

## DEPARTMENT OF COMPUTER SCIENCE AND DESIGN

CD19P02 – FUNDAMENTALS OF IMAGE PROCESSING LABORATORY RECORD

### CD19P02 – FUNDAMENTALS OF IMAGE PROCESSING

|  |  |
| --- | --- |
| **List of Experiments** | |
| **1.** | Practice of important image processing commands – imread(), imwrite(), imshow(), plot() etc. |
| **2.** | Program to perform Arithmetic and logical operations |
| **3.** | Program to implement sets operations, local averaging using neighborhood processing. |
| **4.** | Program to implement Convolution operation. |
| **5.** | Program to implement Histogram Equalization. |
| **6.** | Program to implement Mean Filter. |
| **7.** | Program to implement Order Statistic Filters |
| **8.** | Program to remove various types of noise in an image |
| **9.** | Program to implement Sobel operator. |

**INDEX**

|  |  |  |  |
| --- | --- | --- | --- |
| **EXP.No** | **DATE** | **NAME OF THE EXPERIMENT** | **SIGN** |
| 1 |  | Practice of important image processing commands – imread(), imwrite(), imshow(), plot() etc. |  |
| 2a |  | Program to perform Arithmetic and logical operations |  |
| 2b |  | Program to perform logical operations |  |
| 3a |  | Program to implement sets operations using neighborhood processing. |  |
| 3b |  | Program to implement local averaging using neighborhood processing. |  |
| 4 |  | Program to implement Convolution operation. |  |
| 5 |  | Program to implement Histogram Equalization. |  |
| 6 |  | Program to implement Mean Filter. |  |
| 7 |  | Program to implement Order Statistic Filters |  |
| 8 |  | Program to remove various types of noise in an image |  |
| 9 |  | Program to implement Sobel operator. |  |
| 10 |  |  |  |

### INTRODUCTION TO MATLAB

MATLAB stands for MATrix LABoratory and the software is built up around vectors and matrices. It is

a technical computing environment for high performance numeric computation and visualization. It

integrates numerical analysis, matrix computation, signal processing and graphics in an easy-to-

use environment, where problems and solutions are expressed just as they are written mathematically,

without traditional programming. MATLAB is an interactive system whose basic data element is a matrix

that does not require dimensioning. It enables us to solve many numerical problems in a fraction of the time

that it would take to write a program and execute in a language such as FORTRAN, BASIC, or C. It

also features a family of application specific solutions, called toolboxes. Areas in which toolboxes are

available include signal processing, image processing, control systems design, dynamic systems

simulation, systems identification, neural networks, wavelength communication and others. It can handle

linear, non- linear, continuous-time, discretetime, multivariable and multirate systems.

## How to start MATLAB

Choose the submenu "Programs" from the "Start" menu. From the "Programs" menu, open the "MATLAB"

submenu. From the "MATLAB" submenu, choose "MATLAB".

## Procedure

1. Open Matlab.
2. File New Script.
3. Type the program in untitled window
4. File Save type filename.m in Matlab workspace path.
5. Debug Run.
6. Output will be displayed at Figure dialog box.

## Library Functions

### clc:

Clear command window

Clears the command window and homes the cursor.

### clear all:

Removes all variables from the workspace.

### close all:

Closes all the open figure windows.

### exp:

Y = exp(X) returns the exponential e x for each element in array X.

### linespace:

y = linspace(x1,x2) returns a row vector of 100 evenly spaced points between x1 and x2.

### rand:

X = rand returns a single uniformly distributed random number in the interval (0,1).

### ones:

X = ones(n) returns an n-by-n matrix of ones.

### zeros:

X = zeros(n) returns an n-by-n matrix of zeros.

**plot:**

plot(X,Y) creates a 2-D line plot of the data in Y versus the corresponding values in X.

### subplot:

subplot(m,n,p) divides the current figure into an m-by-n grid and creates an axes for a subplot in the position specified by p.

### stem:

stem(Y) plots the data sequence, Y, as stems that extend from a baseline along the x-axis. The data values are indicated by circles terminating each stem.

### title:

title(str) adds the title consisting of a string, str, at the top and in the center of the current axes.

### xlabel:

xlabel(str) labels the x-axis of the current axes with the text specified by str.

### ylabel:

ylabel(str) labels the y-axis of the current axes with the string, str.

### A Summary of Matlab Commands Used

|  |  |
| --- | --- |
| imread | Read image from graphics file |
| imwrite | Write image to graphics file |
| imfinfo | Information about graphics file |
| imshow | Display Image |
| Implay | Play movies, videos or image sequences |
| gray2ind | Convert grayscale to indexed image |
| ind2gray | Convert indexed image to grayscale image |
| mat2gray | Convert matrix to grayscale image |
| rgb2gray | Convert RGB image or colormap to grayscale |
| imbinarize | Binarize image by thresholding |
| adapthresh | Adaptive image threshold using local first-order statistics |
| otsuthresh | Global histogram threshold using Otsu's  method |
| im2uint16 | Convert image to 16-bit unsigned integers |
| im2uint8 | Convert image to 8-bit unsigned integers |
| imcrop | Crop image |
| imresize | Resize image |
| imrotate | Rotate image |
| imadjust | Adjust image intensity values or colormap |
| imcontrast | Adjust Contrast tool |
| imsharpen | Sharpen image using unsharp masking |
| histeq | Enhance contrast using histogram equalization |
| adapthisteq | Contrast-limited adaptive histogram  equalization (CLAHE) |
| imhistmatch | Adjust histogram of image to match N-bin  histogram of reference image |
| imnoise | Add noise to image |
| imfilter | N-D filtering of multidimensional images |
| fspecial | Create predefined 2-D filter |
| weiner2 | 2-D adaptive noise-removal filtering |
| medfilt2 | 2-D median filtering |
| ordfilt2 | 2-D order-statistic filtering |
| imfill | Fill image regions and holes |
| imclose | Morphologically close image |
| imdilate | Dilate image |
| imerode | Erode image |
| imopen | Morphologically open image |
| imreconstruct | Morphological reconstruction |
| watershed | Watershed transform |
| dct2 | 2-D discrete cosine transform |
| hough | Hough transform |

|  |  |
| --- | --- |
| graydist | Gray-weighted distance transform of grayscale image |
| fft2 | 2-D fast Fourier transform |
| ifftshift | Inverse FFT shift |
| imcomplement | Complement image |
| immultiply | Multiply two images or multiply image by constant |
| imsubtract | Subtract one image from another or  subtract constant from image |
| imdivide | Divide one image into another or divide image by constant |
| imadd | Add two images or add constant to image |

**Ex.No:1 IMPLEMENTATION OF IMAGE PROCESSING COMMANDS**

### Date:

**Aim:**

To Perform important image processing commands using Matlab.

### Software Used:

MATLAB

### Theory:

**Basic Image Processing with MATLAB:**

MATLAB is a very simple software for coding. All data variable in MATLAB are thought a matrix

and matrix operations are used for analyzing them. MATLAB has the different tool

boxes according to application areas. In this section, MATLAB Image Processing Toolbox is

presented and the use of its basic functions for digital image is explained.

**Read, write, show image and plot:**

**imread()**

It is the function is used for reading image. If we run this function with requiring data, image is

converted to a two‐dimensional matrix (gray image is two‐dimensional, but, color image is

three‐dimensional) with rows and columns including gray value in the each cell.

I = imread(‘path/filename.fileextension');

imread() function only needs an image file. If the result of imread() function is equal to a variable, a

matrix variable (I) is created. File name, extension, and directory path that contains image must be

written between two single quotes. If script and image file are in the same folder,path is not necessary.

**imshow()**

The matrix variable of image is showed using imshow() function. If many images show with sequence on the different figure windows, we use “figure” function for opening new window.

