

Image filtering

Robert Haase

With material from

Mauricio Rocha Martins, Norden lab, MPI CBG

Dominic Waithe, Oxford University

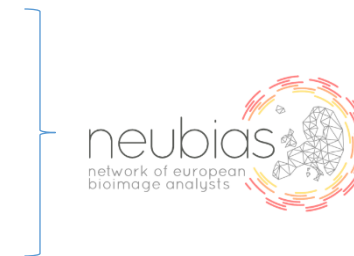
Nuno P Martins, IGC Lisbon

Paulo Aguiar, INEB, Porto

Sebastian Tosi, IRB Barcelona

Benoit Lombardot, Scientific Computing Facility, MPI CBG

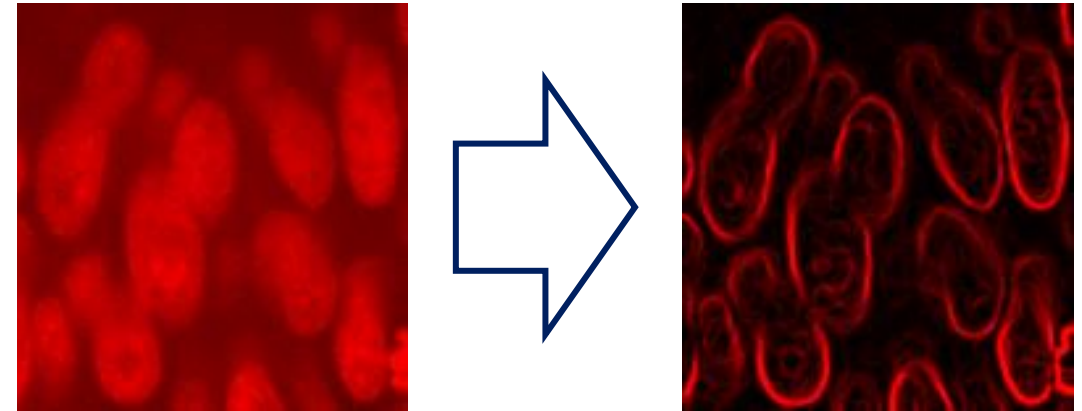
Alex Bird, Dan White, MPI CBG



April 2021

Today

- Filters
- Image math
- Image filtering with Fiji
- Image filtering with Python
- Exercises

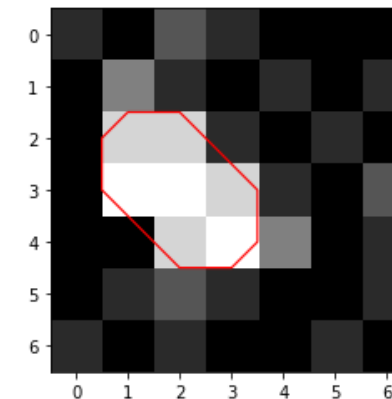


```
import matplotlib.pyplot as plt

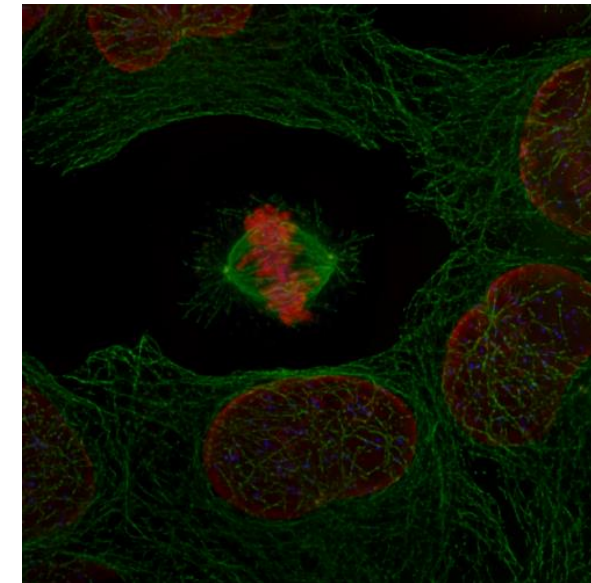
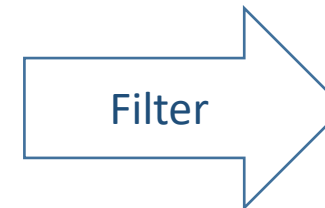
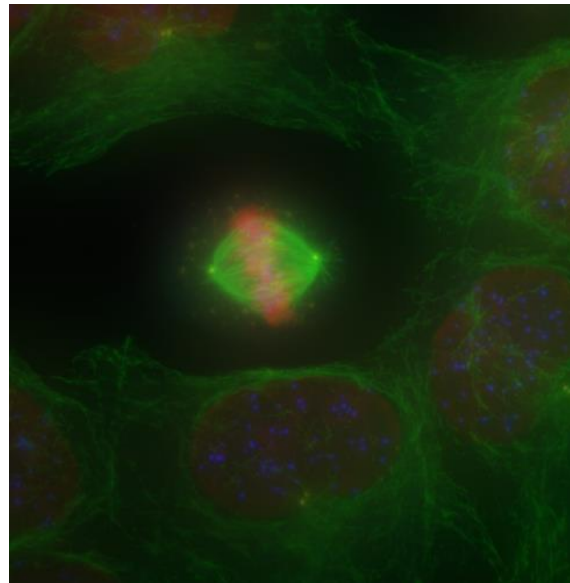
# create a new plot
fig, axes = plt.subplots(1,1)

# add two images
axes.imshow(image, cmap=plt.cm.gray)
axes.contour(binary_image, [0.5], linewidths=1.2, colors='r')

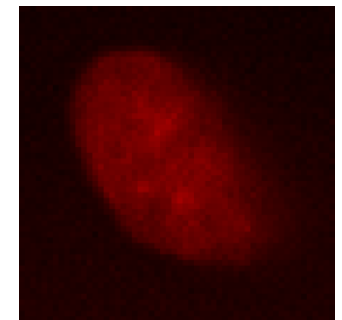
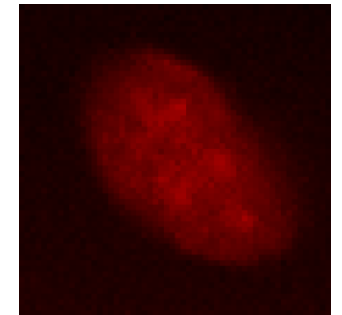
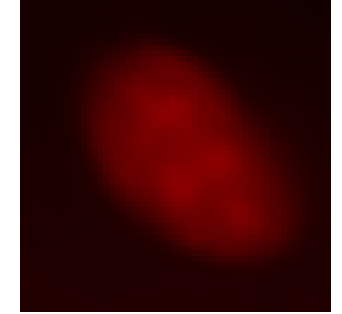
<matplotlib.contour.QuadContourSet at 0x206dc159310>
```



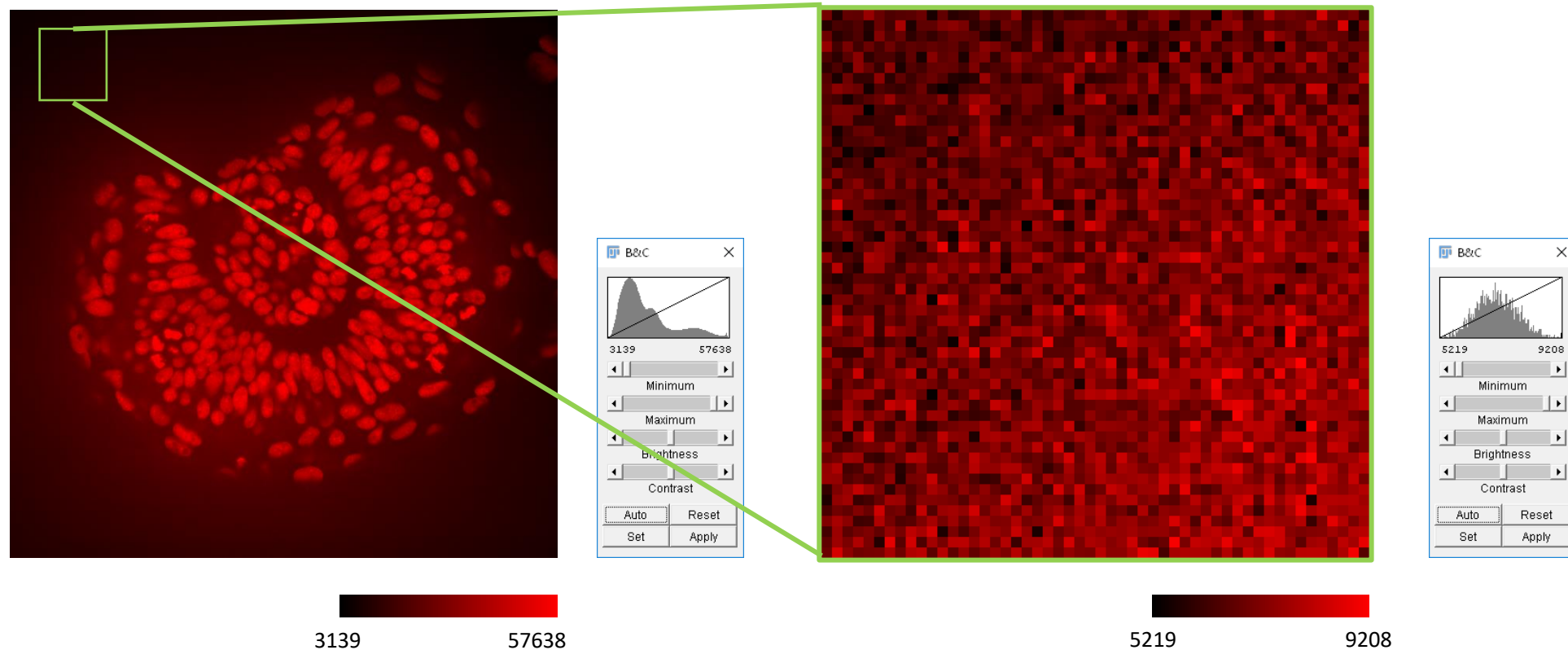
- An image processing filter is an operation on an image.
- It takes an image and produces a new image out of it.
- Filters change pixel values.
- There is no “best” filter. Which filter fits your needs, depends on the context.
- Filters do not do magic. They can not make things visible which are not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



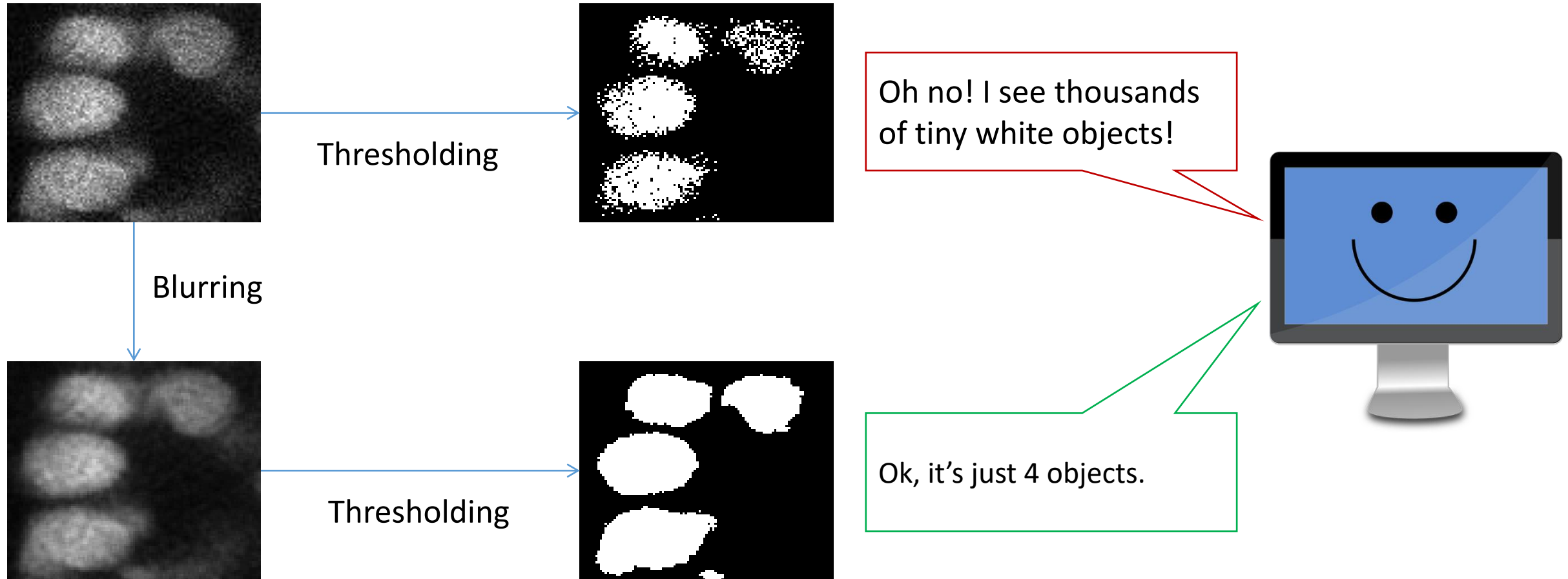
- Noise is a general term for unwanted modifications that a signal may suffer during capture, storage, transmission, processing, or conversion.
- In microscopy, image quality suffers from
 - shot noise: Statistical variation of the photons arriving at the camera
 - dark noise: Statistical variation of 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



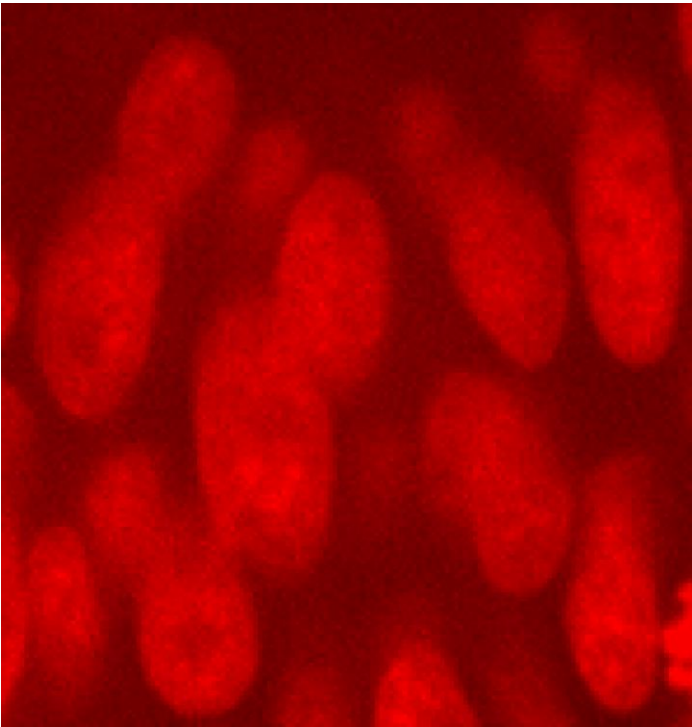
- When dealing with noise in microscopic image processing, we mostly mean the noise visible in the images.



