

Lecture 2:

Image Enhancement

Image Sensing and Acquisition

- The types of images in which we are interested are generated by the combination of an “*illumination*” source and the *reflection* or *absorption* of energy from that source by the elements of the “scene” being imaged.

Why Enhancement?

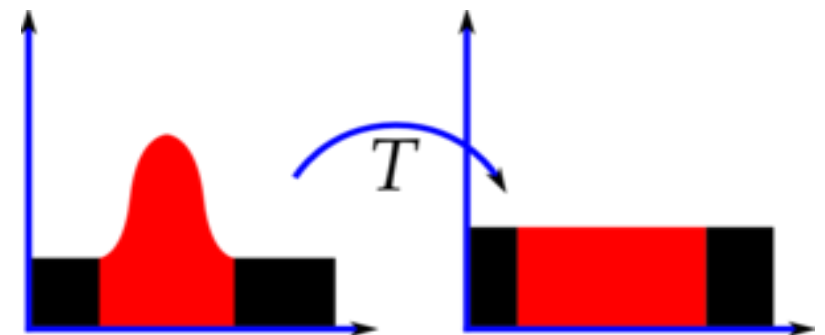
- ❑ Images may suffer from the following degradations:
 - **Poor contrast** due to poor illumination or finite sensitivity of the imaging device
 - **Electronic sensor noise** or atmospheric disturbances leading to broadband noise
 - Aliasing effects due to **inadequate sampling**
 - Finite aperture effects or motion leading to spatial

Cont..

- ❑ There are various and simple algorithms for image enhancement based on lookup tables
 - *Contrast enhancement*
- ❑ Other algorithms also work with simple linear filtering methods
 - *Noise removal*

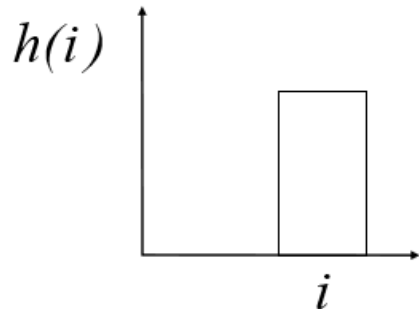
Histogram equalisation

- ❑ Equalization increases the global **contrast** of many images, especially when the usable **data** of the image is represented by close contrast values.
- ❑ Through this adjustment, the **intensities** can be better distributed on the histogram. This allows for areas of lower local contrast to gain a higher contrast.
- ❑ Histogram equalization accomplishes this by effectively **spreading out the most frequent intensity values**.
 - ✓ The method is useful in images with backgrounds and foregrounds that are both bright or both dark. In particular, the method can lead to better views of **bone** structure in **x-ray** images



