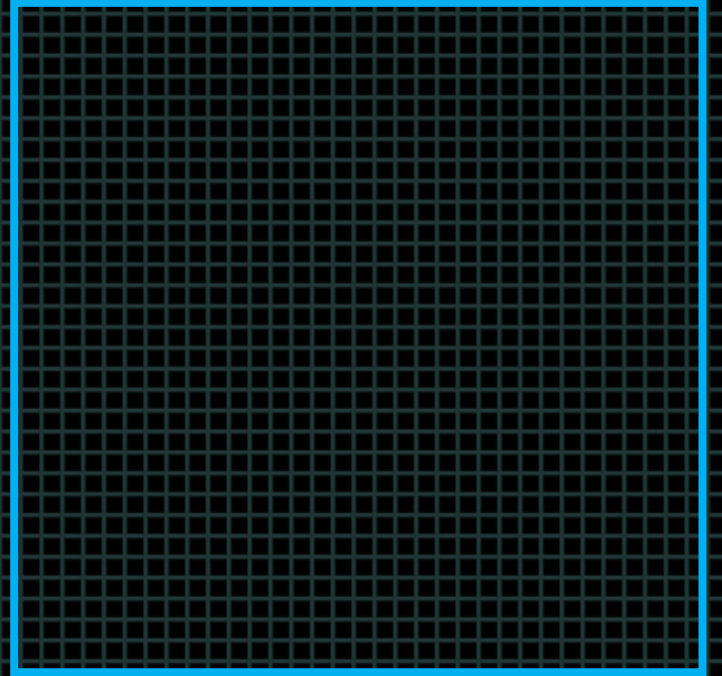
Y : another 2D array



First image



Second image



Result

Adding to images

Normalization

$$g = \frac{L_{max}}{f_{max} - fmin} (f - fmin)$$

f : current pixel in w

W : intermediate result variable

L_{max} : maximum possible intensity

g : corresponding pixel in Z

f_{max} : maximum pixel in W

Addition

$$\begin{matrix} X \\ \begin{bmatrix} 200 & 100 & 100 \\ 0 & 10 & 50 \\ 50 & 250 & 120 \end{bmatrix} \end{matrix} + \begin{matrix} Y \\ \begin{bmatrix} 100 & 220 & 230 \\ 45 & 95 & 120 \\ 205 & 100 & 0 \end{bmatrix} \end{matrix} = \begin{matrix} W \\ \begin{bmatrix} 300 & 320 & 330 \\ 45 & 105 & 170 \\ 255 & 350 & 120 \end{bmatrix} \end{matrix}$$

Normalizing [45,350] range to [0,255]

$$\begin{matrix} Z_a \\ \begin{bmatrix} 213 & 230 & 238 \\ 0 & 50 & 105 \\ 175 & 255 & 63 \end{bmatrix} \end{matrix}$$

$$g = \frac{L_{max}}{f_{max} - f_{min}} (f - f_{min})$$

Truncating all values above 255 in W

$$\begin{matrix} Z_b \\ \begin{bmatrix} 255 & 255 & 255 \\ 45 & 105 & 170 \\ 255 & 255 & 120 \end{bmatrix} \end{matrix}$$

Subtraction

$$\begin{array}{c} X \\ \left[\begin{array}{ccc} 200 & 100 & 100 \\ 0 & 10 & 50 \\ 50 & 250 & 120 \end{array} \right] \end{array}
 \begin{array}{c} Y \\ \left[\begin{array}{ccc} 100 & 220 & 230 \\ 45 & 95 & 120 \\ 205 & 100 & 0 \end{array} \right] \end{array}
 =
 \begin{array}{c} Z_a \\ \left[\begin{array}{ccc} 100 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 150 & 120 \end{array} \right] \end{array}$$

With truncation
 $= X - Y$

$$Z_b = \left[\begin{array}{ccc} 0 & 120 & 130 \\ 45 & 85 & 70 \\ 155 & 0 & 0 \end{array} \right] = Y - X$$

With truncation

$$Z_c = \left[\begin{array}{ccc} 100 & 120 & 130 \\ 45 & 85 & 70 \\ 155 & 150 & 120 \end{array} \right] = |Y - X|$$

Multiplication

- Used to perform brightness adjustment
- Makes each pixel brighter or darker by multiplying its original value by a scalar factor
- Multiplication produces better subjective results than addition.

Logic Operations

- AND
- OR
- XOR
- NOT

Conventions

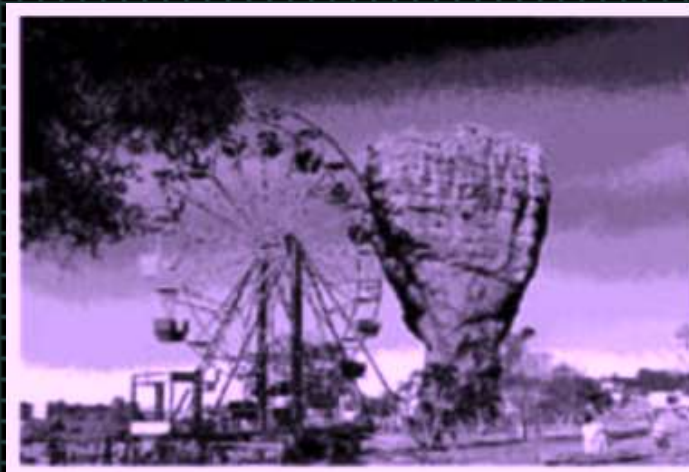
- 1 = true = white pixel
- 0 = false = black pixel



AND



=

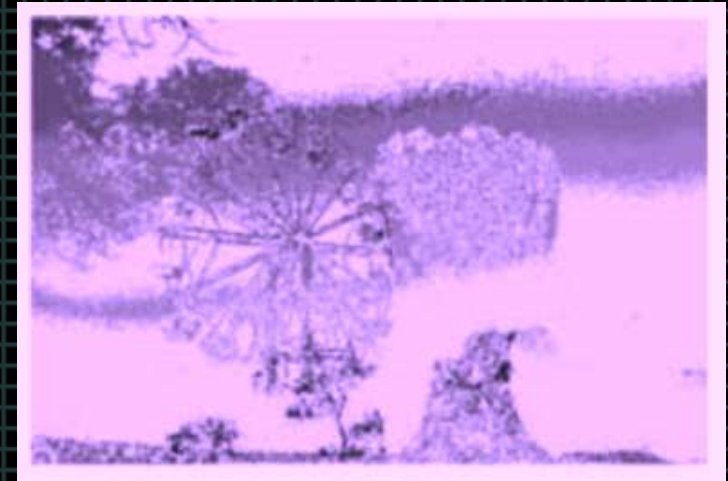




OR



=

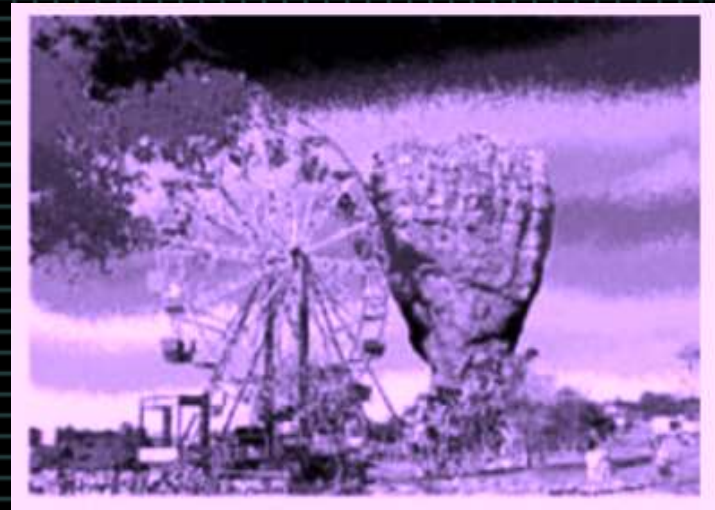




XOR

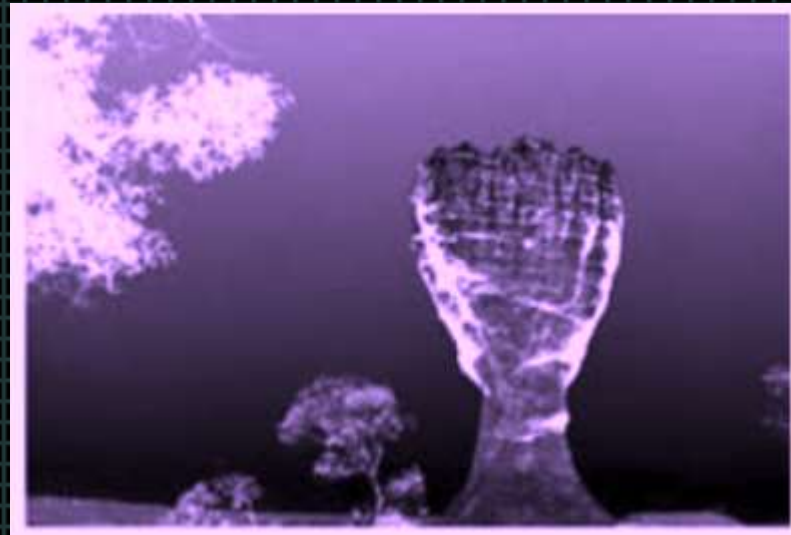


=





NOT →



Linear Shift-Invariance & Convolution

$$g(x, y) = \iint f(x', y') h(x, y; x', y') dx' dy'$$

128 x 128

$128^4 = 268.4$ million

$$h(x, y; x', y') = h(x'', y'') = h(x - x', y - y')$$

The convolution integral

2-D

$$g(x, y) = \int\limits_{-\infty}^{\infty} \int\limits_{-\infty}^{\infty} f(x', y') h(x - x', y - y') dx' dy'$$