**imwrite()**

It is the function is used to create an image. This function only requires a new image file name with extension. If the new image is saved to a specific directory, the path of directory is necessary.

**subplot**

Subplot divides the current figure into rectangular panes that are numbered rowwise. Each pane contains

an axes object which you can manipulate using Axes Properties. Subsequent plots are output to the current

pane. h

= subplot(m,n,p) or subplot(mnp) breaks the figure window into an m-by-n matrix of small axes, selects the

pth axes object for the current plot, and returns the axes handle. The axes are counted along the top row of

the figure window, then the second row, etc.

**impixelinfo**

The function impixelinfo creates a Pixel Information tool in the current figure. The Pixel Information

tool displays information about the pixel in an image that the pointer is positioned over. The tool can

display pixel information for all the images in a figure.

**imageinfo**

The function imageinfo creates an Image Information tool associated with the image in the current figure.

The tool displays information about the basic attributes of the target image in a separate figure. title –

The function title('string') outputs the string at the top and in the center of the current axe**s.**

### Program:

**To read and show the image**

%Clear all workspace clear

%Clear all open images close all

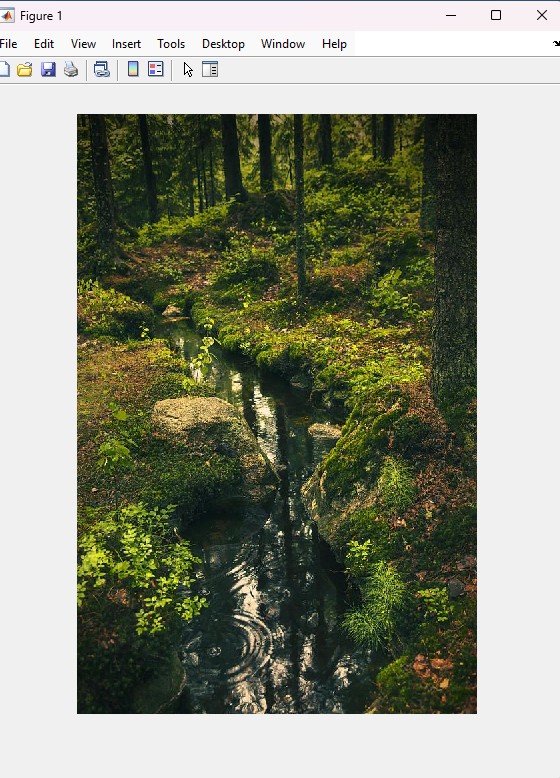
%Clear Command windows clc

%Read Image

I = imread('onion.png');

%Show Image imshow(I);

**Output:**



# Program:

clc;

clear all; close all;

subplot(2,2,1), imshow('cameraman.tif'),title('cameraman.tif'); subplot(2,2,2), imshow('peppers.png'),title('peppers.png'); subplot(2,2,3), imshow('baby.bmp'),title('baby.bmp'); subplot(2,2,4), imshow('butterfly.jpg'),title('butterfly'); impixelinfo;

imageinfo('cameraman.tif'); imageinfo('peppers.png'); imageinfo('baby.bmp'); imageinfo('butterfly.jpg');

# Output:

**Program:**

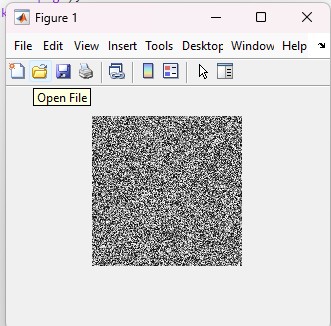
% Write Gray scale Image to PNG

% Write a 150-by-150 array of grayscale values to a PNG file in the current folder. clc;

clear all; close all;

A = rand(150); imwrite(A,'myGray.png'); imshow('mygray.png')

# Output:



**Program:**

%write Indexed Image with MATLAB colormap

% Write image data to a new PNG file with the built-in MATLAB colormap, copper.

% Load sample image data from the file clown.mat. clc;

clear all; close all; load clown.mat

% The image array X and its associated colormap, map, are loaded into the MATLAB workspace. map is a matrix of 81 RGB vectors.

% Define a copper-tone colormap with 81 RGB vectors. Then, write the image data to a PNG file using the new colormap.

newmap = copper(81); imwrite(X,newmap,'copperearth.png');

% imwrite creates the file, copperclown.png, in your current folder.

% View the new file by opening it outside of MATLAB. imshow('copperearth.png');

### Output:

**Result:**

The important image commands have been displayed and studied

### Ex.No:2a IMPLEMENTATION OF AIRTHMETIC OPERATIONS

**Date:**

#### Aim:

To implement arithmetic operations of an image using Matlab.

#### Software Used:

MATLAB

### Theory:

**Imadd**

Add two images or add constant to image

**Syntax:**

Z = imadd(X,Y)

**Description:**

Z = imadd(X,Y) adds each element in array X with the corresponding element in array Y and returns the sum in

the corresponding element of the output array Z. X and Y are real, nonsparse numeric arrays with the same size and class,

or Y is a scalar double. Z has the same size and class as X, unless X is logical, in which case Z is double.

If X and Y are integer arrays, elements in the output that exceed the range of the integer type are truncated, and

fractional values are rounded.

**Example**

Add two uint8 arrays. Note the truncation that occurs when the values exceed 255.

X = uint8([ 255 0 75; 44 225 100]);

Y = uint8([ 50 50 50; 50 50 50 ]);

Z = imadd(X,Y) Z =

255 50 125

94 255 150

**imsubtract**

Subtract one image from another or subtract constant from image

**Syntax**

Z = imsubtract(X,Y)

**Description**

Z = imsubtract(X,Y) subtracts each element in array Y from the corresponding element in array X and returns the difference in the corresponding element of the output array Z. X and Y are real, nonsparse numeric arrays of the same

size and class, or Y is a double scalar. The array returned, Z, has the same size and class as X unless X is logical, in which case Z is double.

If X is an integer array, elements of the output that exceed the range of the integer type are truncated, and fractional

values are rounded.

**Example**

Subtract two uint8 arrays. Note that negative results are rounded to 0. X = uint8([ 255 10 75; 44 225 100]);

Y = uint8([ 50 50 50; 50 50 50 ]);

Z = imsubtract(X,Y) Z =

205 0 25

0 175 50

#### immultiply

Multiply two images or multiply image by constant

#### Syntax

Z = immultiply(X,Y)

#### Description

Z = immultiply(X,Y) multiplies each element in array X by the corresponding element in array Y and returns the

product in the corresponding element of the output array Z.

If X and Y are real numeric arrays with the same size and class, then Z has the same size and class as X. If X is a

numeric array and Y is a scalar double, then Z has the same size and class as X. If X is logical and Y is numeric, then

Z has the same size and class as Y. If X is numeric and Y is logical, then Z has the same size and class as X.

immultiply computes each element of Z individually in double-precision floating point. If X is an integer array,

then elements of Z exceeding the range of the integer type are truncated, and fractional values are rounded. If X and Y

are numeric arrays of the same size and class, you can use the expression X.\*Y instead of immultiply.

#### Example

%Scale an image by a constant factor:

I = imread('moon.tif'); J = immultiply(I,0.5);

subplot(1,2,1), imshow(I) subplot(1,2,2), imshow(J)

#### imdivide

Divide one image into another or divide image by constant

#### Syntax

Z = imdivide(X,Y)

#### Description

Z = imdivide(X,Y) divides each element in the array X by the corresponding element in array Y and returns the result

in the corresponding element of the output array Z. X and Y are real, nonsparse numeric arrays with the same size

and class, or Y can be a scalar double. Z has the same size and class as X and Y, unless X is logical, in which case Z

is double. If X is an integer array, elements in the output that exceed the range of integer type are truncated, and f

ractional values are rounded. If X and Y are numeric arrays of the same size and class, you can use the expression X./

Y instead of imdivide.

