

# Medical Engineering - Imaging Systems

## Image Processing

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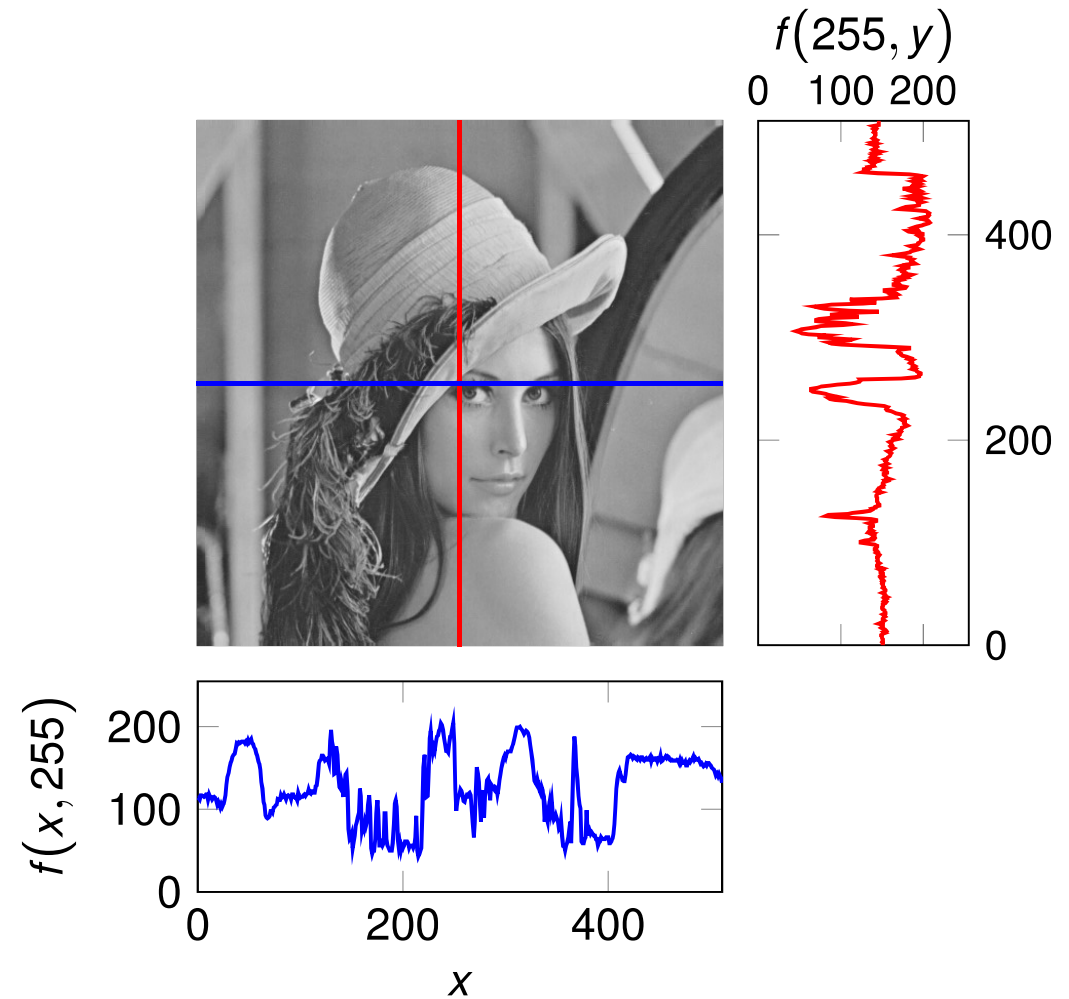
# Image Processing

## The Basics of Image Processing

### Further Readings

## Images as 2-D Signals

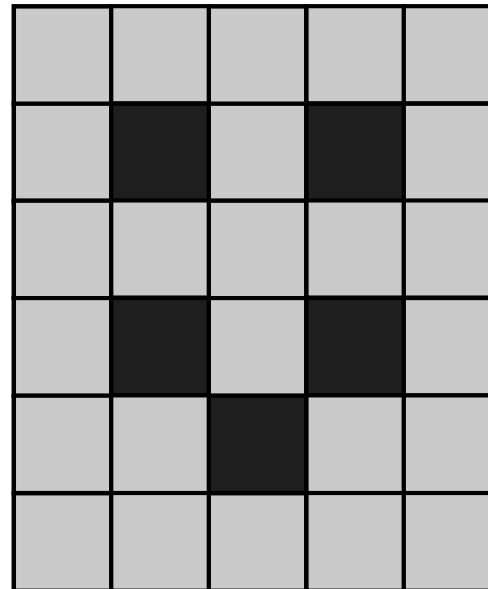
Image  $f$ : a 2-D function  $f(x, y)$   
defined over the discrete *image*  
*domain*  $\Omega \subset \mathbb{Z}^2$



## Image Storage

Images are stored as matrices on a computer.