1-D

$$g(x) = \int\limits_{-\infty}^{\infty} f(x') h(x - x') dx'$$

$$g(x, y) = f(x, y) ** h(x, y) \quad \text{2-D}$$

$$g(x) = f(x) * h(x) \quad \text{1-D}$$

Calculation of a 1-D Convolution Integral

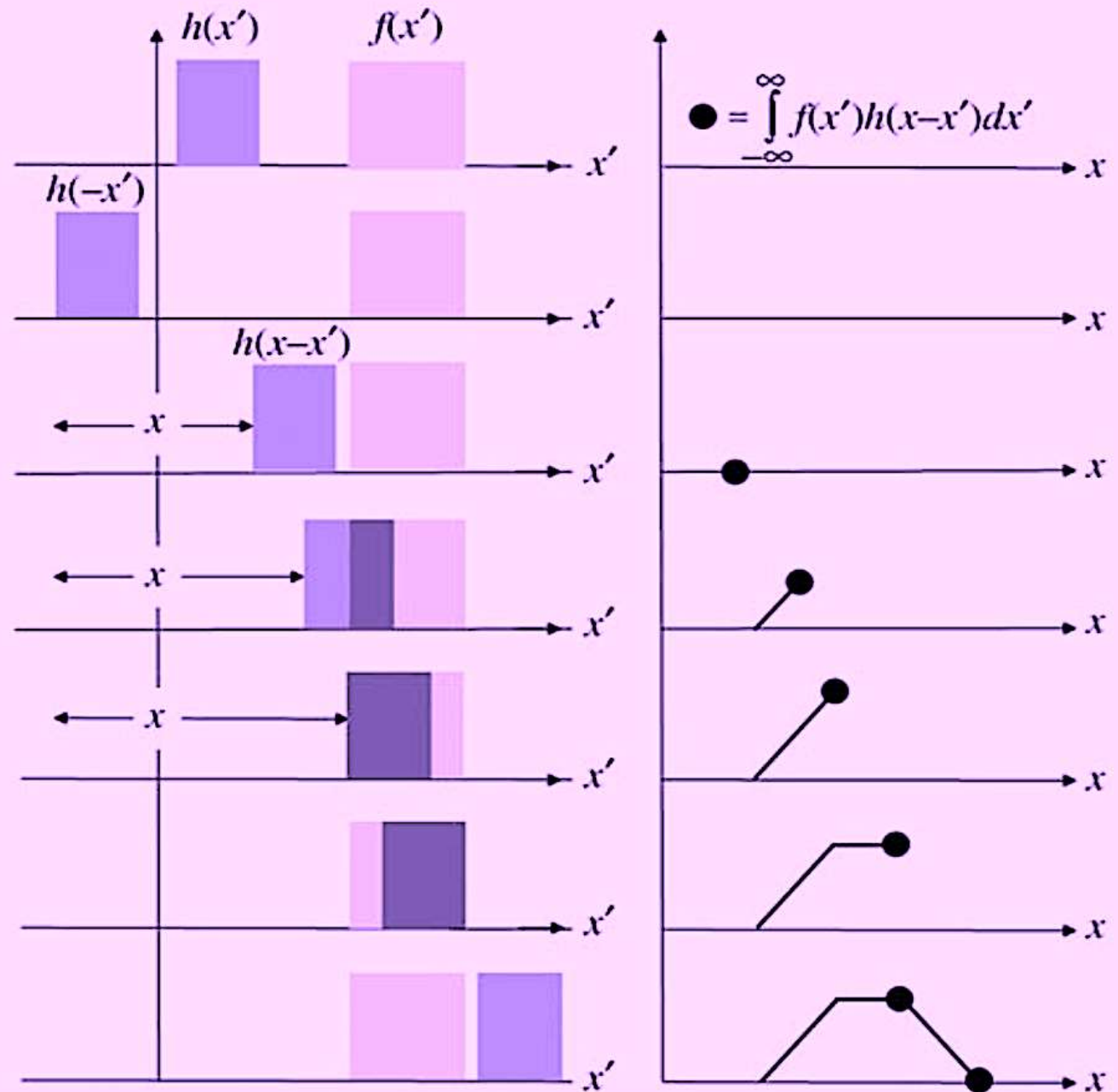
Define f and h

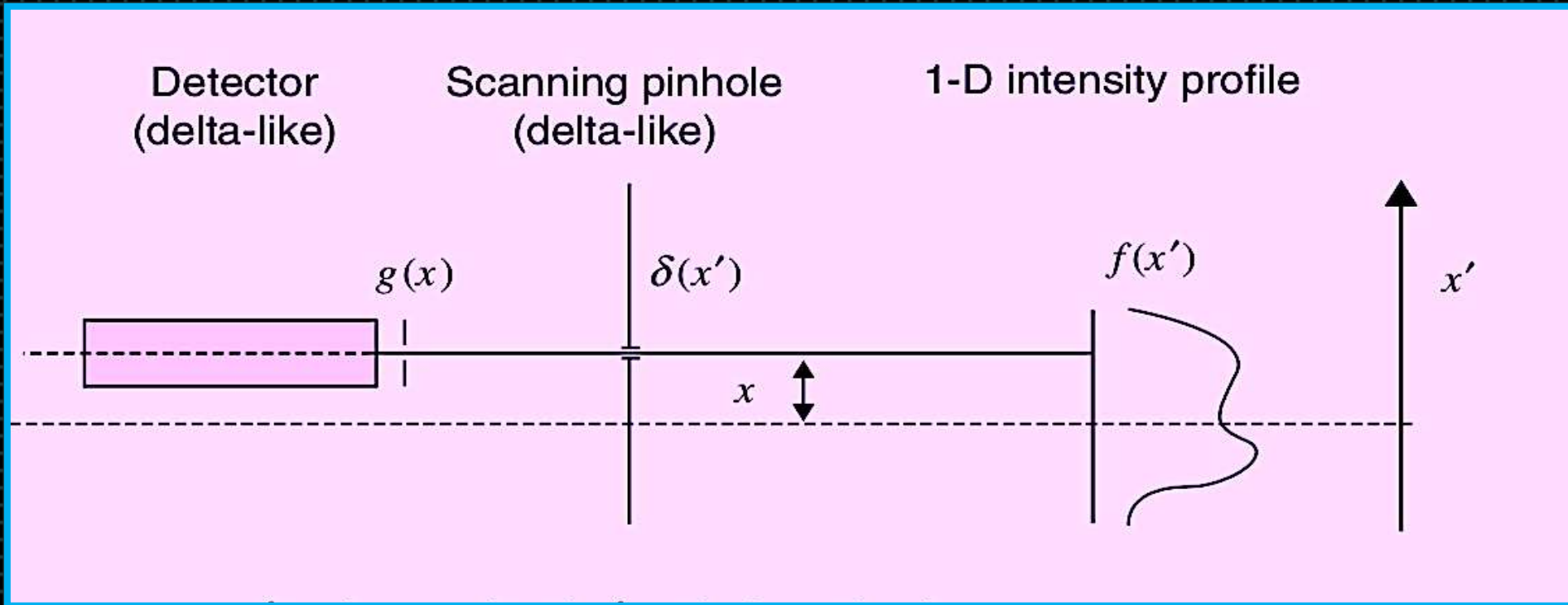
Flip h about y axis

Shift h by amount x
and calculate
overlap integral

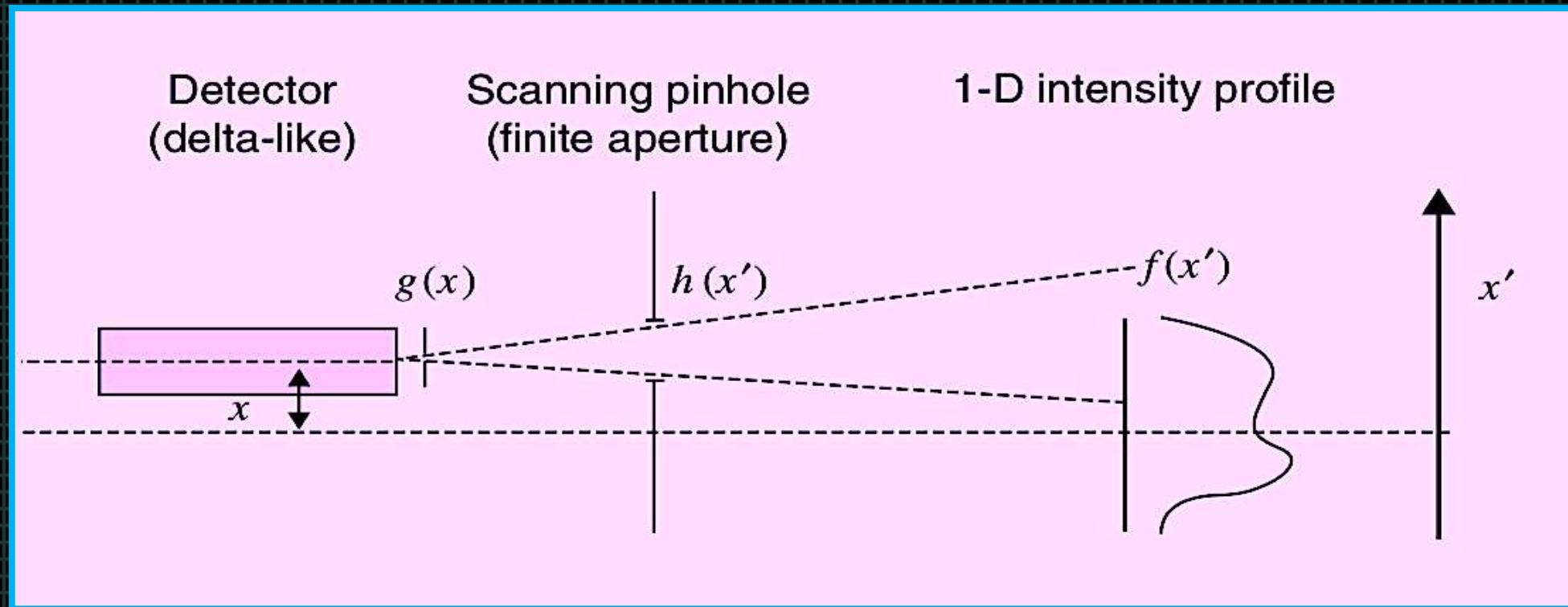
Repeatedly shift
 h by x and calculate
overlap integral

Continue for all
values with non-zero
overlap

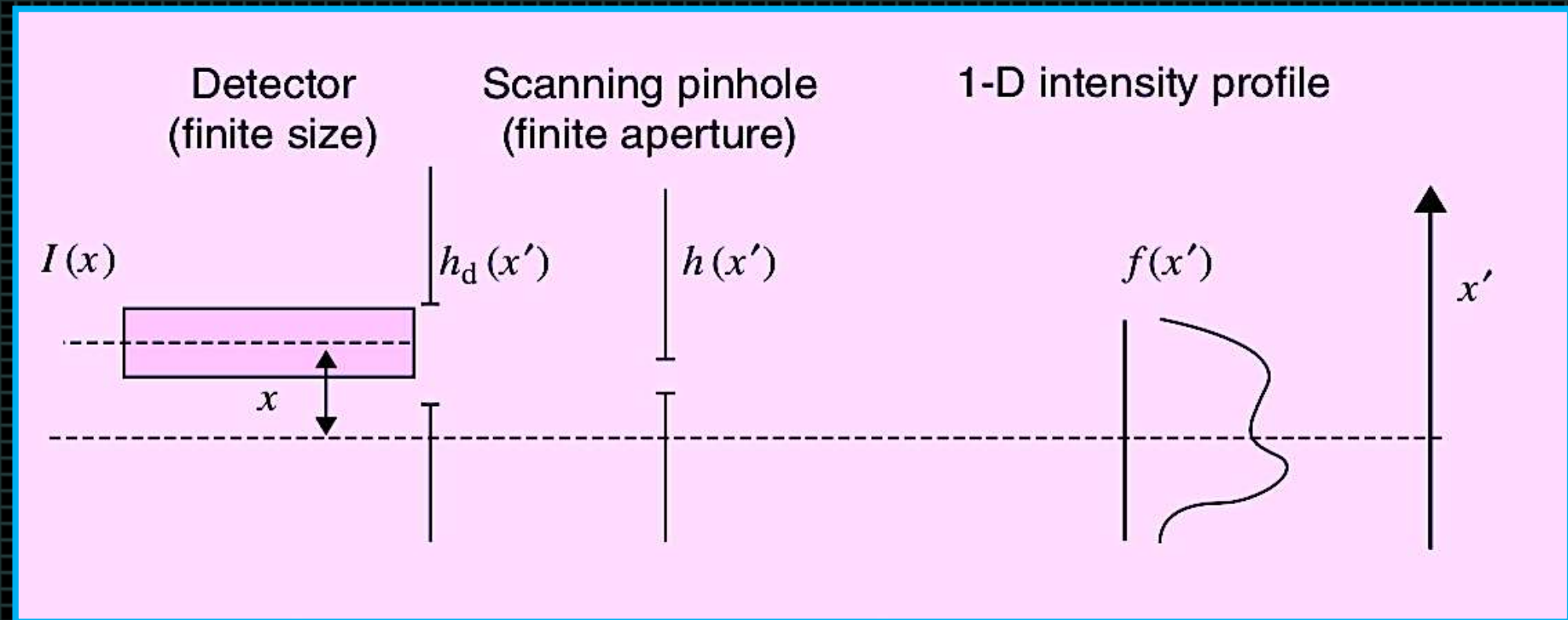




$$g(x) = \int f(x') h(x, x') dx' = \int f(x') \delta(x - x') dx' = f(x)$$

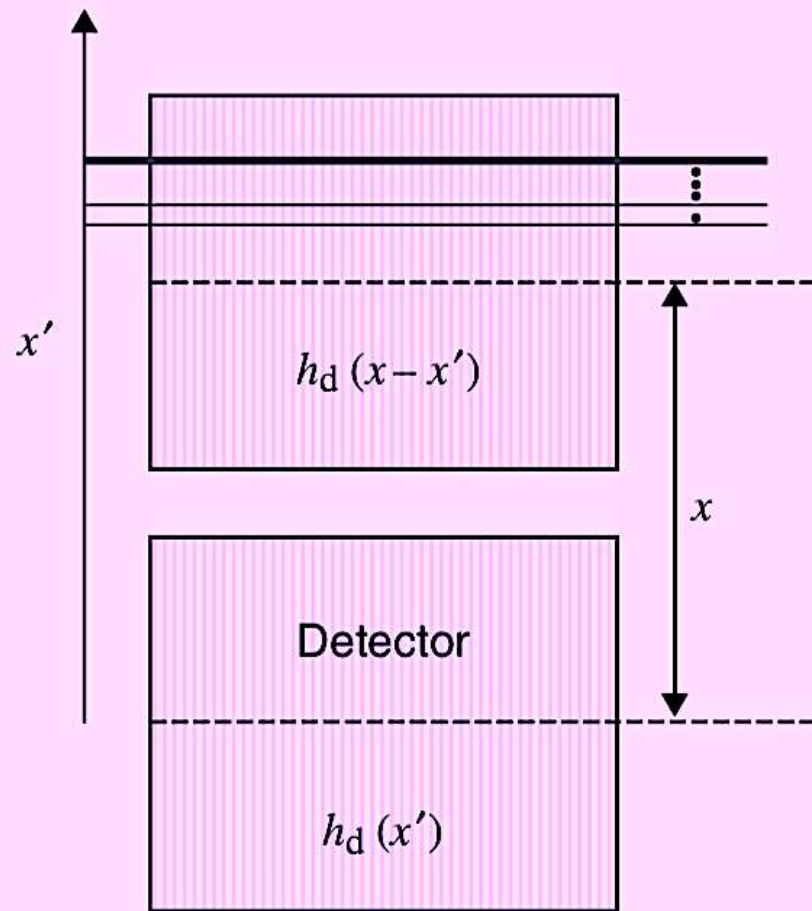


$$g(x) = \int f(x') h(x, x') dx' = \int f(x') h(x - x') dx'$$



$$I(x) = \int g(x') h_d(x, x') dx' \quad \text{where} \quad g(x) = \int f(x') \delta(x - x') dx'$$

