

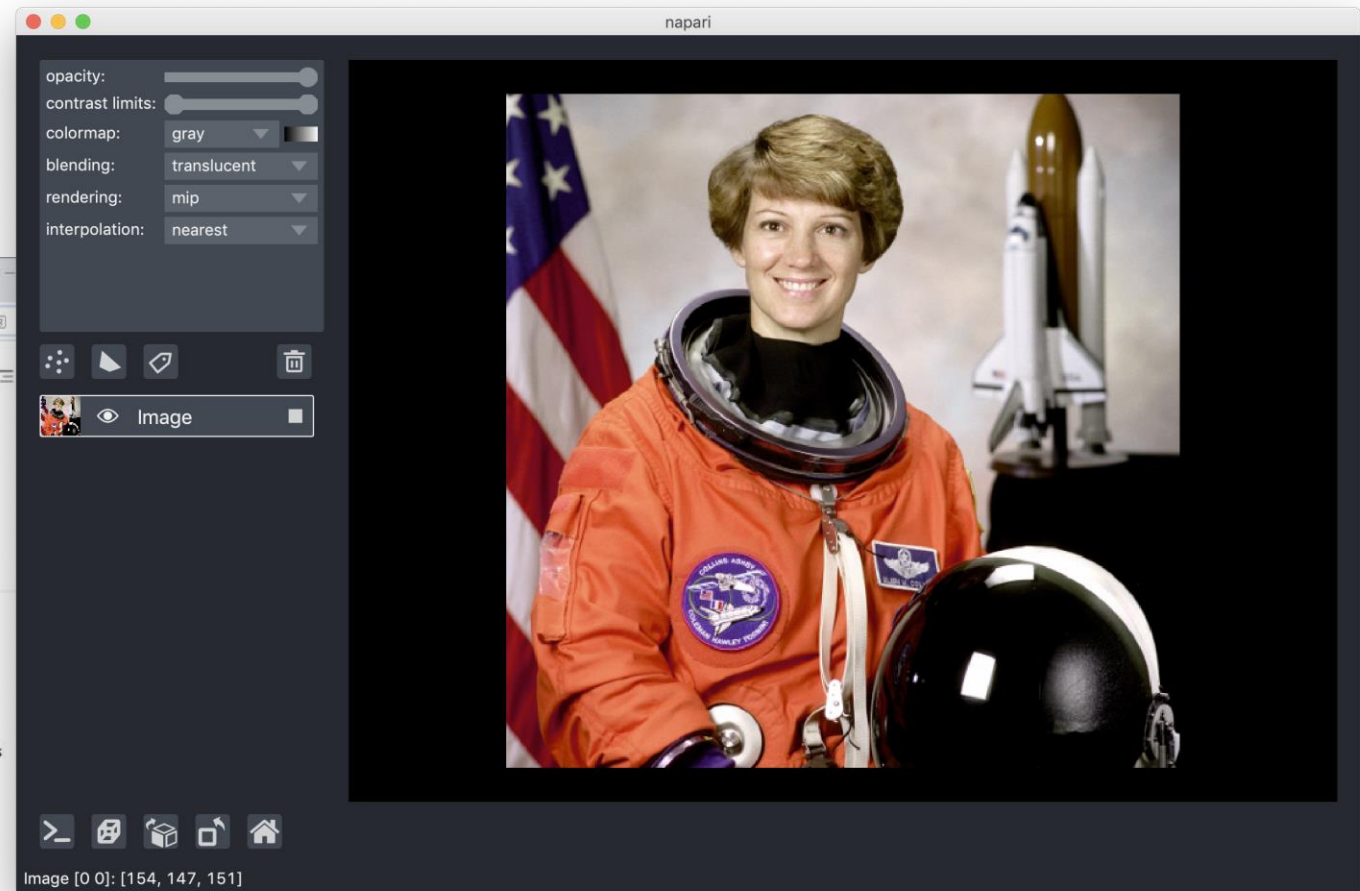
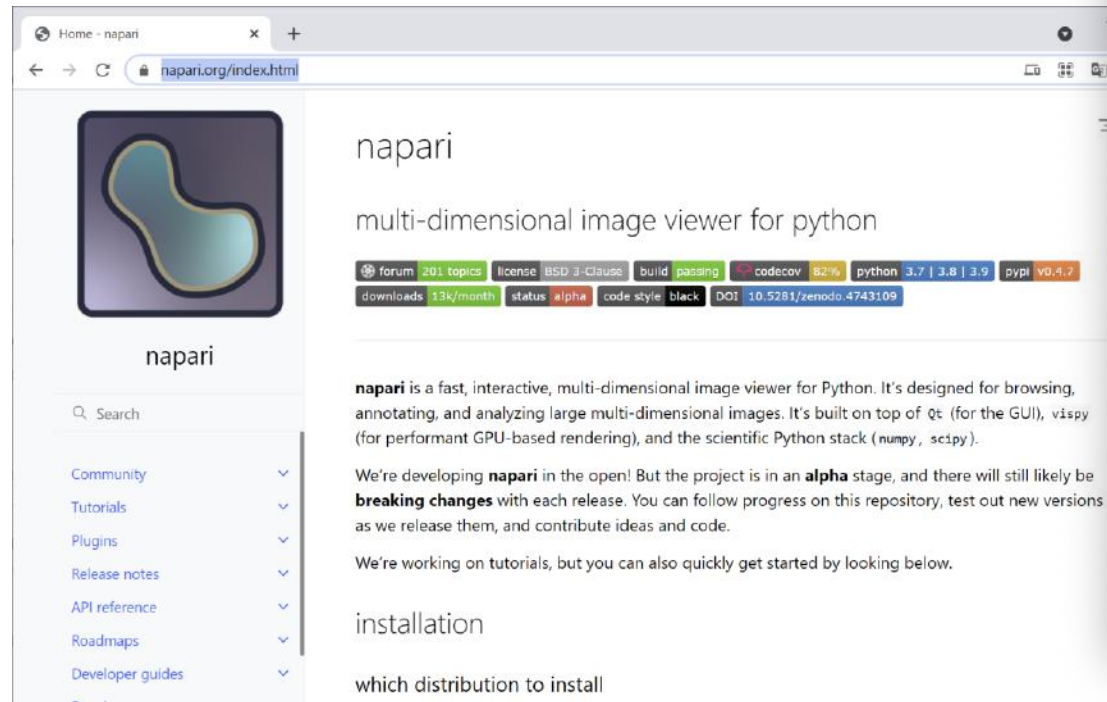
On-the-fly image processing with Python and napari

Robert Haase



Partly based on napari community materials <https://napari.org/>

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



napari user interface

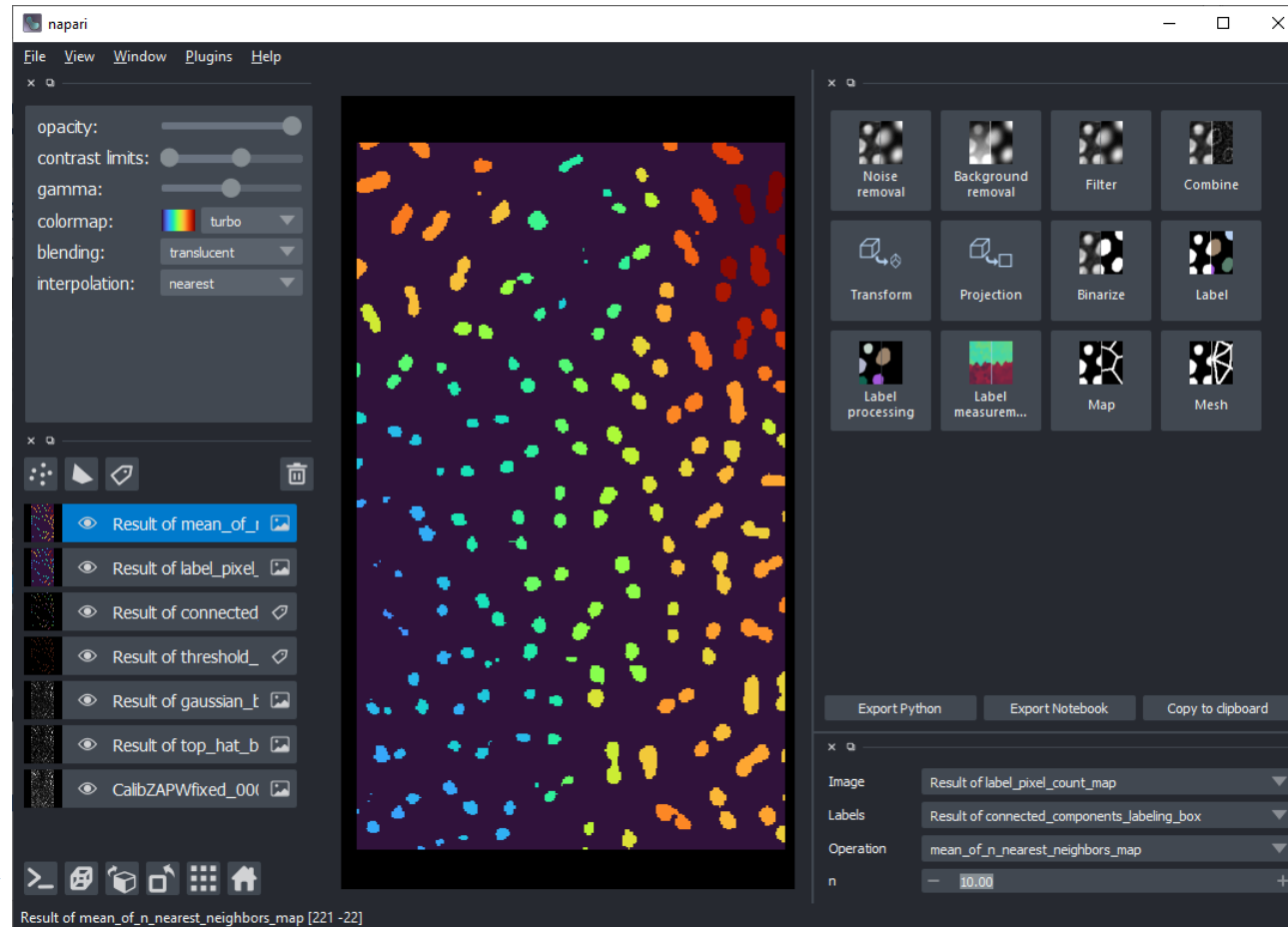
View configuration /
tools

```
layer.opacity = 0.5
```

Layers

```
layer.visible = False
```

Viewer controls

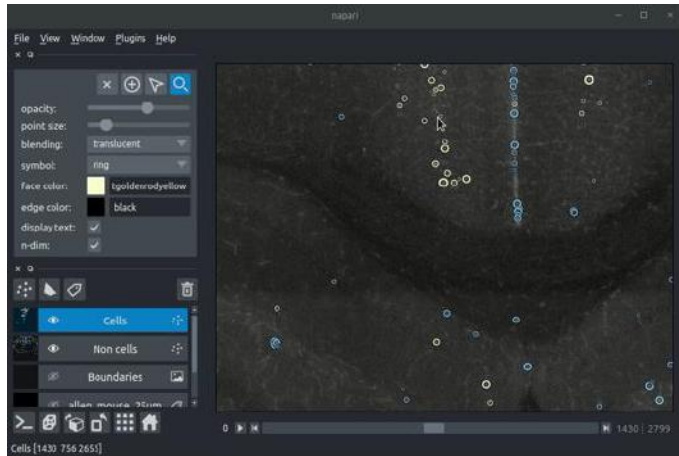


Dock widgets
(custom plugins)

Function widgets
(custom plugins)

