



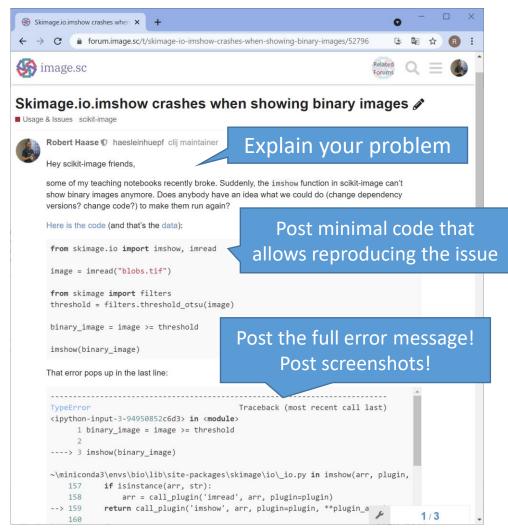
# GPU-accelerated image processing

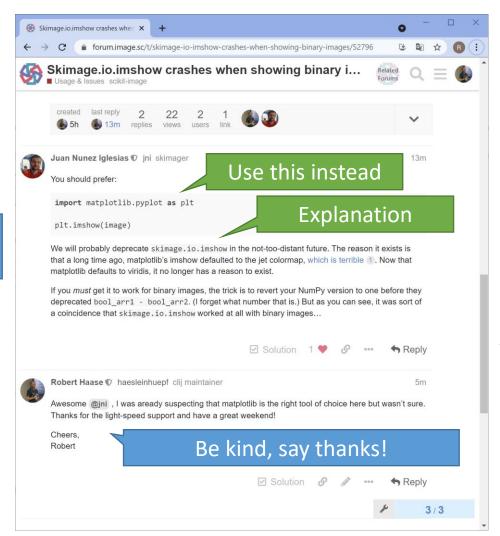
Robert Haase

## The image science community: https://image.sc



In case you run in trouble with image analysis, observe bugs in open-source software or want to know how to analyze your image data: Ask experts!



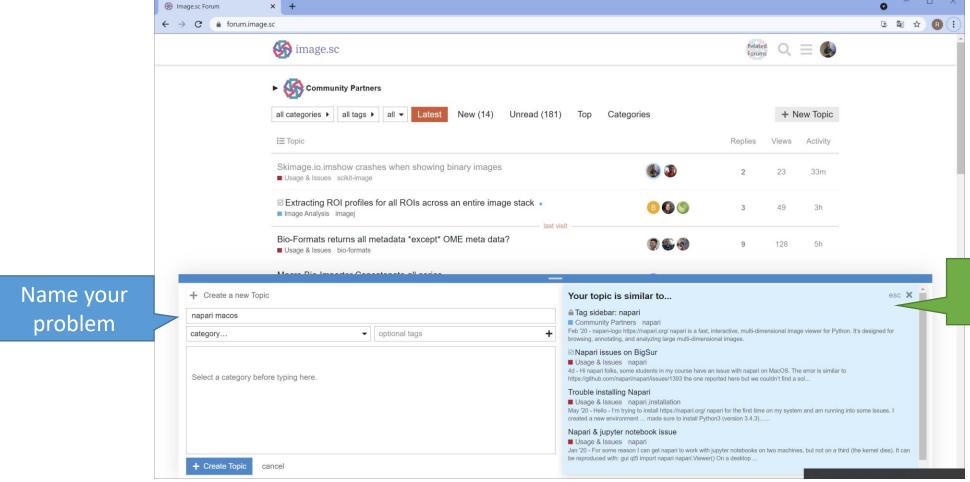




## The image science community: https://image.sc



- In case you run in trouble with image analysis, observe bugs in open-source software or want to know how to analyze your image data: <u>Ask experts!</u>
- Plus: discover other's people similar questions



Discover similar problems

## GPU-accelerated image processing



State-of-the-art software for more than 20 years: ImageJ / Fiji





@haesleinhuepf

## GPU-accelerated image processing



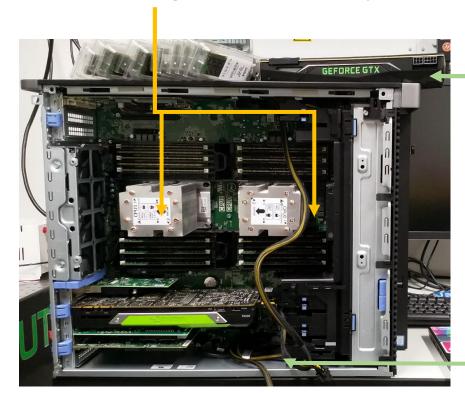
• The dawn of graphics processing units changes how we interact with image data



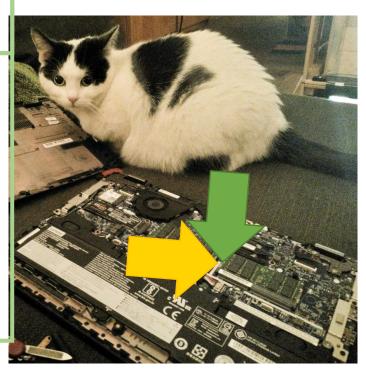


Every computer that has a screen also has a GPU.

#### Central Processing Unit (CPU) Graphics Processing Unit (GPU)



dedicated GPUs



integrated GPUs

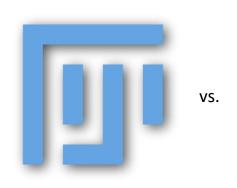


external "eGPUs"

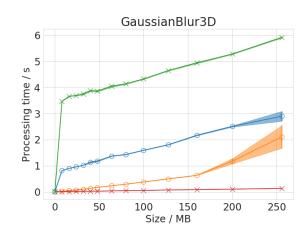
## Why GPU-acceleration?



• 1-2 orders of magnitude speedup are possible









2x Intel Xeon Silver 4110



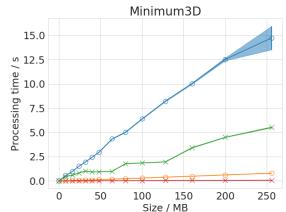
Intel Core i7-8650U



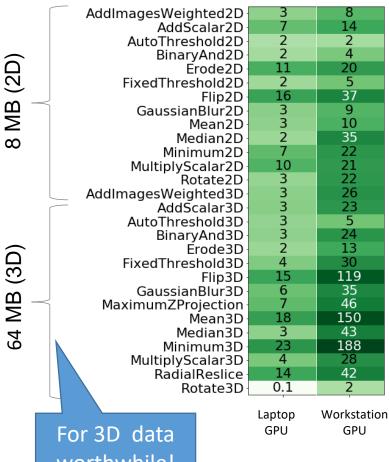
Nvidia Quadro P6000



Intel UHD 620 GPU



#### Speedup factor compared to Laptop CPU



worthwhile!