$$I(x) = f(x') * h(x,) h_d(x)$$

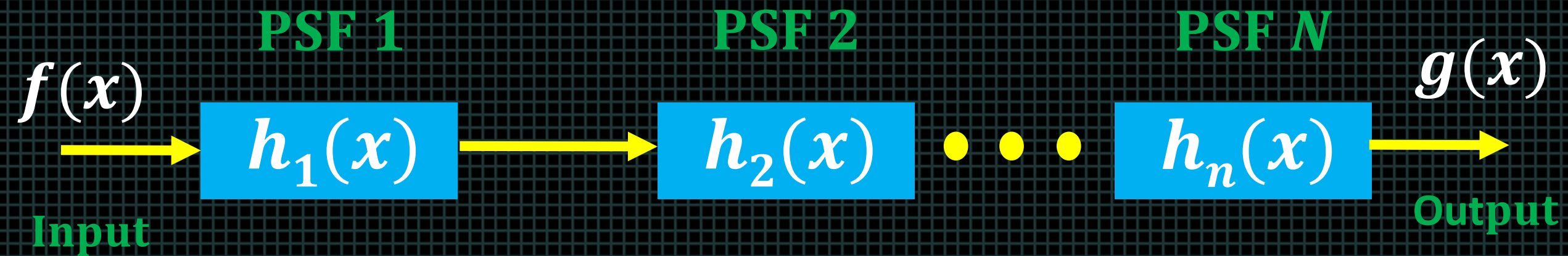


$$dI = g(x') h(x-x')$$

Contribution dI to the total intensity over the detector at point x' is the product of the incident intensity g and the response at that point h

The total recorded intensity is the sum of all such contributions over the detector-

$$I(x) = \int dI = \int_{-\infty}^{\infty} g(x') h(x-x') dx'$$



$$g(x) = f(x) * h_1(x) * h_2(x) \dots * h_n(x)$$

w_1	w_2	w_3
w_4	w_5	w_6
w_7	w_8	w_9

=

-1	-1	-1
-1	8	-1
-1	-1	-1

12	11	12	13	13	9
10	8	10	11	8	13
32	36	40	35	42	40
40	37	38	36	46	41
41	36	89	39	42	39
42	37	39	43	45	38

$$f_i = \sum_{k=1}^9 w_k I_k(i)$$

$$= (-1 \times 10) + (-1 \times 11) + (-1 \times 8) + (-1 \times 40) + (8 \times 35) + (-1 \times 42) + (-1 \times 38) + (-1 \times 36) + (-1 \times 46) = 14$$

$$g_j = \sum_i f_{ij} h_j$$

$$g_{kl} = \sum_j \sum_i f_{ij} h_{kj}$$

Understanding the engineering of image formation

The Pixel

Pixel : Picture element

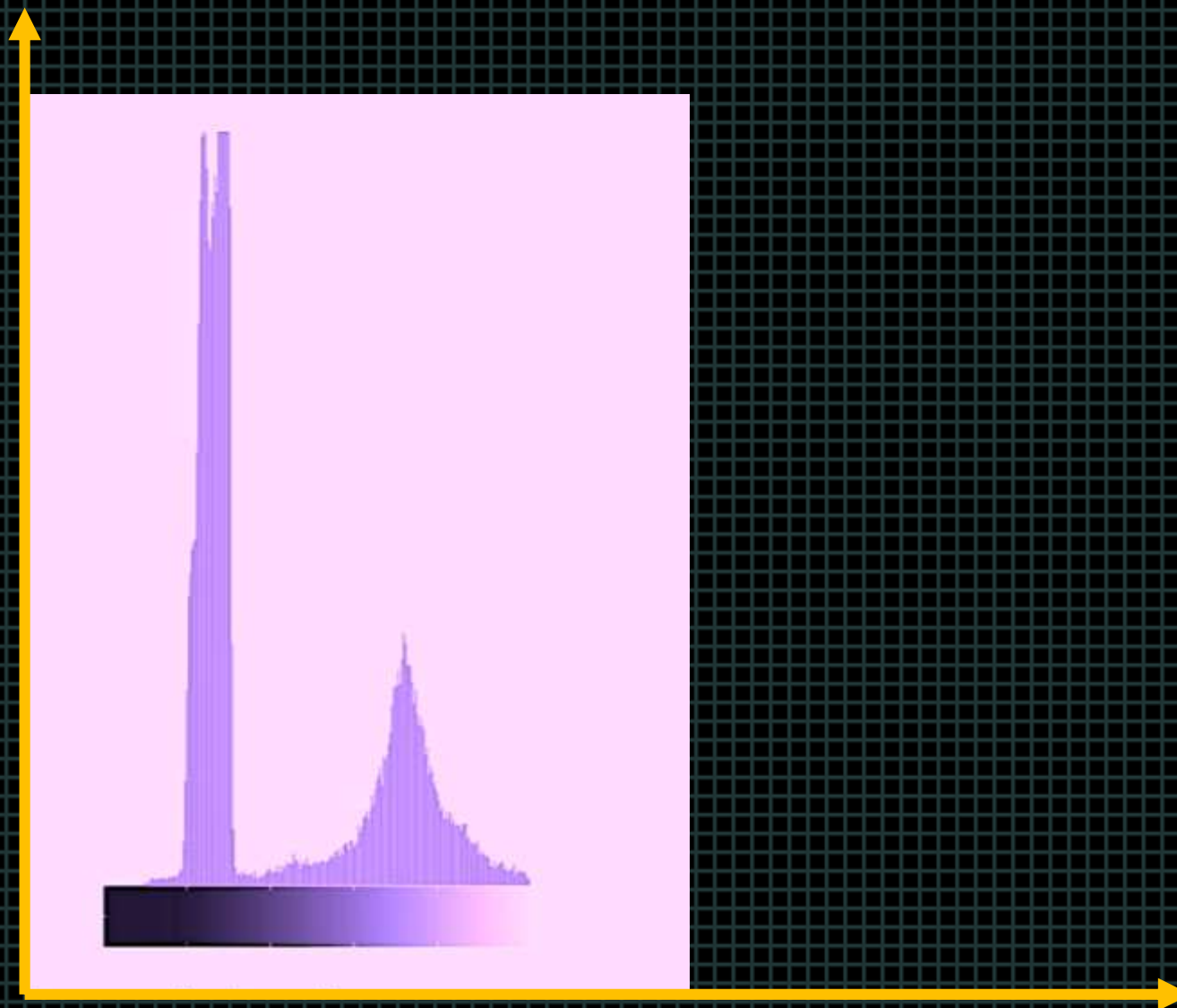
Histogram and Equalization

Introduction to Histograms

An image histogram is a plot of the relative frequency of occurrence of each of the permitted pixel values in the image against the values themselves.

A discrete probability density function which defines the likelihood of a given pixel value occurring within the image.

**Number of times each value
actually occurs within the
particular image**



**Range of values within the image
(0–255 for 8-bit grayscale)**

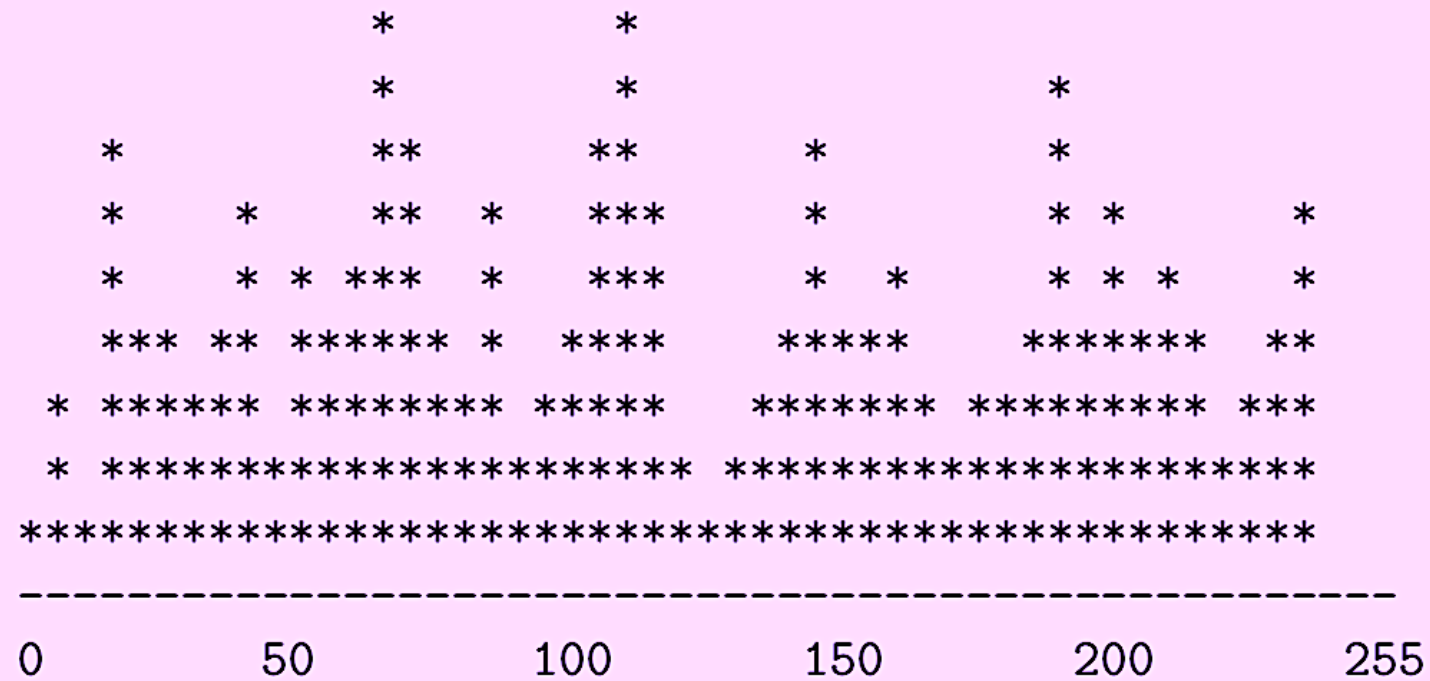
Gray-scale value : (0 -255)

```
initialize all histogram array entries to 0  
  
for each pixel I(i, j) within the image I  
    histogram(I(i,j)) = histogram(I(i,j)) + 1  
end
```

