

# MICROSCOPIA OTTICA IN BIOLOGIA CELLULARE [675SM]

## MICROSCOPY IN CELL BIOLOGY –

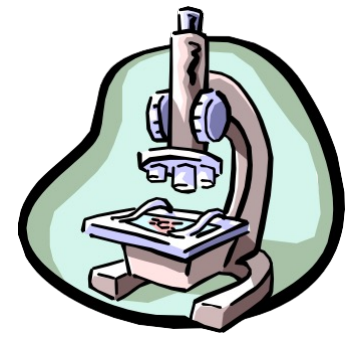
aa 2022/2023, 2<sup>nd</sup> semester

Aula 5A, edificio H2bis ~~14-17~~ 15-18

Agnes Thalhammer  
agnes.thalhammer@units.it

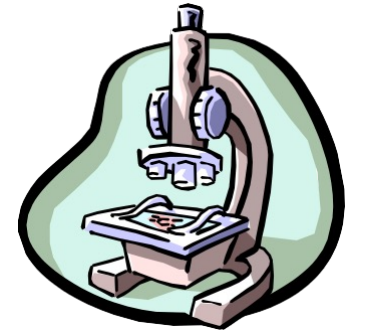


| date     | lesson/lab | aula                    | time                             |
|----------|------------|-------------------------|----------------------------------|
| 01/03/23 | intro      | aula5A, H2bis           | 14-15 (GF and BT),<br>16-17 (NS) |
| 08/03/23 | lesson1    | aula5A, H2bis           | 15-18                            |
| 15/03/23 | lab1       | sala microscopia F2, C1 | 15-18                            |
| 22/03/23 | lesson2    | aula5A, H2bis           | 15-18                            |
| 29/03/23 | lab2       | sala microscopia F2, C1 | 15-18                            |
| 05/04/23 | lesson3    | aula5A, H2bis           | 15-18                            |
| 19/04/23 | lesson4    | aula5A, H2bis           | 15-18                            |
| 26/4/23  | lesson5    | aula5A, H2bis           | 15-18                            |
| 3/5/23   | lab3       | CIMA center             | 15-18                            |
| 10/5/23  | lab4       | aula5A, H2bis           | 15-18                            |



# Live imaging microscopy

# LIVE IMAGING

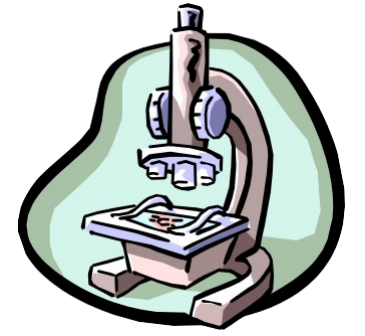


- Allows study of dynamic processes in real time instead of snap-shot of one timepoint
- Results in movie instead of single image
- Dynamic processes of subcellular compartments, cells, cellular networks, organs, entire animals

## Examples for dynamic processes of interest

- Cell migration
- Morphological changes
- Physiological parameters (ion concentrations, vesicle dynamics)

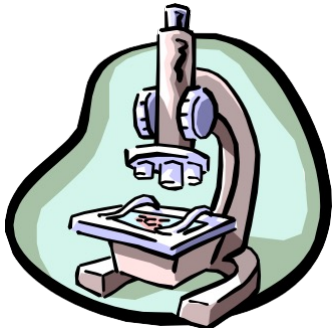
# LIVE IMAGING - CHALLENGES



## Keeping cells alive

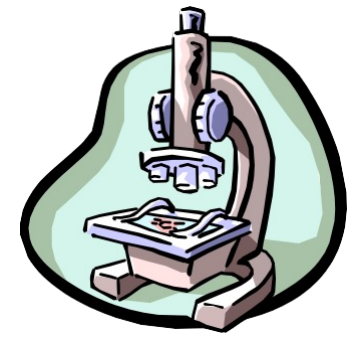
- physiological environment
  - Temperature
  - Ionic and osmotic environment
  - Oxigenation
  - pH (ev. CO<sub>2</sub> concentration for carbonate buffer)
- Mechanical stress
- Phototoxicity
- Chemical toxicity of dyes

# LIVE IMAGING - INCUBATION



| Cell Health Parameter | Short-Term Study      | Longer-Term Study                                                                                                  |
|-----------------------|-----------------------|--------------------------------------------------------------------------------------------------------------------|
| pH                    | HEPES Buffer          | CO <sub>2</sub> Incubator                                                                                          |
| Temperature           | Stage Warmers         | Objective Lens Warmers<br>Cell chambers with integrated heating elements<br>Enclose microscope within a heated box |
| Humidity              | Open cell chambers    | Tightly sealed cell chambers                                                                                       |
| Oxygenation           | Large volume of media | Media changes during study<br>Large media volume                                                                   |
| Osmolarity            | Sealed chamber        | Enclose system within a humidified chamber                                                                         |

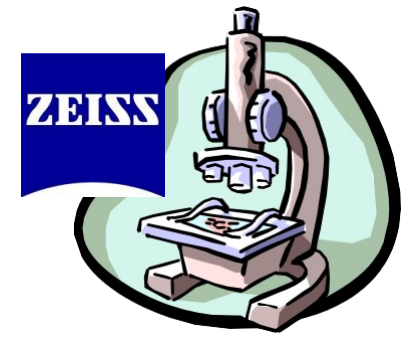
# LIVE IMAGING - INCUBATION



Incubation chambers can surround the microscope stand or just the cell dish. Both can maintain cells at the correct temperature for long periods of time.



# CONTRASTING TECHNIQUES



Optical Techniques to Enhance contrast:

Differential Interference Contrast (DIC)

Phase Contrast

Darkfield

Polarized Light

Fluorescence

Transmitted Light Techniques in Live-Cell Imaging

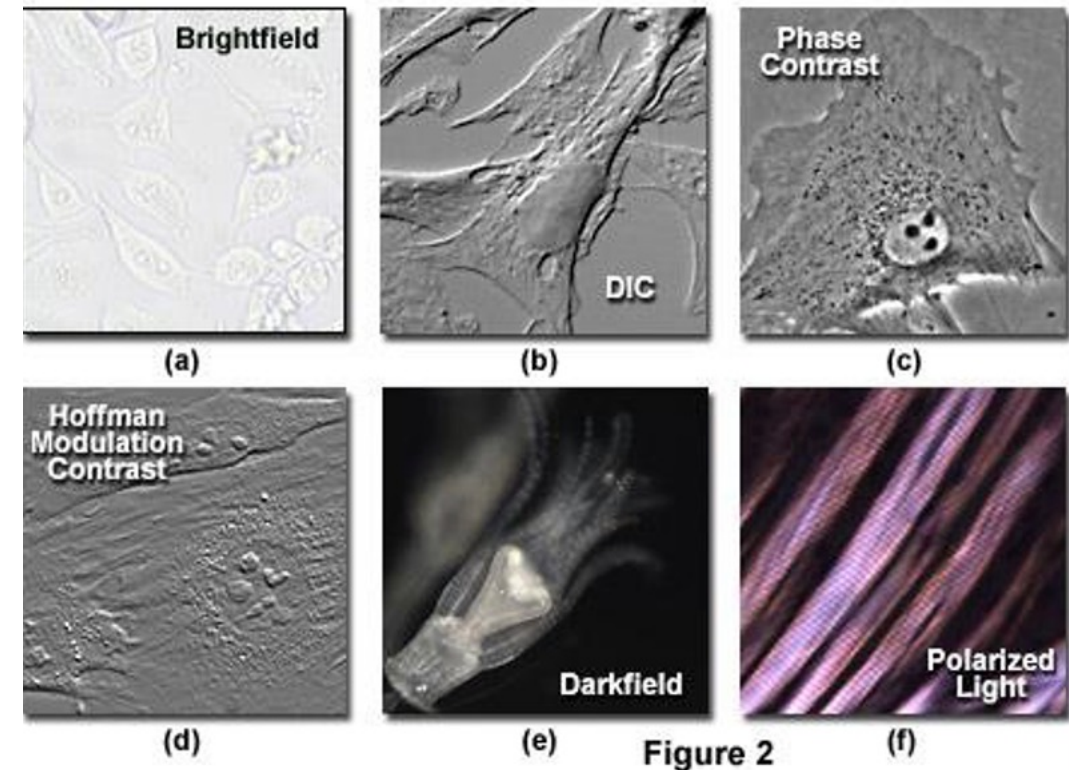
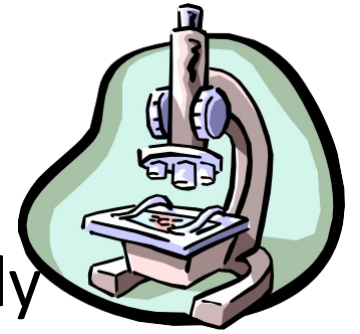


Figure 2

# PHOTOTOXICITY

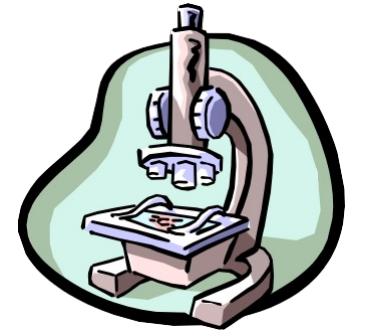


The **central** challenge in live cell imaging – Cells are inherently photosensitive!

- Formation of free radicals by fluorophores only exacerbates the problem
- Free radical formation cannot be prevented, only managed by careful tuning of imaging parameters to reduce light exposure

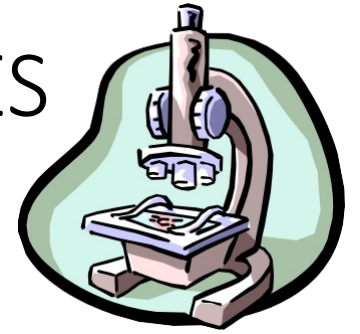
Live cell imaging requires “gentle imaging” - short exposures of low intensity illumination

# FLOURESCENCE LABELLING OF LIVE CELLS



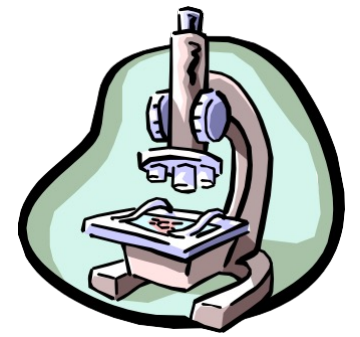
- Detect specific molecules (immunocytochemistry, fluorescent proteins)
- Must keep the specimen transparent
- Targeting live cell (membrane permeability)
- Eventually detect multiple molecules at the same time

# MINIMIZING DAMAGE WHILE GETTING GOOG IMAGES



- Resolution vs Speed vs Sensitivity
- Light Gathering Power and Focus Maintenance
- Which immersion media
- Correct coverslip thickness
- 3D Imaging

# RESOLUTION-SPEED-SENSITIVITY



## – Spatial Resolution

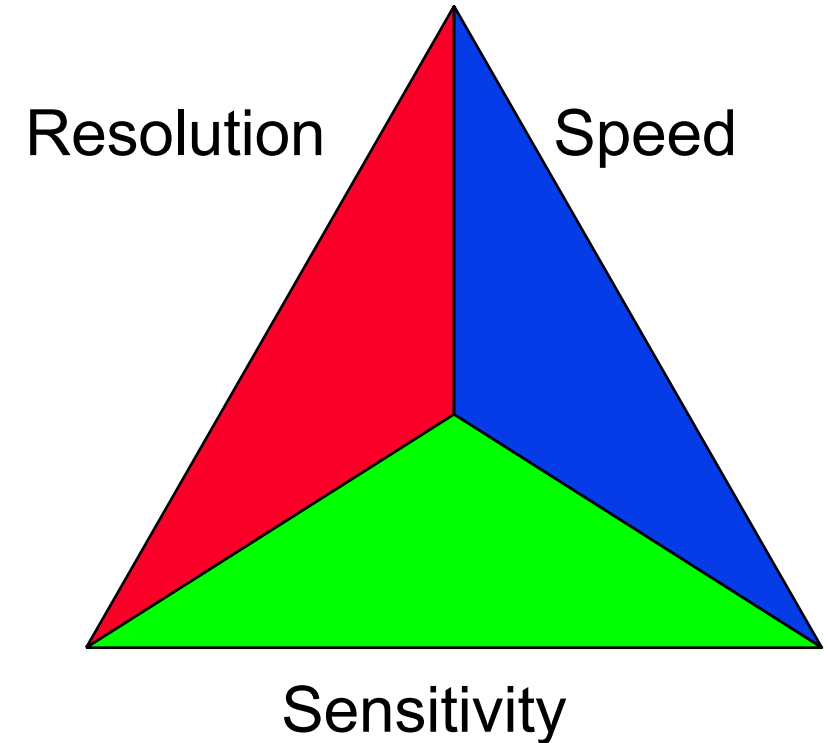
- Depends on application ? optical magnification and lens apperture
- Smaller and more pixels = higher resolution

## – Sensitivity

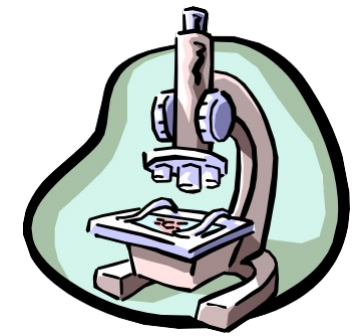
- Larger pixels = higher sensitivity
- Higher sensitivity = shorter exposure time
- Shorter exposure time = higher frame rate
- Larger pixels = lower resolution

## – Speed

- Less pixels = faster readout
- Faster Readout = higher framerate
- Less pixels = lower resolution
- requires fast readout and high sensitivity



# CORRECTION IMMERSION MEDIA



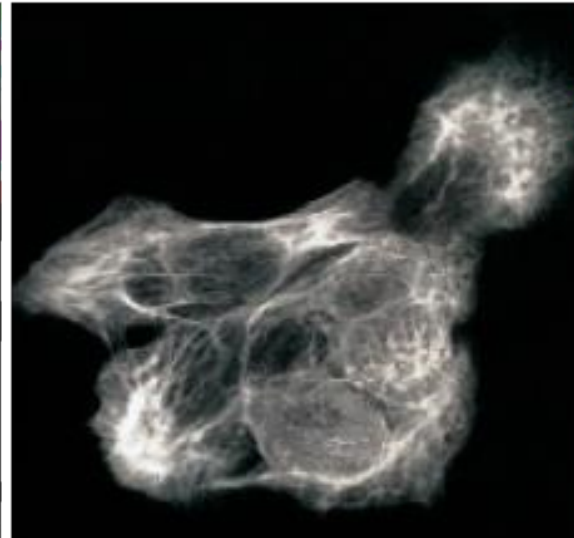
oil immersion objective

NA=1.4



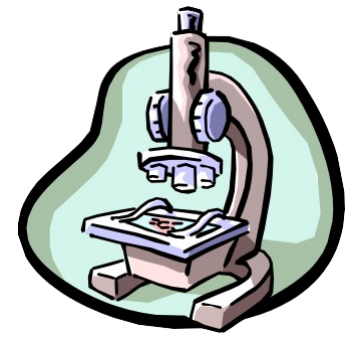
water immersion objective

NA=1.2

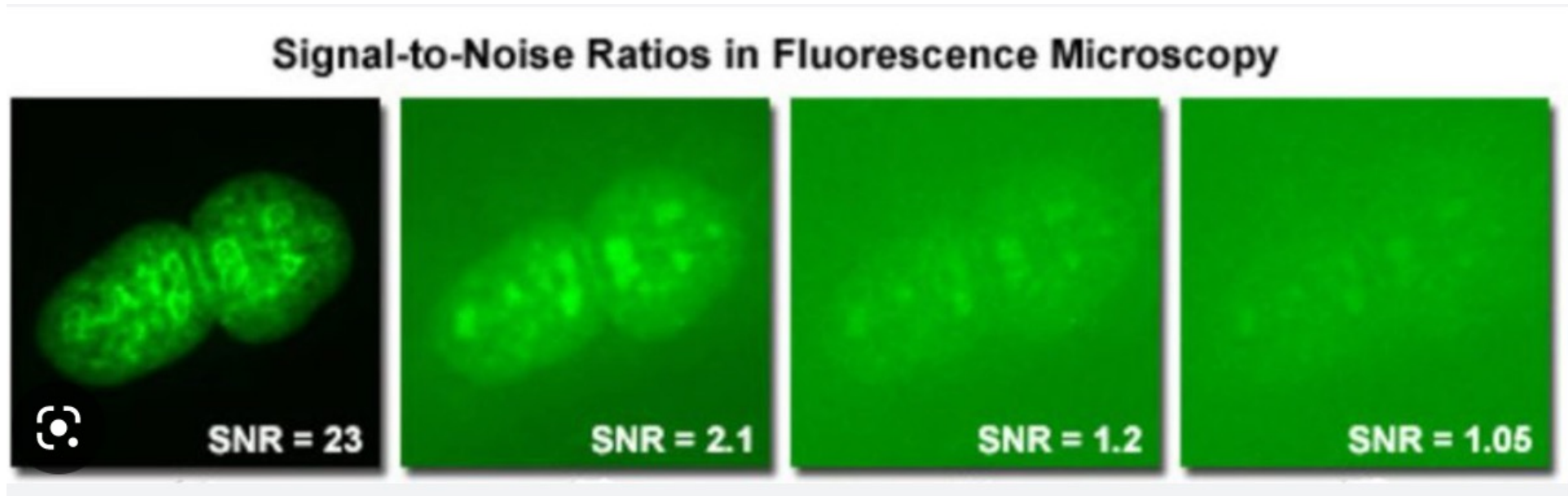


Using water immersion objectives for samples in liquid media captures more signal and preserves resolution better than refractive index mismatched oil immersion objectives.

# SIGNAL-TO-NOISE

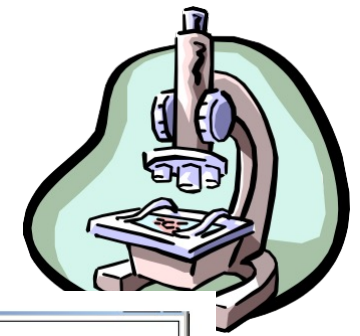


- Camera Sensitivity -
  - Signal needs to be 2-3x brighter than background noise to be detected.

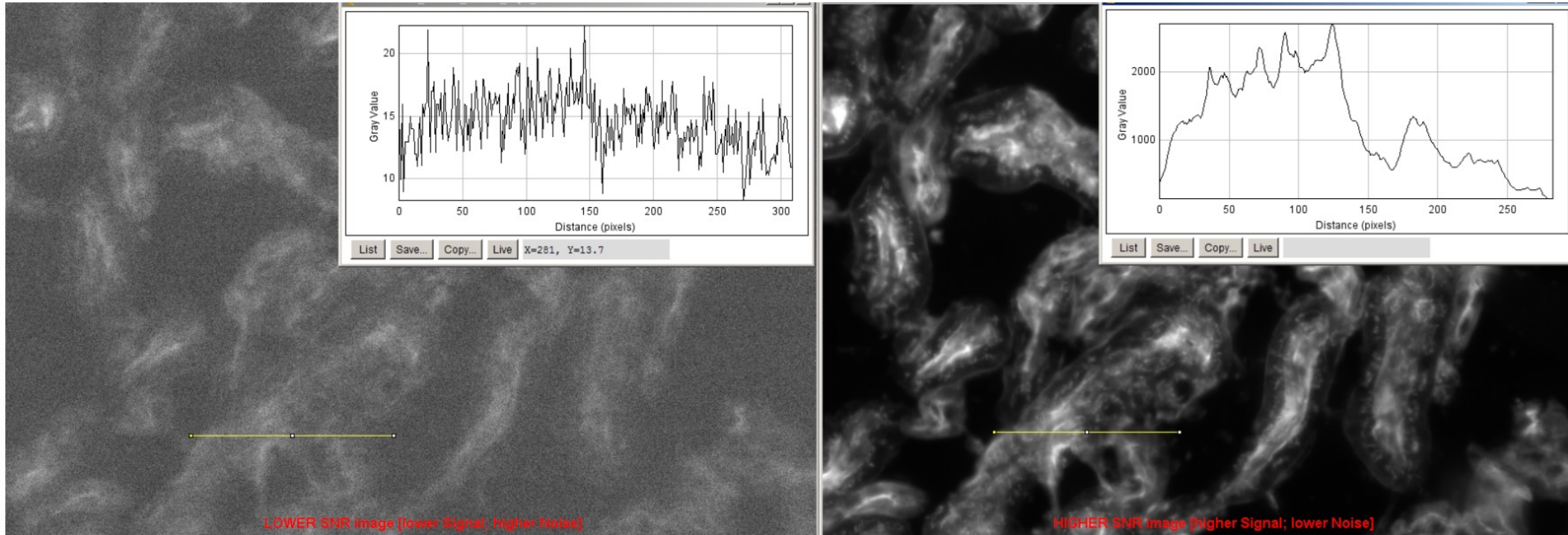


- Increasing the exposure time (integration time) or increasing the illumination intensity produces more signal.

# SIGNAL-TO-NOISE

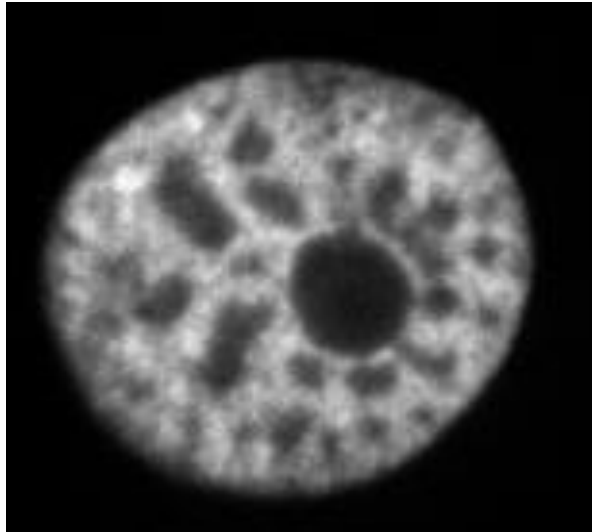
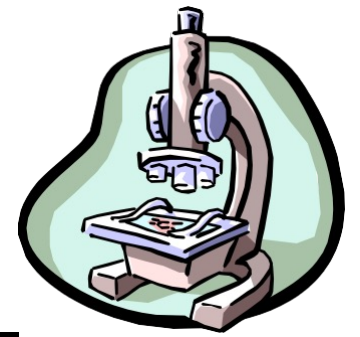


<https://scientificimaging.com/knowledge-base/signal-and-noise-quantitative-explanation/>

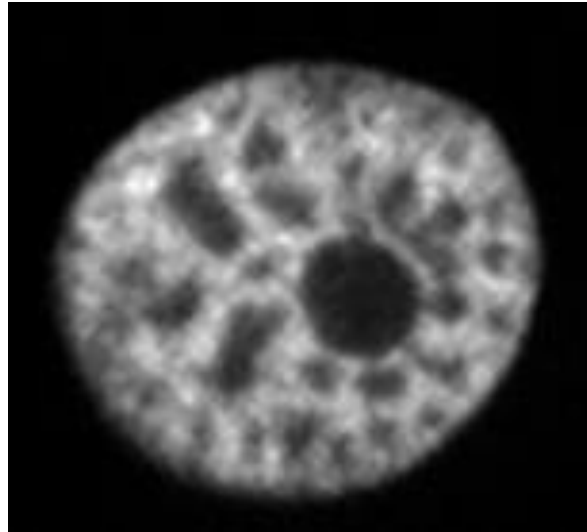


A fluorescence slide was imaged with the same excitation & filters, but with different camera exposures. The LOWER SNR image was obtained with a short=2.5ms exposure and the HIGHER SNR image was obtained with a long=1sec exposure.

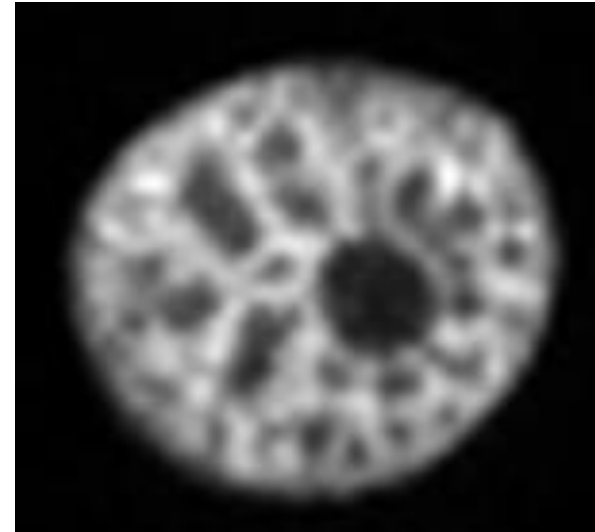
# RESOLUTION VS SPEED



512ms  
1x1 binning



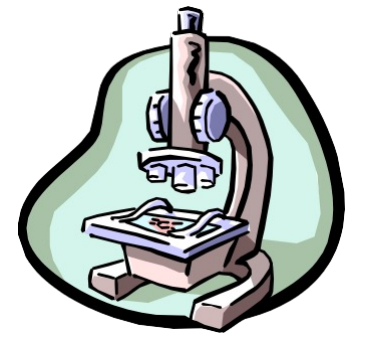
117ms  
2x2 binning



28ms  
4x4 binning

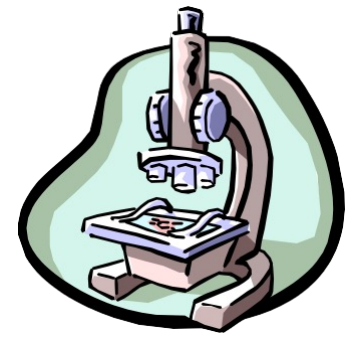
Using binning shortens exposure times reducing cell phototoxicity and fluorophore photobleaching

# CHANGES IN FLOURESCENCE



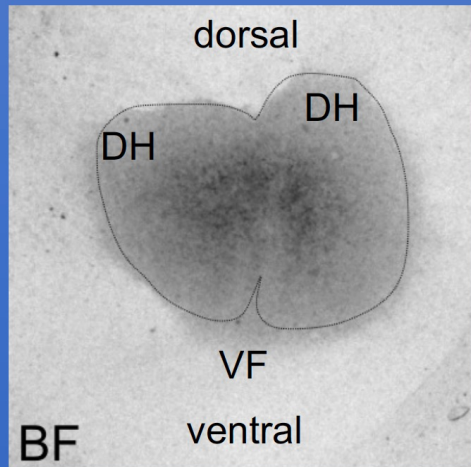
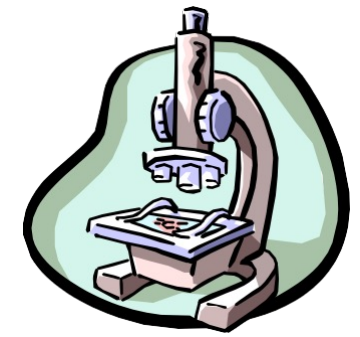
- Changes of fluorescence intensity
  - Examples: calcium imaging, vesicle release
- Changes of fluorescence localisation
  - Examples: Mitochondrial dynamics

# IMAGING CHANGES IN ION CONCENTRATIONS

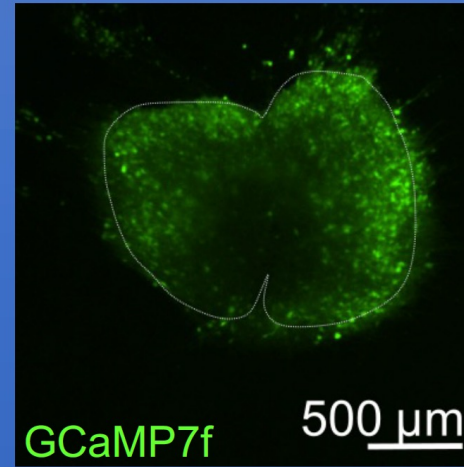
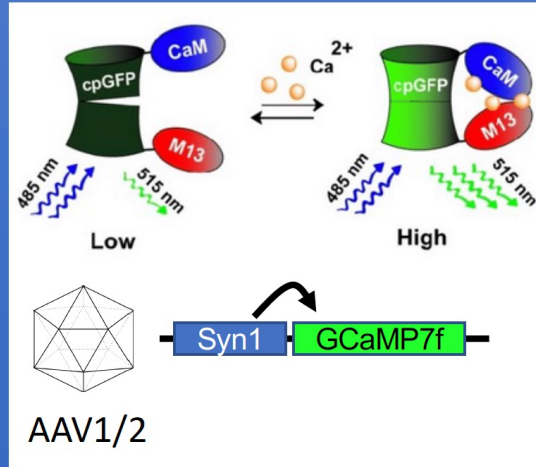


A commonly performed live fluorescence imaging method is ion imaging (calcium, chloride, magnesium, etc) using either fluorescent dyes or proteins especially designed to change their emission behavior upon binding of the respective ion.

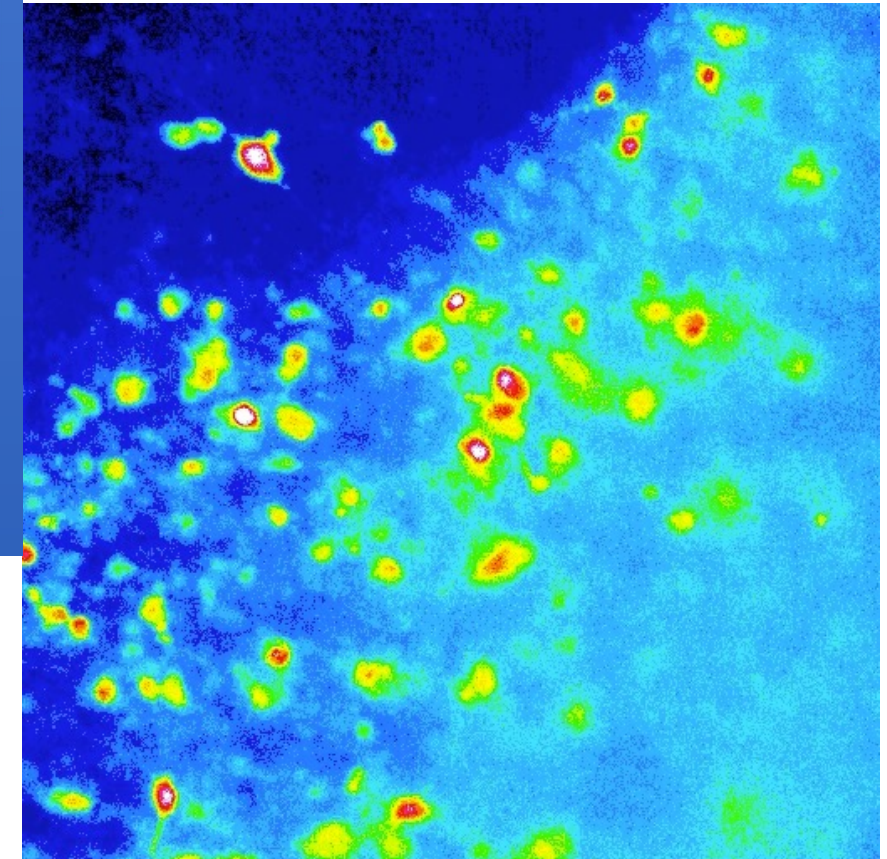
# IMAGING CHANGES IN ION CONCENTRATIONS



Organotypic spinal cord culture at DIV 18. DH, dorsal horn; VF, ventral fissure.

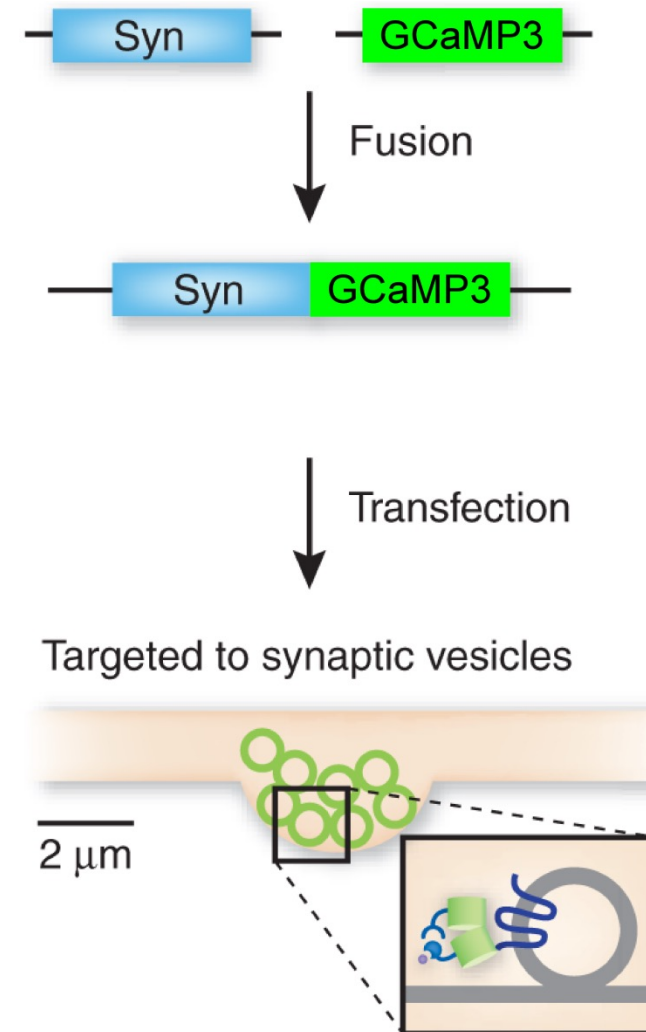
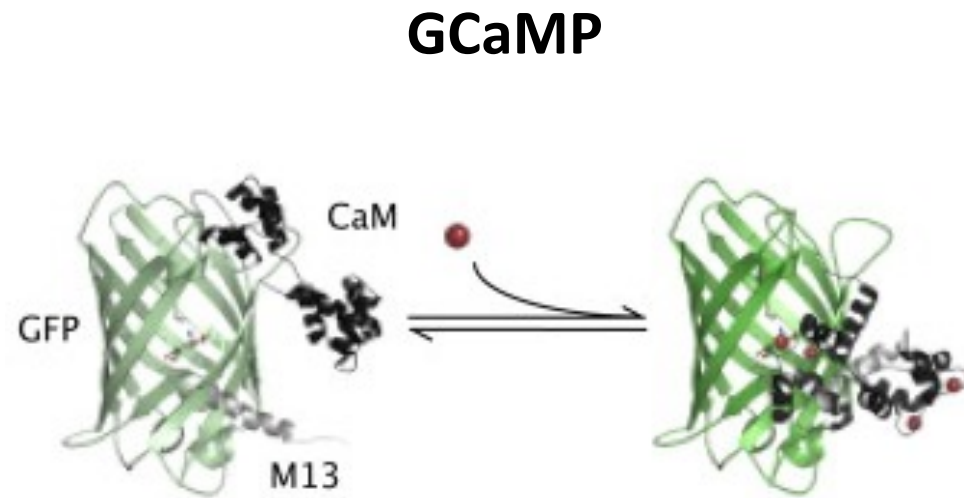


Organotypic spinal cord cultures were infected with AAV1/2 at DIV13, expressed calcium indicator GCaMP7f under control of the neuron-specific promotor Syn1 for 4-6 days. Image of GCaMP7f fluorescence.