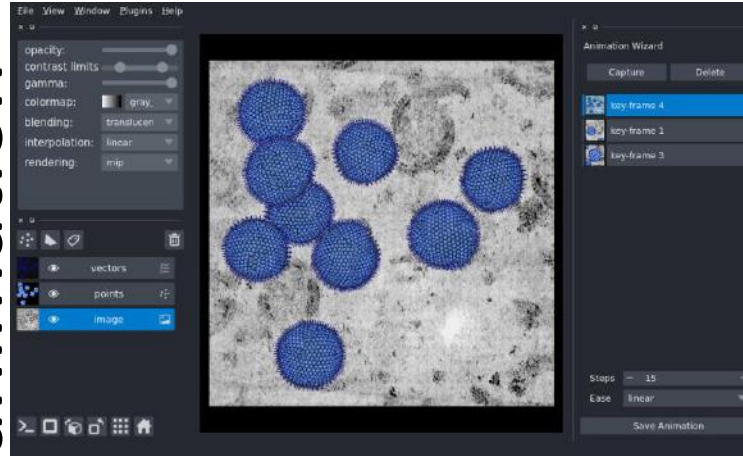
The era of napari plugins has just begun

cellfinder



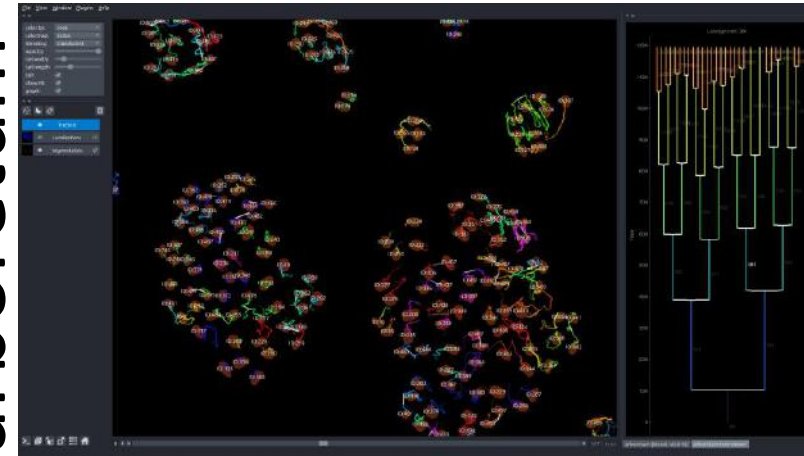
<https://github.com/brainlobe/napari-cellfinder>

animation



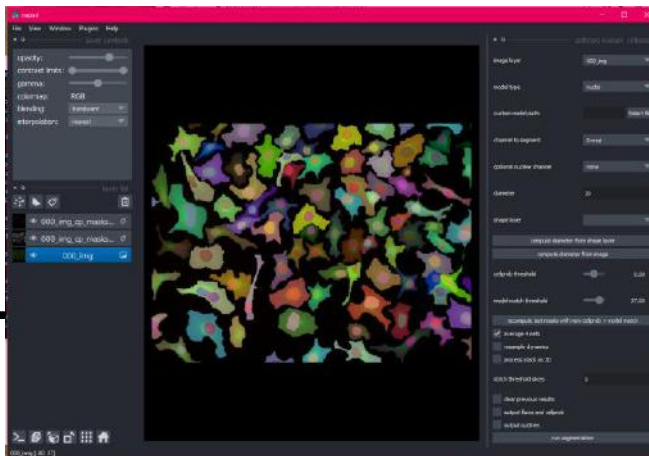
<https://github.com/napari/napari-animation>

arboretum



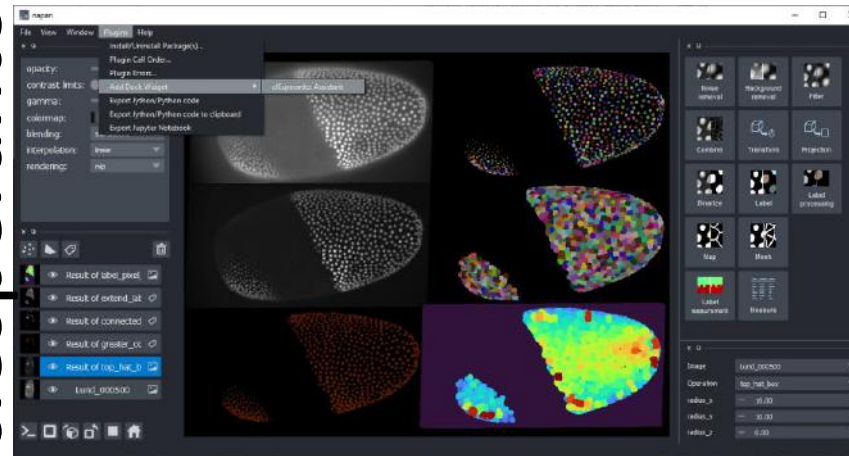
<https://github.com/quantumiot/arboretum>

cellpose



<https://cellpose-napari.readthedocs.io/en/latest/>

clesperanto



https://github.com/clEsperanto/napari_pyclesperanto_assistant

In development: <https://github.com/topics/napari-plugin>

Released: <https://pypi.org/search/?q=&o=&c=Framework+%3A%3A+napari>

Installation

- Install napari via pip

```
pip install napari[all]==0.4.8rc3
```

“all” -> default framework
Alternatively:
“pyqt5” or “pyside2”

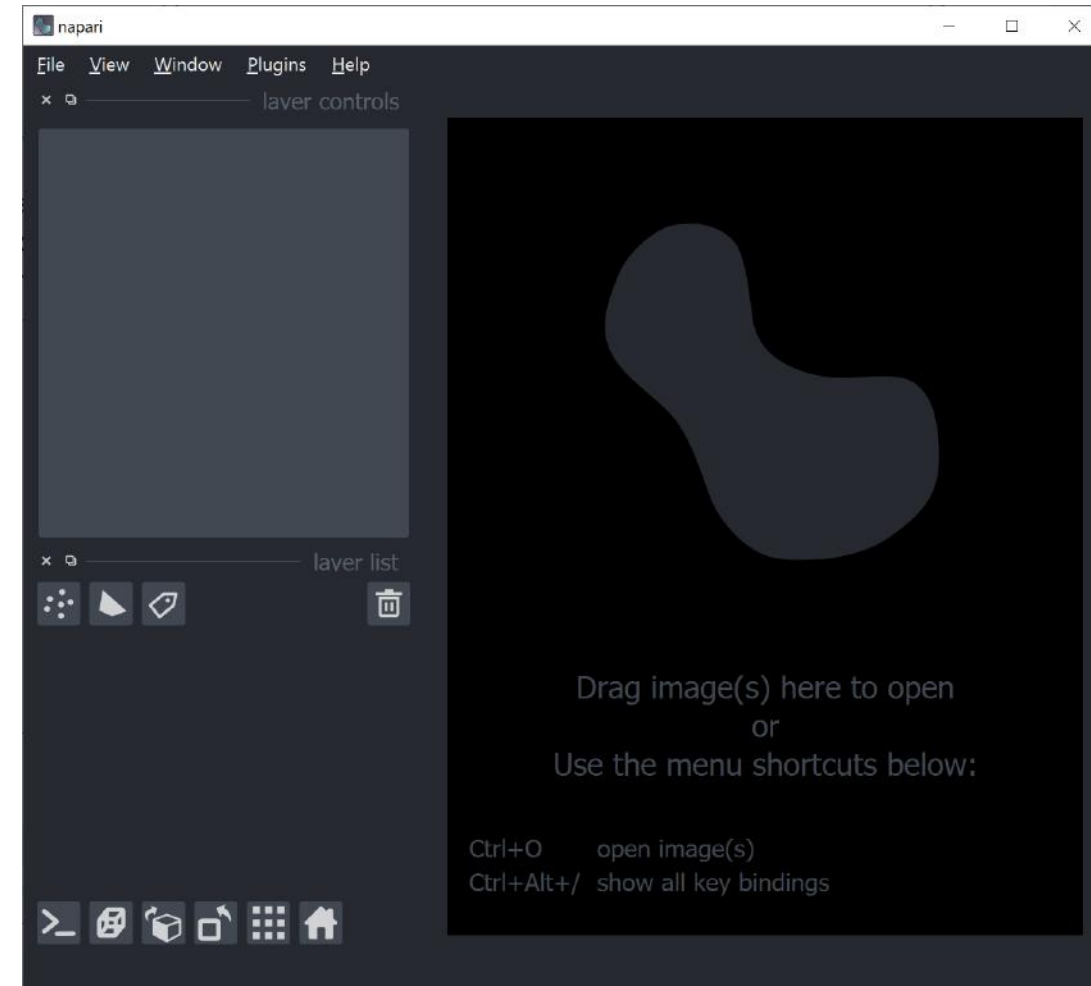
For today only ;-)

- Install napari via conda

```
conda install -c conda-forge napari
```

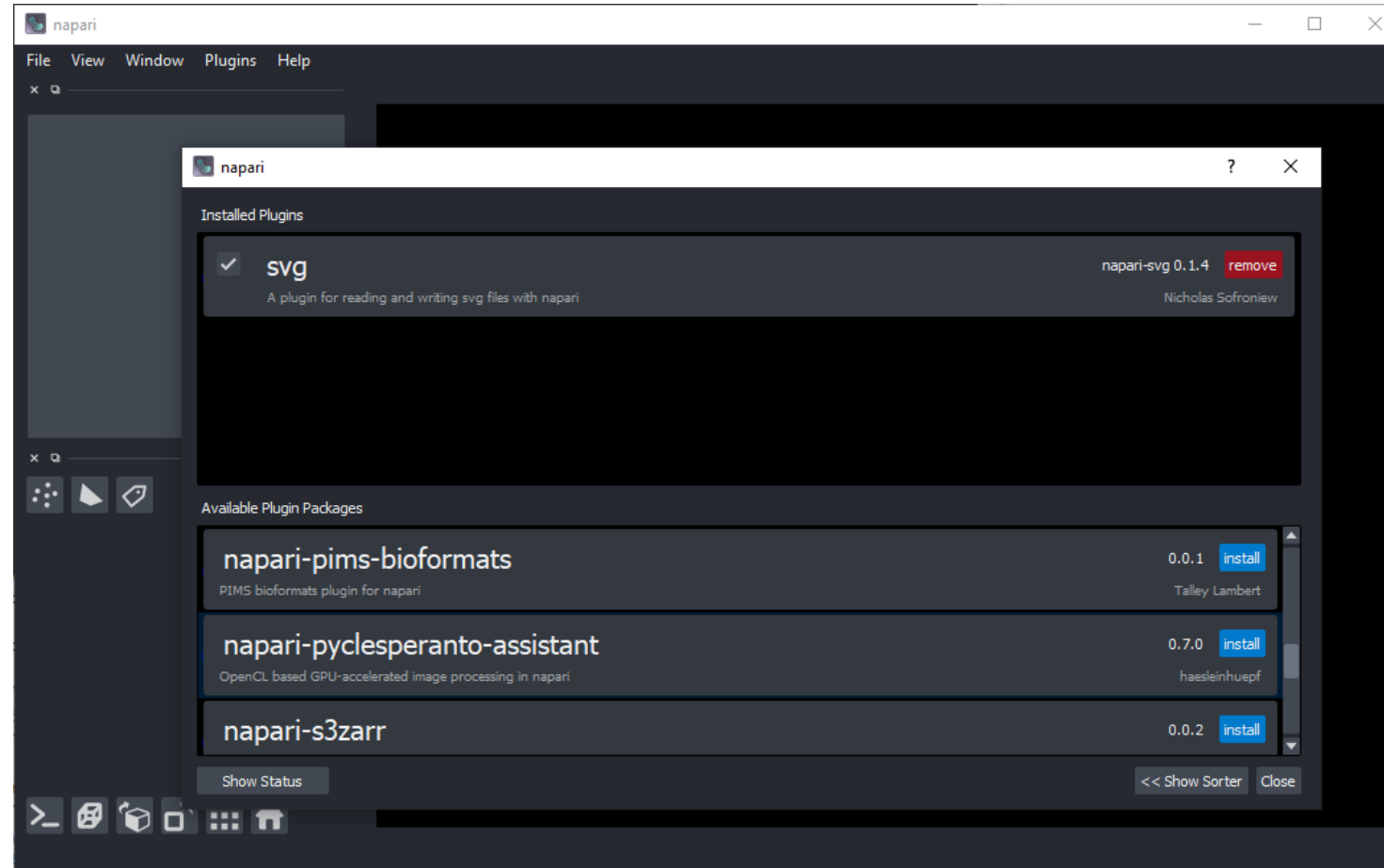
- Test if napari works

```
napari
```



napari plugins

- Install them via conda, pip or the menu *Plugins > Install / Uninstall packages...*