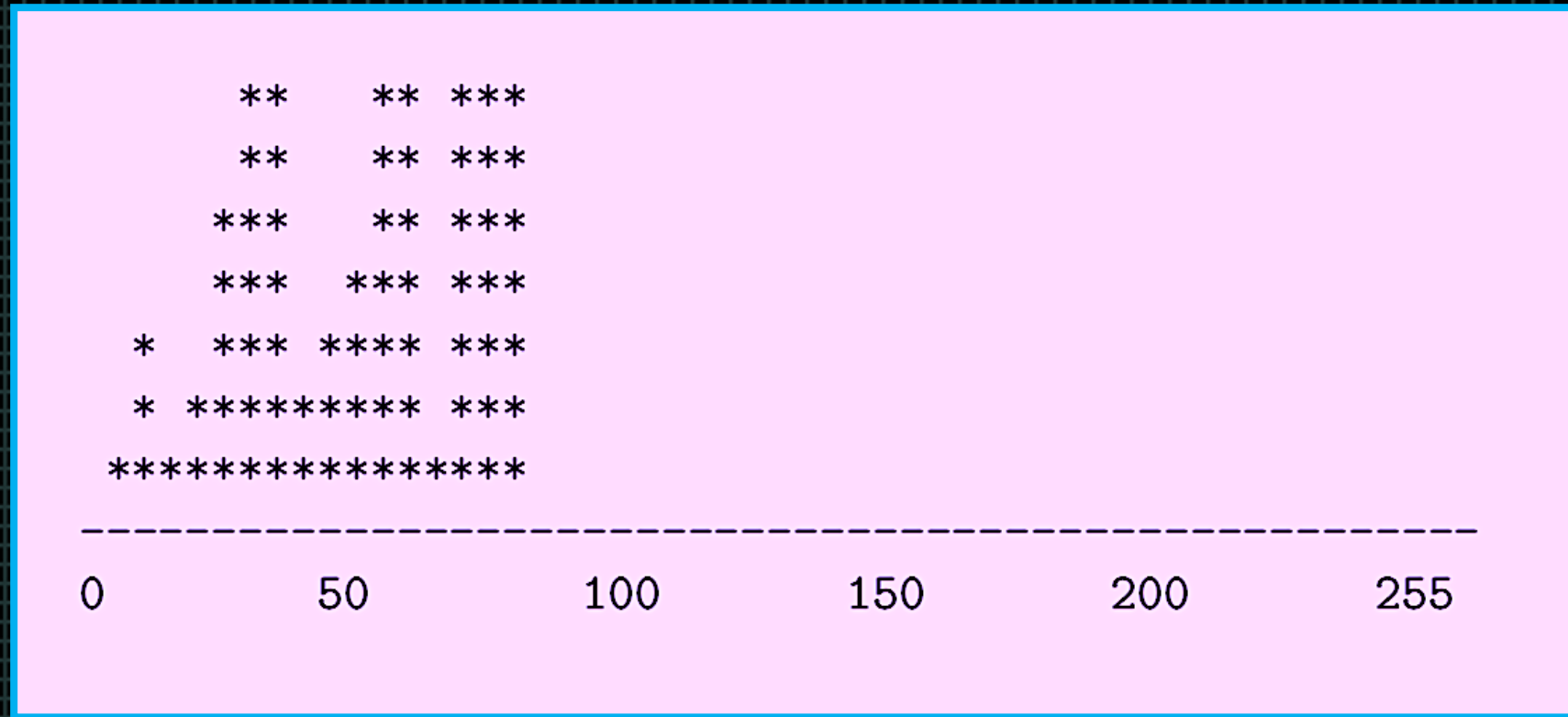

- Basic contrast present in the image
- Potential differences in the color distribution of the image foreground and background scene components.

A histogram uses a bar graph to profile the occurrences of each gray level present in an image

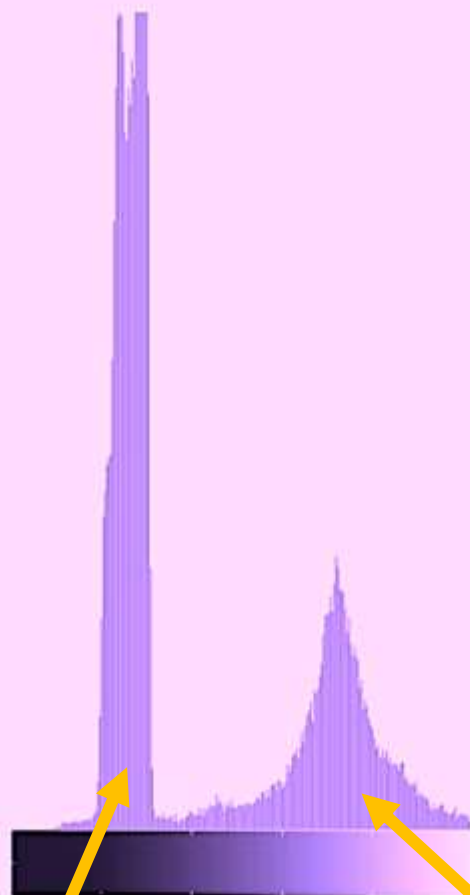
Histogram of a simple gray scale image



Simple Histogram

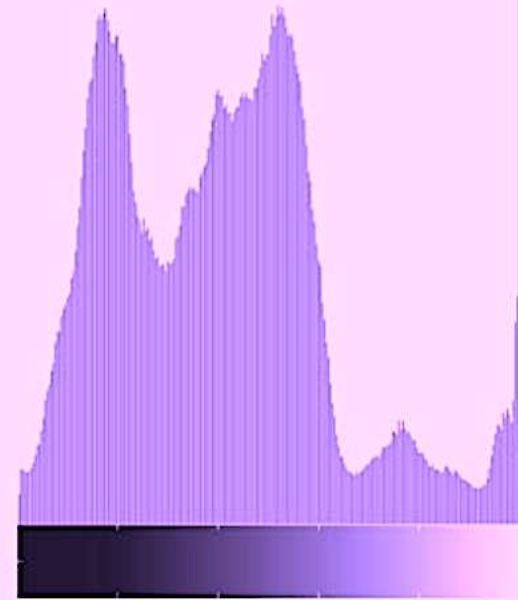


Histogram of a poorly scanned image



**Background
intensity**

Foreground



Histogram of a more complex scene

Introduction to Histogram Equalization

Image with poor contrast



Equalization causes a histogram with bins (vertical lines) grouped closely together to “spread out” into a flat or equalized histogram.

Image with poor contrast



$$b(x, y) = f[c(x, y)]$$

c : image with poor histogram

b : new image with improved histogram

f : transformation function

$$p1(a) = \frac{1}{Area_1} H_1(a)$$

$p1(a)$: probability of finding a pixel with the value “a”
in the image

$Area_1$: area or number of pixels in the image

$H_1(a)$: histogram of the image

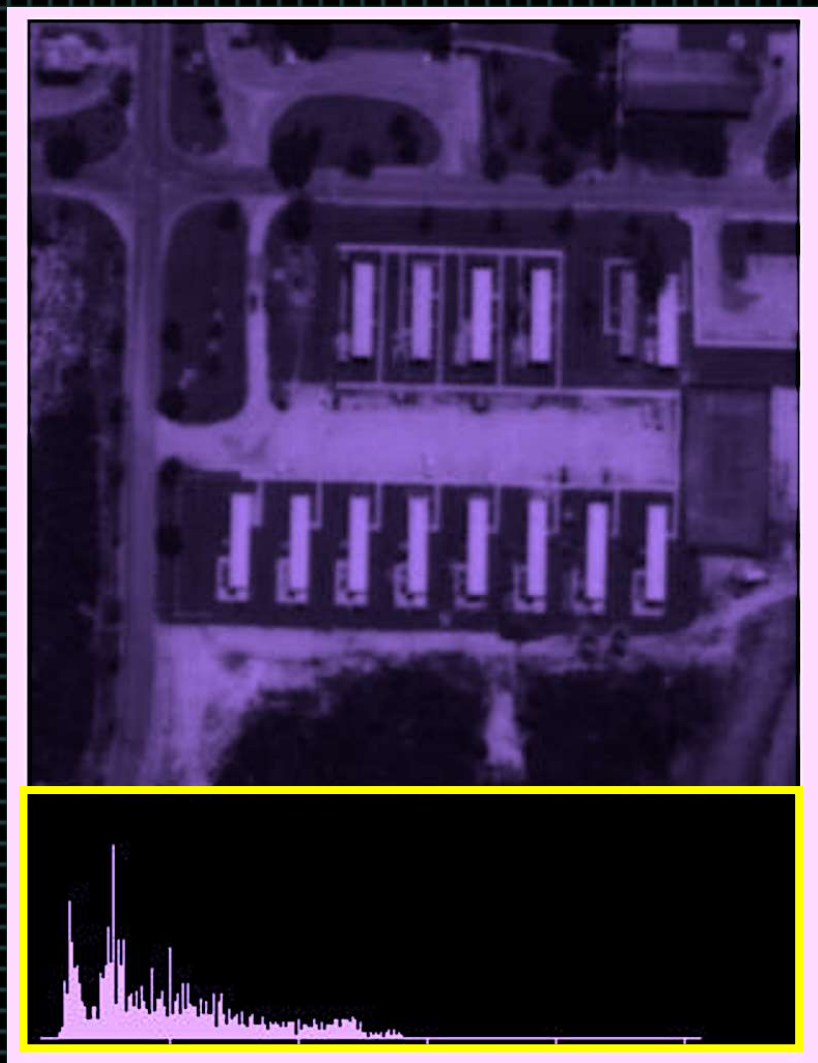
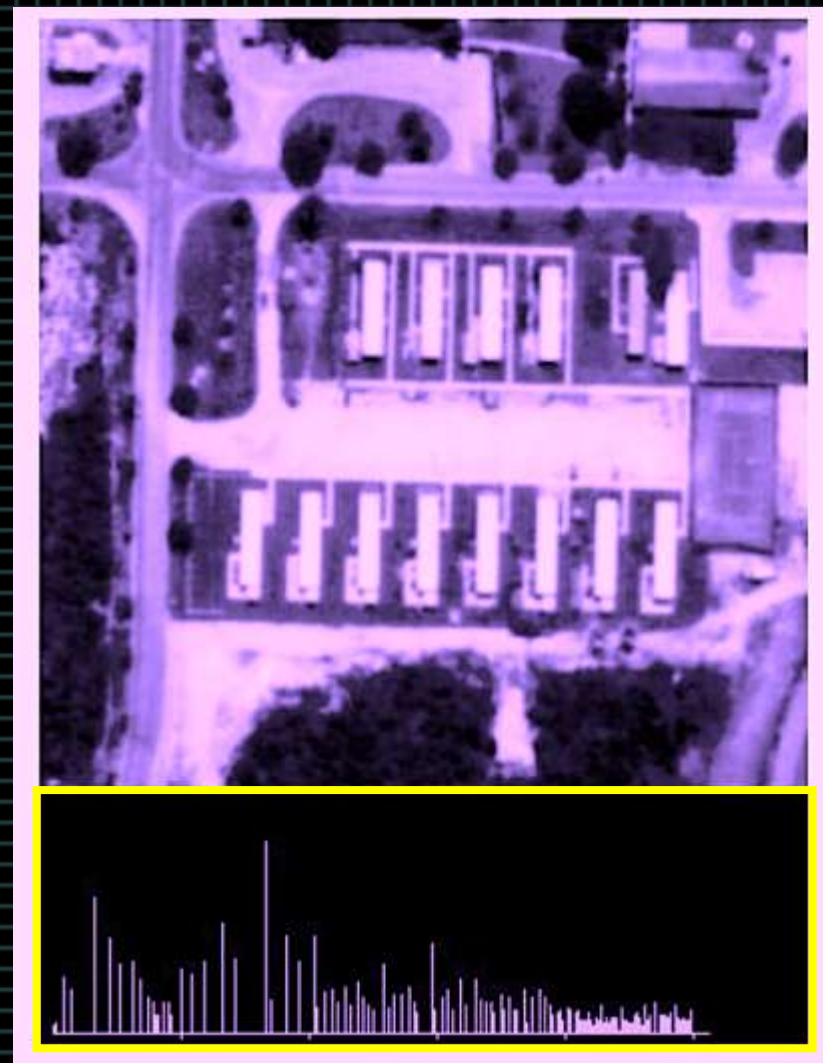
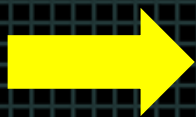
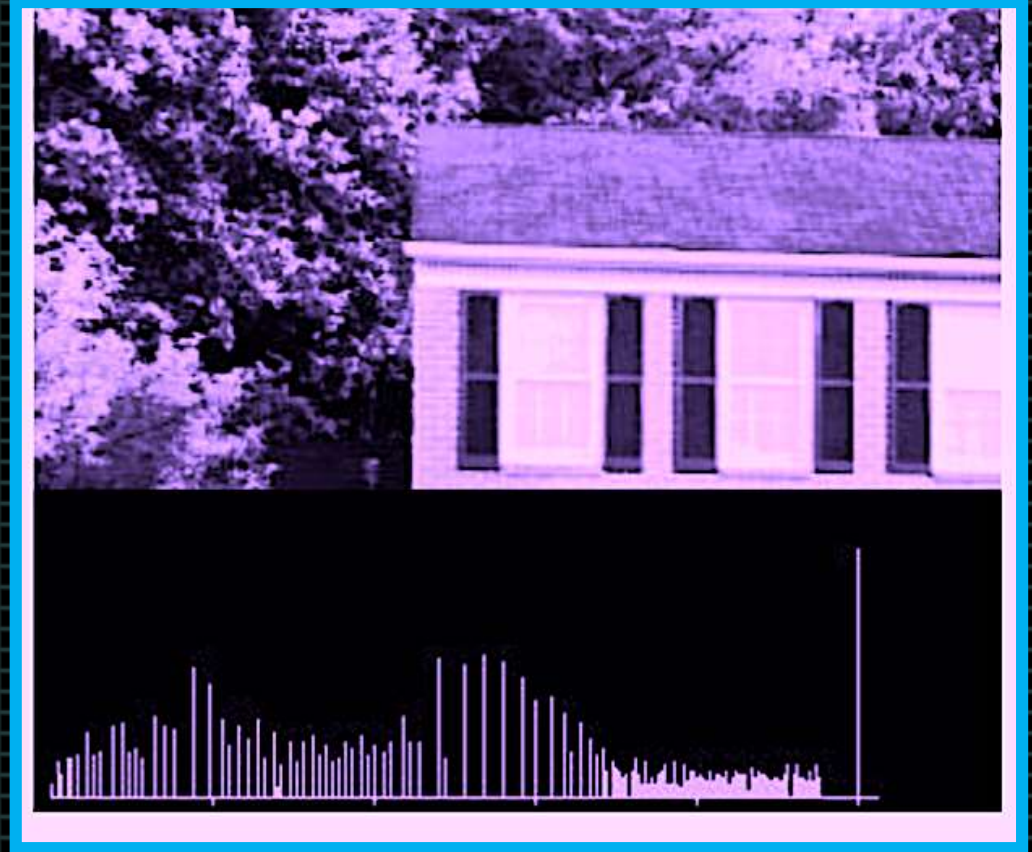
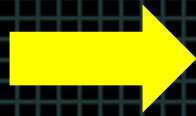
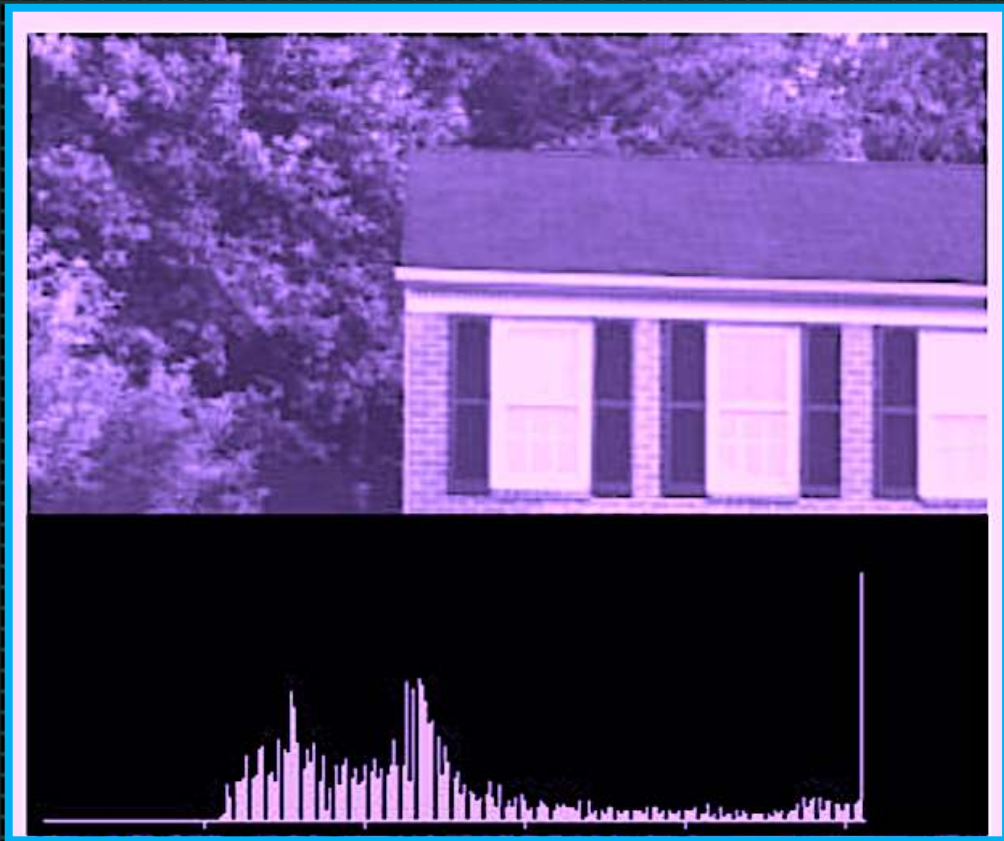


