

# SPATIAL CONVOLUTION AND CORRELATION

# Filtering

- Filtering is a technique for modifying or enhancing an image.
  - Emphasize some features or remove other features
- Includes smoothing, sharpening, edge enhancement etc
- When filtering is used directly on the pixels it is called **spatial filtering**

# Correlation and Convolution

- Correlation and Convolution are related concepts used in **linear spatial filtering**
  - When filtering is used directly on the pixels it is called **spatial filtering**
  - Term **linear** indicates that a pixel is replaced by the linear combinations of its neighbours

# Correlation

- Correlation is a neighbourhood operation in which each output pixel is the weighted sum of neighbouring pixels
- The matrix of weights is called correlation kernel or filter or mask.

# Correlation Example:

- Input Image

1	2	2	3	2
4	2	3	1	2
4	2	1	1	6
3	2	1	4	5
3	2	1	2	1

- Filter / mask:

4	5	1
5	1	2
3	1	3

Correlation Example: Apply correlation to pixel (1, 2) on the given below image using the given filter.

1	2	2	3	2
4	2	3	1	2
4	2	1	1	6
3	2	1	4	5
3	2	1	2	1

Input Image

4	5	1
5	1	2
3	1	3

Filter / mask

$$\text{Output Pixel (1,2)} = 2 \times 4 + 2 \times 5 + 3 \times 1 + 2 \times 5 + 3 \times 1 + 1 \times 2 + 2 \times 3 + 1 \times 1 + 1 \times 3 = 46$$

Output Image

1	2	2	3	2
4	2	46	1	2
4	2	1	1	6
3	2	1	4	5
3	2	1	2	1

# Correlation Example: On entire image

0	0	0	0	0	0	0
0	1	2	2	3	2	0
0	4	2	3	1	2	0
0	4	2	1	1	6	0
0	3	2	1	4	5	0
0	3	2	1	2	1	0
0	0	0	0	0	0	0

Input Image

4	5	1
5	1	2
3	1	3

Filter / mask

Output Image

15	34	30	33	22
25	61	46	67	38
39	67	56	59	42
38	60	46	42	66
24	42	32	38	52

# Convolution

- Convolution is similar to correlation and so, each output pixel is the weighted sum of neighbouring pixels
- The difference is that, the matrix of weights (filter or mask) is rotated  $180^\circ$



# Convolution Example: On entire image

0	0	0	0	0	0	0
0	1	2	2	3	2	0
0	4	2	3	1	2	0
0	4	2	1	1	6	0
0	3	2	1	4	5	0
0	3	2	1	2	1	0
0	0	0	0	0	0	0

Input Image

3	1	3
2	1	5
1	5	4

4	5	1
5	1	2
3	1	3

Filter / mask

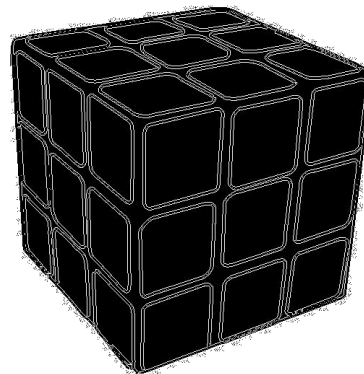
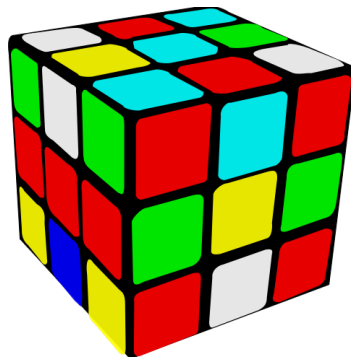
Output Image

39	40	42	33	19
49	54	40	62	46
47	55	45	90	42
46	47	50	68	29
22	27	34	31	22

# EDGE DETECTION

# Edges

- Edges are those places in a image that corresponds to object boundaries
- Variation in intensity level between neighbourhood pixels constitute an edge
- Therefore the objective of edge detector is to measure intensity gradient



# Edge Detection, Why?

- Image sharpening and enhancement
- Data extraction
- Object identification
  - Detect object shape, size etc
  - Finger print matching, medical diagnosis, number plate detection

# Types of edges

- Generally edges are of three types:
  - Horizontal
  - Vertical
  - Diagonal edges

# Edge Detection Steps

1. Smoothing : Noise reduction
2. Enhancement: Edge sharpening
3. Thresholding: Detecting which to discard and which to maintain
4. Localization: Determine the exact location of an edge