Scripting napari

- Initialization

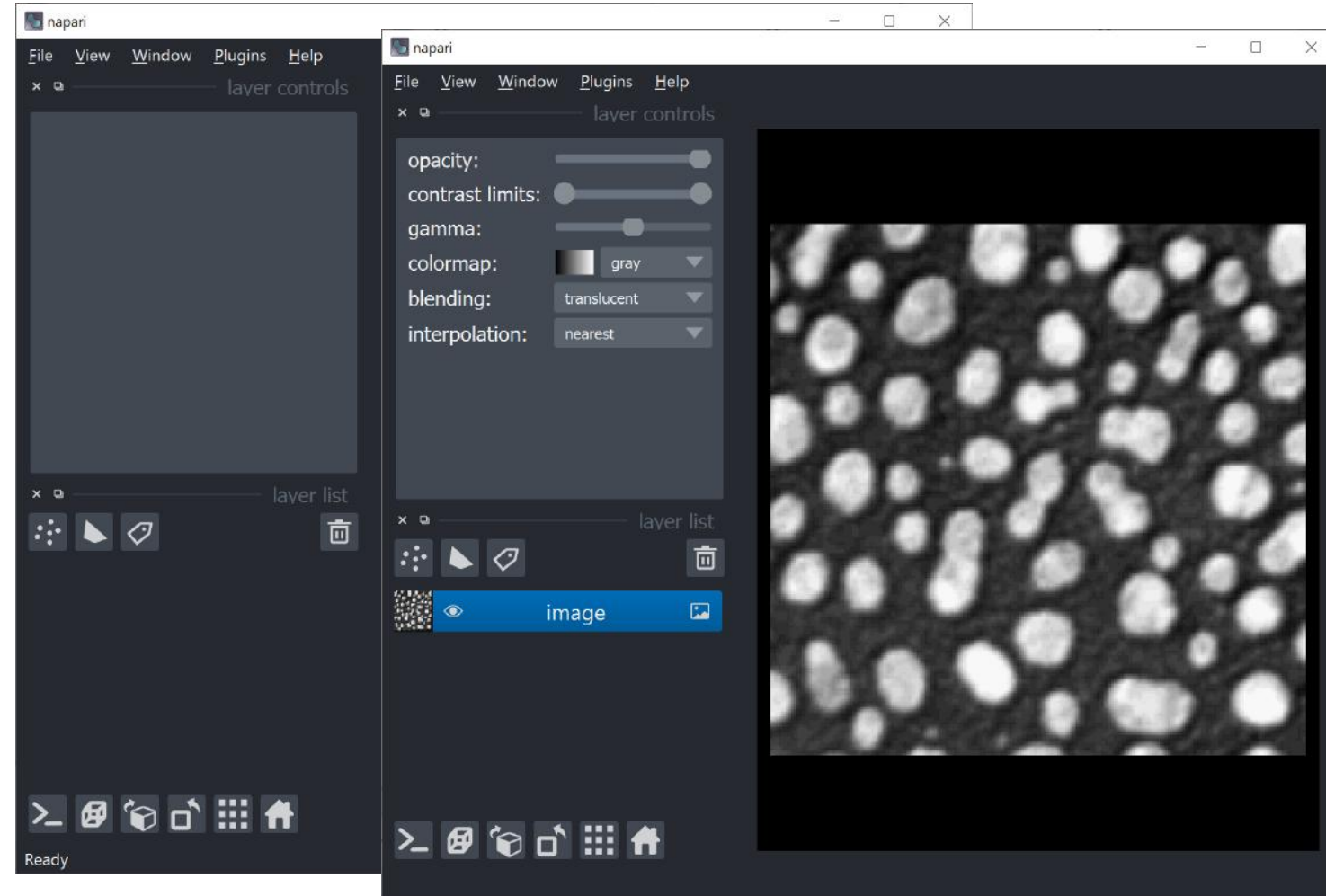
```
import napari

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

# Start it
napari.run()
```

- Adding images

```
viewer.add_image(image)
```

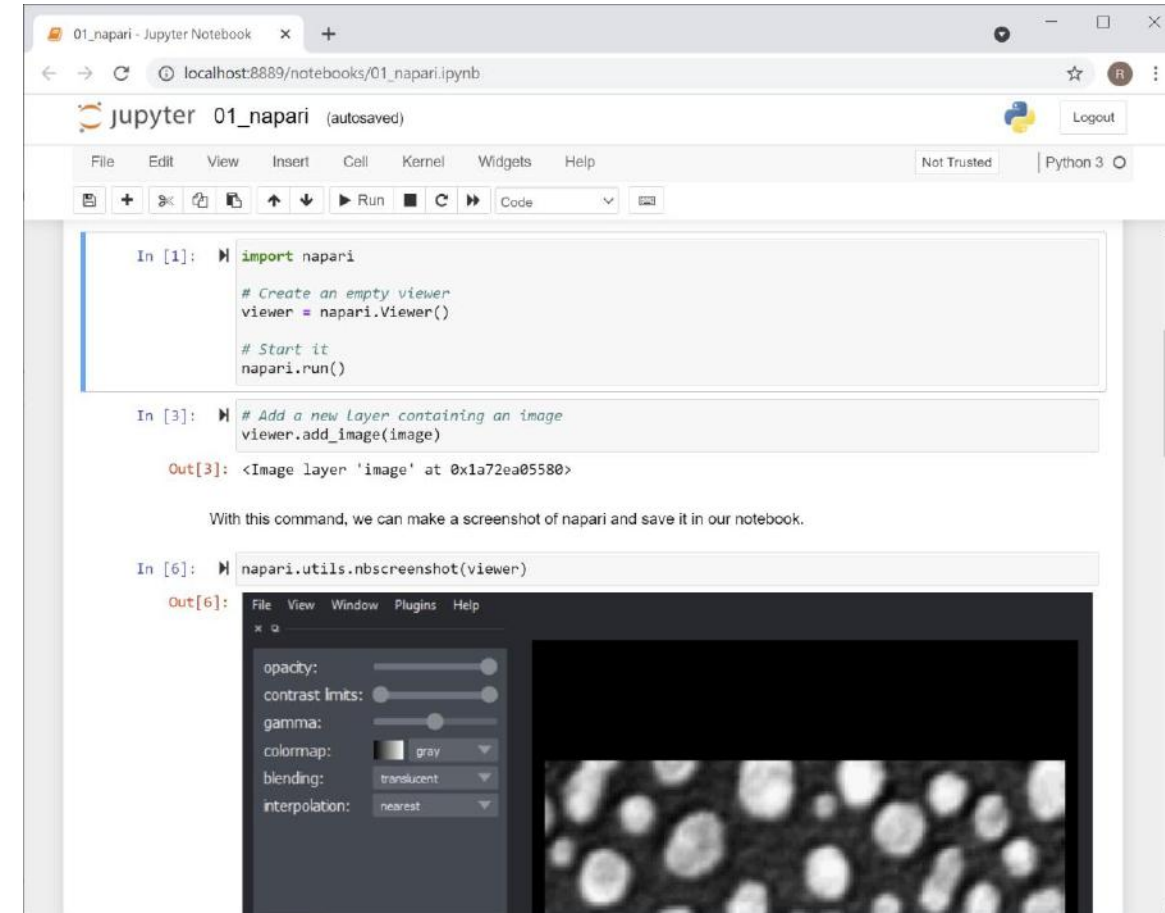


Scripting napari in notebooks

- Make screenshots from napari and put them in your jupyter notebook

```
napari.utils.nbscreenshot(viewer)
```

Place your viewer here



Working with layers

- Removing layers

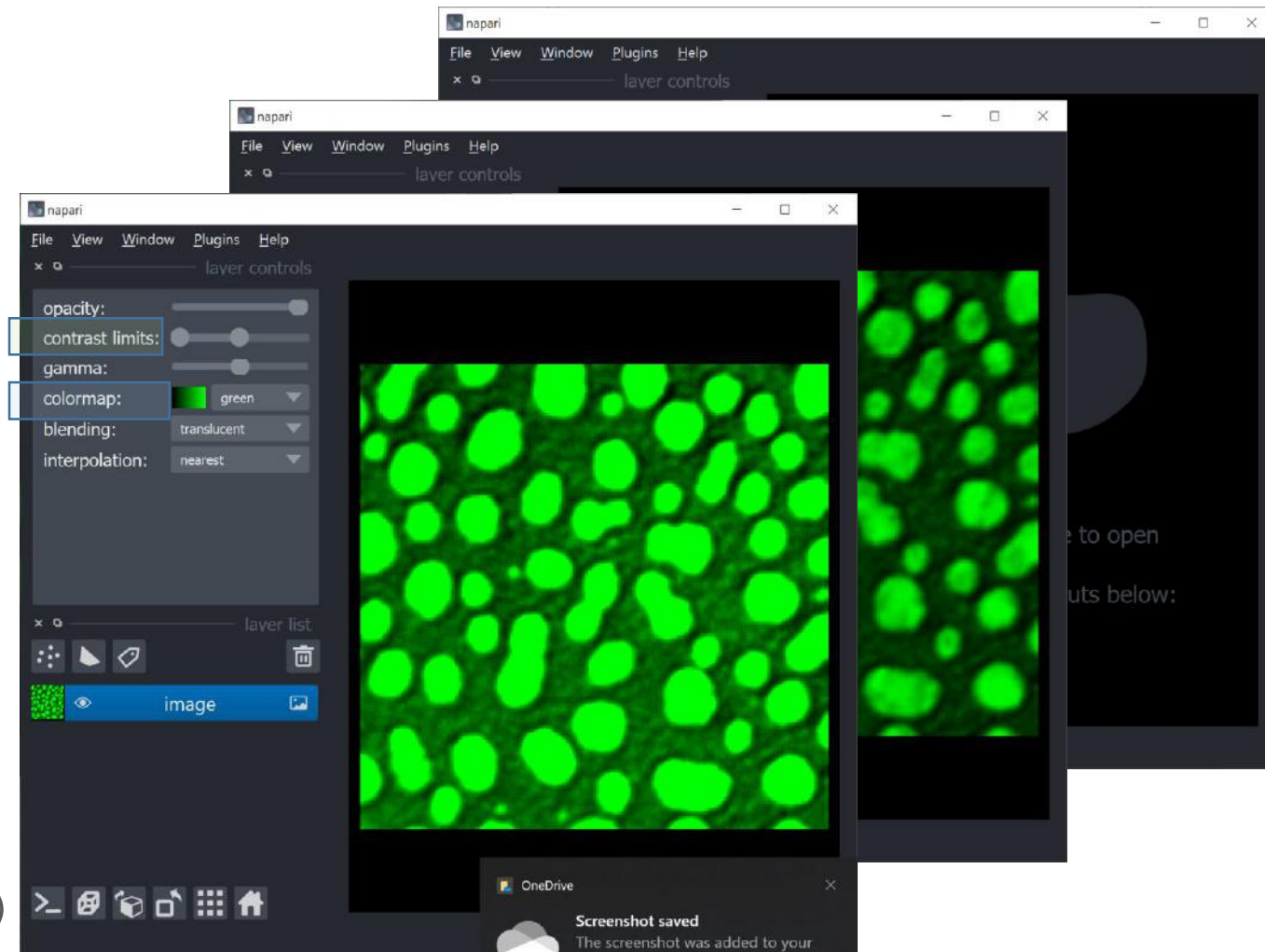
```
for l in viewer.layers:  
    viewer.layers.remove(l)
```

- Modify visualization while adding layers

```
viewer.add_image(image,  
                 colormap='green')
```

- Modify layers after adding

```
layer = viewer.add_image(image)  
layer.colormap = 'green'  
layer.contrast_limits = (0, 128)
```



