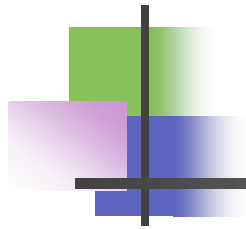


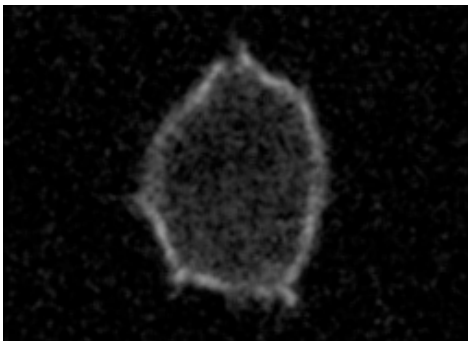
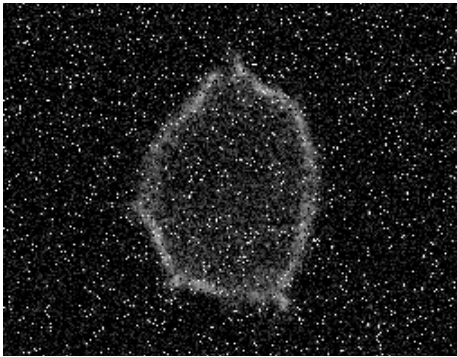
*Basic techniques in IP - part 2:*

# Image processing in the spatial / frequency / time domain.



Filtering - Neighborhood -  
Fourier space - Time series

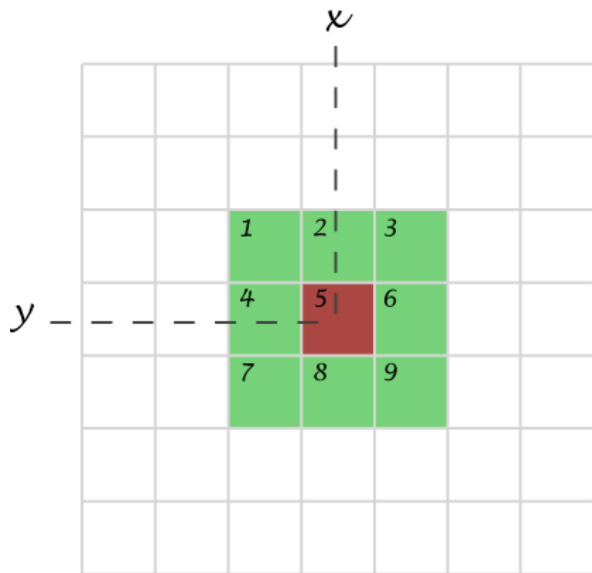
# I. Image processing in the spatial domain



- A. Introduction
  - Neighborhood
  - Operation on neighbors
- B. Spatial filters
  - Mean filter
  - Median filter
  - Edge detection

# A. Introduction

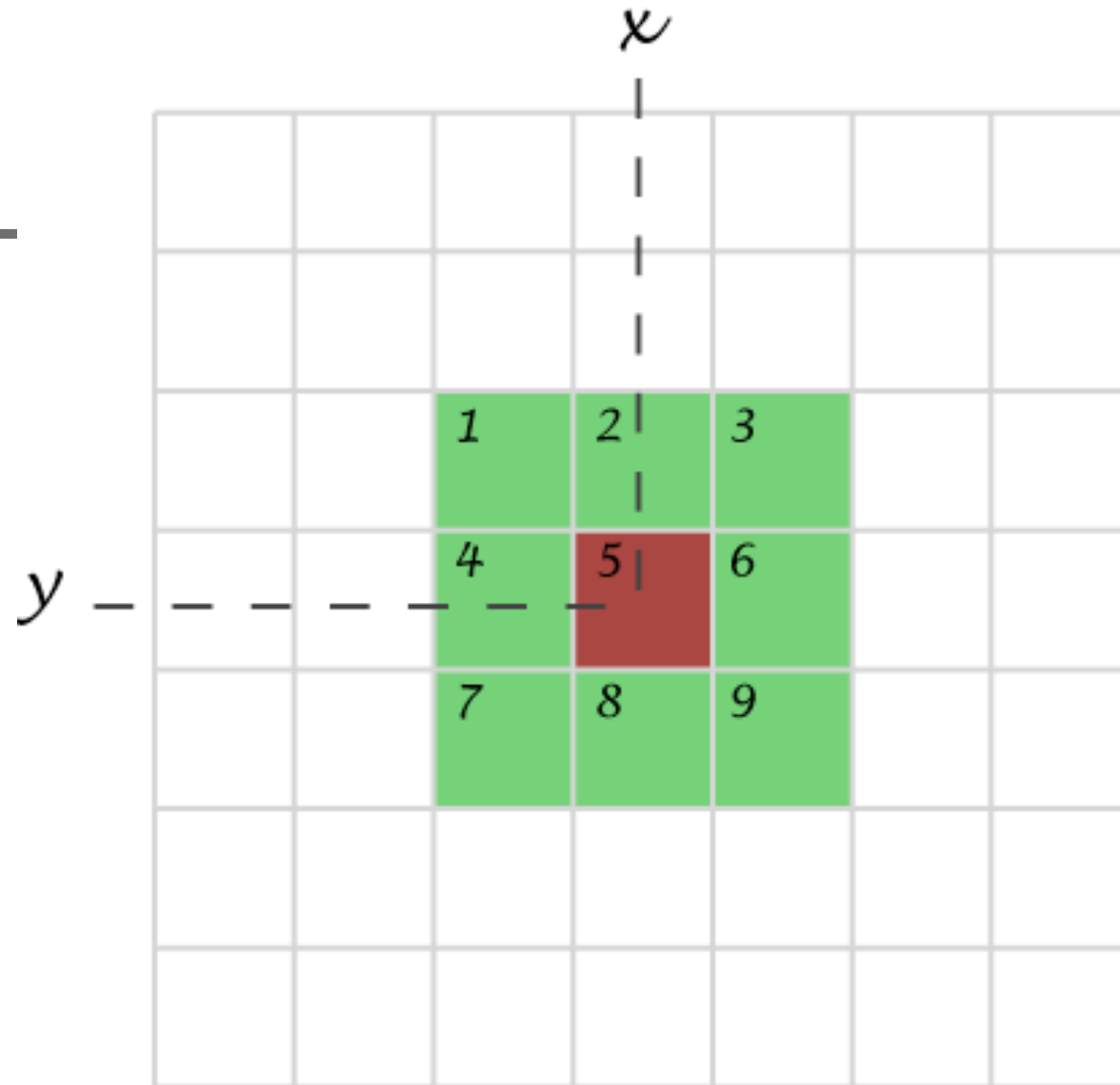
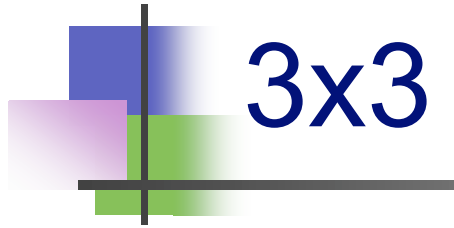
- Definition