## Why GPU-acceleration?



1-2 orders of magnitude speedup are possible

```
# convolve with scikit-image
result_image = None

for i in range(0, 10):
    start_time = time.time()
    result_image = filters.gaussian(test_image, output=result_image, sigma=sigma)
    print("skimage Gaussian duration: " + str(time.time() - start_time))
```

scikit-image

```
skimage Gaussian duration: 0.644662618637085 skimage Gaussian duration: 0.63631272315979 skimage Gaussian duration: 0.6193966865539551 skimage Gaussian duration: 0.6499156951904297 skimage Gaussian duration: 0.6301307678222656 skimage Gaussian duration: 0.6531178951263428 skimage Gaussian duration: 0.6489198207855225 skimage Gaussian duration: 0.6308994293212891 skimage Gaussian duration: 0.7410404682159424 skimage Gaussian duration: 0.8148434162139893
```

clesperanto

```
# convolve with pyclesperanto
result_image_gpu = None

test_image_gpu = cle.push(test_image)

for i in range(0, 10):
    start_time = time.time()
    result_image_gpu = cle.gaussian_blur(test_image_gpu, result_image_gpu, sigma_x=sigma, sigma_y=sigma, sigma_z=sigma)
    print("pyclesperanto Gaussian duration: " + str(time.time() - start_time))
```

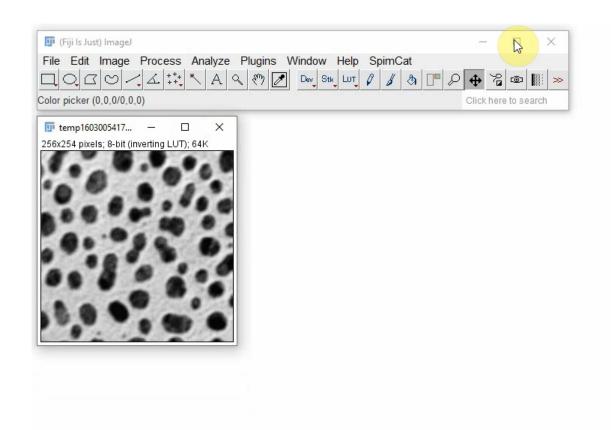
```
pyclesperanto Gaussian duration: 0.026170730590820312 pyclesperanto Gaussian duration: 0.002056121826171875 pyclesperanto Gaussian duration: 0.015659093856811523 pyclesperanto Gaussian duration: 0.019225597381591797 pyclesperanto Gaussian duration: 0.01566314697265625 pyclesperanto Gaussian duration: 0.015616178512573242 pyclesperanto Gaussian duration: 0.01566910743713379 pyclesperanto Gaussian duration: 0.015576839447021484 pyclesperanto Gaussian duration: 0.01562190055847168 pyclesperanto Gaussian duration: 0.023794889450073242
```

https://nbviewer.jupyter.org/github/BiAPoL/Bioimage Analysis with Python/blob/main/gpu acceleration/ 03 why GPU acceleration.ipynb