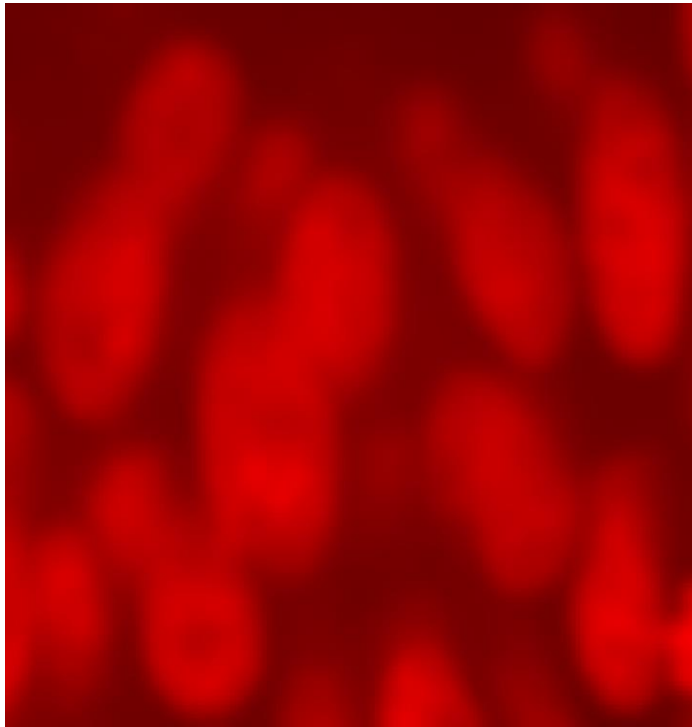
- We need to remove the noise to help the computer *interpreting* the image



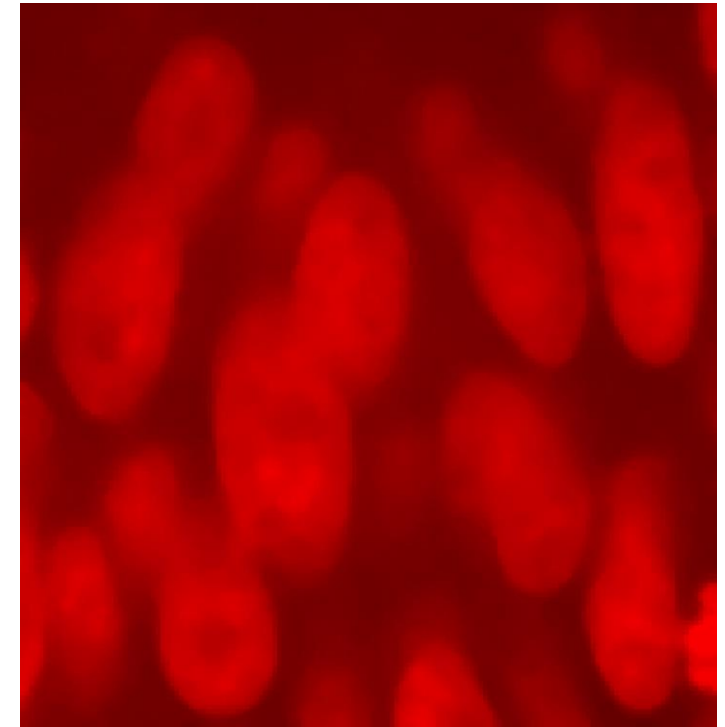
- Noise removal using *filters*



Original image

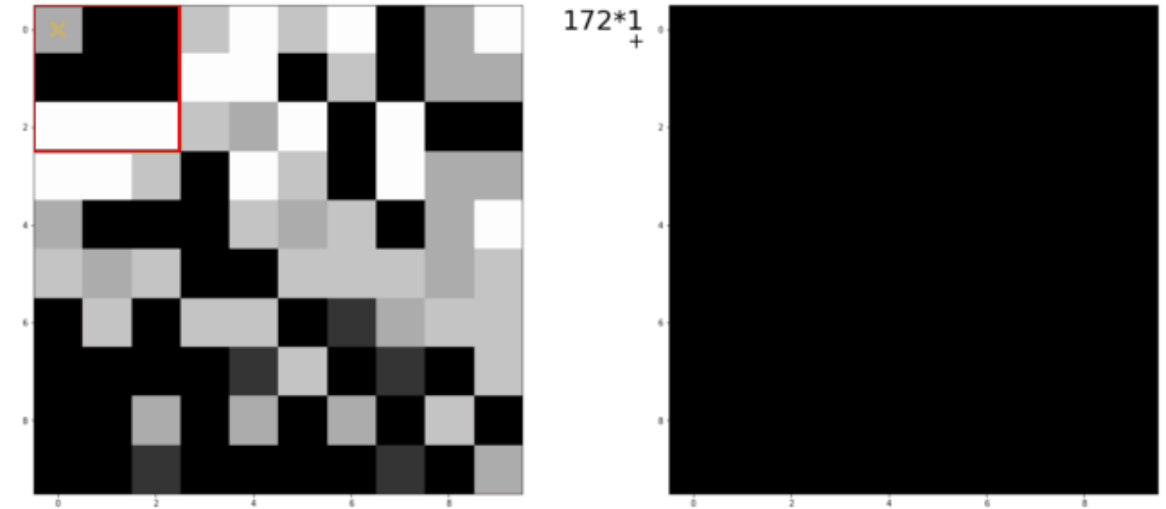


Gaussian blur filtered



Median filtered

- 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



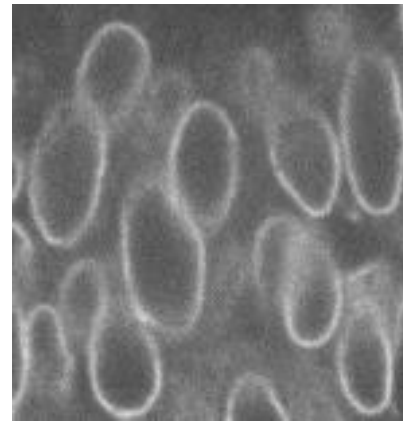
Mean filter, 3x3 kernel

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

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

- Examples

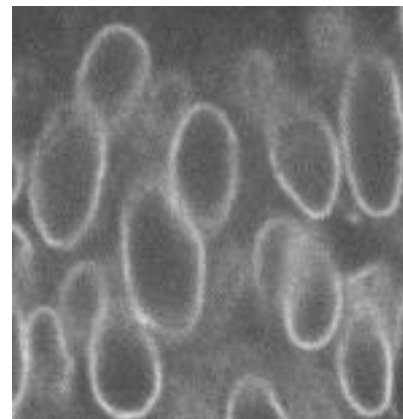
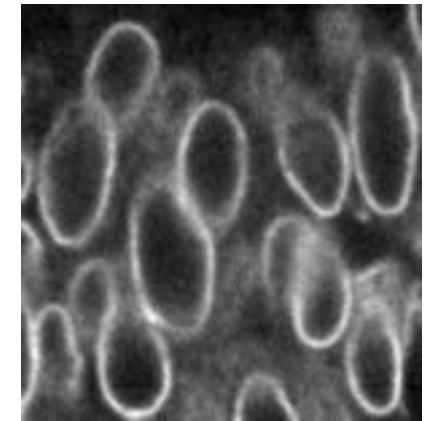
- Mean
- Gaussian blur
- Sobel-operator
- Laplace-filter



*

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

=



*

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

=



- 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

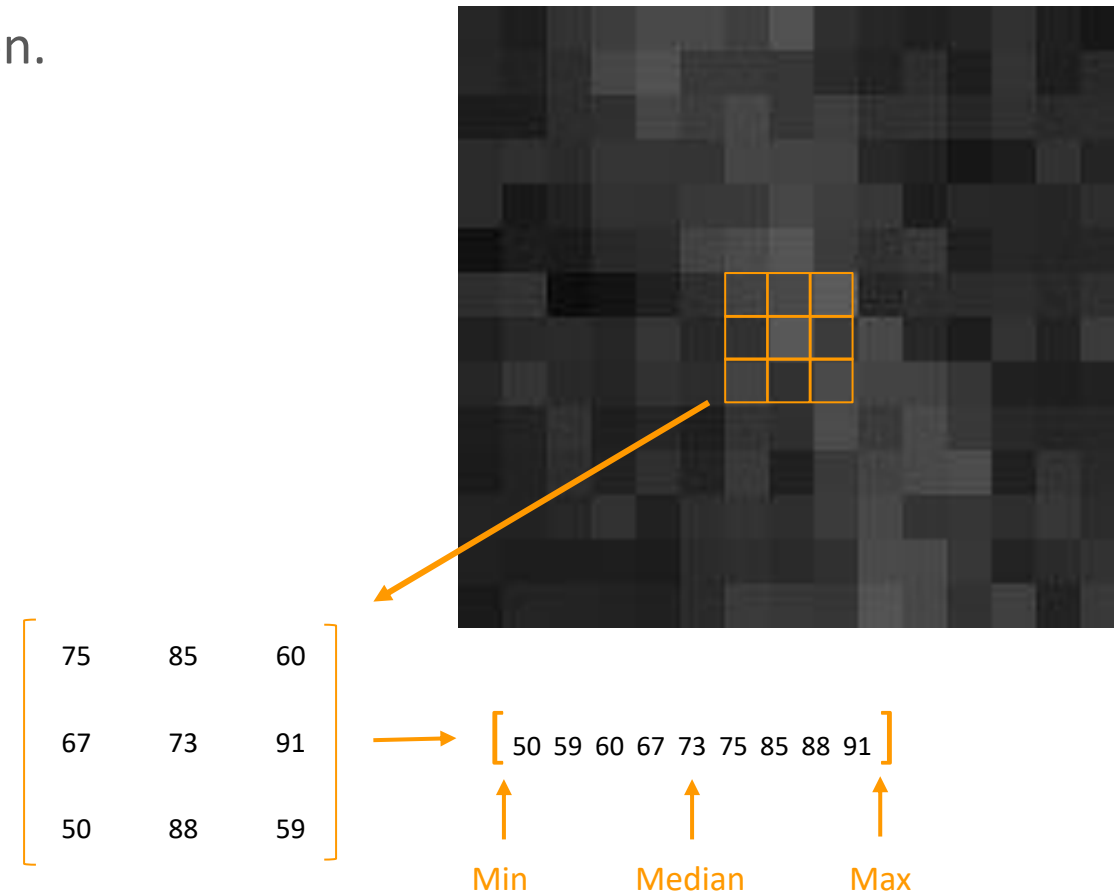


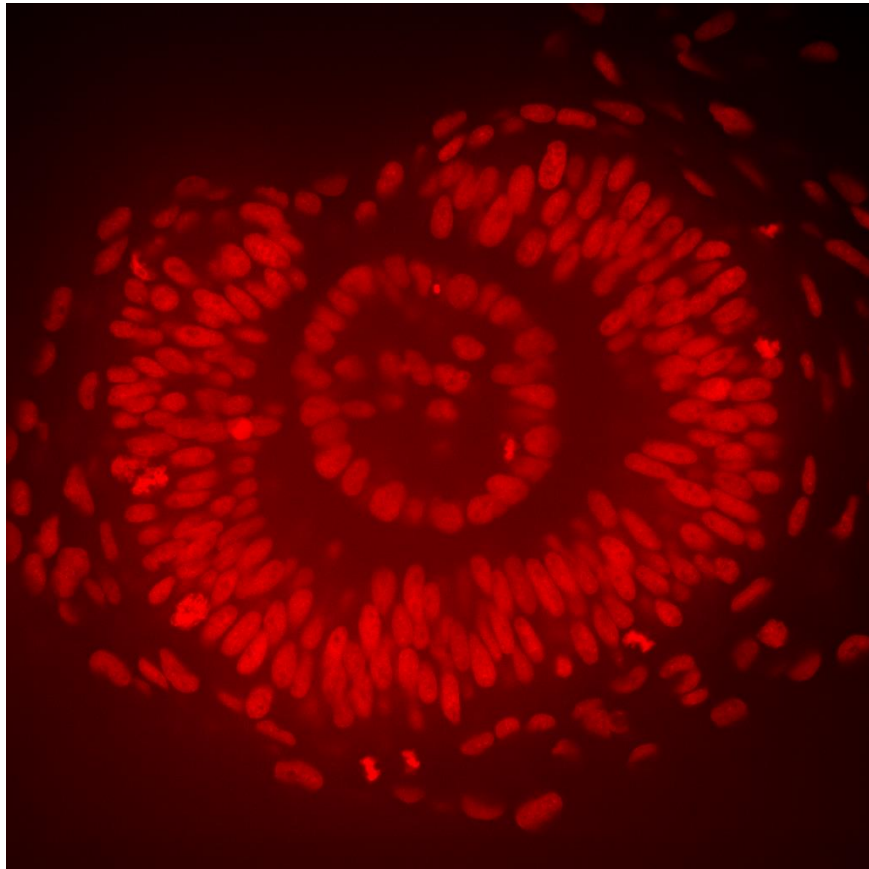
Image math

Robert Haase

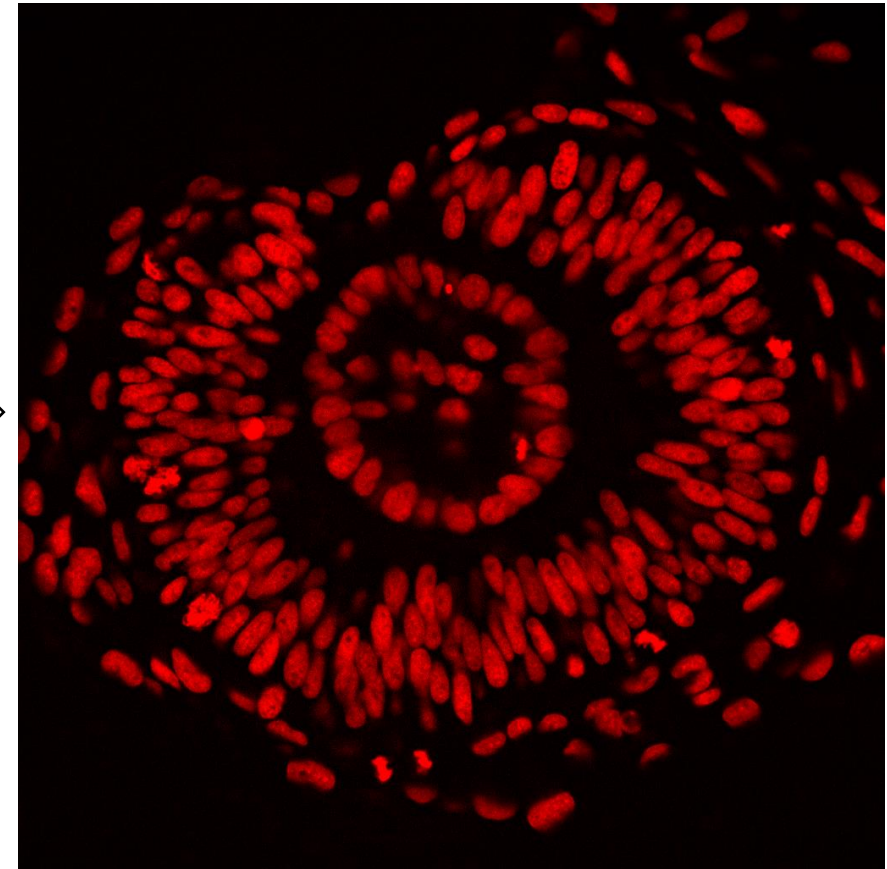
With material from
Mauricio Rocha Martins, Norden lab, MPI CBG

April 2021

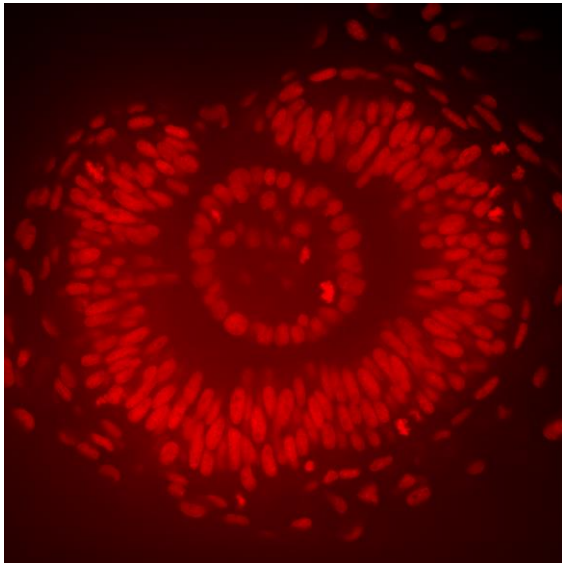
- Differentiating objects is easier if their background intensity is equal.