Histogram equalisation

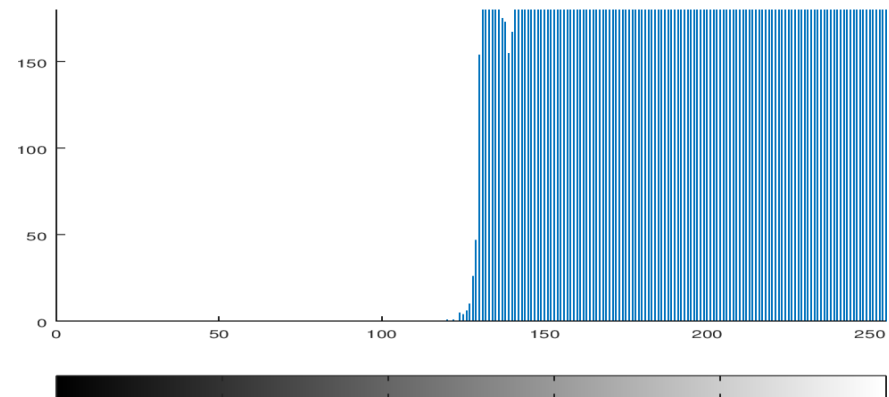
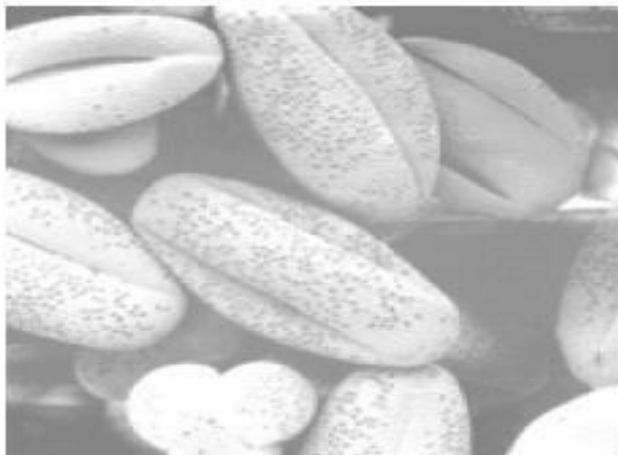
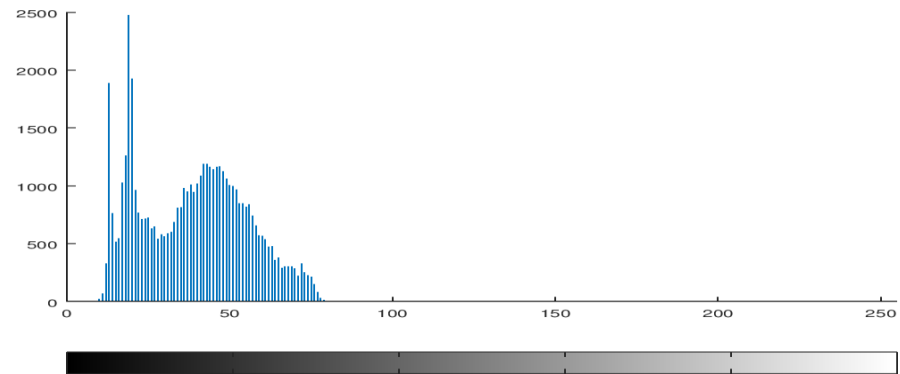
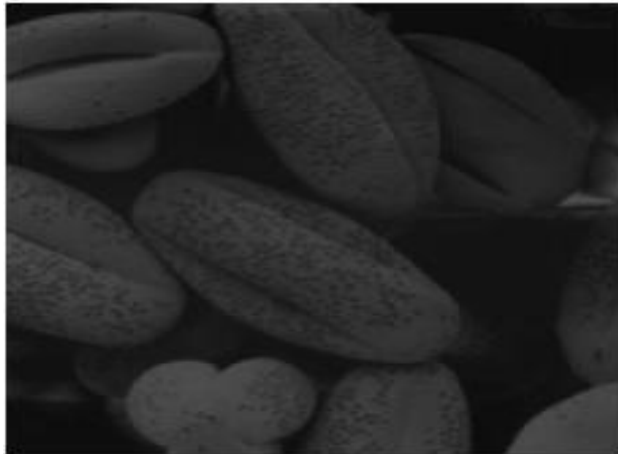
- ❑ In an image of **low contrast**, the image has grey levels **concentrated** in a **narrow band**
 - *The grey-levels are not too dark or too bright but in the middle. And it covers only few grey-level intensity range*
- ❑ Define the grey-level histogram of an image $h(i)$ where :
 - $h(i)$ =number of pixels with grey level = i
- ❑ Graphically, the histogram for a specific grey-level will be:



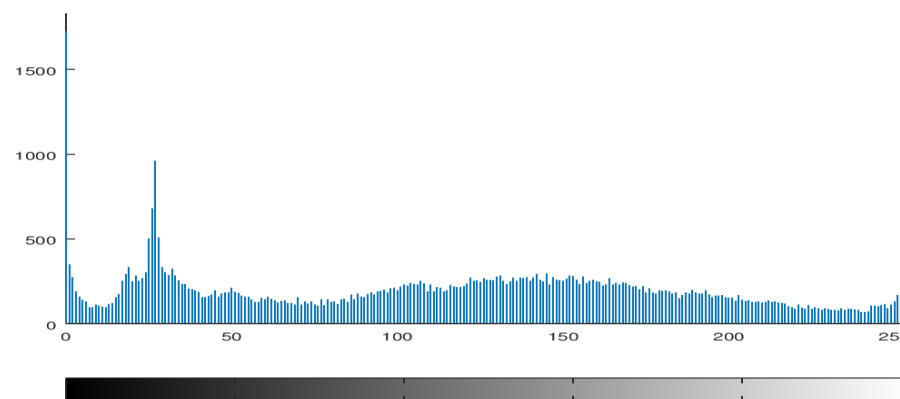
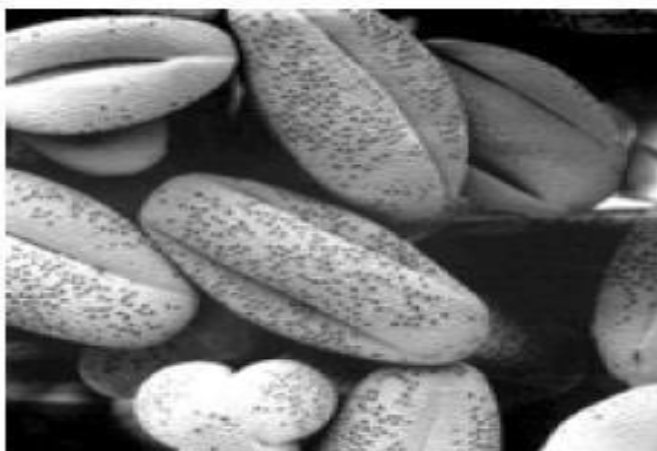
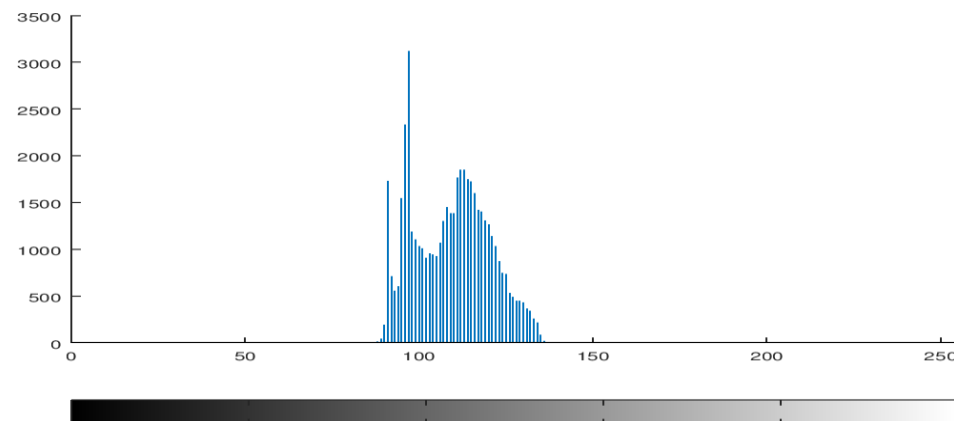
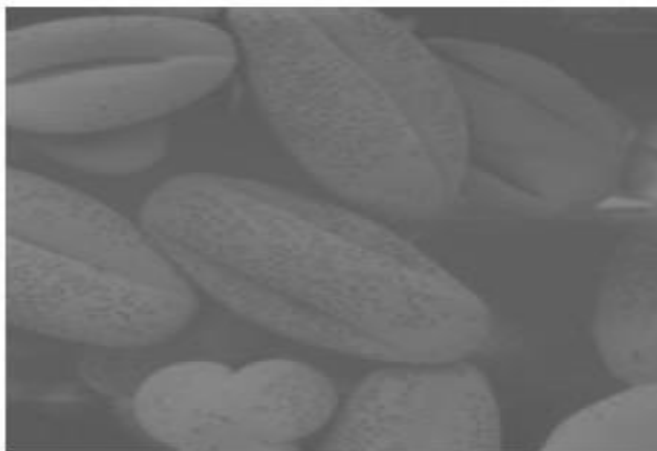
Darker

vs

Bright Image



Low vs High Contrast Image



Histogram equalisation



Original



Black/White

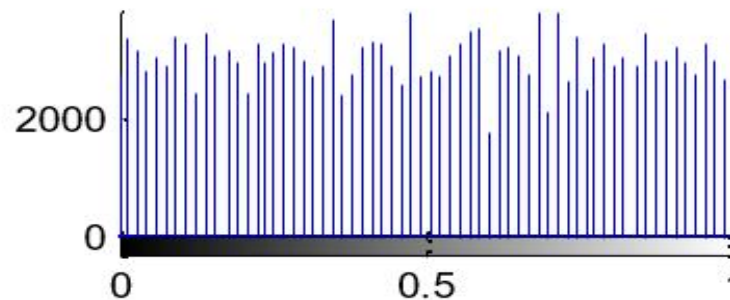
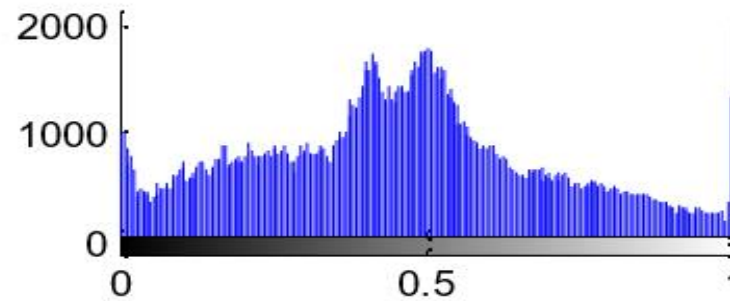