200	200	200	200	200
200	35	200	35	200
200	200	200	200	200
200	35	200	35	200
200	200	35	200	200
200	200	200	200	200



## Gamma Correction – Example



Gamma corrected ( $\gamma = 2.2$ )

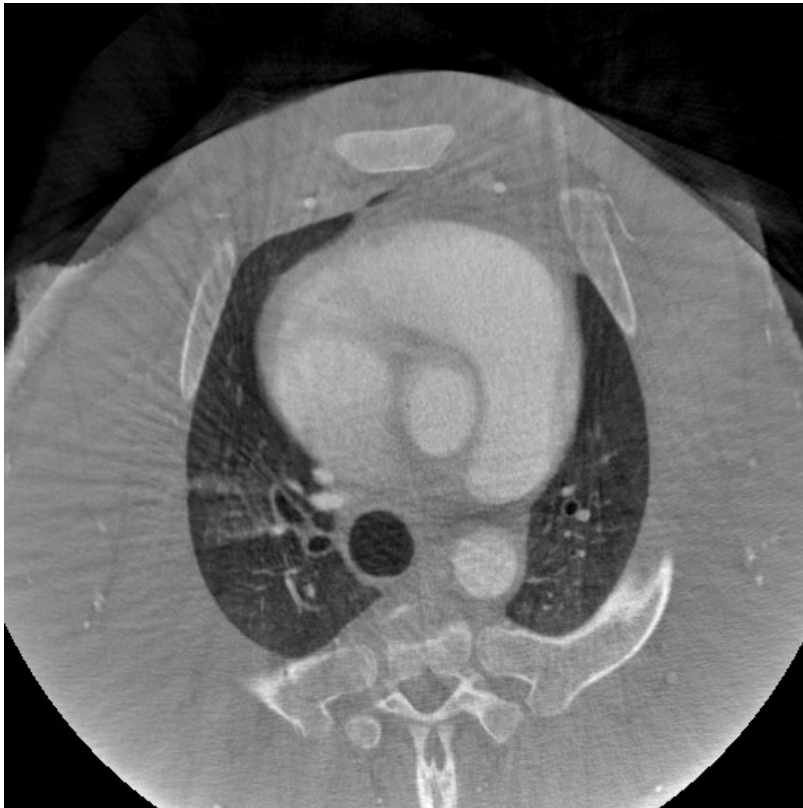


Original

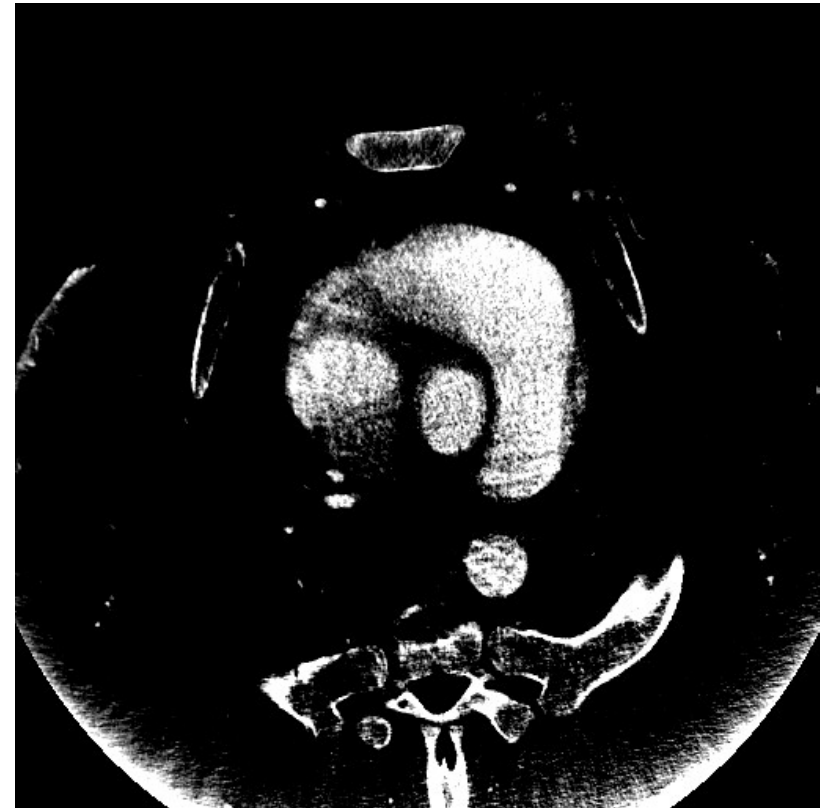


Gamma corrected ( $\gamma = \frac{1}{2.2}$ )

