# Lab 4

Applying filters on images.

| Filters types:

- Linear: (Mean Gaussian) Linear filtering is the filtering method where the value of output pixel is linear combinations of the neighboring input pixels.
- Non-linear: (Min Max Median) Non-linear filter is a filter whose output is not a linear function of its input.

# | Non-Linear filter

### Max Filter (Dilation Filter):

Applying morphological operation using a square-shaped structureing element

• It contains the regions of white pixels expanded or thickened.

\_It is useful for joining broken parts of an object or increasing the size of foreground regions in an image.

\_It works by moving the kernel over the image and replacing each pixel in the neighborhood with the maximum pixel value found in the kernel's corresponding region in the image.

```
kernel_size = 5
max_filtered_image = cv2.dilate(image, np.ones((kernel_size, kernel_size), np.uint8))

cv2.imshow('Max Filtered Image', max_filtered_image)
cv2.waitKey(0)
cv2.destroyAllWindows()
```

- kernel size is 5 => the kernel is a square matrix with both dimensions equal to 5(5x5 matrix): As each pixel in the output image is replaced by the median value of its 5x5 neighborhood.
- cv2.dilate() => function used to perform dilation.
- np.uint8 => used to ensure that the pixel values in the arrays are within the appropriate range for image data, and it also optimizes memory usage by using 8 bits per pixel.
- np.ones((kernel\_size, kernel\_size),np.uint8) => creates a kernel of size 5x5 filled with ones, which means it's a square structuring element of size 5x5.

## Min Filter (Erosion Filter):

It is a morphological operation that shrinks regions of foreground pixels (usually white) in a binary image.

It contains the regions of white pixels reduced or thinned

\_It is useful for tasks like removing small noisy elements or disconnecting connected regions in an image.

\_lt works by replacing each pixel in the neighborhood with the minimum pixel value found in the kernel's corresponding region in the image.

```
In [47]:
    kernel_size = 3
    output_image = cv2.erode(image, np.ones((kernel_size, kernel_size), np.uint8))

#cv2.imshow('Input Image', image)
    cv2.imshow('Min Filtered Image',output_image)
    cv2.waitKey(0)
    cv2.destroyAllWindows()
```

- Applying kernel size 3 to reducing the pixel of the image.
- cv2.erode(imageName, np.ones(,),) => performs the erosion operation on the image.

#### Median Filter:

\_It is useful for removing noise from an image while preserving edges.

\_It works by replacing each pixel value in the image with the median value of the pixel value in the image

```
In [52]: kernel_size = 5

median_filtered_image = cv2.medianBlur(image, kernel_size)

cv2.imshow('Median Filtered Image', median_filtered_image)
cv2.waitKey(0)
cv2.destroyAllWindows()
```

• cv2.medianBlur() => function replaces each pixel value with the median value of the pixel values in the neighborhood.

## | Linear Filter

## Mean Filter(Box Filter):

\_It is useful in bluring the image(smoothing an image and reducing noise).

\_It works by replacing each pixel value in the image with the average (mean) value of the pixel values in its neighborhood defined by the kernel size.

```
In [49]:
    kernel_size = (5, 5)
    mean_filtered_image = cv2.blur(image, kernel_size) #The function arranges the pixels to determine
    cv2.imshow('Mean Filtered Image', mean_filtered_image)
    cv2.waitKey(0)
    cv2.destroyAllWindows()
```

cv2.blur(image, kernel\_size) => applies median filtering operation to image.

## **Gaussian Blur:**

\_It is useful for bluring image.

\_It applies a weighted aveg=rage to each pixel in the image(The weights defined by a Gaussian function).

```
In [50]:
    kernel_size = (5,5)
    blur = cv2.GaussianBlur(image, kernel_size, 0) #0 for standad deviation
    cv2.imshow('Gaussian Blurred Image', blur)
    cv2.waitKey(0)
    cv2.destroyAllWindows()
```

 $\bullet$  cv2.GaussianBlur(image, (5, 5), 0) => Applies the Gaussian blur filtering operation to the input image