## How to GPU-accelerate workflows



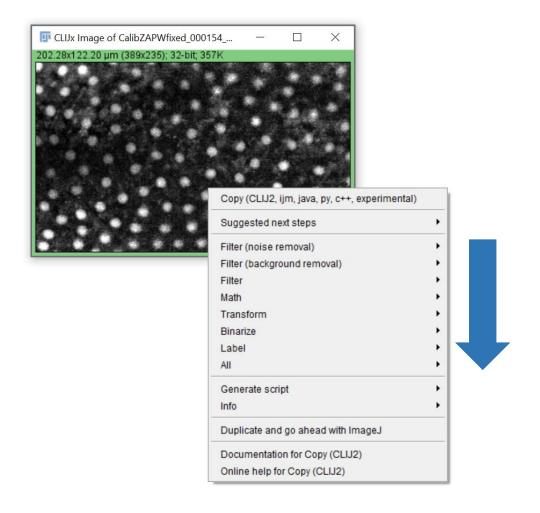
• In ImageJ / Fiji: CLIJ

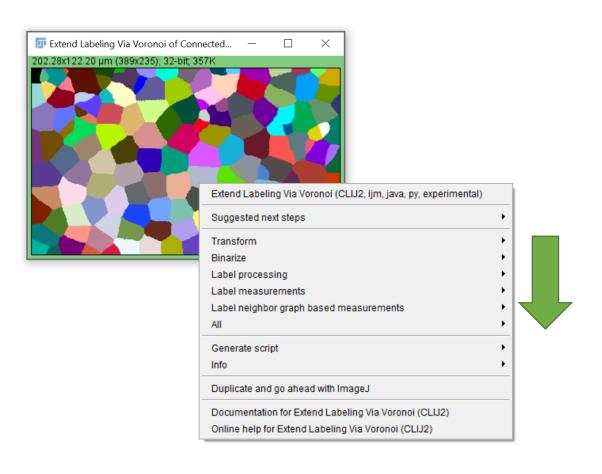


# The right-click menu in CLIJ



• The menu order is intentional: From preprocessing to analysis

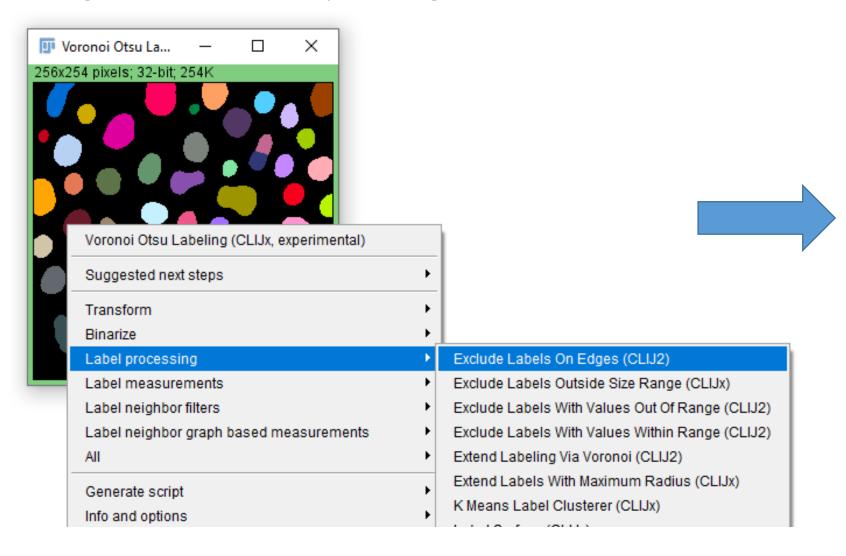




# Post-processing label images: CLIJ



[Right-click menu]: Label processing





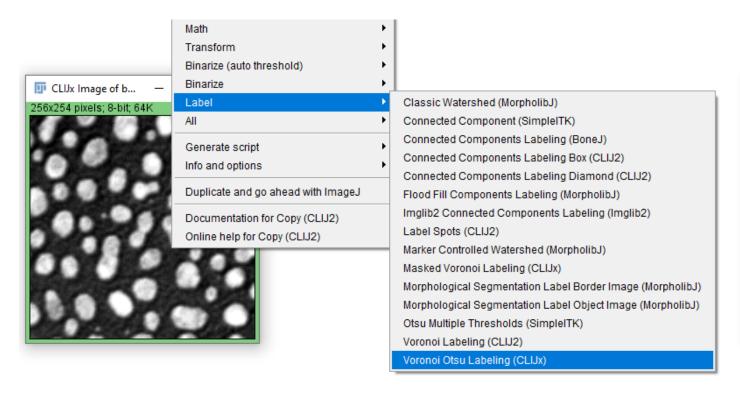
@haesleinhuepf

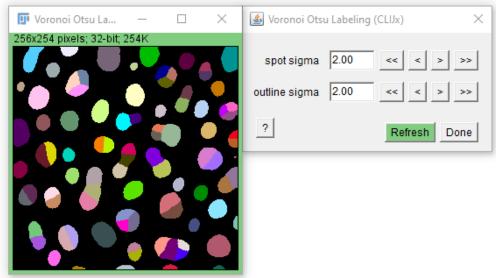
## Short-cuts: Labeling directly from grey-value images: CLIJ



GPU-accelerated image processing...

#### ... is also just image processing

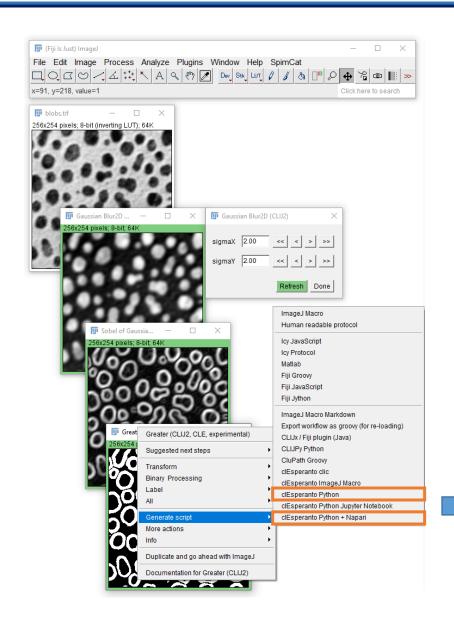




## Code generators in CLIJ



- Right-click > Generate Script > clEsperanto Python
- Copy & Paste it to your Jupyter Notebook
- If you want to run it from Fiji, follow these instructions: https://clij.github.io/assistant /installation#te oki

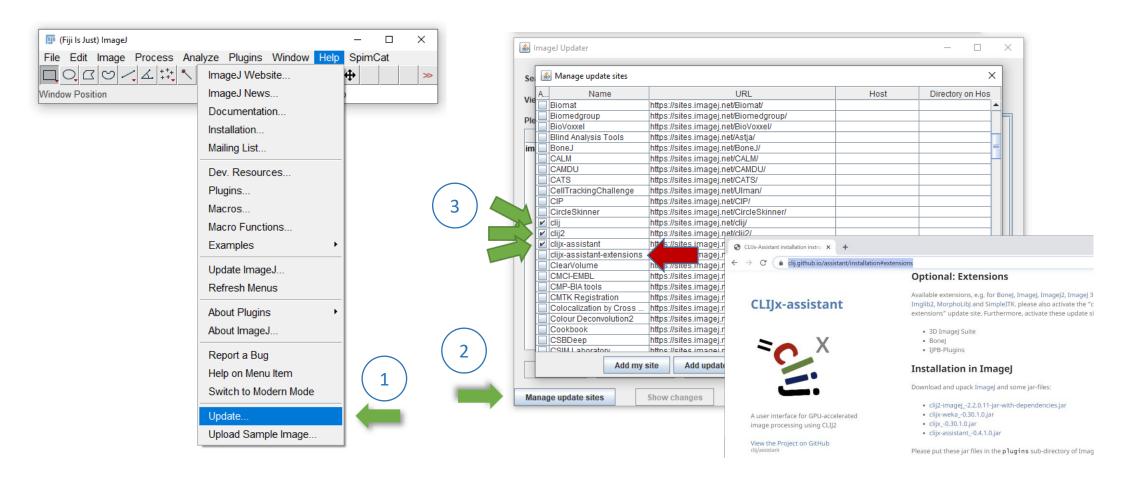




### **CLIJ: Installation**



- Just activate the CLIJ update sites, in menu Help > Update...
- Linux users: you may need to install OpenCL, e.g. apt-get install ocl-icd-devel



## clEsperanto: GPU-accelerated image processing in Python



The CLIJ/Fiji counter-part in python/napari

```
conda install -c conda-forge pyopencl
pip install pyclesperanto-prototype
```

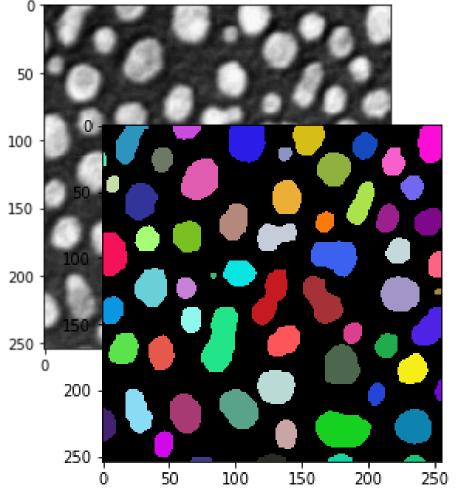
Napari user interface

pip install napari-pyclesperanto-assistant

# The cle gateway



```
from skimage.io import imread, imshow
image = imread("blobs.tif")
imshow(image)
import pyclesperanto prototype as cle
# noise removal
blurred = cle gaussian_blur(image, sigma_x=1, sigma_y=1)
# binarization
binary = cle.threshold_otsu(blurred)
# labeling
labels = cle.connected_components_labeling_box(binary)
# visualize results
cle imshow(labels, labels=True)
```