## Window and Level – Example



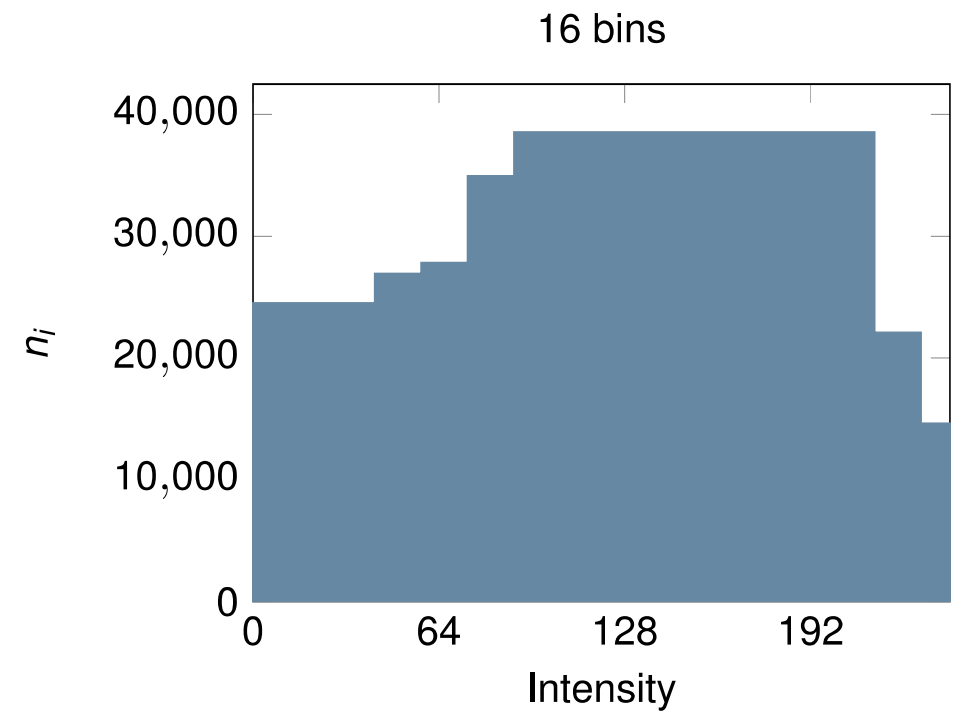
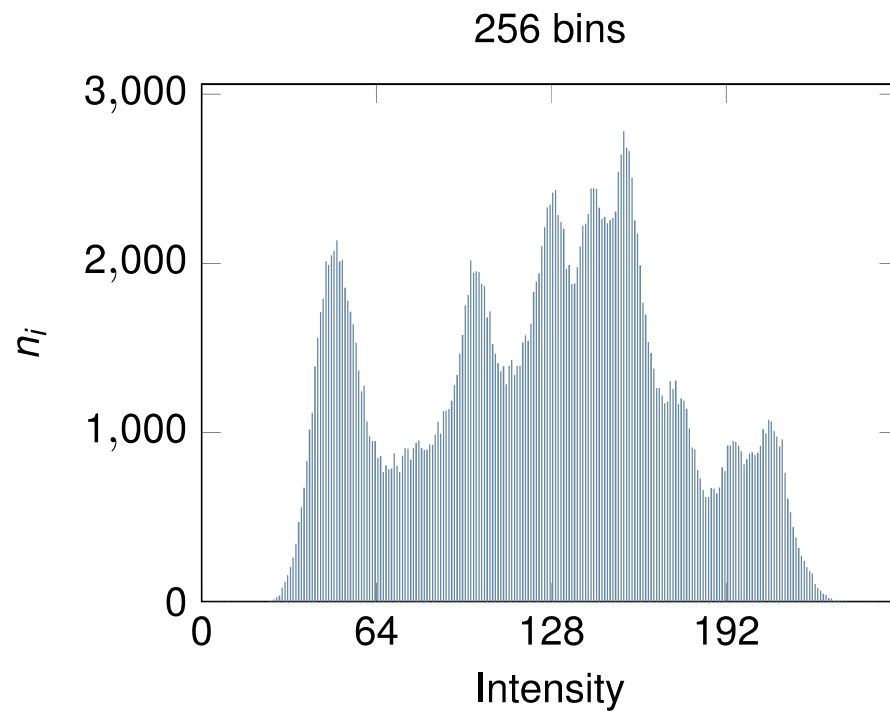
—1000 HU to 1000 HU