#### Example

%Divide two uint8 arrays. Note that fractional values greater than or equal to 0.5 are rounded up to the nearest integer. X = uint8([ 255 10 75; 44 225 100]);

Y = uint8([ 50 20 50; 50 50 50 ]);

Z = imdivide(X,Y) Z =

5 1 2

1 5 2

%Estimate and divide out the background of the rice image. I = imread('rice.png');

background = imopen(I,strel('disk',15)); Ip = imdivide(I,background); imshow(Ip,[])

### Program:

**To perform arithmetic operation in an image**

clc;

close all; clear all;

I = imread('rice.png');

J = imread('cameraman.tif'); K = imadd(I, J);

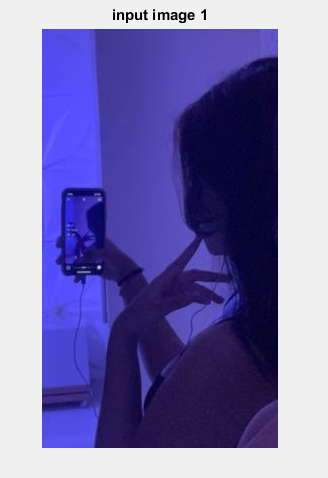
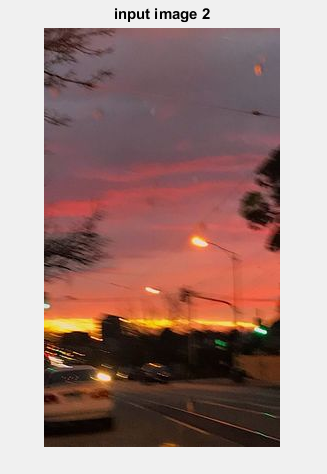
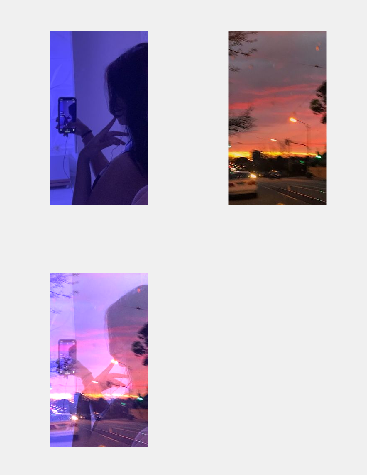
figure;imshow(I);title('input image 1');

figure;imshow(J);title('input image 2'); figure;imshow(K);title('Output image'); subplot(2,2,1);imshow(I);

subplot(2,2,2);imshow(J);

subplot(2,2,3);imshow(K);

**Output:**

**  **

# Program:

% To subtract one image from another, or subtract a constant value from an

% image, use the imsubtract function close all;

clear;

I = imread('rice.png'); %read a gray scale image into workspace background = imopen(I,strel('disk',15)); %Eastimate the Background Ip = imsubtract(I,background);

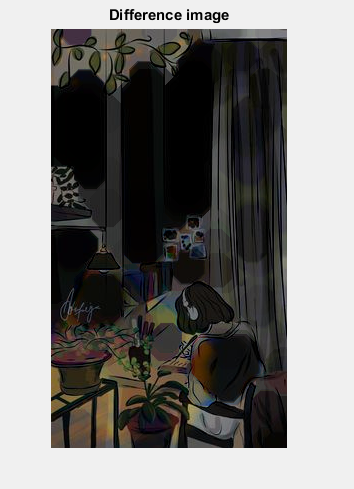
imshow(Ip,[]), title('Difference Image');

Iq = imsubtract(I,50); %subtract a constant value from the image figure

subplot(1,2,1), imshow(I), title('Original Image'); subplot(1,2,2), imshow(Iq), title('Subtracted Image');

# Output:

# 

**Program:**

%Multiply an image by itself.

%converts the class of the image from uint8 to uint16 before performing the multiplication to avoid truncating the results.

clc;

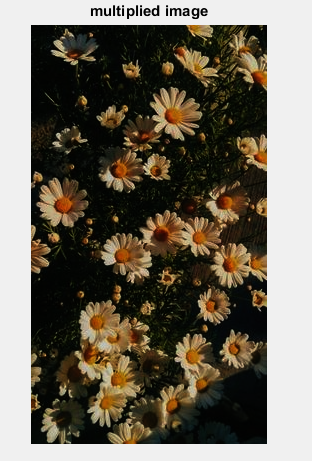
close all; clear all;

I = imread('moon.tif'); I16 = uint16(I);

J = immultiply(I16,I16);

imshow(I), title ('input image'), figure, imshow(J), title ('multiplied image');

# Output:

 ****

**Program:**

%Estimate and divide out the background of the rice image. clc;

clear all; close all;

I = imread('rice.png'); J = imdivide(I,2);

subplot(1,2,1), imshow(I), title('Input Image'); subplot(1,2,2), imshow(J), title('Output Image');

### Output:



**Result:**

Thus the arithmetic operations of an image have been implemented using MATLAB.

### Ex.No:2b IMPLEMENTATION OF LOGICAL OPERATIONS

**Date:**

#### Aim:

To implement logical operations of an image using Matlab.

#### Software Used:

MATLAB

### Theory:

Logical operations apply only to binary images, whereas arithmetic operations apply to multi-valued pixels.

Logical operations are basic tools in binary image processing, where they are used for tasks such as masking,

feature detection, and shape analysis. Logical operations on entire image are performed pixel by pixel. Because the

AND operation of two binary variables is 1 only when both variables are 1, the result at any location in a resulting

AND image is 1 only if the corresponding pixels in the two input images are 1. As logical operation involve only

one pixel location at a time, they can be done in place, as in the case of arithmetic operations. The XOR (exclusive

OR) operation yields a 1 when one or other pixel (but not both) is 1, and it yields a 0 otherwise. The operation is

unlike the OR operation, which is 1, when one or the other pixel is 1, or both pixels are 1.

Logical AND & OR operations are useful for the masking and compositing of images. For example, if

we compute the AND of a binary image with some other image, then pixels for which the corresponding

value in the binary image is 1 will be preserved, but pixels for which the corresponding binary value is 0 will

be set to 0 (erased) . Thus the binary image acts as a mask that removes information from certain parts of

the image. On the other hand, if we compute the OR of a binary image with some other image , the pixels

for which the corresponding value in the binary image is 0 will be preserved, but pixels for which

the corresponding binary value is 1, will be set to 1 (cleared).

#### Logical AND:

**Syntax:**

c = a & b;

Logical And is commonly used for detecting differences in images, highlighting target regions with a binary mask

or producing bit-planes through an image.

#### Logical OR:

**Syntax:**

C = a | b;

It is useful for processing binary-valued images (0 or 1) to detect objects which have moved between frames.

Binary objects are typically produced through application of thresholding to a grey-scale image.

#### Logical NOT:

**Syntax:**

B = ~A

This inverts the image representation. In the simplest case of a binary image, the (black) background pixels

become (white) and vice versa.

#### Logical X OR:

**Syntax:**

C = xor (a,b) ;

It is useful for processing binary-valued images (0 or 1) to detect objects which have moved between frames.

Binary objects are typically produced through application of thresholding to a grey-scale image.