## The cle gateway



• ... allows you to explore available functions for given purposes

```
cle.operations('denoise').keys()
dict keys(['gaussian blur', 'mean box', 'mean sphere'])
cle.operations('background removal').keys()
dict keys(['bottom hat box', 'bottom hat sphere', 'difference of gaussian', 'divide by gaussian background', 'subtract gaussian
background', 'top hat box', 'top hat sphere'])
cle.operations('binarize').keys()
dict keys(['detect label edges', 'detect maxima box', 'detect minima box', 'equal', 'equal constant', 'greater', 'greater const
ant', 'greater_or_equal', 'greater_or_equal_constant', 'label_to_mask', 'local_threshold', 'not_equal', 'not_equal_constant',
'smaller', 'smaller constant', 'smaller or equal', 'smaller or equal constant', 'threshold', 'threshold otsu'])
cle.operations('label').keys()
dict keys(['connected components labeling box', 'connected components labeling diamond', 'label spots', 'masked voronoi labelin
g', 'voronoi labeling', 'voronoi otsu labeling'])
```



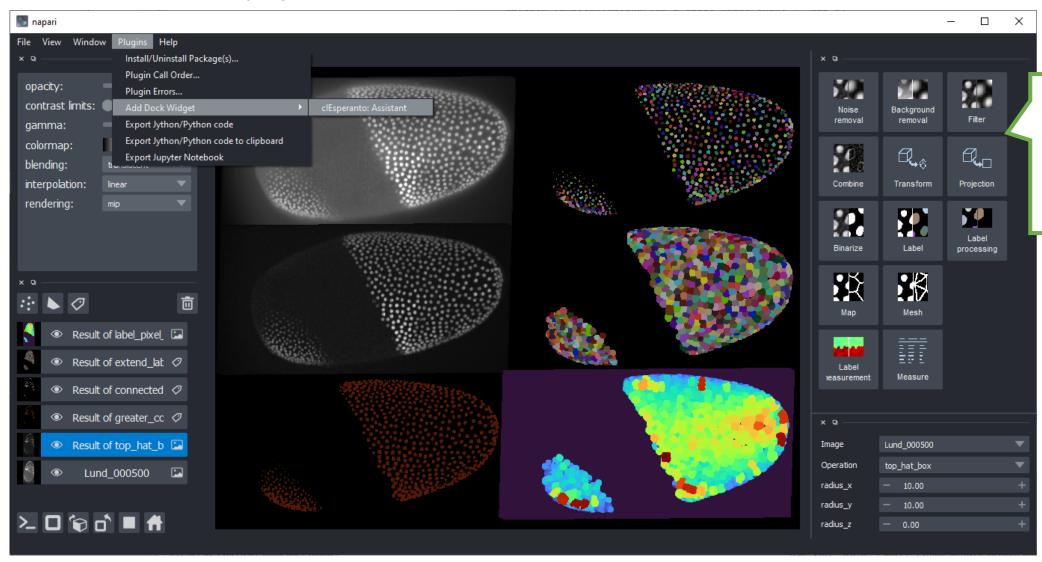
... allows you to read the documentation

```
print(cle.voronoi_otsu_labeling.__doc__)
  Applies two Gaussian blurs, spot detection, Otsu-thresholding and Voronoi-labeling.
      The thresholded binary image is flooded using the Voronoi approach starting from the found local maxima.
      Noise-removal sigma for spot detection and thresholding can be configured separately.
      Parameters
      source : Image
      label_image_destination : Image
      spot_sigma : float
      outline_sigma : float
      Returns
      label_image_destination
      Examples
      >>> import pyclesperanto prototype as cle
      >>> cle.voronoi_otsu_labeling(source, label_image_destination, 10, 2)
      References
      .. [1] https://clij.github.io/clij2-docs/reference voronoiOtsuLabeling
```

## Why GPU-acceleration?



... because it renders projects feasible which were not without GPUs



Lab / Bachelor / Master Projects available here. Get in touch!





# Quantitative measurements

Robert Haase

With material from

Daniela Vorkel (Myers lab MPI CBG)

Cyrus Jin (Dana-Farber Cancer Institute - Harvard BCMP)

Federico M. Gasparoli (NIC@HMS)

Mauricio Rocha Martins, Norden Lab MPI CBG

David Legland (INRAE)

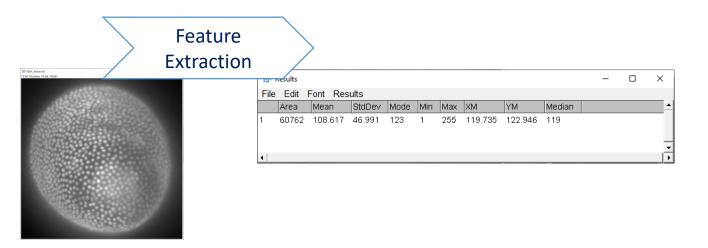
May 2021

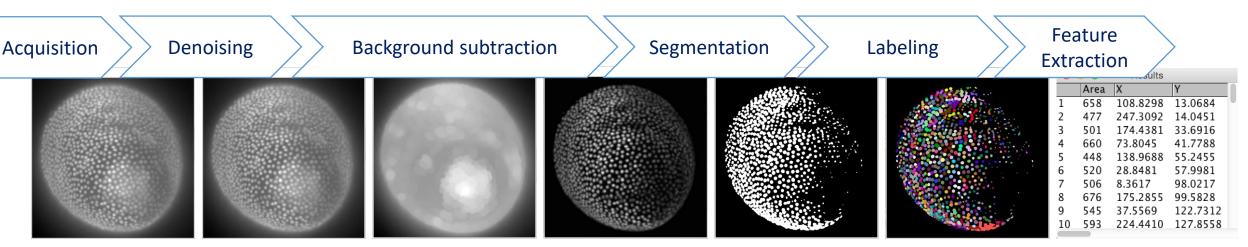


### Feature extraction



- Feature extraction a late processing step in image analysis.
- It can be used for images, or segmented/labelled images





### Feature extraction



- A feature is a countable or measurable property of an image or object.
- Goal of feature extraction is finding a minimal set of features to describe an object well enough to differentiate it from other objects.
- Intensity based features
  - Mean intensity
  - Standard deviation
  - Total intensity
  - Textures
  - ...