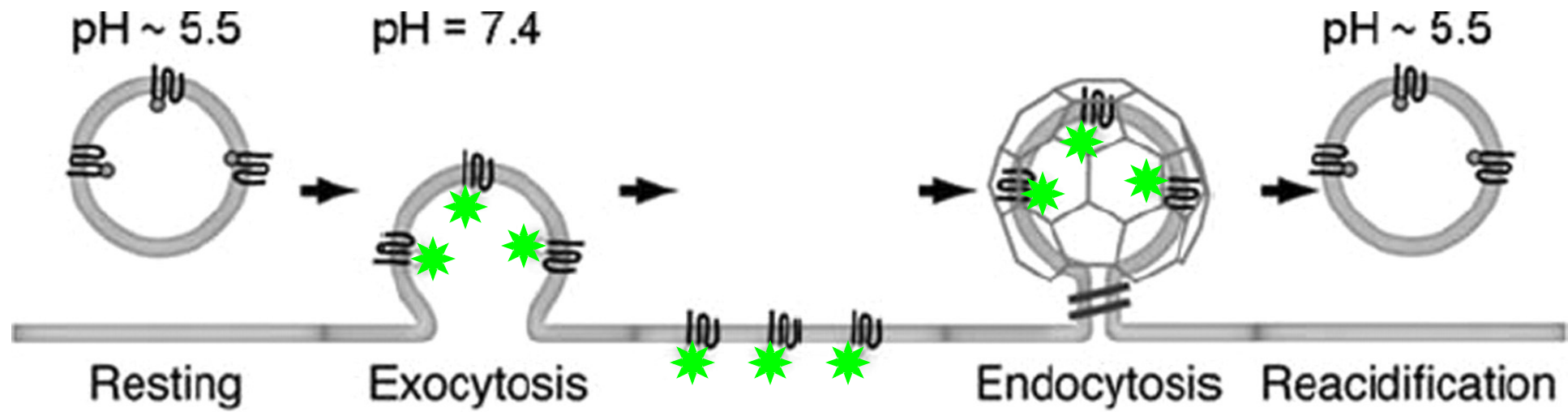


Thalhammer et al, 2022

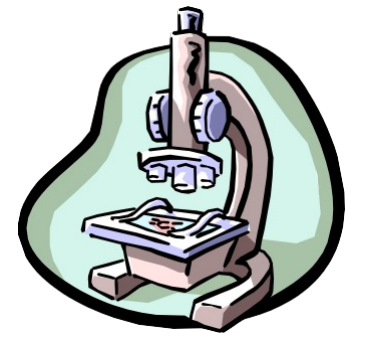
# SYNAPTIC TARGETING OF GCaMPs



# SYNAPTOPHYSIN-PHLUORIN (SYPHY) – IMAGING OF VESICLE RELEASE

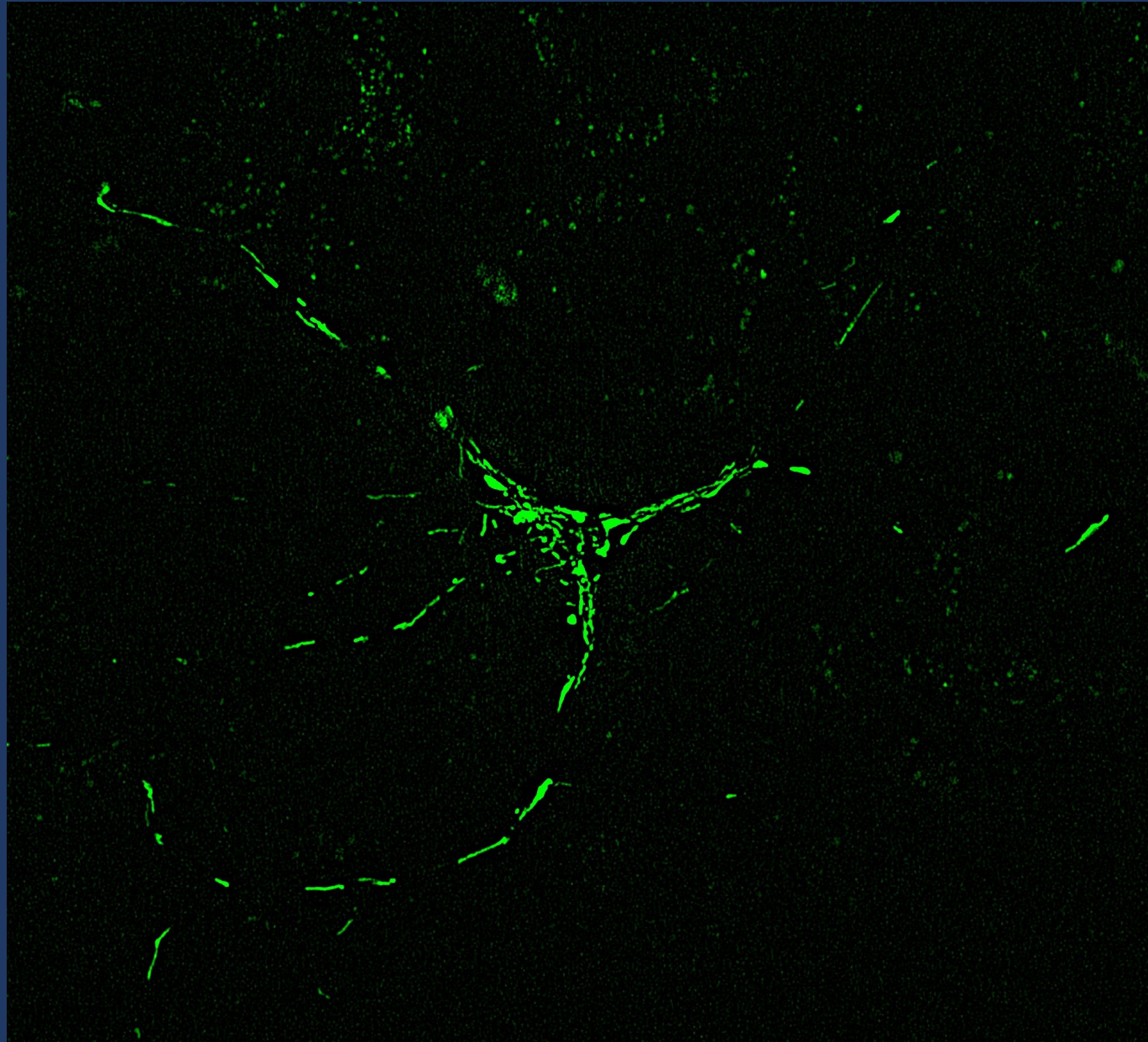


# CHANGES IN FLOURESCENCE

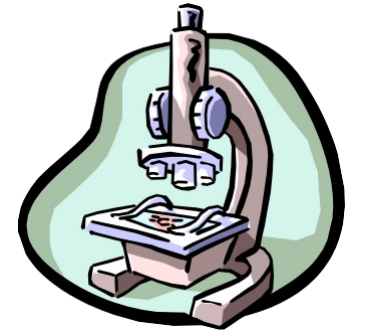


- Changes of fluorescence intensity
  - Examples: calcium imaging, vesicle release
- Changes of fluorescence localisation
  - Examples: Mitochondrial dynamics

# MITOCHONDRIAL PARTICLE DYNAMICS

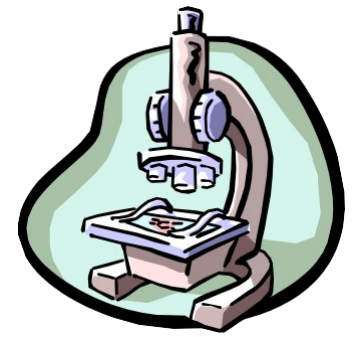


# OTHER LIVE IMAGING TECHNIQUES



- Vesicle fusion/release (TIRF or pH-sensitive GFP)
- FRAP
- FRET
- Photoactivation/Photoconversion

# RESOLUTION-SPEED-SENSITIVITY



## – Spatial Resolution

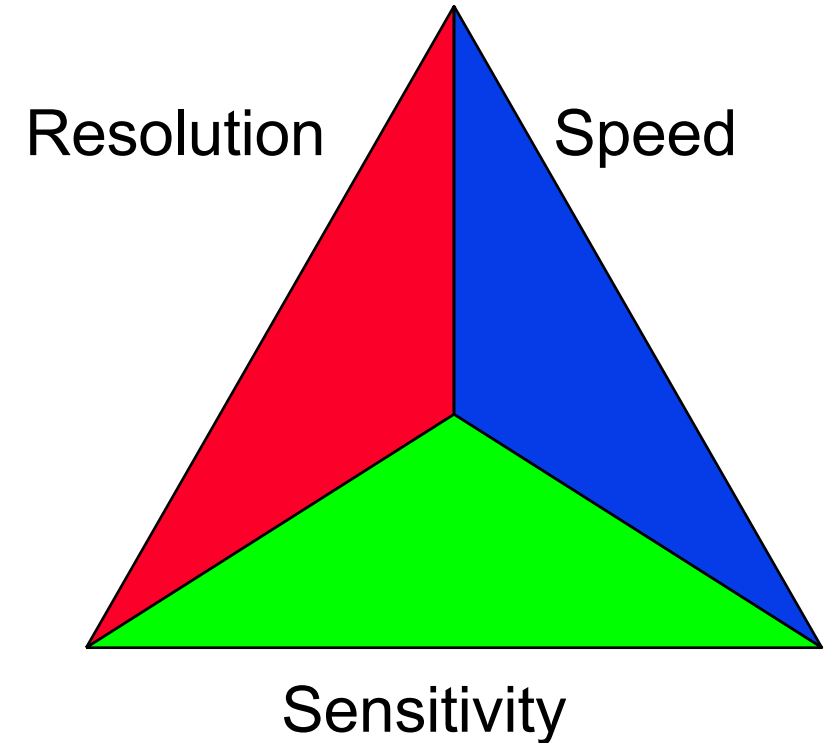
- Depends on application ? optical magnification and lens apperture
- Smaller and more pixels = higher resolution

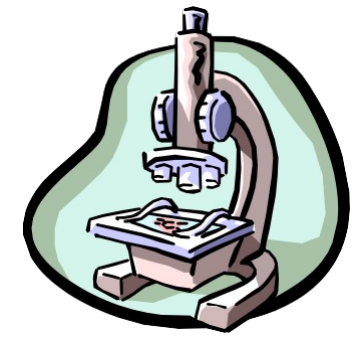
## – Sensitivity

- Larger pixels = higher sensitivity
- Higher sensitivity = shorter exposure time
- Shorter exposure time = higher frame rate
- Larger pixels = lower resolution

## – Speed

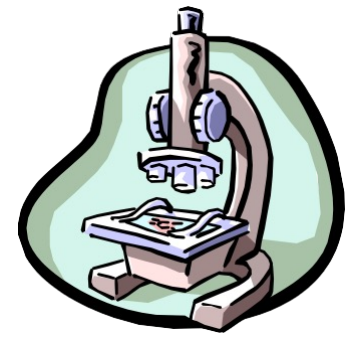
- Less pixels = faster readout
- Faster Readout = higher framerate
- Less pixels = lower resolution
- requires fast readout and high sensitivity





# Image processing and analysis

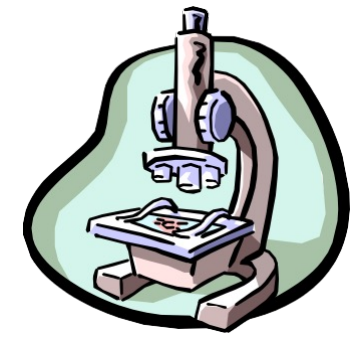
# DIGITAL IMAGES



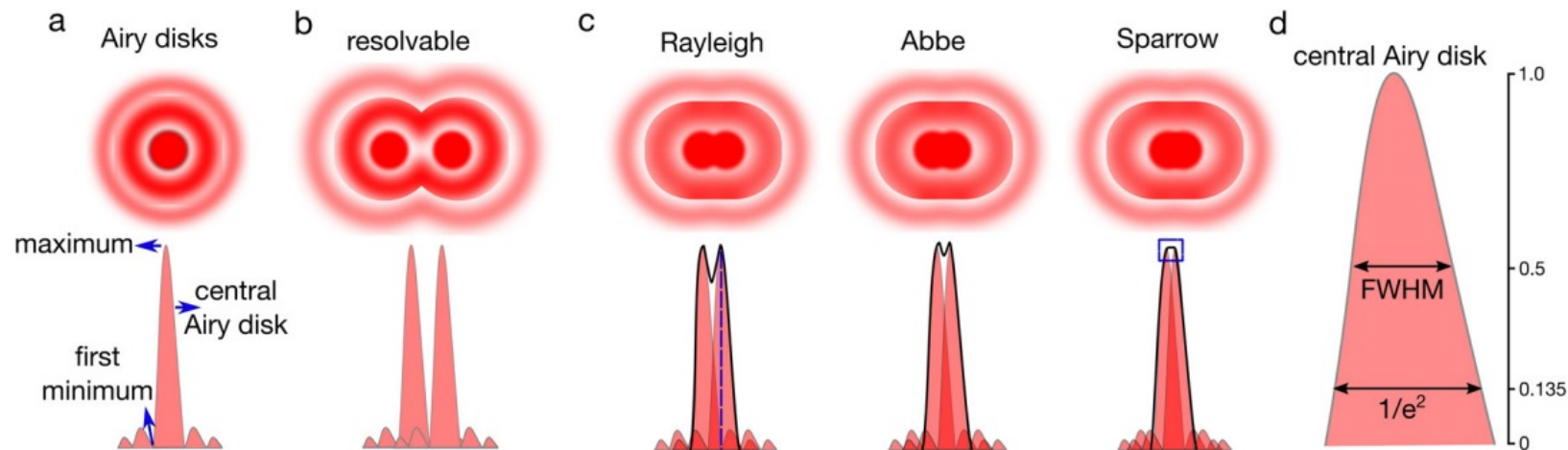
- 2D structure built from pixels
- Can be seen as a matrix comprised of  $m$  columns and  $n$  rows, pixel location denoted by index  $(i,j)$
- for:  $0 \leq i < n$  and  $0 \leq j < m$

|          |          |          |          |
|----------|----------|----------|----------|
| $0,0$    | $0,1$    | $\dots$  | $0,m$    |
| $1,0$    | $1,1$    | $\dots$  | $1,m$    |
| $\vdots$ | $\vdots$ | $\ddots$ | $\vdots$ |
| $n,0$    | $n,1$    | $\dots$  | $n,m$    |

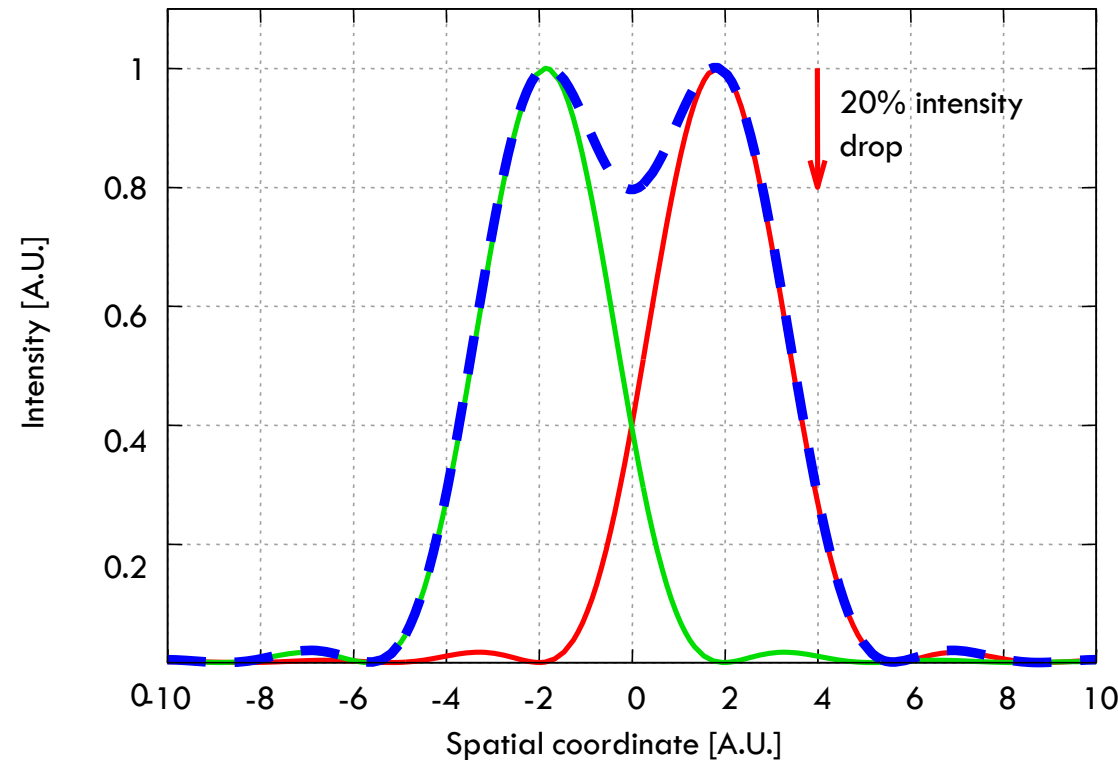
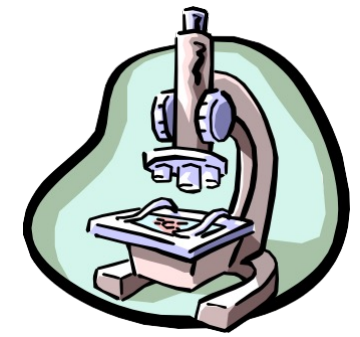
# RESOLUTION



| Microscopy                        | Digital image processing                                                       |
|-----------------------------------|--------------------------------------------------------------------------------|
| Rayleigh criterion                | The spatial resolution of an image is the physical size of a pixel in an image |
| Sparrow criterion                 |                                                                                |
| full width at half maximum (FWHM) |                                                                                |



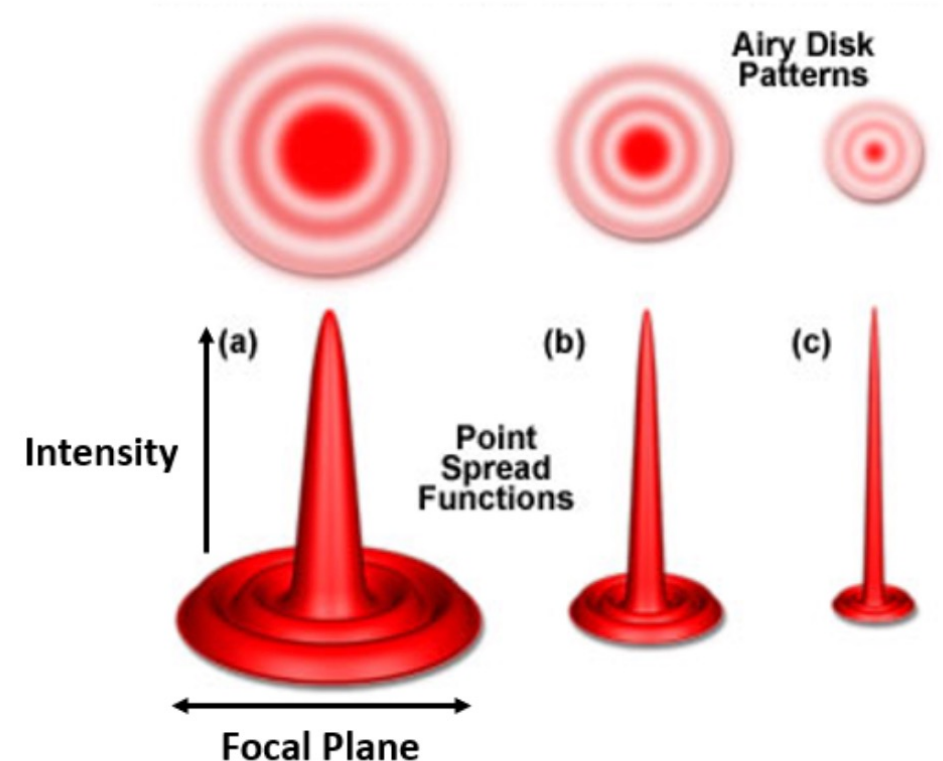
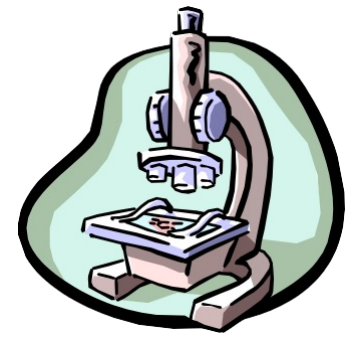
# RAYLEIGH CRITERION



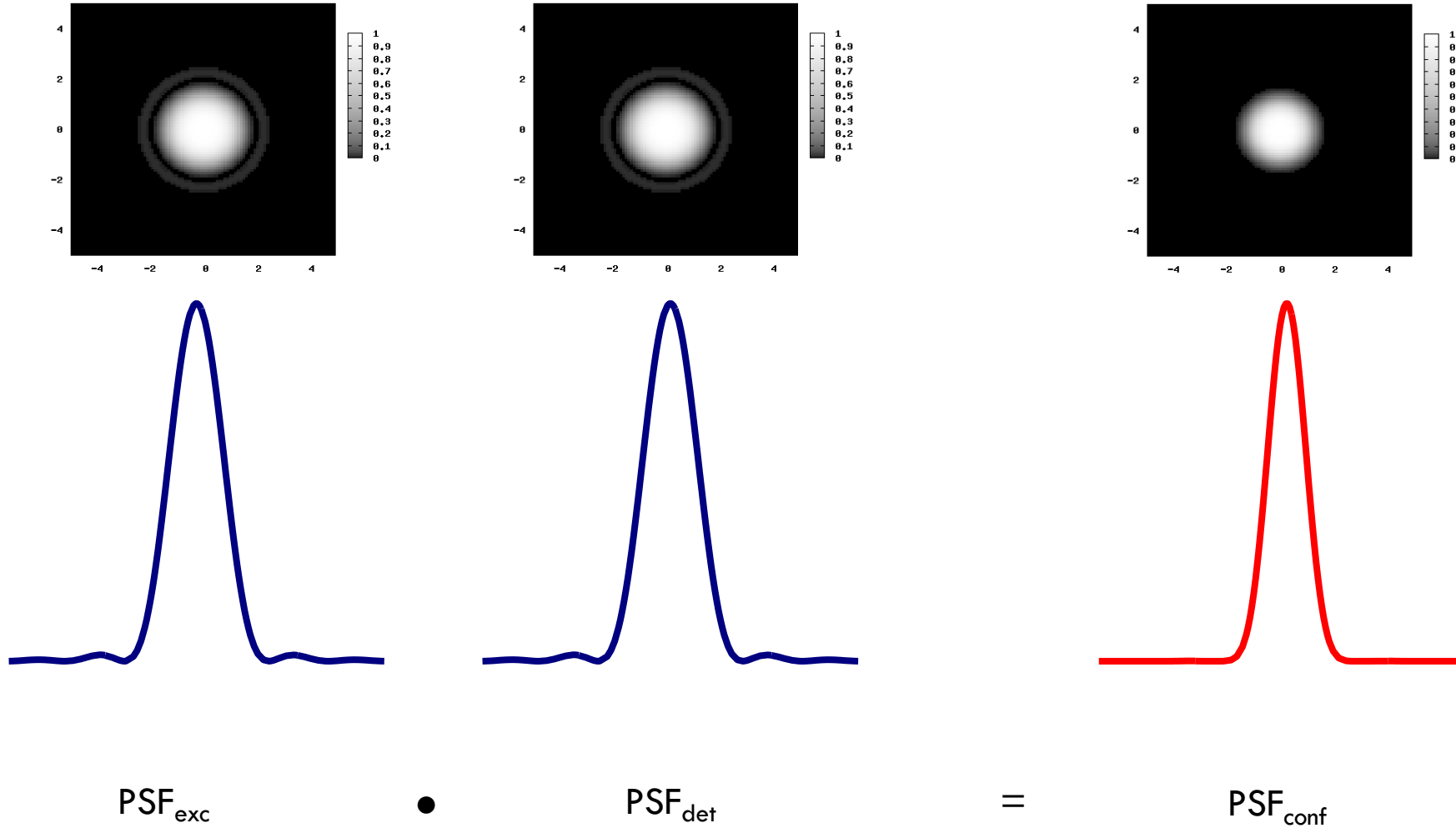
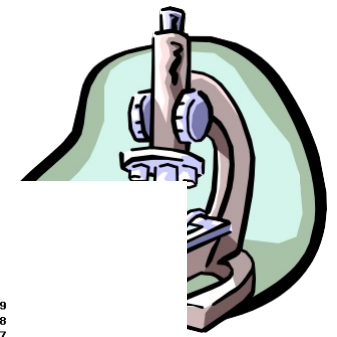
The **Rayleigh criterion** is the generally accepted, although arbitrary, criterion for the minimum resolvable detail – the imaging process is said to be diffraction-limited when the first diffraction minimum of the image of one source point coincides with the maximum of another.

# POINT-SPREAD FUNCTION

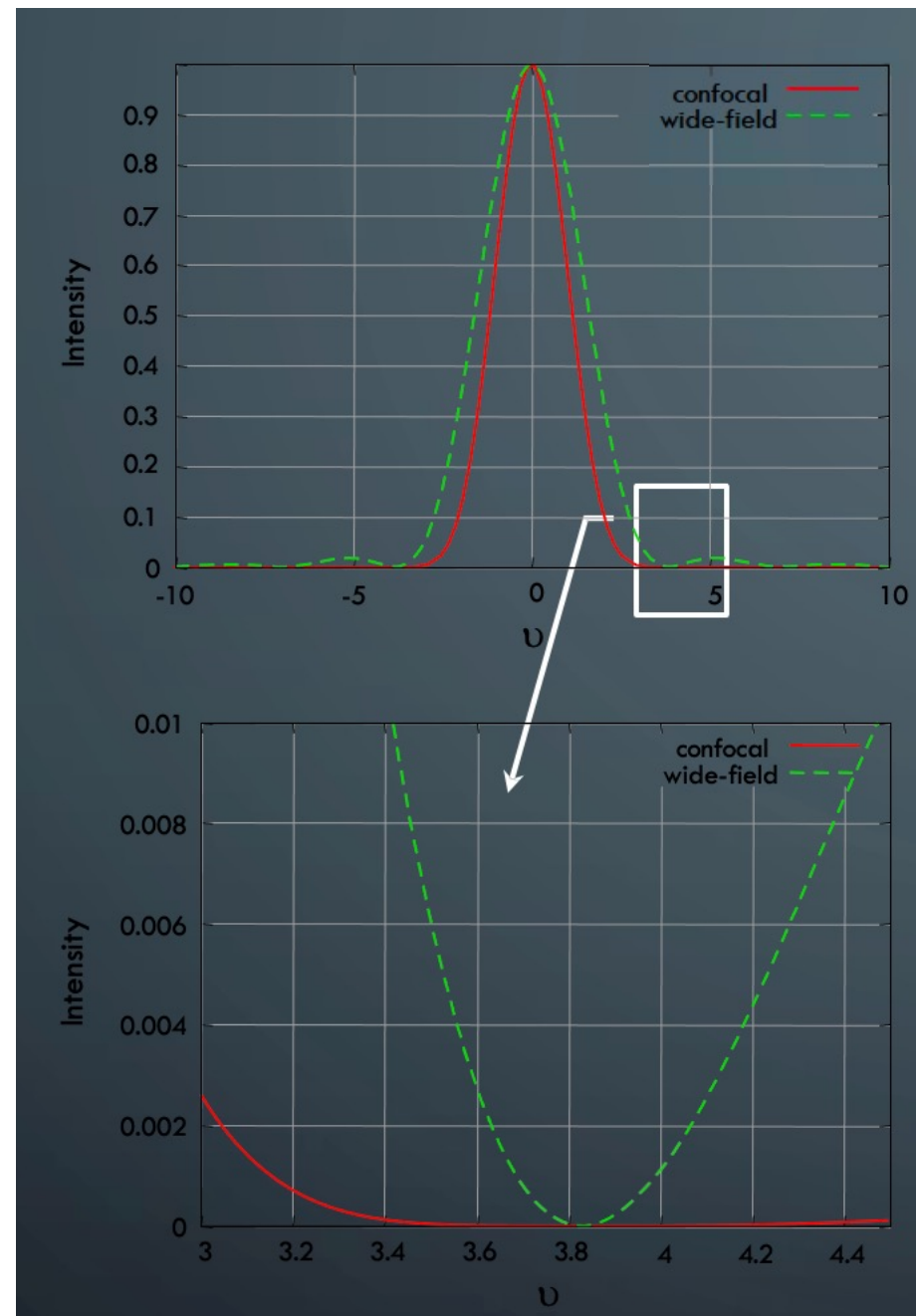
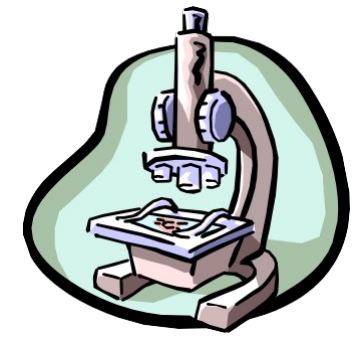
- The point spread function (PSF) describes 3D light distribution in an image of a point source (for a given lens)
- An x-y slice through the center of the wide-field point spread function reveals a set of concentric rings: the so-called Airy disk that is commonly referenced in texts on classical optical microscopy



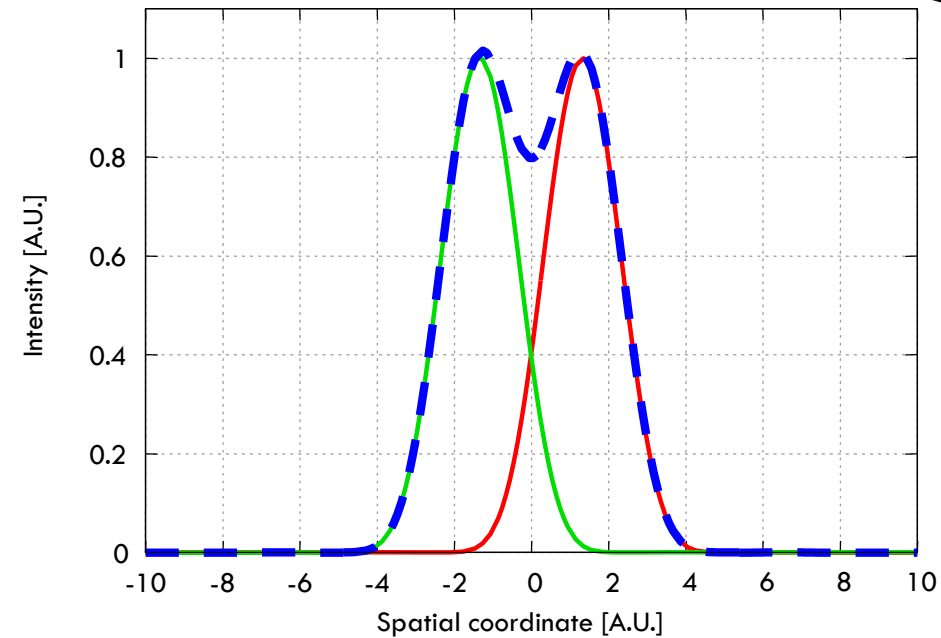
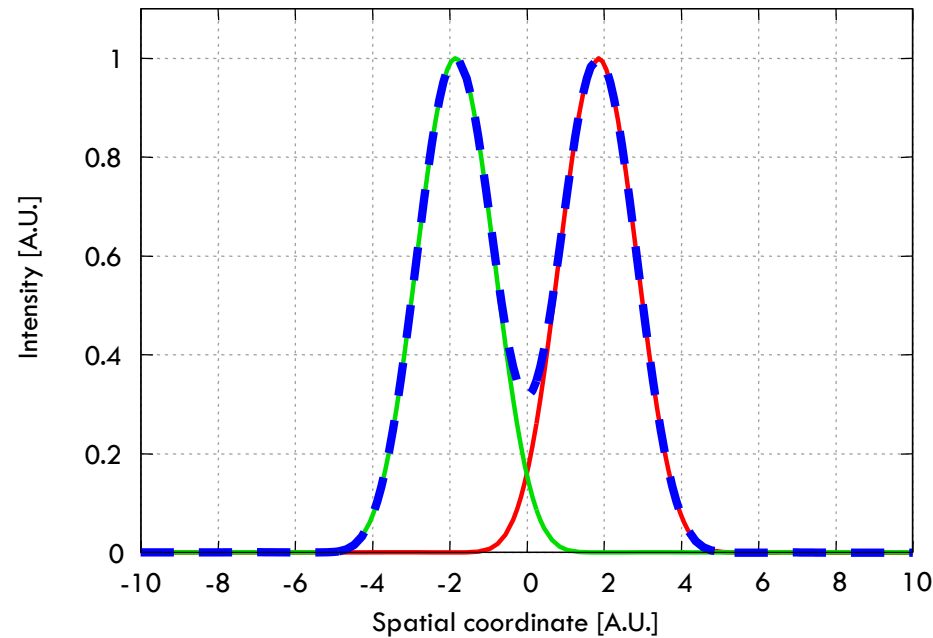
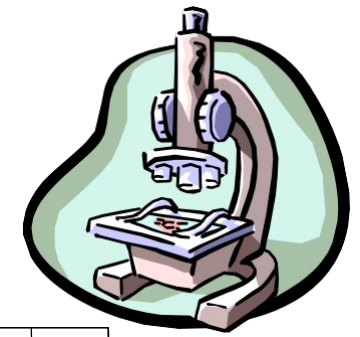
# POINT-SPREAD FUNCTION - CONFOCAL