### Program:

**To perform AND operation in an image**

% Create an image of zeros

imageSize = [200, 200]; % Specify image size

i = zeros(imageSize); % Initialize image with zeros

% Define the region where you want to set ones rowStart = 50;

rowEnd = 150;

colStart = 80;

colEnd = 120;

% Set ones in the specified region i(rowStart:rowEnd, colStart:colEnd) = 1;

% Create an image of zeros

imageSize = [200, 200]; % Specify image size

j = ones(imageSize); % Initialize image with ones

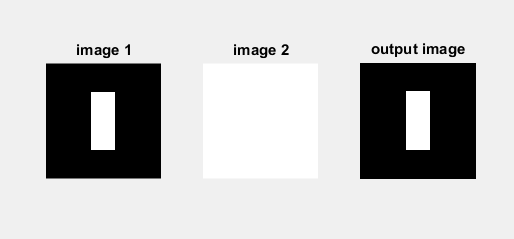
% Perform logical AND operation between image 'i' and image 'j' resultImage = i & j;

% Display the image1, image 2, and output image subplot(1, 3, 1), imshow(i), title('Image 1');

subplot(1, 3, 2), imshow(j), title('Image 2');

subplot(1, 3, 3), imshow(resultImage), title('Output Image');

**Output:**



# Program:

### To perform OR operation in an image

% Create an image of zeros

imageSize = [200, 200]; % Specify image size

i = zeros(imageSize); % Initialize image with zeros

% Define the region where you want to set ones rowStart = 50;

rowEnd = 150;

colStart = 80;

colEnd = 120;

% Set ones in the specified region i(rowStart:rowEnd, colStart:colEnd) = 1;

% Create an image of zeros

imageSize = [200, 200]; % Specify image size

j = ones(imageSize); % Initialize image with ones

% Perform logical OR operation between image 'i' and image 'j' resultImage = i | j;

% Display the image1, image 2, and output image subplot(1, 3, 1), imshow(i), title('Image 1')

subplot(1, 3, 2), imshow(j), title('Image 2');

subplot(1, 3, 3), imshow(resultImage), title('Output Image');

# Output:

# 

**Program:**

### To perform NOT operation in an image

% Create an image of zeros

imageSize = [200, 200]; % Specify image size

i = zeros(imageSize); % Initialize image with zeros

% Define the region where you want to set ones rowStart = 50;

rowEnd = 150;

colStart = 80;

colEnd = 120;

% Set ones in the specified region i(rowStart:rowEnd, colStart:colEnd) = 1;

resultImage = ~i ;

% Display the image1, image 2, and output image subplot(2, 2, 1), imshow(i), title('Input Image ');

subplot(2, 2, 2), imshow(resultImage), title('Output Image');

# Output:



**Program:**

### To perform XOR operation in an image

% Create an image of zeros

imageSize = [200, 200]; % Specify image size

i = zeros(imageSize); % Initialize image with zeros

% Define the region where you want to set ones rowStart = 50;

rowEnd = 150;

colStart = 80;

colEnd = 120;

% Set ones in the specified region i(rowStart:rowEnd, colStart:colEnd) = 1;

% Create an image of zeros

imageSize = [200, 200]; % Specify image size

j = ones(imageSize); % Initialize image with ones

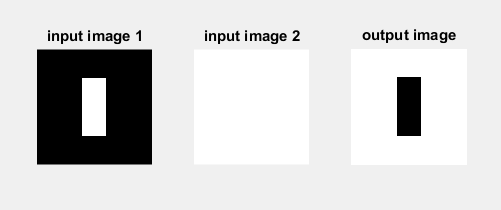
% Perform logical AND operation between image 'i' and image 'j' resultImage = xor(i,j);

% Display the image1, image 2, and output image subplot(1, 3, 1), imshow(i), title('Image 1');

subplot(1, 3, 2), imshow(j), title('Image 2');

subplot(1, 3, 3), imshow(resultImage), title('Output Image');

# Output:



### Result:

Thus the logical operations of an image have been implemented using MATLAB.

### Ex.No:3a IMPLEMENTATION OF SET OPERATIONS

**Date:**

#### Aim:

To implement Set operations of an image using Matlab.

#### Software Used:

MATLAB

### Theory:

Set operations in MATLAB refer to various mathematical operations performed on the pixel values

of two or more images. These operations allow you to combine or manipulate the pixel values

to achieve different effects. Here's an overview of some common set operations in MATLAB

image processing.

#### Union:

**Syntax:**

*unionImage = max(image A, image B);*

The union of two images is obtained by taking the maximum pixel value at each corresponding pixel

position from the input images. This operation can be used for merging images or enhancing certain features.

#### Interssection:

**Syntax:**

*intersectionImage = min(image A, image B);*

The intersection of two images is obtained by taking the minimum pixel value at each corresponding

pixel position from the input images. This operation highlights common features between the images.

### Complement:

**Syntax:**

*ComplementImage = 255 – image;*

The complement of an image is obtained by subtracting each pixel value from the maximum pixel value

(often 255 for 8-bit images). This operation results in an image with inverted pixel values.

### Difference:

**Syntax:**

*differenceimage = abs (image A – image B) ;*

The difference between two images is obtained by taking the absolute difference between their pixel

values. This operation can be used for highlighting dissimilarities between images.

### Program:

**To perform Convolution operation in an image**

clc;

clear all; close all;

a=imread('pic1.jpeg'); subplot(2,4,1); imshow(a); title('Original Image'); b=rgb2gray(a); subplot(2,4,2); imshow(b);

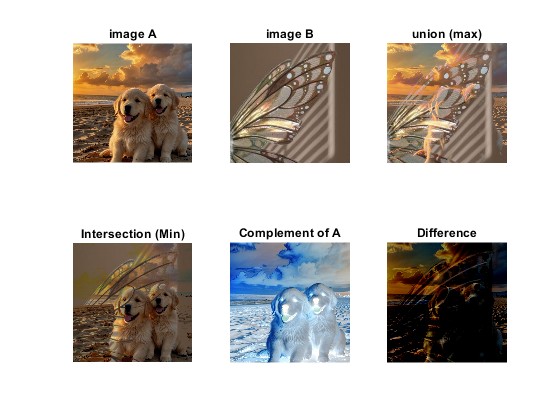
title('Gray Scale Image'); c=imnoise(b,'salt & pepper',0.1); subplot(2,4,6);

imshow(c);

title('Salt and Pepper Noise'); h1=1/9\*ones(3,3); c1=conv2(c,h1,'same');

subplot(2,4,3); imshow(uint8(c1)); title('3x3 Smoothing'); h2=1/25\*ones(5,5); c2=conv2(c,h2,'same');

subplot(2,4,7); imshow(uint8(c2)); title('5x5 Smoothing'); **Output:**



### Result:

Thus, the convolution operations of an image have been implemented using MATLAB.

### Ex.No:3b IMPLEMENTATION OF LOCAL AVERAGING

**USING NEIGHBORHOOD PROCESSING**

### Date:

#### Aim:

To implement local averaging using neighborhood processing in an image using Matlab.

#### Software Used:

MATLAB

### Theory:

Local averaging using neighborhood processing is a fundamental technique in image processing. It involves smoothing or blurring an image by computing the average value of pixels in a local neighborhood around each pixel. The goal is to reduce noise and fine details in the image while preserving its overall structure. Here's the theory behind the process.

#### Neighborhood Selection:

In this technique, a fixed-size neighborhood (also known as a kernel or filter) is defined around each pixel in the

image. This neighborhood is typically square or rectangular and can vary in size. Common neighborhood sizes are

3x3, 5x5, or 7x7, but the choice depends on the specific application and desired level of smoothing.

#### Kernel Creation:

A kernel is created with values that represent the weights assigned to each pixel within the neighborhood. For

local averaging, all values in the kernel are typically set to 1, and the sum of the kernel values is often normalized to 1

by dividing each value by the total number of values in the kernel. This ensures that the operation doesn't change

the overall brightness of the image.

#### Convolution Operation:

To perform local averaging, a convolution operation is applied to the image. Convolution is a mathematical operation

that combines two functions to produce a third function. In image processing, the convolution operation combines

the pixel values in the neighborhood with the corresponding values in the kernel. The result is a weighted sum of

pixel values, which effectively represents the average value of the pixels in the neighborhood.

#### Pixel Replacement:

The new value for the pixel at the center of the neighborhood is computed based on the weighted sum, and it replaces

the original pixel value. This process is repeated for every pixel in the image.