- Shape based / spatial features
  - Area / Volume
  - Roundness
  - Solidity
  - Circularity / Sphericity
  - Elongation
  - Centroid
  - Bounding box
  - •
- Mixed features
  - Center of mass
  - Local minima / maxima

- Spatio-temporal features
  - Displacement,
  - Speed,
  - Acceleration,
  - •

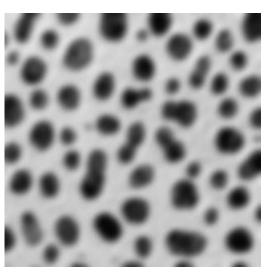
- Others
  - Overlap
  - Colocalisation
  - Networkanalysis
  - •

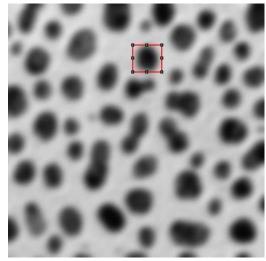
## Intensity based features

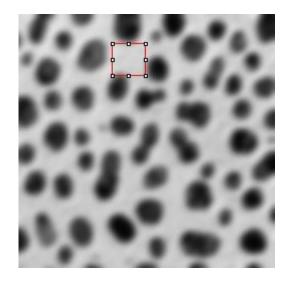


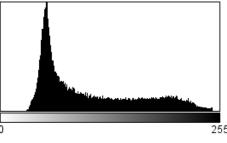
- Min / max
- Median
- Mean
- Mode
- Variance
- Standard deviation

- Can be derived from pixel values
- Don't take spatial relationship of pixels into account
- See also:
  - descriptive statistics
  - histogram



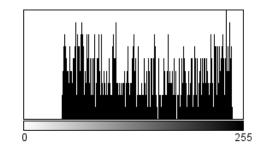




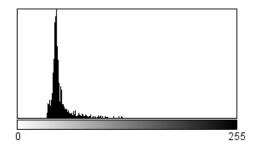


Count: 65024 Mean: 103.301 StdDev: 57.991

Min: 29 Max: 248 Mode: 53 (1663)



Count: 783 Mean: 141.308 StdDev: 61.876 Min: 44 Max: 243 Mode: 236 (9)



Count: 1056 Mean: 49.016 StdDev: 12.685 Min: 34 Max: 122 Mode: 45 (120)

## Center of mass



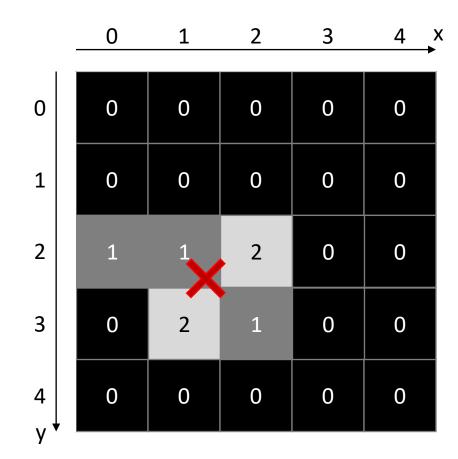
- Relative position in an image weighted by pixel intensities
  - x, y ... pixel coordinates
  - w ... image width
  - h ... image height
  - μ ... mean intensity
  - g<sub>x,y</sub> ... pixel grey value
  - $x_m$ ,  $y_m$  ... center of mass coordinates

$$\mu = \frac{1}{wh} \sum_{y=0}^{h-1} \sum_{x=0}^{w-1} g_{x,y}$$

$$x_m = \frac{1}{wh\mu} \sum_{y=0}^{h-1} \sum_{x=0}^{w-1} x \ g_{x,y}$$

$$y_m = \sum_{wh\mu} \sum_{y=0}^{h-1} \sum_{x=0}^{w-1} y \ g_{x,y}$$

"sum intensity"
"total intensity"



$$x_m = 1/7 (1.0 + 1.1 + 2.2 + 2.1 + 1.2) = 1.3$$

$$y_m = 1/7 (1.2 + 1.2 + 2.3 + 2.2 + 1.3) = 2.4$$

## Center of geometry / centroid



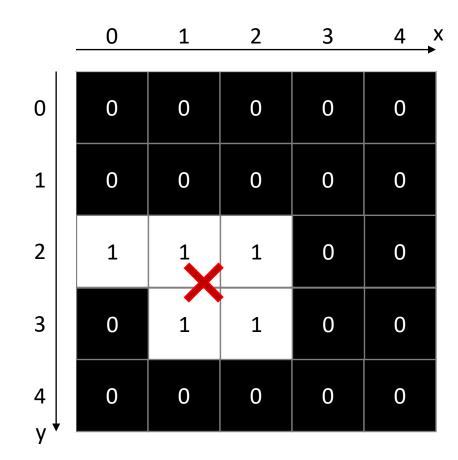
- Relative position in an image weighted by pixel intensities
- Special case of center of mass for binary images
  - x, y ... pixel coordinates
  - w ... image width
  - h ... image height
  - μ ... mean intensity
  - $g_{x,y}$  ... pixel grey value, integer in range [0;1]
  - $x_m$ ,  $y_m$  ... center of mass coordinates

$$\mu = \frac{1}{wh} \sum_{v=0}^{h-1} \sum_{x=0}^{w-1} g_{x,y}$$

$$x_m = \frac{1}{wh\mu} \sum_{v=0}^{h-1} \sum_{x=0}^{w-1} x \ g_{x,y}$$

$$y_m = \sum_{wh\mu} \sum_{y=0}^{h-1} \sum_{x=0}^{w-1} y \, g_{x,y}$$

Number of white pixels



$$x_m = 1/5 (1.0 + 1.1 + 1.2 + 1.1 + 1.2) = 1.2$$

$$y_m = 1/5 (1.2 + 1.2 + 1.3 + 1.2 + 1.3) = 2.4$$

## Bounding rectangle / bounding box



- Position and size of the smallest rectangle containing all pixels of an object
  - $x_b$ ,  $y_b$  ... position of the bounding box
  - w<sub>h</sub> ... width of the bounding box
  - h<sub>b</sub> ... height of the bounding box

variable	value
$x_b$	0
$y_b$	2
$w_b$	3
h <sub>b</sub>	2

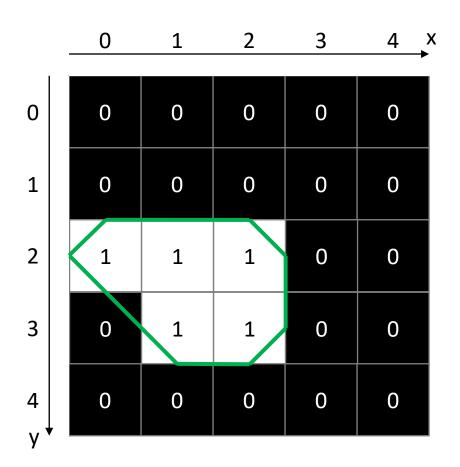
	0	1	2	3	4 ×
0	0	0	0	0	0
1	0	0	0	0	0
2	1	1	1	0	0
3	0	1	1	0	0
4 y	0	0	0	0	0