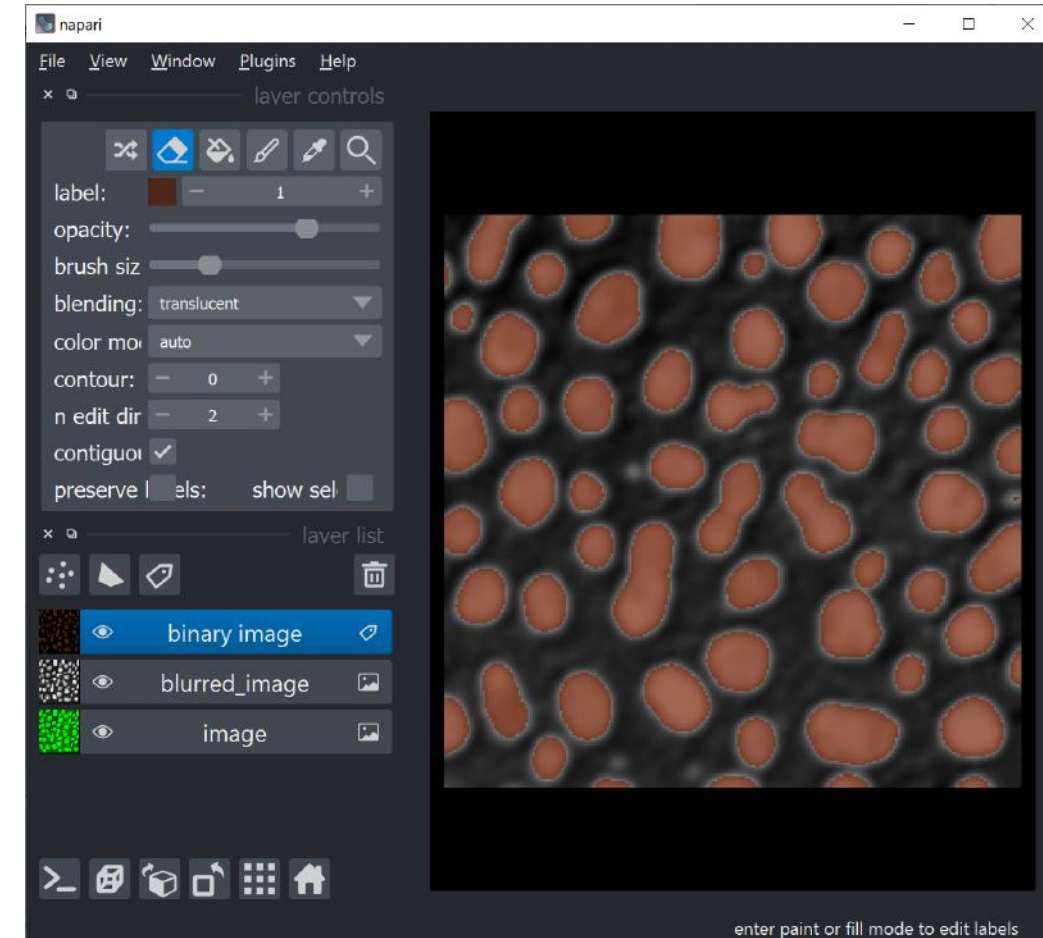
Visualizing image segmentation

- Binary images and **label** images visualized as label layers

```
from skimage.filters import threshold_otsu
threshold = threshold_otsu(blurred_image)
binary_image = blurred_image > threshold

# Add a new labels layer containing an image
viewer.add_labels(binary_image,
                  name="binary image")
```

Name your layers to keep track
of what they contain



Visualizing image segmentation

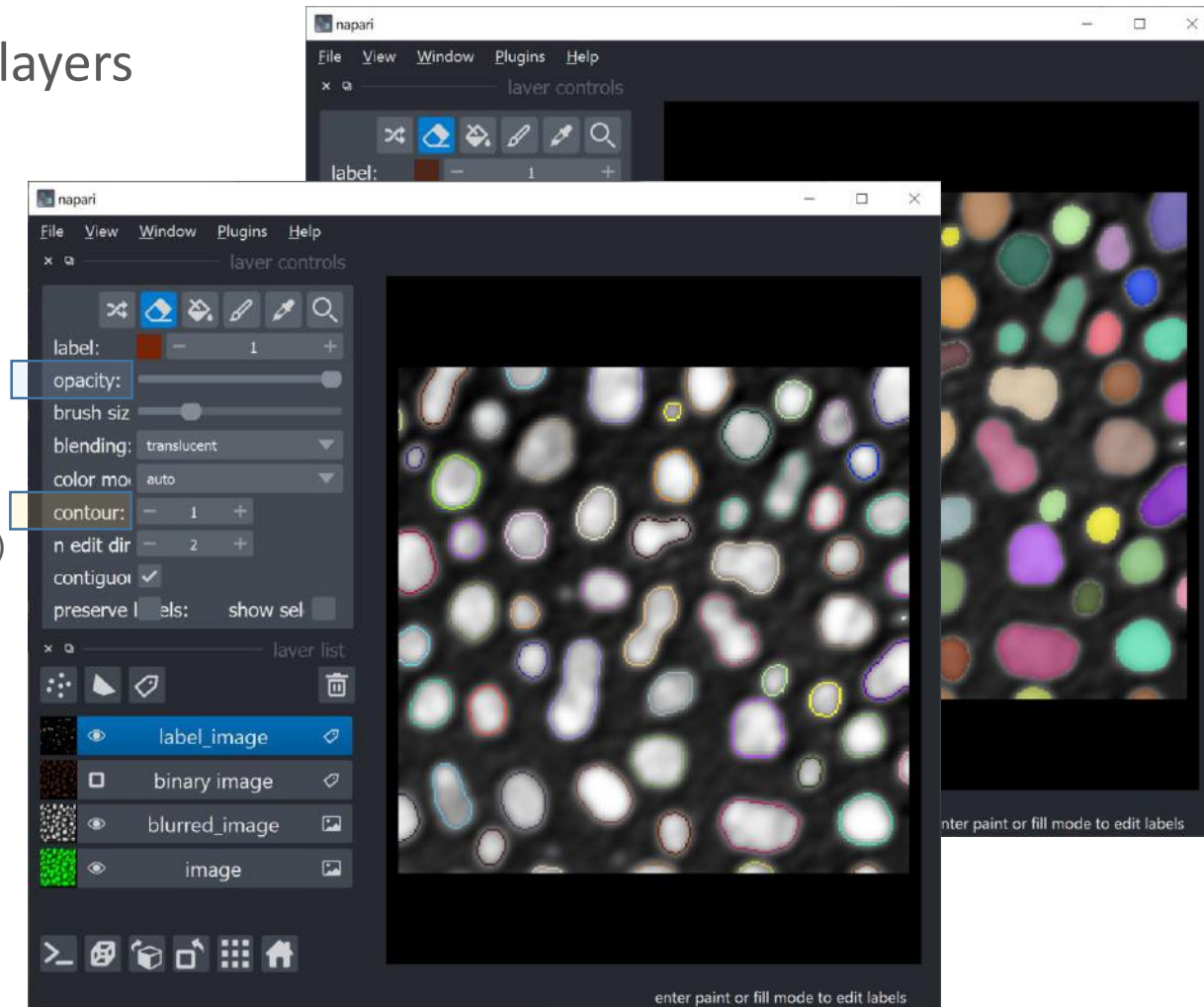
- Binary images and label images visualized as label layers

```
from skimage.measure import label
label_image = label(binary_image)

# add labels to viewer
label_layer = viewer.add_labels(label_image)
```

- Visualize contours instead of the overlay

```
label_layer.contour = 1
label_layer.opacity = 1
```



Points layers

- There is also other layer types
 - Shapes
 - **Points**
 - Surfaces
 - Tracks
 - Vectors

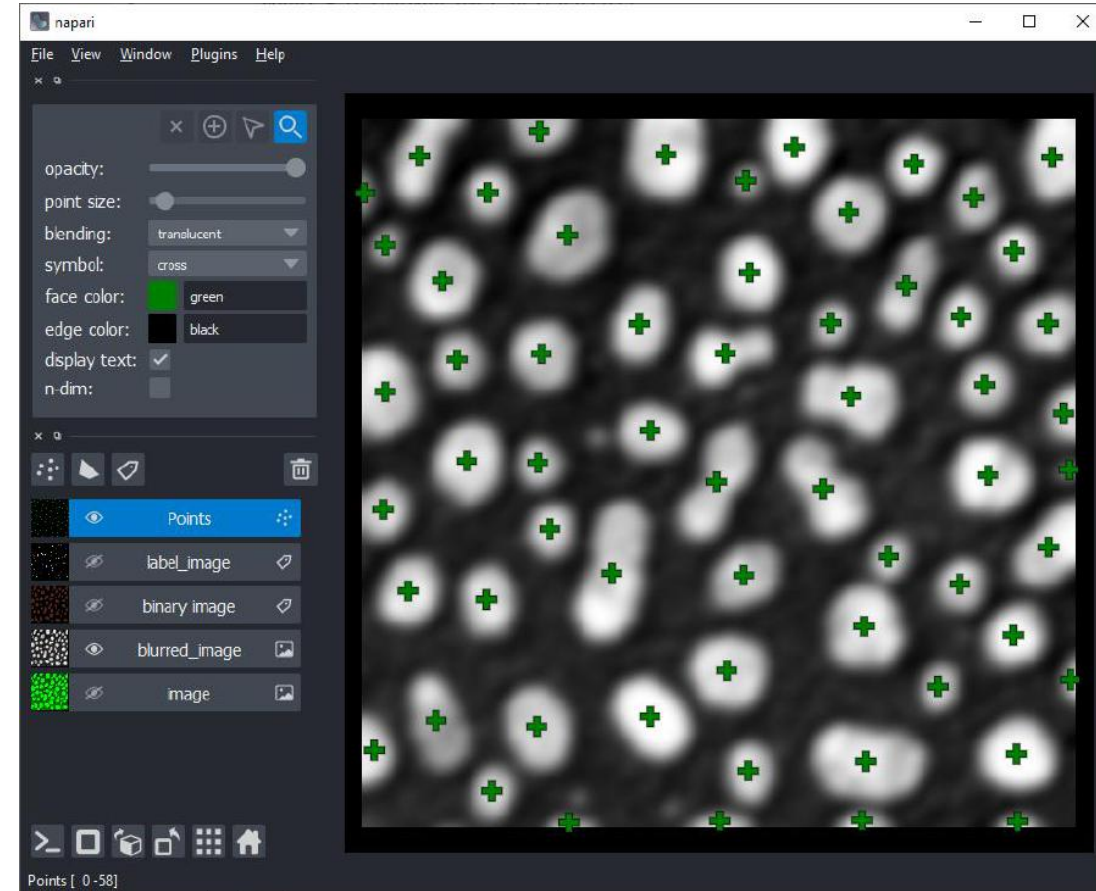
```
from skimage.measure import regionprops
```

```
statistics = regionprops(label_image)
```

```
points = [s.centroid for s in statistics]
```

```
# add points to viewer
```

```
label_layer = viewer.add_points(points, face_color='green', symbol='cross', size=5)
```