#### Smoothing Effect:

The convolution operation effectively smooths the image by averaging pixel values in local regions. Pixels with

strong noise or high-frequency details are averaged with their neighbors, leading to a blurring effect that reduces

the impact of noise and enhances the visibility of larger-scale features in the image.

#### Adjustable Smoothing:

The degree of smoothing can be controlled by adjusting the size of the neighborhood and the values in the kernel.

Larger neighborhoods or kernels with larger values will produce more significant smoothing, while

smaller neighborhoods or kernels with smaller values will result in less smoothing.

Local averaging using neighborhood processing is a simple yet powerful technique with a wide range of applications

in image processing, such as noise reduction, edge-preserving smoothing, and feature extraction. It's a building block

for more advanced filtering and processing techniques used in computer vision, image enhancement, and

computer graphics.

### Program:

**To perform local averaging using neighborhood processing in an image**

% Read the input image

inputImage = imread('image.jpg'); % Replace with your image's path

% Define the neighborhood size

neighborhoodSize = 3; % Adjust the neighborhood size as needed

% Apply local averaging using a filter

filter = fspecial('average', neighborhoodSize); averagedImage = imfilter(inputImage, filter);

% Display the original and averaged images subplot(1, 2, 1);

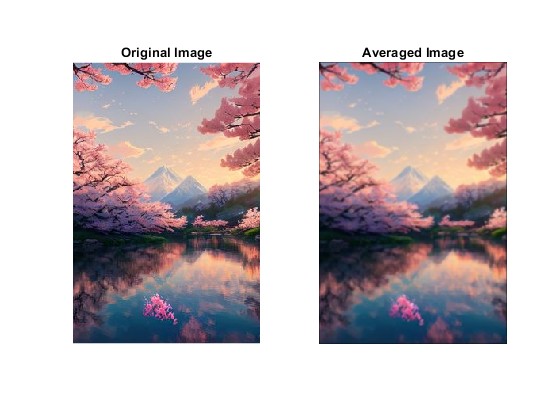
imshow(inputImage); title('Original Image');

subplot(1, 2, 2); imshow(averagedImage); title('Averaged Image');

% Save the averaged image imwrite(averagedImage, 'averaged\_image.jpg');

disp('Averaged image saved as "averaged\_image.jpg"');

### Output:

****

**Result:**

Thus, the local averaging using neighborhood processing of an image has been implemented using MATLAB.

### Ex.No:4 IMPLEMENTATION OF CONVOLUTION OPERATION

**Date:**

#### Aim:

To implement Convolution operation of an image using Matlab.

#### Software Used:

MATLAB

### Theory:

Convolution and correlation are the two fundamental mathematical operations involved in linear

filters based on neighbourhood-oriented image processing algorithms.

Convolution

Convolution processes an image by computing, for each pixel, a weighted sum of the values of

that pixel and its neighbours. Depending on the choice of weights, a wide variety of

image processing operations can be implemented.

Different convolution masks produce different results when applied to the same input image.

These operations are referred to as filtering operations and the masks as spatial filters. Spatial filters

are often named based on their behaviour in the spatial frequency. Low-pass filters (LPFs) are

those spatial filters whose effect on the output image is equivalent to attenuating the high-

frequency components (fine details in the image) and preserving the low-frequency components

(coarser details and homogeneous areas in the image). These filters are typically used to either blur

an image or reduce the amount of noise present in the image. Linear low-pass filters can

be implemented using 2D convolution masks with non-negative coefficients.

[High-pass filter](https://www.electronicsforu.com/resources/learn-electronics/band-stop-high-low-pass-filter)s (HPFs) work in a complementary way to LPFs, that is, these preserve or enhance

high-frequency components with the possible side-effect of enhancing noisy pixels as well.

High-frequency components include fine details, points, lines and edges. In other words, these

highlight transitions in intensity within the image. There are two in-built functions in

[MATLAB](https://www.electronicsforu.com/?s=matlab)’s Image Processing Toolbox (IPT) that can be used to implement 2D convolution: conv2

and filter2.

* 1. **conv2** computes 2D convolution between two matrices. For example, C=conv2(A,B) computes the
  2. two-dimensional convolution of matrices A and B. If one of these matrices describes a two-
  3. dimensional finite impulse response (FIR) filter, the other matrix is filtered in two dimensions.
  4. **filter2** function rotates the convolution mask, that is, 2D FIR filter, by 180° in each direction to create

a convolution kernel and then calls conv2 to perform the convolution operation.

### Program:

**To perform Convolution operation in an image**

clc;

clear all; close all;

a=imread('pic1.jpeg'); subplot(2,4,1); imshow(a); title('Original Image'); b=rgb2gray(a); subplot(2,4,2); imshow(b);

title('Gray Scale Image'); c=imnoise(b,'salt & pepper',0.1); subplot(2,4,6);

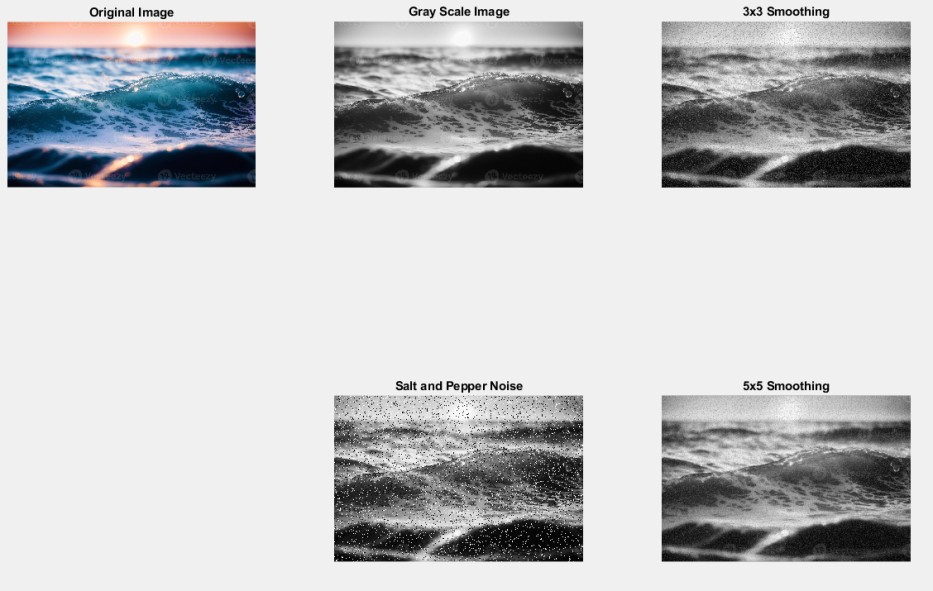
imshow(c);

title('Salt and Pepper Noise'); h1=1/9\*ones(3,3); c1=conv2(c,h1,'same');

subplot(2,4,3); imshow(uint8(c1)); title('3x3 Smoothing'); h2=1/25\*ones(5,5); c2=conv2(c,h2,'same');

subplot(2,4,7); imshow(uint8(c2)); title('5x5 Smoothing');

**Output:**



### Result:

Thus, the convolution operations of an image have been implemented using MATLAB

### Ex.No:5 IMPLEMENTATION OF HISTOGRAM EQUALIZATION

**Date:**

#### Aim:

To implement Histogram equalization of an image using Matlab.

#### Software Used:

MATLAB

### Theory:

Histogram of an image is a plot of number of occurrences of gray level in the image against the gray level value. For

dark image, histogram is concentrated in the lower (dark) side of the gray scale. For bright image, histogram

is concentrated on higher side of the gray scale. Equalization is a process that attempts to spread out the gray levels in

an image so that they are evenly distributed across the range.