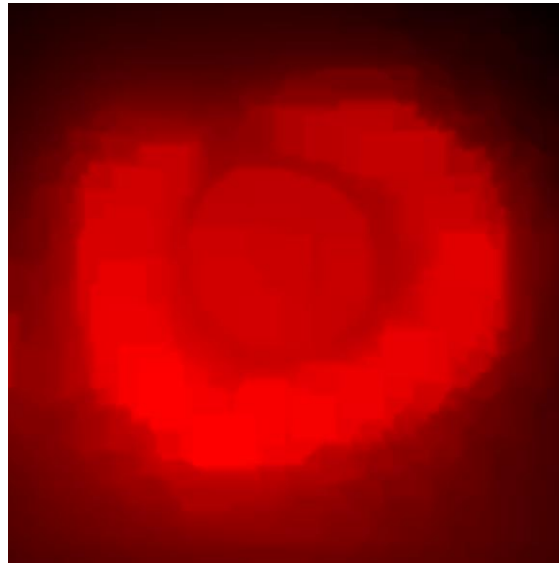
Subtract
background



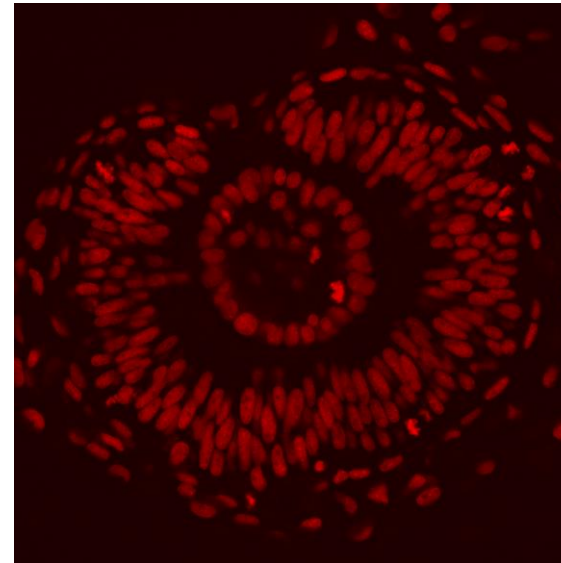
- Depending on the effect we want to correct for, it might make sense to divide an image by its background.



Original image



Extracted
background



Subtracted
background

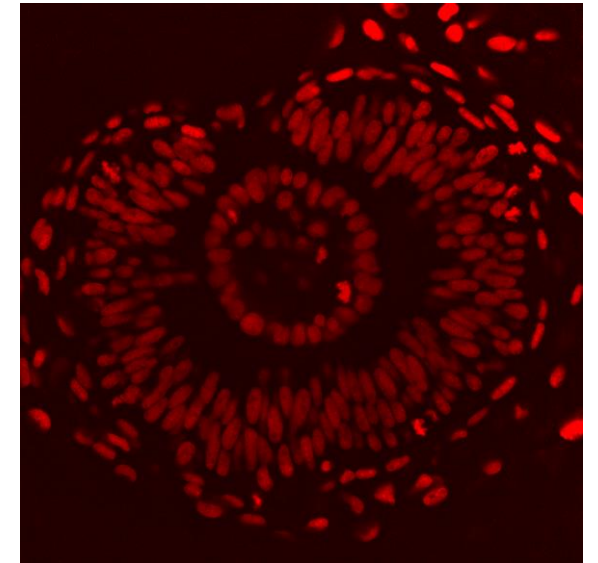
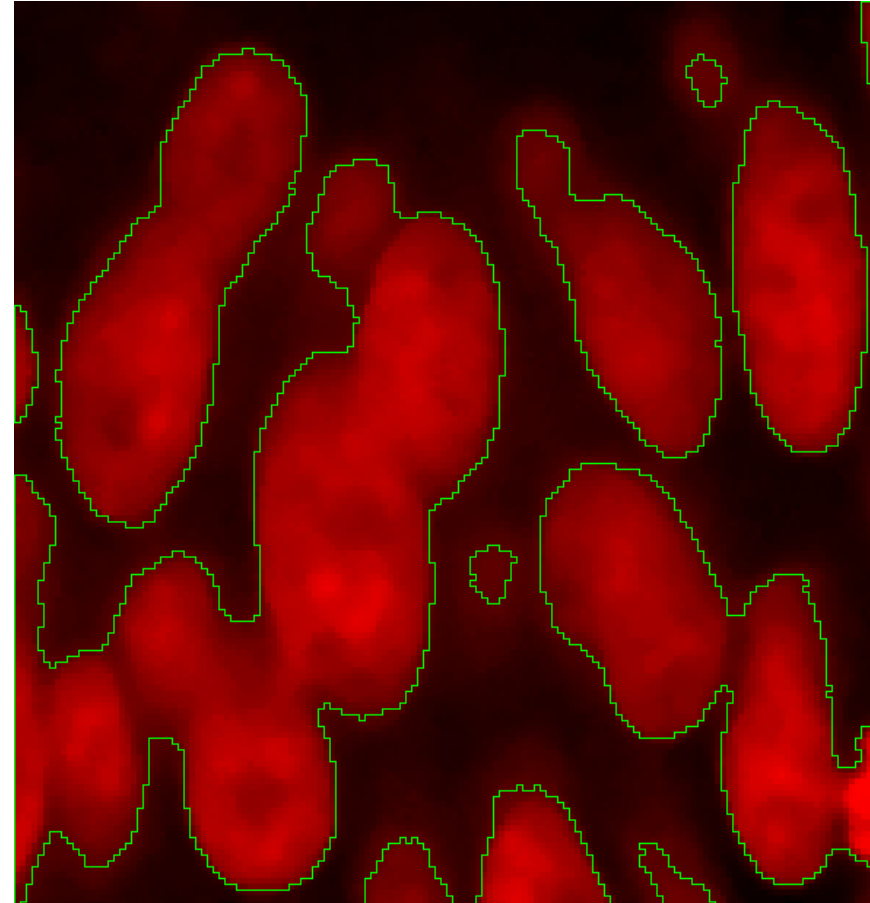
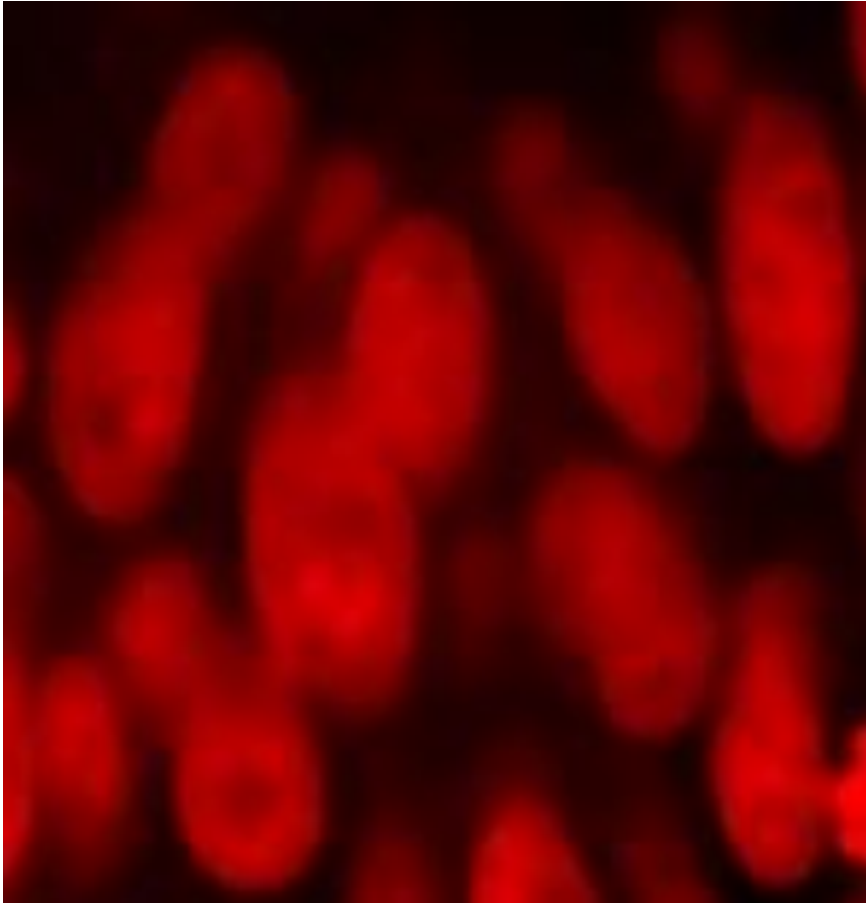
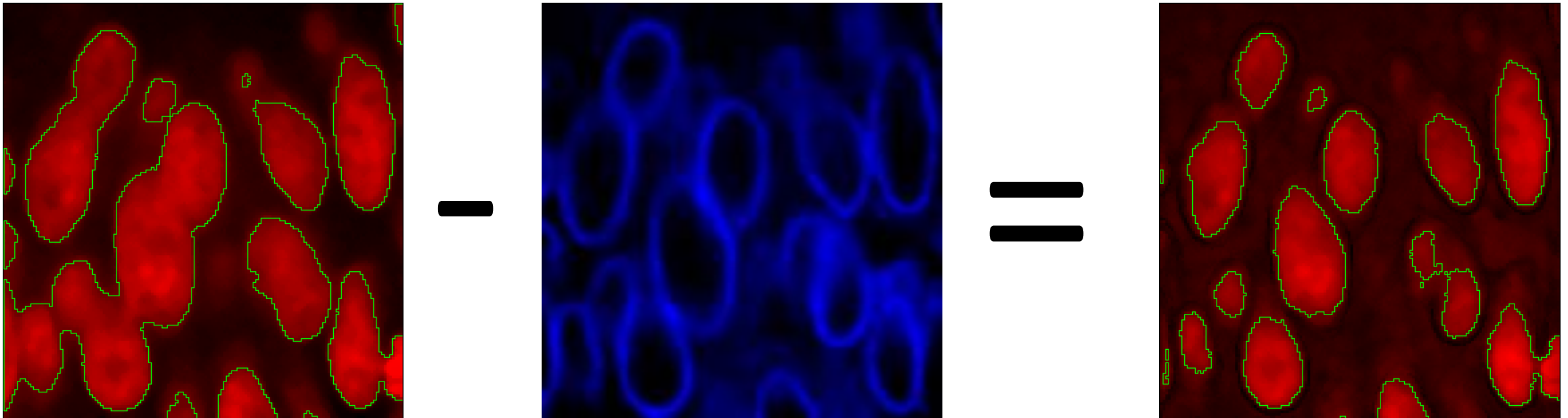


Image divided by
background

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

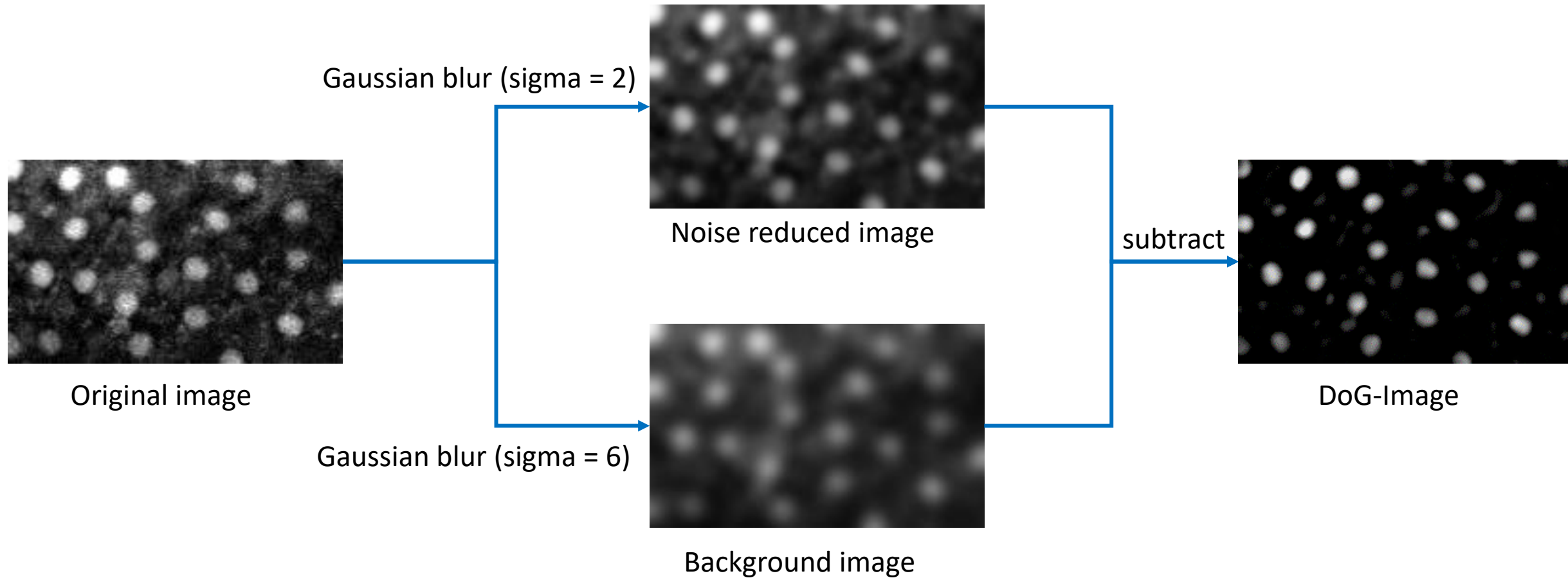


- Segmentation of nuclei can be improved by subtracting a channel visualizing nuclei envelope from the nuclei channel.



Difference-of-Gaussian (DoG)

- Improve image in order to detect bright objects.

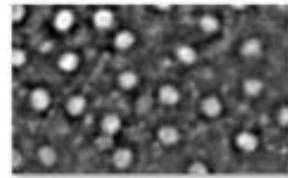


Difference-of-Gaussian (DoG)