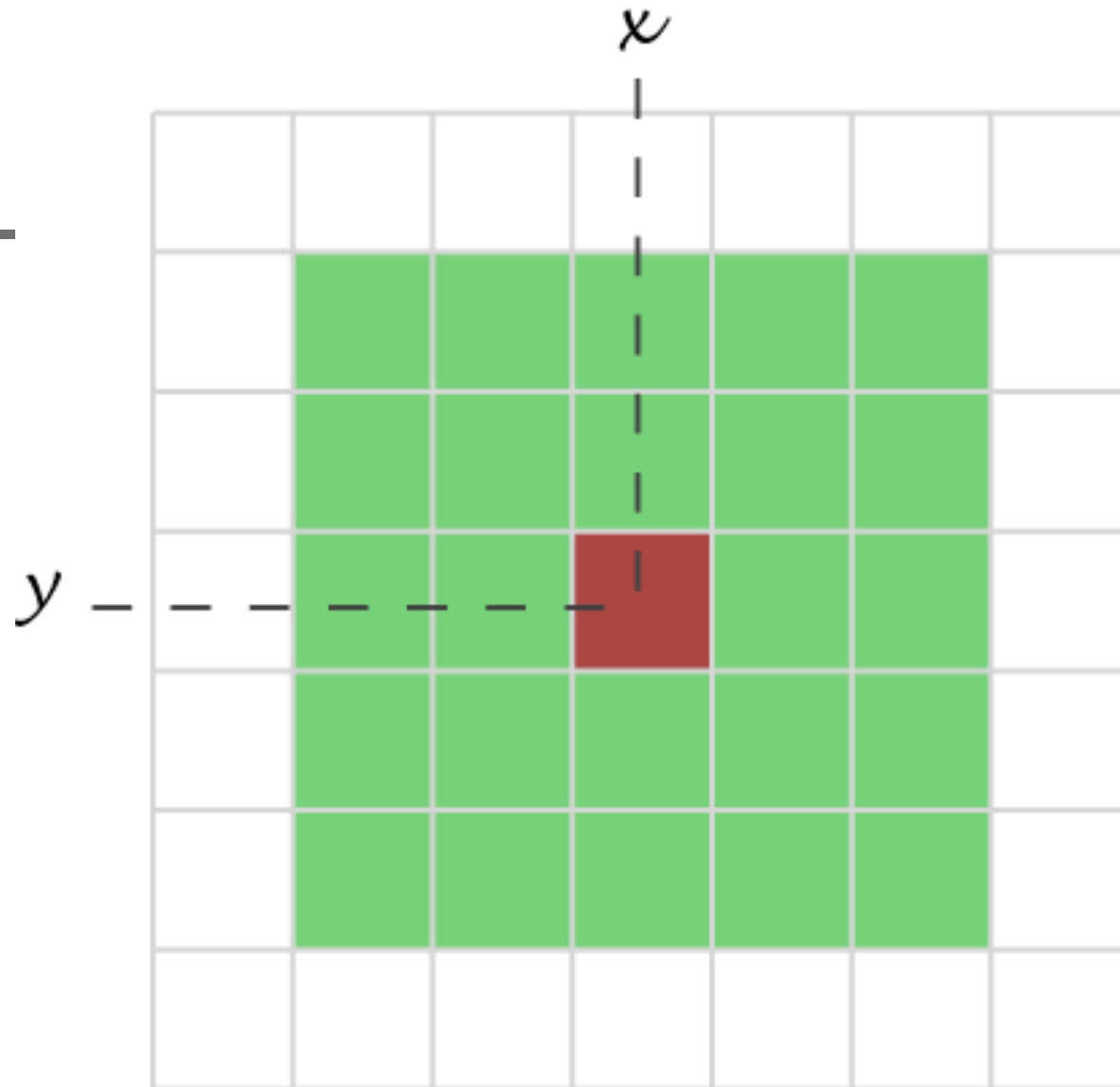
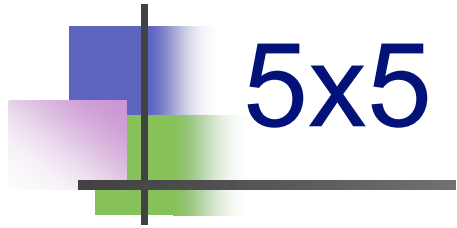


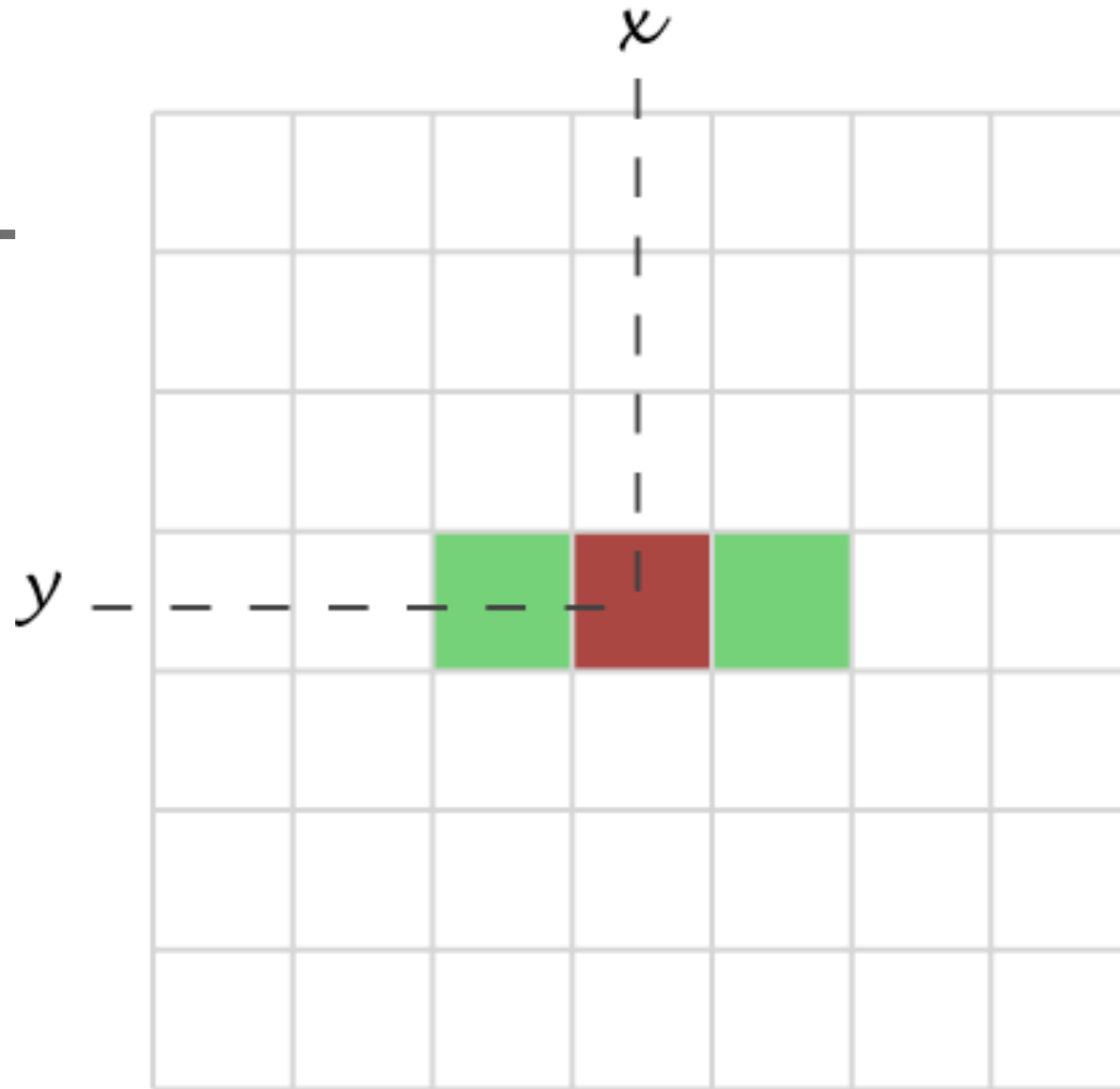
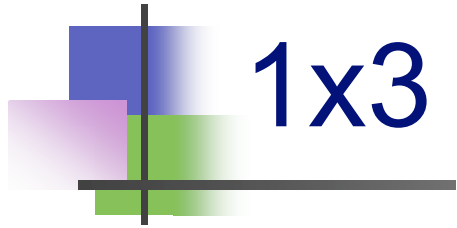
Neighborhood (or kernel):  
pixels that matters