# RESOLUTION - CONFOCAL

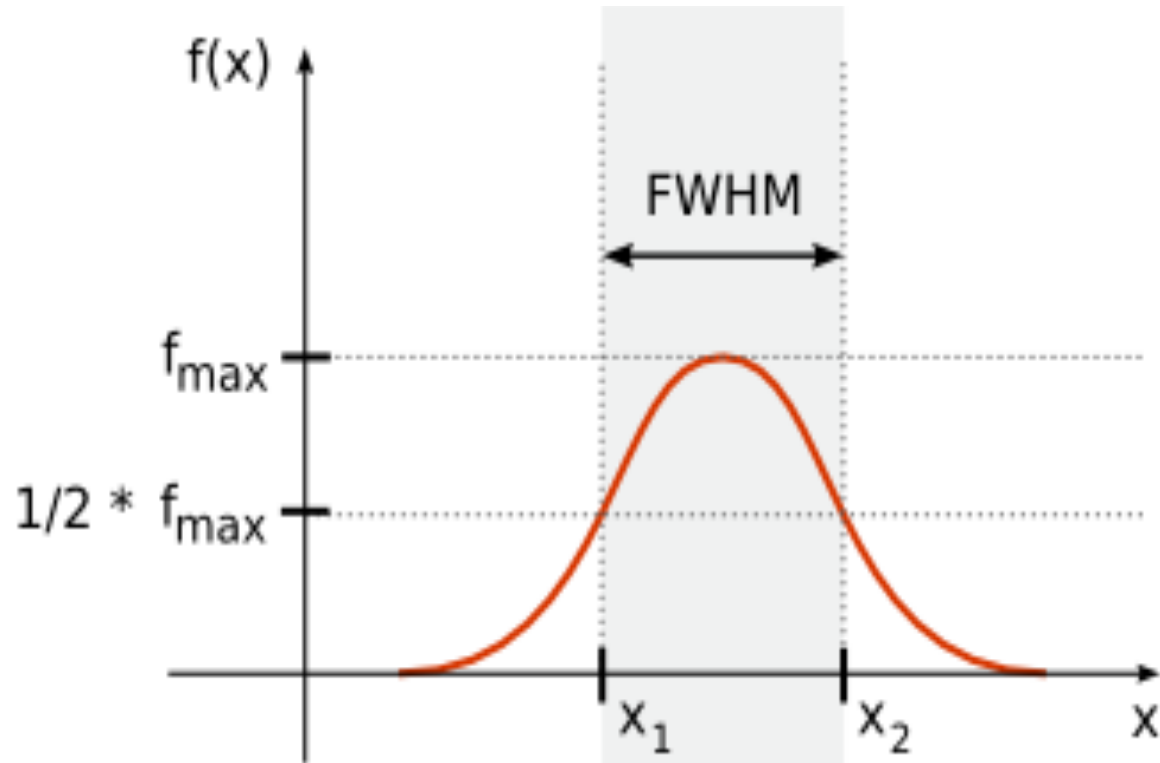
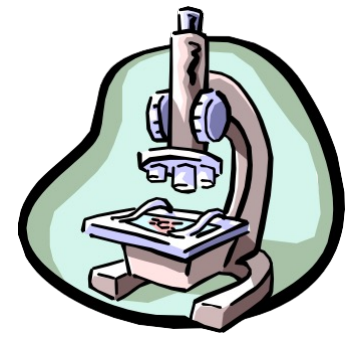


# RESOLUTION - CONFOCAL



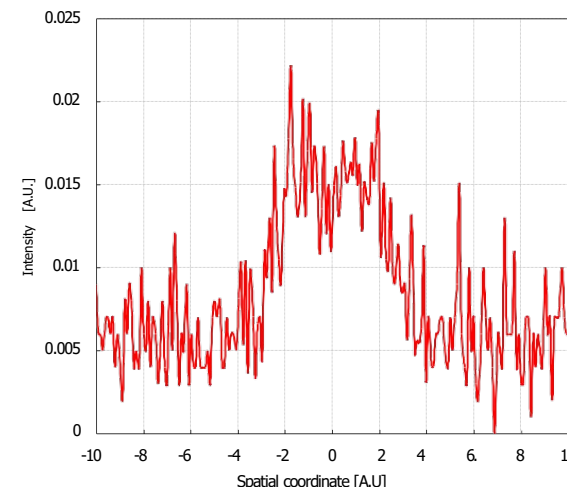
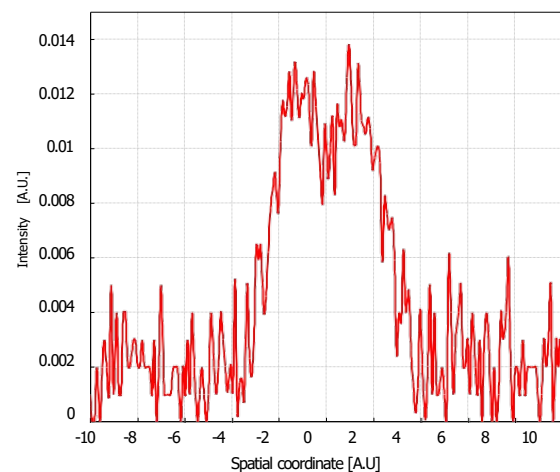
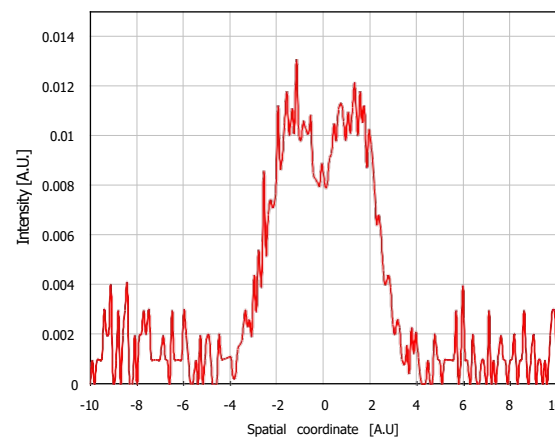
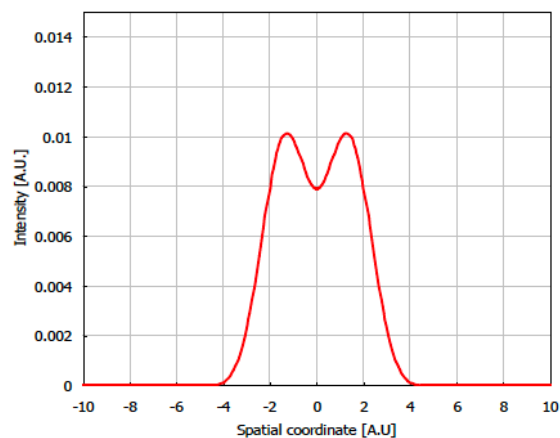
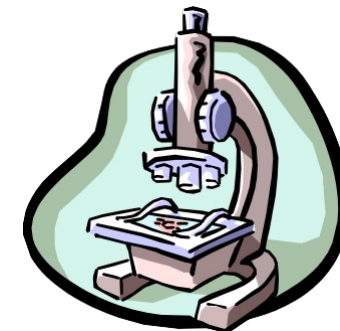
Rayleigh criterion cannot be used directly to define the improvement of resolution in a confocal microscope. The position of the first minimum does not change. The drop in intensity is much larger, though. In fact, we can move two intensity distributions a little bit closer (1.4 times), and still get the required 20% drop in intensity.

# RESOLUTION - CONFOCAL



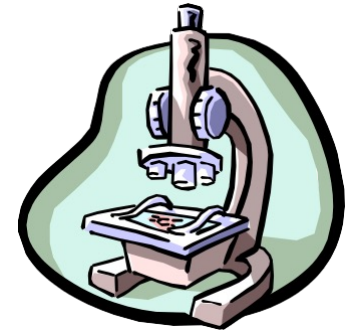
The full width at half maximum (FWHM) is a parameter commonly used to describe the width of a "bump" on a curve or function. It is given by the distance between points on the curve at which the function reaches half its maximum value.

# SIGNAL-TO-NOISE RATIO AND RESOLUTION



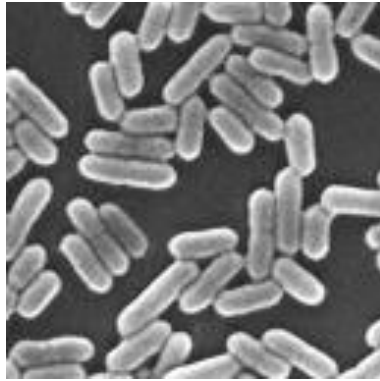
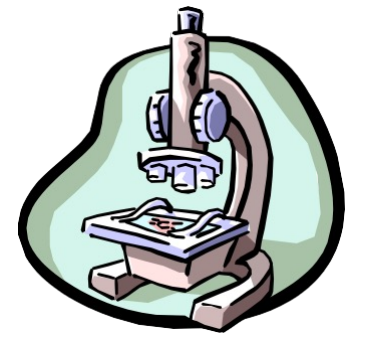
The influence of Poisson noise on two intensity distributions separated spatially according to the Rayleigh criterion.

# DIGITAL RESOLUTION IN ELECTRONIC PUBLISHING

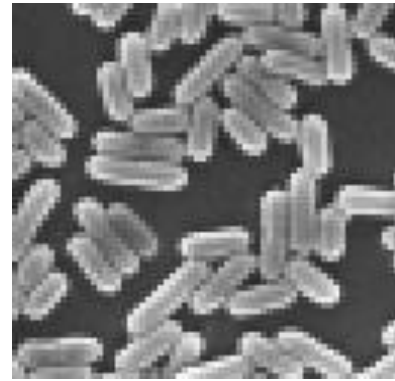


- **Image resolution** is measured in pixels per inch (ppi).
- An image with a resolution of 100 ppi, contains 10,000 pixels in a Square inch (100 pixels wide x 100 pixels high = 10,000).
- The higher the resolution, the more pixels in the image. A 2-by-2-inch image with a resolution of 100 ppi would have 40000 pixels.
- The same image with a resolution of 300 ppi would have 360,000 Pixels in the same 2-by-2-inch area.

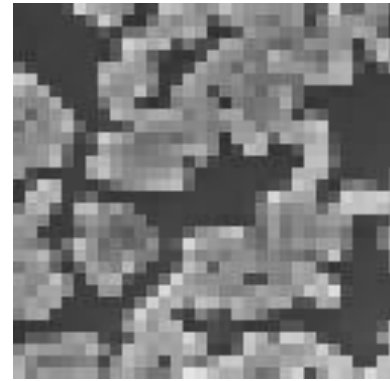
# DIGITAL RESOLUTION



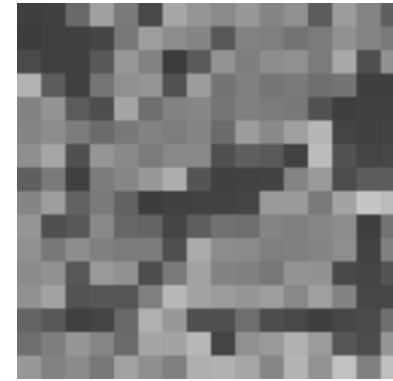
128x128



64x64



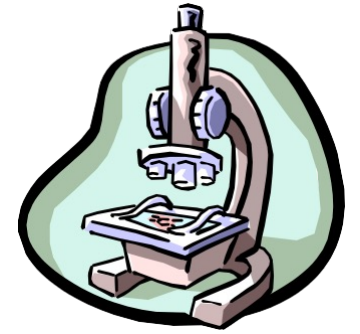
32x32



16x16

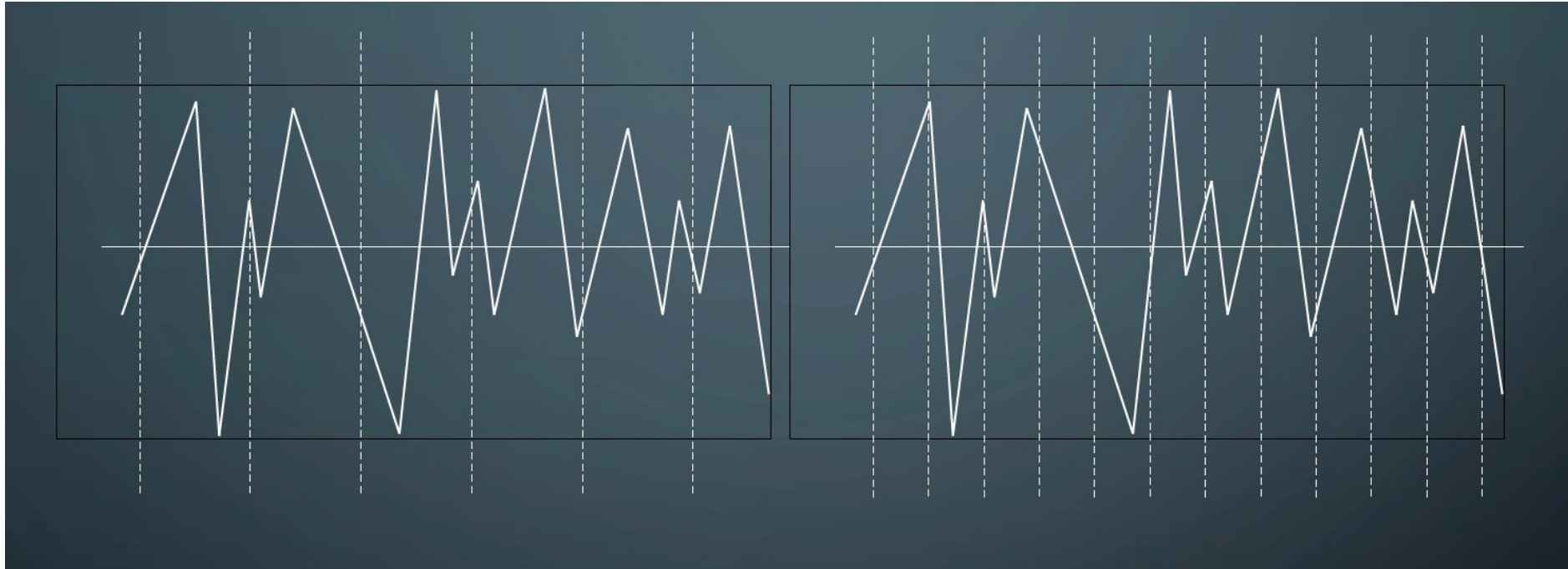
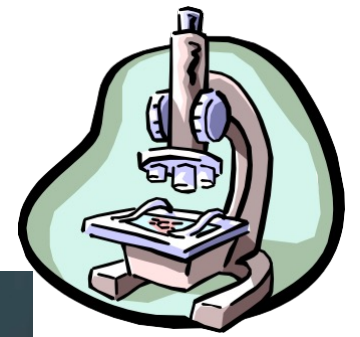


# SAMPLING THEORY

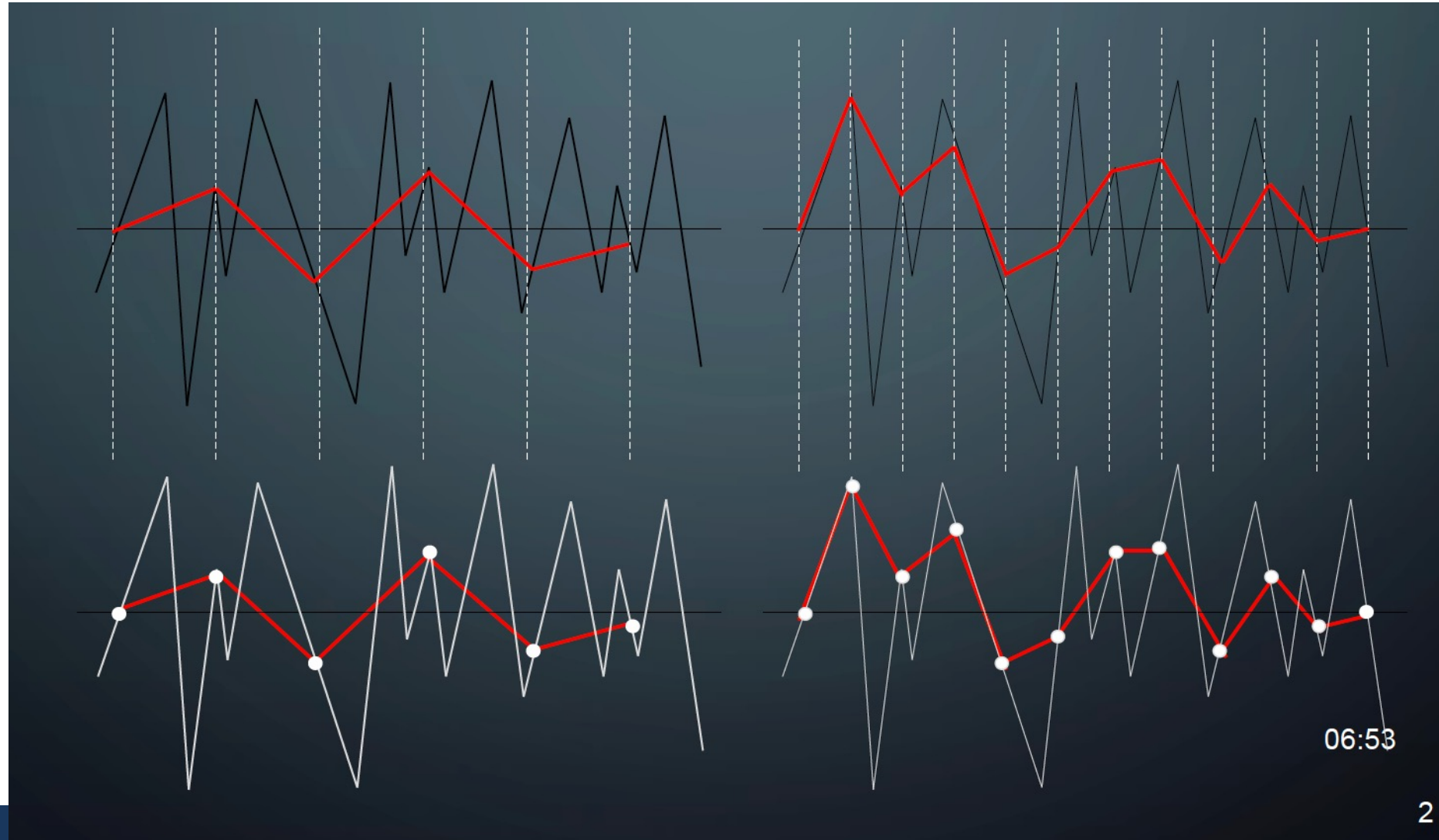
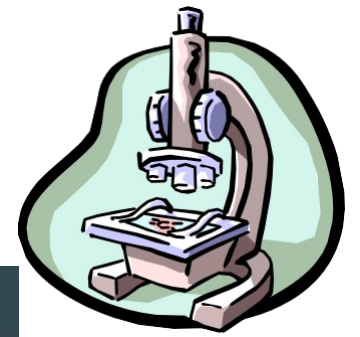


- Nyquist Theorem
  - sampling frequency ( $f$ ) required to represent the true identity of the sample: to capture the periodic components of frequency  $f$  in a signal we need to sample at least  $2f$  times
- Nyquist claimed that the rate was  $2f$ ; in reality the rate is  $2.3f$  (~at least 2 times the highest frequency)

# SAMPLING AND NYQUIST

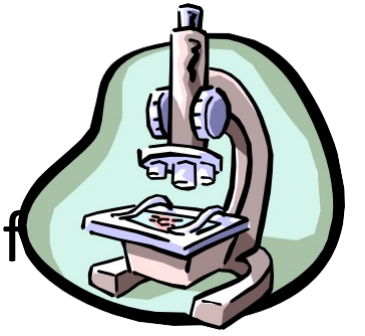


# SAMPLING AND NYQUIST



2

# SAMPLING AND NYQUIST



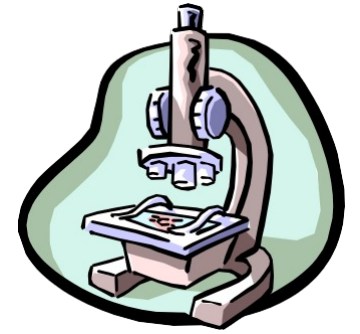
An image that is digitized at a rate of 600 dpi will only be 300 dpi if enlarged 2X and printed

- practical sampling guideline for printed images is that the final image should have at least the resolution of the printing device

For light and electron micrographs of biological tissue, prints with a final resolution of 600 dpi will preserve most detail. However, fine granular structures such as autoradiography silver grains or colloidal gold particles used in immunolabeling will not be as sharp as those in a traditional photograph

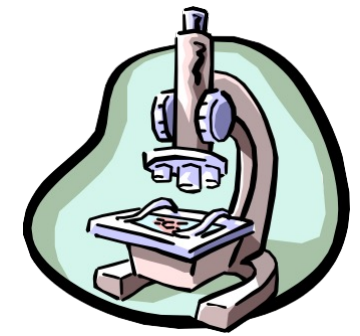
- the final print resolution may need to be 1200 dpi or higher.

# BIT RESOLUTION OR PIXEL DEPTH



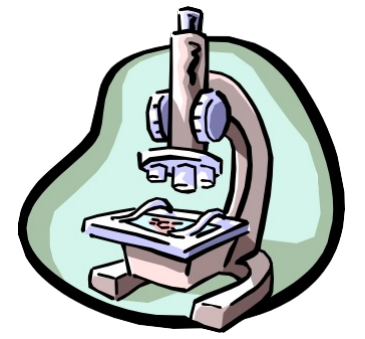
- This is a measurement of the number of **bits of information** per pixel.
- The pixel depth will determine how much color or gray scale Information is available for each pixel.
- Greater pixel depth means more available colors and more accurate color representation.
- Pixels in binary images have a depth of 1 (on or off), and are black and white images
- A pixel with a bit depth of 8 has  $2^8$ , or 256, possible values; and a pixel with a bit depth of 24 has  $2^{24}$ , or 16 million possible values (red=8, green=8, blue=8).
- Common values for pixel depth range from 1 to 24 bits per pixel.

# QUANTIZATION

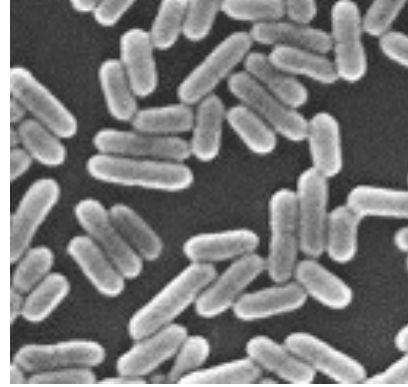


- A set of  $n$  quantization levels comprises the integers:  
 $0, 1, 2, \dots, n-1$
- $0$  and  $n-1$  are usually black and white respectively, with intermediate levels rendered in various shades of gray.
- Quantization levels are commonly referred to as **gray levels**.
- $n$  is usually an integral power of two:  $n=2^b$ , where  $b$  is the number of bits used for quantization.
- If  $b=1$  then the image is **binary**

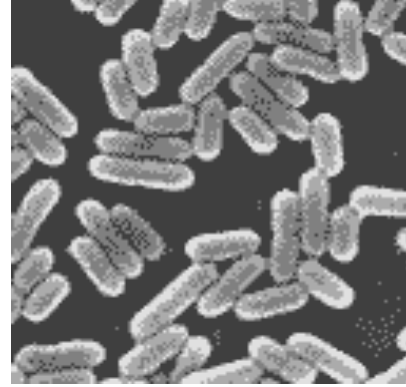
# QUANTIZATION



256



16

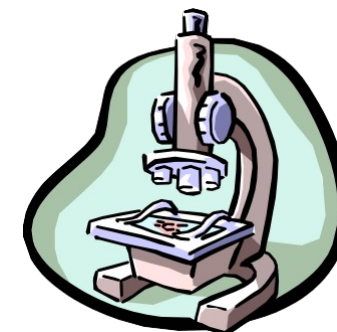


4

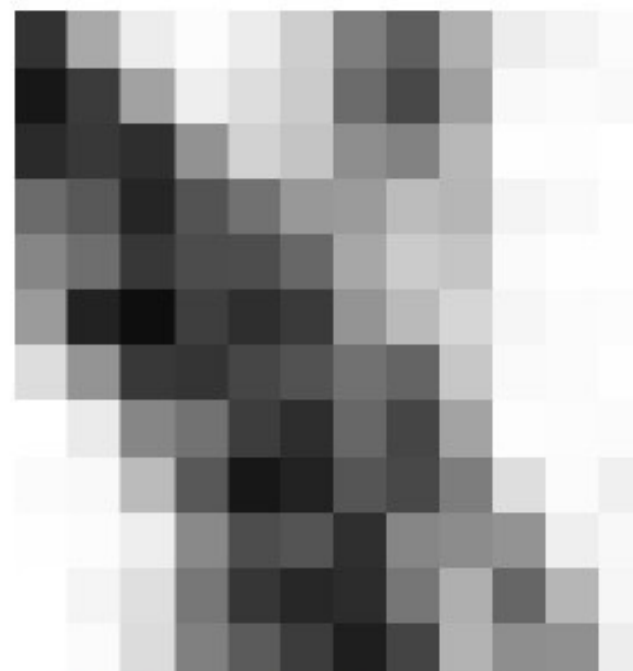


2





(a) Original image

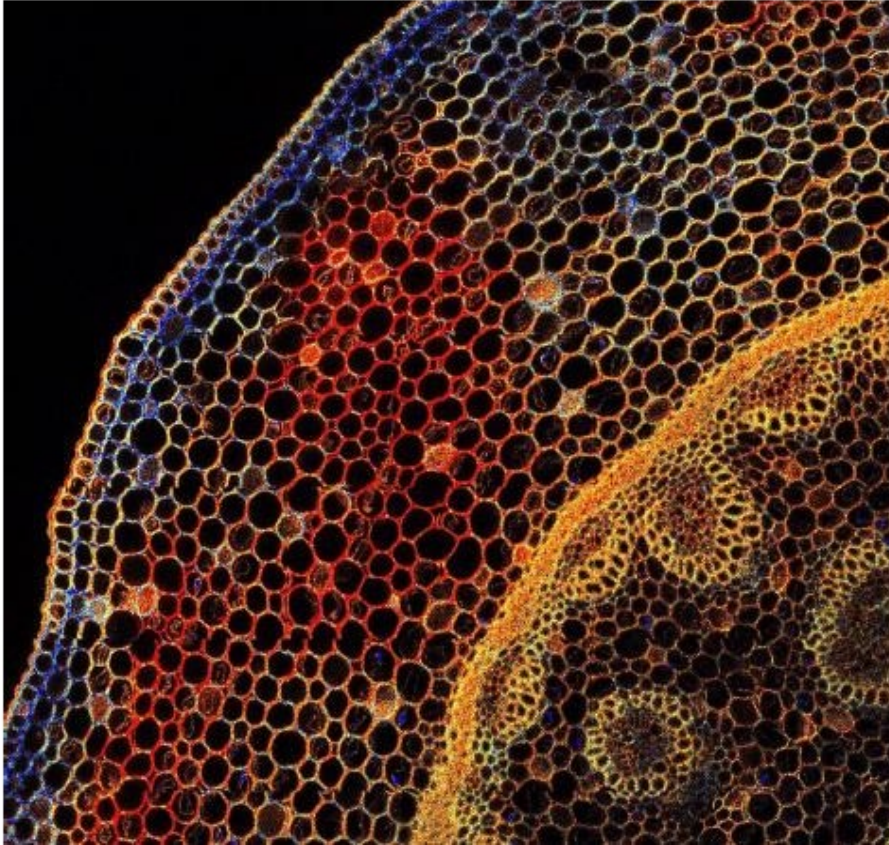


(b) Enlarged view from (a)

|     |     |     |     |     |     |     |     |     |     |     |     |
|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| 50  | 169 | 237 | 252 | 236 | 206 | 125 | 94  | 176 | 237 | 244 | 251 |
| 23  | 59  | 162 | 238 | 221 | 203 | 107 | 72  | 160 | 249 | 251 | 249 |
| 42  | 56  | 46  | 146 | 210 | 196 | 142 | 130 | 184 | 254 | 253 | 255 |
| 107 | 88  | 37  | 83  | 113 | 152 | 155 | 188 | 182 | 244 | 249 | 254 |
| 134 | 110 | 55  | 76  | 77  | 103 | 167 | 203 | 197 | 250 | 254 | 254 |
| 155 | 34  | 14  | 62  | 46  | 58  | 148 | 186 | 214 | 246 | 251 | 252 |
| 221 | 148 | 56  | 52  | 70  | 82  | 113 | 100 | 199 | 250 | 251 | 254 |
| 255 | 235 | 134 | 114 | 61  | 45  | 103 | 69  | 163 | 253 | 252 | 251 |
| 251 | 249 | 187 | 88  | 23  | 34  | 85  | 72  | 125 | 222 | 251 | 240 |
| 254 | 252 | 238 | 137 | 77  | 84  | 47  | 134 | 140 | 147 | 239 | 248 |
| 255 | 245 | 223 | 119 | 54  | 39  | 44  | 118 | 175 | 102 | 182 | 246 |
| 255 | 251 | 220 | 128 | 91  | 60  | 30  | 68  | 179 | 142 | 144 | 237 |

(c) Pixel values of (b)

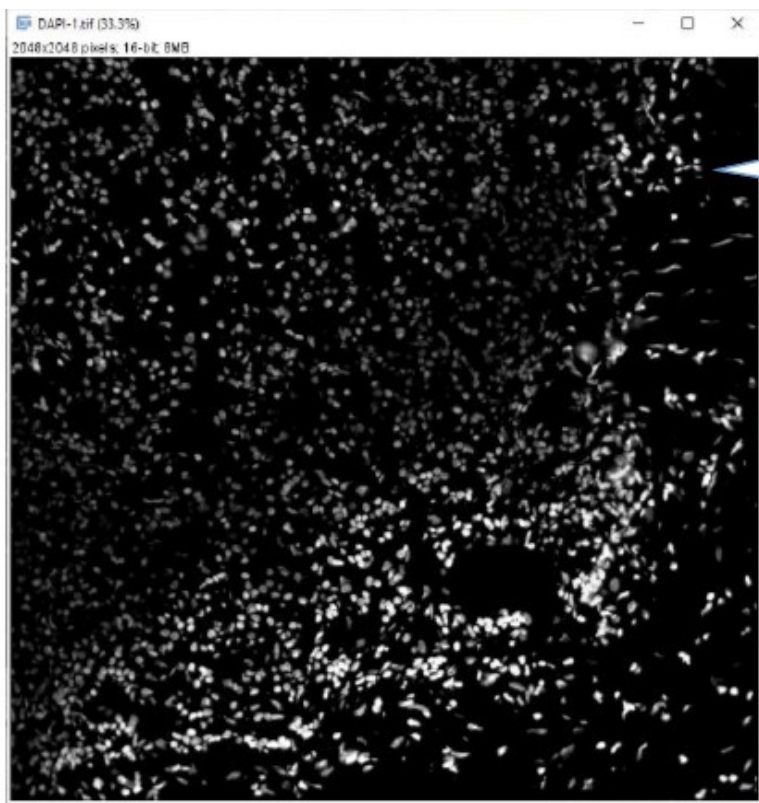
# Deriving quantitative information from images of biological samples



How many cells  
are there in the  
outer ring?

How high is the  
intensity in the  
inner ring?

How is the signal in  
the blue channel  
distributed?



Nuclei in this  
image are ...

... more dense  
than in this image.

Use automation for  
less subjective  
analysis.