## Perimeter



- Length of the outline around an object
- Depends on the actual implementation

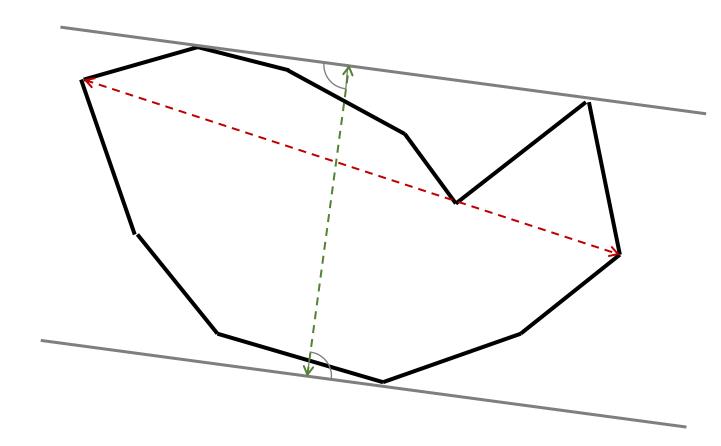
	0	1	2	3	4 X
0	0	0	0	0	0
1	0	0	0	0	0
2	1	1	1	0	0
3	0	1	1	0	0
4 y	0	0	0	0	0



## Feret's diameter



- Feret's diameter describes the maximum distance between any two points of an outline.
- The minimum caliper ("Minimum Feret") describes the shortest distance, the object would fit through.
- Feret and Minimum Feret do not need to be perpendicular to each other!

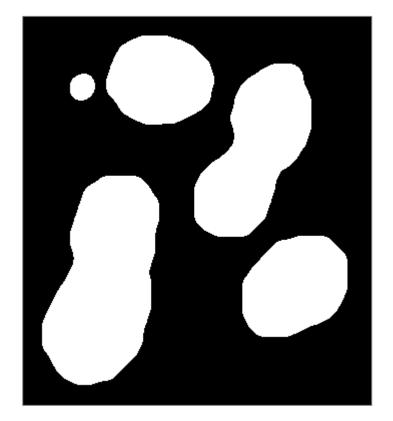


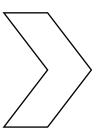
28

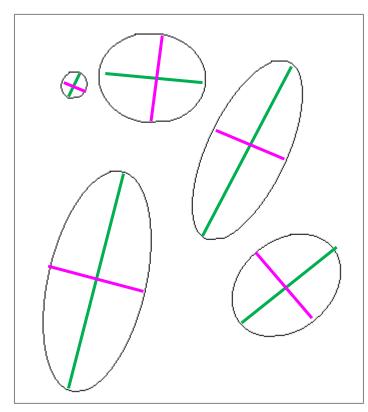
# Fit ellipse



- For every object, find the optimal ellipse simplifying the object.
- Major axis ... long diameter
- Minor axis ... short diameter
- Major and minor axis are perpendicular to each other







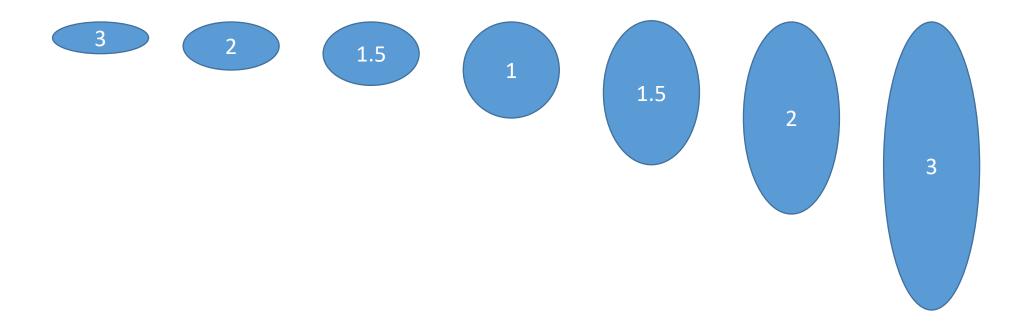
29

## Aspect ratio



• The aspect ratio describes the elongation of an object.

AR = major / minor



# Roundness and circularity

Pol Physics of Life TU Dresden

- The definition of a circle leads us to measurements of circularity and roundness.
- In case you use these measures, define them correctly. They are not standardized!