“ Transformation or set of transformations where a new image is obtained by *neighborhood operations*.”

↳ The intensity of a pixel in the new image depend on the intensity values of “neighbor pixels”.

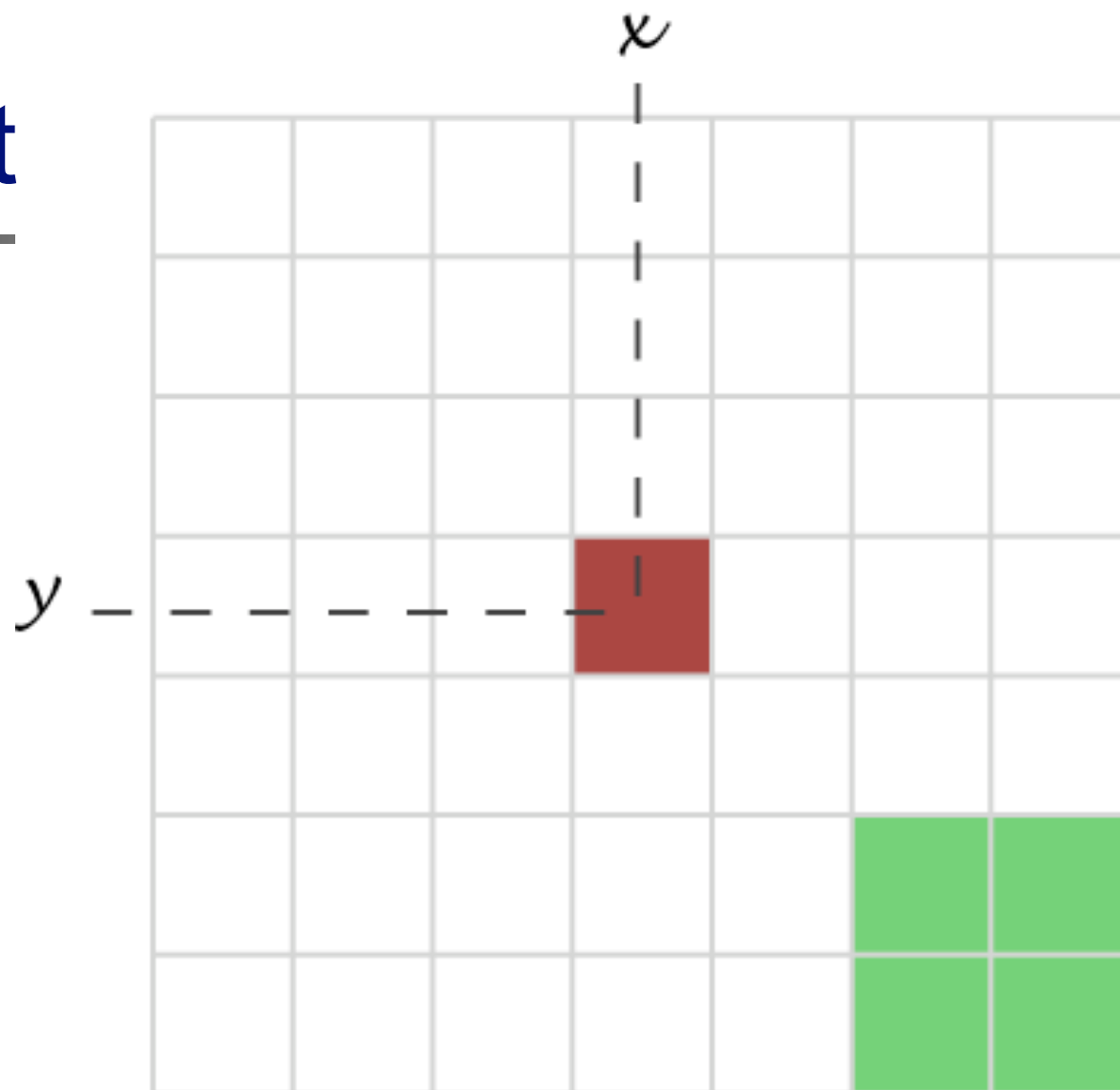




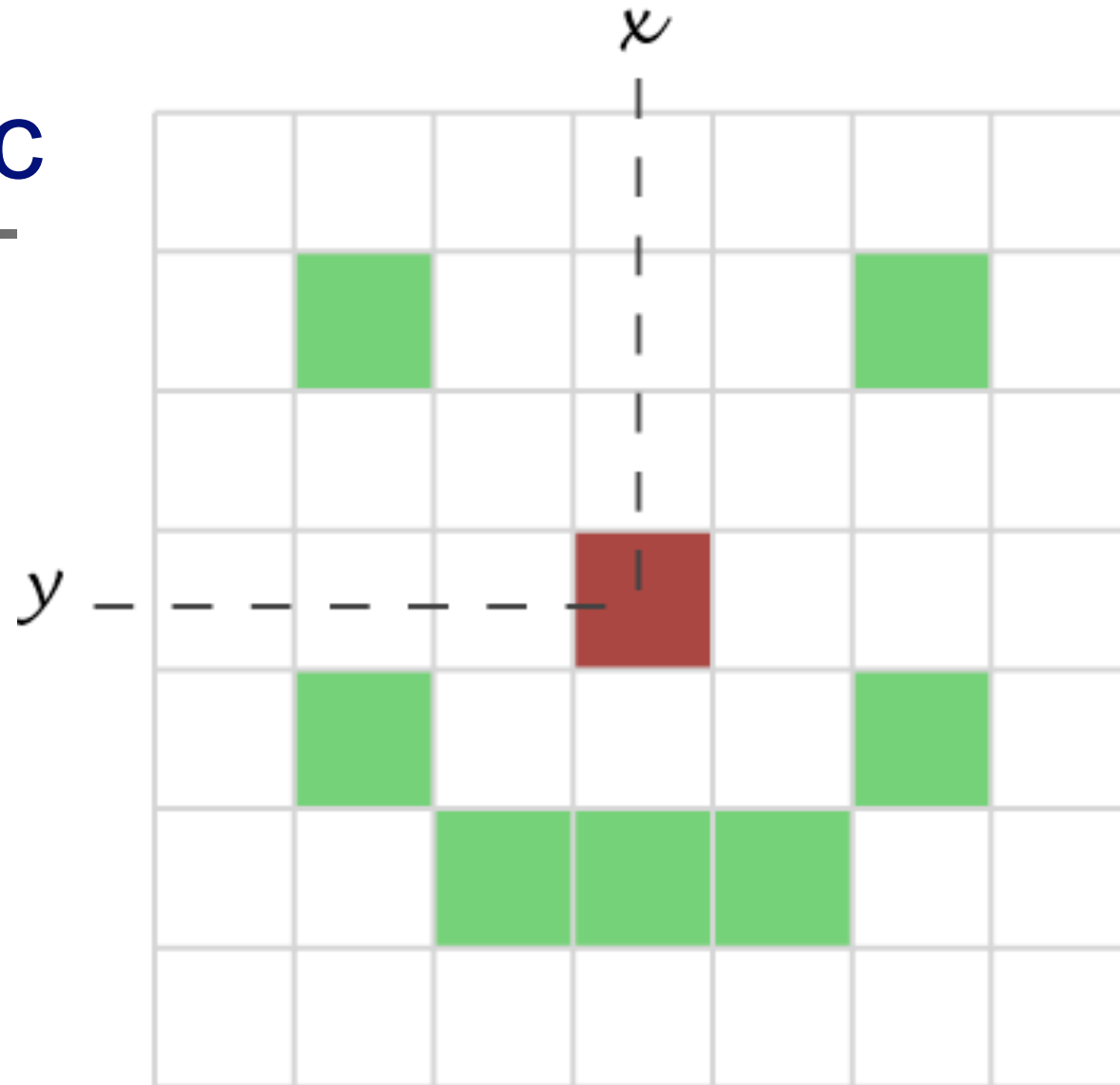




2x2  
shift



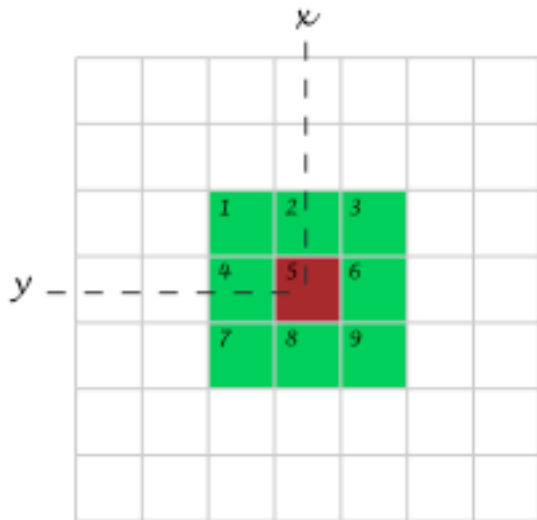