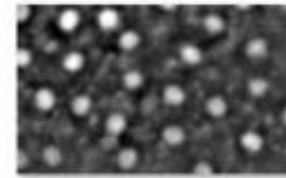
- Example DoG images



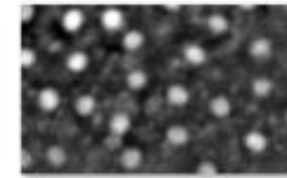
dog-1-1



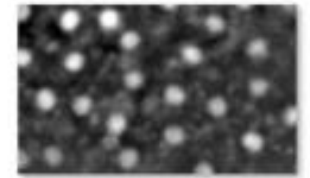
dog-1-4



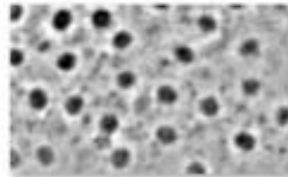
dog-1-7



dog-1-10



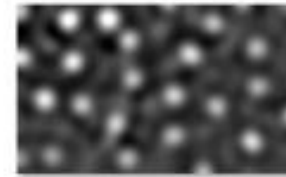
dog-1-13



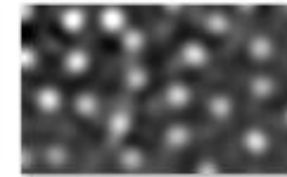
dog-4-1



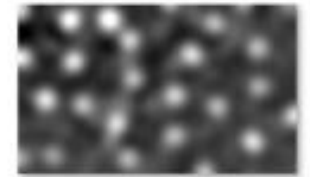
dog-4-4



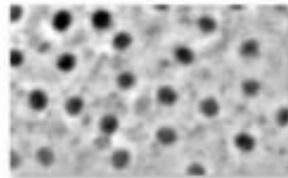
dog-4-7



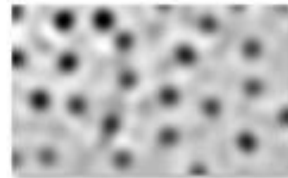
dog-4-10



dog-4-13



dog-7-1



dog-7-4



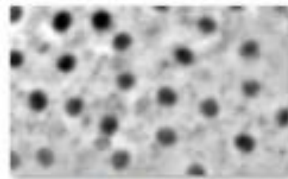
dog-7-7



dog-7-10



dog-7-13



dog-10-1



dog-10-4



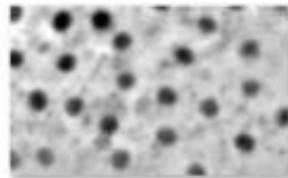
dog-10-7



dog-10-10



dog-10-13



dog-13-1



dog-13-4



dog-13-7

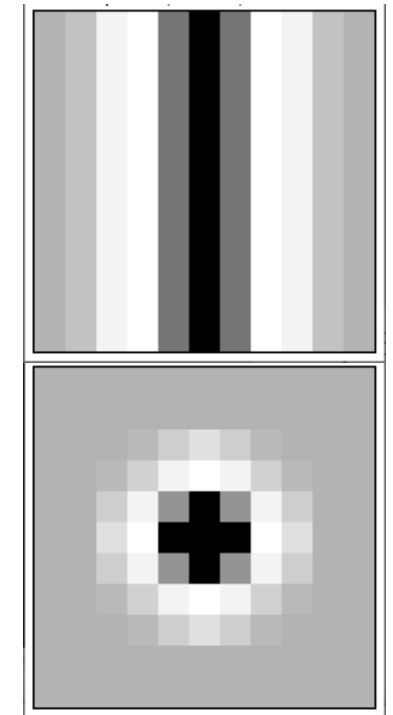
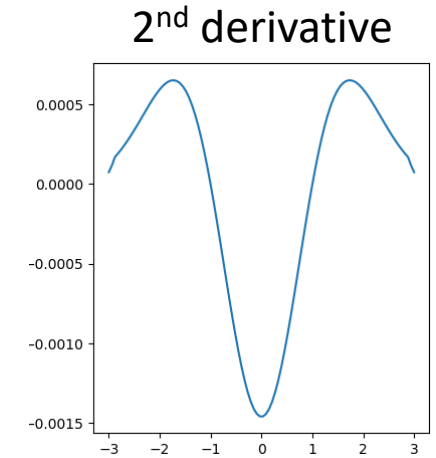
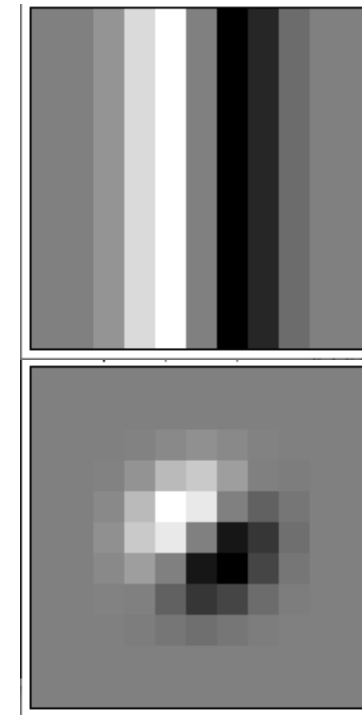
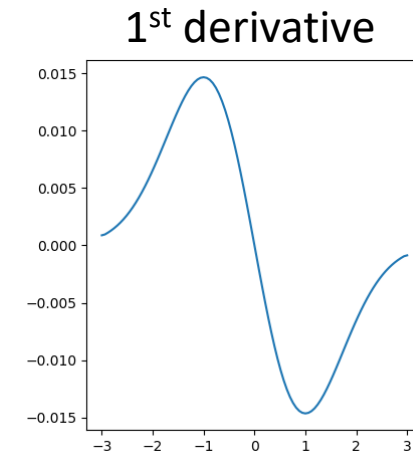
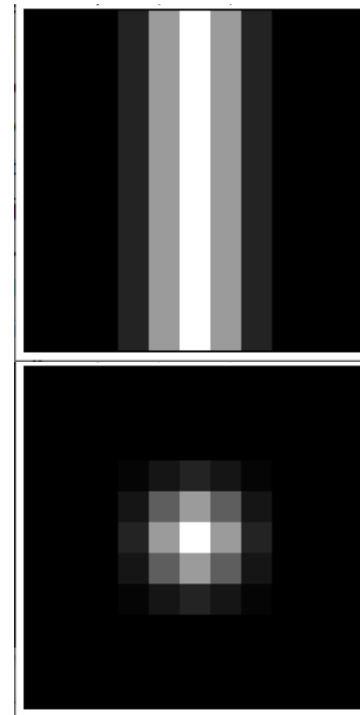
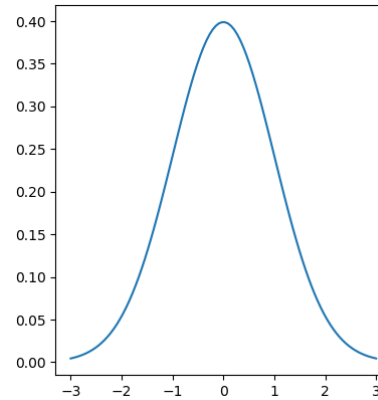
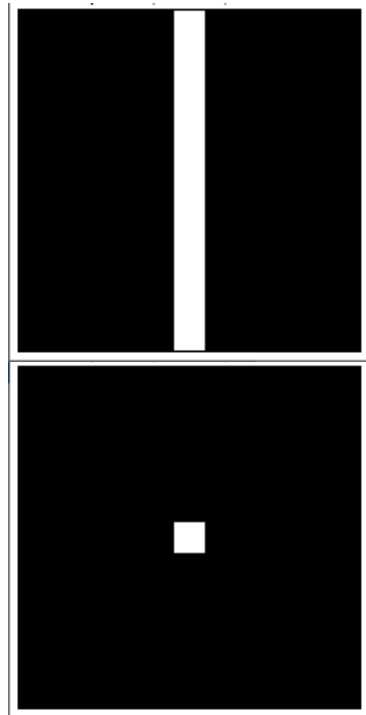


dog-13-10



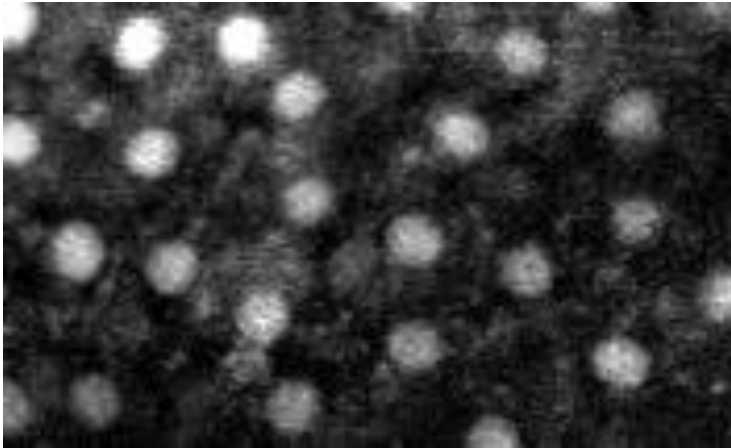
dog-13-13

- *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)

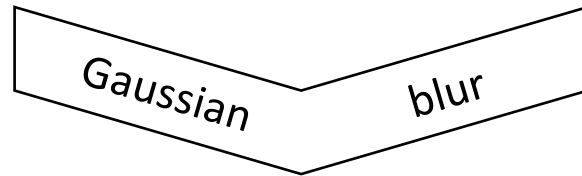
Laplace filter



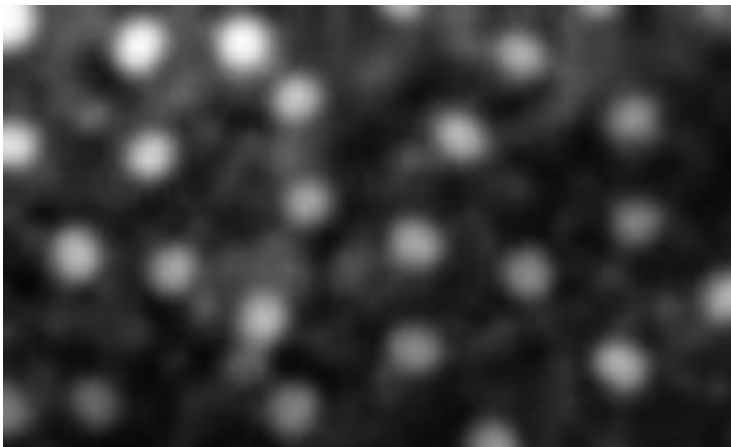
$$* \begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix} =$$



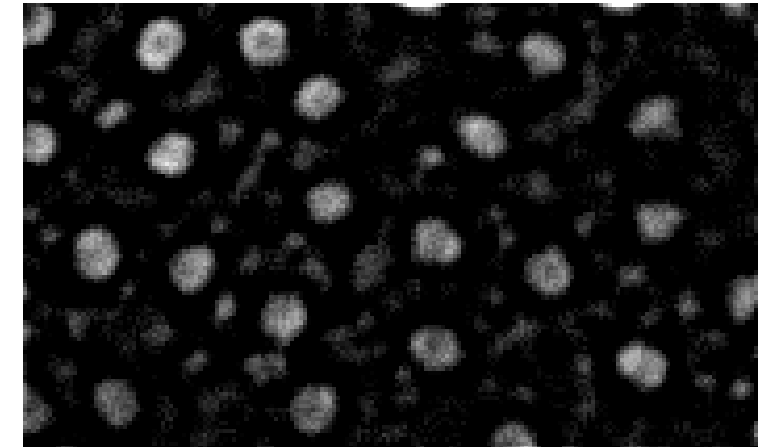
Laplace filtered image



Laplacian of Gaussian filter



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



LoG image

Image Filtering in Fiji

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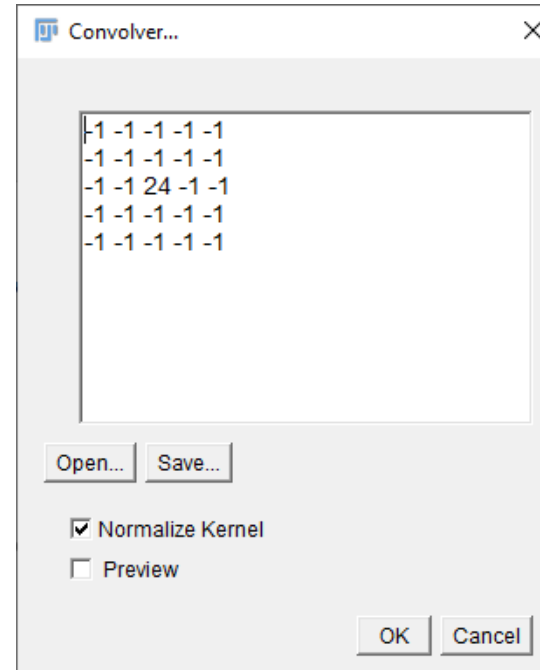
Alex Bird, Dan White, MPI CBG



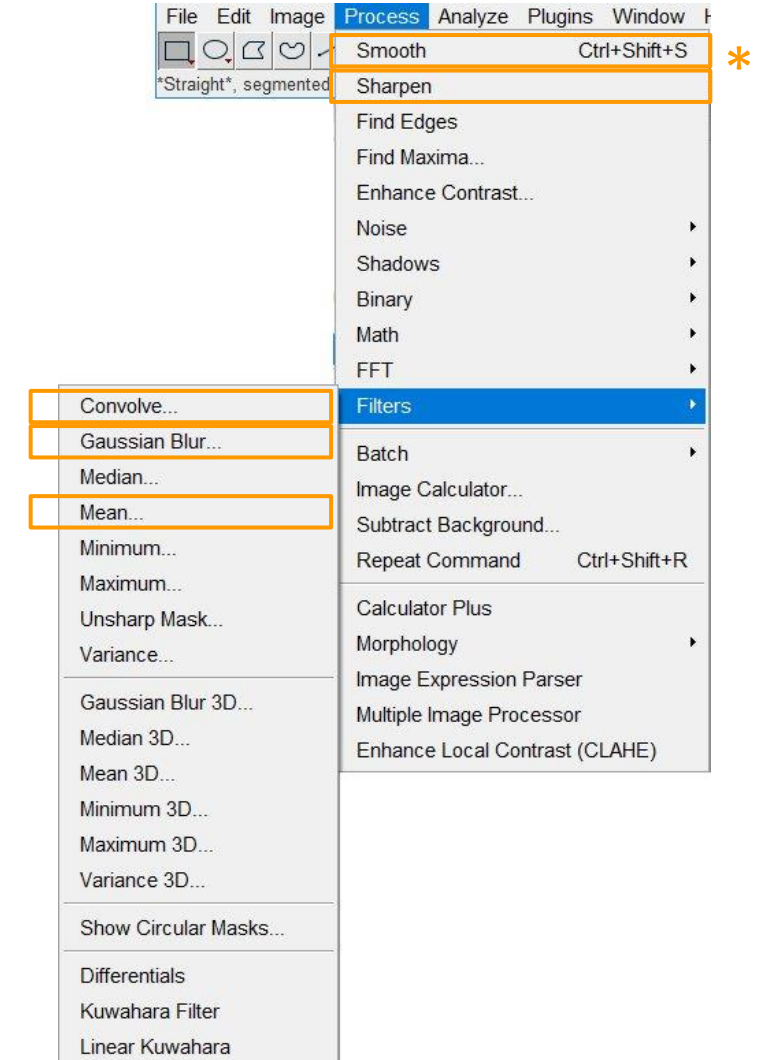
April 2020

Linear filters:

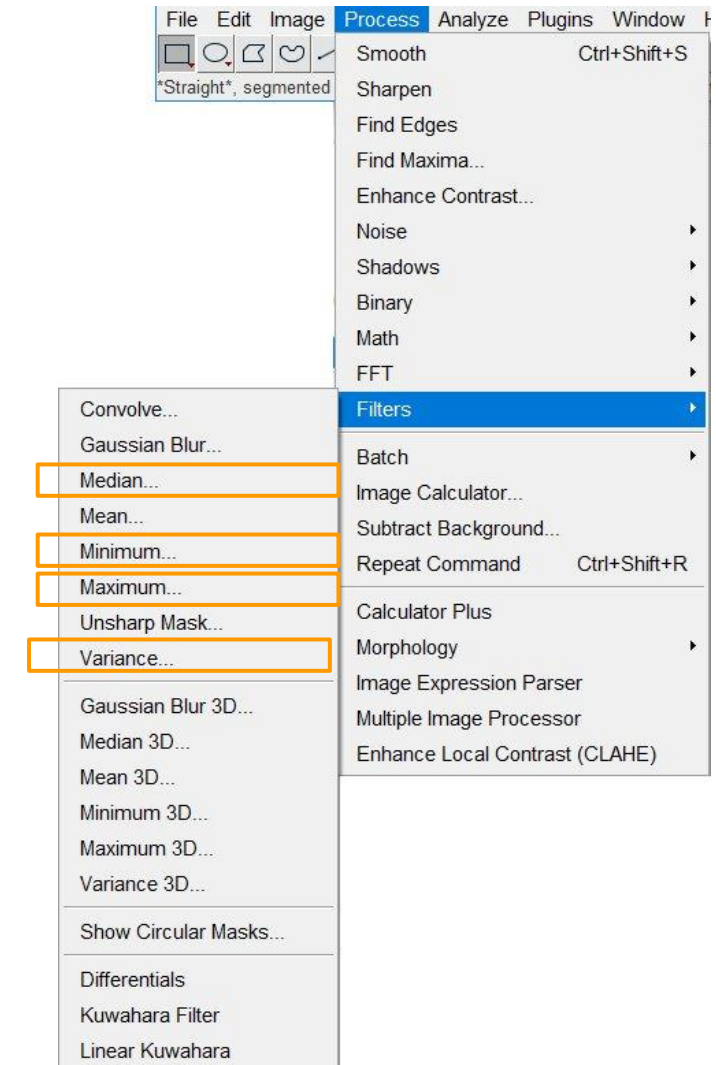
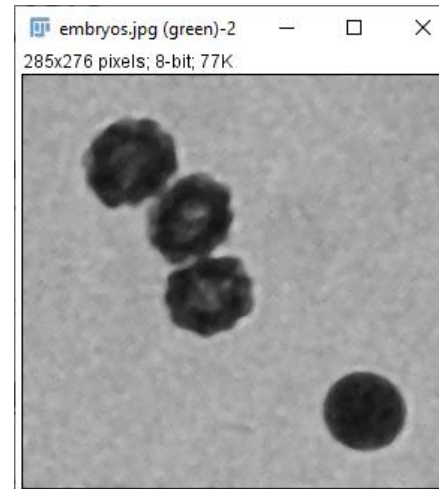
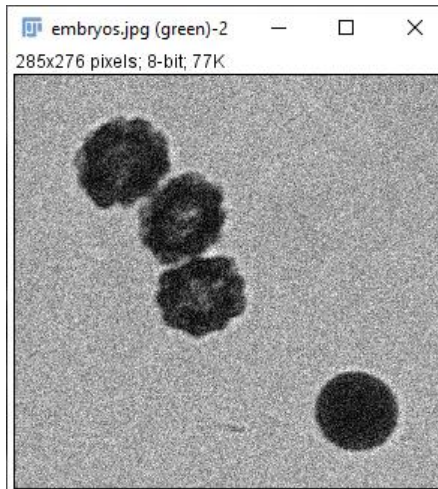
- Convolve...
 - Allows us to specify our kernel