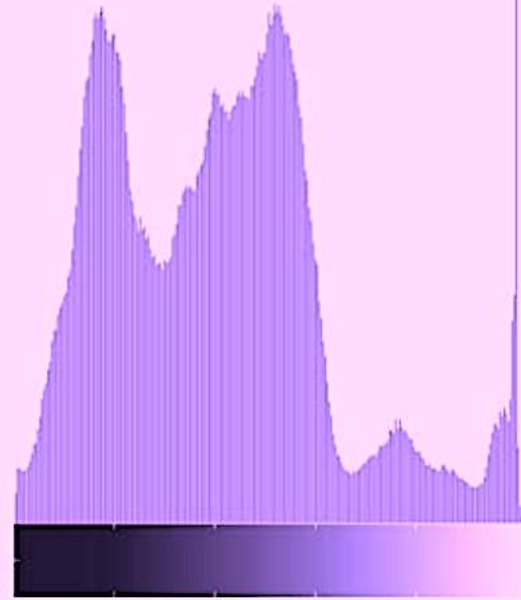
Image with poor contrast



Equalized image



Histogram for threshold selection



Histogram of a more complex scene

Adaptive thresholding

- Different threshold at each pixel location .

$$t = \text{mean} + C$$

$$t = \text{median} + C$$

$$\text{floor}\left(\frac{\text{max} - \text{min}}{2}\right) + C$$

Image Enhancement Techniques

Enhancement by filtering

Spatial domain
filtering

Filtering on the actual pixel
rather than in the frequency domain

Image Filters

```
graph TD; A[Image Filters] -.-> B[Linear Filters]; A -.-> C[Non Linear Filters];
```

Linear Filters

Non Linear Filters

Connectivity

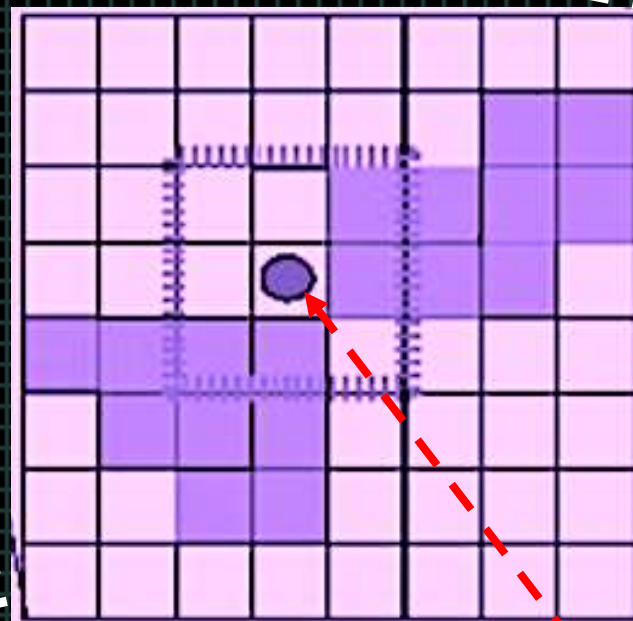
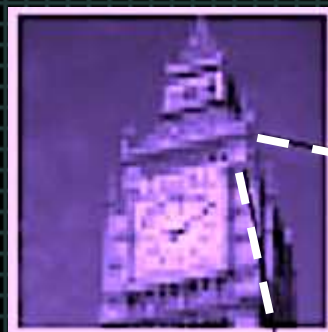
Deciding which pixels are connected to each other

E.g.

4 – connectivity : N, W, E, S

8 – connectivity : N, NW, W, NE,
SE, E, SW, S

NW	N	NE
W	(i,j)	E
SW	S	SE



*3 X 3 neighborhood
centered at this point*

The Filter Kernel

- Also known as the *mask*

w_1	w_2	w_3
w_4	w_5	w_6
w_7	w_8	w_9

=

-1	-1	-1
-1	8	-1
-1	-1	-1

12	11	12	13	13	9
10	8	10	11	8	13
32	36	40	35	42	40
40	37	38	36	46	41
41	36	89	39	42	39
42	37	39	43	45	38

$$f_i = \sum_{k=1}^9 w_k I_k(i)$$

$$= (-1 \times 10) + (-1 \times 11) + (-1 \times 8) + (-1 \times 40) + (8 \times 35) + (-1 \times 42) + (-1 \times 38) + (-1 \times 36) + (-1 \times 46) = 14$$

Filtering with an $N \times N = 3 \times 3$ kernel

Row and Column Indices

$$f(x, y) = \sum_{i=I_{min}}^{I_{max}} \sum_{j=J_{min}}^{J_{max}} w(i, j) I(x + i, y + j)$$

$i = 0, j = 0$: *center pixel of the kernel*

$(I_{max} - I_{min} + 1, J_{max} - J_{min} + 1)$: *size of kernel center pixel*

Linear Indices

$$f_i = \sum_{k=1}^N w_k I_k(i)$$

$I_k(i)$: *neighborhood pixel of the i th image pixel*
 k : *a linear index running over the neighborhood*

w_1	w_2	w_3
w_4	w_5	w_6
w_7	w_8	w_9

=

-1	-1	-1
-1	8	-1
-1	-1	-1

12	11	12	13	13	9
10	8	10	11	8	13
32	36	40	35	42	40
40	37	38	36	46	41
41	36	89	39	42	39
42	37	39	43	45	38

$$f_i = \sum_{k=1}^9 w_k I_k(i)$$

$$= (-1 \times 10) + (-1 \times 11) + (-1 \times 8) + (-1 \times 40) + (8 \times 35) + (-1 \times 42) + (-1 \times 38) + (-1 \times 36) + (-1 \times 46) = 14$$

Filtering with an $N \times N = 3 \times 3$ kernel

Linear Filtering Steps

Define filter kernel



Slide kernel over image



Multiply pixels under the kernel by weights



Copy resulting value to same location in new image

Geometric Operations

Overview of Geometric Operations

- **Modify geometry of an image by repositioning pixels**

Rotate

Flip

Crop

Resize

Common uses of geometric operations

- **Correcting geometric distortions**
- **Creating special effects**
- **As part of image registration**

Geometric
Operations

```
graph TD; A[Geometric Operations] --> B[Mapping Functions]; A --> C[Interpolation Methods];
```

Mapping Functions

Interpolation Methods

Mapping and Affine Transformation

$$f(x, y) \rightarrow g(x', y')$$

Mapping function:

$$(x', y') = T(x, y)$$

$$x' = T_x(x, y)$$

$$y' = T_y(x, y)$$

Where:

$T_x, T_y =$ polynomials in x and y

Affine transformation:

$$x' = a_0x + a_1y + a_2$$

$$y' = b_0x + b_1y + b_2$$

$$\begin{aligned}x' &= a_0x + a_1y + a_2 \\ y' &= b_0x + b_1y + b_2\end{aligned}$$

$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} a_0 & a_1 & a_2 \\ b_0 & b_1 & b_2 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