## B. The mean filter

Simplest filter: the value of a pixel is replaced by the intensity mean computed over neighbors pixels

$$a_i^* = \frac{1}{N_{\Omega}} \sum_{j \in \Omega} a_j$$



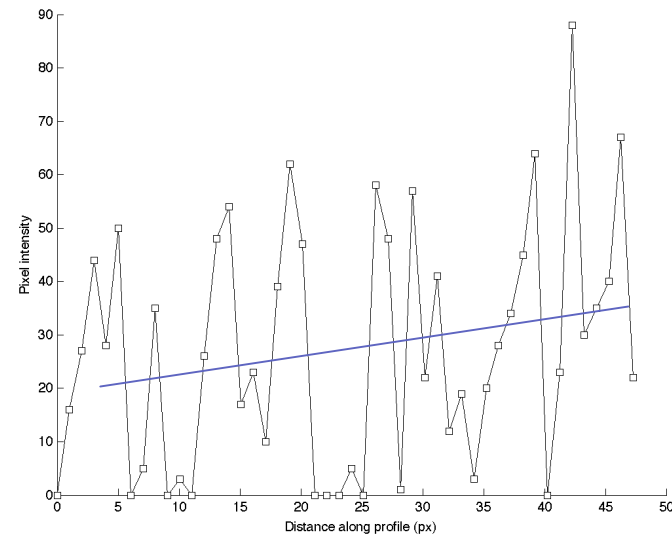
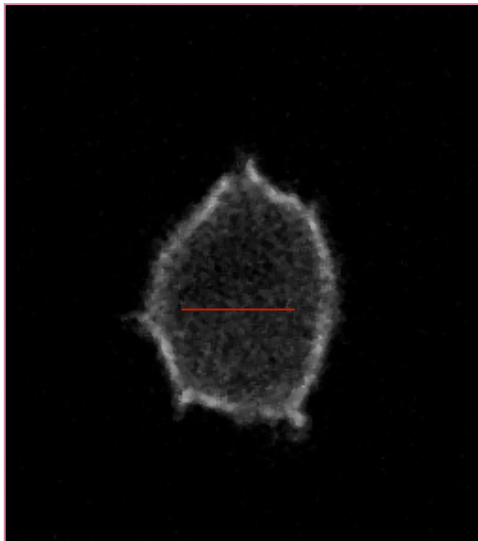
3x3 example:

$$a_i^* = \frac{1}{9} (a_1 + a_2 + a_3 + a_4 + a_5 + a_6 + a_7 + a_8 + a_9)$$

# The mean filter

what is it good for?