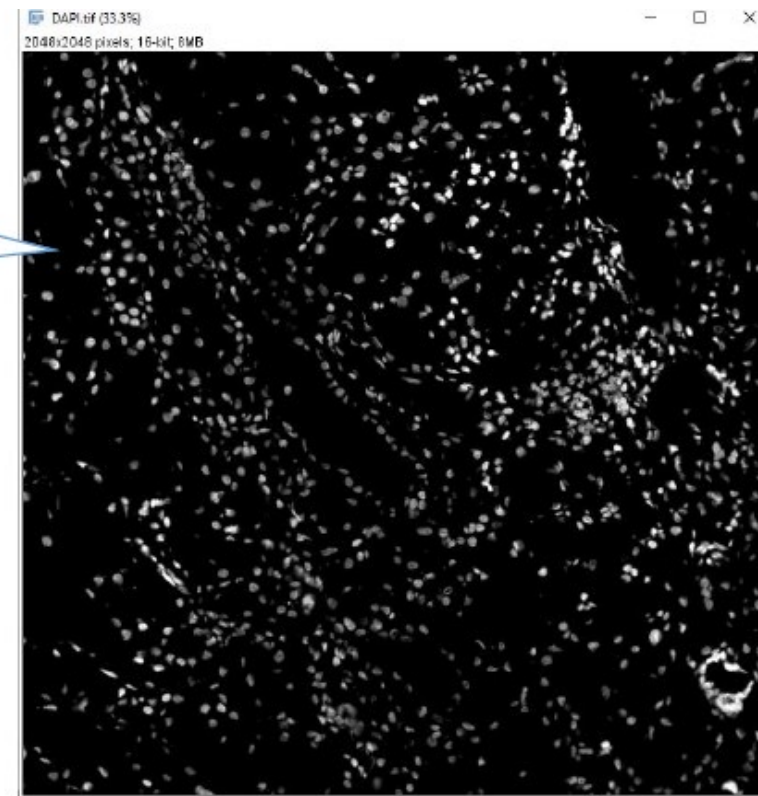
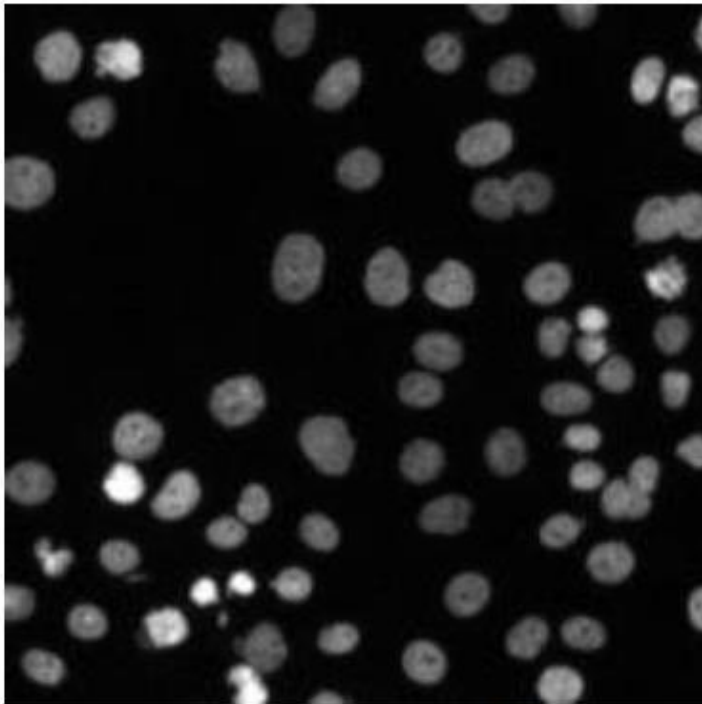


Image data source: Pascual-Reguant, Anna. (2021). Immunofluorescence staining of a human kidney (#2, peri-tumor area) obtained by MELC [I

Algorithms must be reliable. Visualization helps gaining trust in methods

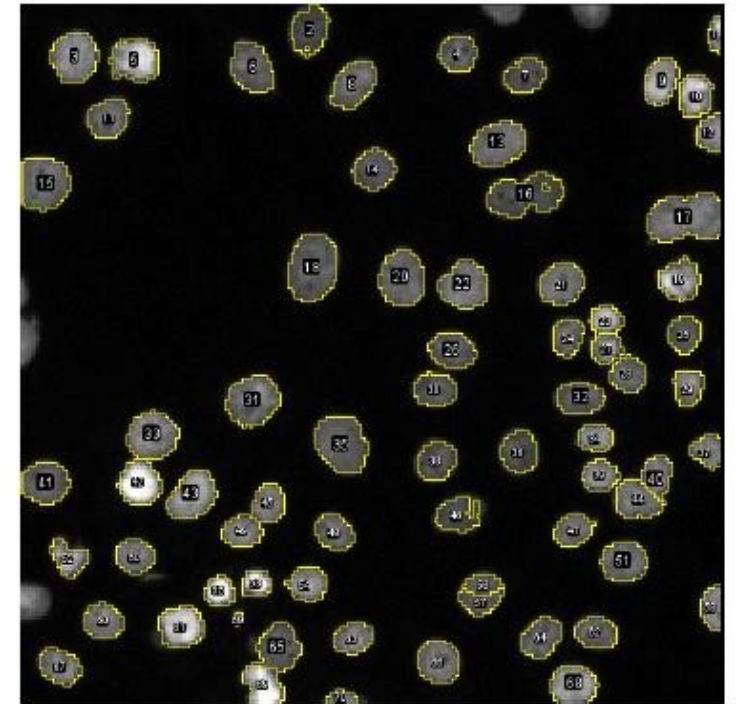
Original image



Label image



Overlay



There are 70 nuclei

# QUANTITATIVE MICROSCOPY IS SUPPOSED TO BE:

- **Quantitative**
  - We derive numbers from images which describe physical properties of the observed sample.
- **Objective**
  - The derived measurement does not depend on who did the measurement. The measurement is free of interpretation.
- **Reliable (trustworthy / validated)**
  - We are confident that the measurement is describing what it is supposed to describe.
- **Reproducible**
  - Somebody else can do the experiment under *different conditions* and gets similar measurements. For this, documentation is decisive!
- **Repeatable**
  - We can do the same experiment twice under the *same conditions* and get similar measurements.

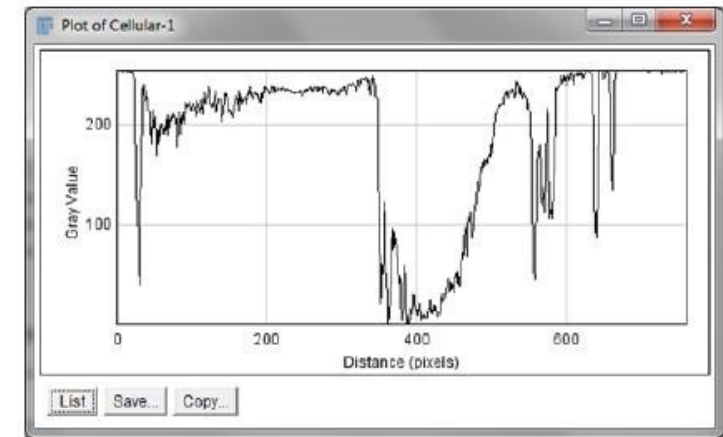
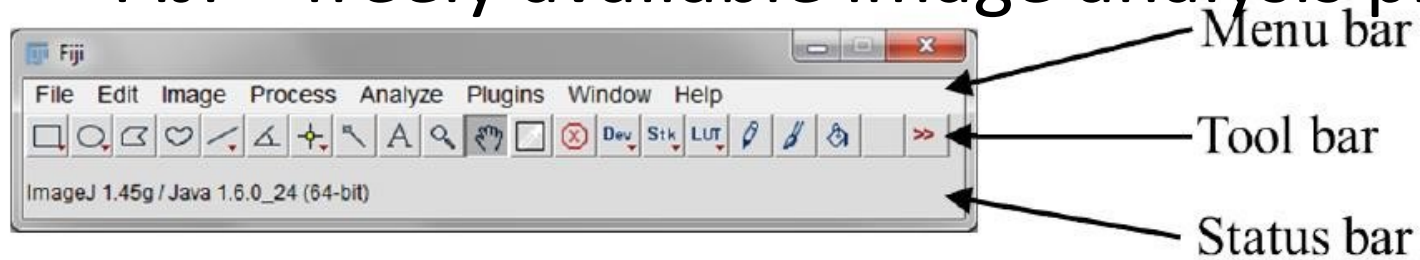
# IMAGE ANALYSIS IS PART OF THE EXPERIMENT

- Think about how to analyze your images before starting the experiment.
- Consider adapting your experiment so that quantitative image analysis can be performed easily.
- Think about controls, counter-proves, an easy to falsify null-hypothesis
- Do simple experiments.
- How can you exclude yourself from the experiment?
- Think of blinding yourself or fully automate analysis.
- One experiment usually answers just one or less questions.

# COMMON QUESTIONS IN QUANTITATIVE MICROSCOPY

1. Numbers (particles, spots, nuclei. etc)
2. Morphometry (length, Feret diameters, roundness, etc)
3. Densitometry (intensity base analysis on fluorophores)
4. Colocalization
5. Live imaging & tracking

# FIJI – freely available image analysis program



Plot window

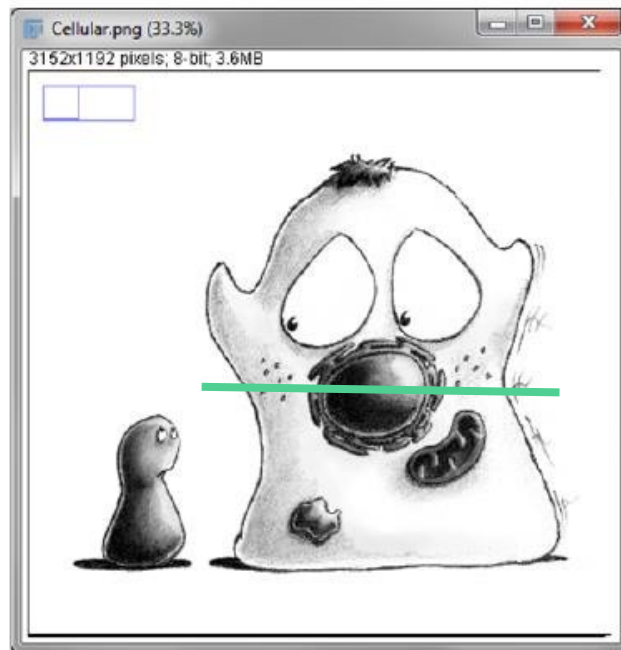
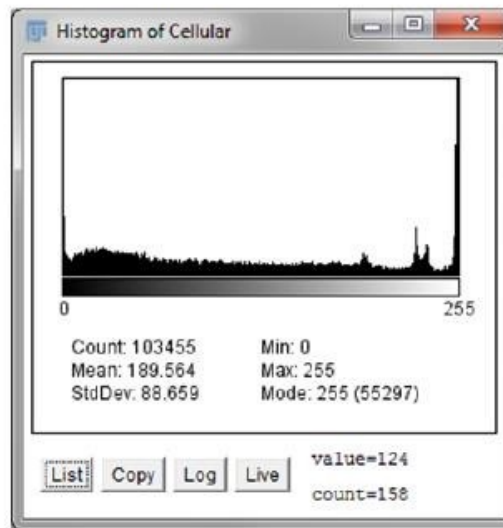


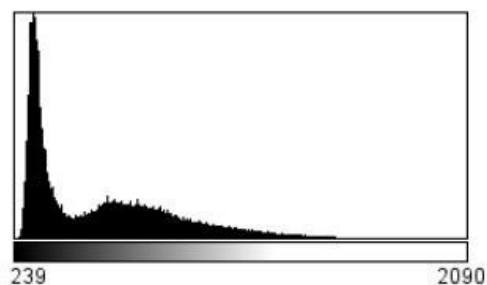
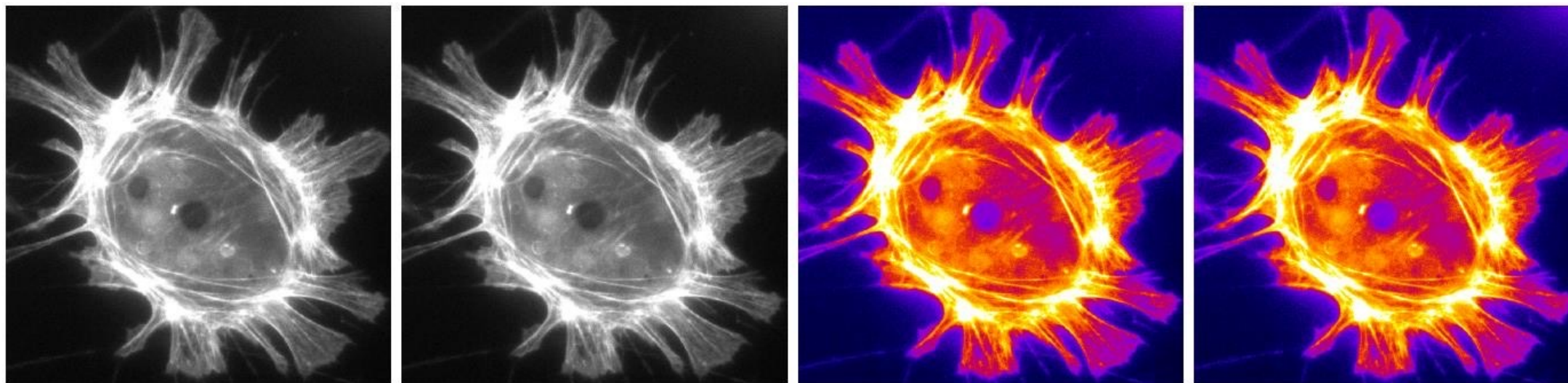
Image window



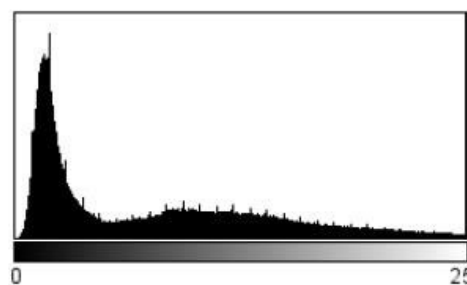
Histogram window

|   | Area  | Mean    | StdDev | Min | Max | Major   | Minor   | Angle  |
|---|-------|---------|--------|-----|-----|---------|---------|--------|
| 1 | 18972 | 249.590 | 27.457 | 14  | 255 | 209.879 | 115.095 | 0      |
| 2 | 42758 | 140.995 | 93.854 | 0   | 254 | 329.914 | 165.016 | 0      |
| 3 | 4007  | 128.757 | 75.740 | 0   | 251 | 80.927  | 63.043  | 90.000 |
| 4 | 4705  | 75.605  | 43.936 | 0   | 255 | 104.933 | 57.090  | 90.000 |
| 5 | 4705  | 92.641  | 50.162 | 0   | 221 | 104.933 | 57.090  | 90.000 |

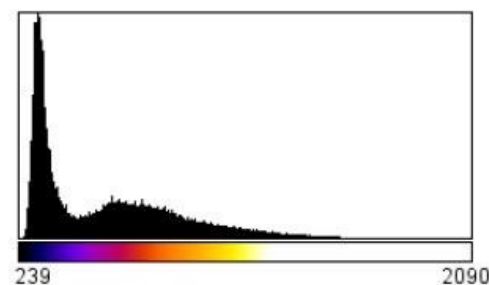
Results table



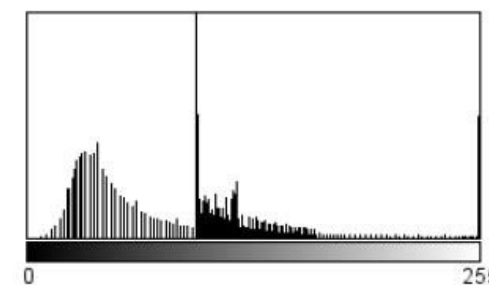
Count: 339864  
Mean: 591.429  
StdDev: 306.524  
Bins: 256  
Min: 239  
Max: 2090  
Mode: 313 (14617)  
Bin Width: 7.230



Count: 339864  
Mean: 82.006  
StdDev: 71.418  
Min: 0  
Max: 255  
Mode: 255 (10308)



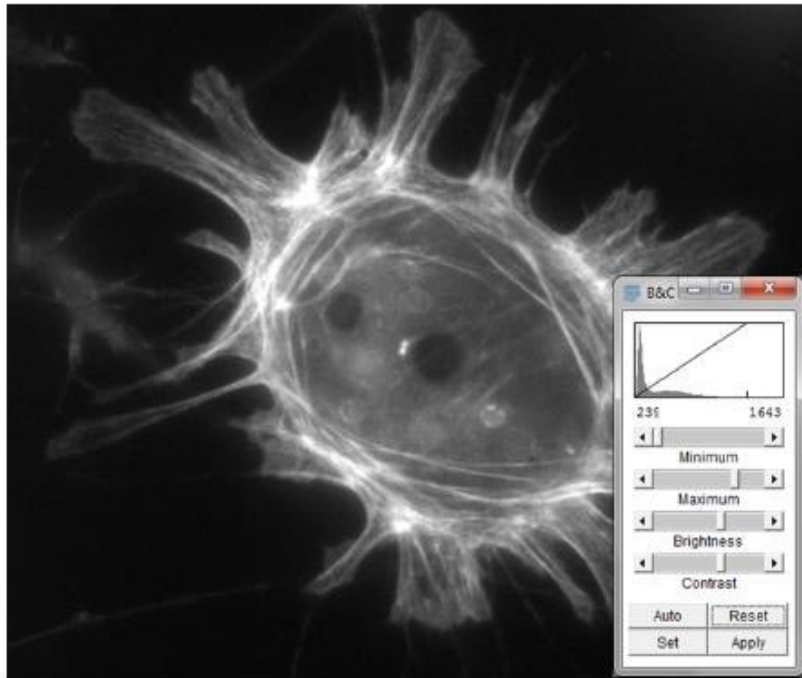
Count: 339864  
Mean: 591.429  
StdDev: 306.524  
Bins: 256  
Min: 239  
Max: 2090  
Mode: 313 (14617)  
Bin Width: 7.230



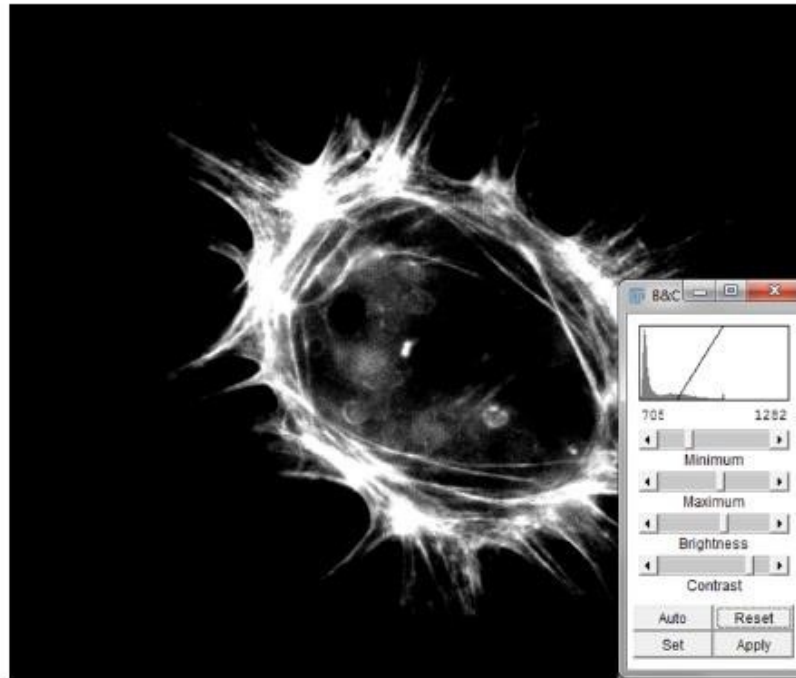
Count: 339864  
Mean: 90.544  
StdDev: 56.392  
Min: 0  
Max: 255  
Mode: 95 (21862)

(a) 16-bit (Grays LUT) (b) 8-bit (Grays LUT) (c) 16-bit (Fire LUT) (d) 8-bit (RGB)

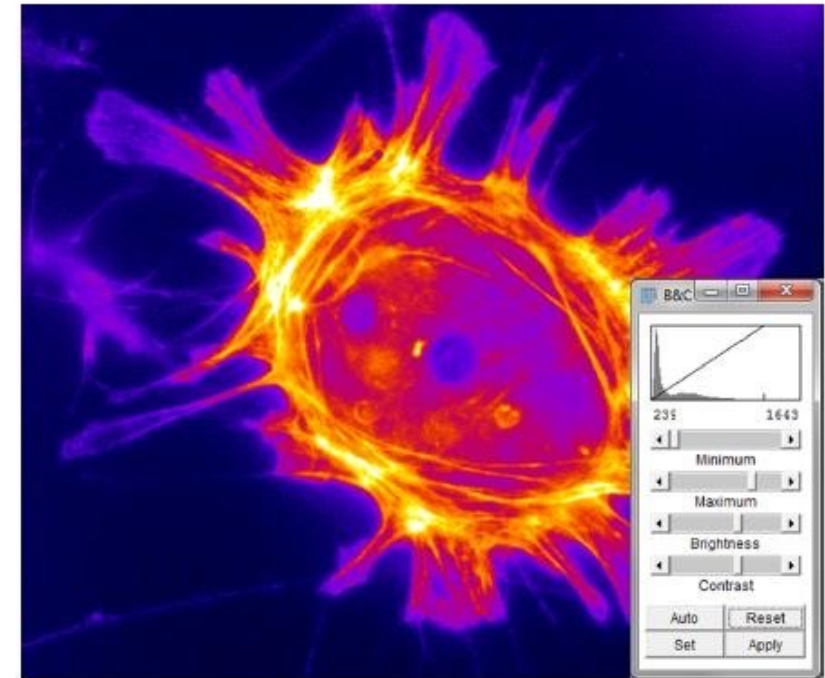
- The same image can be displayed in different ways by adjusting the contrast settings or the LUT.
- Nevertheless, despite the different appearance, the values of the pixels are the same in all three images.



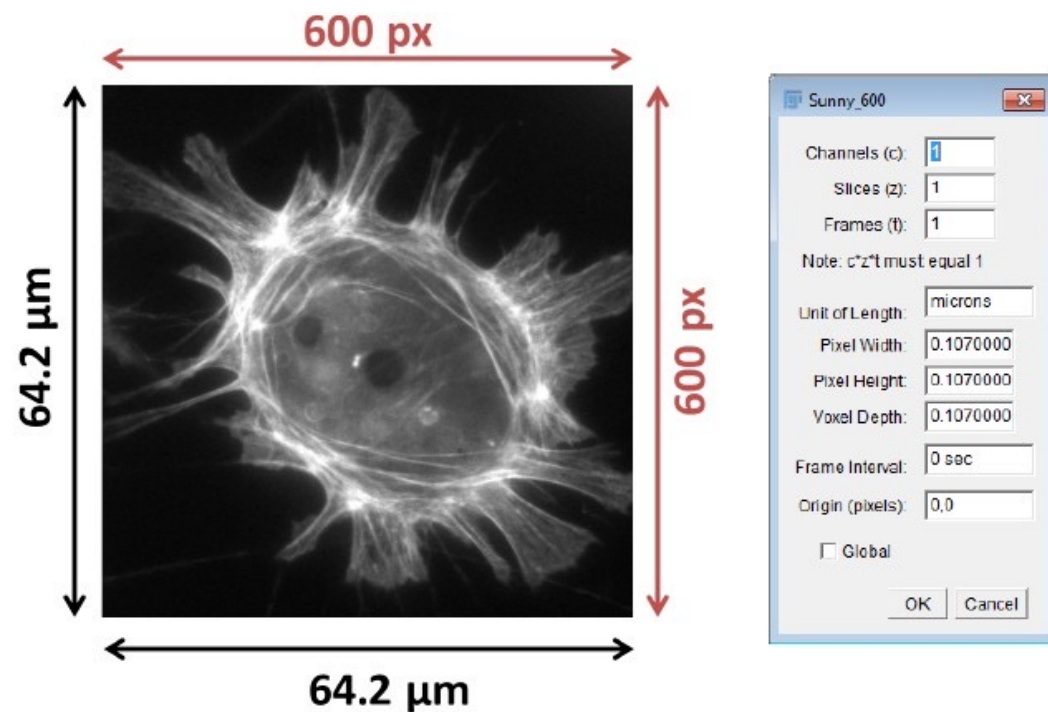
(a) Grayscale



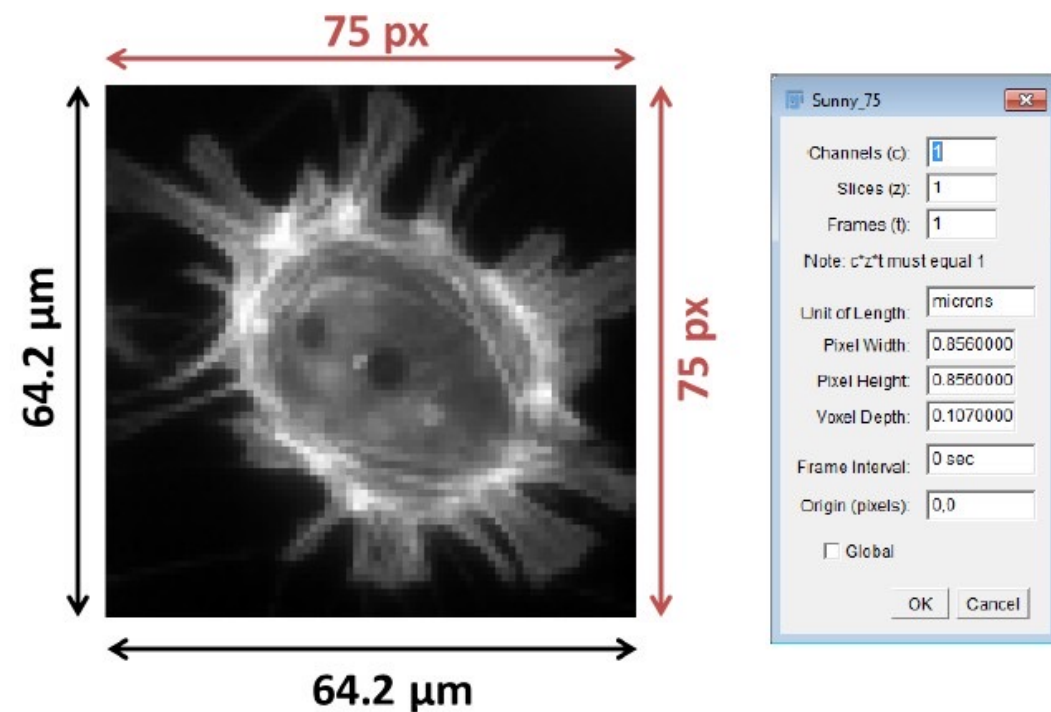
(b) Grayscale (high contrast)



(c) Fire LUT

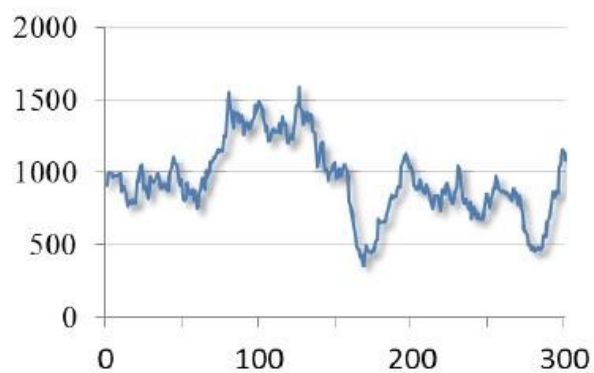


(a)  $600 \times 600$  pixel image and its properties



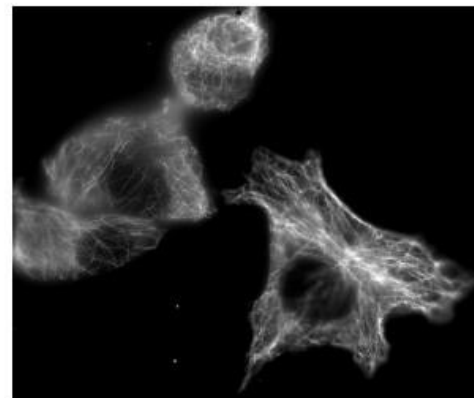
(b)  $75 \times 75$  pixel image and its properties