Tables

- Support for `tables` was added recently. Basic function

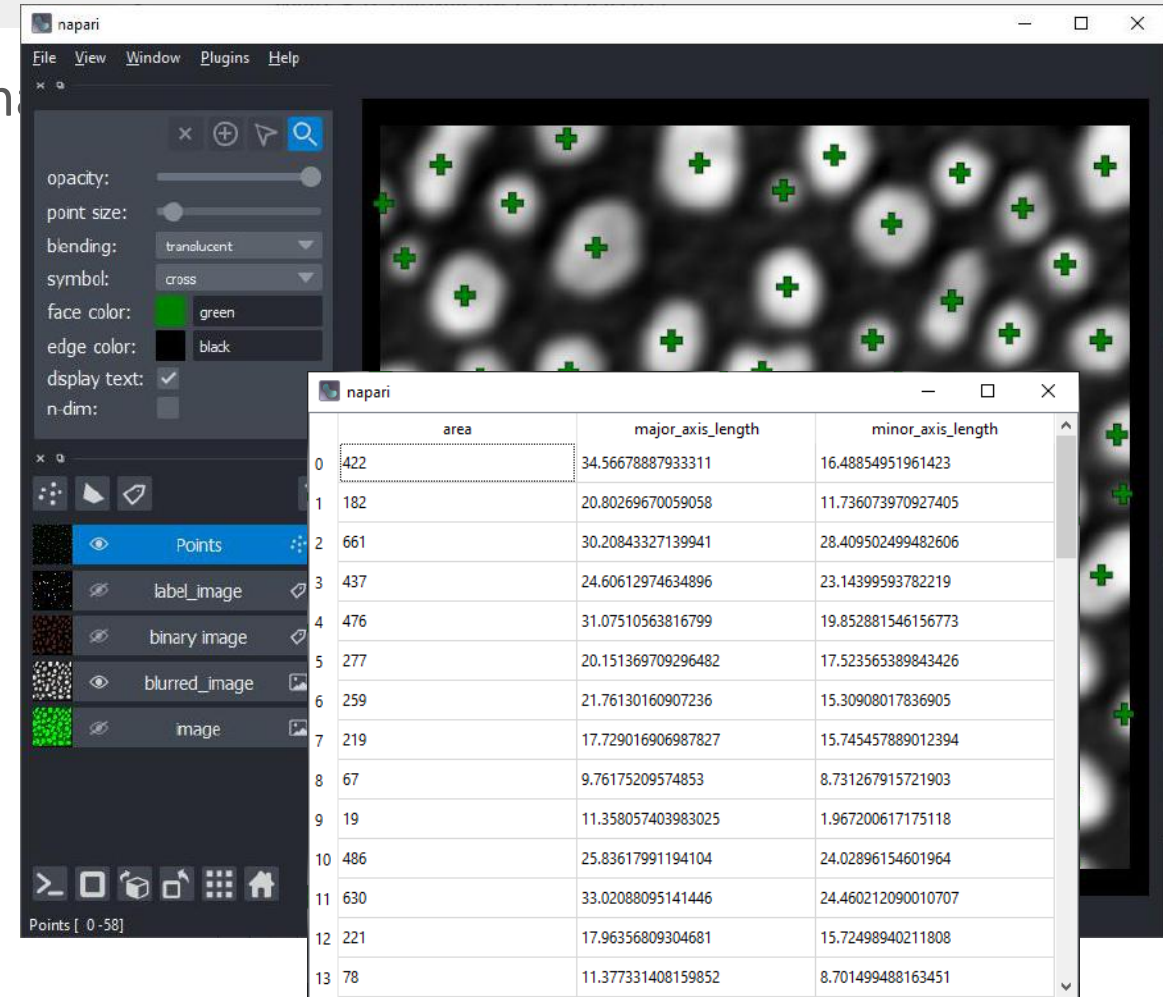
```
from skimage.measure import regionprops_table
```

```
statistics = regionprops_table(label_image,  
                               properties=['area',  
                                          'major_axis_length',  
                                          'minor_axis_length'])
```

```
# Show statistics as table on screen
```

```
from magicgui.widgets import Table
```

```
Table(statistics).show()
```



Interactive pixel classification

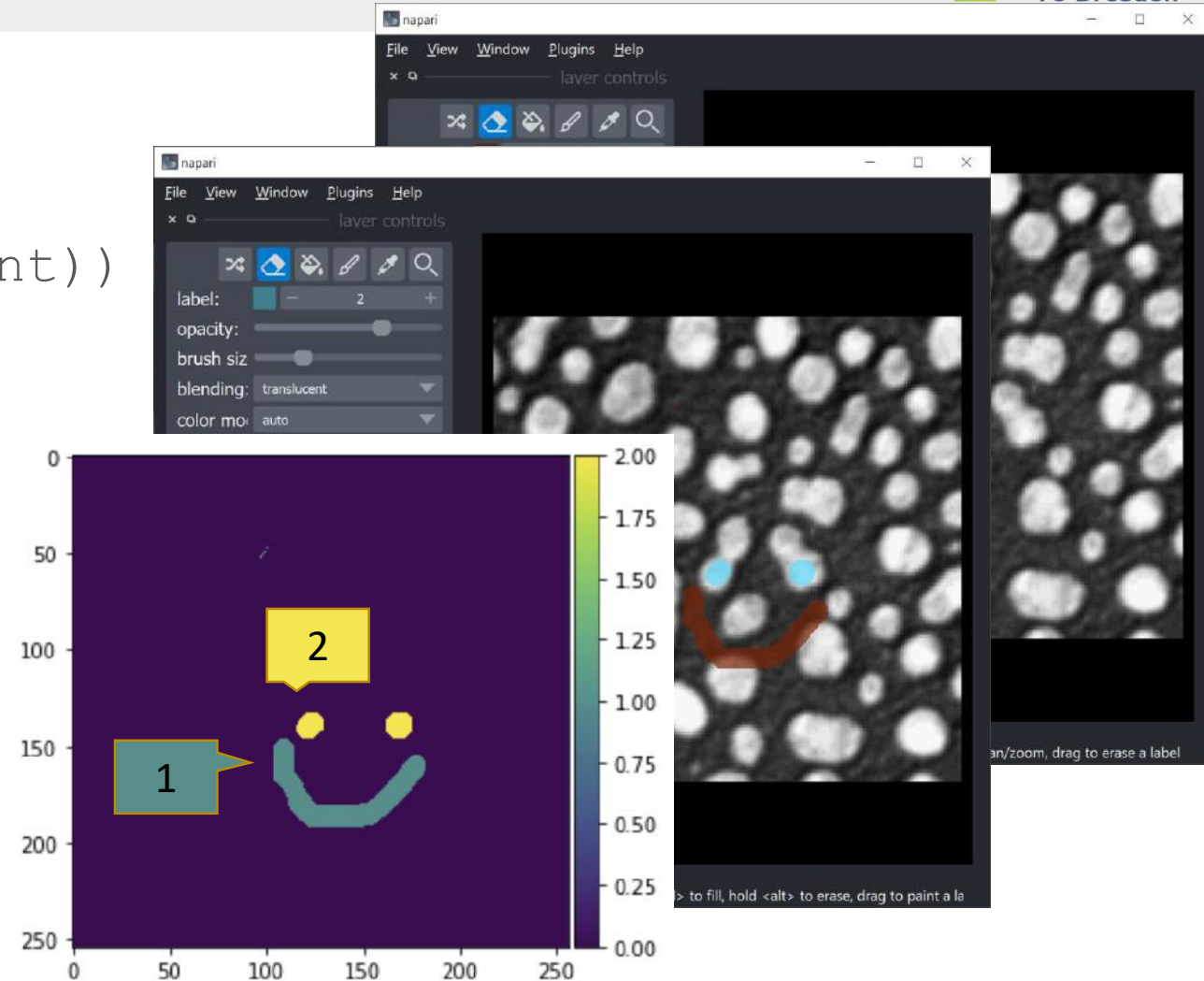
- Prepare an empty layer for annotations

```
labels = viewer.add_labels(  
    np.zeros(image.shape).astype(int))
```

- Read annotations

```
manual_annotations = labels.data
```

```
from skimage.io import imshow  
imshow>manual_annotations,  
      vmin=0, vmax=2)
```



Interactive pixel classification

- Pixel classification using scikit-learn

```
# for training, we need to generate features
```

```
feature_stack = generate_feature_stack(image)
```

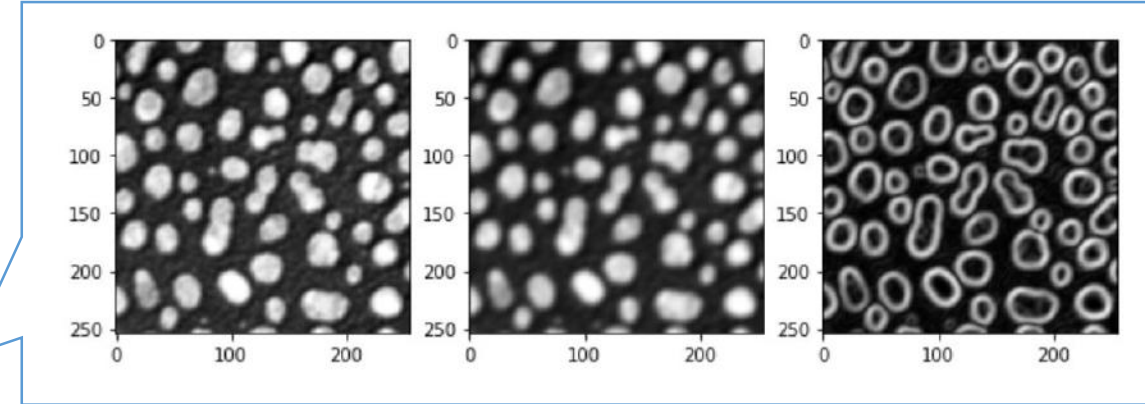
```
X, y = format_data(feature_stack, manual_annotations)
```

```
# train classifier
```

```
from sklearn.ensemble import RandomForestClassifier
```

```
classifier = RandomForestClassifier(max_depth=2, random_state=0)
```

```
classifier.fit(X, y)
```



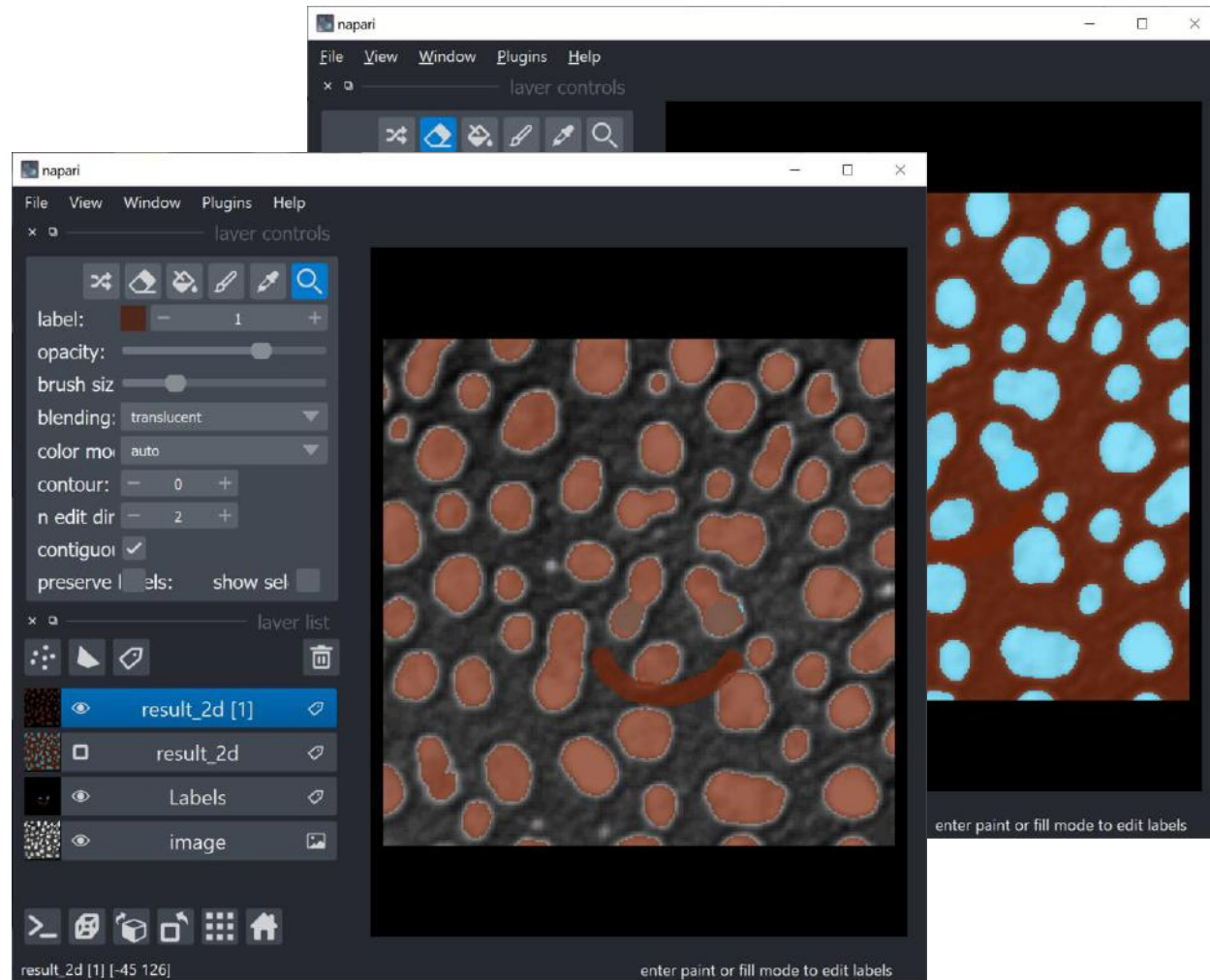
Interactive pixel classification

- Pixel classification using scikit-learn

```
# process the whole image and show result
result_1d = classifier.predict(feature_stack.T)
result_2d = result_1d.reshape(image.shape)
```

```
# make sure background = 0
result_2d = result_2d - 1
```

```
viewer.add_labels(result_2d)
```