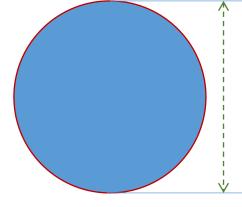
Diameter

d

Circumference

 $C = \pi d$ 

$$A = \frac{\pi d^2}{4}$$



$$roundness = \frac{4 * A}{\pi \; major^2}$$

$$circularity = \frac{4\pi * A}{perimeter^2}$$

Roundness = 1 Circularity = 1

Area

Roundness ≈ 1 Circularity ≈ 1 Roundness < 1 Circularity < 1

• In 3D: Circularity -> Sphericity

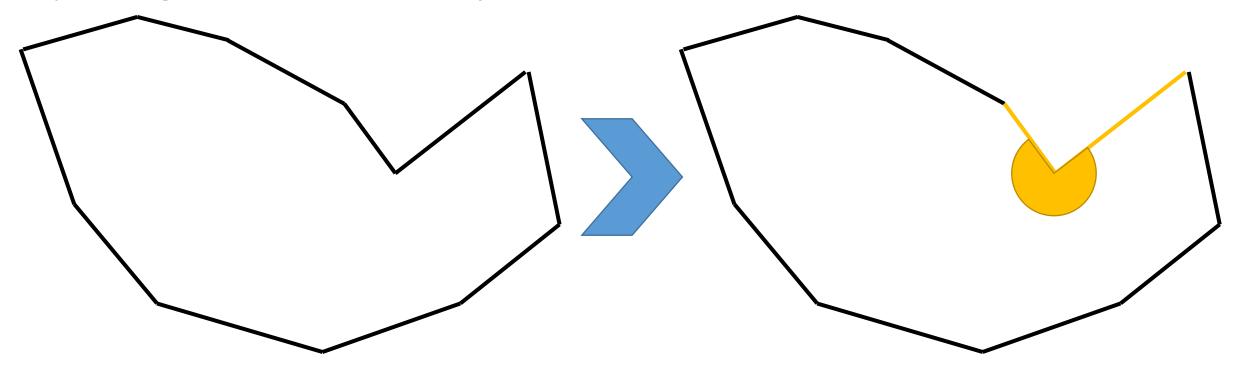


$$roundness = \frac{4 * A}{\pi \; major^2}$$

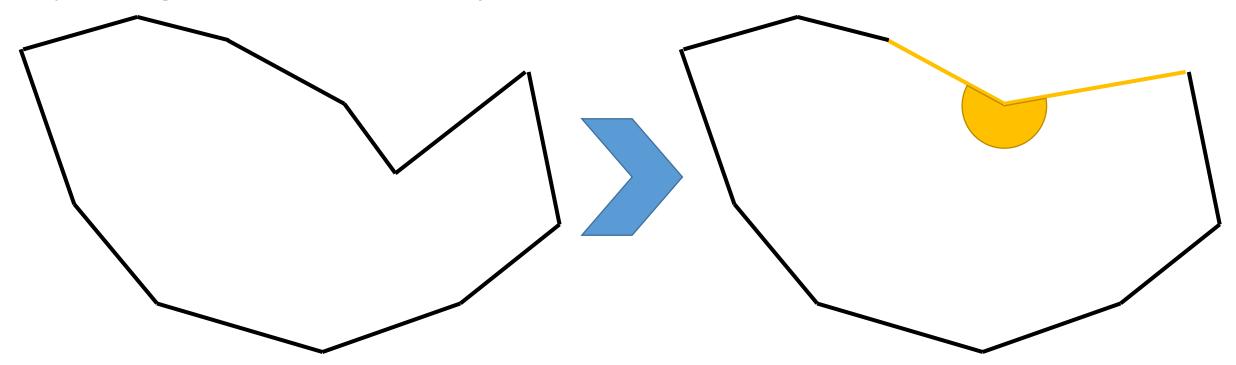
$$circularity = \frac{4\pi * A}{perimeter^2}$$