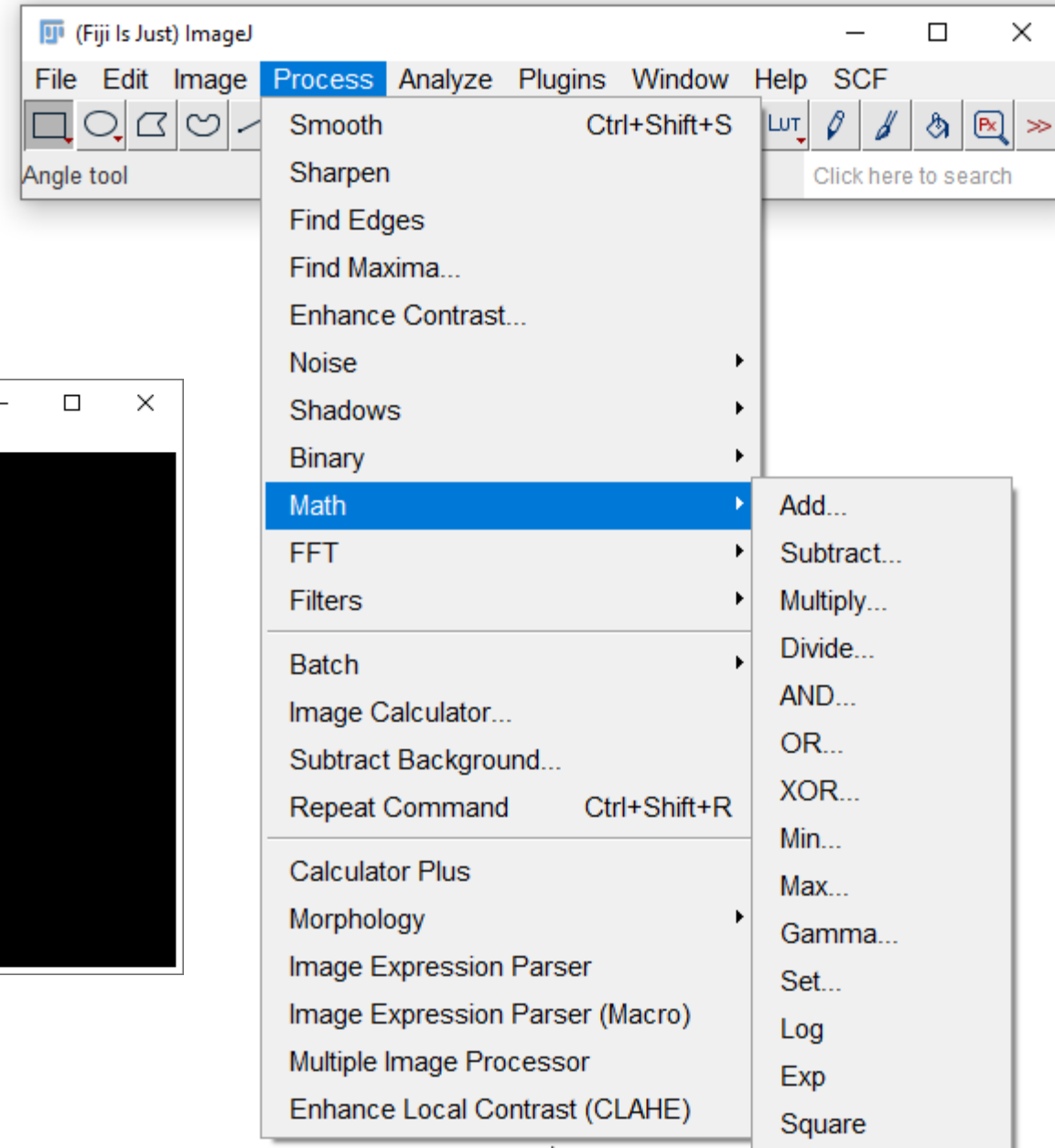
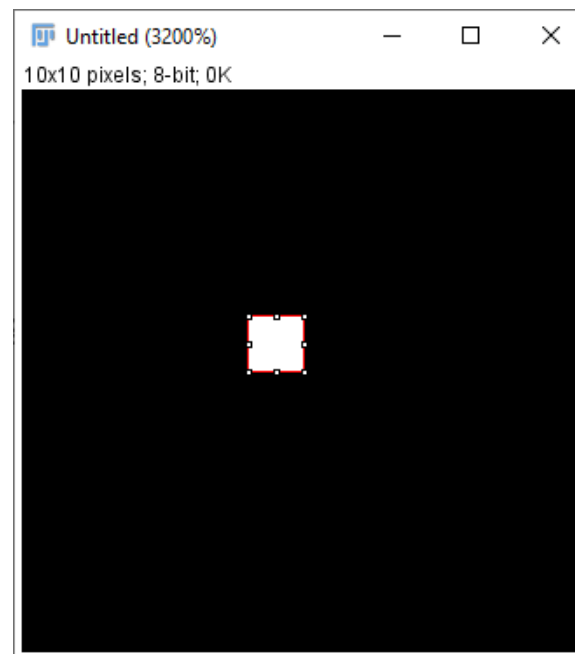
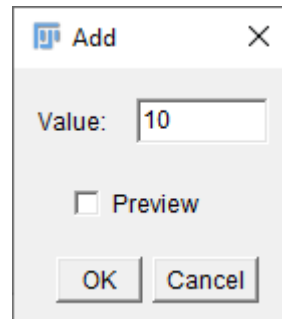
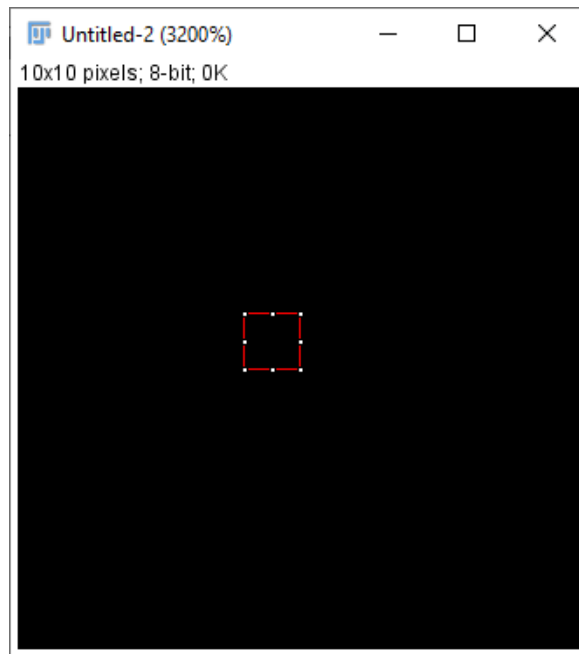
- Gaussian blur
- Mean
 - * Smooth is a 3x3 Mean filter



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

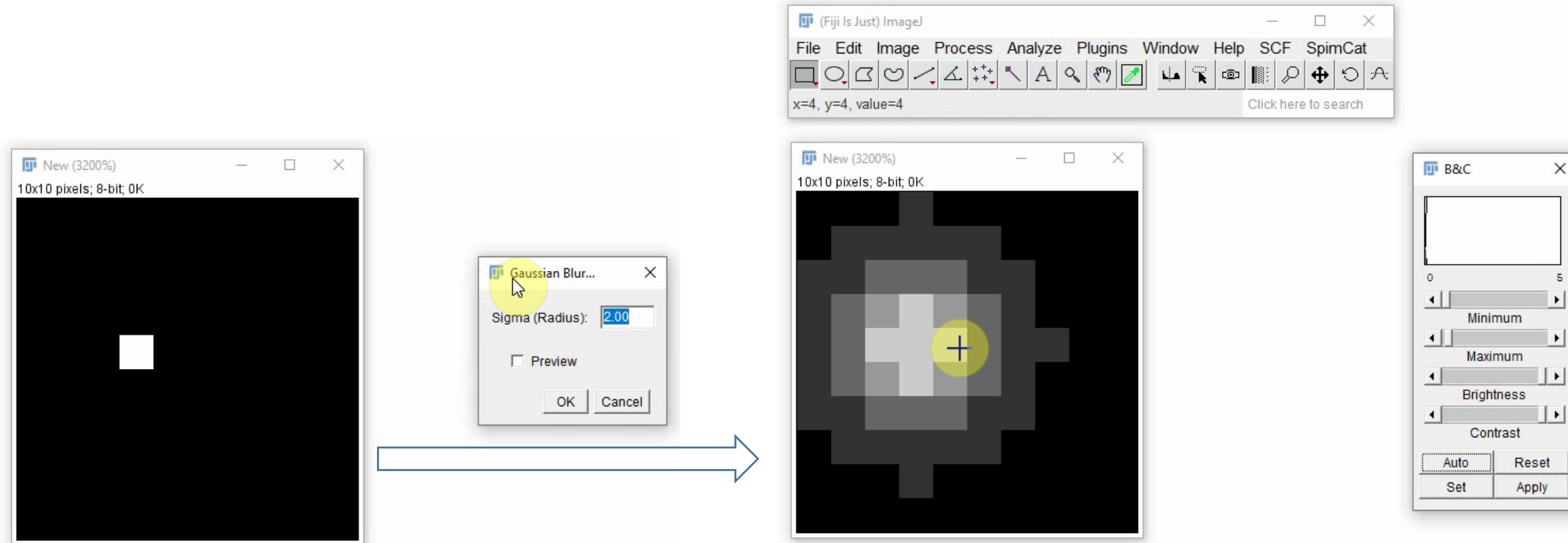


- All basic math operations for pixels are available in ImageJ/Fiji.
- If you want to simulate a binary image: Select some black pixels and add 255 to make them white.

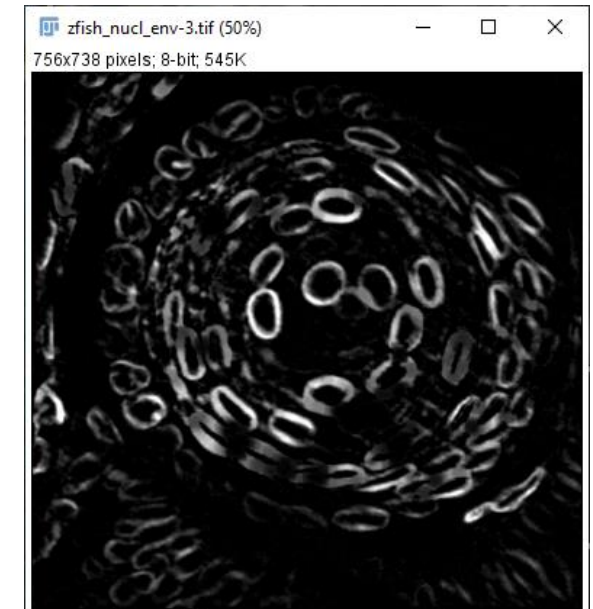
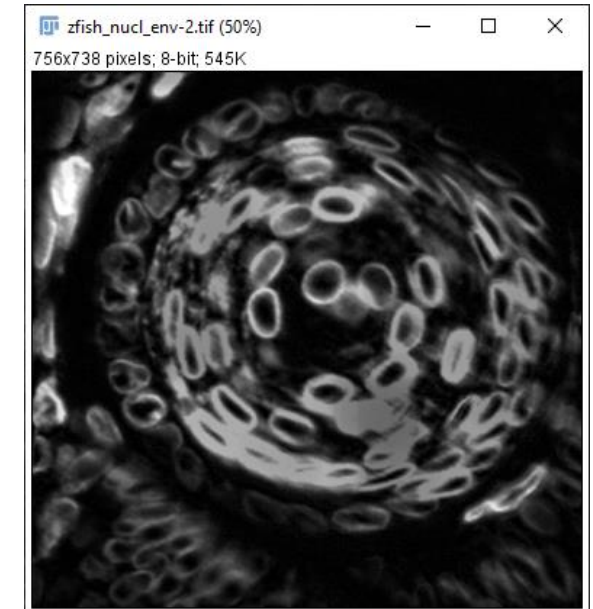
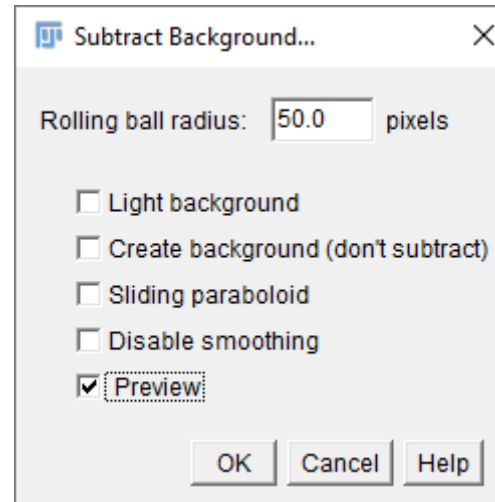


Hint: Image/pixel math

- Use images with single pixels > 0 and apply filters to understand what the filters do

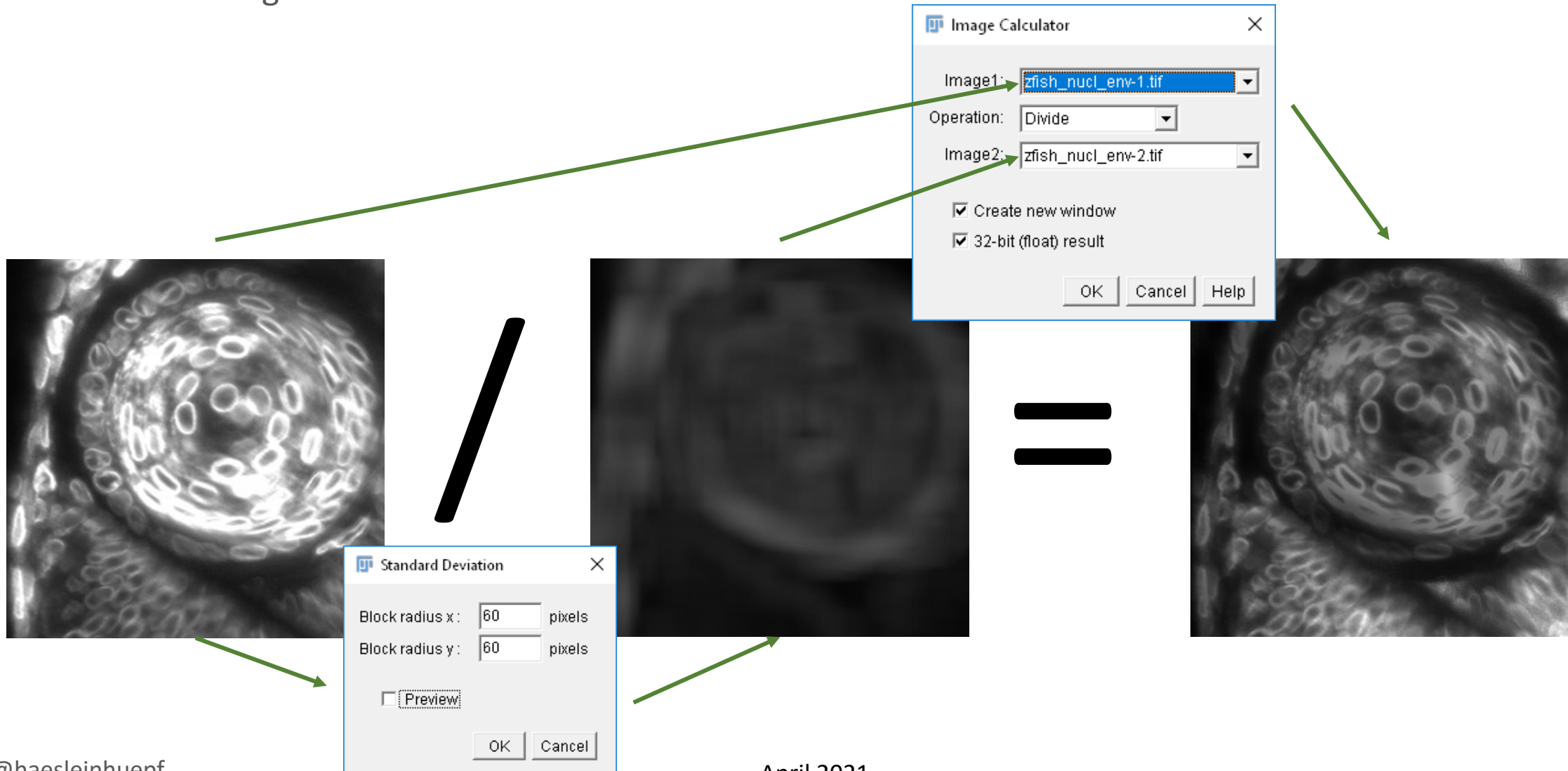


- Process > Subtract background...
 - The rolling ball radius should be a bit larger than the objects you are analyzing.
 - Try the preview checkbox while changing parameters!