102

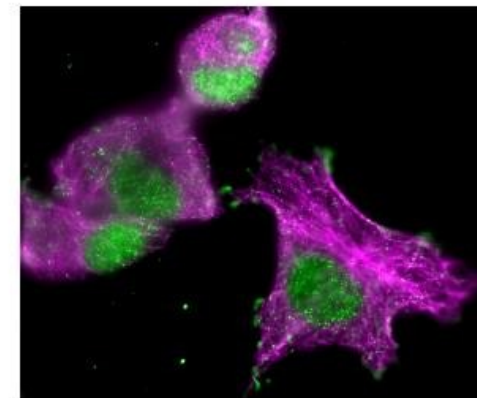


(a) 0 dimensional

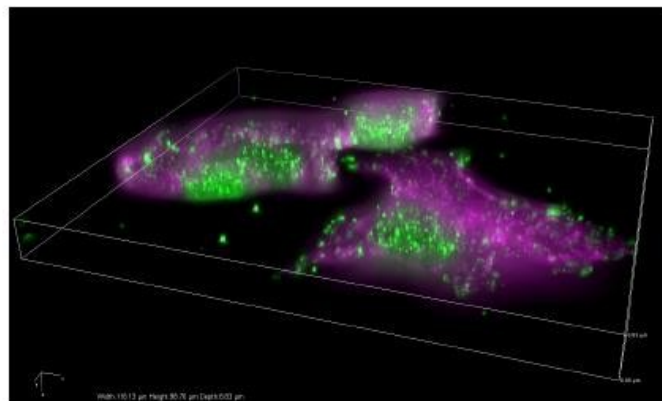
(b) 1 dimensional



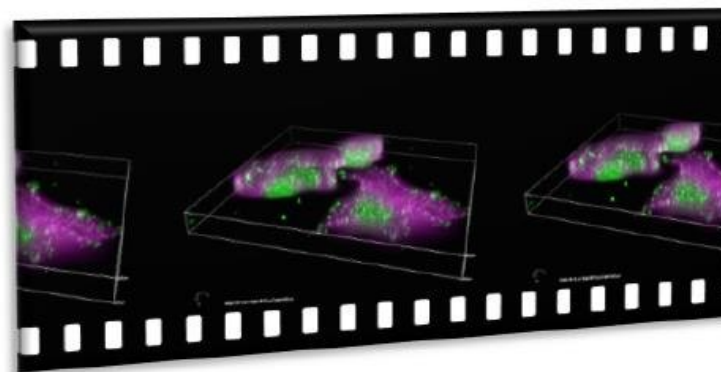
(c) 2 dimensional



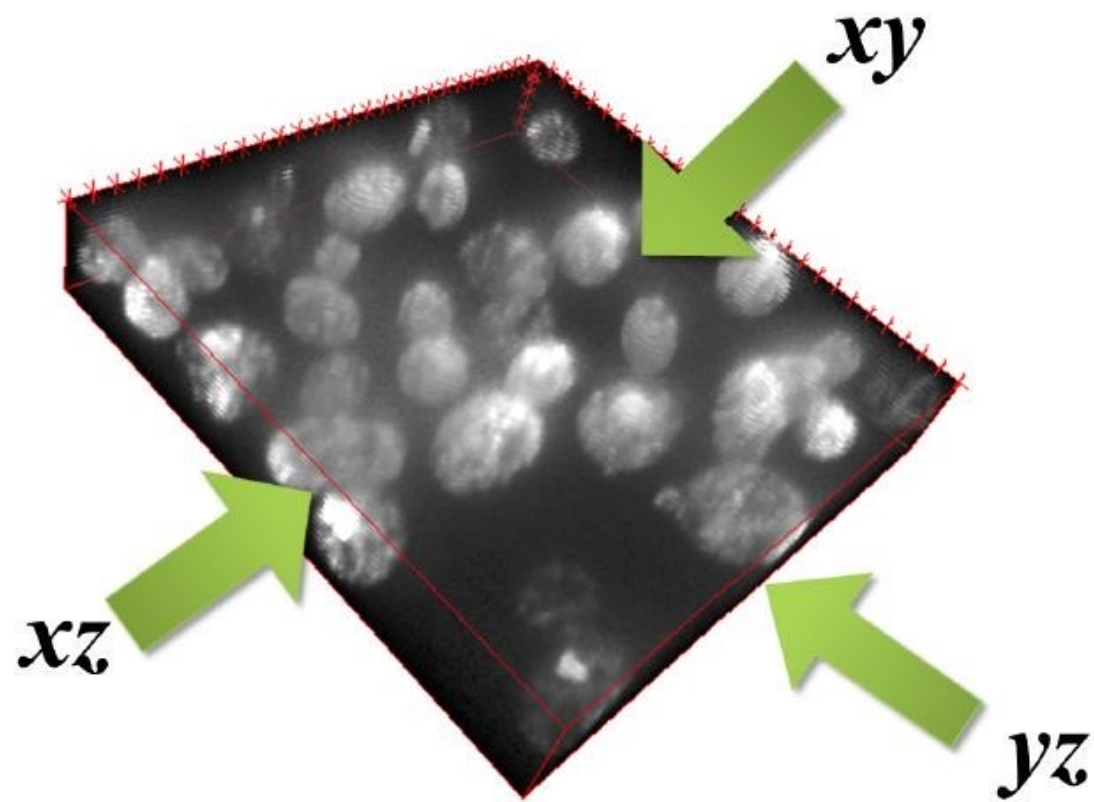
(d) 3 dimensional



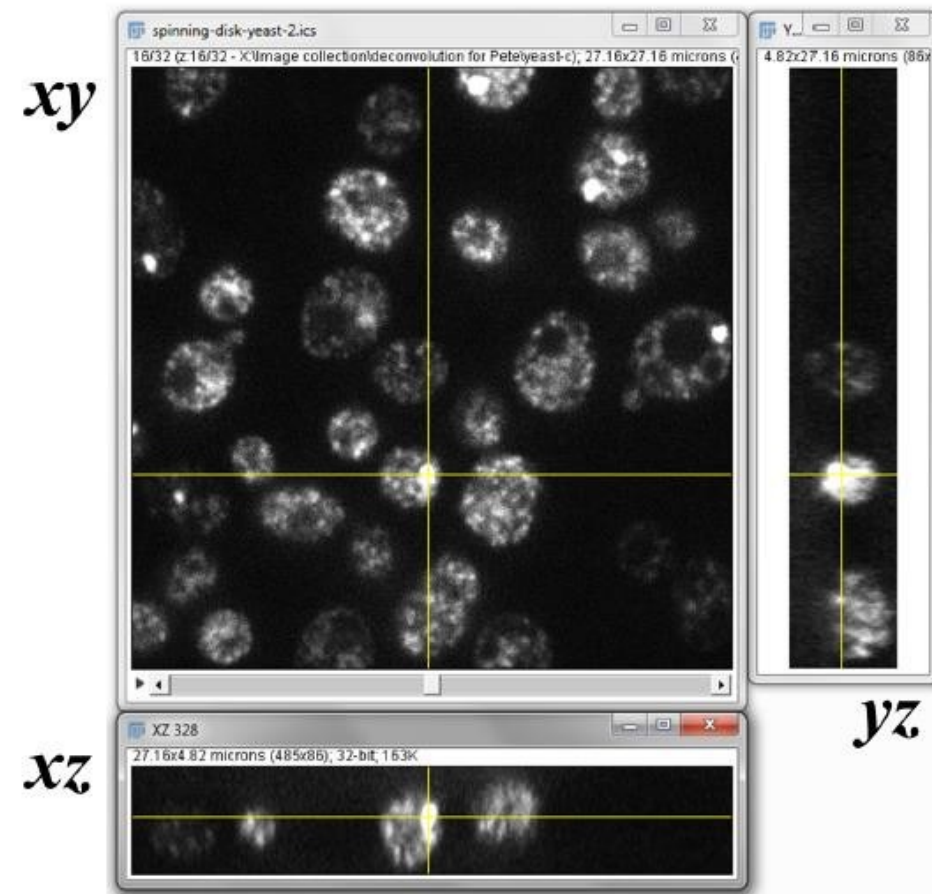
(e) 4 dimensional



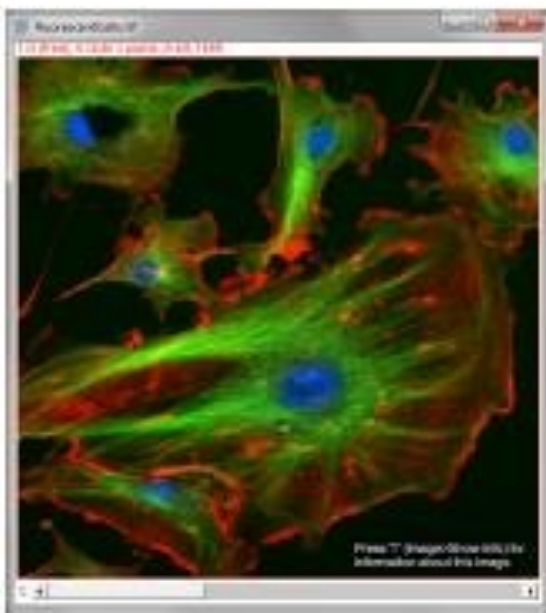
(f) 5 dimensional



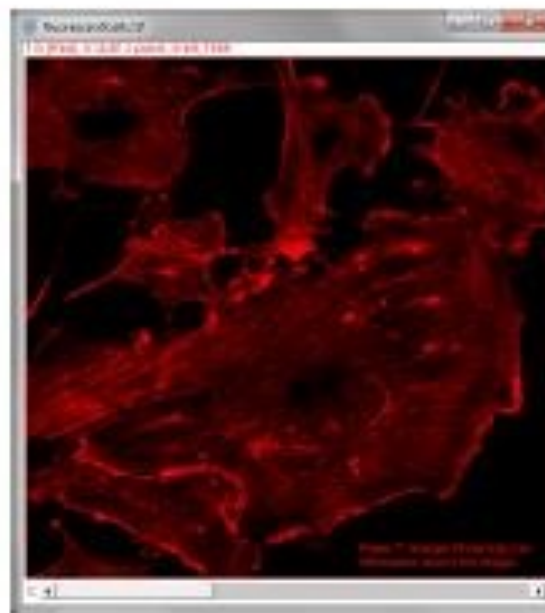
(a) Volume rendering



(b) Orthogonal views



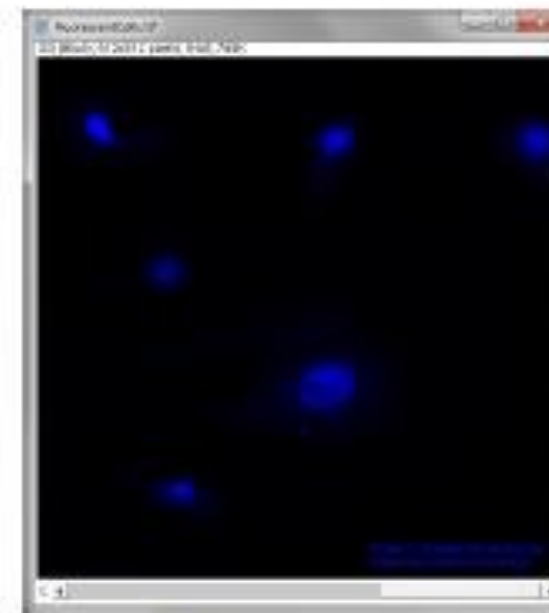
(a) Composite image



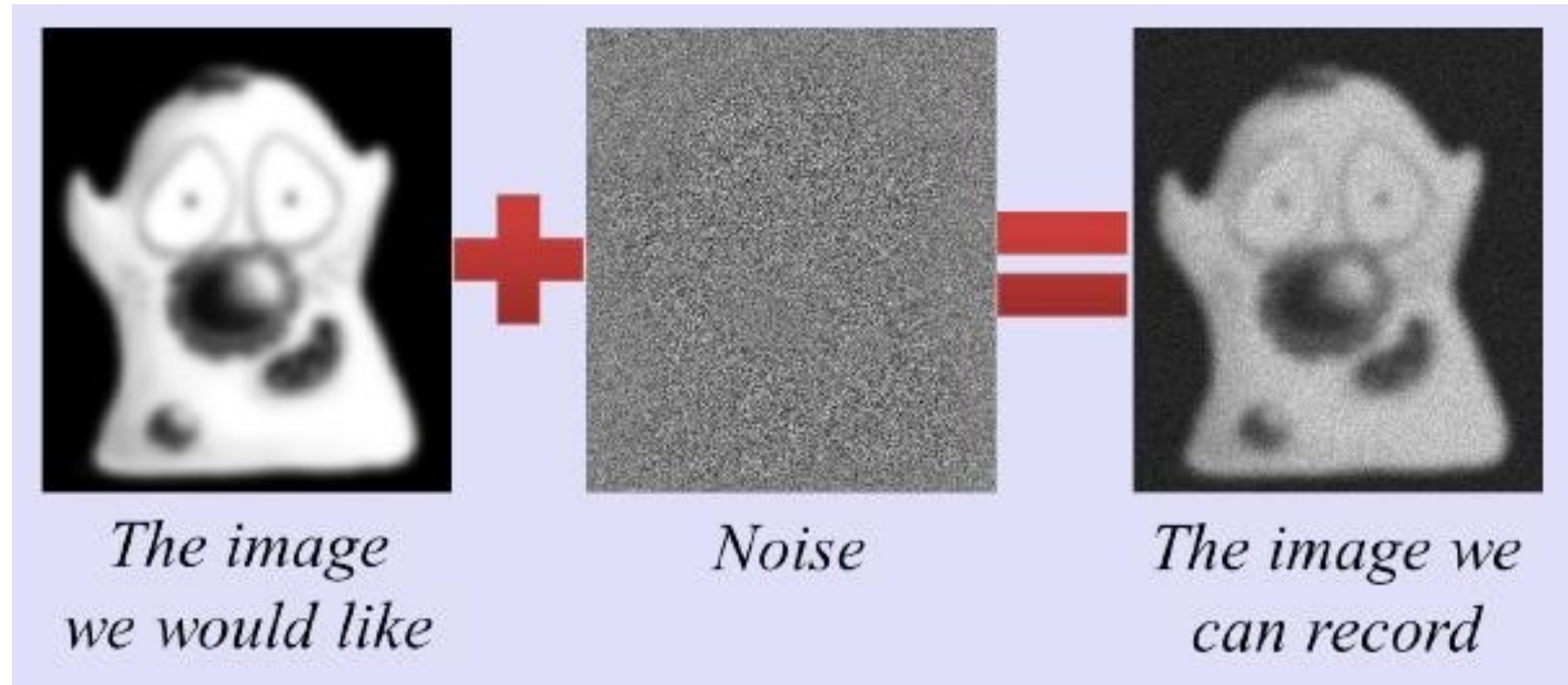
(b) Red channel

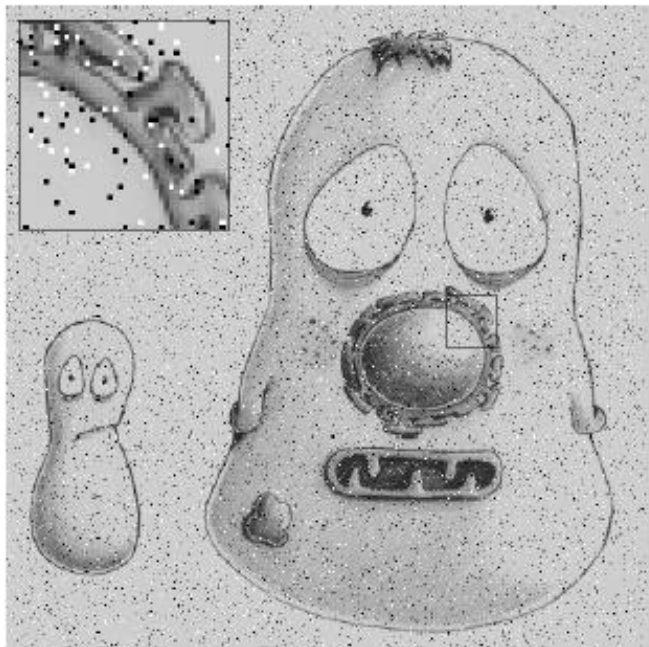


(c) Green channel

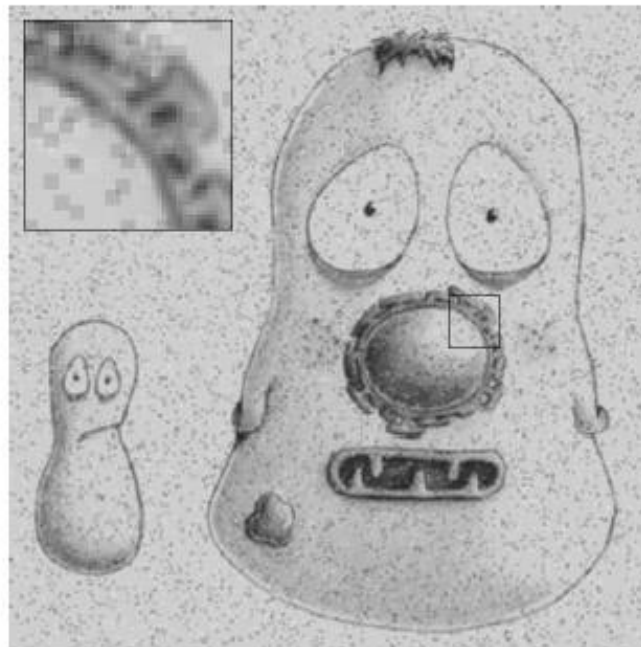


(d) Blue channel

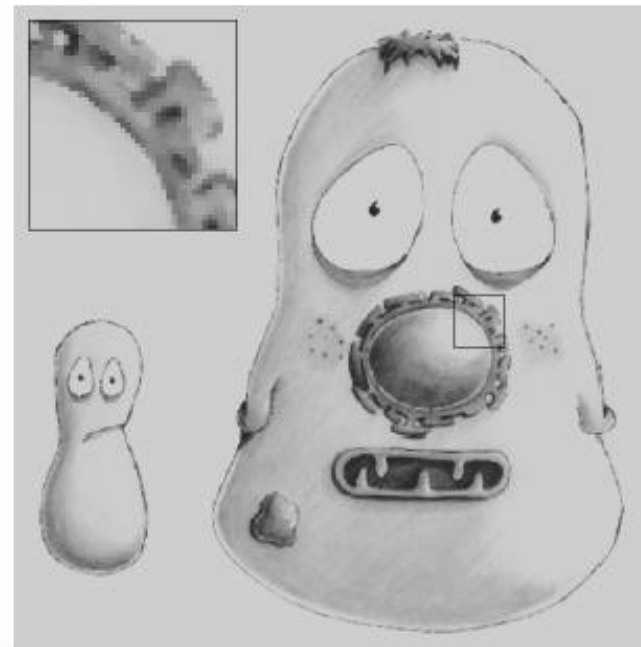




(a) Speckled image



(b) Mean filter



(c) Median filter

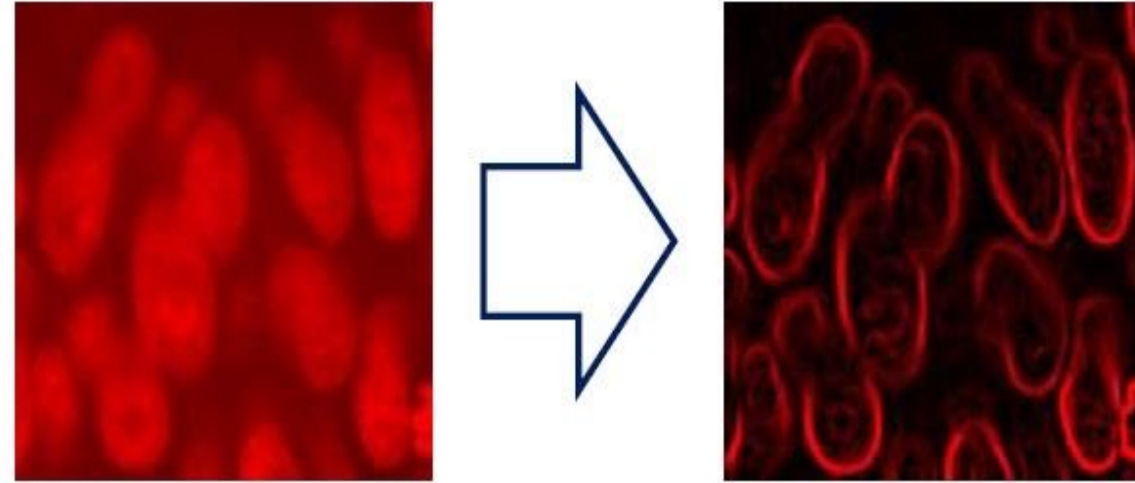
# Basic Intensity Quantification with ImageJ

- Images mostly needed to be transformed into quantifiable data
- ImageJ is useful for getting information from images, including pixel intensity.
- There are a number of different ways to get intensity information from images using the base package of ImageJ (no plugins required).

# IMAGE FILTERING

Today

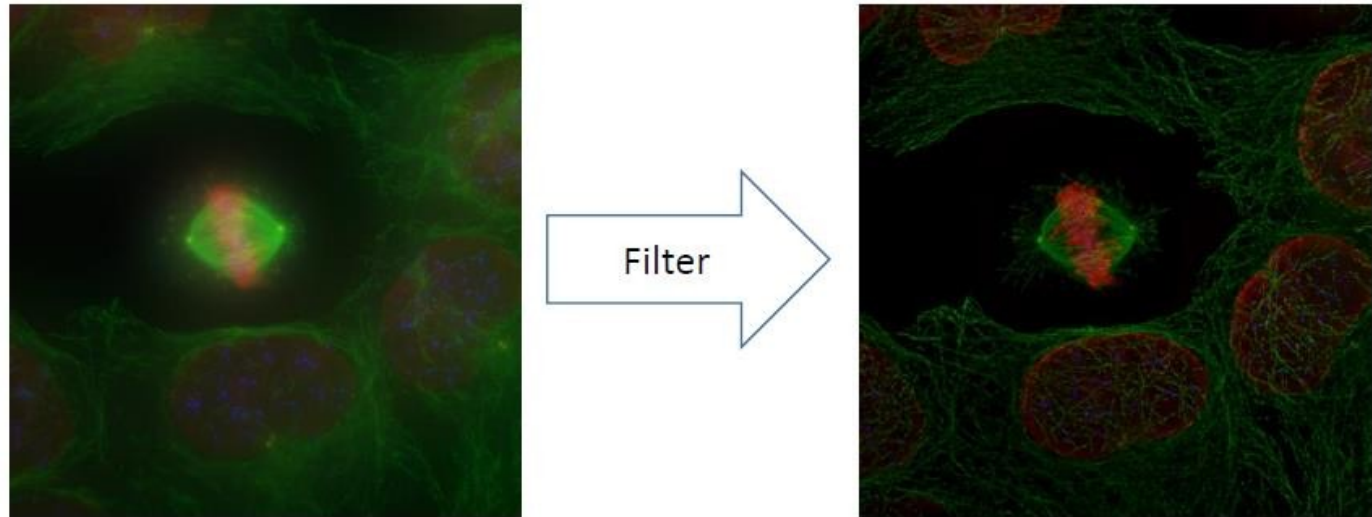
- Filters
- Image math



# FILTERS

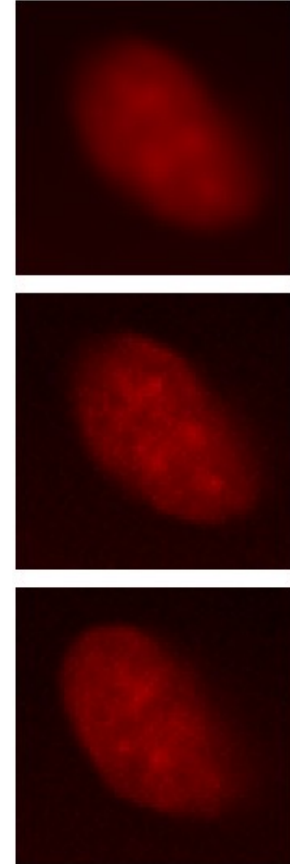
- An image processing filter is an operation on an image.
- It takes an image and produces a new image out of it.
- Filters change pixel values.
- There is no “best” filter. Which filter fits your needs, depends on the context.
- Filters do not do magic. They can not make things visible which are not in the image.

- Application examples
  - Noise-reduction
  - Artefact-removal
  - Contrast enhancement
  - Correct uneven illumination



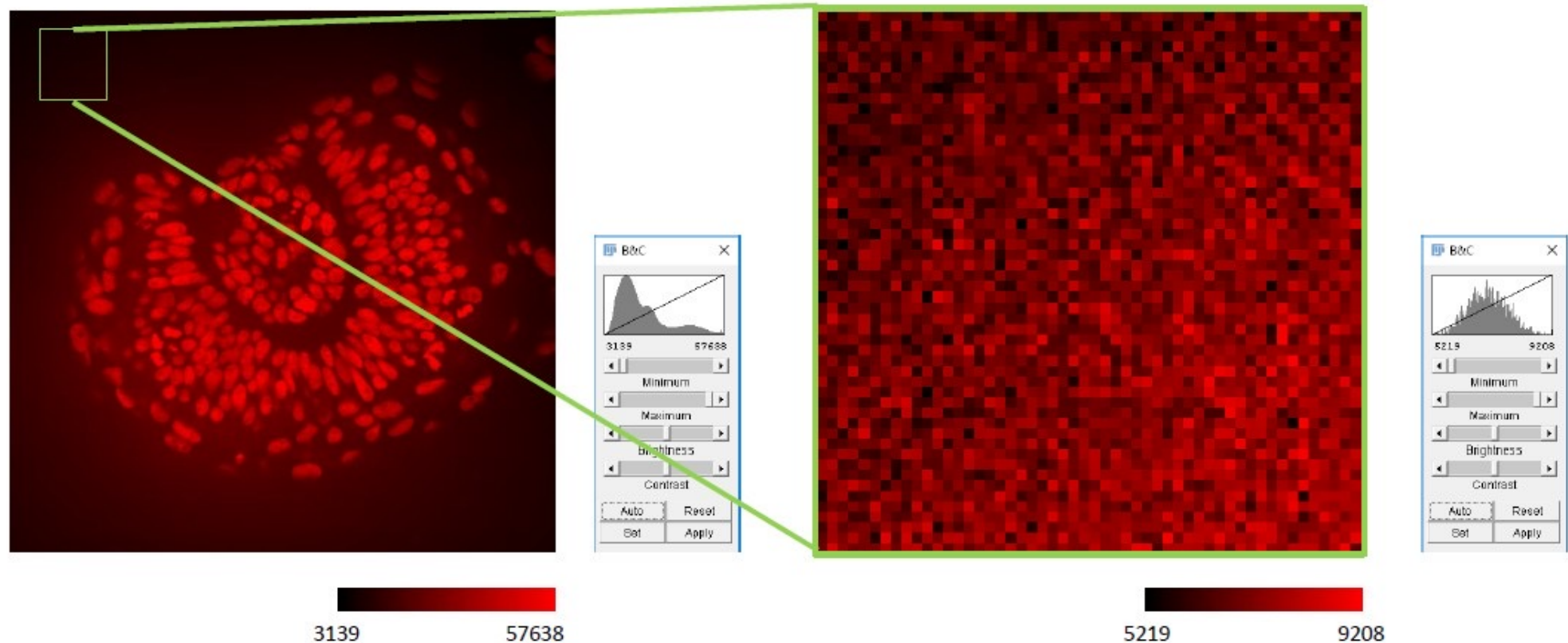
# EFFECTS HARMING IMAGE QUALITY

- Noise is a general term for unwanted modifications that a signal may suffer during capture, storage, transmission, processing, or conversion.
- In microscopy, image quality suffers from
  - shot noise: Statistical variation of the photons arriving at the camera
  - dark noise: Statistical variation of how many electrons are generated if a photon arrives in a camera pixel (temperature dependent).
  - read out noise: introduced by the electronics, especially the pixel and the analog-digital-converter
  - Physical/optical effects: aberrations, defocus
  - Biological/physiological/structural effects: motion, diffusion



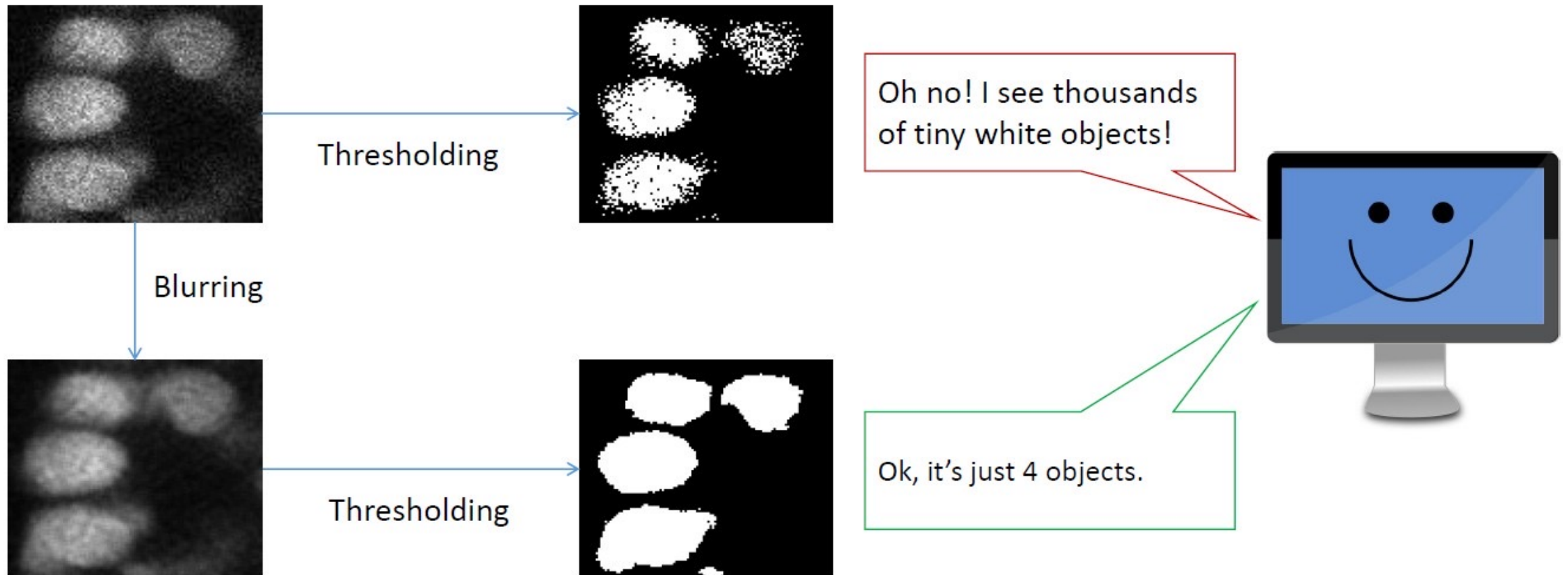
# NOISE

- When dealing with noise in microscopic image processing, we mostly mean the noise visible in the images.



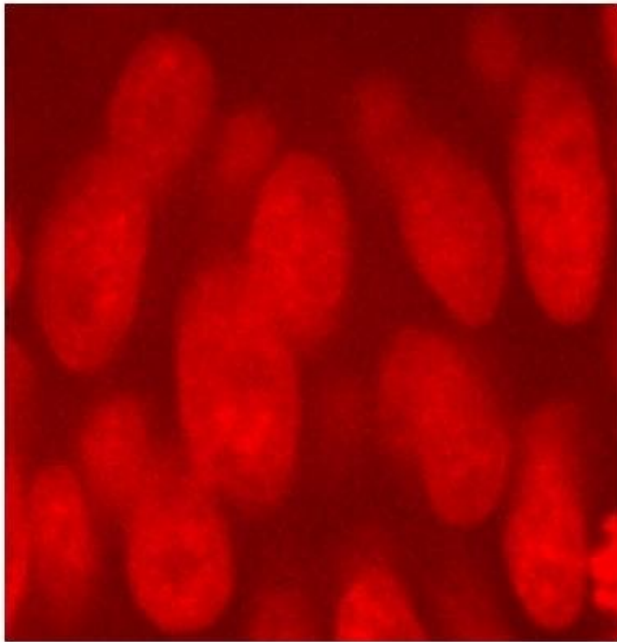
# IMAGE CORRECTION/NOISE REMOVAL

- We need to remove the noise to help the computer *interpreting* the image

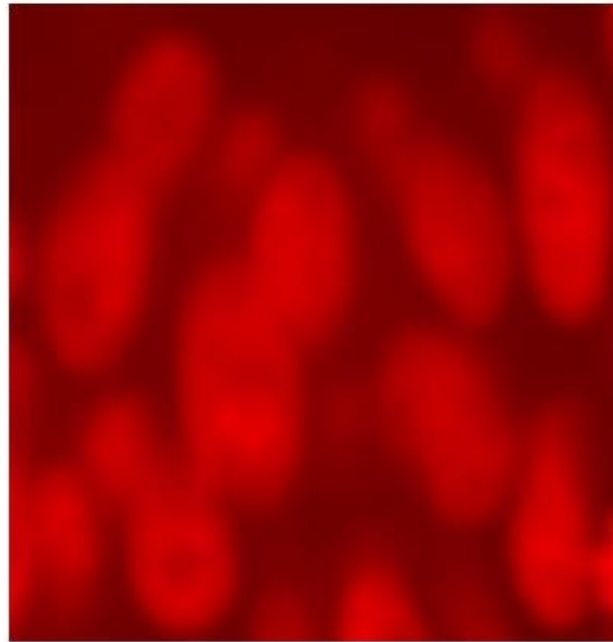


# IMAGE CORRECTION/NOISE REMOVAL

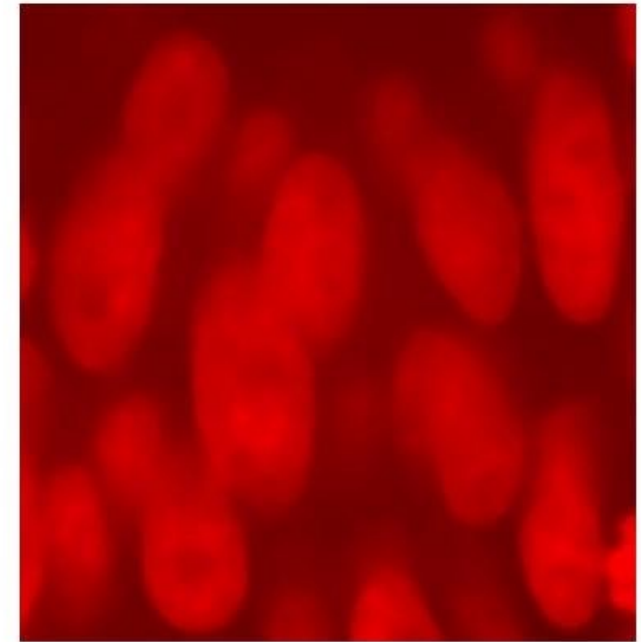
- Noise removal using *filters*



Original image



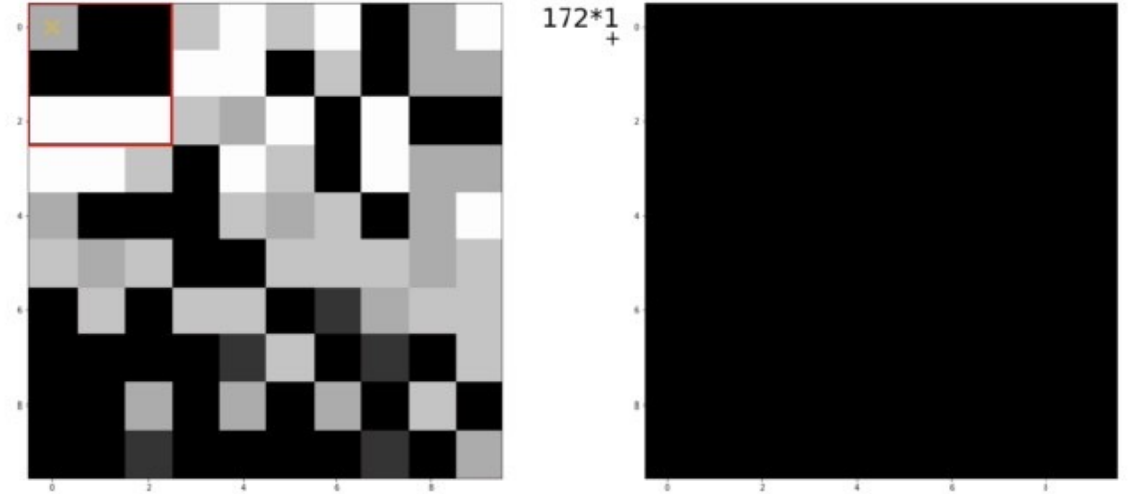
Gaussian blur filtered



Median filtered

# LINEAR FILTERS

- Linear filters replace each pixel value with a linear combination of surrounding pixels
  - Basically, linear filtering is a Convolution
  - It needs a kernel (weight template)
  - Result: new image where each pixel is replaced by the weighted sum of pixel values in the neighbourhood.