200 HU to 500 HU

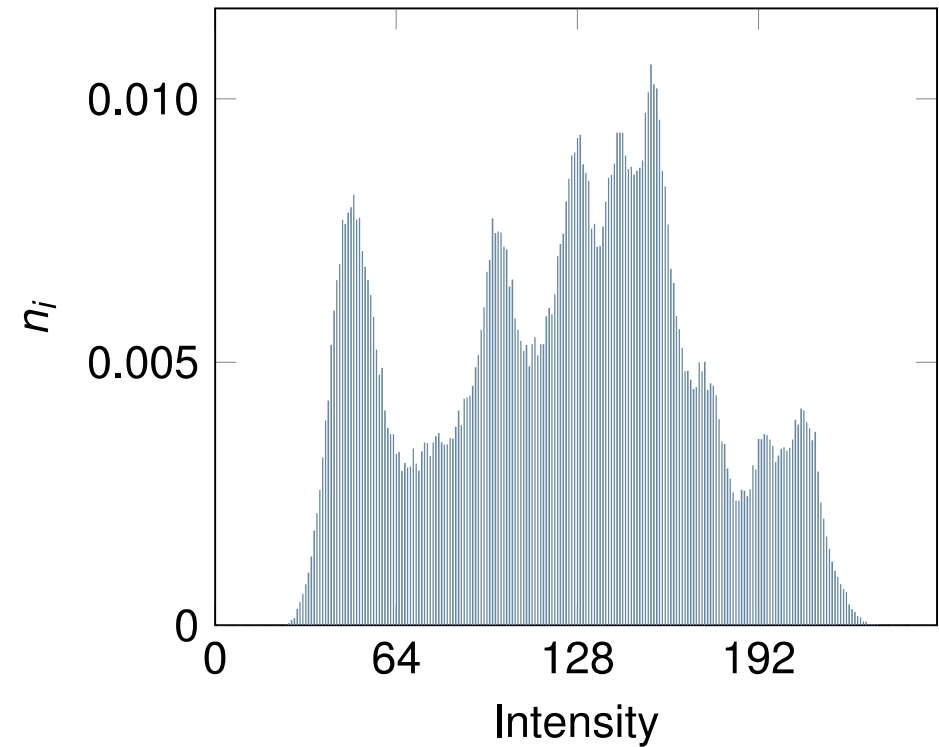
## Histograms of Images

Histograms show the distribution of intensity values, grouped into *bins*.