Background removal

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



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

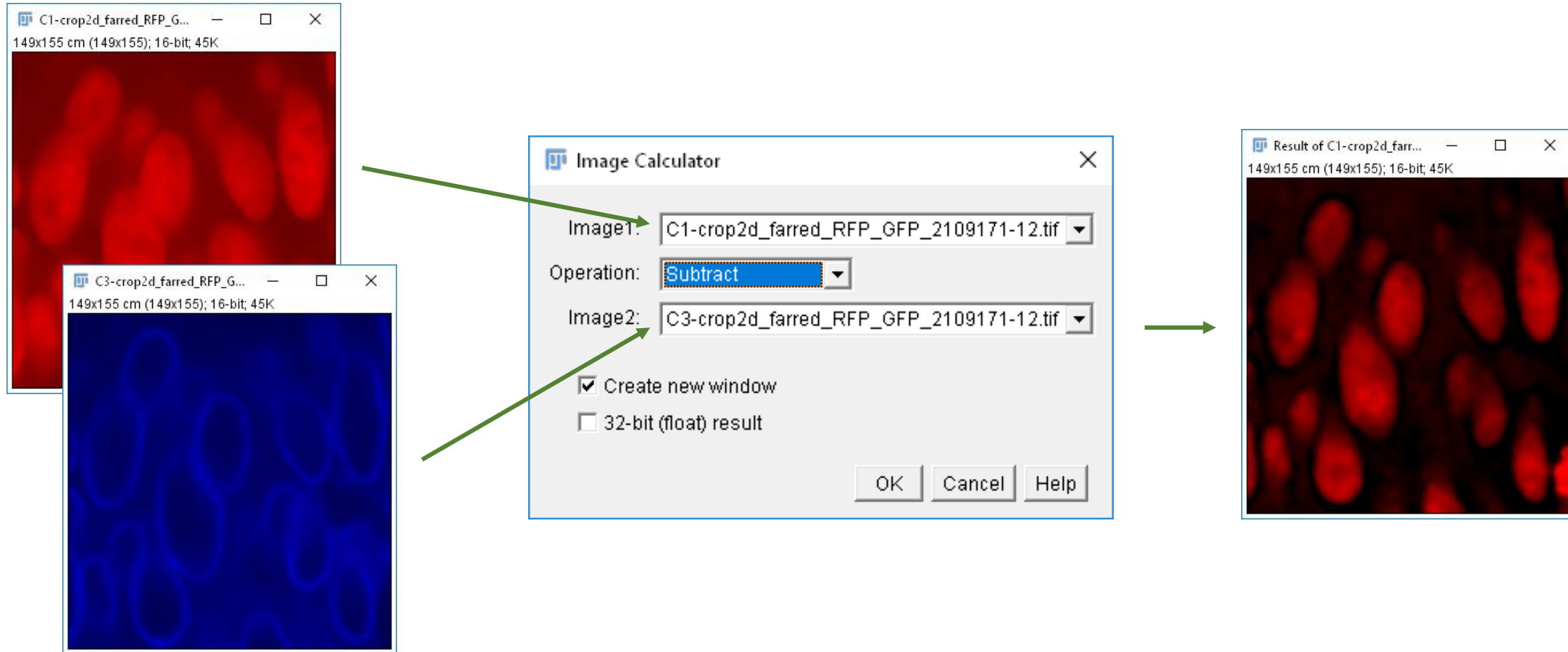




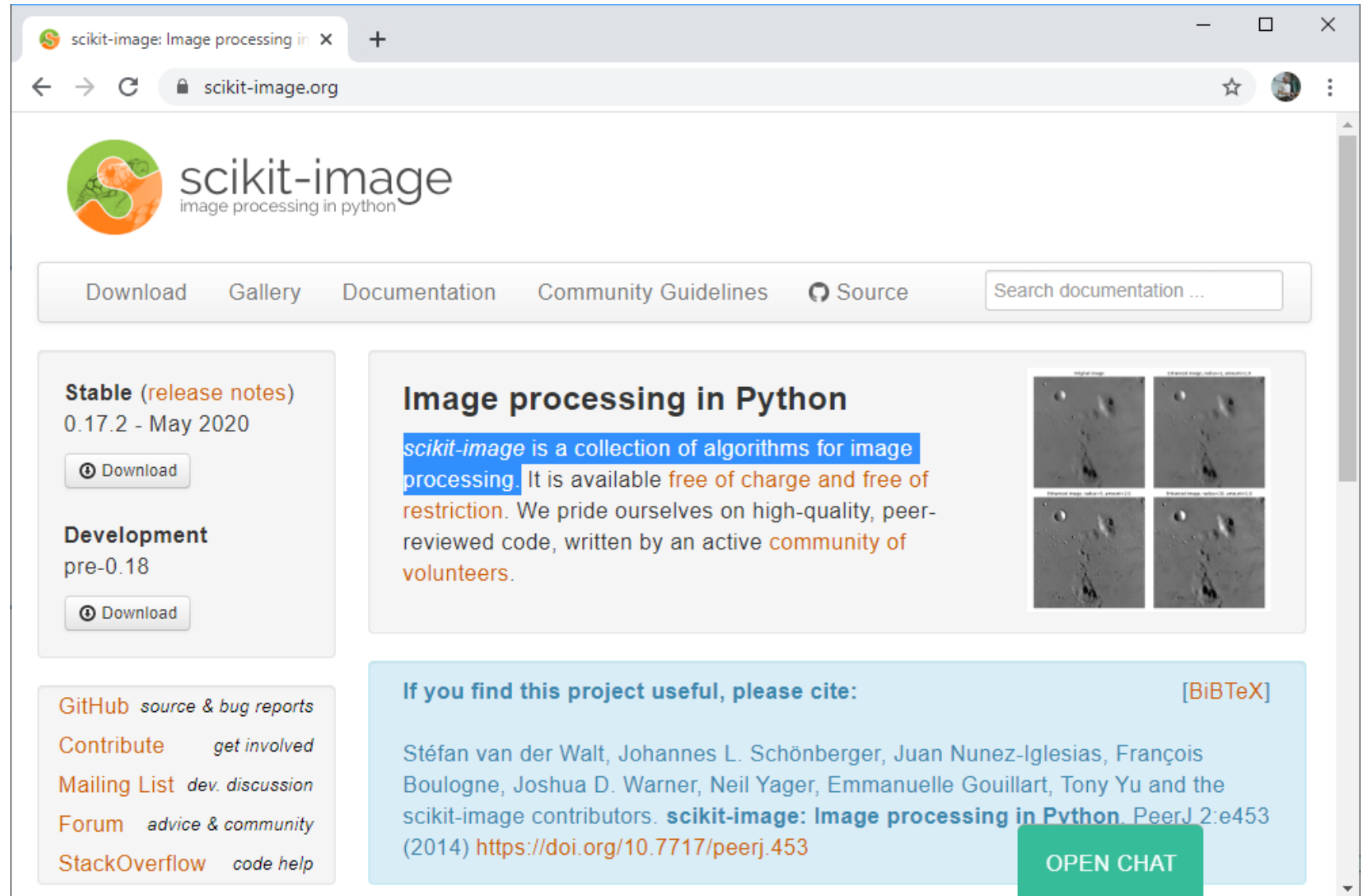
Image Filtering in Python using scikit-image

Robert Haase

April 2021

- *scikit-image* is a collection of algorithms for image processing.


```
conda install scikit-image
```



The screenshot shows the official website for scikit-image. The browser address bar displays 'scikit-image.org'. The website features a navigation bar with links for 'Download', 'Gallery', 'Documentation', 'Community Guidelines', and 'Source', along with a search bar for documentation. The main content area includes a 'Stable' release section (0.17.2 - May 2020) with a 'Download' button, and a 'Development' section (pre-0.18) also with a 'Download' button. A central banner titled 'Image processing in Python' describes scikit-image as a collection of algorithms for image processing, available free of charge and free of restriction, and highlights its high-quality, peer-reviewed code. To the right of this banner is a 2x2 grid of image processing examples. At the bottom, there is a citation section with the text 'If you find this project useful, please cite:' followed by a list of authors and a citation to a PeerJ paper, and a green 'OPEN CHAT' button.

scikit-image: Image processing in x +

← → ↻ 🔒 scikit-image.org ☆ 👤 ⋮

 **scikit-image**
image processing in python

Download Gallery Documentation Community Guidelines 🔍 Source 🔍 Search documentation ...

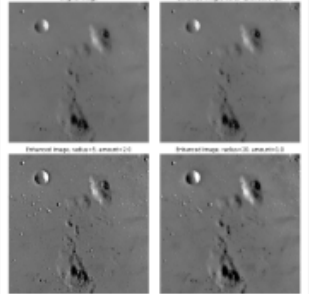
Stable (release notes)
0.17.2 - May 2020
[Download](#)

Development
pre-0.18
[Download](#)

[GitHub](#) source & bug reports
[Contribute](#) get involved
[Mailing List](#) dev. discussion
[Forum](#) advice & community
[StackOverflow](#) code help

Image processing in Python

scikit-image is a collection of algorithms for image processing. It is available free of charge and free of restriction. We pride ourselves on high-quality, peer-reviewed code, written by an active community of volunteers.



If you find this project useful, please cite: [BiBTeX]

Stéfan van der Walt, Johannes L. Schönberger, Juan Nunez-Iglesias, François Boulogne, Joshua D. Warner, Neil Yager, Emmanuelle Gouillart, Tony Yu and the scikit-image contributors. **scikit-image: Image processing in Python**. PeerJ 2:e453 (2014) <https://doi.org/10.7717/peerj.453>

[OPEN CHAT](#)

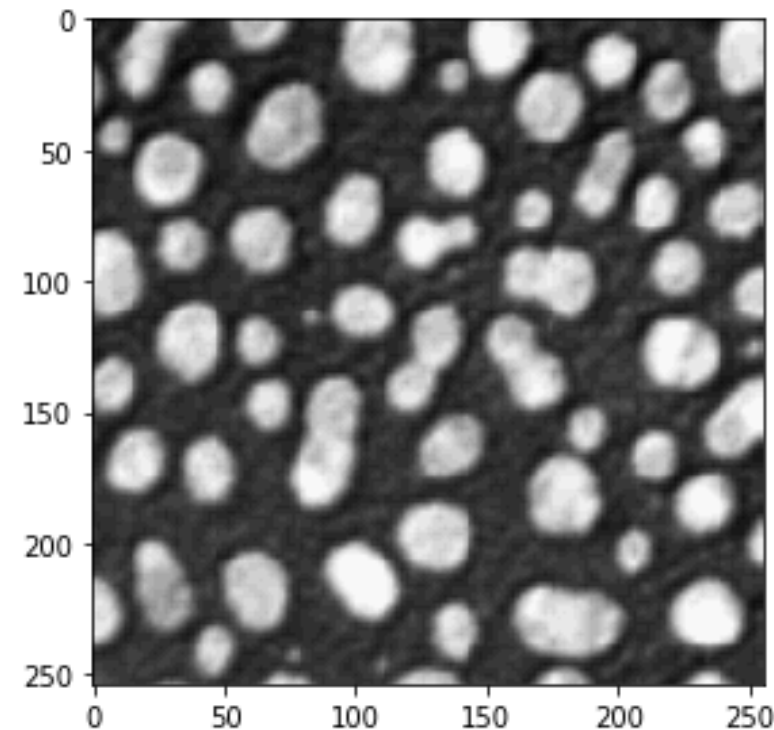
- Open images

```
from skimage.io import imread  
image = imread("blobs.tif")
```

- Visualize images

```
from skimage.io import imshow  
imshow(image)
```

<matplotlib.image.AxesImage at 0x245e7e5b220>



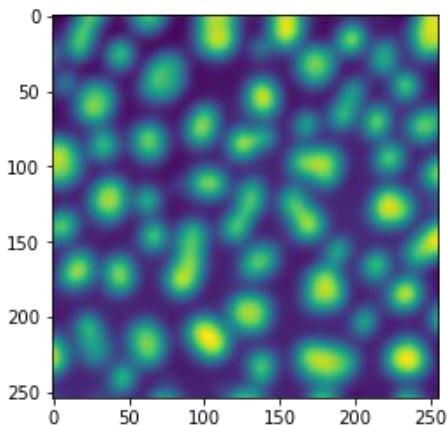
- Explore available filters (lectures 02, 07) and combine them

```
from skimage import filters  
filters.|
```

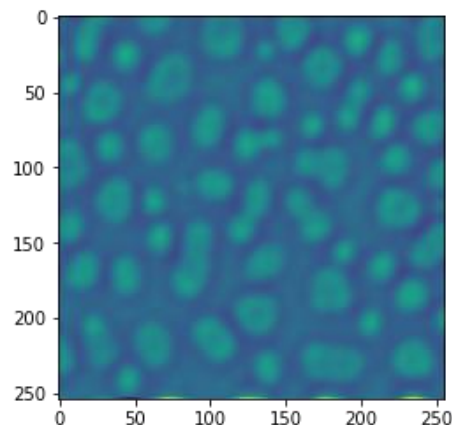
- LPIFilter2D
- median
- meijering
- prewitt
- prewitt_h
- prewitt_v
- rank
- rank_order
- ridges
- roberts

```
from skimage import filters  
  
# Gaussian blur  
gaussian_blurred_image = filters.gaussian(image, 5)  
plt.imshow(gaussian_blurred_image)  
plt.show()  
  
# LoG  
laplacian_of_gaussian = filters.laplace(gaussian_blurred_image)  
plt.imshow(laplacian_of_gaussian)  
plt.show()  
  
# another Gaussian blur  
gaussian_blurred_image_2 = filters.gaussian(image, 10)  
plt.imshow(gaussian_blurred_image_2)  
plt.show()  
  
# DoG  
difference_of_gaussian = gaussian_blurred_image - gaussian_blurred_image_2  
plt.imshow(difference_of_gaussian)  
plt.show()
```

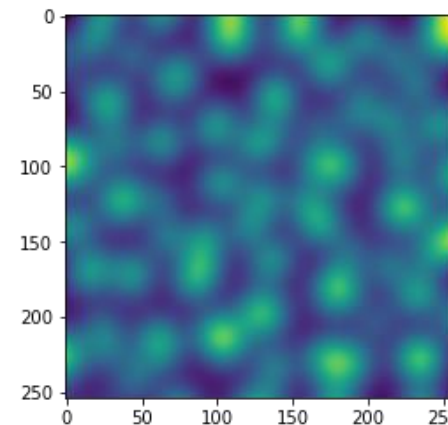
Gaussian blur (sigma=5)



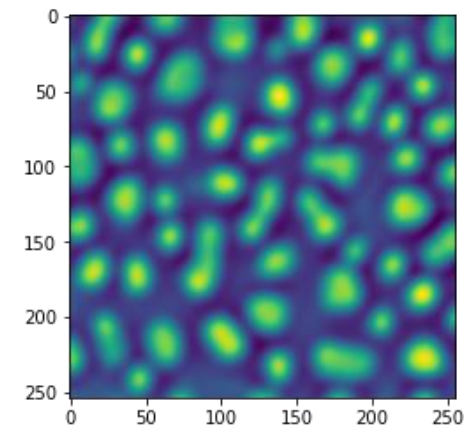
LoG



Gaussian blur (sigma=10)



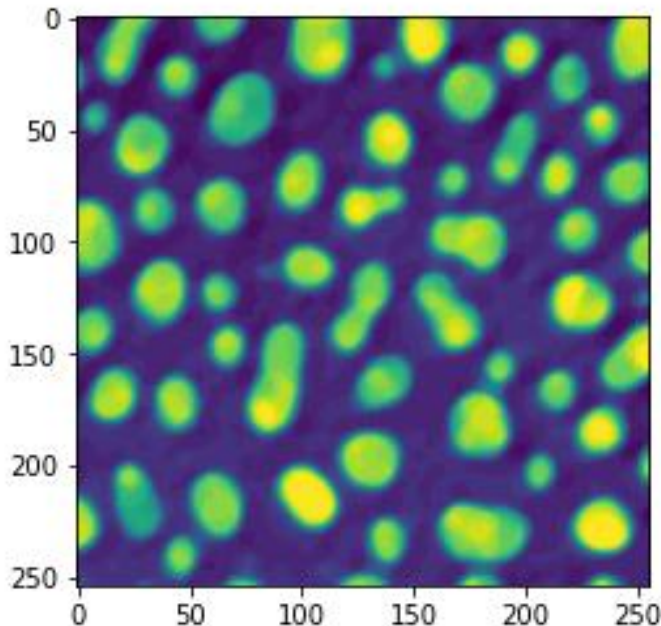
DoG



- Some filters ask for structuring elements

```
from skimage.morphology import disk

median_filtered = filters.median(image, disk(5))
plt.imshow(median_filtered)
plt.show()
```



Structuring
element

Module: filters — skimage v0.17.2

scikit-image.org/docs/stable/api/skimage.filters.html#skimage.filters.median

median

`skimage.filters.median(image, selem=None, out=None, mode='nearest', cval=0.0, behavior='ndimage')` [\[source\]](#)

Return local median of an image.