- Kernels are matrices describing a linear filter

Mean filter, 3x3 kernel

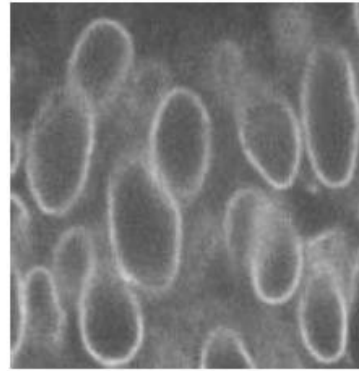
$$\begin{bmatrix} 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \end{bmatrix}$$

# LINEAR FILTERS

- Terminology:
  - “We convolve an image with a kernel.”
  - Convolution operator: \*

- Examples

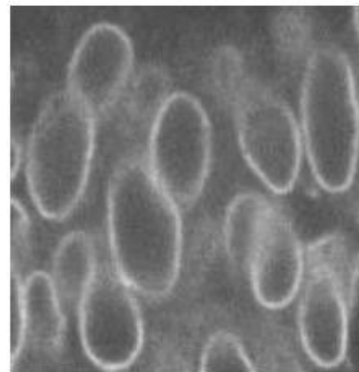
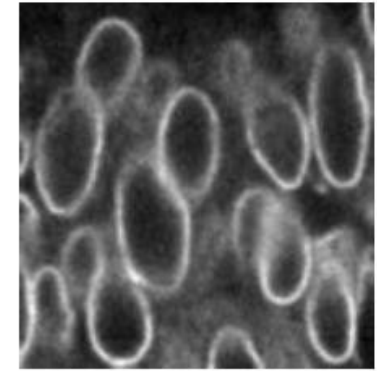
- Mean
- Gaussian blur
- Sobel-operator
- Laplace-filter



\*

$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & 8 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

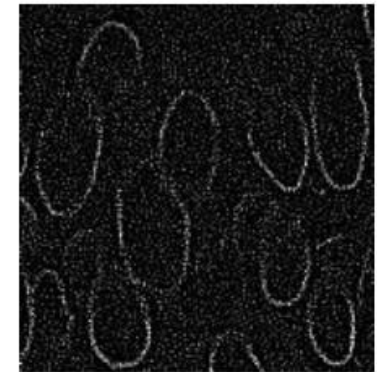
=



\*

$$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$

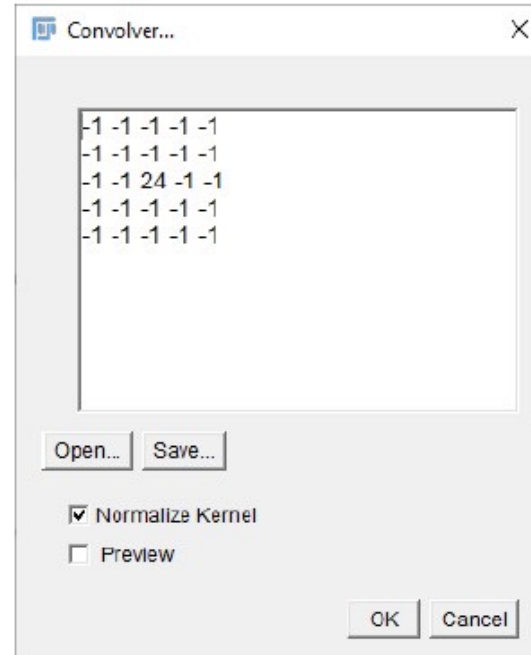
=



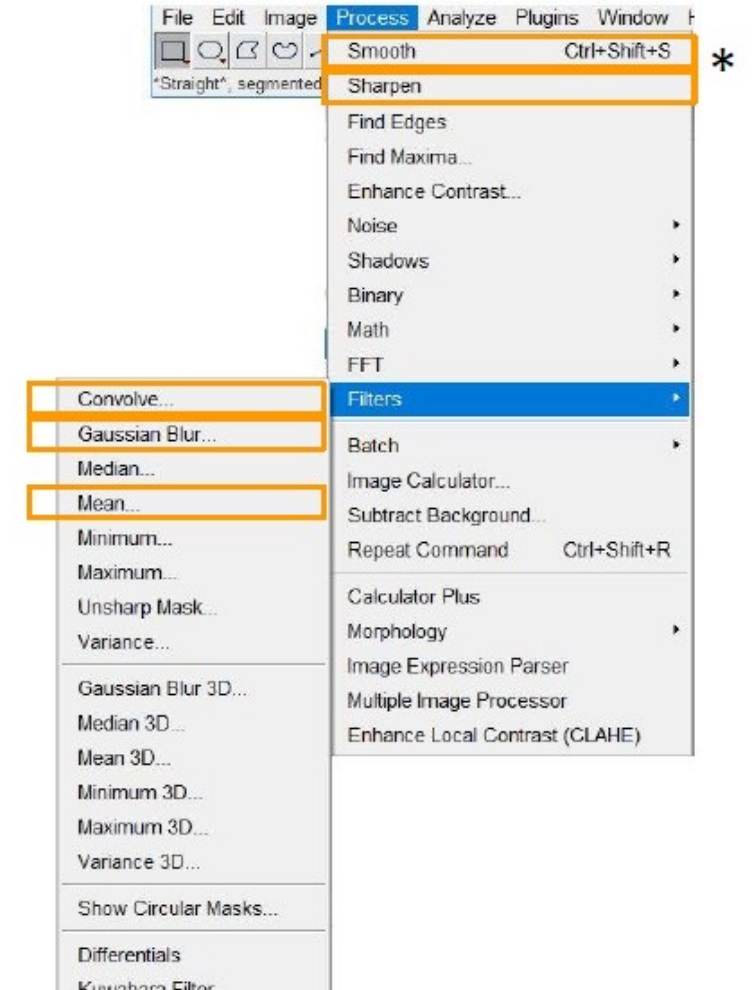
# LINEAR FILTERS

Linear filters:

- Convolve...
  - Allows us to specify our kernel



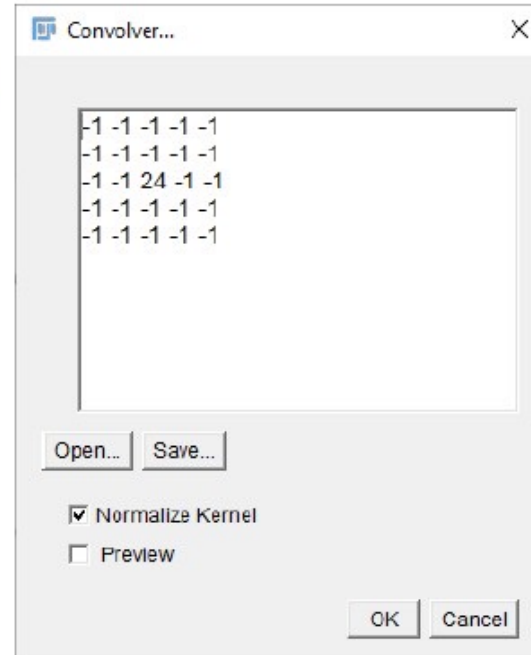
- Gaussian blur
- Mean
  - \* Smooth is a 3x3 Mean filter



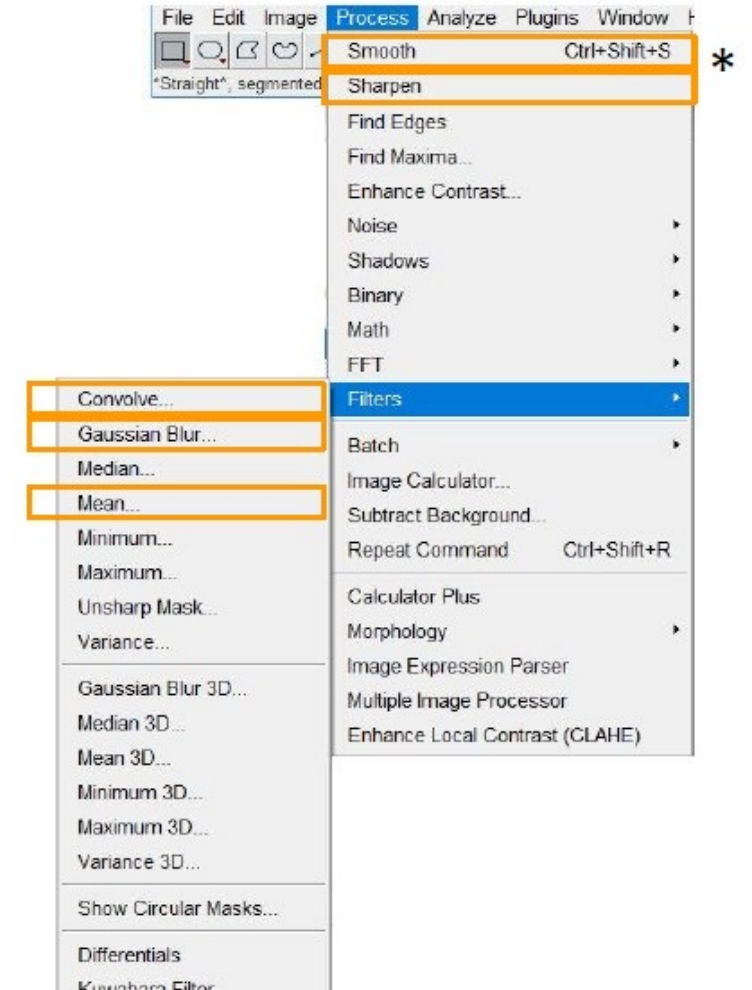
# LINEAR FILTERS

Linear filters:

- Convolve...
  - Allows us to specify our kernel

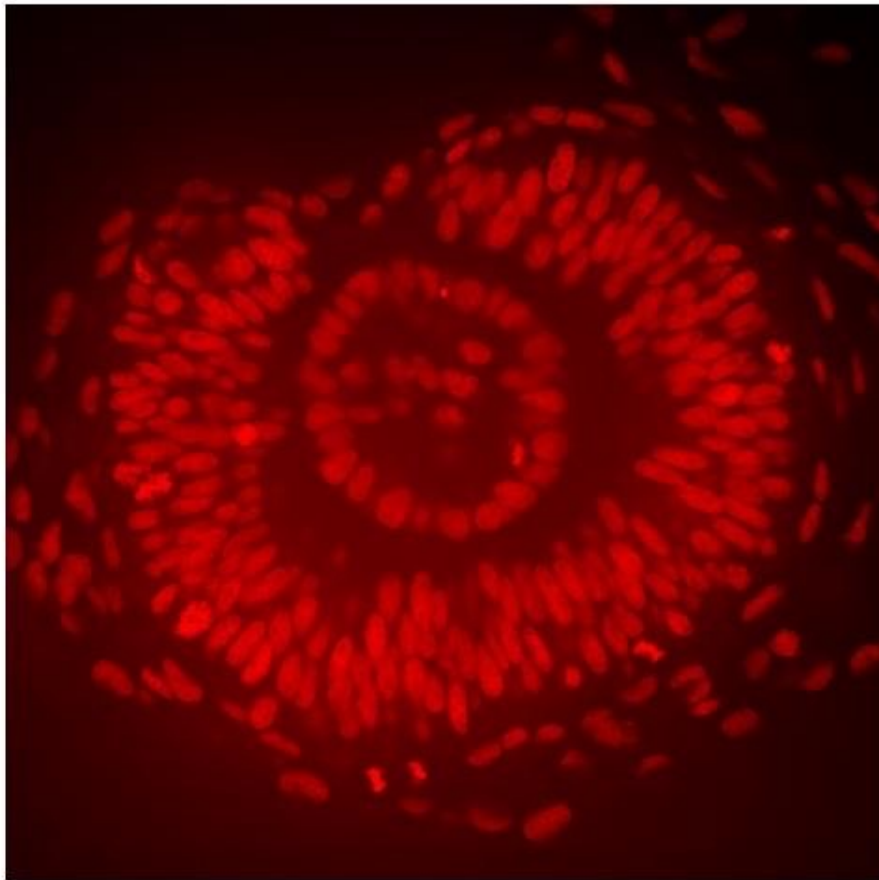


- Gaussian blur
- Mean
  - \* Smooth is a 3x3 Mean filter

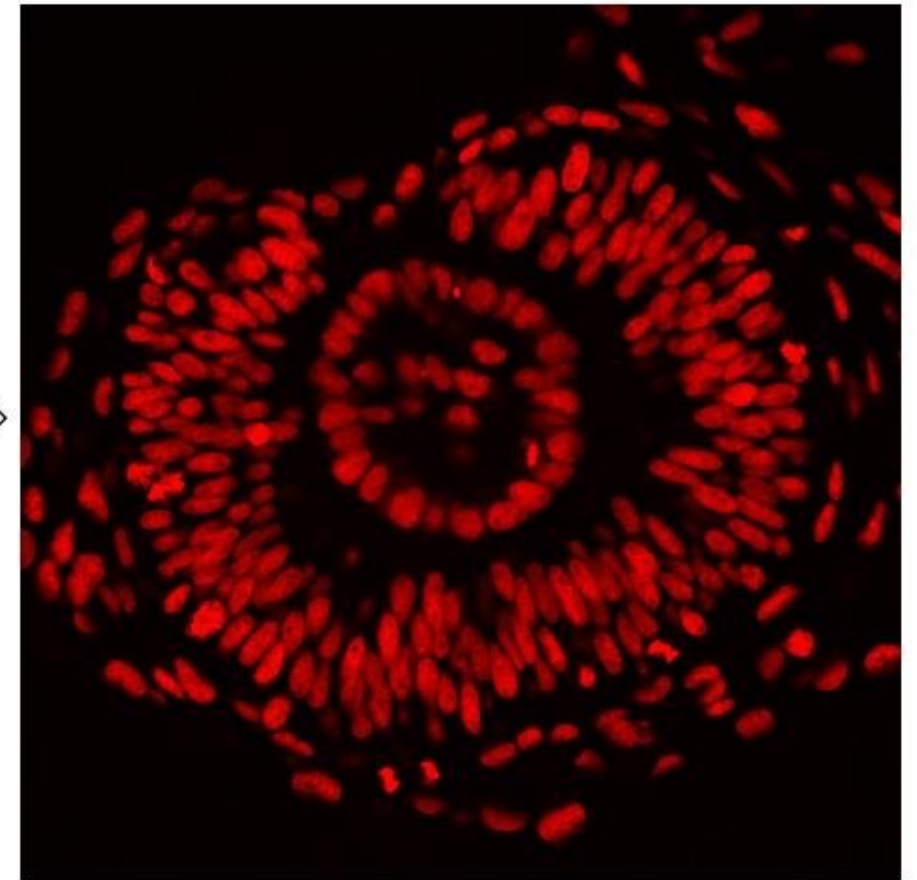


# BACKGROUND REMOVAL

- Differentiating objects is easier if their background intensity is equal.

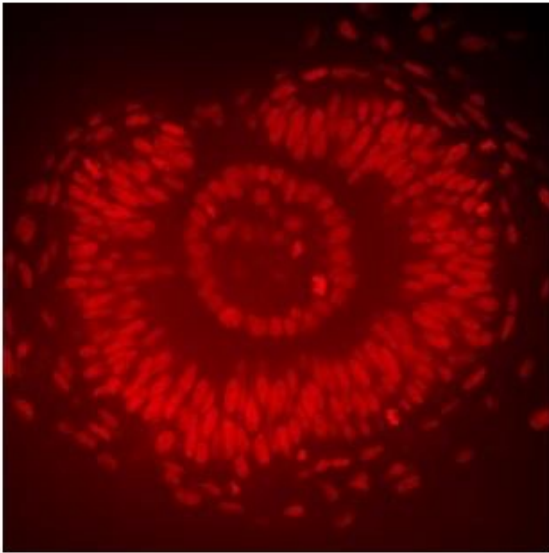


Subtract  
background

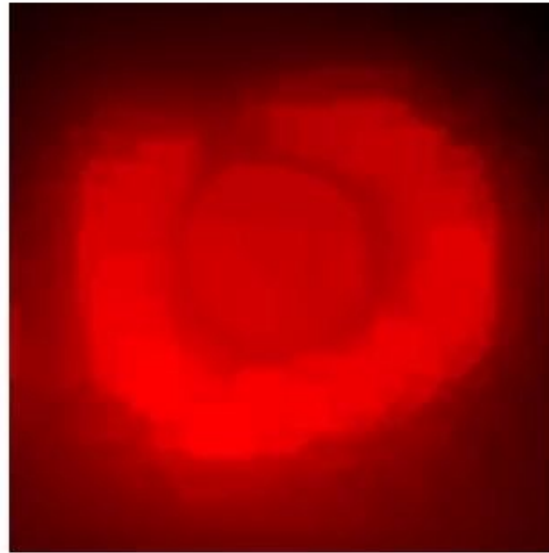


# BACKGROUND REMOVAL

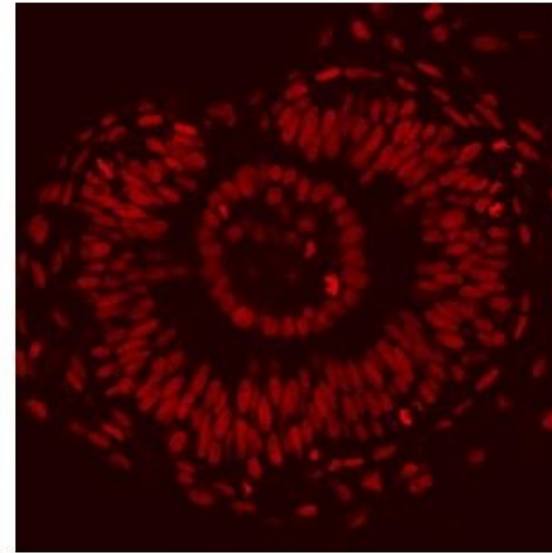
- Depending on the effect we want to correct for, it might make sense to divide an image by its background.



Original image



Extracted  
background



Subtracted  
background

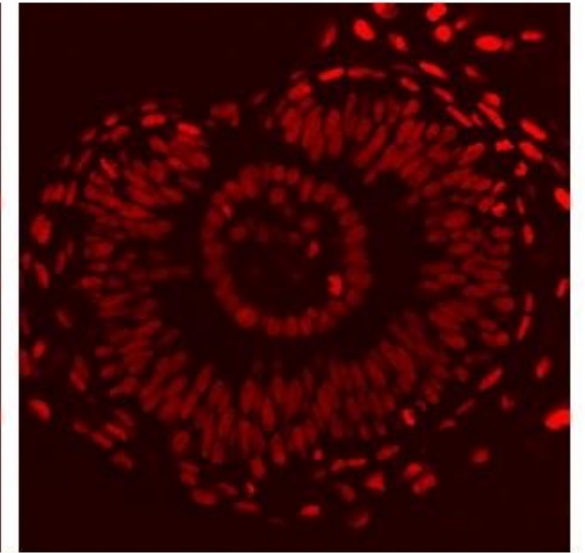
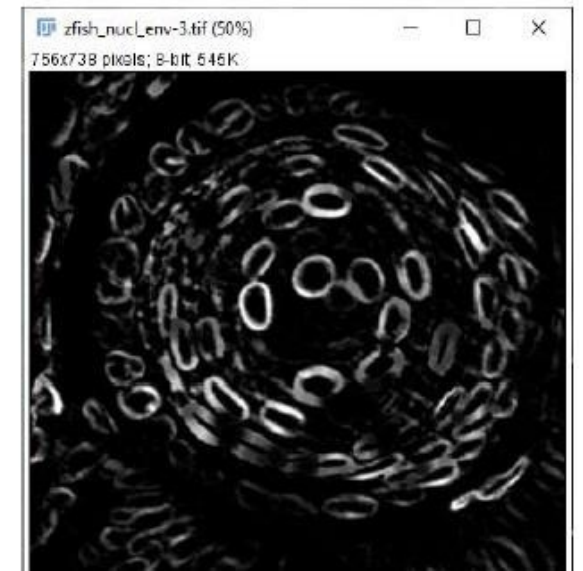
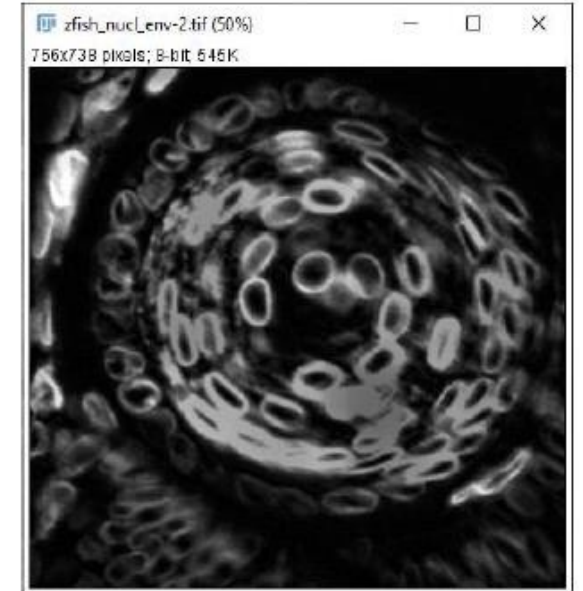
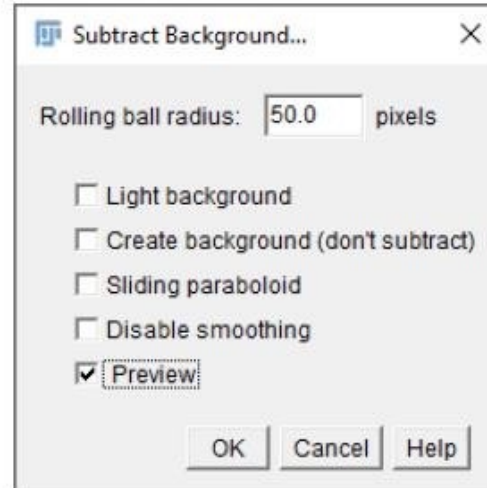


Image divided by  
background

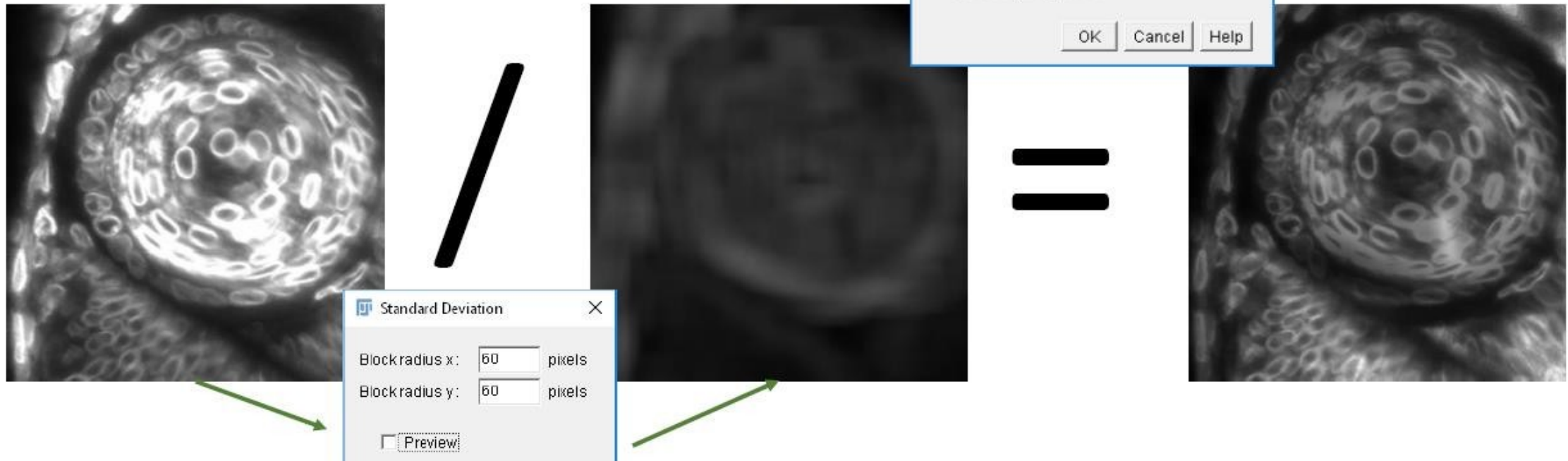
# BACKGROUND REMOVAL

- Process > Subtract background...
  - The rolling ball radius should be a bit larger than the objects you are analyzing.
  - Try the preview checkbox while changing parameters!



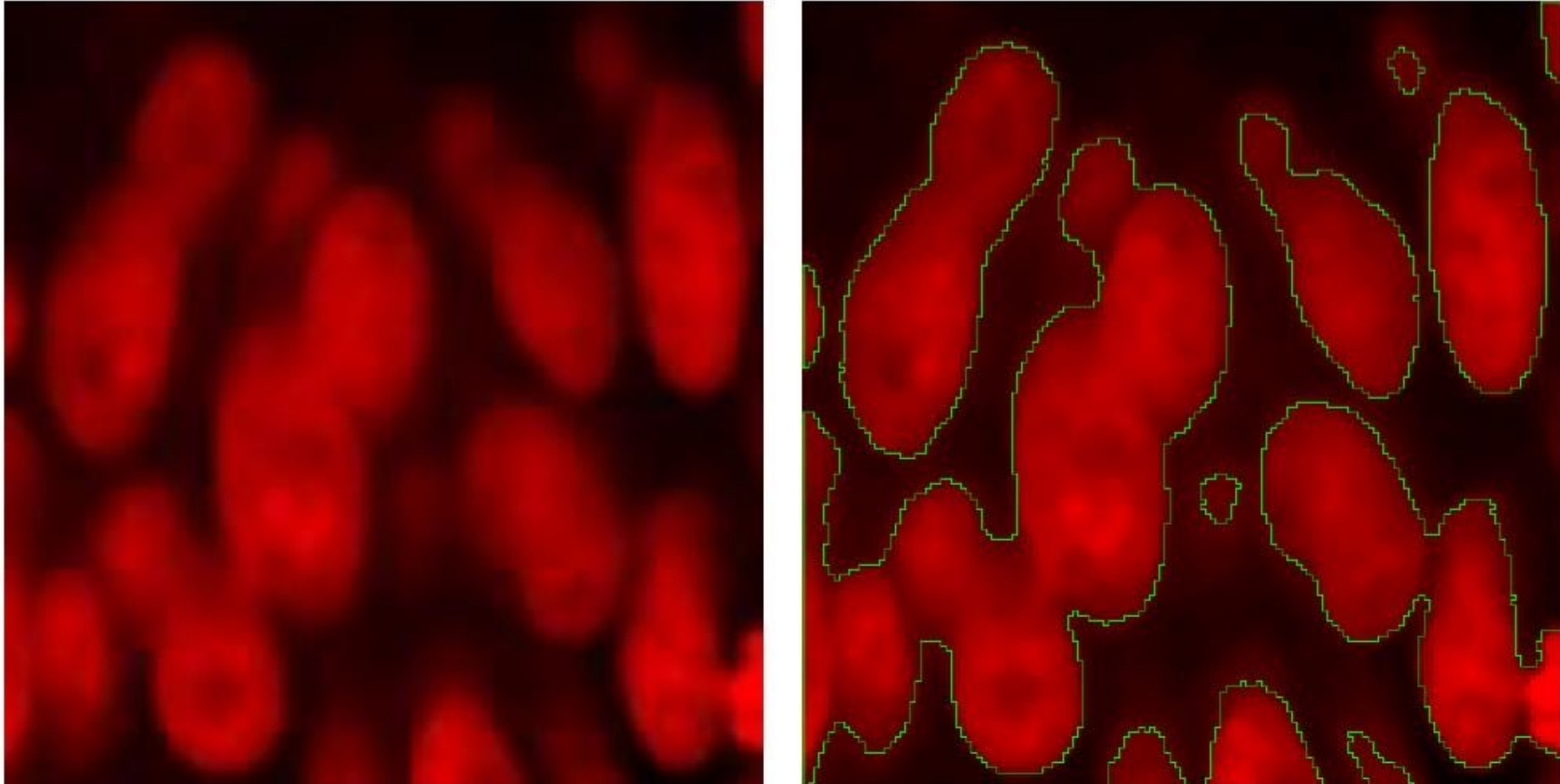
# BACKGROUND REMOVAL

- Plugins > Integral image filters > Standard deviation
- Process > Image calculator



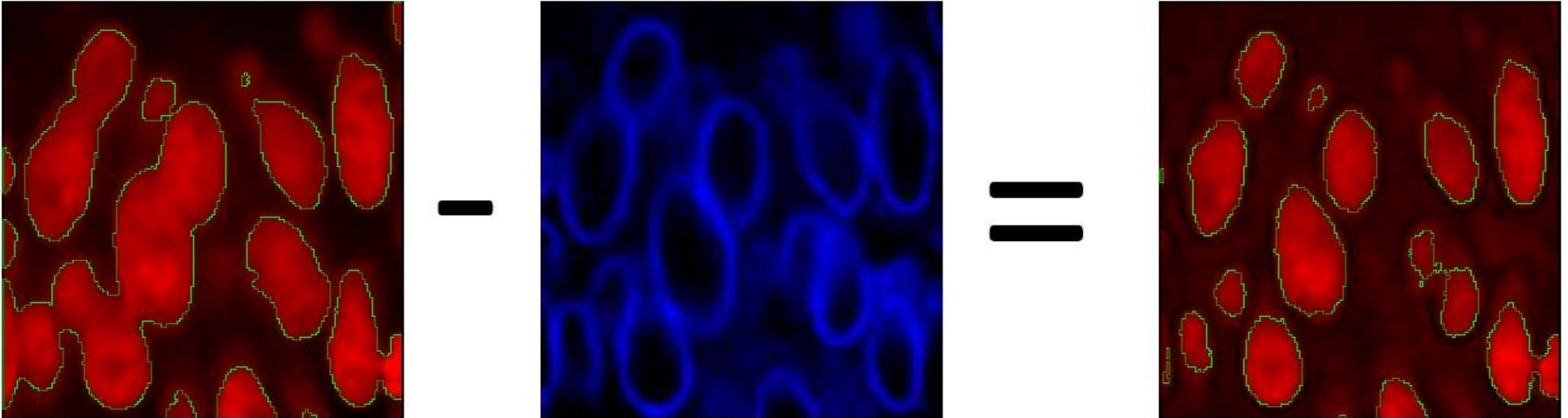
# MATHEMATICAL OPERATIONS ON IMAGES

- It might be hard to differentiate objects from single images...



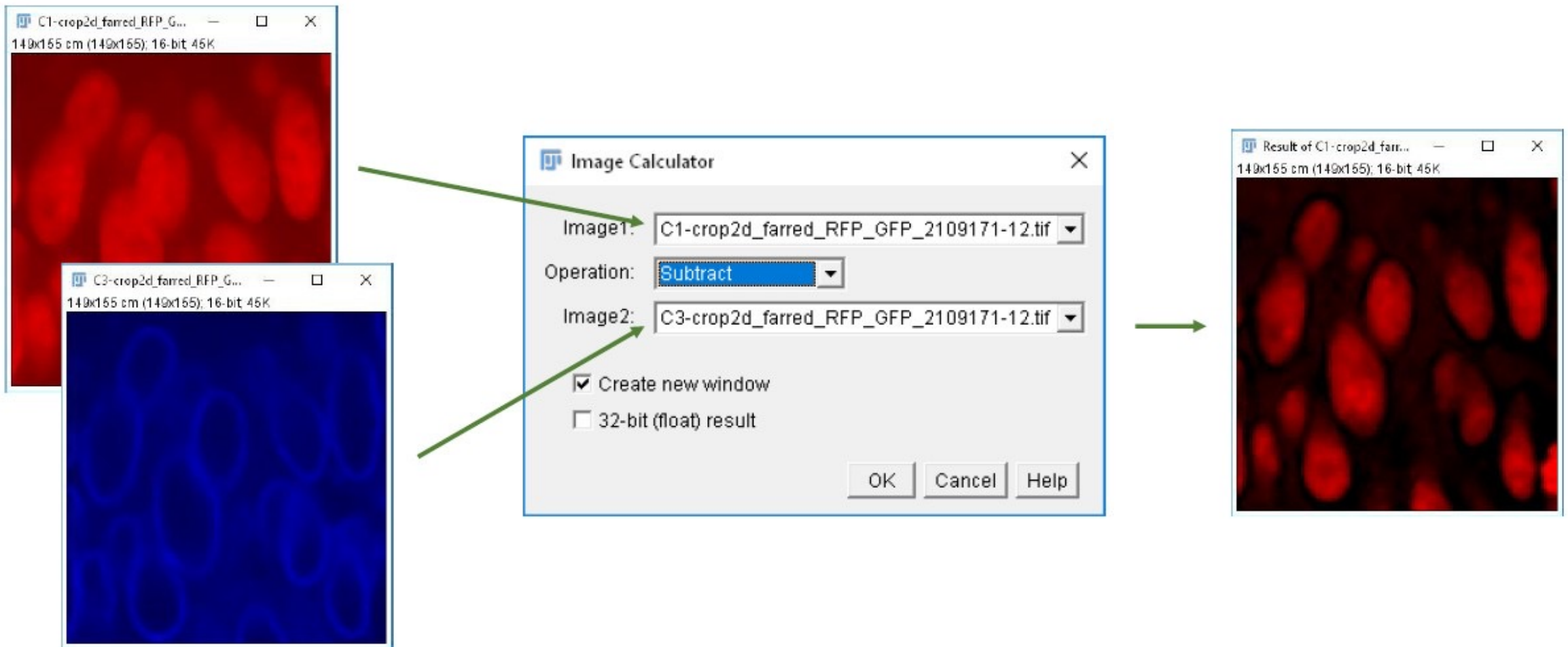
# MATHEMATICAL OPERATIONS ON IMAGES

- Segmentation of nuclei can be improved by subtracting a channel visualizing nuclei envelope from the nuclei channel.



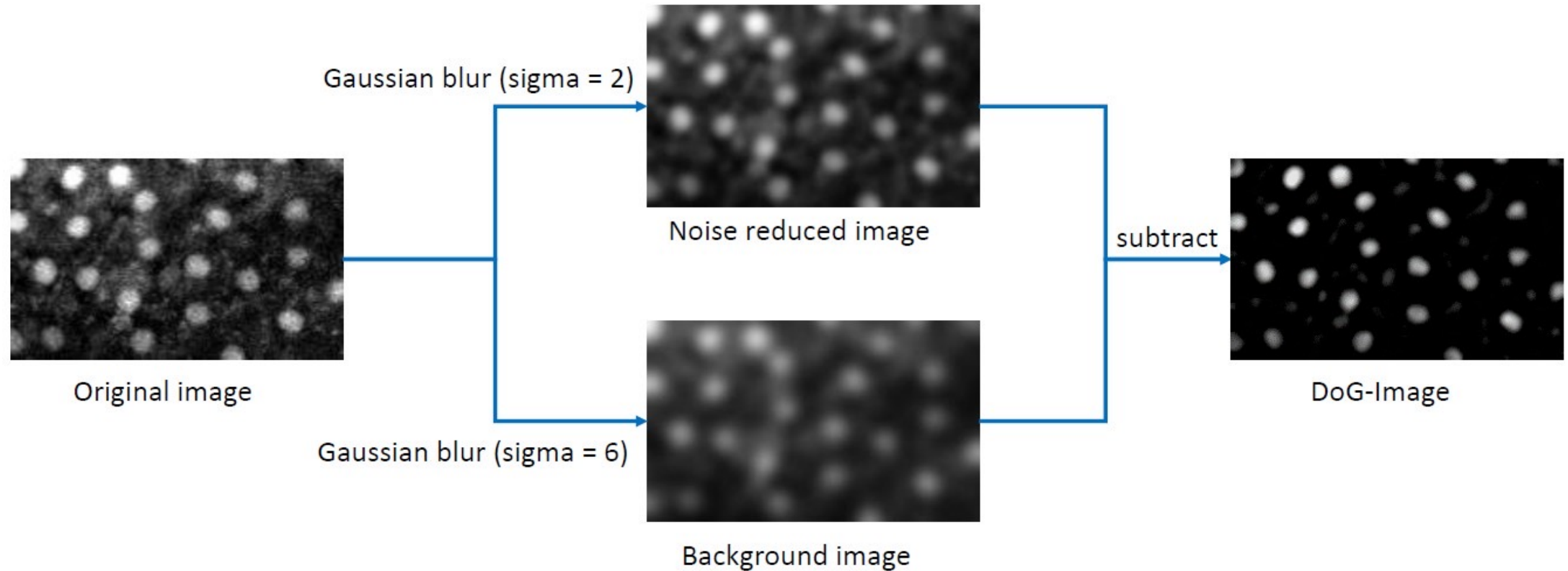
# MATHEMATICAL OPERATIONS ON IMAGES

- Use the *Image Calculator* to subtract, add, multiply or divide images.



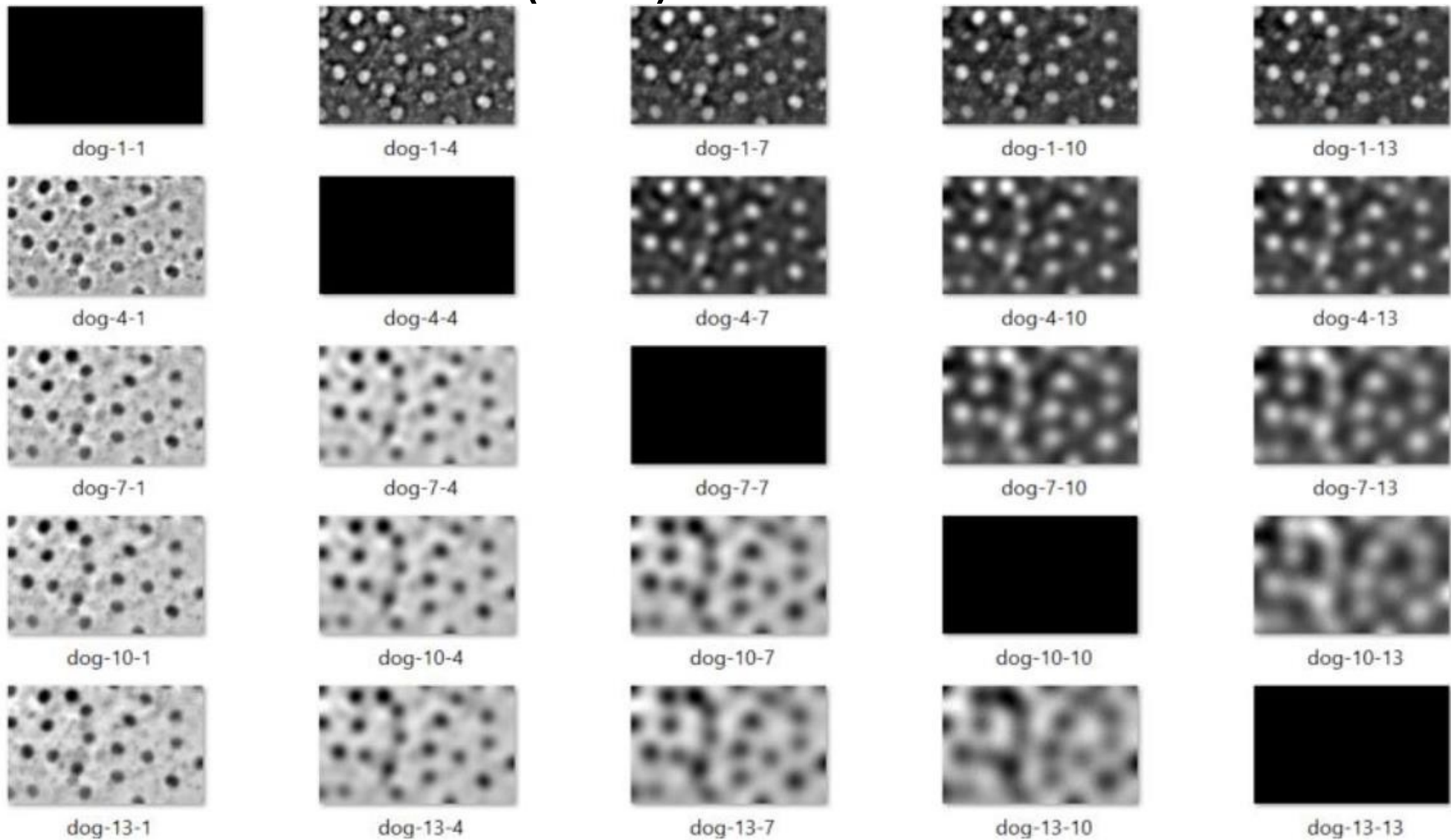
# DIFFERENCE OF GAUSSIAN (DoG)

- Improve image in order to detect bright objects.