Grey Level



Histogram Equalized

Histogram Equalized Image

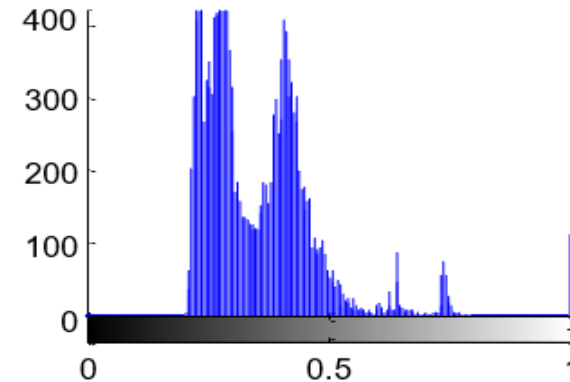


Improving a Low Contrast Image

Original



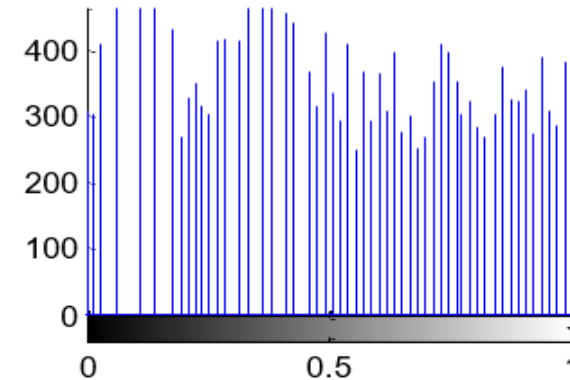
Histogram of Low Contrast Image



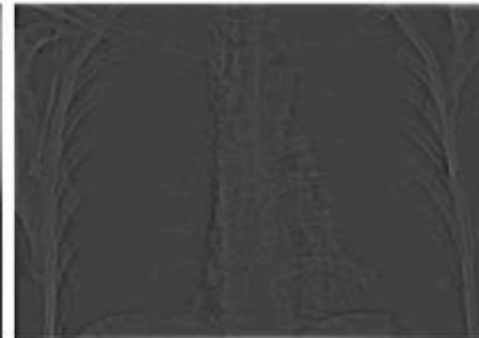
Histogram Equalized



Histogram of Equalized Image



Histogram equalisation



Gaussian highpass
filtering

High-frequency
emphasis filtering



Histogram
Equalisation

Grey level transformation

- ❑ There are three basic grey level transformation.
 - Linear
 - Logarithmic
 - Power – law
- ❑ Linear transformation includes simple *identity* and *negative* transformation.
 - In Identity transformation, each value of the input image is directly mapped to each other value of output image. That results in the same input image and output image.
 - In negative transformation, each value of the input image is subtracted from the $L-1$ and mapped onto the output image.

Image Filtering

- Simple image operators can be classified as:
 - **'pointwise'** which changes a pixel independent of the others;
 - **'neighbourhood'** (filtering) which changes the pixel value by consulting some or all of its neighbours
- *Histogram equalisation is a pointwise operation*
- More general filtering operations use neighbourhoods of pixels

Spatial domain filtering

- ❑ Some **neighborhood** operations work with
 - the values of the image pixels in the neighborhood, and
 - the corresponding values of a subimage that has the same dimensions as the neighborhood window.
- ❑ The subimage is called a **filter** (or mask, kernel, template, window).
- ❑ The values in a filter subimage are referred to as **coefficients**, rather than pixels.

Spatial domain filtering

❑ Operation:

- modify the pixels in an image based on some function of the pixels in their neighborhood.