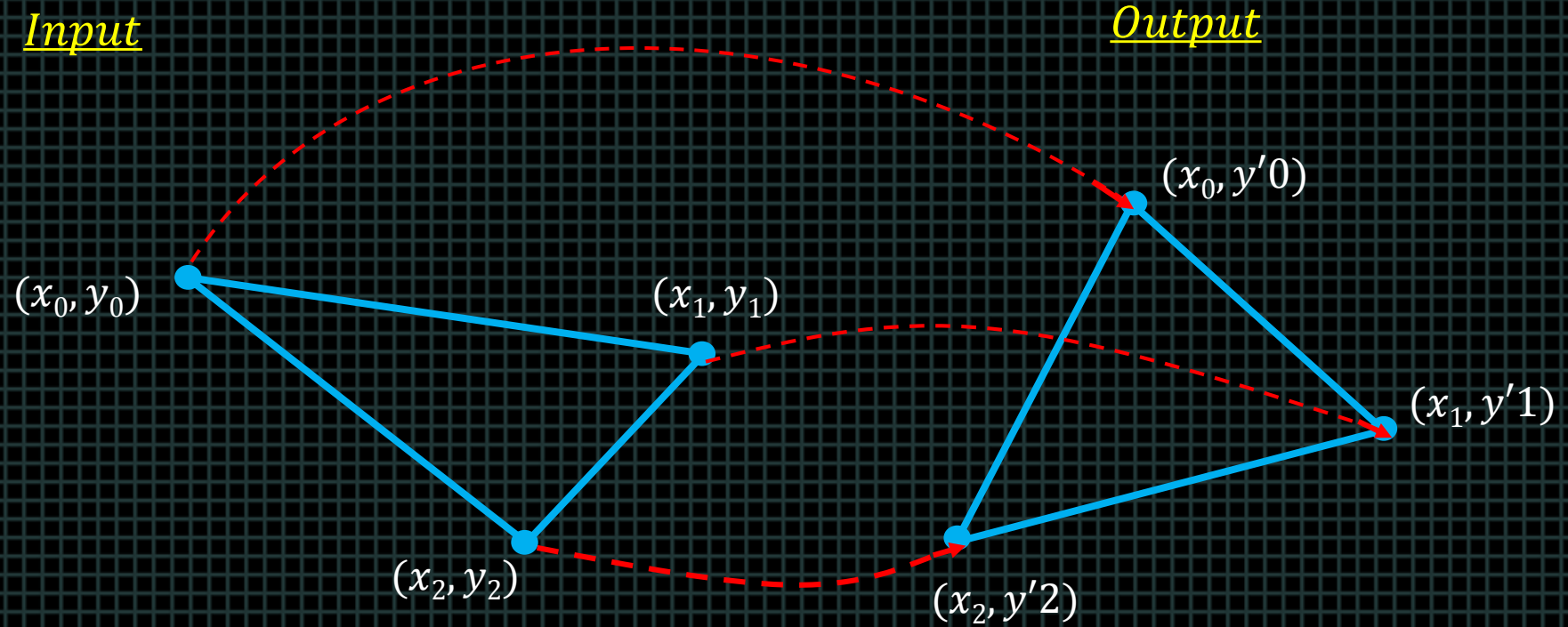
Summary of Affine Transformation Coefficients

Transformation	a_0	a_1	a_2	b_0	b_1	b_2
Translation by Δ_x, Δ_y	1	0	Δ_x	0	1	Δ_y
Scaling by a factor $[s_x, s_y]$	s_x	0	0	0	s_y	0
Counterclockwise rotation by angle θ	$\cos \theta$	$\sin \theta$	0	$-\sin \theta$	$\cos \theta$	0
Shear by a factor $[sh_x, sh_y]$	1	sh_y	0	sh_x	1	0

E.g.:

Generate the affine transformation matrix for :

- *rotation by 30°*
- *scaling by a factor 3.5 in both dimensions*
- *translation by $[25, 15]$ pixels*
- *Shear by a factor $[2, 3]$*



Mapping one triangle onto another by
an affine transformation.

$$\cos 30^\circ = 0.866 \text{ and } \sin 30^\circ = -0.500$$

rotation by 30°

(a) $= \begin{bmatrix} 0.866 & -0.500 & 0 \\ 0.500 & 0.866 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

scaling by a factor 3.5,3.5

(b) $= \begin{bmatrix} 3.5 & 0 & 0 \\ 0 & 3.5 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

translation by [25,15] pixels

(c) $= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 25 & 15 & 1 \end{bmatrix}$

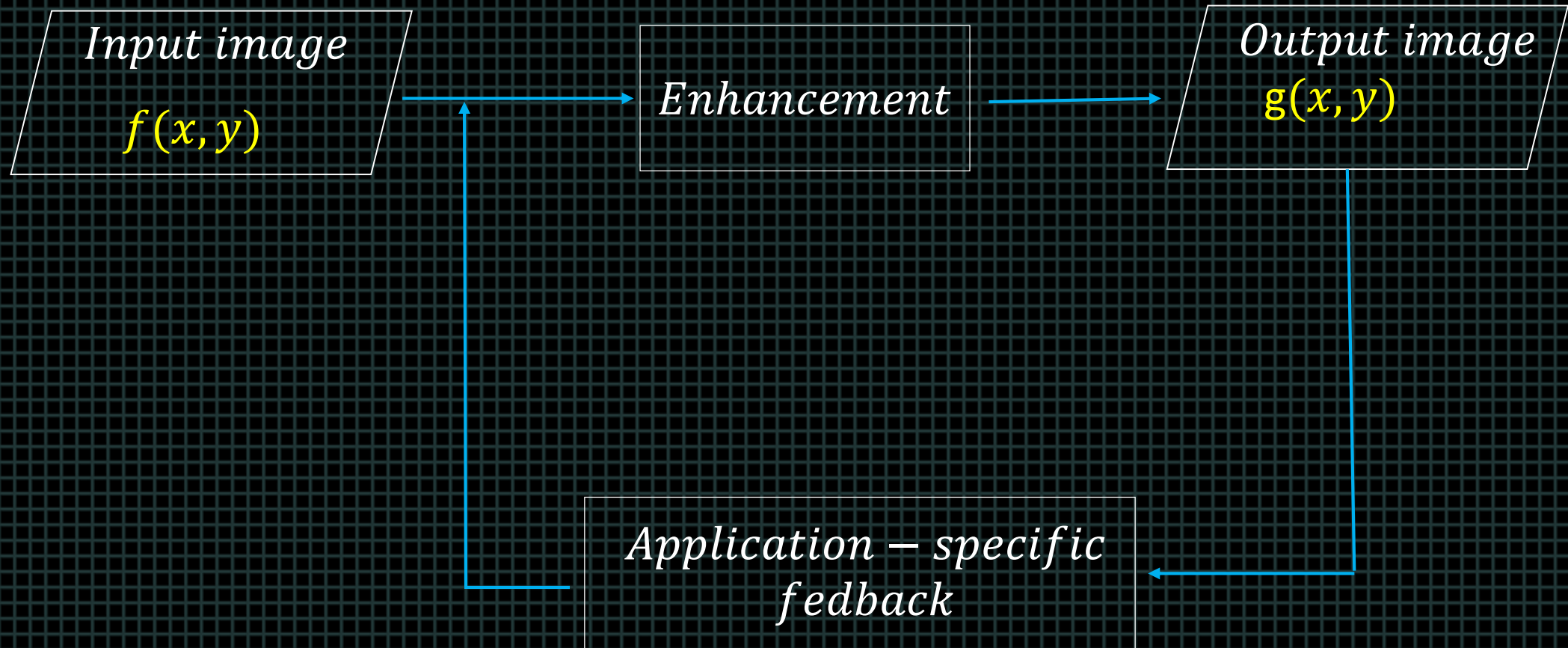
Shear by a factor [2,3]

(d) $= \begin{bmatrix} 1 & 3 & 0 \\ 2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

Transformation	a_0	a_1	a_2	b_0	b_1	b_2
Translation by Δ_x, Δ_y	1	0	Δ_x	0	1	Δ_y
Scaling by a factor $[s_x, s_y]$	s_x	0	0	0	s_y	0
Counterclockwise rotation by angle θ	$\cos \theta$	$\sin \theta$	0	$-\sin \theta$	$\cos \theta$	0
Shear by a factor $[sh_x, sh_y]$	1	sh_y	0	sh_x	1	0

$$\begin{bmatrix} a_0 & a_1 & a_2 \\ b_0 & b_1 & b_2 \\ 0 & 0 & 1 \end{bmatrix}$$

Gray - Level Transformations



Overview of gray-level transformations

Spatial transformation :

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

where :
 $g(x, y)$: *processed image*
 $f(x, y)$: *original image*
 T : *operator on $f(x, y)$*

where :
 $s = T[r]$
 r : *original gray level of a pixel*
 s : *resulting gray level of a pixel*

$$s = c \cdot r + b$$

where :
r: origian pixel value
s: resulting pixel value
c: constant for controlling the contrast of output image

Neighborhood Processing

Steps:

- Define a reference point in the input image
- Perform reference point neighborhood operation in input image
- Apply result of operation to the pixel of same coordinates in the output image.
- Repeat the process for every pixel in the input image

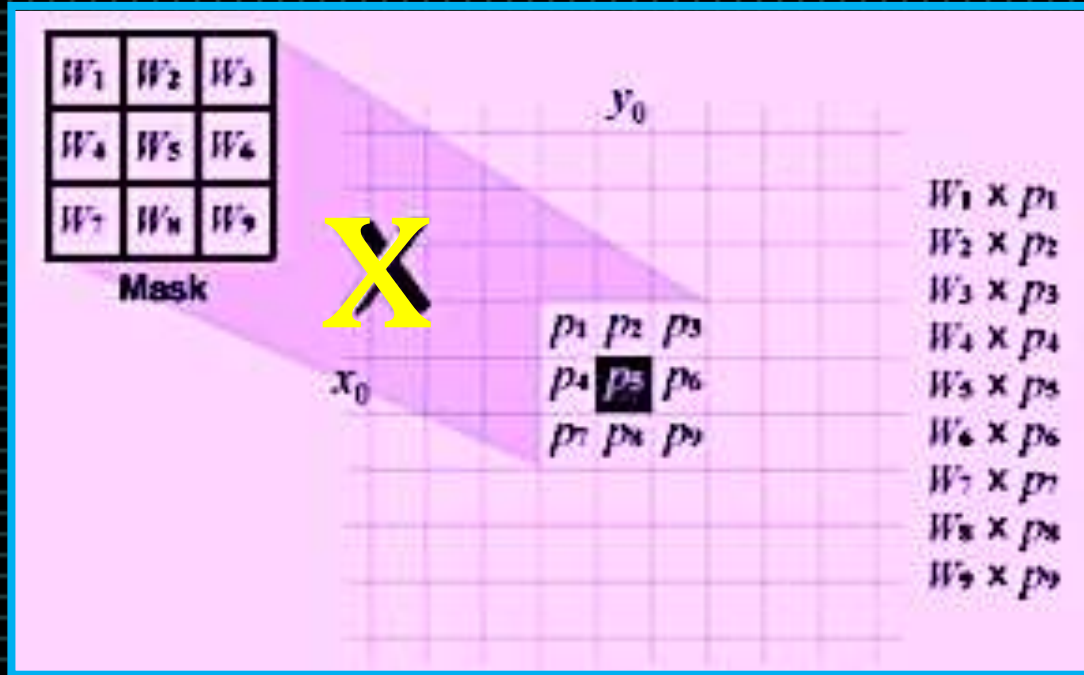


Linear Filters

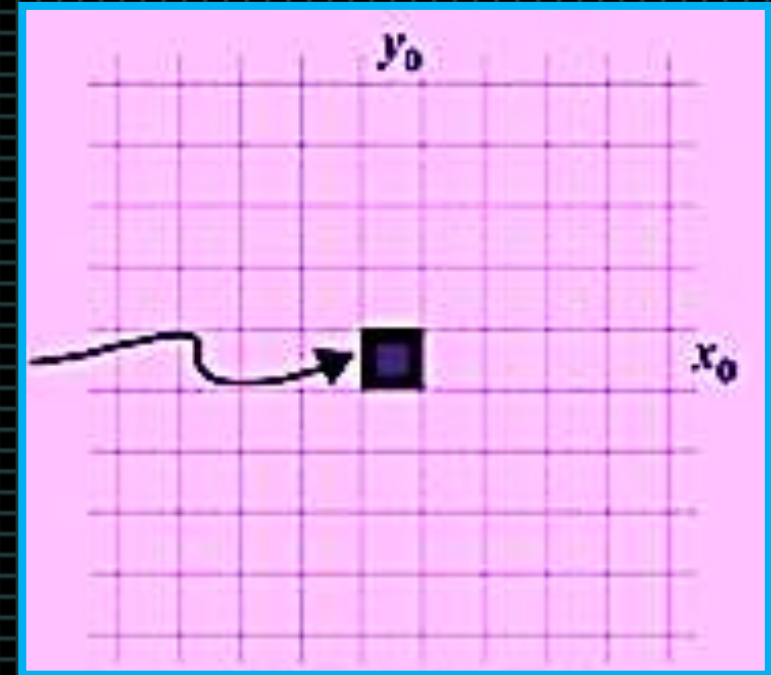
A diagram on a dark blue grid background. Two white rectangular boxes are positioned horizontally. The left box contains the text 'Linear Filters' and the right box contains 'Nonlinear Filters'. A dashed white line starts from the top-right corner of the left box, extends diagonally upwards to a central point, and then extends diagonally downwards to the top-left corner of the right box, forming a peak shape.