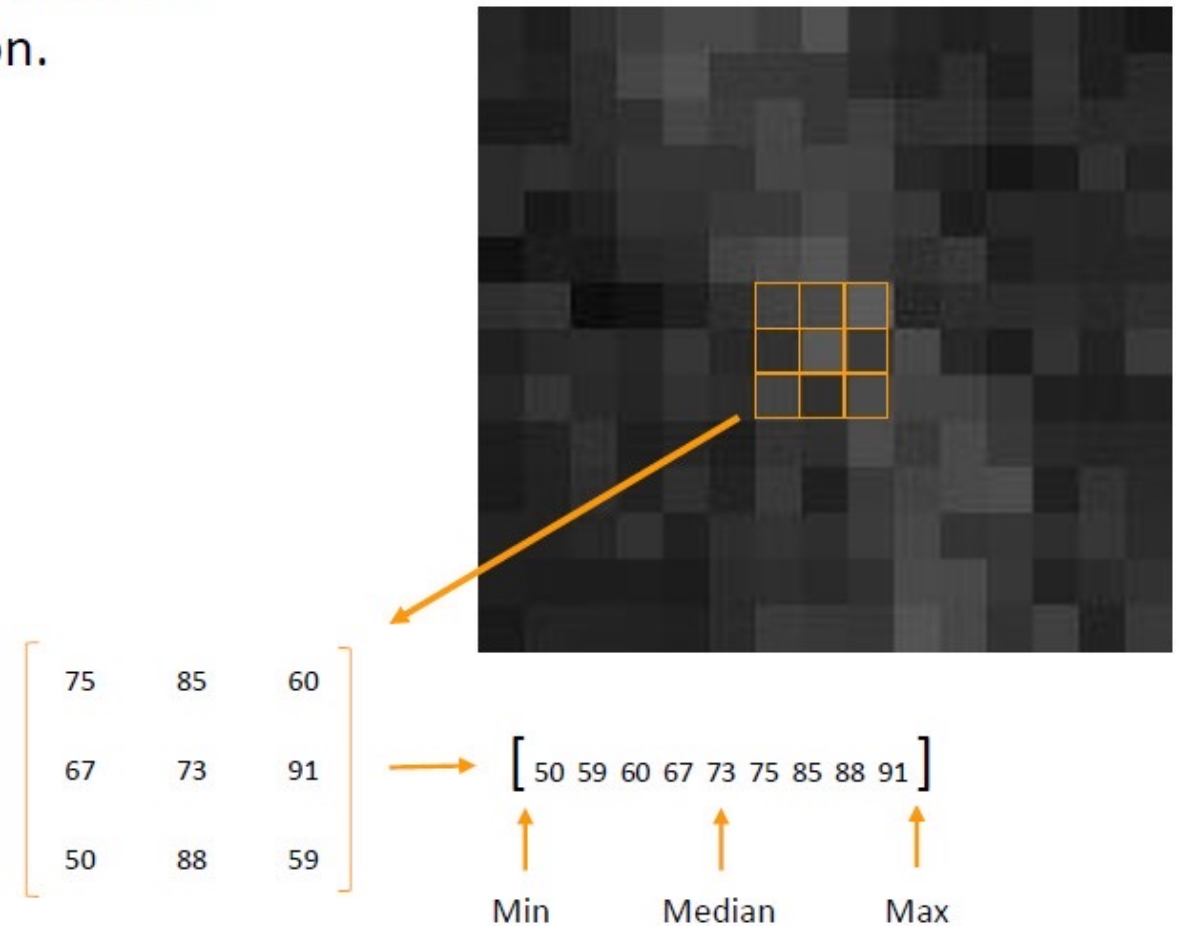
# DIFFERENCE OF GAUSSIAN (DoG)

- Example DoG images



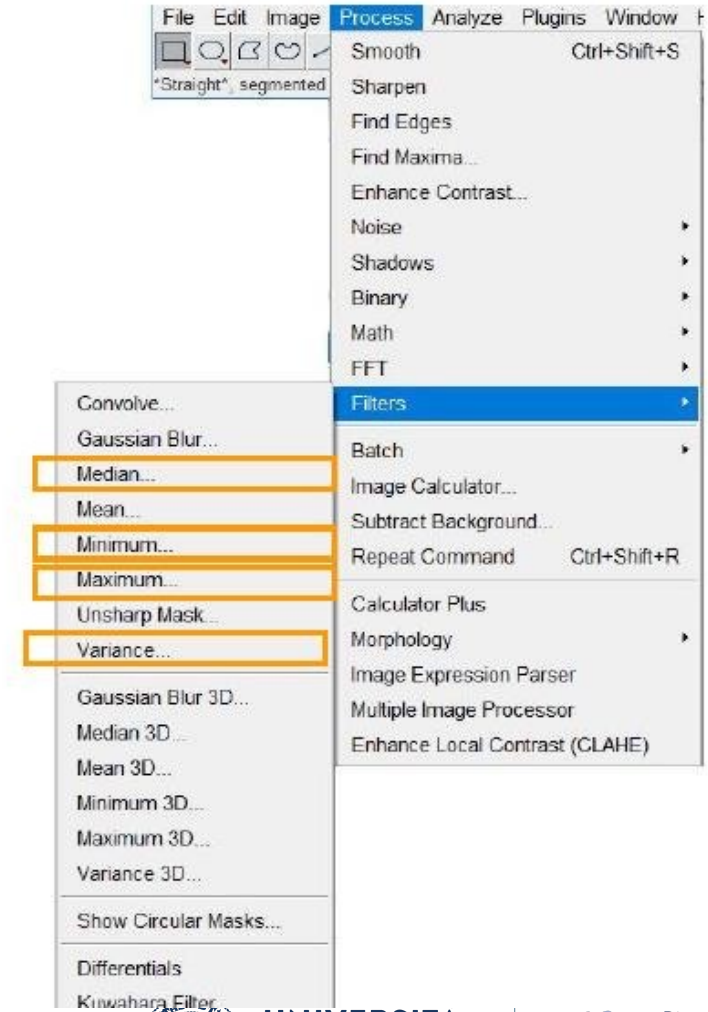
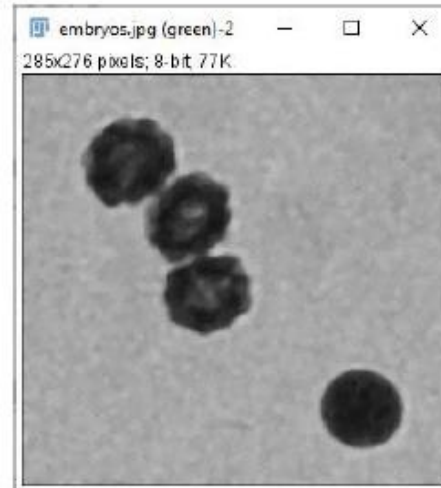
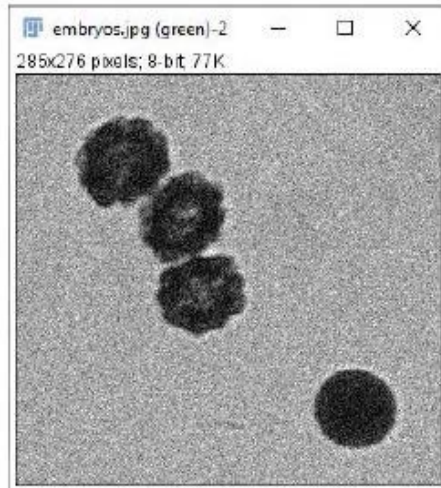
# NON-LINEAR FILTERS

- Non linear filters also replace pixel value inside as rolling window but in a non linear function.
- Examples: order statistics filters
  - Min
  - Median
  - Max
  - Variance
  - Standard deviation



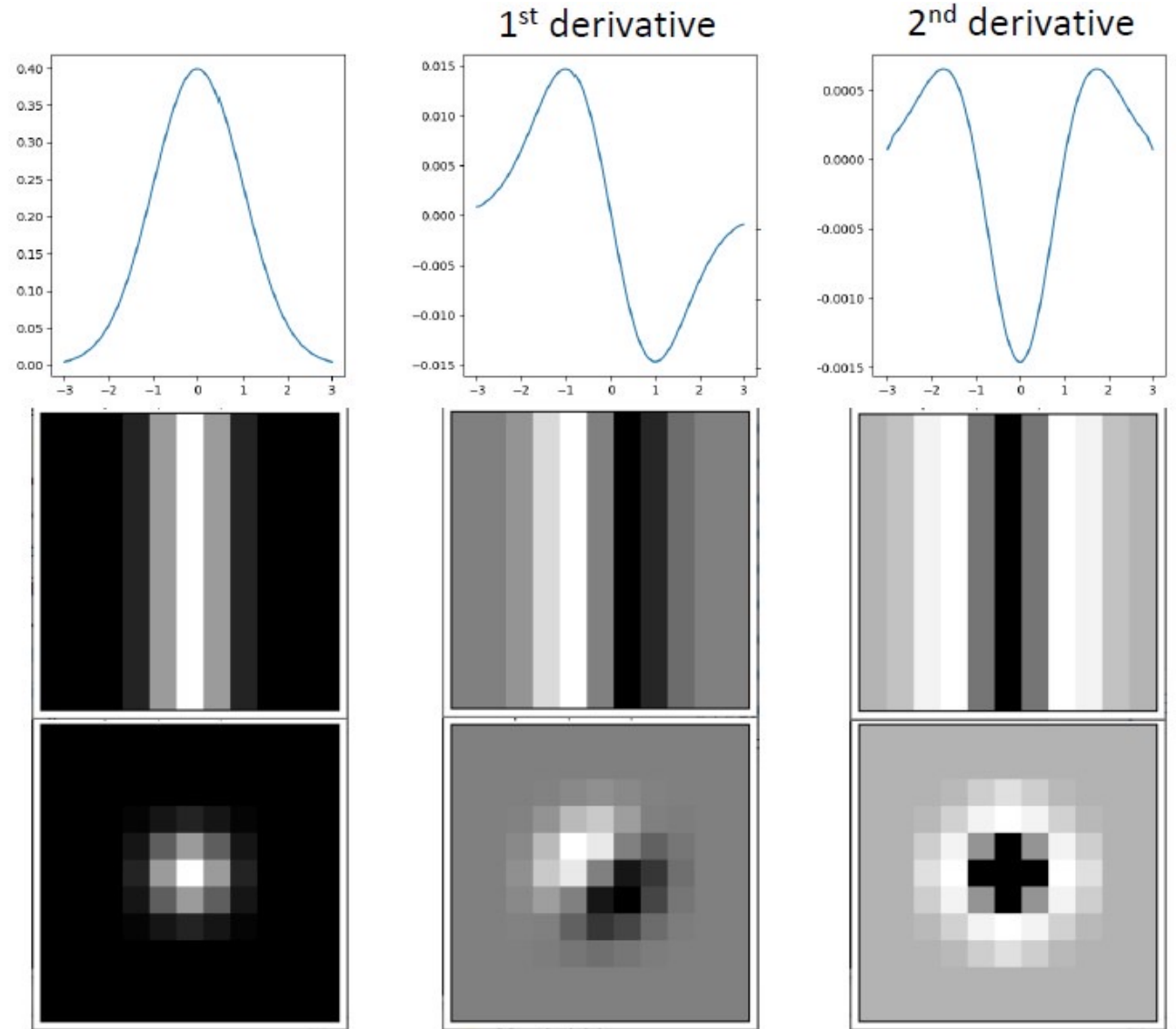
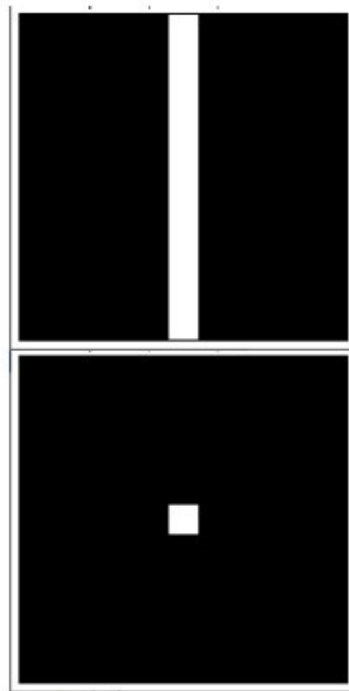
# NON-LINEAR FILTERS

- Non linear filters also replace pixel value inside as rolling window but in a non linear function.
- Examples:
  - Min
  - Median
  - Max
  - Variance



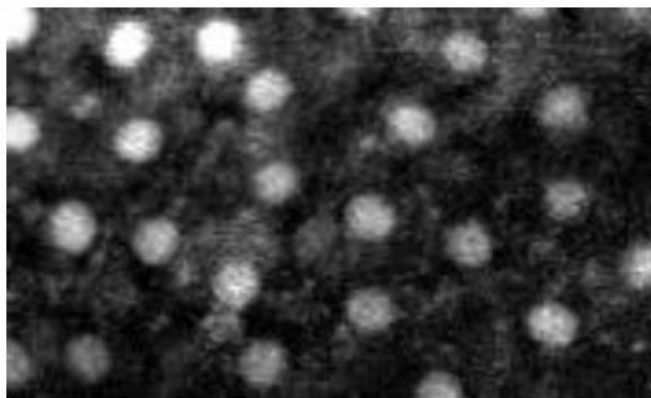
# LAPLACE FILTERS

- *Second derivative of a Gaussian blur filter*
- Used for edge-detection and edge enhancement
- Also known as the Mexican-hat-filter



# LAPLACIAN OF GAUSSIAN (LoG)

Laplace filter



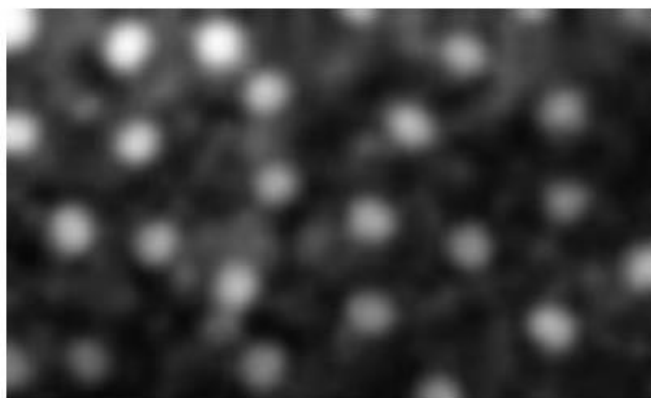
$$* \begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix} =$$



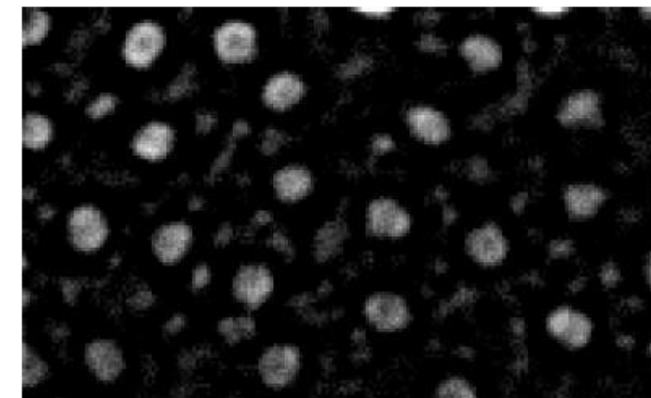
Laplace filtered image



Laplacian of Gaussian filter



$$* \begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix} =$$

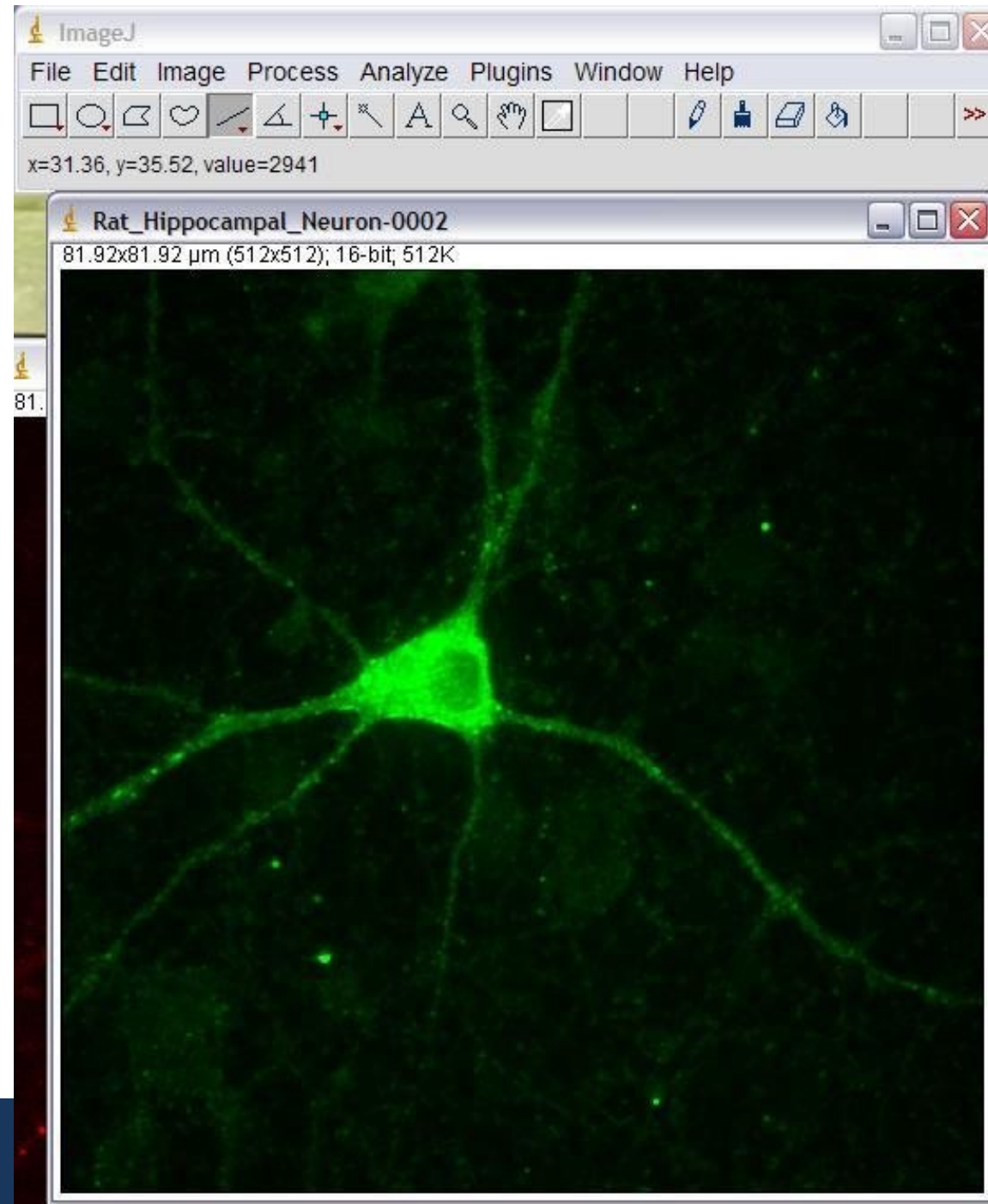


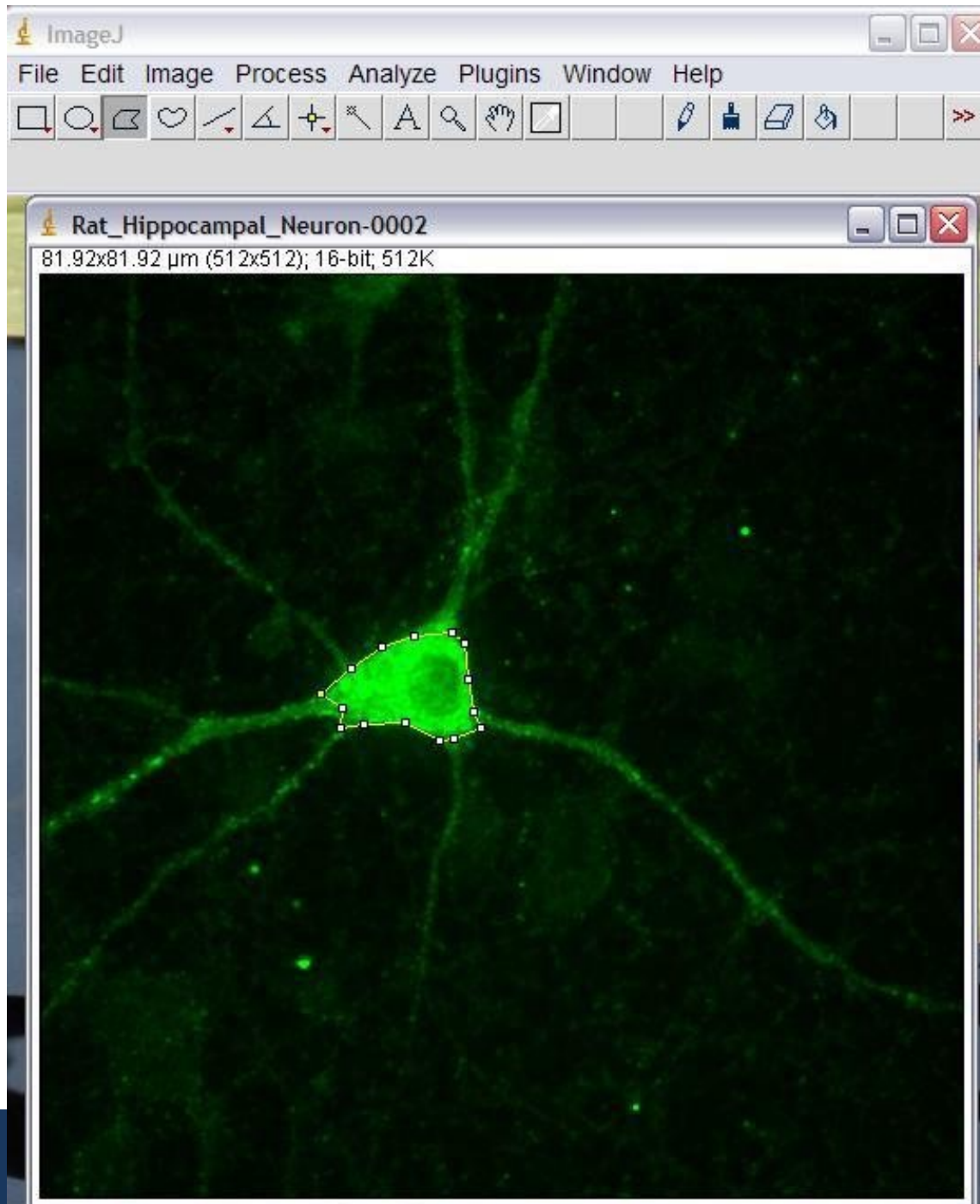
LoG image

# Basic Intensity Quantification with ImageJ

- Images mostly needed to be transformed into quantifiable data
- ImageJ is useful for getting information from images, including pixel intensity.
- There are a number of different ways to get intensity information from images using the base package of ImageJ (no plugins required).

Quantify Gray Levels Across  
an Entire Image or Single  
Object/Region

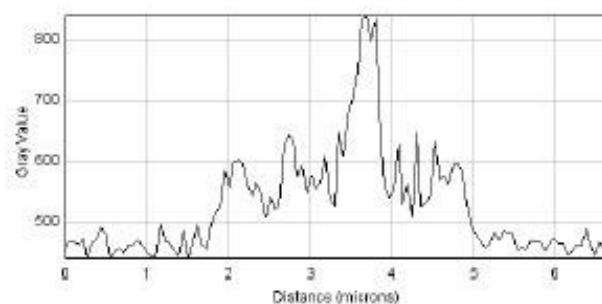
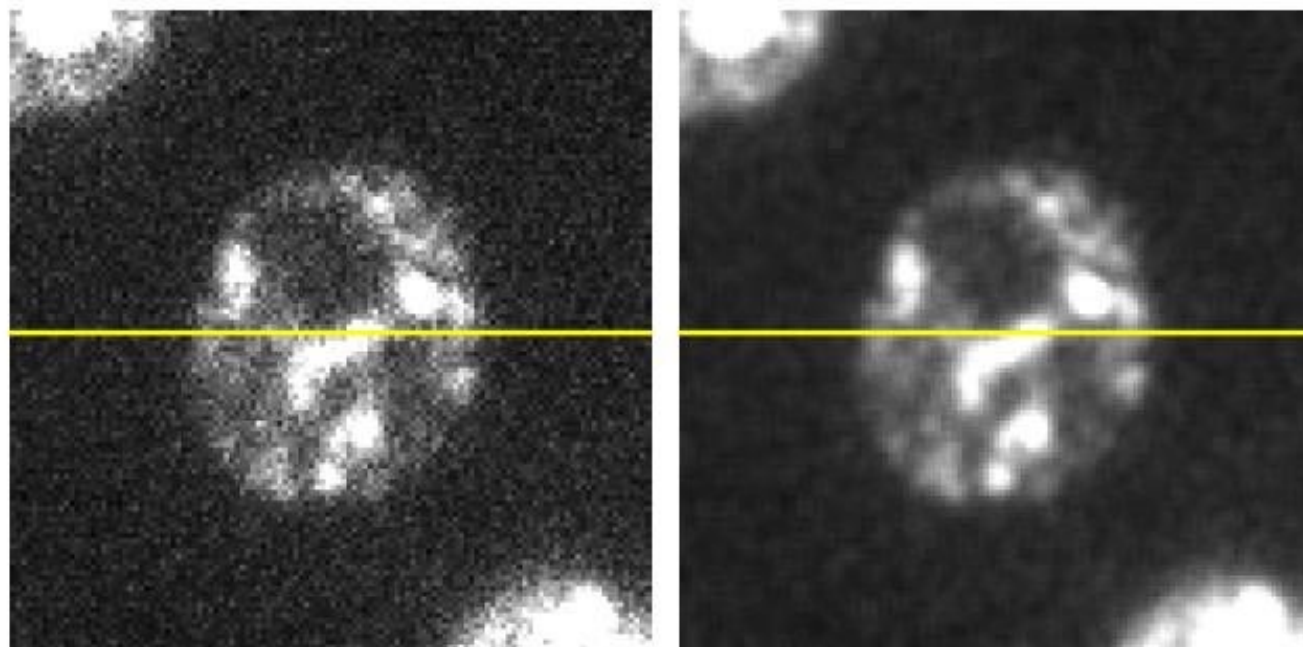




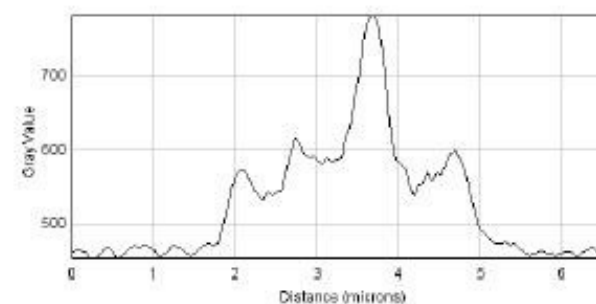
If you want to limit your measured area to just your object you draw a region of interest (ROI) around your object with one of the drawing tools (in the toolbar) and then

Analyze

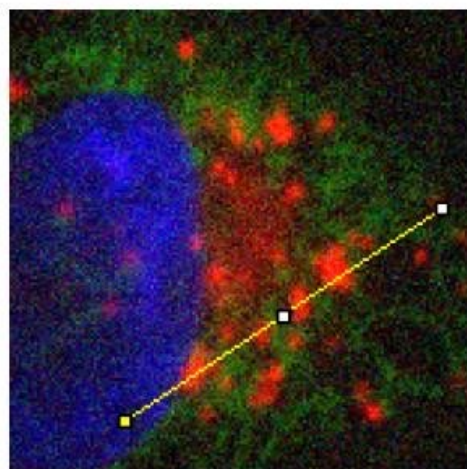
Measure



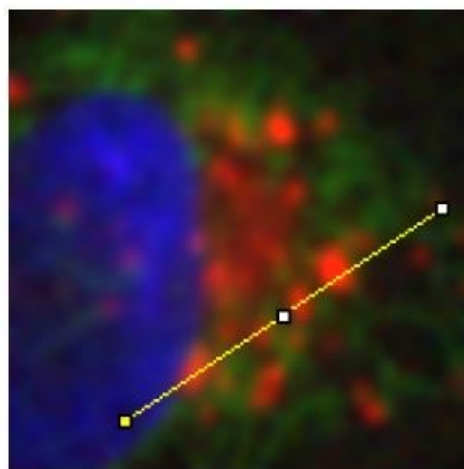
(a) Original image



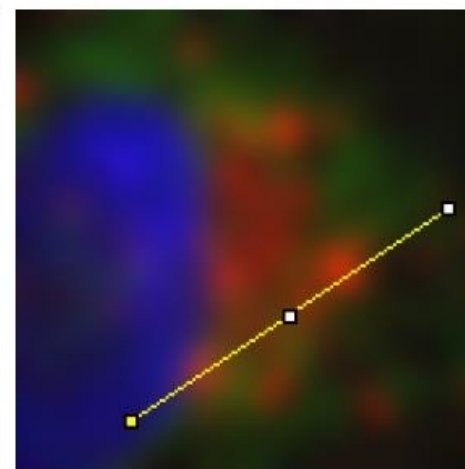
(b) Mean filtered



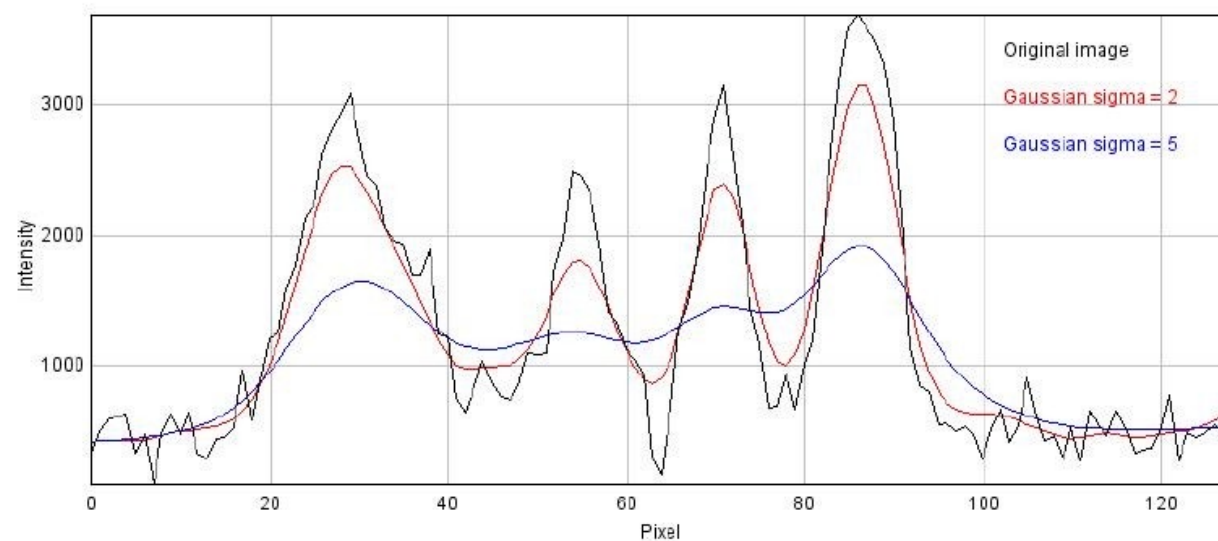
(a) Original image



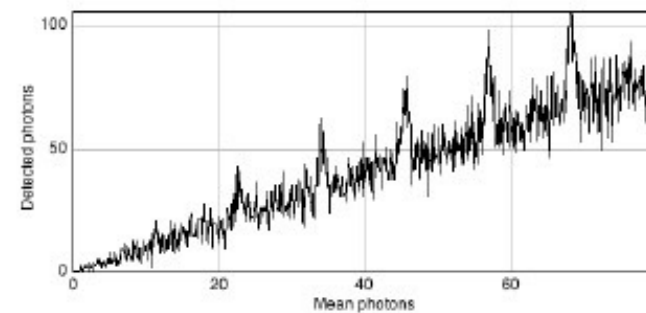
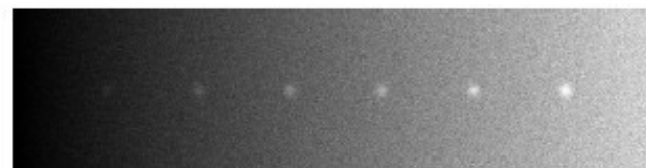
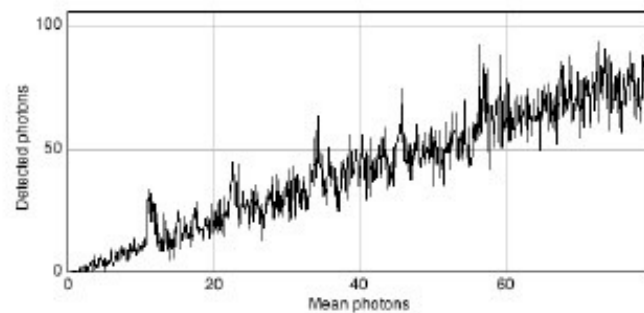
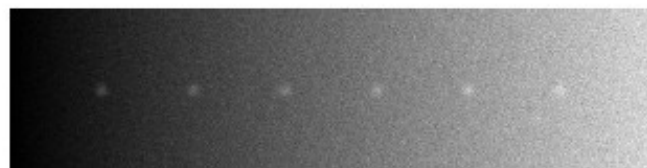
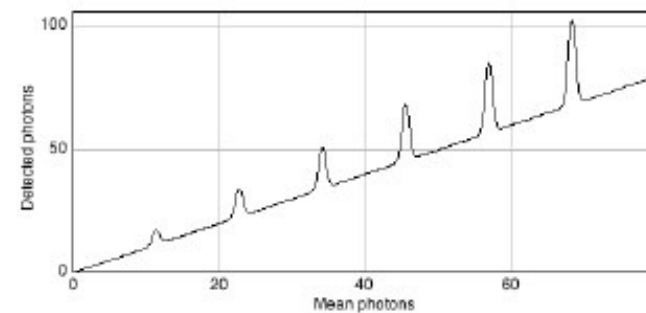
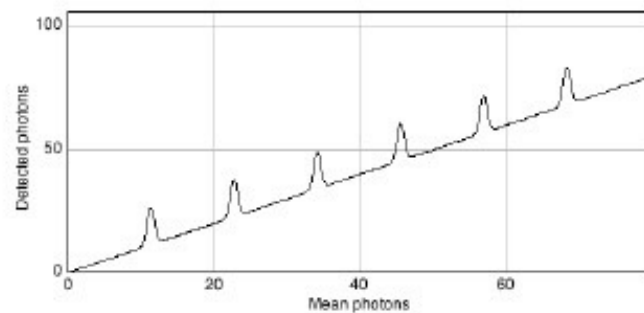
(b) Gaussian  $\sigma = 2$