## Histogram Equalization

Intensities occupy only a small range in the histogram

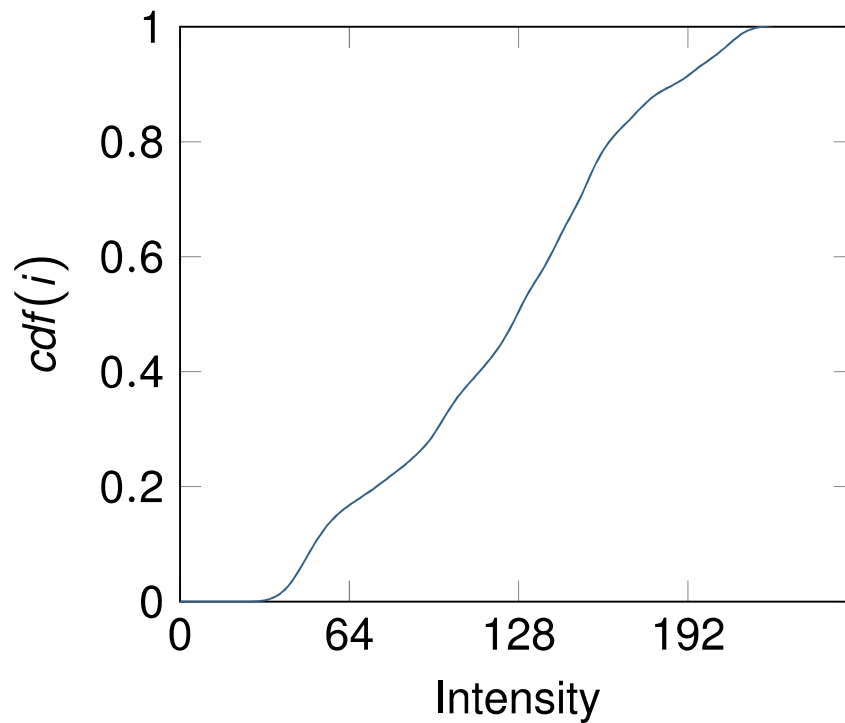




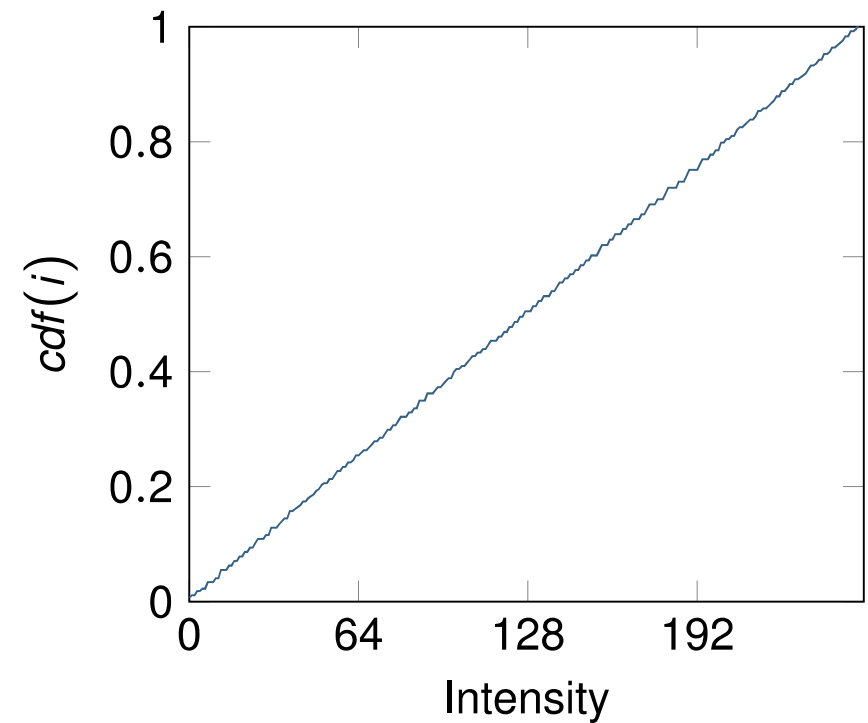
# Histogram Equalization

## Goal of Histogram Equalization

Achieve a uniform distribution of intensities  $\Rightarrow$  **linear CDF**

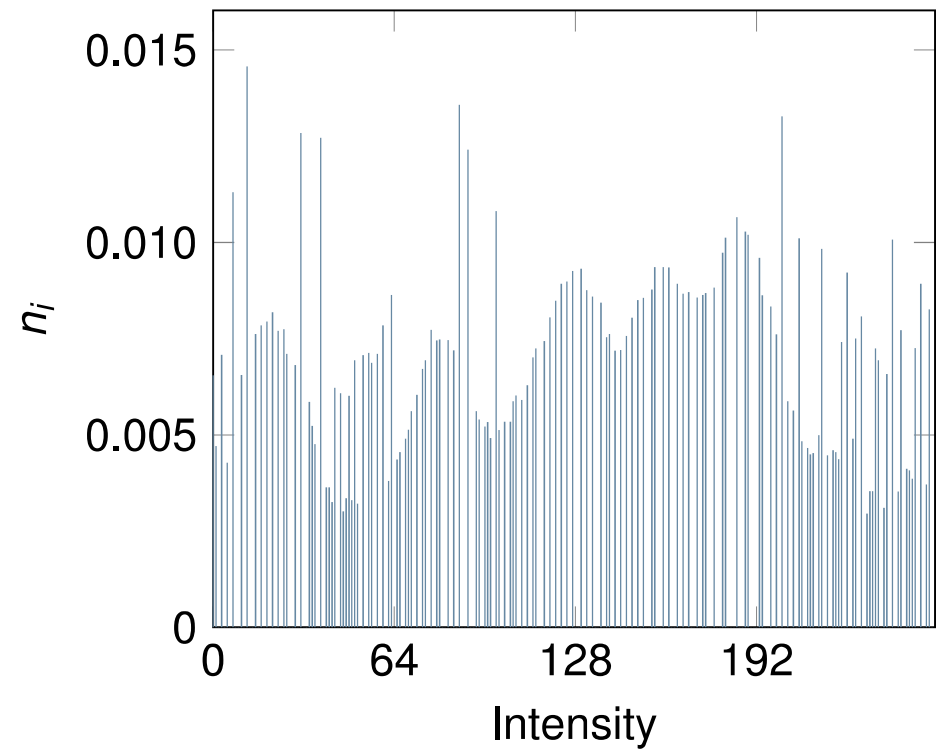


$\Rightarrow$



## Histogram Equalization

The histogram may contain gaps after histogram equalization!