Nonlinear Filters

Convolution and Correlation



Σ



Neighborhood processing : linear filtering

Convolution

$$A * B = \sum_{j=-\infty}^{\infty} A(j) \bullet B(x - j)$$

Example :

Let : $A = \{0, 1, 2, 3, 1, 0\}$ $B = \{1, 3, -1\}$

Solution

$$(0 \times (-1)) + (0 \times 3) + (1 \times 1) = 1$$

A		0	1	2	3	2	1	0
B	-1	3	1					
A * B		1						

Array B mirrored and center value aligned with 1st value of array A

$$(0 \times (-1)) + (1 \times 3) + (2 \times 1) = 5$$

A	0	1	2	3	2	1	0
B	-1	3	1				
A * B	1	5					

Array B shifted one position to the right

$$(1 \times (-1)) + (2 \times 3) + (3 \times 1) = 8$$

A	0	1	2	3	2	1	0
B		-1	3	1			
A * B	1	5	8				

Array B shifted another position to the right

$$(1 \times (-1)) + (2 \times 3) + (3 \times 1) = 8$$

A	0	1	2	3	2	1	0
B		-1	3	1			
A * B	1	5	8				

Array B shifted another position to the right

$$(2 \times (-1)) + (3 \times 3) + (2 \times 1) = 9$$

A	0	1	2	3	2	1	0
B			-1	3	1		
A * B	1	5	8	8			

Array B shifted another position to the right

$$(3 \times (-1)) + (2 \times 3) + (1 \times 1) = 4$$

<i>A</i>	0	1	2	3	2	1	0
<i>B</i>				-1	3	1	
<i>A</i> * <i>B</i>	1	5	8	8	4		

Array B shifted another position to the right

$$(2 \times (-1)) + (1 \times 3) + (0 \times 1) = 1$$

A	0	1	2	3	2	1	0
B					-1	3	1
A * B	1	5	8	8	4	1	

Array B shifted another position to the right

$$(1 \times (-1)) + (0 \times 3) + (0 \times 1) = -1$$

A	0	1	2	3	2	1	0	
B						-1	3	1
$A * B$	1	5	8	8	4	1	-1	

Array B shifted another position to the right

$$A = \{0, 1, 2, 3, 1, 0\}$$

$$B = \{1, 3, -1\}$$

$$A * B = \{1 \quad 5 \quad 8 \quad 8 \quad 4 \quad 1 \quad -1\}$$

2-D Convolution and Correlation

$$g(x, y) = \sum_{k=-\infty}^{\infty} \sum_{j=-\infty}^{\infty} h(j, k) \bullet f(x - j, y - k)$$

OR

$$g(x, y) = \sum_{k=-n_2}^{n_2} \sum_{j=-m_2}^{m_2} h(j, k) \bullet f(x - j, y - k)$$

m_2 = half of mask's width

n_2 = half of mask's height

$$m_2 = [m/2] \qquad n_2 = [n/2]$$

A =

B =

-2×0	-1×0	0×0
-1×0	1×5	1×8
0×0	1×5	2×2

$$A * B =$$

$$(-2x^0) + (-1x^0) + (0x^0) + (-1x^0) + (1x^5) + (1x^8) + (0x^0) + (1x^5) + (2x^2)$$



Correlation

$$A \odot B = \sum_{j=-\infty}^{\infty} A(j) \bullet B(x+j)$$

2-D Correlation

$$g(x, y) = \sum_{k=-\infty}^{\infty} \sum_{j=-\infty}^{\infty} h(j, k) \bullet f(x+j, y+k)$$

$$g(x, y) = \sum_{k=-n_2}^{n_2} \sum_{j=-m_2}^{m_2} h(j, k) \bullet f(x+j, y+k)$$

m_2 = half of mask's width
 n_2 = half of mask's height
 $m_2 = \lceil m/2 \rceil$

Image Smoothing (Lowpass Filtering)

- Attenuates high frequency components
- Attenuates fine details in image
- Preserve coarse details in image and homogeneous areas

Mean Filter

- Known as *neighborhood averaging* or *spatial smoothing filter*.

$$h(x, y) = \begin{bmatrix} 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \end{bmatrix} = \frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

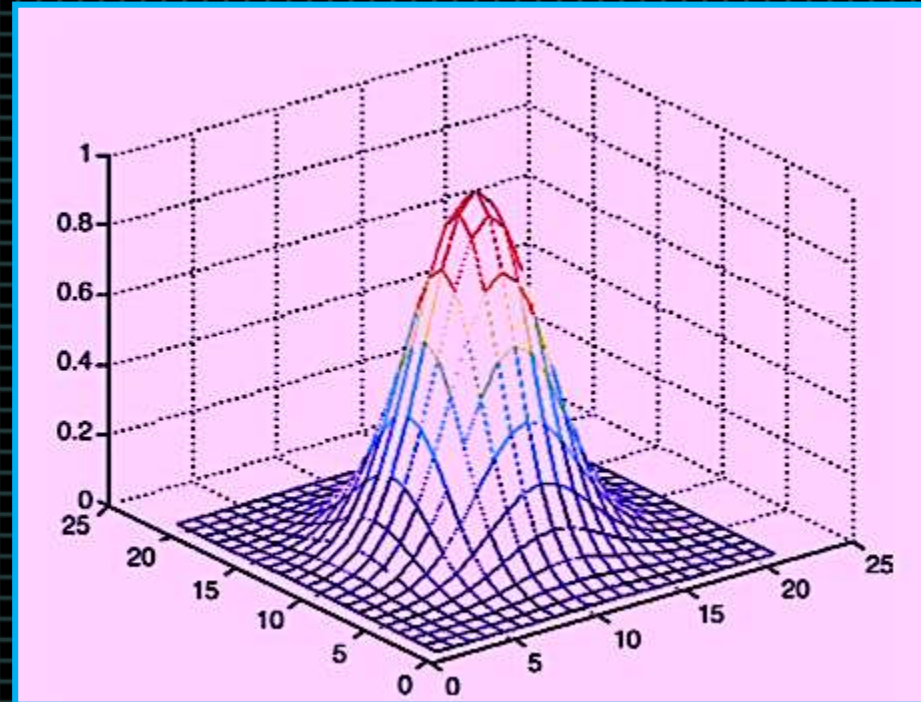
Modified

$$h(x, y) = \begin{bmatrix} 0.075 & 0.125 & 0.075 \\ 0.125 & 0.2 & 0.125 \\ 0.075 & 0.125 & 0.075 \end{bmatrix}$$

Gaussian Blur Filter

$$h(x, y) = \exp \left[-\frac{(x^2 + y^2)}{2\sigma^2} \right]$$

σ = controls the overall shape of the curve

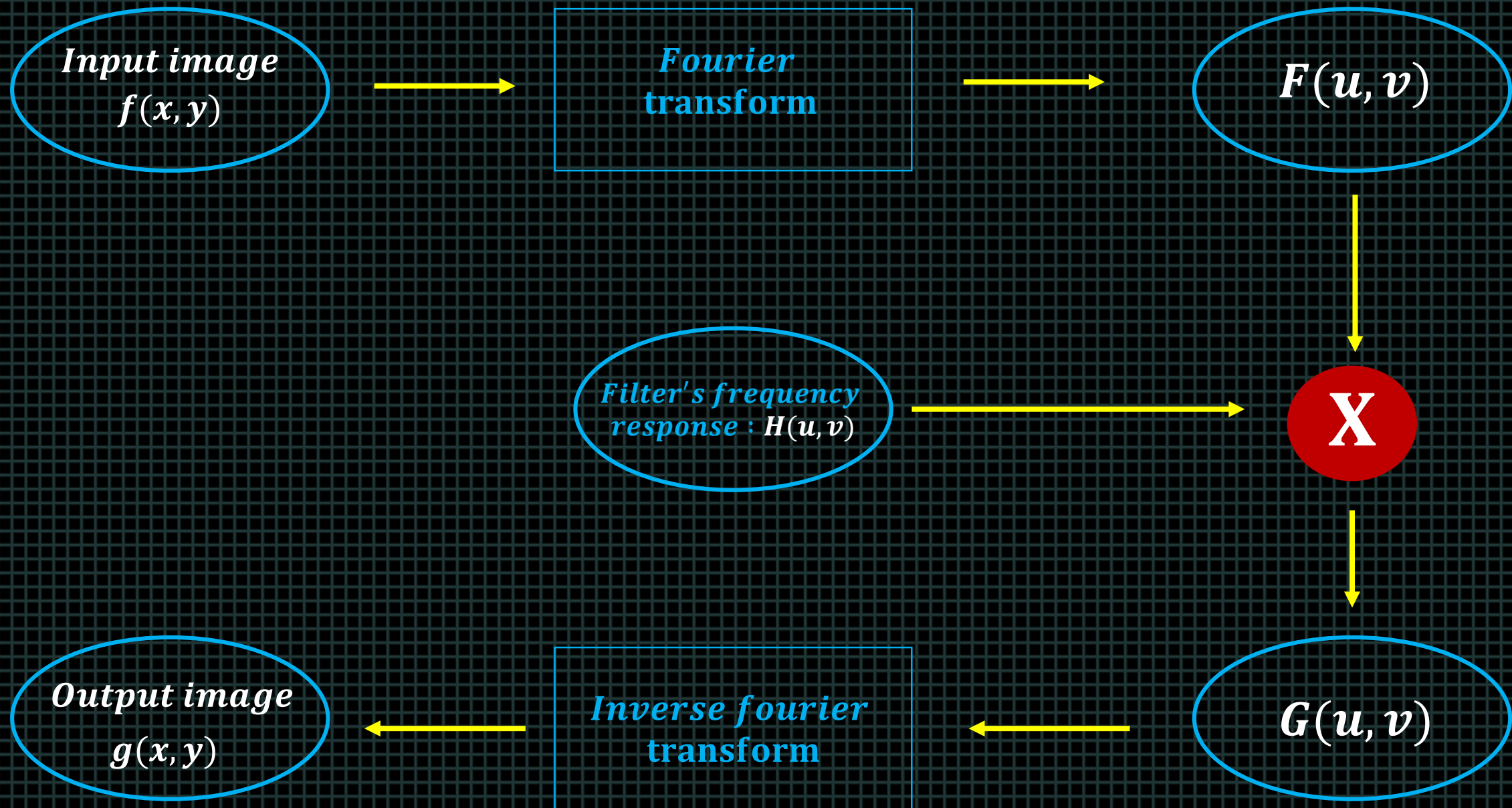


2D Gaussian function (with $\sigma = 3$)

Image Sharpening (High-pass Filtering)

- Attenuates low frequency components
- Attenuates coarse details in image
- Preserve fine details in an image and homogeneous areas

Fourier Transform and Frequency Domain processing.