#### Histogram Processing:

The contrast of an image can be modified by manipulating its histogram. A popular method is via Histogram

equalization. Here, the given histogram is manipulated such that the distribution of pixel values is evenly spread over

the entire range 0 to K-1. Histogram equalization can be done at a global or local level. In the global level the

histogram of the entire image is processed whereas at the local level, the given image is subdivided and the histograms

of the subdivisions (or sub images) are manipulated individually. When histogram equalization is applied locally,

the procedure is called Adaptive HistogramEqualization.

### Program:

**To perform Histogram Equalization in an image**

clc;

clear all; close all;

a= imread('peppers.png'); subplot(4,2,1); imshow(a); title('original image'); b=rgb2gray(a); subplot(4,2,3); imshow(b);

title('gray scale image'); subplot(4,2,4);

imhist(b); title('histogram'); subplot(4,2,5); c=histeq(b); imshow(c);

title('histogram equalisation image'); subplot(4,2,6);

imhist(c);

title('histogram equalisation'); subplot(4,2,7); f=adapthisteq(b);

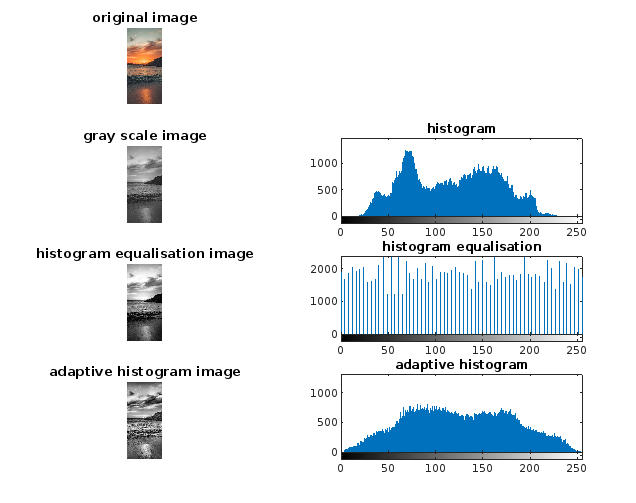
imshow(f);

title('adaptive histogram image'); subplot(4,2,8);

imhist(f);

title('adaptive histogram');

### Output:

****

**Program:**

clc ; clear; close;

img= imread ('D:\cameraman.jpg'); img=rgb2gray(img);

//I = imresize (img ,[256 ,256]) ; [ count , cells ]= imhist (img) ; Iheq = imhistequal(img); [count1,cells1 ]= imhist (Iheq); corrbsameimg = corr2(img,Iheq) disp(corrbsameimg);

x = xcorr ( count , count ) ; x1 = xcorr ( count , count1 ) ; subplot(2,1,1);

plot2d3 ( 'gnn' ,1: length ( x ) ,x ,2); title('correlation b/w histograms of original image'); subplot(2,1,2);

plot2d3 ('gnn' ,1: length ( x1 ) ,x1 ,2);

title('correlation b/w histogram of original and equalized image')

### Output:

### 

**Result:**

Thus, the Histogram equalization of an image have been implemented using MATLAB.

### Ex.No:6 IMPLEMENTATION OF MEAN FILTER

**Date:**

#### Aim:

To implement mean filter in an image reduce noise in digital images using Matlab.

#### Software Used:

MATLAB

### Theory:

When an image is acquired by a web camera or other imaging system, normally the vision system for which it is

intended is unable to use it directly. The image may be corrupted by random variations in intensity, variations

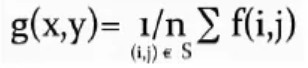
in illumination, poor contrast or noise that must be handle with in the early stages of vision processing. Therefore,

mean filter is one of the techniques which is used to reduce noise of the images.

This is a local averaging operation and it is a one of the simplest linear filter. The value of each pixel is replaced by

the average of all the values in the local neighborhood. Let f(i,j) is a noisy image then the smoothed image g(x,y) can

be obtained by,



Where S is a neighborhood of (x,y) and n is the number of pixels in S.

### Program:

clc;

close all;

clear all;

inputImage = imread('cameraman.tif'); % Load the input image

filterSize = 5; % Define the filter size (e.g., 3x3, 5x5, etc.)

% Pad the image to handle border pixels

paddedImage = padarray(inputImage, [filterSize, filterSize], 'replicate');

% Initialize the output image

outputImage = zeros(size(inputImage));

% Apply the mean filter

for i = 1:size(inputImage, 1)

for j = 1:size(inputImage, 2)

% Extract the neighborhood

neighborhood = paddedImage(i:i+filterSize-1, j:j+filterSize-1);

% Calculate the mean value of the neighborhood

meanValue = mean(neighborhood(:));

% Set the output pixel value

outputImage(i, j) = meanValue;

end

end

% Display the original and filtered images

subplot(1, 2, 1);

imshow(inputImage);

title('Original Image');

subplot(1, 2, 2);

imshow(uint8(outputImage));

title('Mean Filtered Image');

### Output:

### 

**Result:**

### The noise in an image is reduced using a mean filter, and it has been implemented using MATLAB

### 

### Ex.No:7 IMPLEMENTATION OF ORDER STATISTICS FILTERS

**Date:**

#### Aim:

To implement Order Statistics filters in an image using Matlab.

#### Software Used:

MATLAB

### Theory:

Order statistic filters are non-linear spatial filters whose response is based on the ordering(ranking) of the pixels

contained in the image area encompassed by the filter, and then replacing the value in the center pixel with the

value determined by the ranking result. The different types of order statistics filters include Median Filtering, Max

and Min filtering and Mid-point filtering.

#### Median Filtering:

The median filter selects the middle value when the neighborhood values are sorted, making it effective at noise

reduction and preserving edges.

K = (N+1)/2

Replaces the value of a pixel by the median of the pixel values in the neighborhood of that pixel.

#### Maximum Filtering:

The maximum filter selects the maximum value from the neighborhood, which enhances bright features and

suppresses dark features. (K=N)

The maximum filtering is achieved using the following equation f(x,y) = max g(s,t)

#### Minimum Filtering:

This filter selects the minimum value from the neighborhood, effectively enhancing dark features and suppressing

bright features. (K=1)

The minimum filtering is achieved using the following equation f(x,y) = min g(s,t)

### Program:

**To perform order Statistics Filters in an image**

clc; clear all; close all;

b = imread('C:\Users\indhu\Downloads\peppers.jpg'); subplot(2,3,1);

imshow(b); title('Original Image');

a=rgb2gray(b);

a = im2double(a);

a = imnoise(a,'salt & pepper',0.02); subplot(2,3,2);

imshow(a); title('Noise Image'); I = medfilt2(a); subplot(2,3,3); imshow(I);

title('Median filtered Image'); x=rand(size(a));

a(x(:)< 0.05)=0;

max\_Img = ordfilt2(a,9,ones(3,3)); subplot(2,3,4);

imshow(max\_Img);

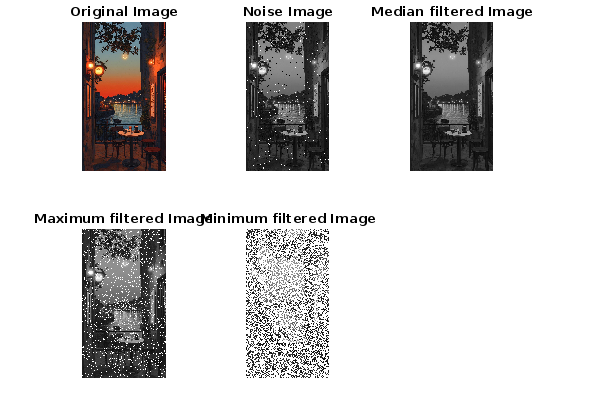
title('Maximum filtered Image'); a(x(:)< 0.95)=255;

min\_Img = ordfilt2(a,1,ones(3,3)); subplot(2,3,5);

imshow(min\_Img);

title('Minimum filtered Image');

### Output:



**Result:**

The different Order Statistics filters in an image have been implemented using MATLAB

### Ex.No:8 REMOVE VARIOUS TYPES OF NOISE IN AN IMAGE

**Date:**

#### Aim:

To Remove Various types of Noise in an Image an image using Matlab.