(c) Gaussian  $\sigma = 5$

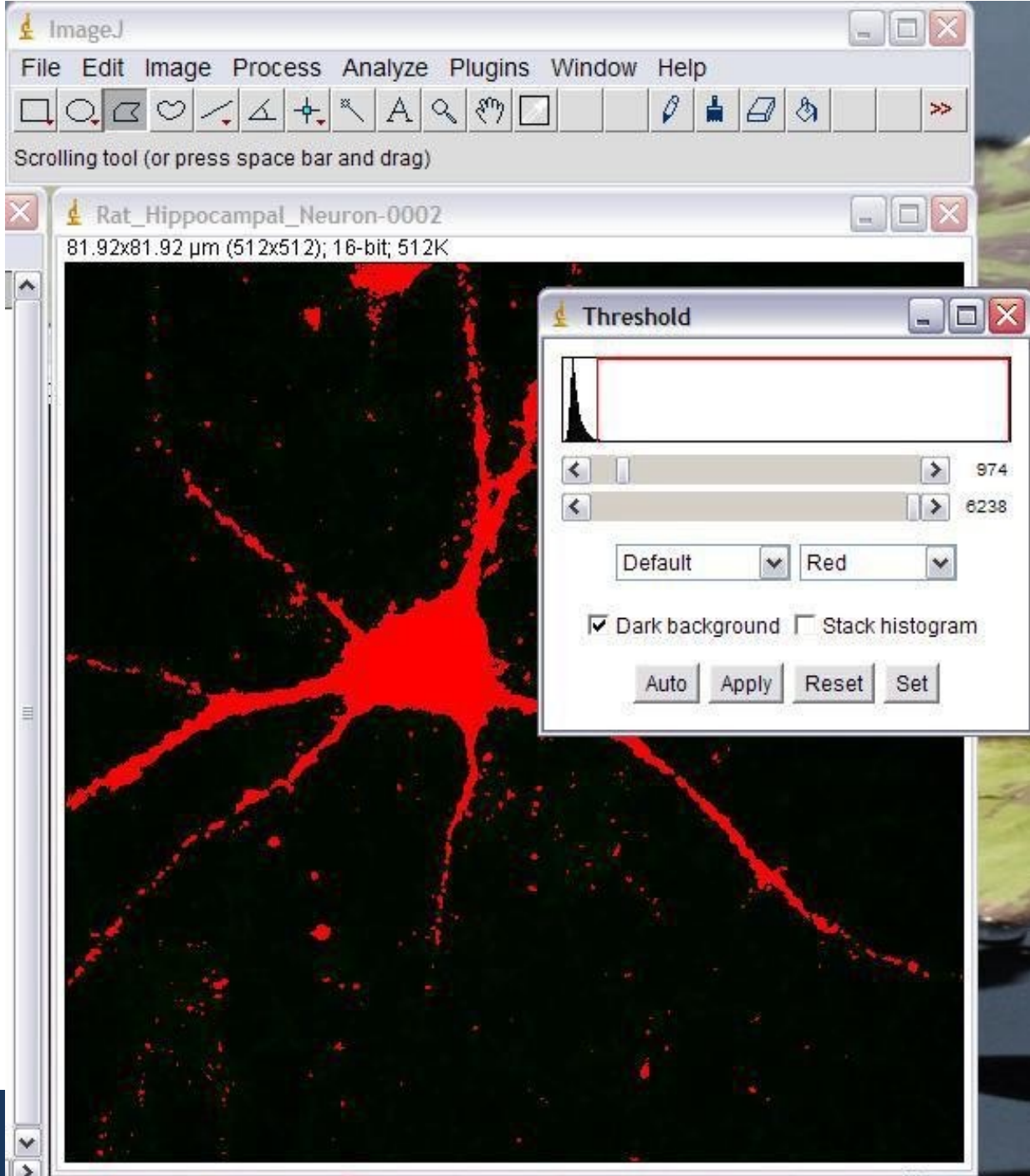


(d) Profile plots of the intensity in the red channel of the image



(a) Seeing spots with the same absolute brightness

(b) Seeing spots with the same relative brightness



Alternatively, you can go to  
Analyze > Set Measurements

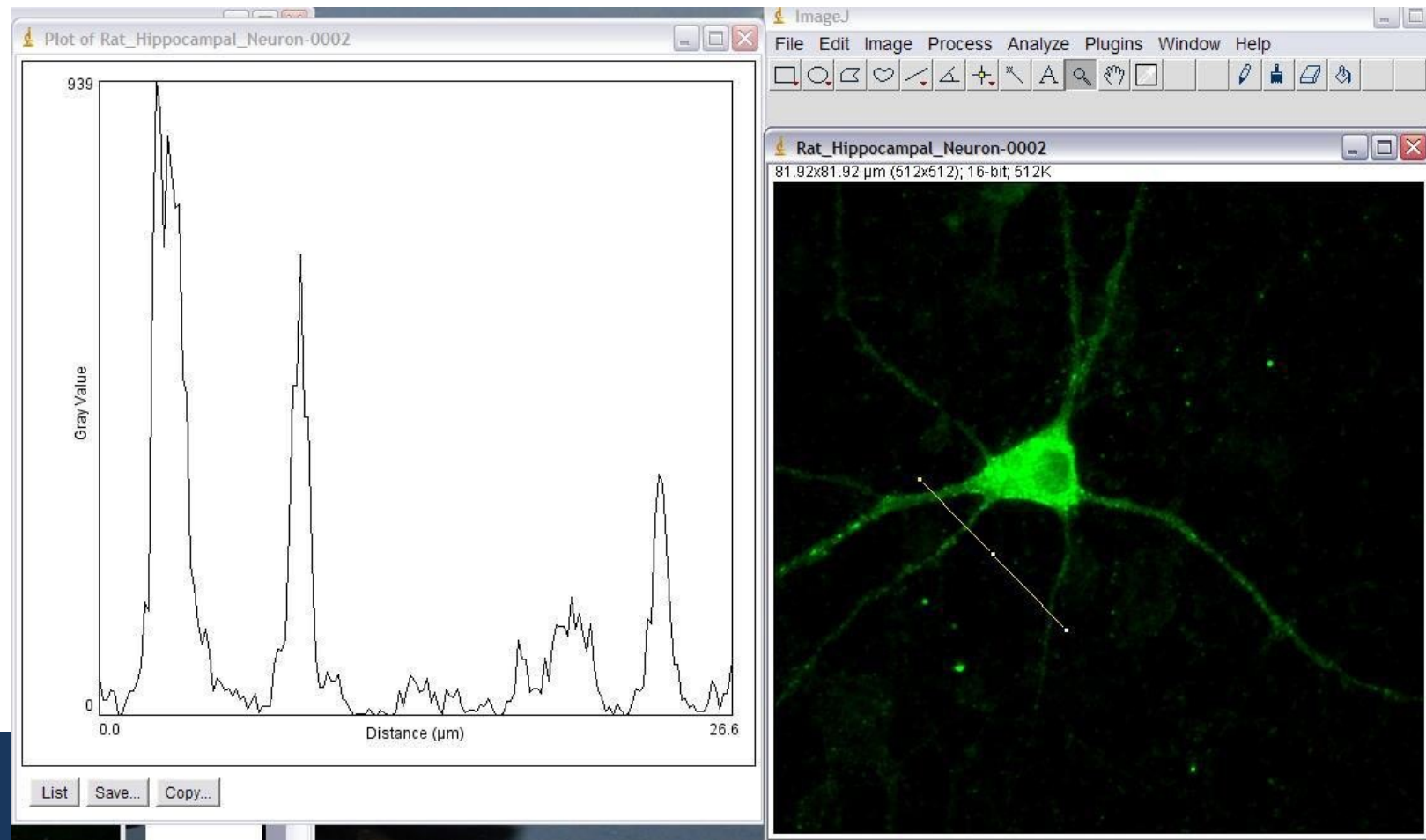
and check off the box next to  
“Limit to Threshold.”

Then use

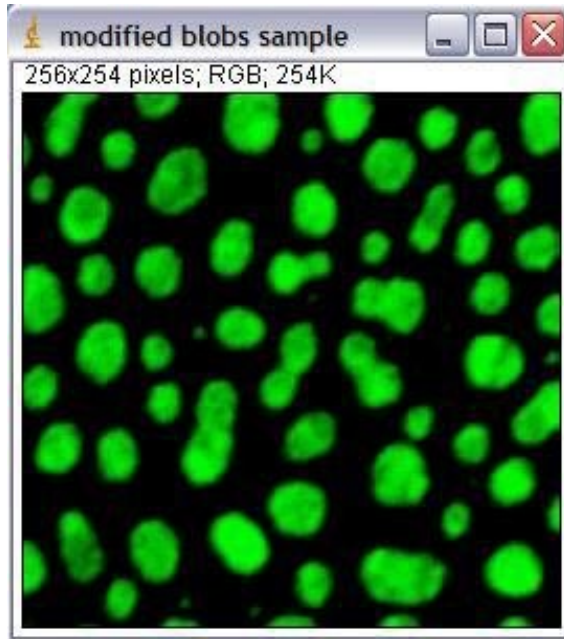
Image > Adjust > Threshold

to highlight the area you want to  
analyze, and then  
>Analyze

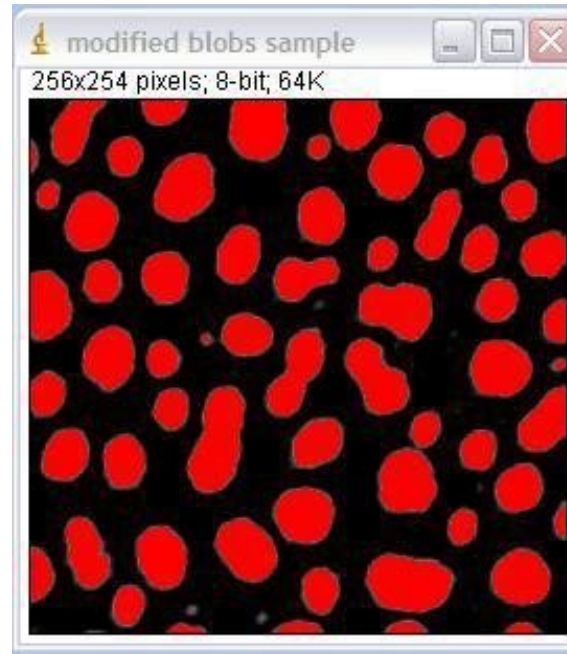
- You can use Analyze > Plot Profile to create a plot of intensity values across features in your image.
- In the example below, the plot gives the intensity values along the line drawn across three cell processes.



# To Quantify Gray Levels for Each Object in Images with Multiple Objects



convert to grayscale

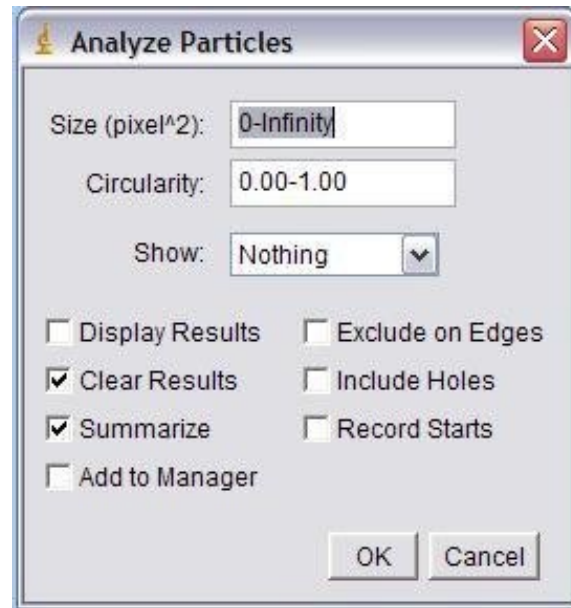


create a binary image,

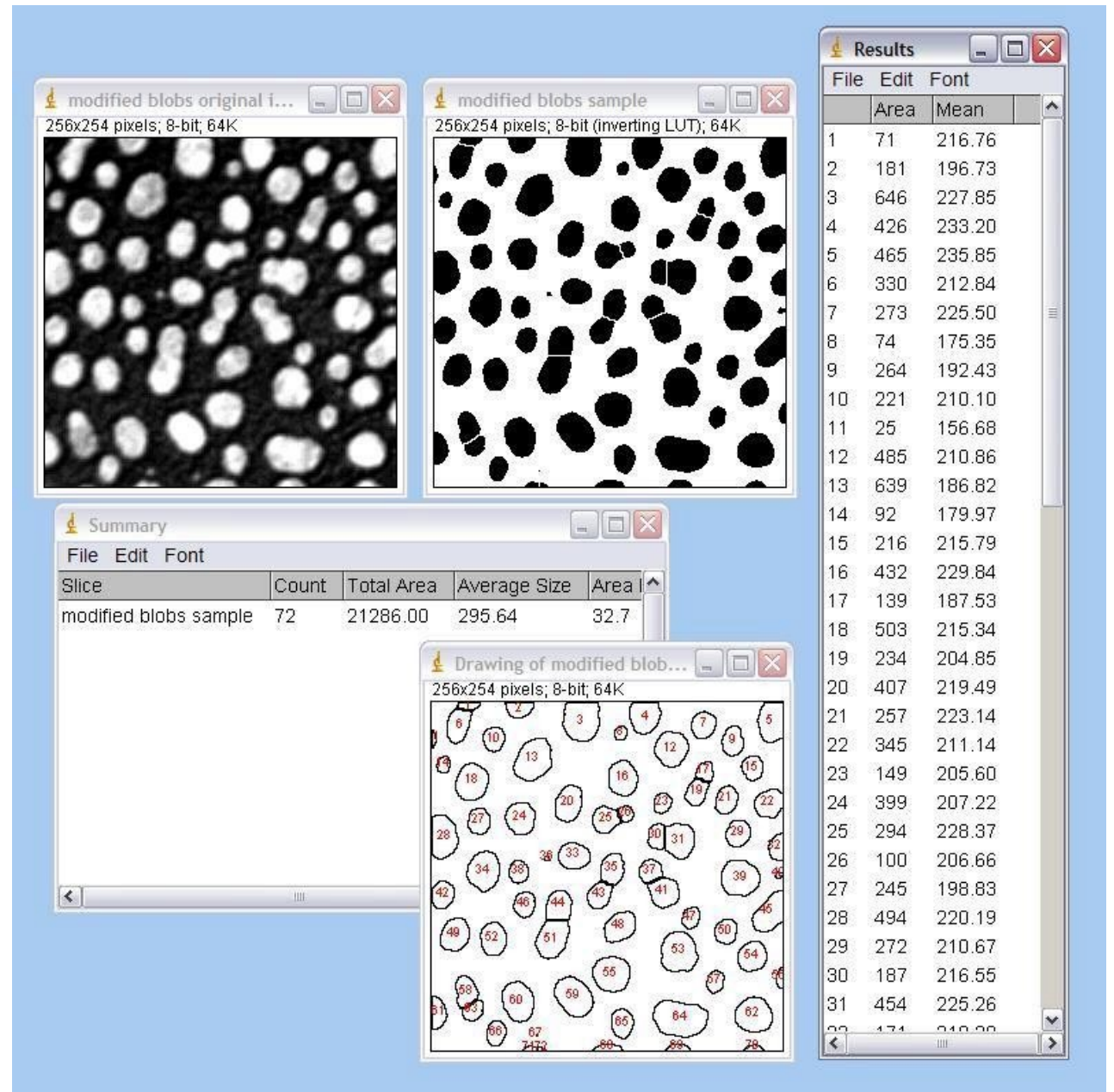


Process > Binary >  
Watershed

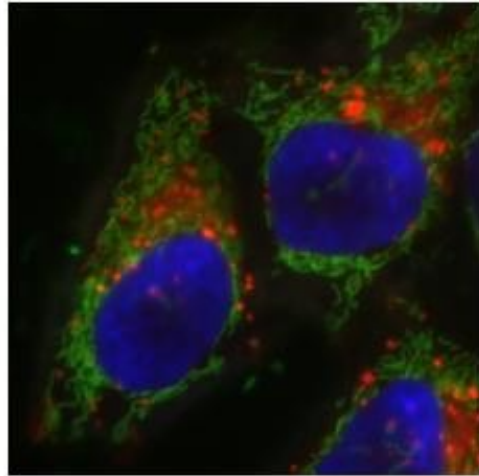
## Analyze > Analyze Particles



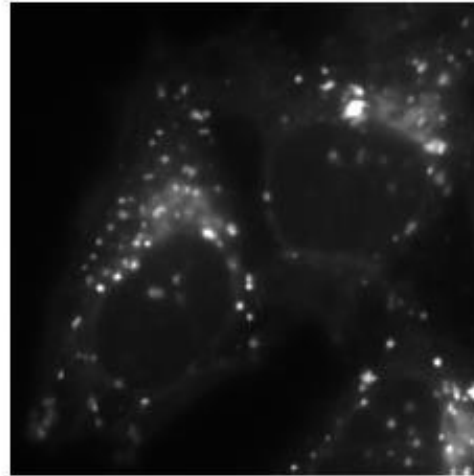
>



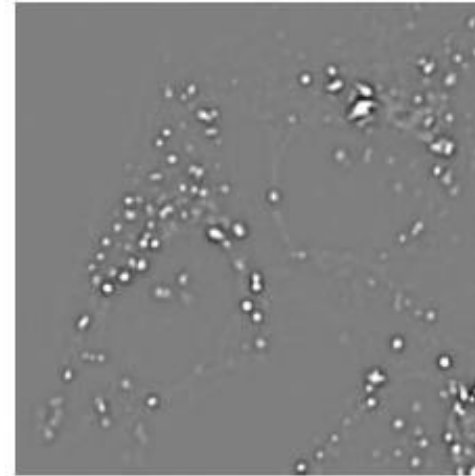
An simple image analysis workflow for detecting and measuring small spots, applied to the red channel of the sample image HeLa Cells.



(a) Original image



(b) Extract channel



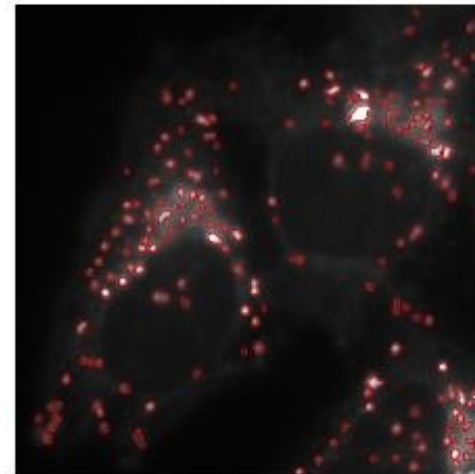
(c) Apply filters



(d) Apply threshold



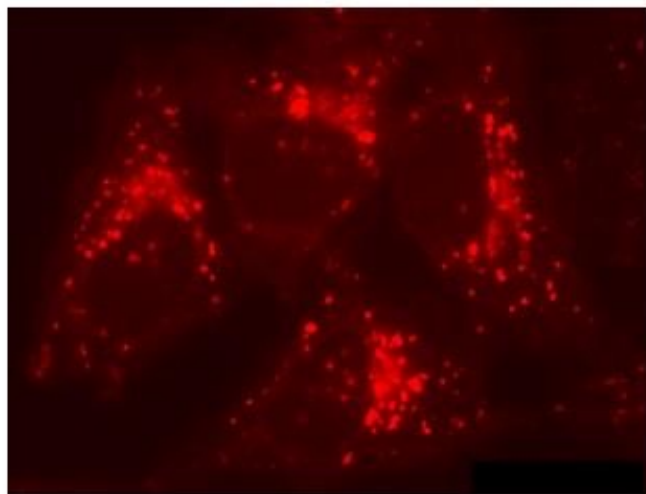
(e) Refine detection



(f) Relate (e) to (b)

| Results                |      |          |         |      |      |         |
|------------------------|------|----------|---------|------|------|---------|
| File Edit Font Results |      |          |         |      |      |         |
|                        | Area | Mean     | StdDev  | Min  | Max  | XM      |
| 159                    | 32   | 773.219  | 74.170  | 656  | 930  | 349.029 |
| 160                    | 37   | 1510.459 | 365.040 | 926  | 2380 | 375.628 |
| 161                    | 11   | 1395.279 | 53.660  | 1303 | 1471 | 382.494 |
| 162                    | 26   | 1588.077 | 77.288  | 1459 | 1710 | 395.341 |
| 163                    | 26   | 752.269  | 152.285 | 578  | 1058 | 276.604 |
| 164                    | 38   | 1654.500 | 307.428 | 1396 | 2535 | 402.042 |
| 165                    | 48   | 1122.333 | 224.585 | 836  | 1591 | 353.846 |
| 166                    | 55   | 2185.236 | 401.208 | 1483 | 3027 | 362.335 |
| 167                    | 33   | 1672.788 | 145.517 | 1380 | 1881 | 392.653 |
| 168                    | 26   | 739.038  | 113.867 | 578  | 994  | 331.600 |
| 169                    | 33   | 798.394  | 200.941 | 527  | 1212 | 263.840 |
| 170                    | 22   | 612.691  | 30.626  | 570  | 679  | 312.422 |
| 171                    | 34   | 1588.912 | 121.553 | 1331 | 1761 | 363.410 |
| 172                    | 44   | 1034.546 | 194.112 | 756  | 1474 | 346.720 |
| 173                    | 37   | 1205.106 | 469.627 | 629  | 2290 | 247.993 |

(g) Measure



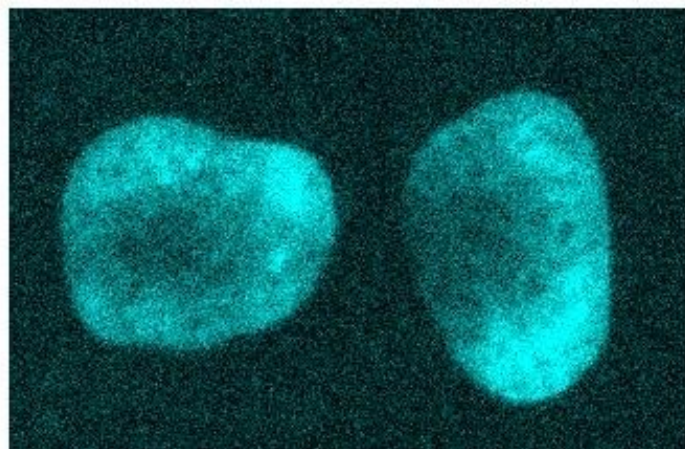
(a) Image



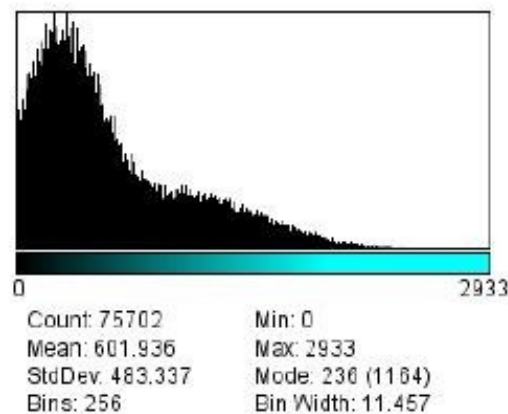
(b) Low threshold



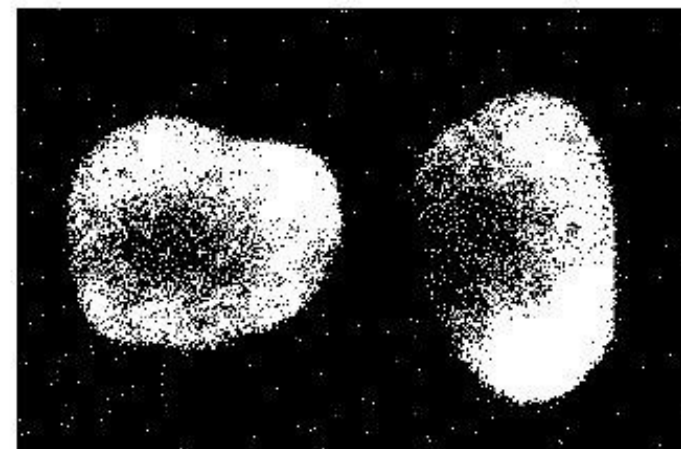
(c) High threshold



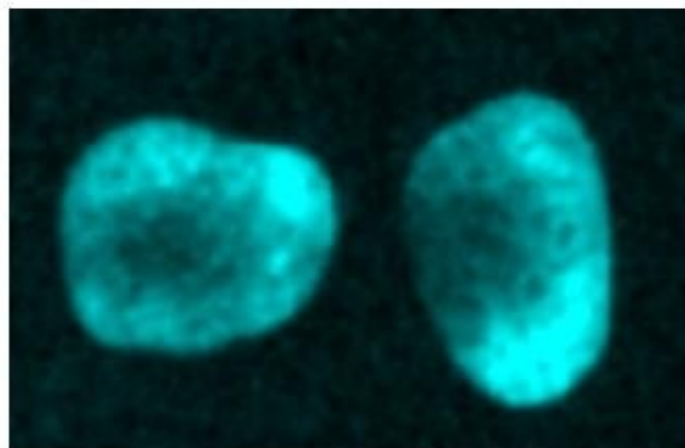
(a) Noisy image



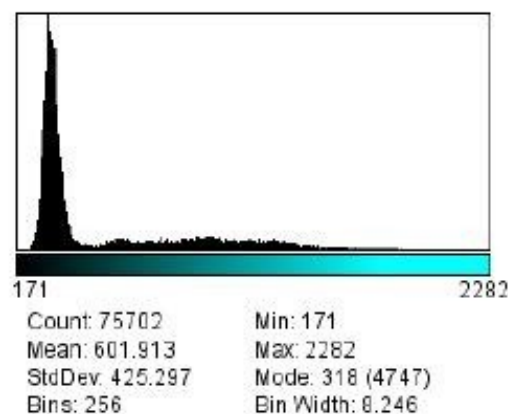
(b) Histogram of (a)



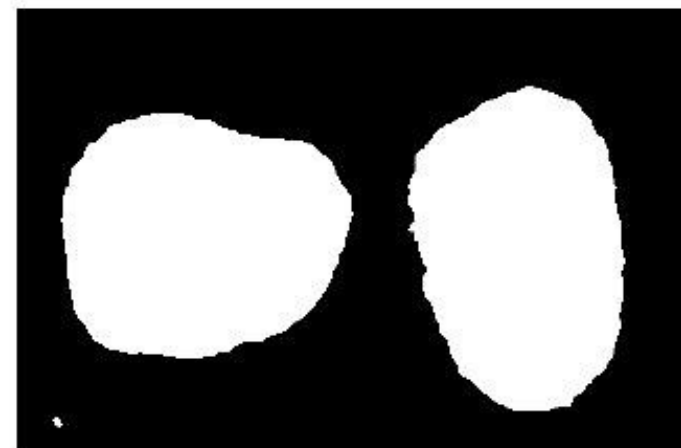
(c) Threshold applied to (a)



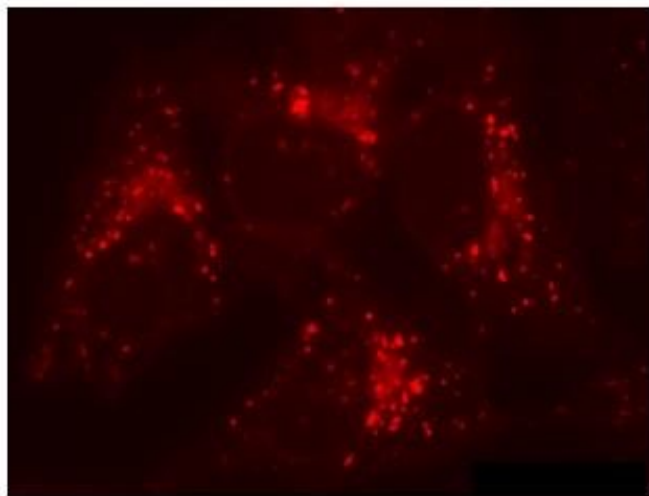
(d) Gaussian filtered image



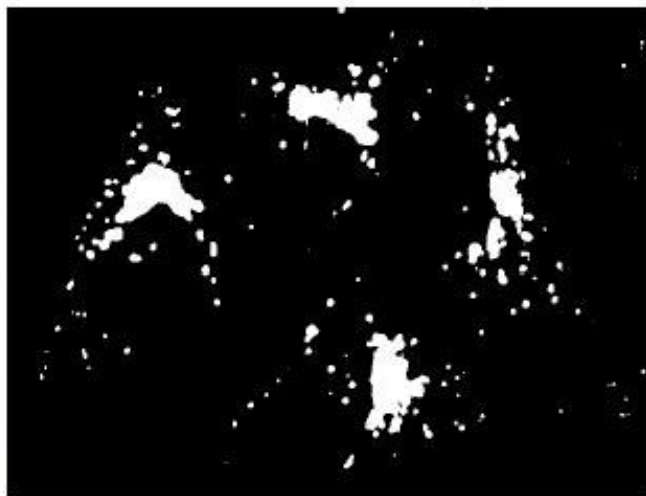
(e) Histogram of (d)



(f) Threshold applied to (d)



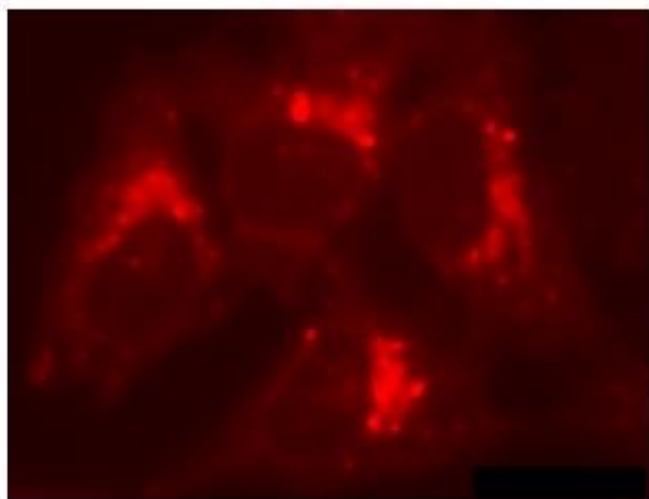
(a) Original image



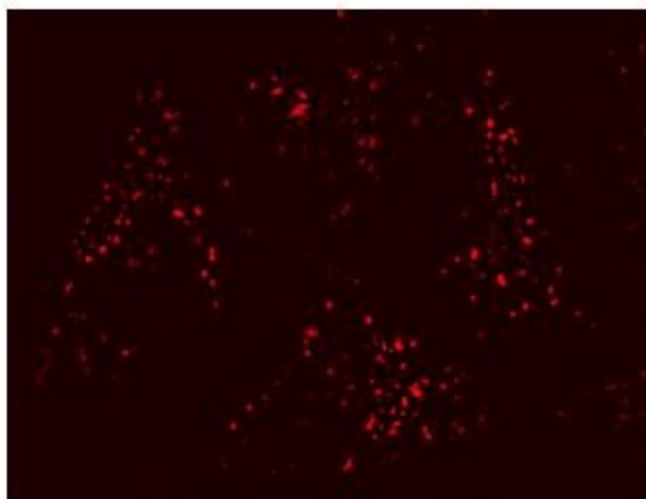
(b) Otsu's threshold



(c) Triangle threshold



(d) Median filtered image



(e) Result of (a)-(d)



(f) Triangle threshold of (e)

# ImageJ How to Measure Mean Fluorescence Intensity Over Timelapse Image Stack

<https://youtu.be/GHvndpGQKe4>

## Short introduction to histogram processing

[https://youtu.be/nIRhHb04u\\_k](https://youtu.be/nIRhHb04u_k)