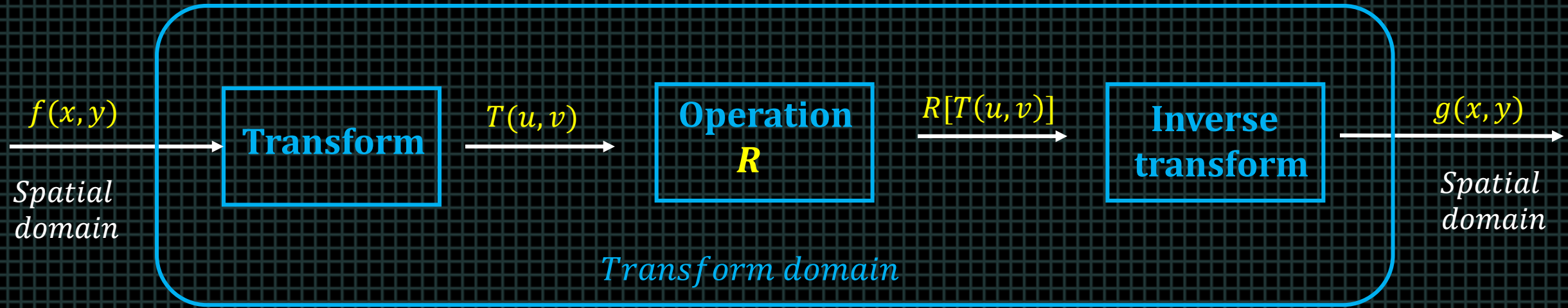


$$g(x, y) = f(x, y) * h(x, y)$$

$$G(u, v) = F(u, v) H(u, v)$$

where G, F , and $H \rightarrow$ Fourier transform of g, f and h respectively

Fourier Transform:



2D Forward Transform:

$$T(x, y) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) \bullet r(x, y, u, v)$$

where :

$$\left. \begin{array}{l} u = 0, 1, 2, \dots, M-1 \\ v = 0, 1, 2, \dots, N-1 \end{array} \right\} \text{transform variables}$$

$f(x, y)$: input image

$r(x, y, u, v)$: forward transformation kernel

2D Inverse Transform (Recovering original image)

$$f(x, y) = \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} T(u, v) \bullet s(x, y, u, v)$$

where :

$$x = 0, 1, 2, \dots, M - 1$$

$$y = 0, 1, 2, \dots, N - 1$$

$s(x, y, u, v)$: inverse transform kernel

Transform pair:

$$T(x, y) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) \bullet r(x, y, u, v)$$

$$f(x, y) = \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} T(u, v) \bullet s(x, y, u, v)$$



original image



after applying lpf

Ideal Low-pass Filtering

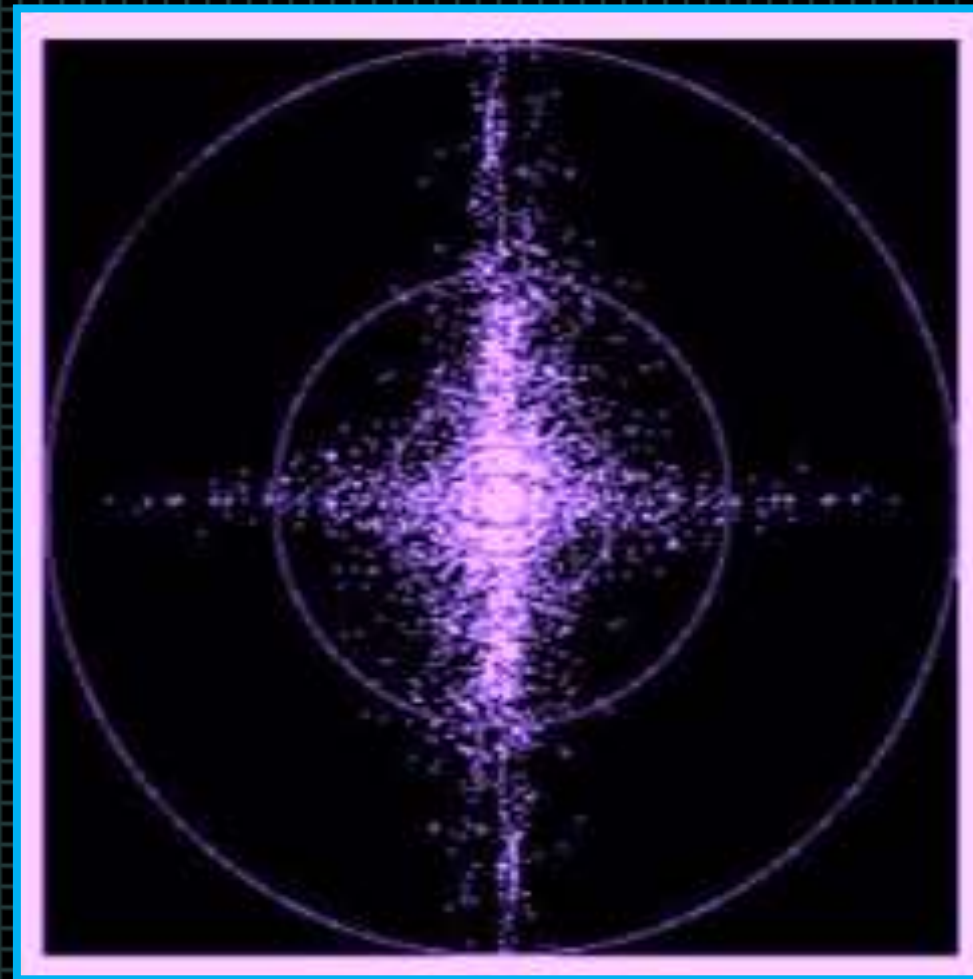
$$H_1(u, v) = \begin{cases} 1 & \text{if } D(u, v) \leq D_0 \\ 0 & \text{if } D(u, v) > D_0 \end{cases}$$

where :

$$D(u, v) = \sqrt{u^2 + v^2} \quad : \text{ distance btw a point of coordinates and the origin of the 2d frequency plot}$$



original image



fourier spectrum



8 pixels



16 pixels



32 pixels



64 pixels

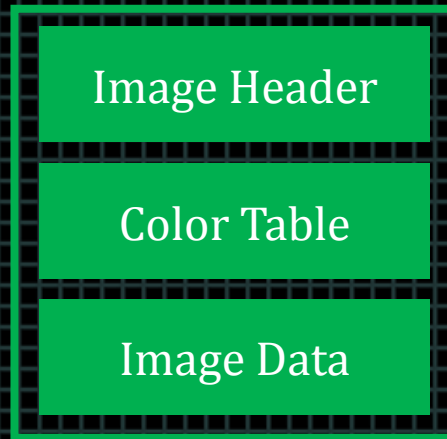


128 pixels

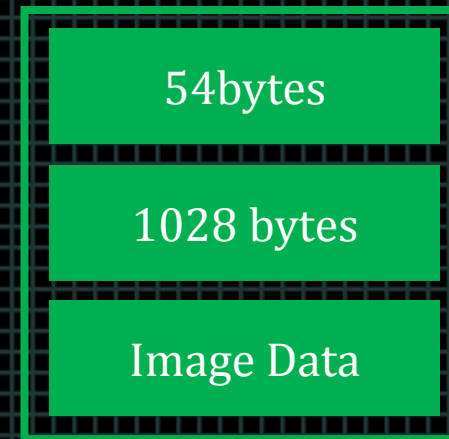
Projects

The bitmap Image

Image



BMP



BMP (Windows) Header Format

offset	size	description
0	2	signature, must be 4D42 hex
2	4	size of BMP file in bytes (unreliable)
6	2	reserved, must be zero
8	2	reserved, must be zero
10	4	offset to start of image data in bytes
14	4	size of BITMAPINFOHEADER structure, must be 40
18	4	image width in pixels
22	4	image height in pixels
26	2	number of planes in the image, must be 1
28	2	number of bits per pixel (1, 4, 8, or 24)
30	4	compression type (0=none, 1=RLE-8, 2=RLE-4)
34	4	size of image data in bytes (including padding)
38	4	horizontal resolution in pixels per meter (unreliable)
42	4	vertical resolution in pixels per meter (unreliable)
46	4	number of colors in image, or zero
50	4	number of important colors, or zero

Operators

Operator

Function that acts on elements of a set to produce other elements of the same space.

Prewitt Operator

- Used to detect vertical and horizontal edges

$$\begin{bmatrix} -1 & 0 & 1 \\ -1 & 0 & 1 \\ -1 & 0 & 1 \end{bmatrix}$$

Vertical Edges

$$\begin{bmatrix} -1 & -1 & -1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

Horizontal Edges

Sobel Operator

- Used to detect vertical and horizontal edges

$$\begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}$$

Vertical Edges

$$\begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$

Horizontal Edges

Kirsch Operator

- Also known as direction masks
- Edge detects in 8 compass directions

+5	+5	+5
-3	0	-3
-3	-3	-3

N

+5	+5	-3
+5	0	-3
-3	-3	-3

NW

+5	-3	-3
+5	0	-3
+5	-3	-3

W

-3	-3	-3
+5	0	-3
+5	+5	-3

SW

-3	-3	+5
-3	0	+5
-3	-3	+5

E

-3	+5	+5
-3	0	+5
-3	-3	-3

NE

-3	-3	-3
-3	0	-3
+5	+5	+5

S

-3	-3	+5
-3	0	+5
-3	+5	+5

SE

Robinson Operator

- Also known as direction masks
- Edge detects in 8 compass directions

-1	0	1	0	1	2	1	2	1	2	1	0
-2	0	2	-1	0	1	0	0	0	1	0	-1
-1	0	1	-2	-1	0	-1	-2	-1	0	-1	-2
<u>N</u>			<u>NW</u>			<u>W</u>			<u>SW</u>		
-1	-2	-1	-2	-1	0	1	0	-1	0	-1	-2
0	0	0	-1	0	1	2	0	-2	1	0	-1
1	2	1	0	1	2	1	0	1	2	1	0
<u>E</u>			<u>NE</u>			<u>S</u>			<u>SE</u>		

Laplacian Operator

- 2nd order derivative mask

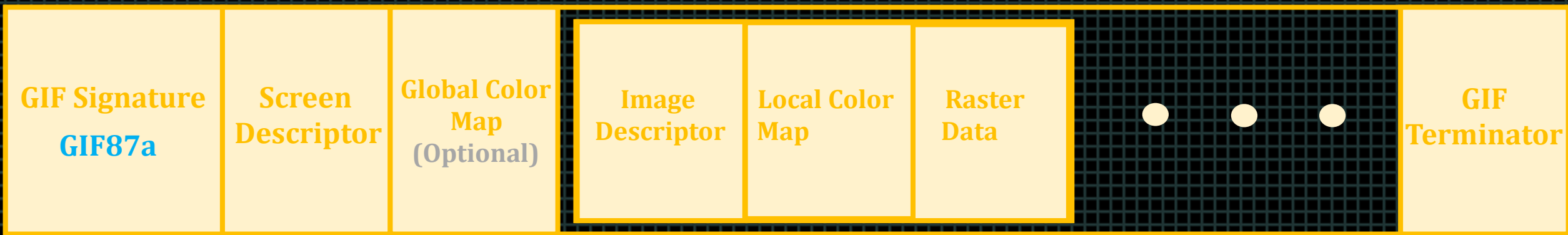
$$\begin{bmatrix} 0 & -1 & 0 \\ -1 & 4 & -1 \\ 0 & -1 & 0 \end{bmatrix}$$

Negative

$$\begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

Positive

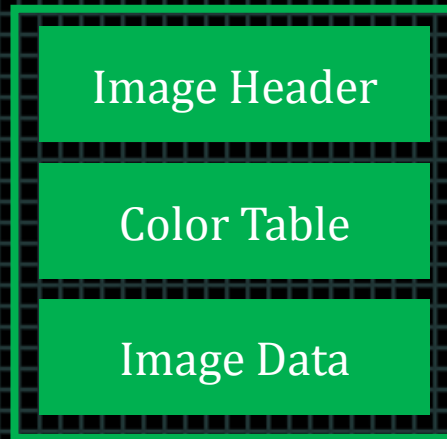
GIF



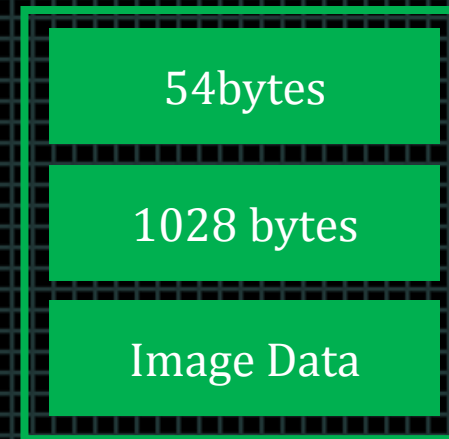
BMP

The bitmap Image

Image



BMP



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46	4	number of colors in image, or zero
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Filtering in the Frequency Domain

Simple Image Operations

Warping and Morphing

Basic Texture Operations

Manipulating Shapes

Edge Detection

Image Restoration

Morphological Processing

Image Compression and Coding

Feature Extraction and Representation

Feature Extraction and Representation

Image Classification

Image Data Basics

Signal Statistics and Noise

Signal:

How one parameter
Relates to another