## Histogram Equalization



Before



After

# Derivatives of Images

## Continuous derivative

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}$$

## Discrete derivatives

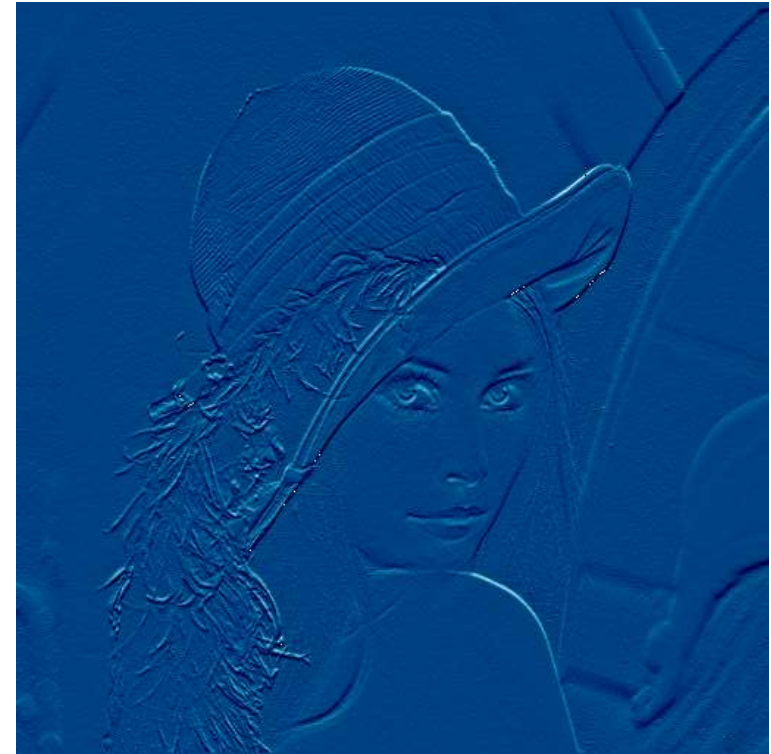
Several choices:

forward difference	$\Delta_x f(x) = f(x+1) - f(x)$
central difference	$\delta_x f(x) = f(x+1) - f(x-1)$
backwards difference	$\nabla_x f(x) = f(x) - f(x-1)$

## Derivatives of Images – Example



Derivative in x-direction



Derivative in y-direction

# Image Filtering

## Filters as operators

A filter  $\mathcal{H}$  can be applied on an image  $f$ :

$$\mathcal{H}(f(x, y)) = r(x, y)$$

It can have the same properties as in 1-D:

linearity	$\mathcal{H}\{\alpha \cdot f(x, y)\} = \alpha \cdot \mathcal{H}\{f(x, y)\}$
	$\mathcal{H}\{f_1(x, y) + f_2(x, y)\} = \mathcal{H}\{f_1(x, y)\} + \mathcal{H}\{f_2(x, y)\}$
shift-invariance	$\mathcal{H}\{f(x - x_0, y - y_0)\} = r(x - x_0, y - y_0)$

# Linear, Shift-invariant Filters

## Filter kernels

Linear, shift-invariant filters are characterized by **filter kernels**  $k$  and are applied to an image by a convolution.

## Convolution with a filter

Image  $f$ , filter kernel  $k$

$$\mathcal{H}\{f\}(x, y) = f * k = \sum_{i=-\frac{w_k}{2}}^{\frac{w_k}{2}} \sum_{j=-\frac{h_k}{2}}^{\frac{h_k}{2}} f(x-i, y-j) \cdot k(i, j)$$

## Efficient Filtering

### Convolution in the Fourier domain

For large images and / or filters:

1. Transform  $f$  and  $k$  to the Fourier domain

$$F = \mathcal{F}(f)$$

$$K = \mathcal{F}(k)$$

2. Like in 1-D, convolution then is a simple multiplication

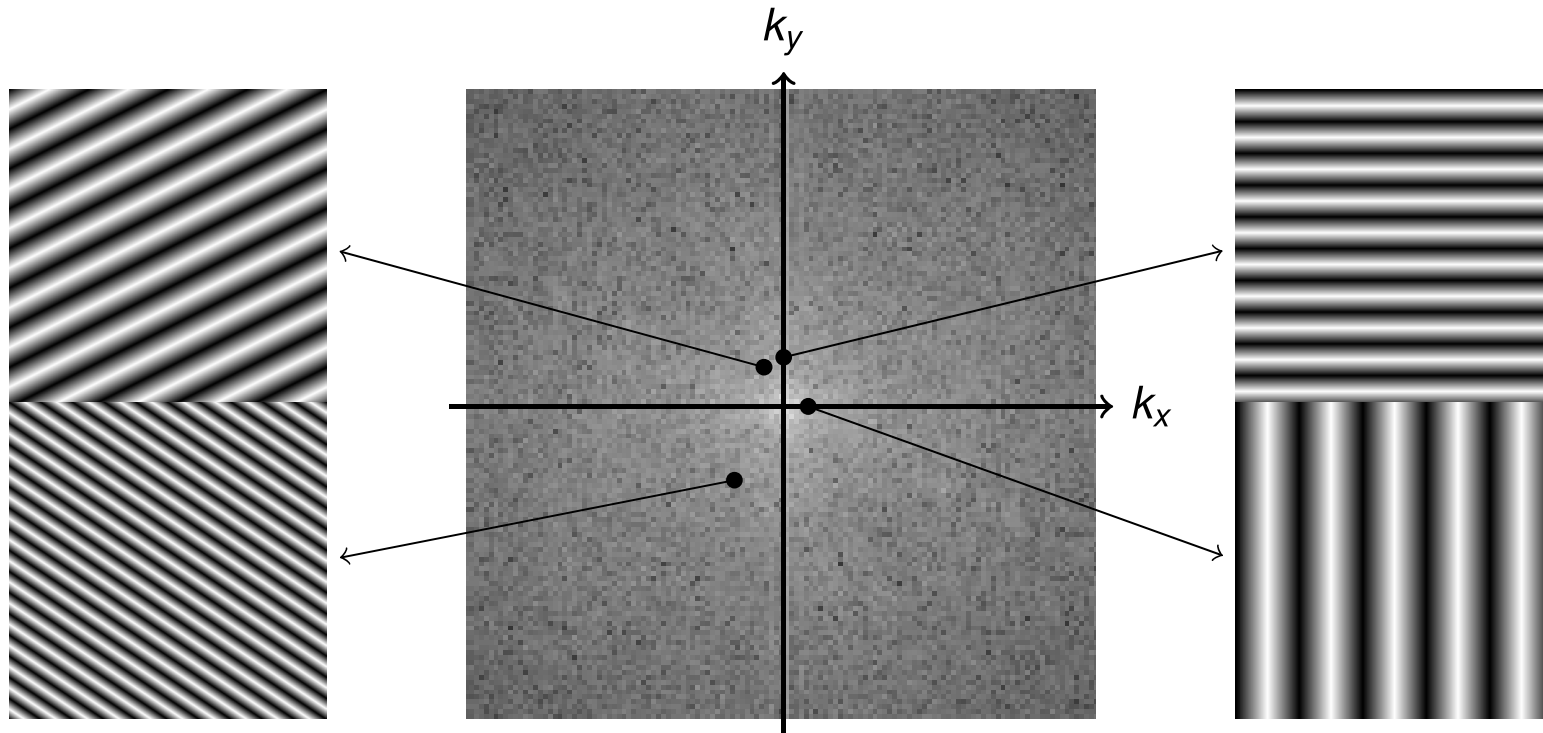
$$\mathcal{F}\{f * k\} = F \cdot K$$

3. Get the filtered image by transforming the result back to the spatial domain



## 2-D Fourier Space

- 2-D Fourier Space with phase patterns associated with some positions:

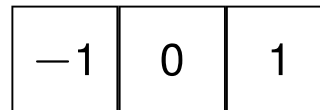


## Filter Kernel – Example: Central Difference

Derivatives are usually calculated by applying a filter.

The kernel can be constructed from the corresponding equation:

$$\delta_x f(x, y) = -1 \cdot f(x-1, y) + 0 \cdot f(x, y) + 1 \cdot f(x+1, y)$$



## Filter Kernels – Derivative Filters

### First order derivatives

$$\delta_x \longrightarrow \begin{array}{|c|c|c|} \hline -1 & 0 & 1 \\ \hline \end{array}$$

$$\delta_y \longrightarrow \begin{array}{|c|} \hline -1 \\ \hline 0 \\ \hline 1 \\ \hline \end{array}$$

### Second order derivative

$$\delta_x^2 \longrightarrow \begin{array}{|c|c|c|} \hline 1 & -2 & 1 \\ \hline \end{array}$$

$$\delta_y^2 \longrightarrow \begin{array}{|c|} \hline 1 \\ \hline -2 \\ \hline 1 \\ \hline \end{array}$$

# Smoothing Filters

Used to reduce noise in images.