Noise removal - typically Gaussian noise.



(typ. Appears for strong labeling, short exposure time)



# The mean filter

## properties

---

Main property: low-pass filter

- kernel size influence
- number of successive applications

Cases where it fails

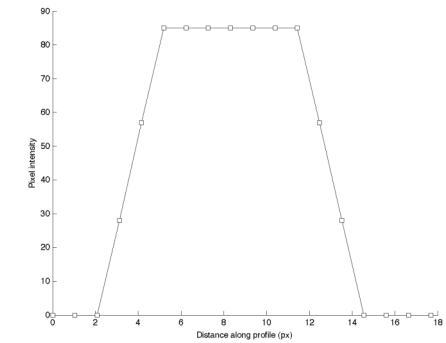
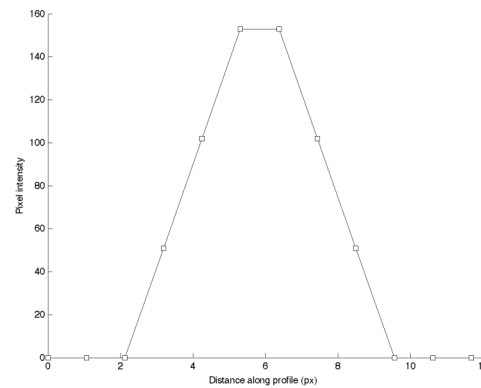
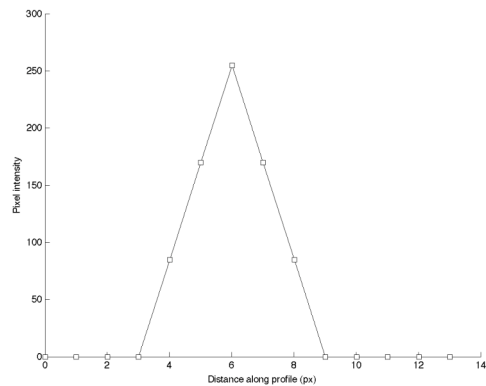
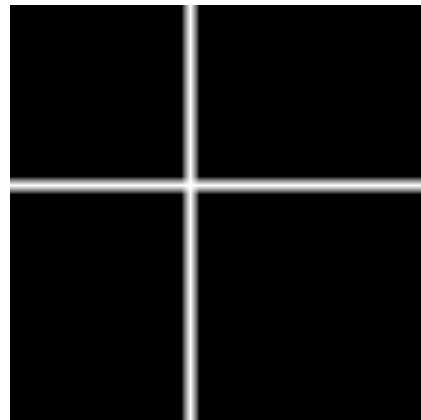
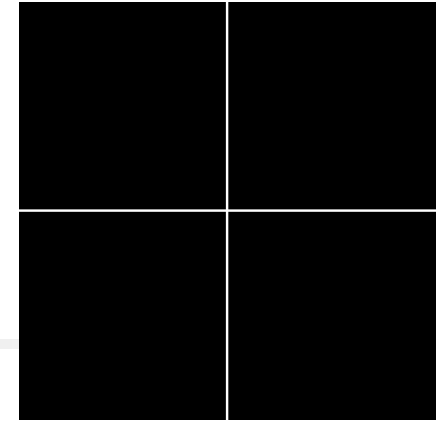
- salt & pepper noise

practical

# The mean filter

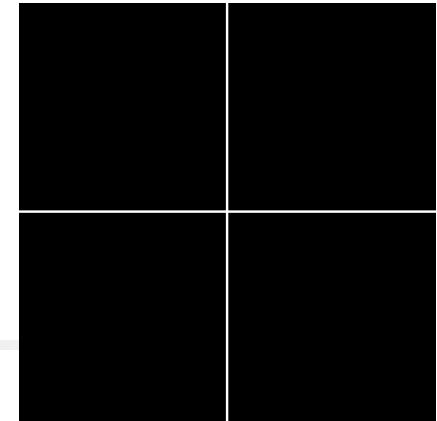
properties - kernel size

Zoom 10x

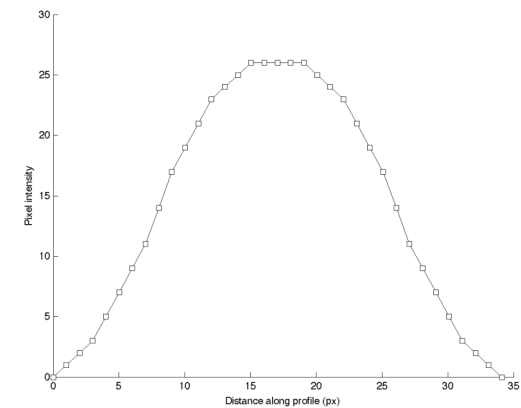
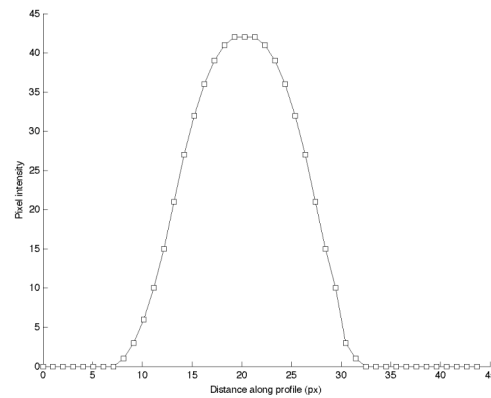
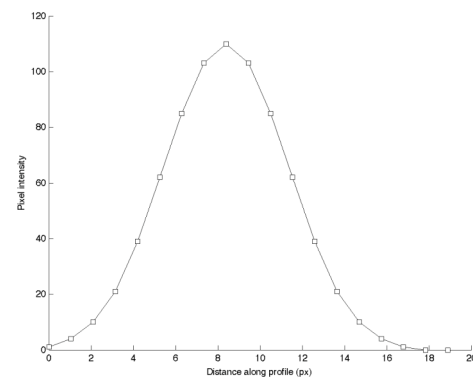
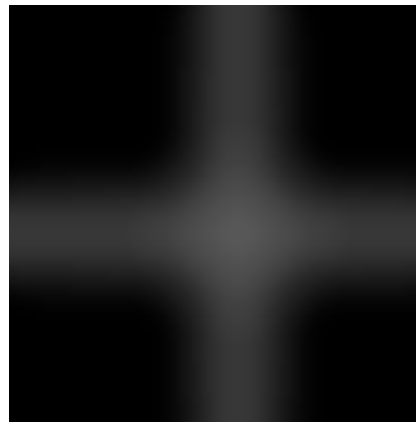
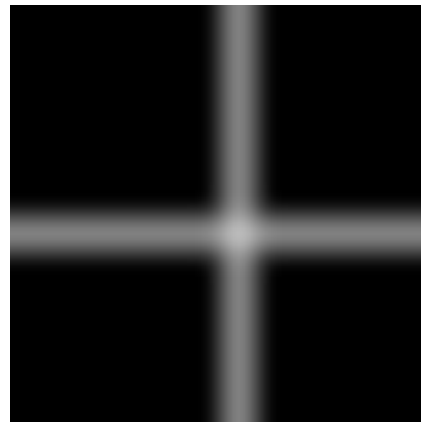