#### Software Used:

MATLAB

### Theory:

Image noise is the random variation of brightness or color information in images produced by the sensor and circuitry of

a scanner or digital camera. Image noise can also originate in film grain and in the unavoidable shot noise of an

ideal photon detector .Image noise is generally regarded as an undesirable by-product of image capture. Although

these unwanted fluctuations became known as "noise" by analogy with unwanted sound they are inaudible and such

as dithering. The types of Noise are following.

* Salt and Pepper Noise
* Gaussian Noise
* Rayleigh Noise
* Erlang Noise
* Exponential Noise
* Uniform Noise

#### Salt and Pepper Noise:

An image containing salt-and-pepper noise will have dark pixels in bright regions and bright pixels in dark regions.

This type of noise can be caused by dead pixels, analog-to-digital converter errors, bit errors in transmission, etc. This

can be eliminated in large part by using dark frame subtraction and by interpolating around dark/bright pixels.

#### Gaussian Noise:

The standard model of amplifier noise is additive, Gaussian, independent at each pixel and independent of the

signal intensity. In color cameras where more amplification is used in the blue color channel than in the green or

red channel, there can be more noise in the blue channel. Amplifier noise is a major part of the "read noise" of an

image sensor, that is, of the constant noise level in dark areas of the image.

#### Rayleigh Noise:

Rayleigh noise is characterized by a Rayleigh probability distribution. This distribution is commonly used to model

the amplitude of a signal that has passed through a random medium, resulting in attenuation and phase shifts.

Rayleigh noise is characterized by an intensity distribution, similar to the Rayleigh distribution in signal processing.

The distribution describes the probability of various pixel intensity values in the presence of noise.

#### Erlang Noise:

Erlang noise, also known as the Erlang distribution, is a statistical model used to describe the behavior of certain types

of noise or random processes. In image processing, Erlang noise is not as commonly encountered as other noise

models like Gaussian or Rayleigh noise. It is a continuous probability distribution that is often used to model the sum

of independent exponential random variables. It is also known as the gamma distribution when the shape parameter

is an integer. In

image processing, Erlang noise can be used to model variations in pixel intensities, especially when the image

acquisition process involves cumulative effects. This is different from many other noise models that assume each pixel

is independently affected

#### Exponential Noise:

Exponential noise, also known as exponential distribution, is a statistical model that describes random variations in

pixel intensities in digital images. This type of noise can be encountered in image processing due to various factors, and

it is important to understand and address it. Exponential noise is characterized by the exponential probability

distribution. This distribution is often used to model the time between events in a Poisson process, but it can also

describe random variations in Image intensities.

#### Uniform Noise:

Uniform noise, also known as uniform distribution, is a statistical model used to describe variations in pixel intensities

in digital images. It is one of the simpler noise models and is often encountered in image processing due to

various sources of noise. Uniform noise follows the uniform probability distribution, which is characterized by a

constant probability density over a specified range of Values. In image processing, uniform noise can be used to

model variations in pixel intensities that result from various factors, such as sensor noise, quantization errors, or

other sources of interference during image acquisition.

### Program:

**To remove various types of noise in an image**

#### Salt and Pepper Noise:

clc; clear all; close all;

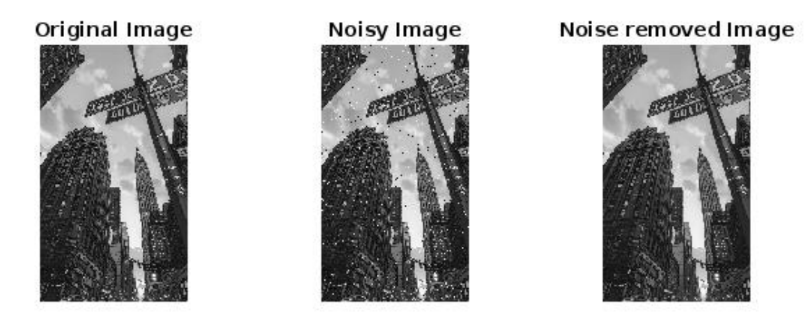
I = imread('eight.tif');

J = imnoise(I,'salt & pepper',0.02); subplot(2,3,1);

imshow(I) title('Original Image'); subplot(2,3,2) imshow(J)

title('Noisy Image'); Kmedian = medfilt2(J); subplot(2,3,3); imshow(Kmedian); title('Noise removed Image');

### Output:

****

#### Gaussian Noise:

clc; close all; clear all;

RGB = imread('saturn.png'); I = im2gray(RGB);

J = imnoise(I,'gaussian',0,0.025);

K = wiener2(J,[5 5]);

subplot(2,3,1); imshow(I) title('Original Image'); subplot(2,3,2); imshow(J)

title('Added Gaussian Noise'); subplot(2,3,3);

imshow(K);

title(' Wiener Filtered Image');

### Output:

### 

#### Rayleigh Noise:

clc; close all; clear all;

RGB = imread('saturn.png'); I = im2gray(RGB);

J = imnoise(I,'gaussian',0,0.025); K = wiener2(J,[5 5]);

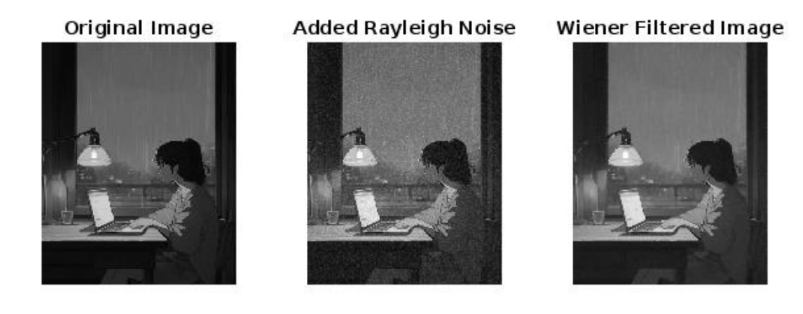
subplot(2,3,1); imshow(I) title('Original Image'); subplot(2,3,2); imshow(J)

title('Added Gaussian Noise'); subplot(2,3,3);

imshow(K);

title(' Wiener Filtered Image');

### Output:

****

#### d.Erlang Noise:

clc; close all; clear all;

I = imread('eight.tif'); scale = 10;

shape= 5;

sizeSignal = size(I);

erlangNoise = scale\*gamrnd(shape, 1, sizeSignal); noisy = double(I) + erlangNoise;

noisy = min(max(noisy, 0), 255); noisy = uint8(noisy); denoised=medfilt2(noisy); figure;

subplot(2, 3, 1); imshow(I); title('Input Image'); subplot(2, 3, 2); imshow(noisy); title('Noisy Image'); subplot(2, 3, 3); imshow(denoised);

title('Denoised Image');

### Output:

****

#### d.Uniform Noise:

I = imread('eight.tif'); minValue = 0;

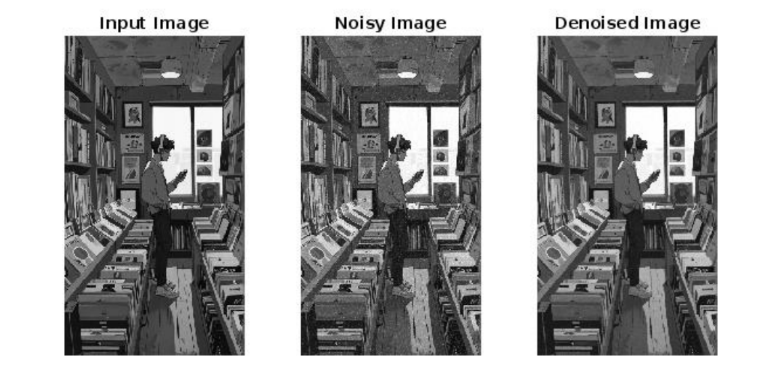
maxValue = 255; sizeImage = size(I);

uniformNoise = (maxValue - minValue) \* rand(sizeImage) + minValue; noisy = double(I) + uniformNoise;

noisy = min(max(noisy, 0), 255); noisy = uint8(noisy); denoised=medfilt2(noisy); figure;

subplot(1, 2, 1); imshow(noisy); title('Noisy Image'); subplot(1, 2, 2); imshow(denoised); title('Denoised Image');

### Output:

****

**Result:**

Thus, the various types of noise in an image have been removed and implemented using MATLAB.