Parameters

image : array-like
Input image.

selem : ndarray, optional
If `behavior=='rank'`, `selem` is a 2-D array of 1's and 0's. If `behavior=='ndimage'`, `selem` is a N-D array of 1's and 0's with the same number of dimension than `image`. If `None`, `selem` will be a N-D array with 3 elements for each dimension (e.g., vector, square, cube, etc.)

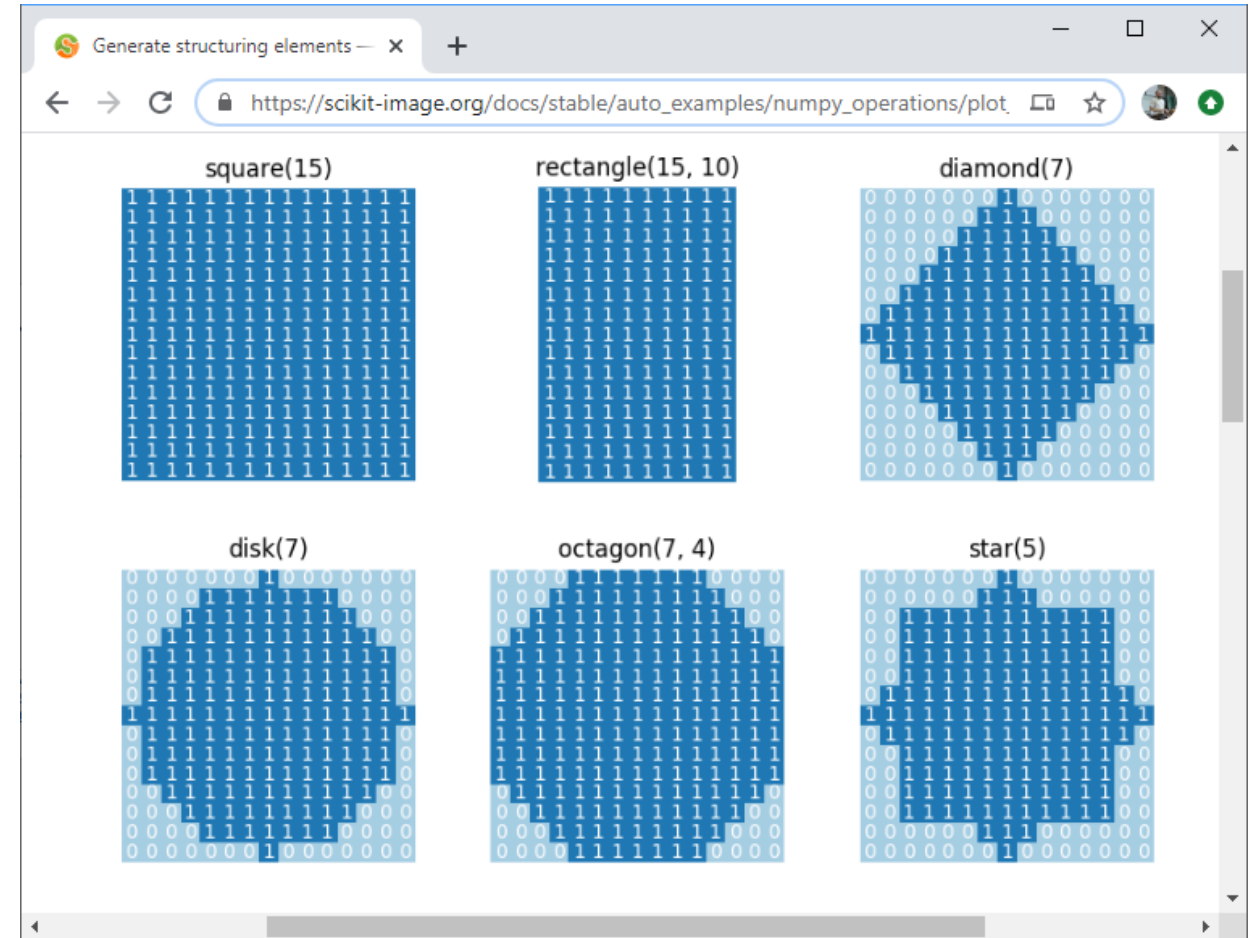
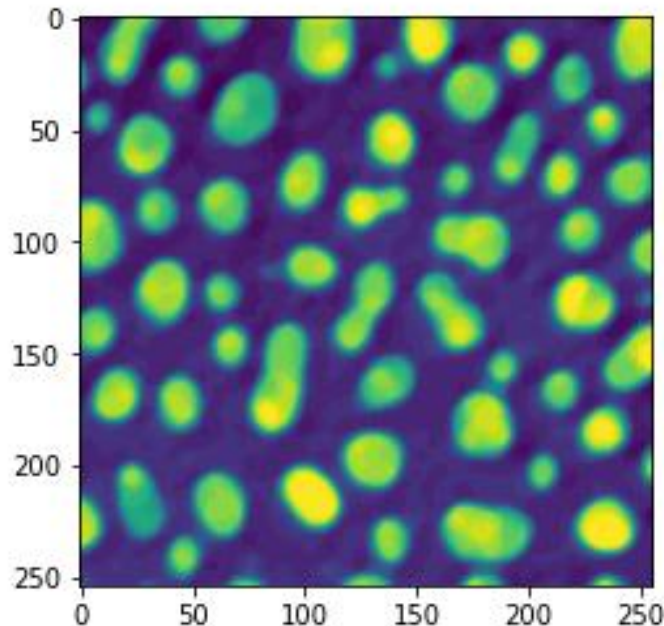
out : ndarray, (same dtype as image), optional
If `None`, a new array is allocated.

mode : {'reflect', 'constant', 'nearest', 'mirror', 'wrap'}, optional
The mode parameter determines how the array borders are handled, where `cval` is the value when mode is equal to 'constant'. Default is 'nearest'.
New in version 0.15: `mode` is used when `behavior='ndimage'`.

- Some filters ask for structuring elements

```
from skimage.morphology import disk

median_filtered = filters.median(image, disk(5))
plt.imshow(median_filtered)
plt.show()
```



https://scikit-image.org/docs/stable/auto_examples/numpy_operations/plot_structuring_elements.html#sphx-gl-r-auto-examples-numpy-operations-plot-structuring-elements-py

Image visualization: subplots

- Use matplotlib to put images side-by-side

```
median_filtered = filters.median(noisy_mri, disk(1))
mean_filtered = filters.rank.mean(noisy_mri, disk(1))
gaussian_filtered = filters.gaussian(noisy_mri, sigma=1)

fig, axs = plt.subplots(2, 3, figsize=(15,10))

# first row
axs[0, 0].imshow(median_filtered)
axs[0, 0].set_title("Median")
axs[0, 1].imshow(mean_filtered)
axs[0, 1].set_title("Mean")
axs[0, 2].imshow(gaussian_filtered)
axs[0, 2].set_title("Gaussian")

# second row
axs[1, 0].imshow(median_filtered[50:100, 50:100])
axs[1, 1].imshow(mean_filtered[50:100, 50:100])
axs[1, 2].imshow(gaussian_filtered[50:100, 50:100])
```

row column

https://matplotlib.org/stable/api/as_gen/matplotlib.pyplot.subplots.html

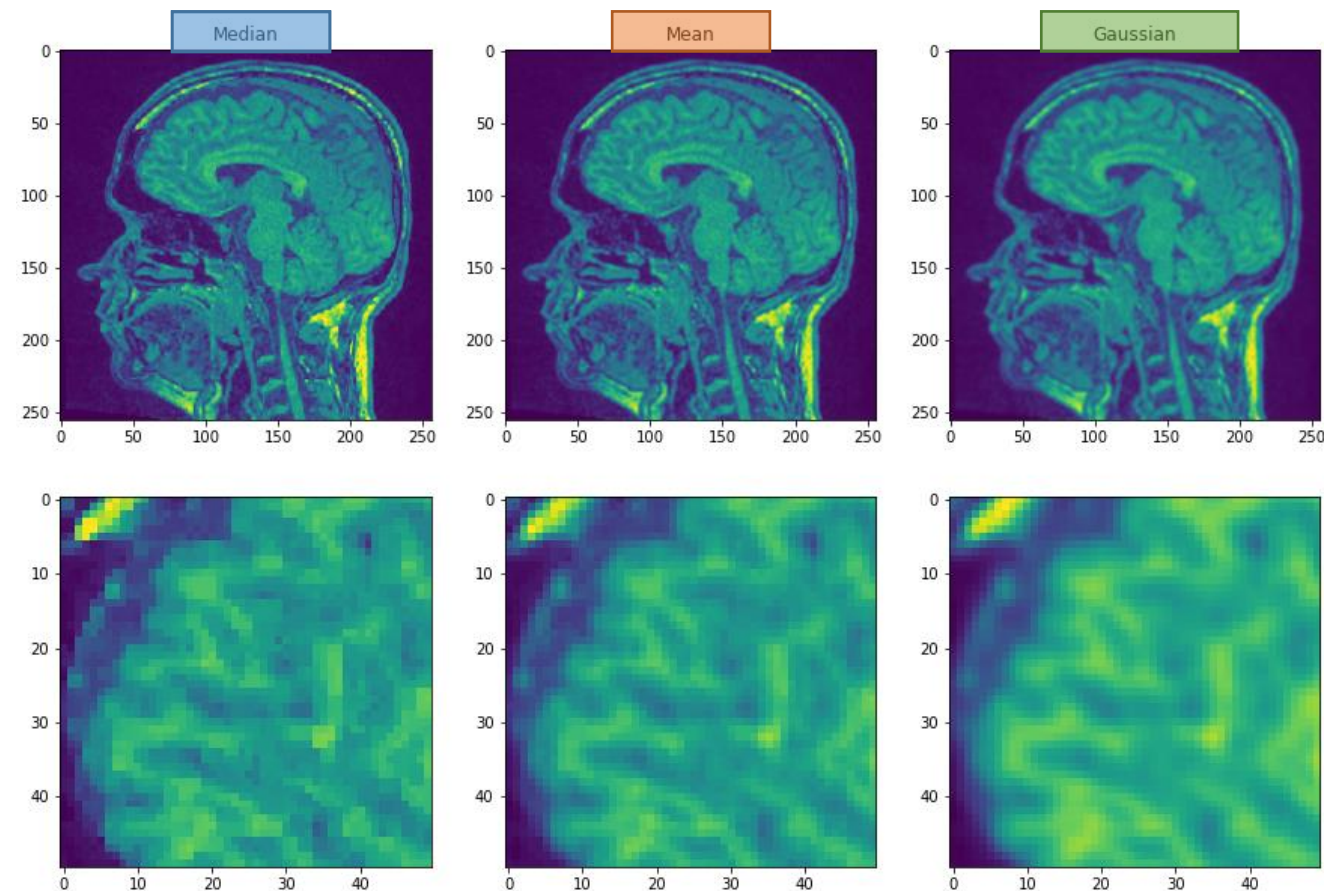
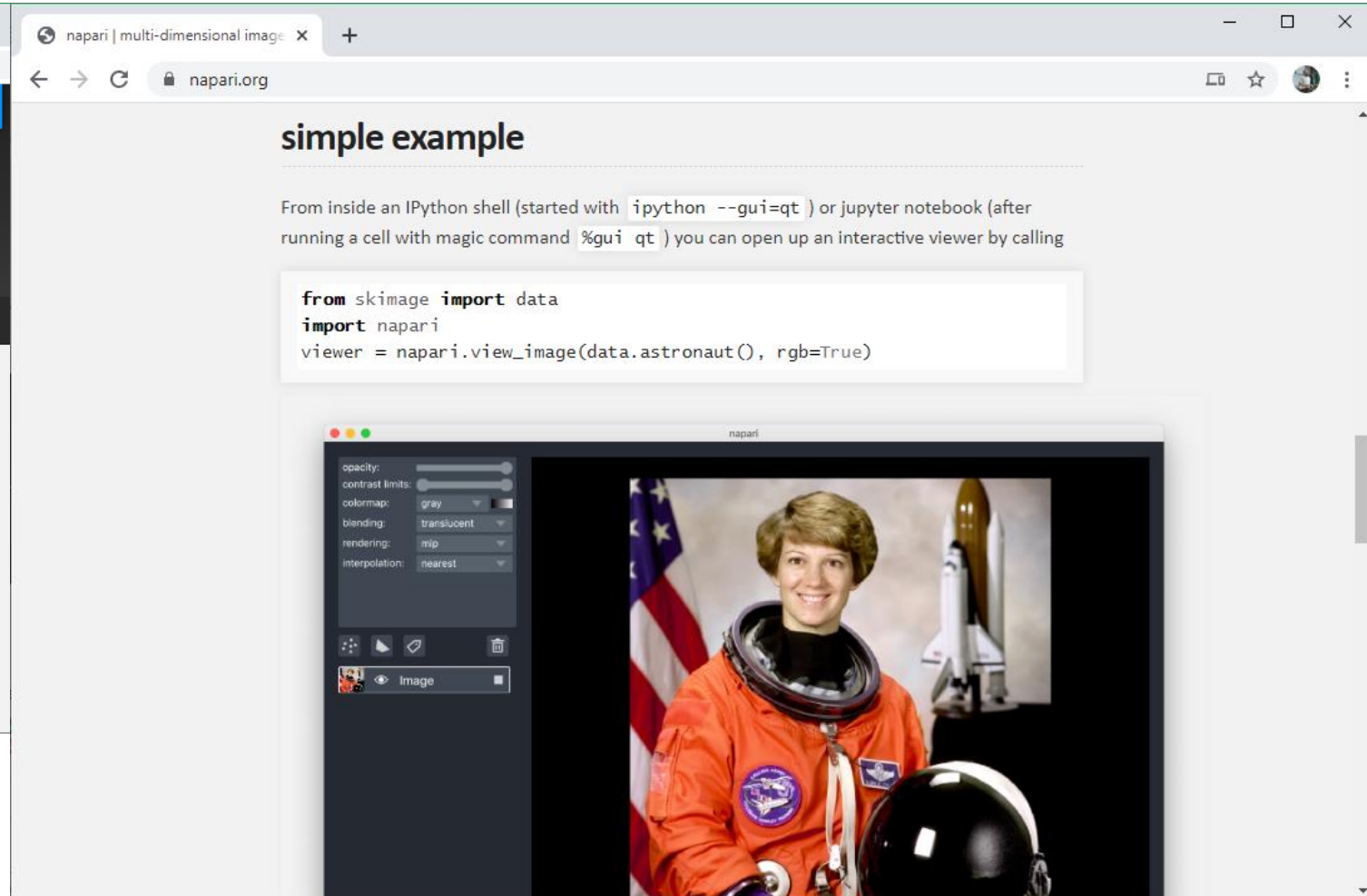
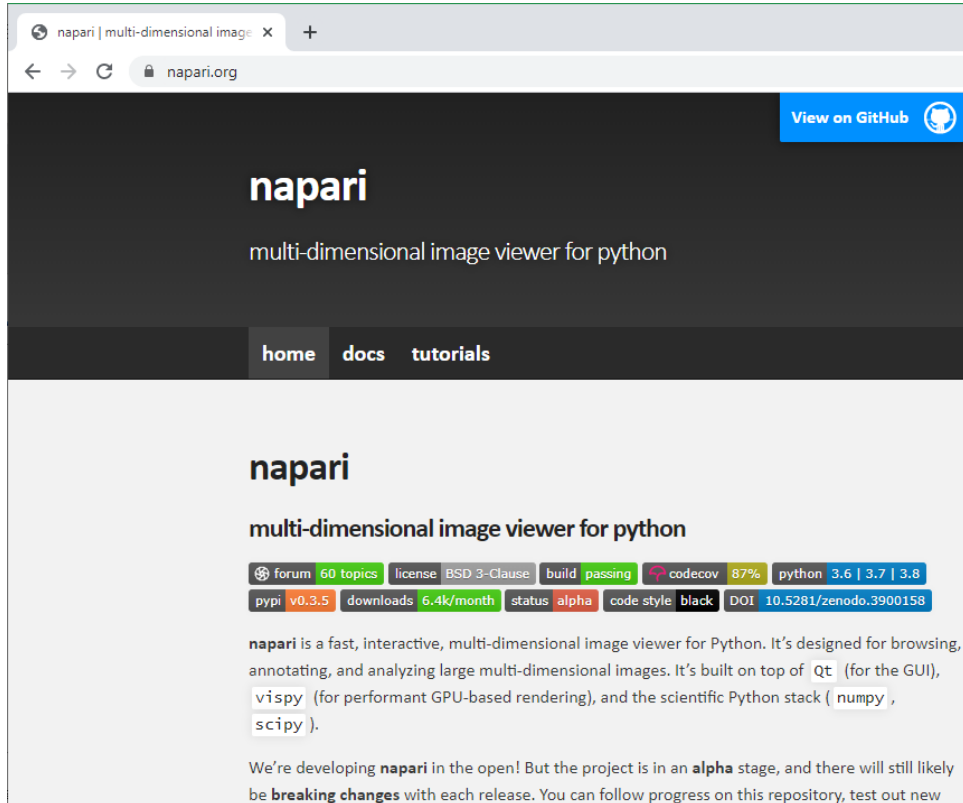