## Convex hull



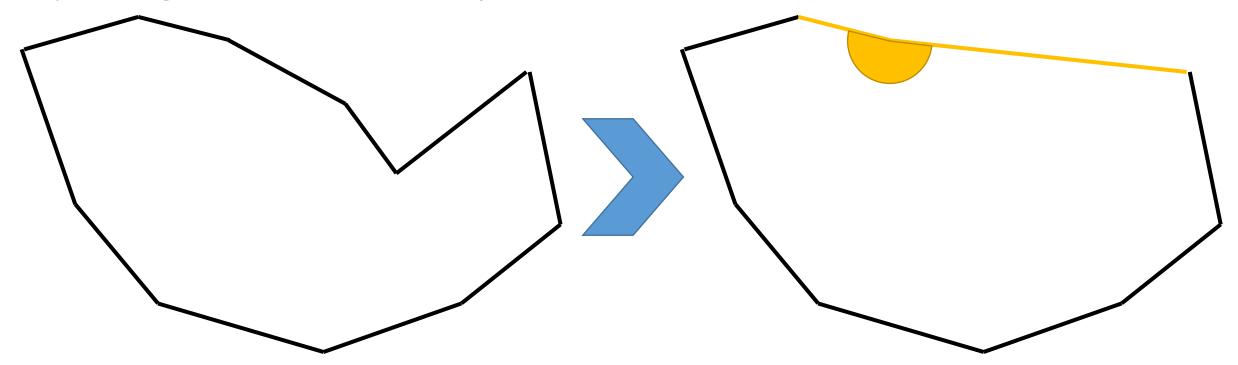






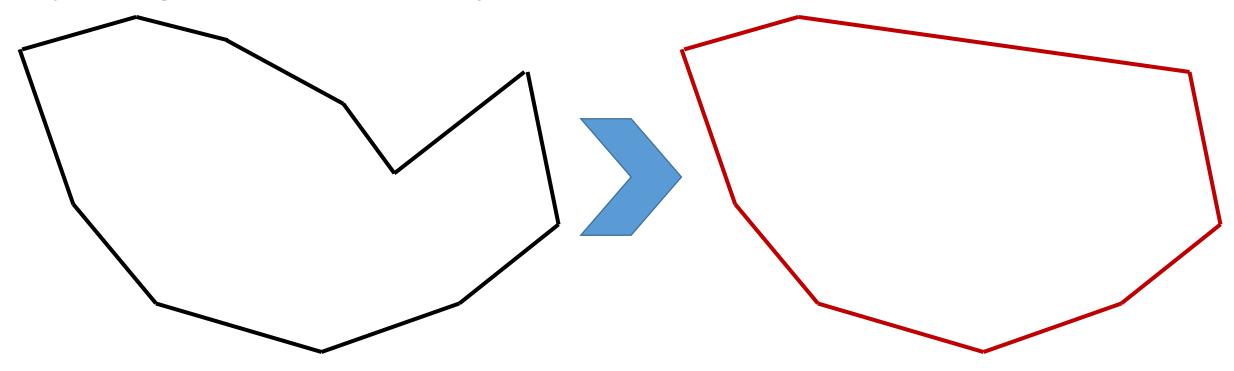
## Convex hull





## Convex hull



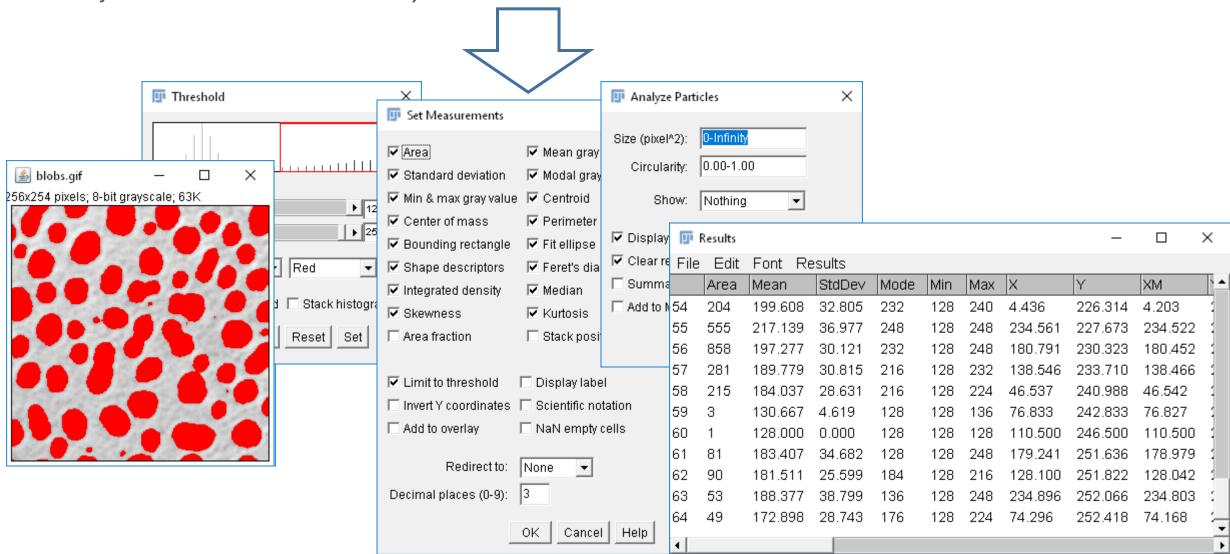


$$solidity = \frac{A}{A_{convexHull}}$$

## Measurements in 2D: The Particle Analyzer



• Tell Fiji what to measure with *Analyze > Set Measurements* 



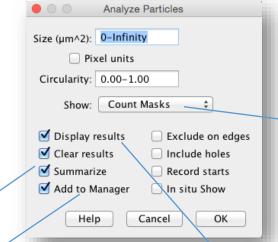
## Measurements in 2D: The Particle Analyzer

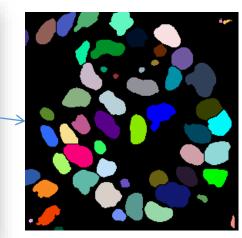


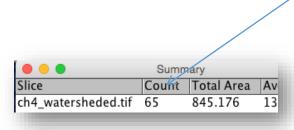
- The particle analyser
  - performs connected components labelling,
  - can make selections and
  - Measure properties of particles (feature extraction)

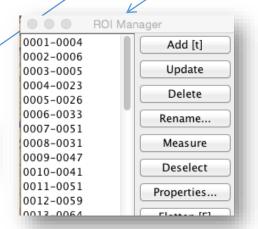
• Analyze > Analyse particles...

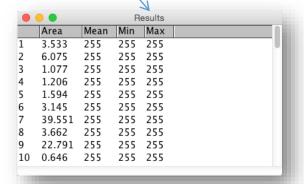








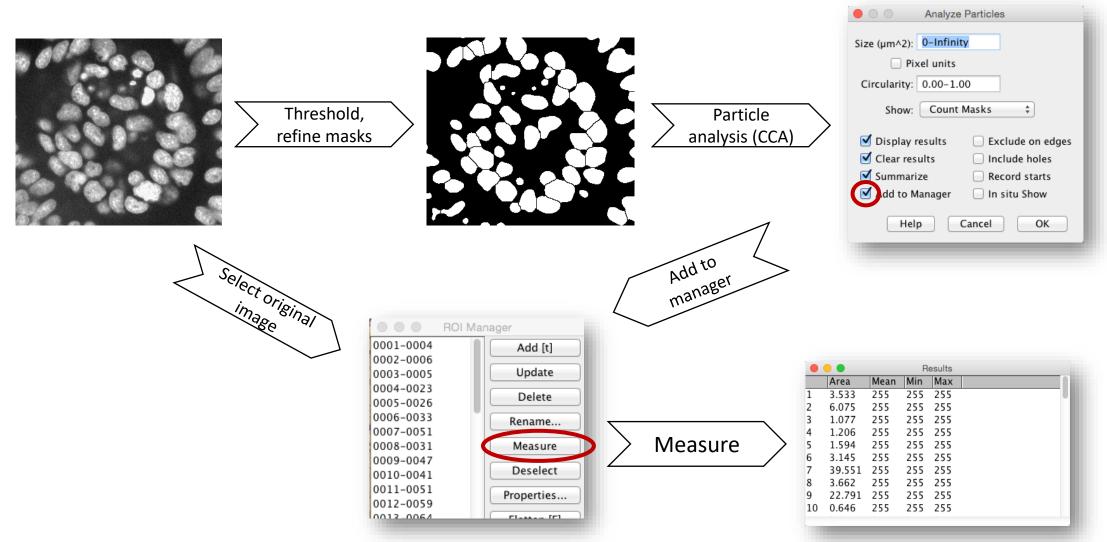




## Measurements in 2D: The Particle Analyzer



Suggested workflow



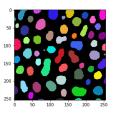
## Measurements with scikit-image: regionprops



#### Regionprops to collect measurements

from skimage import measure

We start with a label image



# analyse objects

properties = measure.regionprops (label\_image, intensity\_image=image)

Reorganize in a dictionary of lists

Visualize as table, a.k.a. pandas DataFrame

```
import pandas as pd
dataframe = pd DataFrame (statistics)
```

area	mean	major_axis	aspect_ratio
429	191.440559	34.779230	2.088249
183	179.846995	20.950530	1.782168
658	205.604863	30.198484	1.067734
433	217.515012	24.508791	1.061942
472	213.033898	31.084766	1.579415
213	184.525822	18.753879	1.296143
79	184.810127	18.287489	3.173540
88	182.727273	21.673692	4.021193
52	189.538462	14.335104	2.839825
48	173.833333	16.925660	4.417297
	429 183 658 433 472  213 79 88 52	429 191.440559 183 179.846995 658 205.604863 433 217.515012 472 213.033898 213 184.525822 79 184.810127 88 182.727273 52 189.538462	429       191.440559       34.779230         183       179.846995       20.950530         658       205.604863       30.198484         433       217.515012       24.508791         472       213.033898       31.084766              213       184.525822       18.753879         79       184.810127       18.287489         88       182.727273       21.673692         52       189.538462       14.335104

#### Save to disk

dataframe.to csv("blobs analysis.csv")

https://pandas.pydata.org/

## Summary

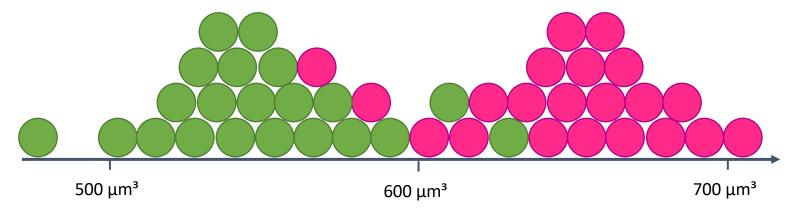


#### Today, you learned

- GPU-acceleration
- Quantitative measurements

#### Coming up next:

Introduction to Biostatistics with Anna Poetsch



Cell volume V



45