# Edge detection methods

- First order derivative/Gradient Methods
  - Roberts Operator
  - Prewitt Operator
  - Sobel Operator
  - Robinson
  - Kirsch
- Second order derivative
  - Laplacian of Gaussian
  - Difference of Gaussian

# Gradient Methods

- An image gradient is a directional change in the intensity or colour in an image
- It includes both magnitude and direction
  - Magnitude of the gradient is defined by

$$M = \sqrt{G_x^2 + G_y^2}$$

- Direction is given by

$$\theta = \tan^{-1} \left( \frac{G_x}{G_y} \right)$$



# Roberts Edge Detector

- Roberts operator performs a simple two dimensional spatial gradient measurement on an image
- It uses a pair of 2x2 convolution mask

$$G_x = \begin{array}{|c|c|} \hline 1 & 0 \\ \hline 0 & -1 \\ \hline \end{array}$$

$$G_y = \begin{array}{|c|c|} \hline 0 & 1 \\ \hline -1 & 0 \\ \hline \end{array}$$

# Roberts Edge Detector

- Algorithm:
  - Read the image
  - Convolve the image with  $G_x$
  - Convolve the image with  $G_y$
  - Compute the edge magnitude
  - Compare the edge magnitude with a threshold value
    - If edge magnitude is higher, assign it as a possible edge point

# Roberts Edge Detector : Example

A =

3	6	1	5
3	5	7	0
4	3	4	3
2	5	7	5

G<sub>x</sub> =

1	0
0	-1

G<sub>y</sub> =

0	1
-1	0



# Roberts Edge Detector : Example

$$A = \begin{bmatrix} 3 & 6 & 0 & 1 & 0 & 5 & 1 & 0 \\ 3 & 5 & -1 & 7 & -1 & 0 & 0 & -1 \\ 4 & 3 & & 4 & & 3 & & \\ 2 & 5 & & 7 & & 5 & & \end{bmatrix} \quad Gx = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \quad Gy = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$$

$$M = \sqrt{Gx^2 + Gy^2}$$

$$Gx = -2$$

$$Gy = 3$$

$$M = \sqrt{(-2)^2 + 3^2} = 3.61 \cong 4$$

4	4	2

# Roberts Edge Detector : Example

A =

3	6	1	5
3	5	7	0
4	3	4	3
2	5	7	5

Gx =

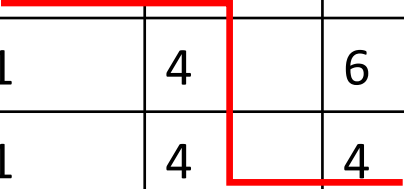
1	0
0	-1

Gy =

0	1
-1	0

M =

4	4	2
1	4	6
1	4	4



# Prewitt Edge Detector

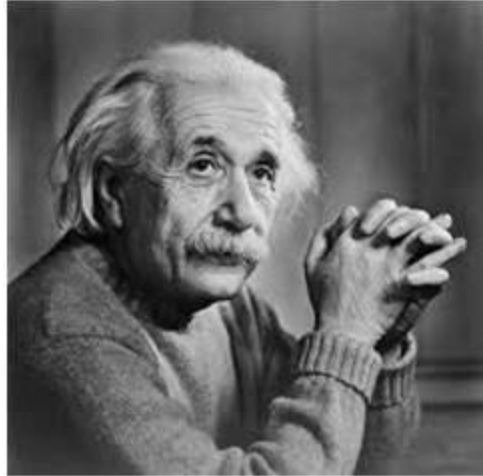
- It uses 3x3 convolution mask

$$G_x = \begin{array}{|c|c|c|} \hline -1 & 0 & 1 \\ \hline -1 & 0 & 1 \\ \hline -1 & 0 & 1 \\ \hline \end{array}$$

$$G_y = \begin{array}{|c|c|c|} \hline -1 & -1 & -1 \\ \hline 0 & 0 & 0 \\ \hline 1 & 1 & 1 \\ \hline \end{array}$$

- Prewitt method takes the central difference of the neighbouring pixels

# Prewitt Edge Detector



After applying vertical mask



After applying horizontal mask

# Sobel Edge Detector

- It uses 3x3 convolution mask

Gx =

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

Gy =

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

- Sobel method takes the central difference of the neighbouring pixels



# Gradient based Edge Detection

## **Advantages**

- Simple to implement
- Capable of detecting edges and their direction

## **Disadvantages**

- Susceptible to noise
- Not very accurate



THANK YOU