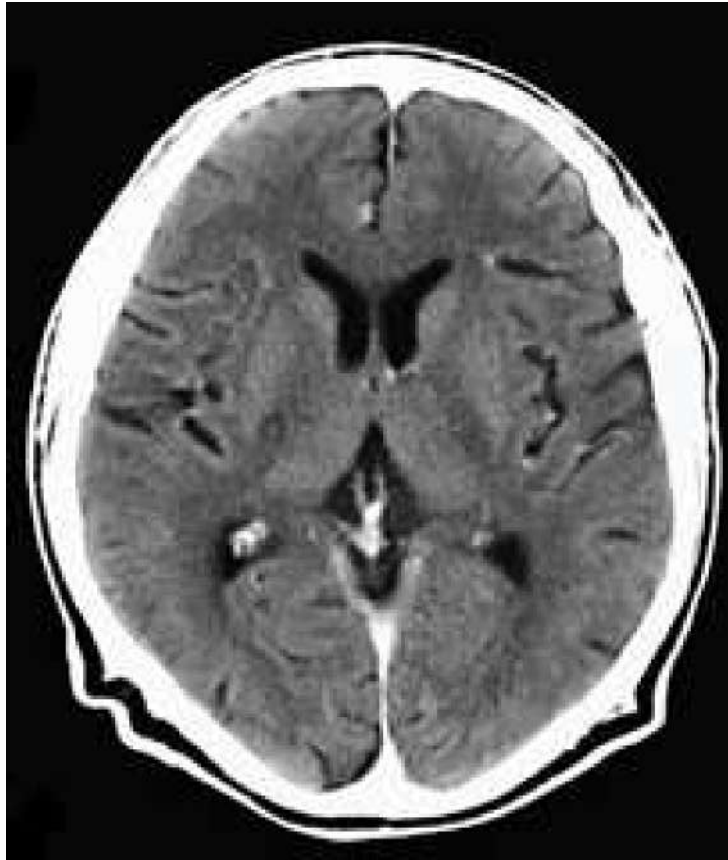
## Mean filter

$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$
$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$
$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$

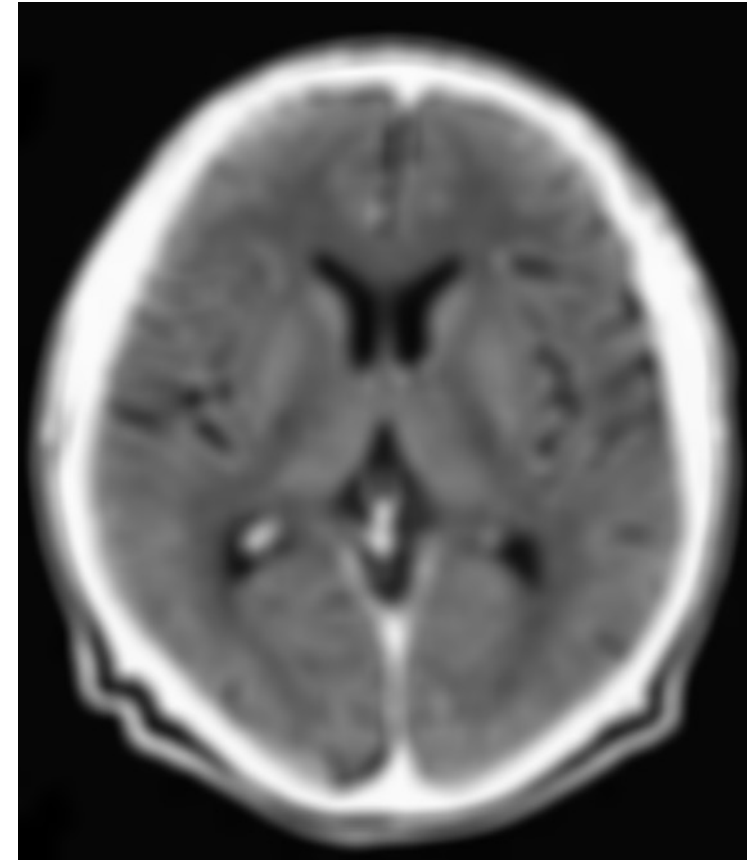
## Gaussian filter

$\frac{1}{52}$	1	1	2	1	1
	1	2	4	2	1
	2	4	8	4	2
	1	2	4	2	1
	1	1	2	1	1

## Gaussian Filtering



Original



Blurred