# The mean filter

properties - number of application



Zoom 10x



# The mean filter

properties - linear filtering

The mean filter is a linear filter:

“ *The new pixel value depend on a linear combination of neighbor pixel values*”

$$i_{x,y}^* = \sum_{i,j \in \Omega} \alpha_{ij} i_{x+i,y+j}$$

Neighbor pixels  
Filter coefficients

$\alpha_{1,1}$	$\alpha_{1,2}$	$\alpha_{1,3}$
$\alpha_{2,1}$	$\alpha_{2,2}$	$\alpha_{2,3}$
$\alpha_{3,1}$	$\alpha_{3,2}$	$\alpha_{3,3}$

→ another notation for 3x3 kernel



# The mean filter

## properties - linear filtering 2

---

Lot of properties:

- Suppose one image is made of a combination of images (e.g.  $I = a_1 \times I_1 + a_2 \times I_2$ , like in color combine or contrast change) → applying a linear filter to it is the same as applying it to all parts, then making the combination.
- When applying a succession of linear filters → order in which linear filters are applied does not matter.
- Mathematical framework underlying it → convolution (see later).





# The mean filter

## summary

---

- simplest filter - fast
  - is a linear filter
  - average noise, does not eliminate it
  - good against Gaussian noise
  - blur images - small details are lost in the average
  - smooth edges dramatically
  - trivial convergence
- } Low-pass filter



## C. The median filter

---

The value of a pixel is replaced by the *median* of the pixel intensity in neighbors pixels

Take neighborhood  
(e.g. 3x3)

5	112	86
235	88	211
137	233	108

Sort it

5  
86  
88  
108  
112  
137  
211  
233  
235

Take median

112

# The median filter

noise elimination

Original:

5	9	6	6	9	5	9
9	5	9	7	8	7	9
8	9	8	6	7	9	9
9	9	7	200	9	6	9
6	5	8	6	9	6	7
9	7	9	9	8	6	7
7	9	5	6	7	6	6

outlier

Median filtered:

0	5	6	6	6	7	0
5	8	7	7	7	9	7
8	9	8	8	7	9	7
6	8	8	8	7	9	6
6	8	8	9	8	7	6
6	7	7	8	6	7	6
0	7	6	6	6	6	0

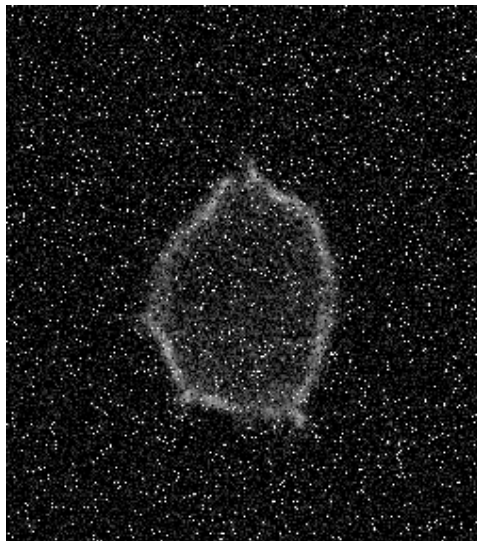
The outlier value has completely been removed from the dataset

# The median filter

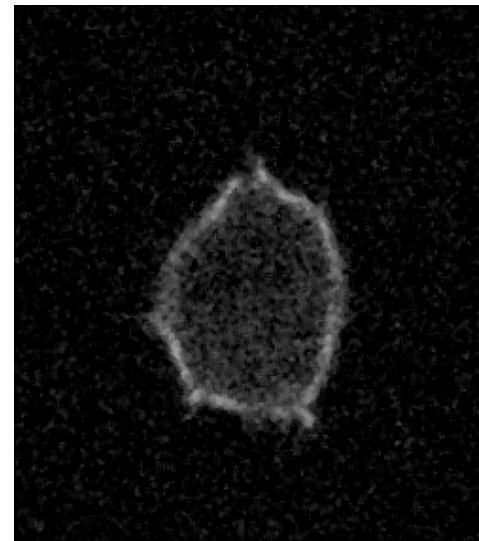
what is it good for?

“Salt & pepper” noise removal

Original:



Median filtered:



(typ. Appears for very weak labeling)



# The median filter

## properties

---

- Typically good for “Salt & pepper” noise removal
- *Eliminates* noise
- Slower than mean and similar (not such a problem anymore)
- Not linear
- Edge-preserving
- Non-trivial convergence



## D. The Sobel filter

---

Combination of 2 filters:

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

$g_x$

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

$g_y$

$$\text{output} = \sqrt{g_x^2 + g_y^2}$$

# The Sobel filter

What does it do?

(on  $g_x$  only)

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

Uniform image:

1	1	1	1	1
1	1	1	1	1
1	1	1	1	1
1	1	1	1	1
1	1	1	1	1

→ 0

Edge:

1	1	1	0	0
1	1	1	0	0
1	1	1	0	0
1	1	1	0	0
1	1	1	0	0

→ 4

**Edge detector!**



# The Sobel filter

## edge detector

---

$g_x$  detects edges in the X direction

$g_y$  detects edges in the Y direction

The output is ~ the edge gradient magnitude

$$\text{output} = \sqrt{g_x^2 + g_y^2}$$





Relax...

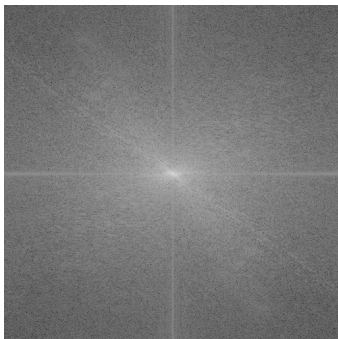
---

# II. Image processing in the Fourier domain

Before:



After:



- A. The Fourier transform
  - Introduction
  - Examples and measurements
- B. The *inverse* Fourier transform
  - Principles
  - Filtering



# A. The Fourier transform

---

- The Fourier transform is a way to obtain a new *representation* of the data.
- It is best suited for data with *repetitive patterns* and highlights these patterns.
- It's mathematical definition (1D, discrete) is:

$$\hat{I}_k = \sum_n I_n e^{-2i\pi k / n}$$

(we don't have to care too much)

- It is best explained with a music player.