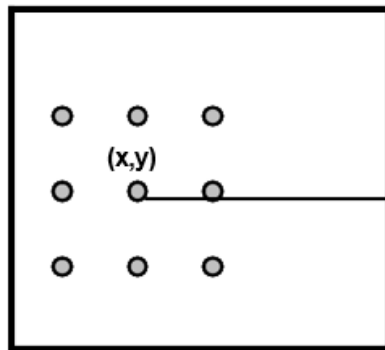
❑ Simplest:

- **linear filtering** (replace each pixel by a linear combination of its neighbors).

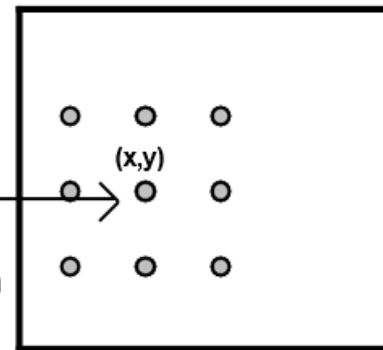
Linear spatial filtering is often referred to as “convolving an image with a filter”.

Image Filtering

Input image

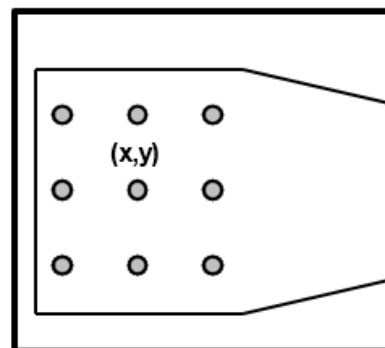


Output image

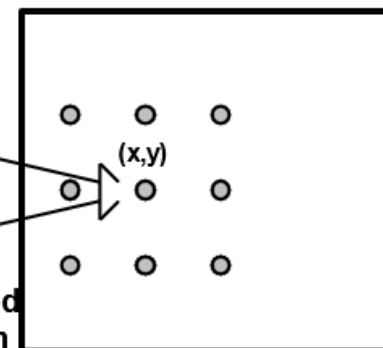


pointwise
transformation

Input image



Output image



neighbourhood
transformation

Linear filtering and convolution

□ Convolution involves:

1. **overlap**

2. **multiply**

3. **add** with 'convolution mask'

$$H = \begin{pmatrix} \frac{1}{9} & \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} & \frac{1}{9} \end{pmatrix}$$

Spatial domain filtering

3	3	3
3	3	3
3	3	3

❑ What is the value of the center pixel?

3	4	3
2	3	3
3	4	2

❑ What assumptions are you making to infer the center value?

Spatial domain filtering

- ❑ Be careful about indices, image borders and padding during implementation.