### Ex. No:9 IMPLEMENTATION OF SOBEL OPERATOR

**Date:**

#### Aim:

To implement SOBEL operator in digital images for edge detection using Matlab.

#### Software Used:

MATLAB

### Theory:

The Sobel operator is a fundamental tool in image processing for edge detection and gradient estimation. It

is used to find edges or boundaries in images by measuring the rate of change of intensity at each pixel.

The theory behind the Sobel operator involves convolution with a pair of kernels to compute the gradients in

both the horizontal and vertical directions. Here is a detailed explanation of the theory behind the Sobel operator.

#### Gradient Calculation

The Sobel operator is designed to compute the gradient of an image. The gradient represents the rate of change of

pixel intensities, which is essential for identifying edges or abrupt changes in an image

#### Convolution Operation

The core operation of the Sobel operator involves convolution. Convolution is a mathematical operation that

combines two functions to produce a third. In image processing, it is used to apply a kernel or filter to an image.

#### Sobel Kernels

The Sobel operator uses two 3x3 convolution kernels, one for detecting changes in the horizontal direction (Sobel-X)

and the other for changes in the vertical direction (Sobel-Y).

Sobel-X Kernel:

-1 0 1 2 0 2 -1 0 1

Sobel-X Kernel:

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

#### Gradient Computation

To calculate the gradient at a given pixel, the Sobel operator convolves the image with both the Sobel-X and Sobel-

Y kernels separately.

The result of these two convolutions provides the horizontal gradient (Gx) and the vertical gradient (Gy) at each pixel.

#### Edge Detection

The Sobel operator highlights edges by emphasizing areas where the gradient magnitude (G) is high. A high

gradient magnitude indicates a rapid change in pixel intensities, which is characteristic of edges or boundaries.

#### Thresholding

To extract significant edges, a threshold can be applied to the gradient magnitude. Pixels with a gradient magnitude

above a certain threshold are considered part of an edge, while pixels with lower magnitudes are often treated as non-

edge pixels.

#### Noise Sensitivity

The Sobel operator is sensitive to noise, as noise can create small variations that may be mistaken for edges. Preprocessing steps, such as Gaussian smoothing, are sometimes applied to reduce noise before applying the operator.

#### Applications

The Sobel operator is widely used in image processing and computer vision tasks, including object detection,

feature extraction, image segmentation

### Program:

**To perform Sobel operator in an image**

% MATLAB code to detect edges using Sobel operator

% Load the input image

a = imread('peppers.png');

% Convert the input image to grayscale b = rgb2gray(a);

% Convert the grayscale image to double precision gray\_img = double(b);

% Specify the Sobel operator kernels

h\_kernel = [-1, 0, 1; -2, 0, 2; -1, 0, 1]; % Horizontal kernel

v\_kernel = [-1, -2, -1; 0, 0, 0; 1, 2, 1]; % Vertical kernel

% Use the Sobel operator kernels to calculate gradients c = imfilter(gray\_img, h\_kernel);

d = imfilter(gray\_img, v\_kernel);

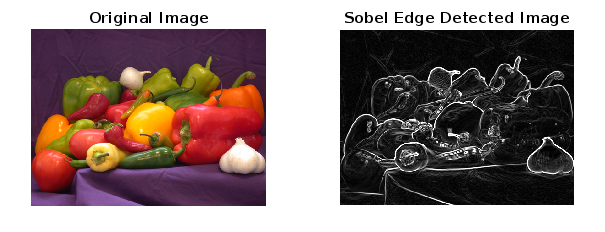
% Calculate the gradient magnitude gradient\_magnitude = sqrt(c.^2 + d.^2);

% Display the original image and the edge detected image figure;

subplot(2, 2, 1); imshow(a); title('Original Image');

subplot(2, 2, 2); imshow(uint8(gradient\_magnitude)); title('Sobel Edge Detected Image');

### Output:



**Result:**

The SOBEL operator in digital images for edge detection has been implemented using MATLAB

PROJECT

REAL-TIME IMAGE BLUR DETECTION

**Problem Statement:**

In the field of photography and image processing, capturing sharp and clear images is essential. However, due to various factors such as motion blur, focus errors, or camera shake, images often turn out blurry, reducing their quality and usability. Detecting image blur in real-time is a challenging task, especially for photographers and content creators who need to ensure the quality of their images before proceeding with further processing or publishing.

Traditional methods for evaluating image sharpness often involve manual inspection, which is time-consuming and subjective. A reliable automated solution for blur detection would save time and improve efficiency, allowing photographers to quickly identify whether an image needs to be retaken.

**Objective:**

The objective of this project is to develop a MATLAB-based tool that can automatically detect whether an image is "Blurry" or "Sharp." This tool will use image processing techniques such as the variance of the Laplacian filter, Fourier Transform, and edge detection to analyze the sharpness of an image. The solution should be robust enough to handle various types of images and provide real-time feedback to users.

This tool will benefit photographers and content creators by providing a quick and automated solution for assessing image clarity, ensuring high-quality output and minimizing time lost on blurry images.

**Code:**

[file, path] = uigetfile({'\*.jpg;\*.png;\*.bmp', 'Image Files (\*.jpg, \*.png, \*.bmp)'});

if isequal(file, 0)

disp('User selected Cancel');

else

img = imread(fullfile(path, file)); % Load the selected image

imshow(img); % Display the image

title('Loaded Image');

end

gray\_img = rgb2gray(img); % Convert to grayscale

laplacian\_filter = fspecial('laplacian', 0.2); % Create a Laplacian filter

laplacian\_response = imfilter(double(gray\_img), laplacian\_filter); % Apply the filter

variance\_laplacian = var(laplacian\_response(:)); % Calculate the variance

threshold = 100; % Example threshold for sharpness; you may need to adjust this

if variance\_laplacian < threshold

label = 'Blurry';

else

label = 'Sharp';

end

% Display the result

figure;

imshow(img);

title(['Image Quality: ', label]);

cam = webcam; % Connect to the webcam

while true

img = snapshot(cam); % Capture an image

gray\_img = rgb2gray(img); % Convert to grayscale

laplacian\_response = imfilter(double(gray\_img), laplacian\_filter);

variance\_laplacian = var(laplacian\_response(:));

if variance\_laplacian < threshold

label = 'Blurry';

else

label = 'Sharp';

end

% Display the image with label

imshow(img);

title(['Image Quality: ', label]);

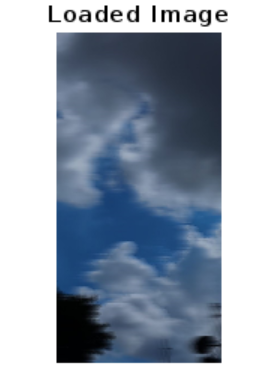
pause(0.5); % Pause to control the frame rate

end

clear cam; % Release the camera when done

Quit;

**Output:**

Then,

Capture images from a webcam in real-time, analyze them, and display live feedback on image sharpness.(Optional)