Key bindings

- Bind short-cuts to operations

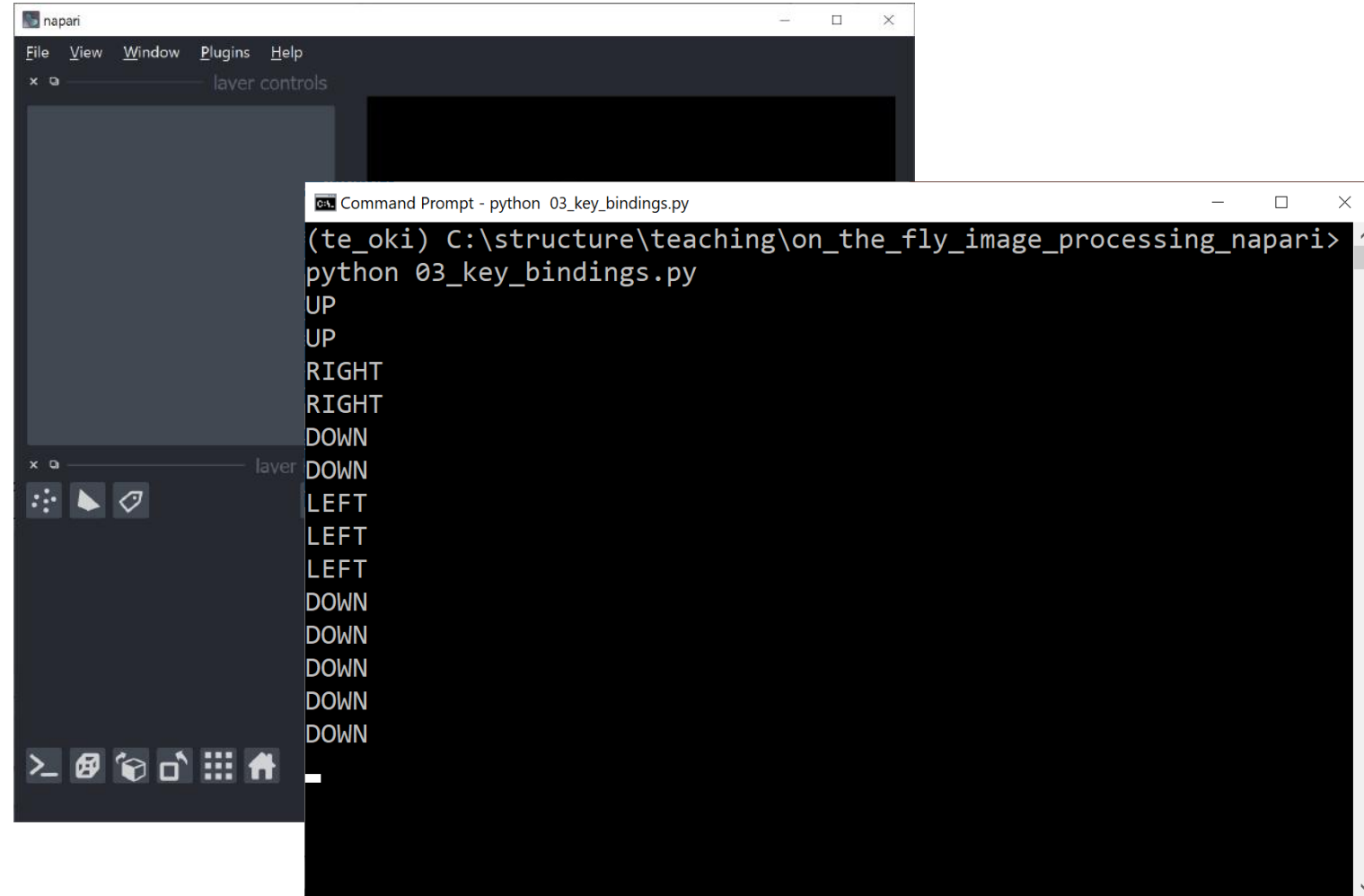
```
import napari

viewer = napari.Viewer()

@viewer.bind_key("w")
def up(event):
    print("UP")

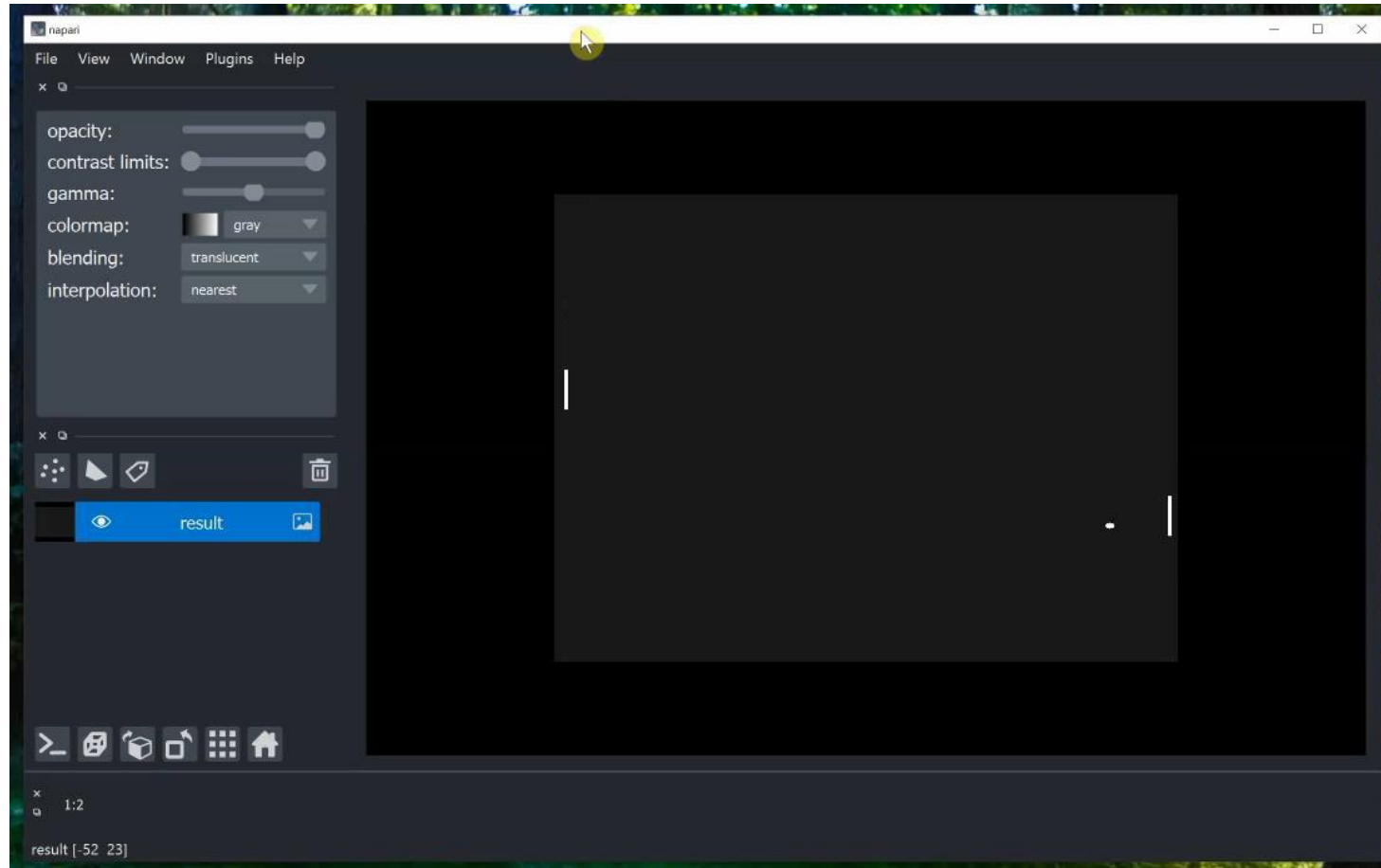
def left(event):
    print("LEFT")

viewer.bind_key("a", left)
```



Key bindings

- Control microscope stages, or



Multi-threading

- Python code typically runs in the `main thread`
- `Multi-threading` might be necessary for
 - User-interaction
 - Parallel processing
 - Device handling (camera, stage,...)
- Read more:
<https://napari.org/guides/stable/threading.html>

```
import napari
import time
from napari._qt.qthreading import thread_worker
viewer = napari.Viewer()
```

```
@thread_worker
def loop_run():
    while True: # endless loop
        print("Hello world", time.time())
        time.sleep(0.5)
        yield
```

Stops here once
napari closes

```
# Start the loop
worker = loop_run()
worker.start()
```

```
# Start napari
napari.run()
```

Multi-threading: processing images

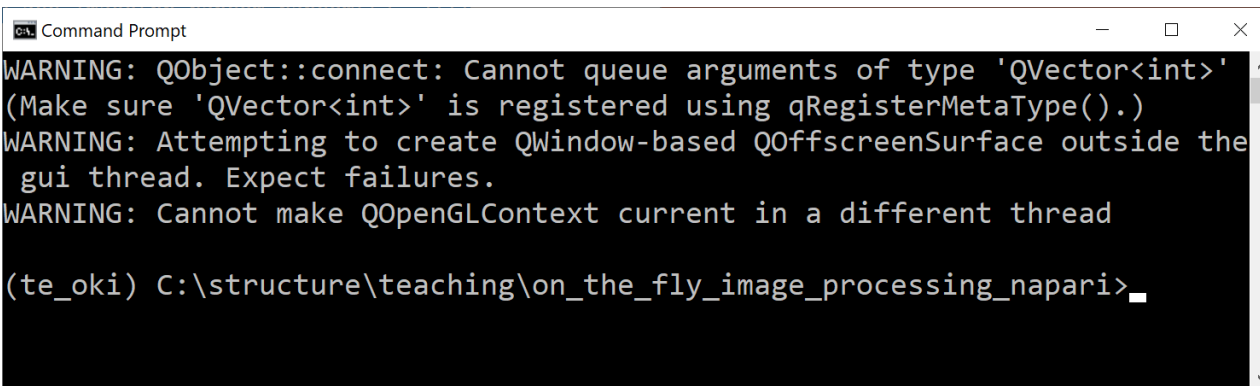
- Pitfall: You cannot access the viewer from other threads but the main thread.

```
import napari
import time
from napari._qt.qthreading import thread_worker
viewer = napari.Viewer()
```

```
@thread_worker
def loop_run():
    while True: # endless loop
        viewer.add_image(np.random.random((2, 2)))
        time.sleep(0.5)
        yield
```

```
# Start the loop
worker = loop_run()
worker.start()
```

```
# Start napari
napari.run()
```



```
Command Prompt
WARNING: QObject::connect: Cannot queue arguments of type 'QVector<int>'
(Make sure 'QVector<int>' is registered using qRegisterMetaType().)
WARNING: Attempting to create QWindow-based QOffscreenSurface outside the
gui thread. Expect failures.
WARNING: Cannot make QOpenGLContext current in a different thread
(te_oki) C:\structure\teaching\on_the_fly_image_processing_napari>
```

Multi-threading: processing images

- Solution: `yield` results
 - “Return something but continue processing”
 - a.k.a. Generator pattern
 - <https://wiki.python.org/moin/Generators>
- Add/update layers on demand

```
def update_layer(image):  
    """  
    Updates the image in the layer 'result'  
    or adds this layer.  
    """  
    try:  
        viewer.layers['result'].data = image  
    except KeyError:  
        viewer.add_image(image, name='result')
```

```
import napari  
import time  
from napari._qt.qthreading import thread_worker  
viewer = napari.Viewer()  
  
@thread_worker  
def loop_run():  
    while True: # endless loop  
        yield np.random.random((2, 2))  
        time.sleep(0.5)  
  
# Start the loop  
worker = loop_run()  
worker.yielded.connect(update_layer)  
worker.start()
```