zero



fixed/clamp



periodic/wrap



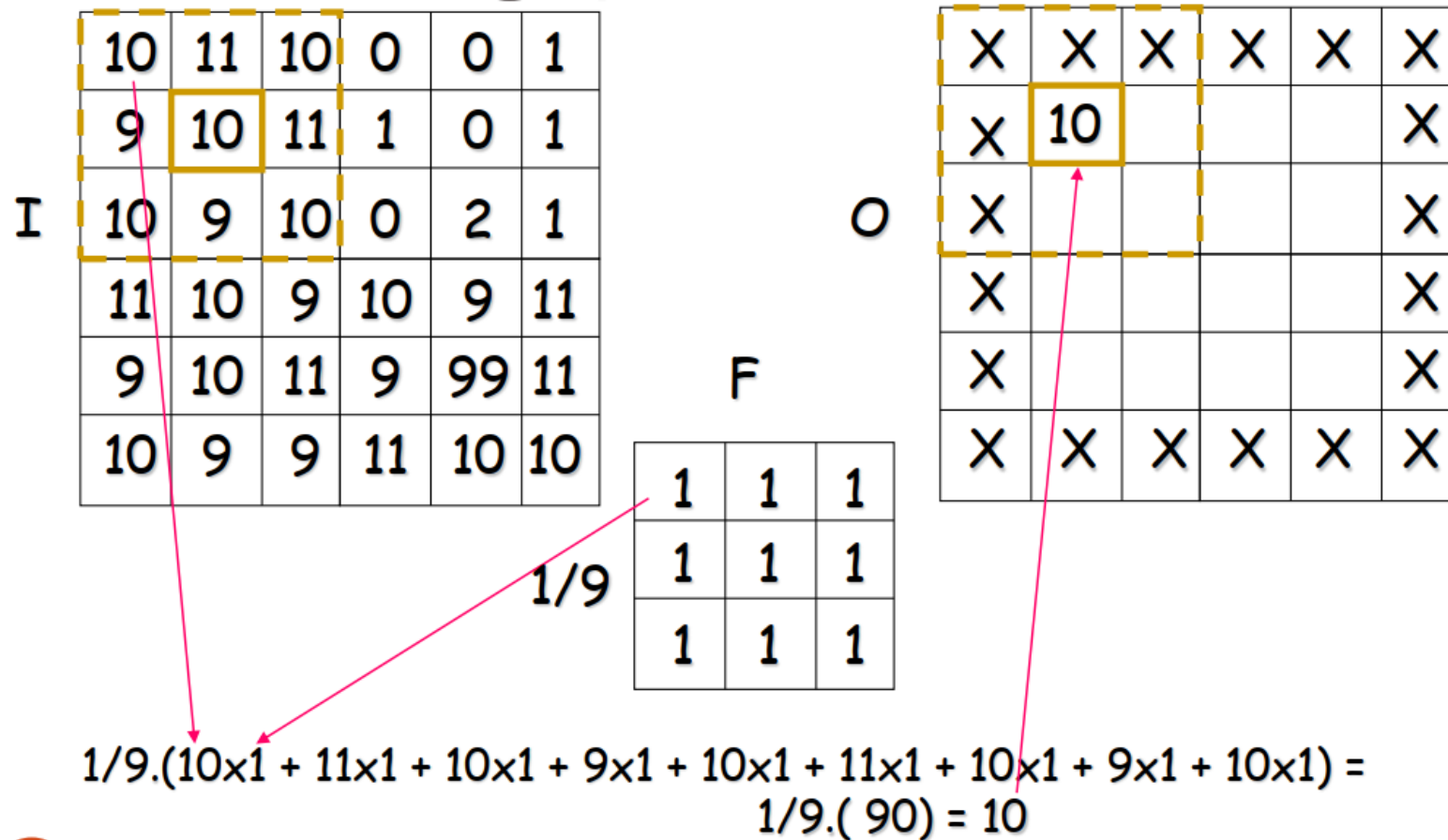
reflected/mirror

Border padding examples.

Smoothing spatial filters

- ❑ Often, an image is composed of
 - some underlying ideal structure, which we want to detect and describe,
 - together with some random noise or artifact, which we would like to remove.
- ❑ Smoothing filters are used for **blurring** and for **noise reduction**.
- ❑ Linear smoothing filters are also called **averaging filters**.

Smoothing spatial filters



Smoothing spatial filters

I	10	11	10	0	0	1
	9	10	11	1	0	1
	10	9	10	0	2	1
	11	10	9	10	9	11
	9	10	11	9	99	11
	10	9	9	11	10	10

O	X	X	X	X	X	X
	X					X
	X					X
	X					X
	X			20		X
	X	X	X	X	X	X

F	1	1	1
	1	1	1
	1	1	1

1/9

$$1/9.(10 \times 1 + 9 \times 1 + 11 \times 1 + 9 \times 1 + 99 \times 1 + 11 \times 1 + 11 \times 1 + 10 \times 1 + 10 \times 1) = 1/9.(180) = 20$$

Order-statistic filters

I

10	11	10	0	0	1
9	10	11	1	0	1
10	9	10	0	2	1
11	10	9	10	9	11
9	10	11	9	99	11
10	9	9	11	10	10

O

X	X	X	X	X	X
X	10				X
X					X
X					X
X					X
X	X	X	X	X	X

10, 11, 10, 9, 10, 11, 10, 9, 10

sort →

9, 9, 10, 10, 10, 10, 10, 11, 11

median

Order-statistic filters

I

10	11	10	0	0	1
9	10	11	1	0	1
10	9	10	0	2	1
11	10	9	10	9	11
9	10	11	9	99	11
10	9	9	11	10	10

O

X	X	X	X	X	X
X					X
X					X
X					X
X				10	X
X	X	X	X	X	X

10, 9, 11, 9, 99, 11, 11, 10, 10

sort →

9, 9, 10, 10, 10, 11, 11, 11, 99

median

Common 3x3 Filters

- Low/High pass filter

$$\frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \quad \begin{bmatrix} -1 & -1 & -1 \\ -1 & 9 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$