# The Fourier transform

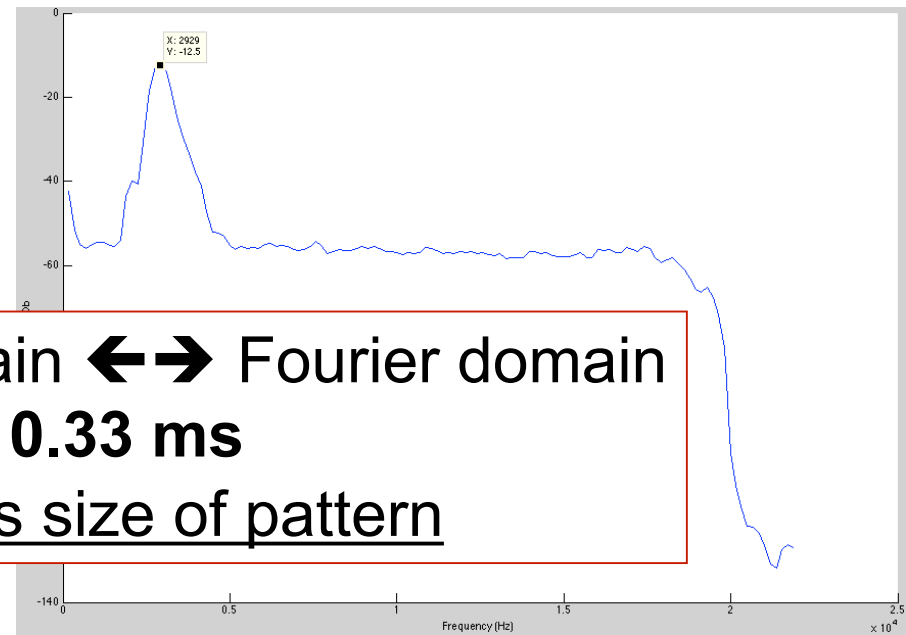
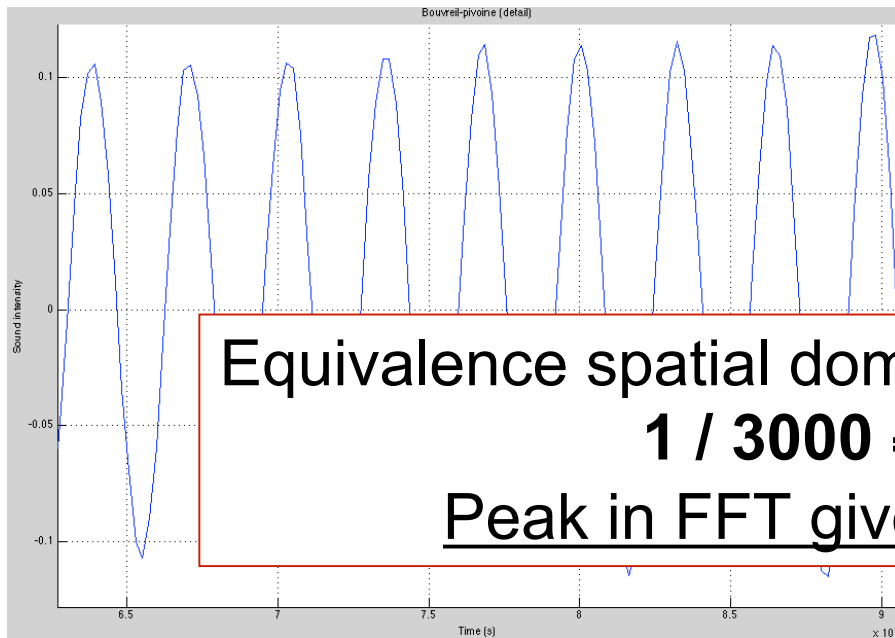
## the Bouvreuil-Pivoine

Bird song.

Delay between peaks:  
~ 0.35 ms

FFT of this looks like:

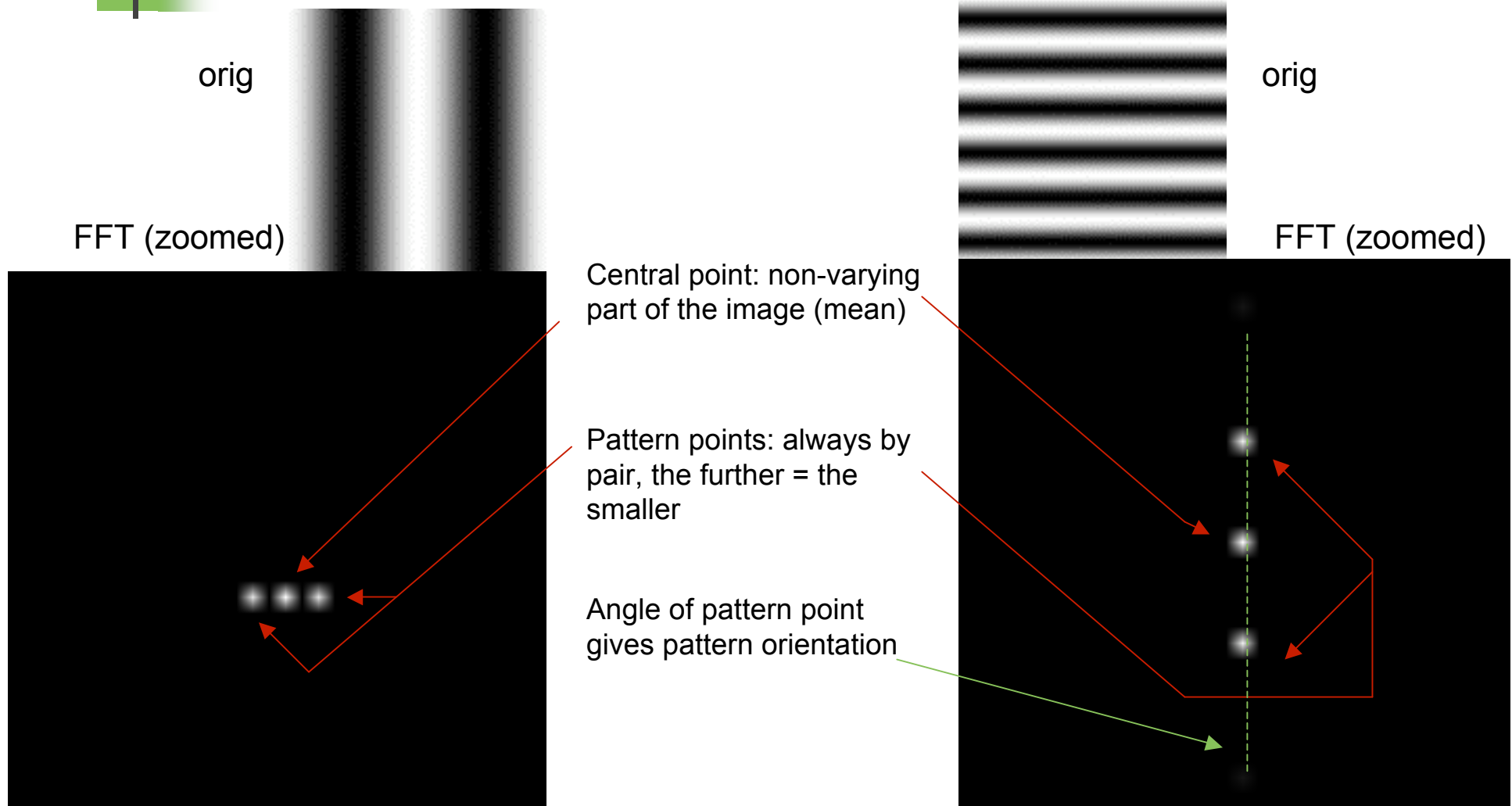
Peak in FFT:  
~ 3 kHz



Equivalence spatial domain ↔ Fourier domain  
 $1 / 3000 = 0.33 \text{ ms}$   
Peak in FFT gives size of pattern

# The Fourier transform

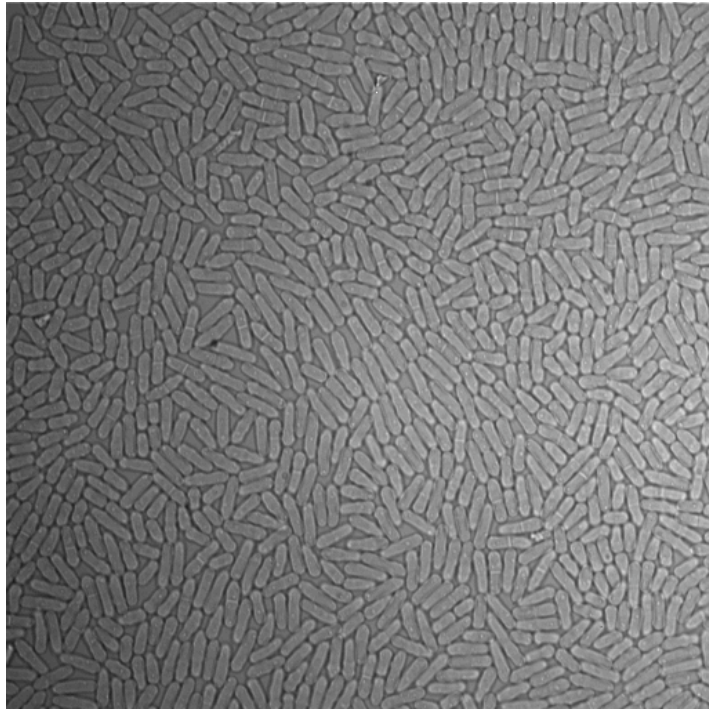
## in 2D (images)



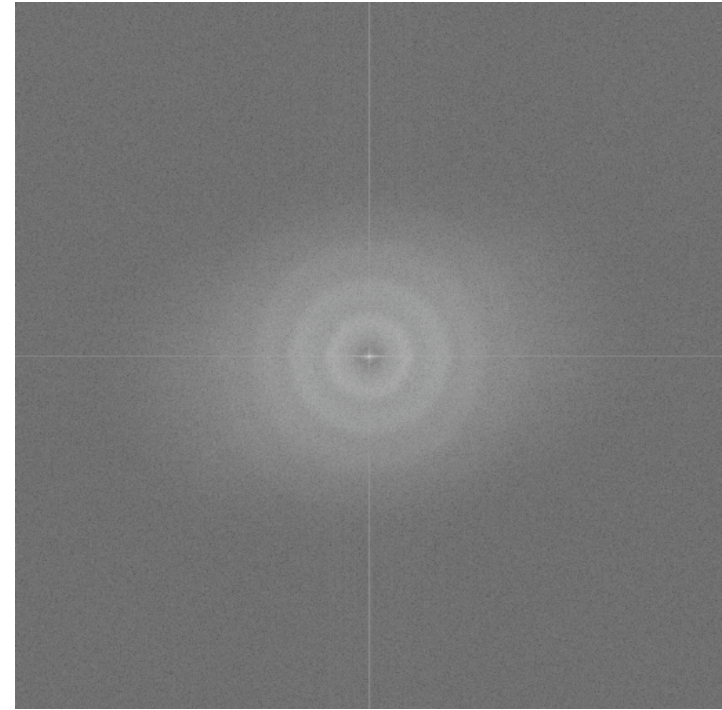
# The Fourier transform

real images

... are rarely that clear



S pombe cells (*Tolic lab*)



FFT

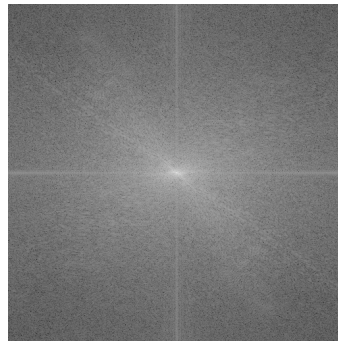
# B. The *inverse* Fourier transform

Because the Fourier image and the real image contain the same amount of info, it is possible to generate a real image from its Fourier representation:

Before:



After:



Changed her mind:





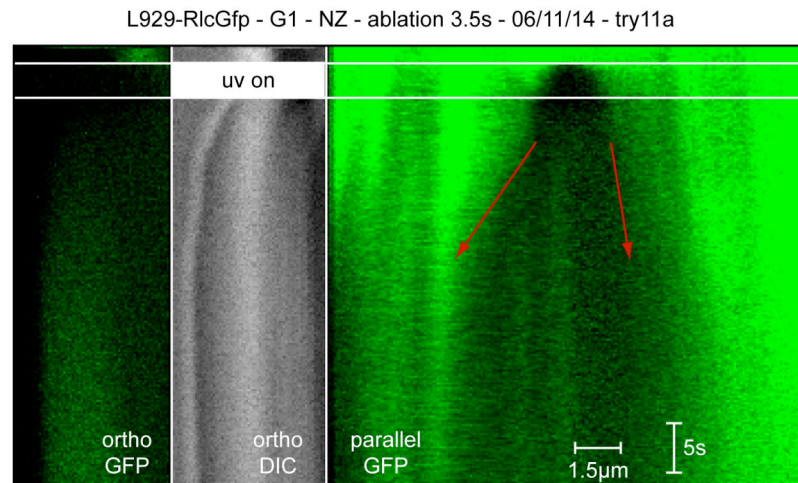
Relax...

---



# III. Time domain

Dealing with multiple images files (a.k.a. *stacks*):  
timelapse movies, 3D stacks, ...



total speed of  
cortex movement:  
17.0  $\mu$ m/mn

- Intensity over time
- Kymographs



# Motion blur

---

Motion blur = 1x10 neighborhood + average



Blackboard