Big thanks to



Talley J. Lambert
(HMS)
@TalleyJLambert

Reading camera output

- Image acquisition: OpenCV

```
from cv2.cv2 import VideoCapture
from skimage.color import rgb2gray
```

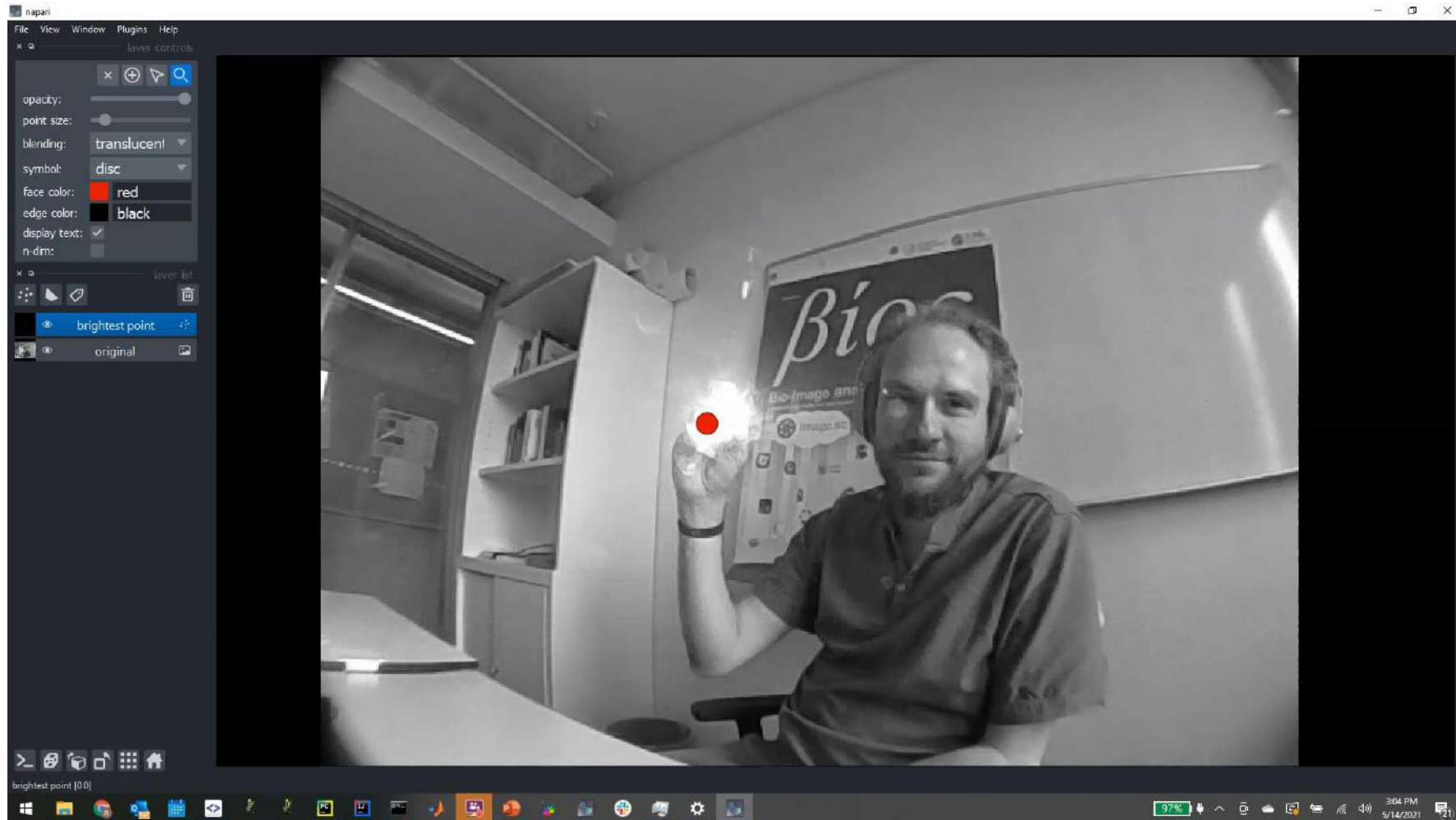
```
camera = VideoCapture(camera_index)
```

```
def acquire_image(camera : VideoCapture):
    _, picture = camera.read()
    if picture is None:
        return
```

```
# convert to single channel image
return rgb2gray(picture)
```



Multi-threading: processing camera images



- magicgui automatically generates a user interface for your function
- Read more: <https://napari.org/magicgui/>

```
def gaussian_blur(image, sigma = 2):  
    """  
    Apply a gaussian blur to an image.  
    """  
    return gaussian(image, sigma)
```

- `magicgui` automatically generates a user interface for your function
- Read more: <https://napari.org/magicgui/>

```
@magicgui(call_button='Run')
```

Input parameters

Return type

```
def gaussian_blur(image : ImageData, sigma : float = 2) -> ImageData:
```

```
    """
```

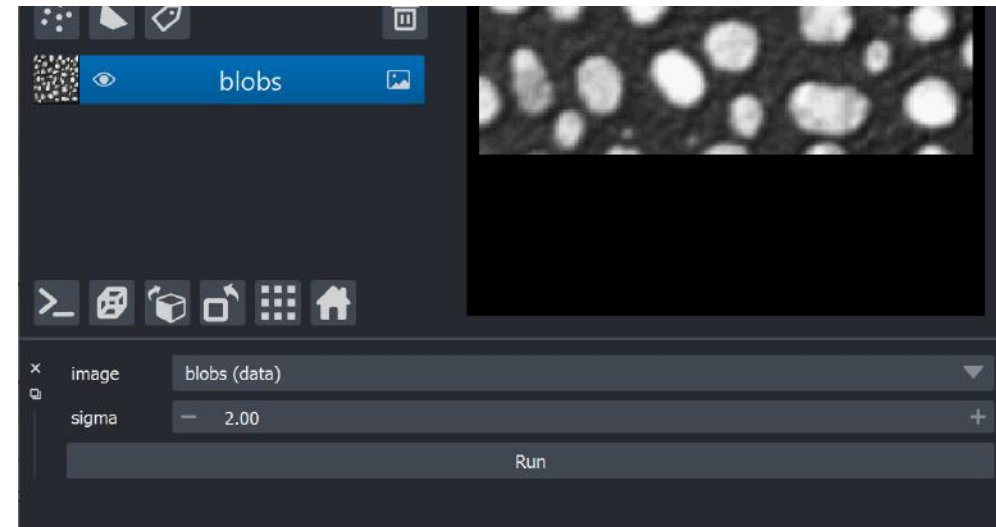
```
    Apply a gaussian blur to an image.
```

```
    """
```

```
    return gaussian(image, sigma)
```

```
# add user interface to napari viewer
```

```
viewer.window.add_dock_widget(gaussian_blur)
```



Dock widgets

- **Dock widgets** are user-interface plugins that snap to napari's window

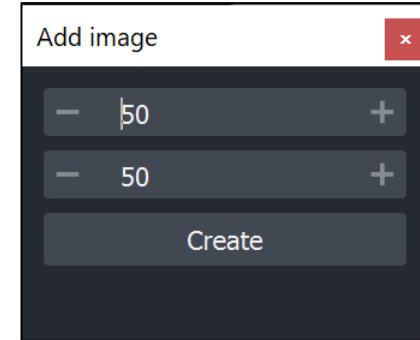
- They have a **title**

```
import napari
from qtpy.QtWidgets import QWidget
```

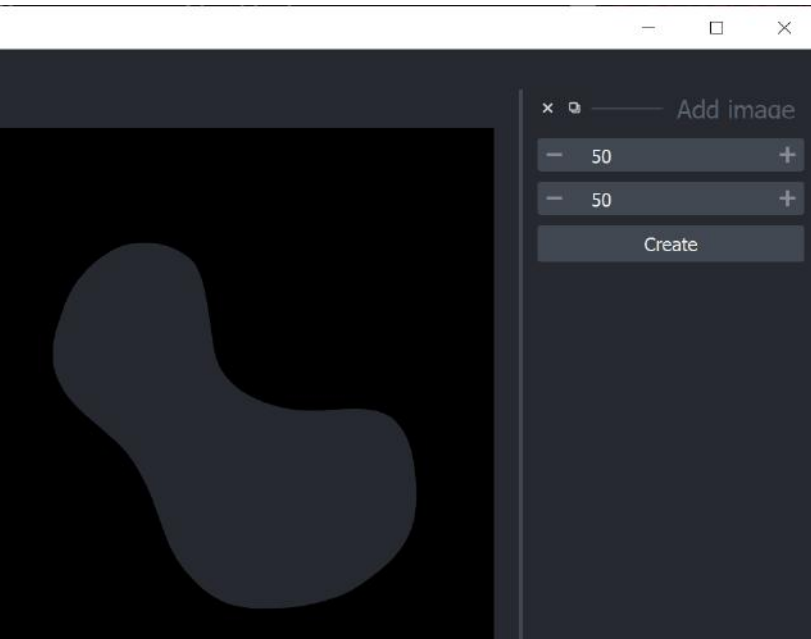
```
viewer = napari.Viewer()
```

```
class AddImageWidget(QWidget):
    def __init__(self, napari_viewer: napari.Viewer):
        super().__init__()
        self._viewer = napari_viewer
        self.setObjectName("Add image")
```

```
# add the dock widget to the viewer
viewer.window.add_dock_widget(AddImageWidget(viewer))
```



Undockable ;-)



Dock widgets

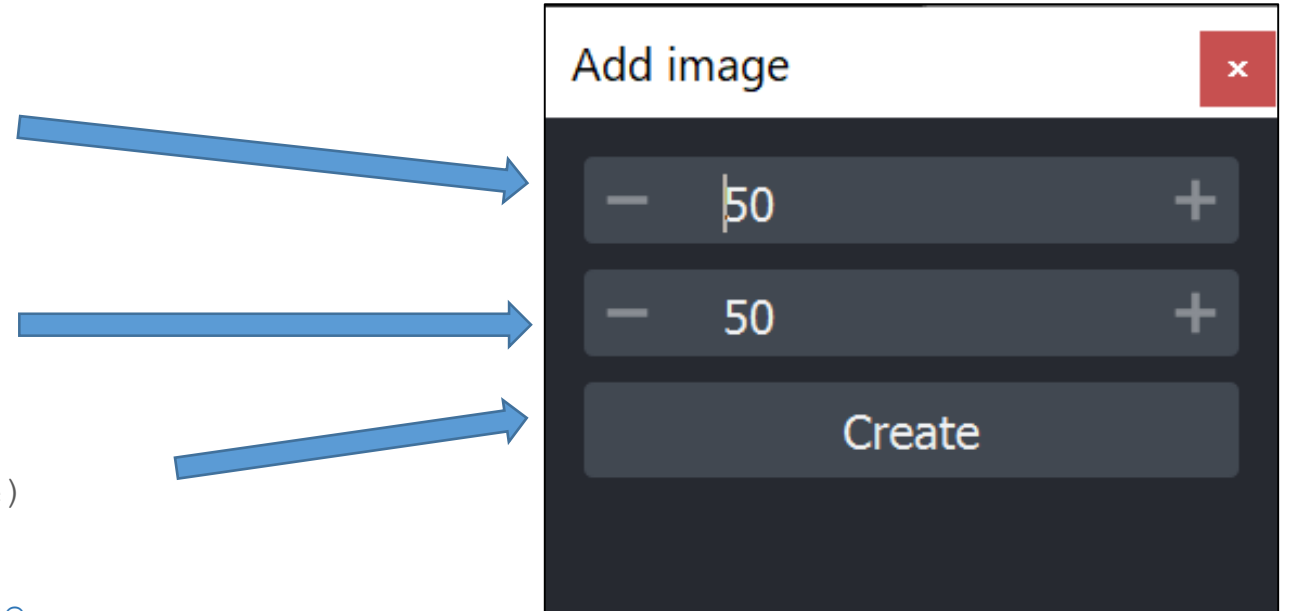
- User interfaces are made with Qt for python. Qt is the “cute” library. Read more: <https://doc.qt.io/qtforpython/>

```
self.width_spinner = QSpinBox()  
self.width_spinner.setValue(50)
```

```
self.height_spinner = QSpinBox()  
self.height_spinner.setValue(50)
```