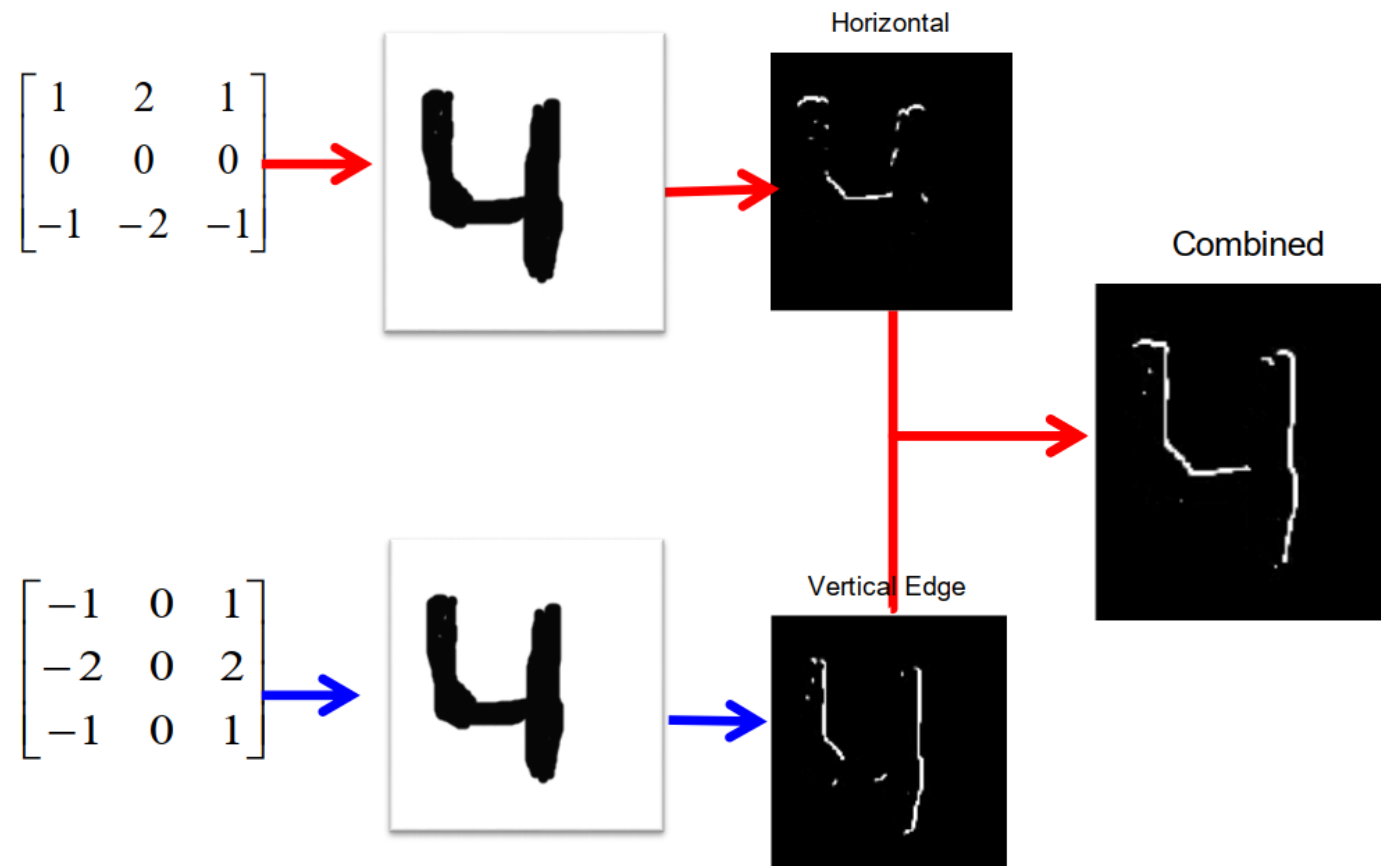
- Blur operator

$$\frac{1}{13} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 1 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

- H/V Edge detector

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

Edge Detection



Smoothing spatial filters

- ❑ Common types of noise:
 - **Salt-and-pepper noise**: contains random occurrences of black and white pixels.
 - **Impulse noise**: contains random occurrences of white pixels.
 - **Gaussian noise**: variations in intensity drawn from a Gaussian normal distribution.



Original



Salt and pepper noise



Impulse noise



Gaussian noise

Linear filtering and convolution

Original



Noisy



Filtered
 $\sigma=1.5$



Filtered
 $\sigma=3.0$



Conclusion

- ❑ We have looked at basic (low level) image processing operations
 - Enhancement
 - Filtering
- ❑ These are usually important pre-processing steps carried out in computer vision systems



The End