Image visualization in python using napari

Robert Haase

April 2021

- Multi-dimensional image viewer in python
- <https://napari.org/>



- Initialisation

```
▶ %gui qt
```

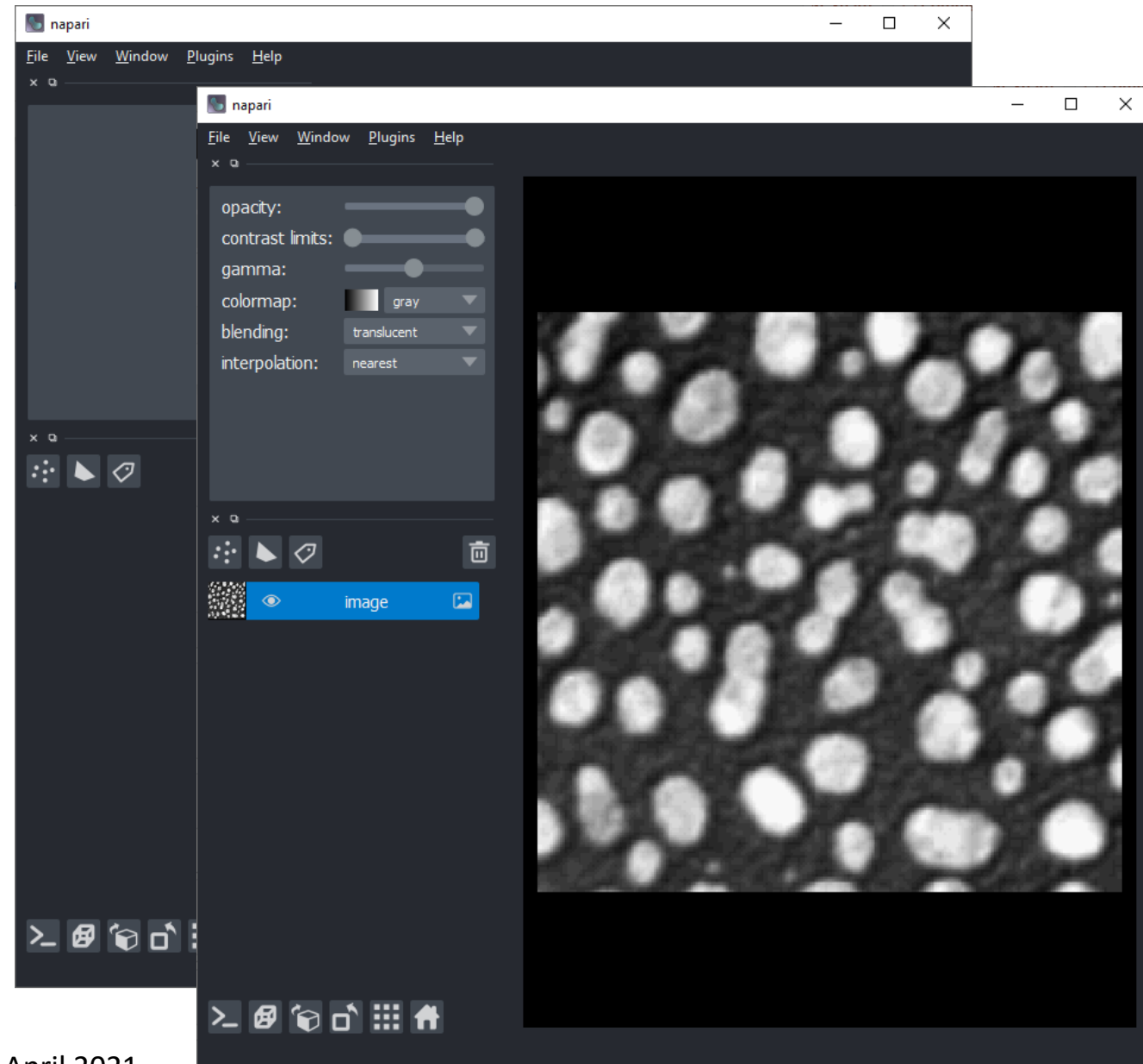
- Open a window

```
import napari

# Create an empty viewer
viewer = napari.Viewer()
```

- Add an image (layer)

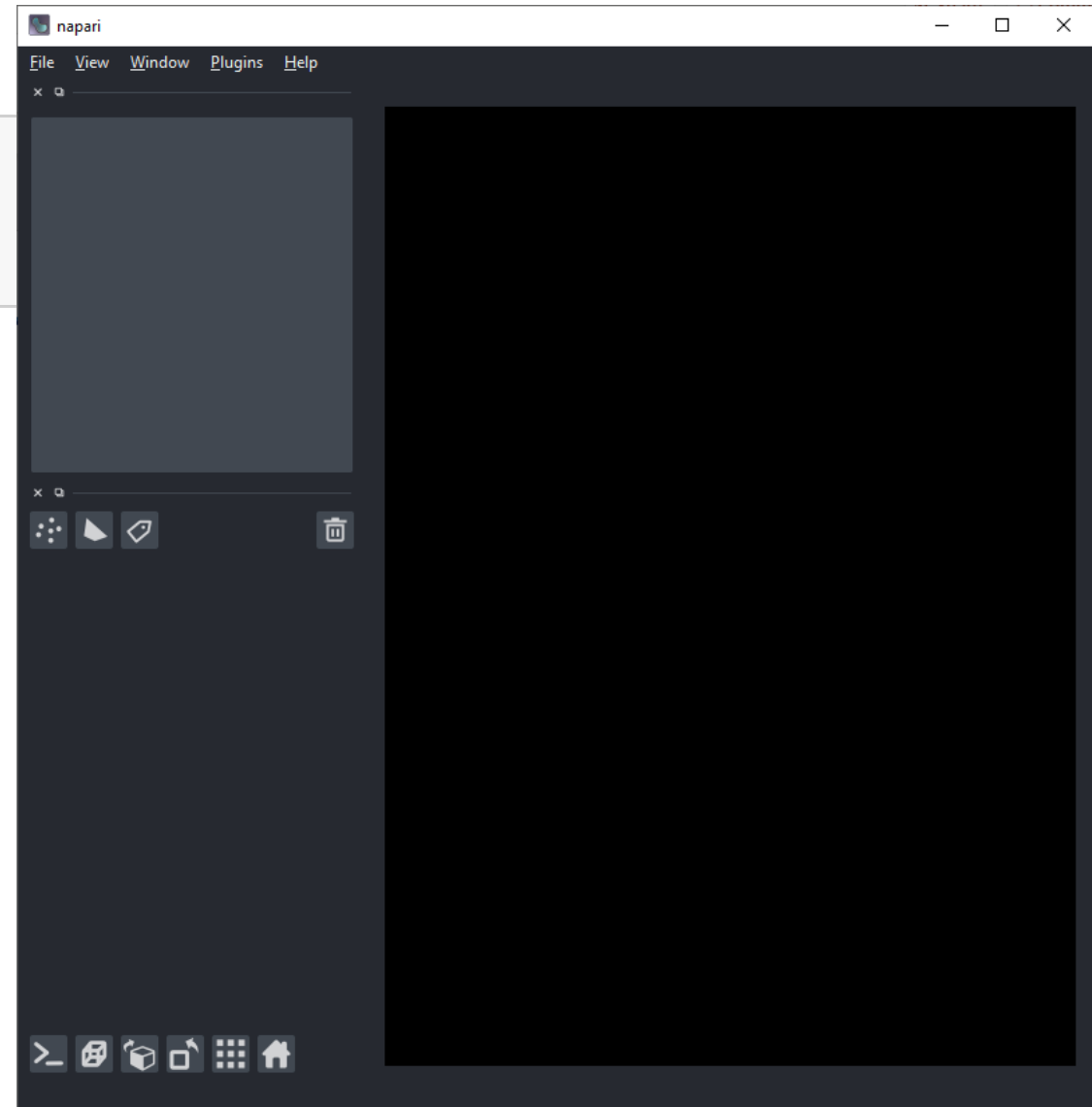
```
▶ # Add a new layer containing an image
viewer.add_image(image);
```



Removing layers

- Remove all layers from napari

```
# Remove all layers to start from scratch  
for l in viewer.layers:  
    viewer.layers.remove(l)
```



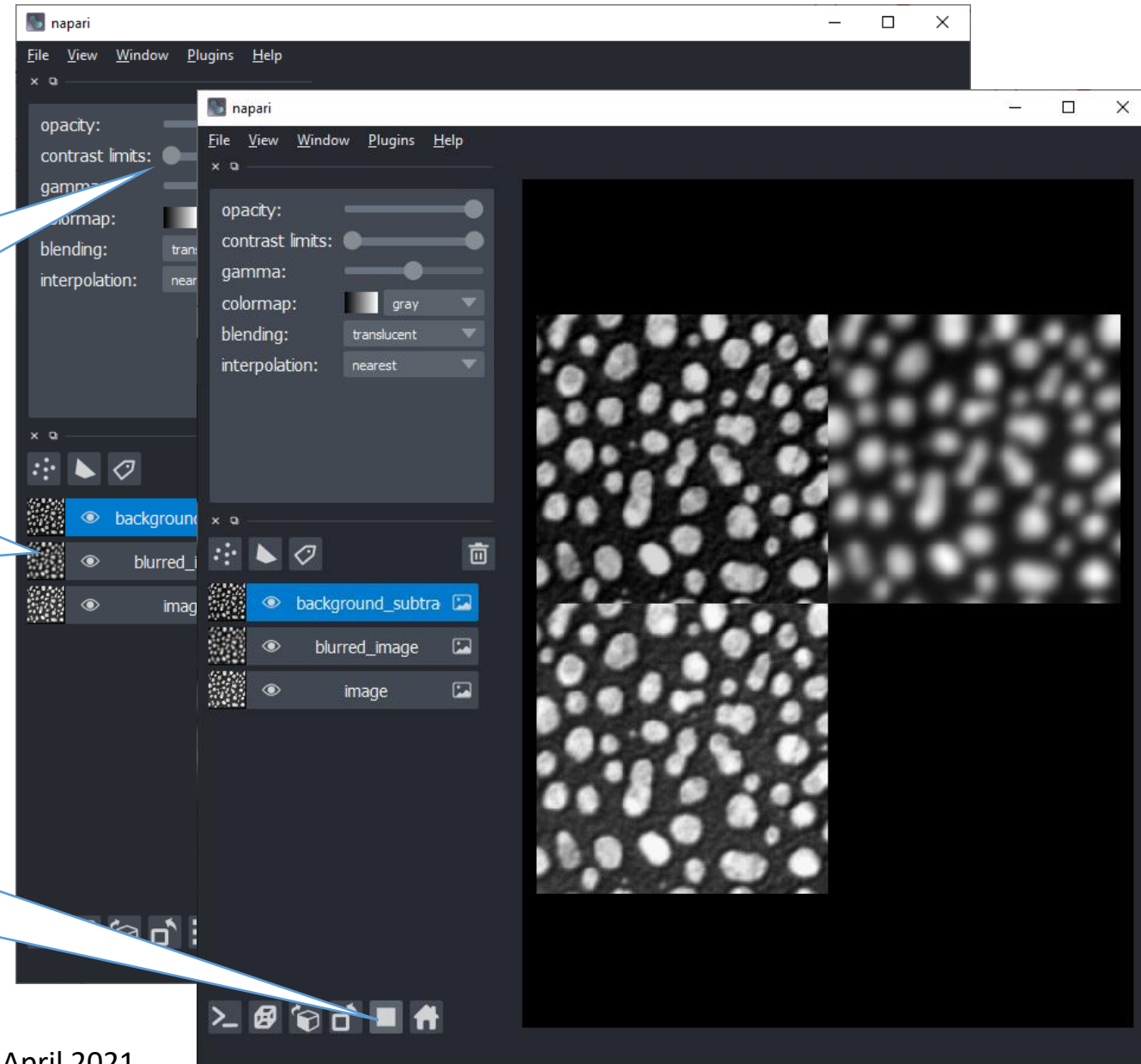
Add multiple images as layers

- Add layers to napari to visualize intermediate processing results on top of each other or side by side

Change
Brightness
and contrast
here

Toggle
visualization
of layers
here

Gallery view



When to use what? (opinionated slide)

- ImageJ / Fiji
 - Established platform for 20 years
 - Right tool for quick exploration of image data
 - Classical algorithms
 - Hundreds of plugins available
 - Not compatible with python
 - Workflow development limited to ImageJ compatible functionality
 - Pixel-by-pixel comparison complicated
- napari
 - Under development
 - Limited functionality (yet)
 - Small number of plugins available, e.g. for deep-learning
 - Compatible with python
 - Workflows can include any kind of data analysis tools from the python ecosystem
 - Layer-concept makes comparing images easy



- Image filtering
 - Noise removal
 - Background subtraction
- Tools
 - ImageJ / Fiji
 - Python
 - Scikit-image
 - napari

Coming up next

- Image Segmentation
 - Thresholding
 - Mask refinement
 - Connected component analysis