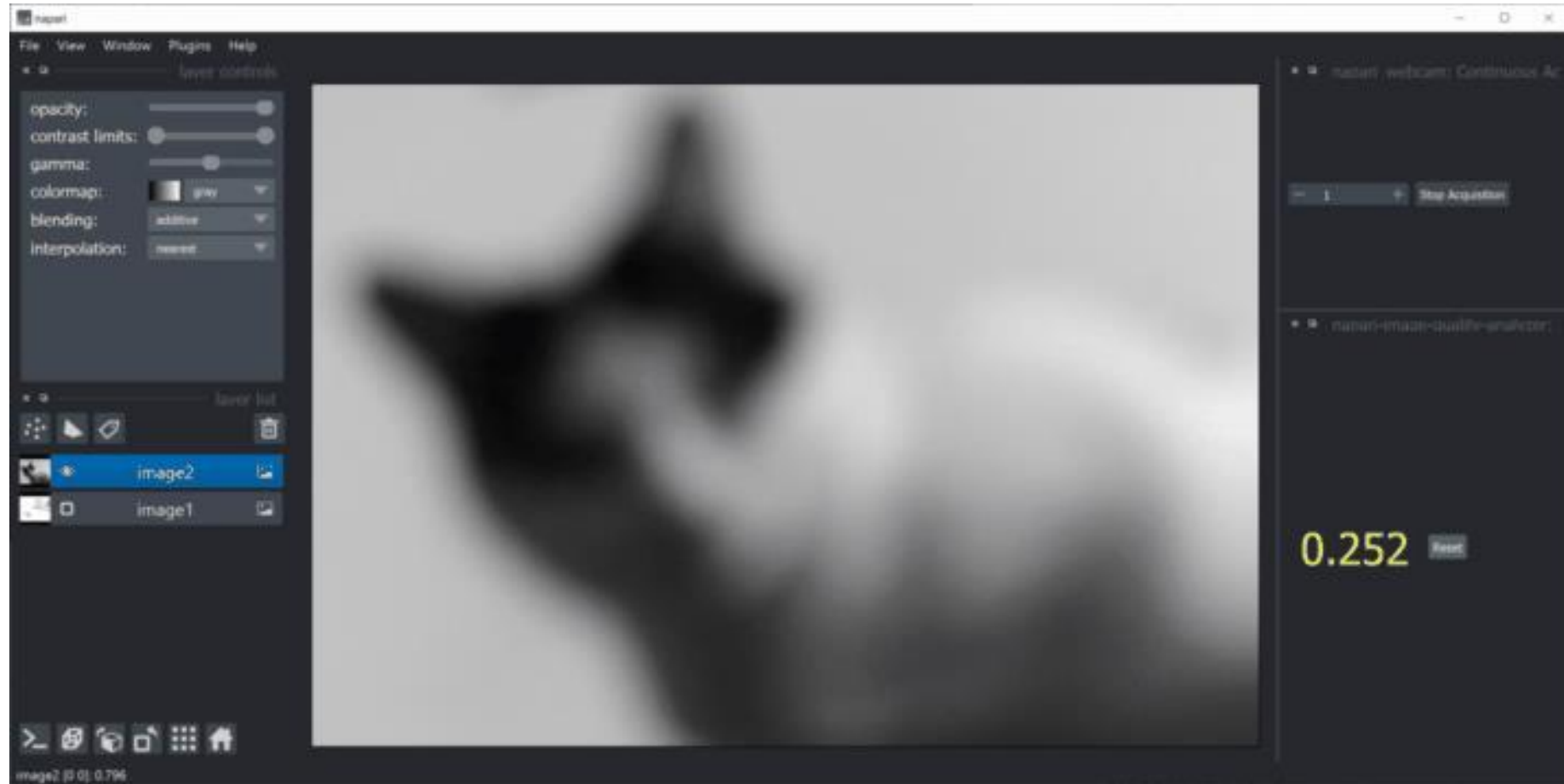
```
self.create_button = QPushButton("Create")  
self.create_button.clicked.connect(self._create)
```

```
# put spinners and buttons in the user interface  
self.setLayout(QVBoxLayout(self))  
self.layout().addWidget(self.width_spinner)  
self.layout().addWidget(self.height_spinner)  
self.layout().addWidget(self.create_button)
```



On the fly processing + dock widgets

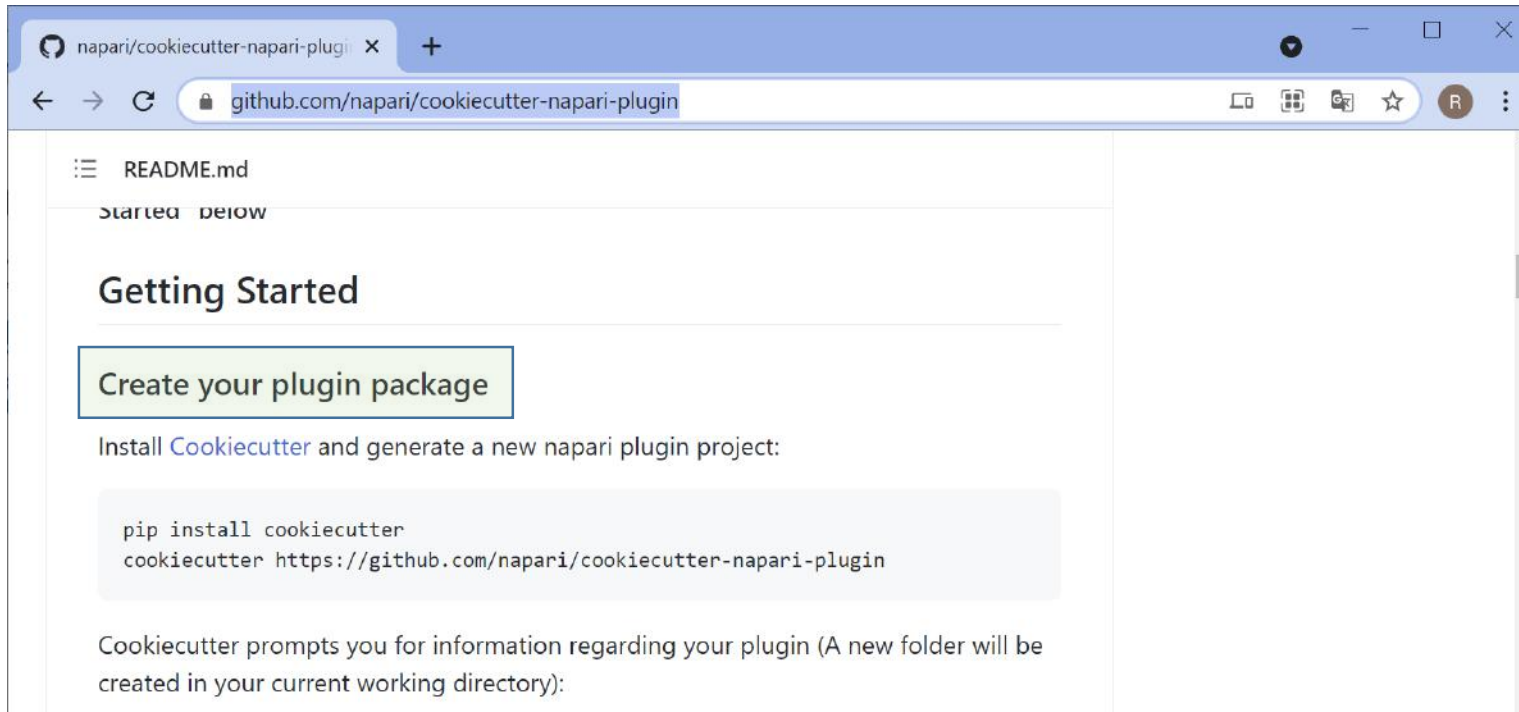
- Measure image quality on the fly



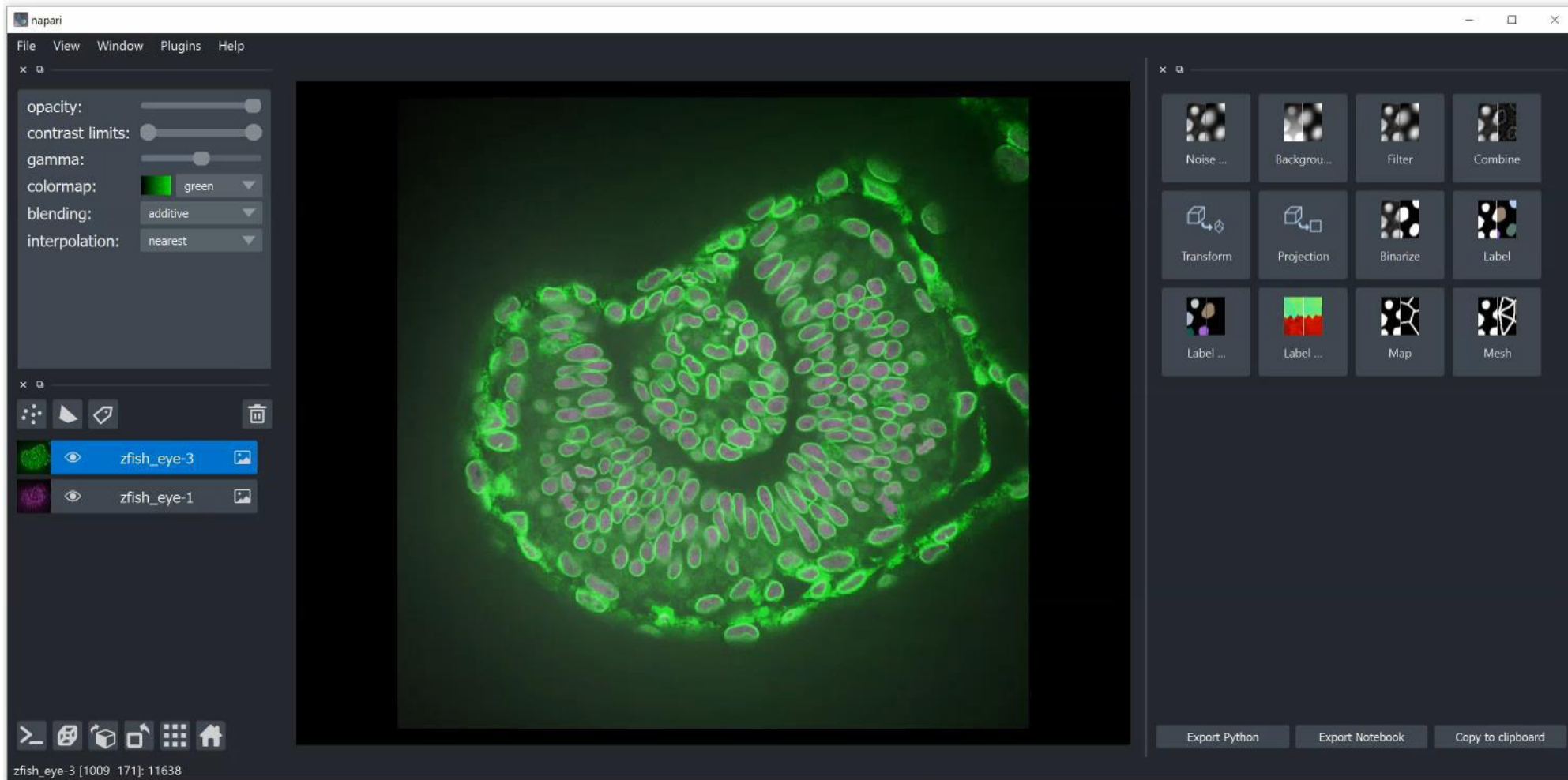
Napari plugins

- Instead of adding functions and dock widgets to the napari viewer:
 - Make plugins!
 - Use the cookiecutter!

Learn from Rafa
in our session
on Friday!



clEsperanto: Building image data flow graphs in napari



Fantastic
support by



Talley J. Lambert
(HMS)
@TalleyJLambert



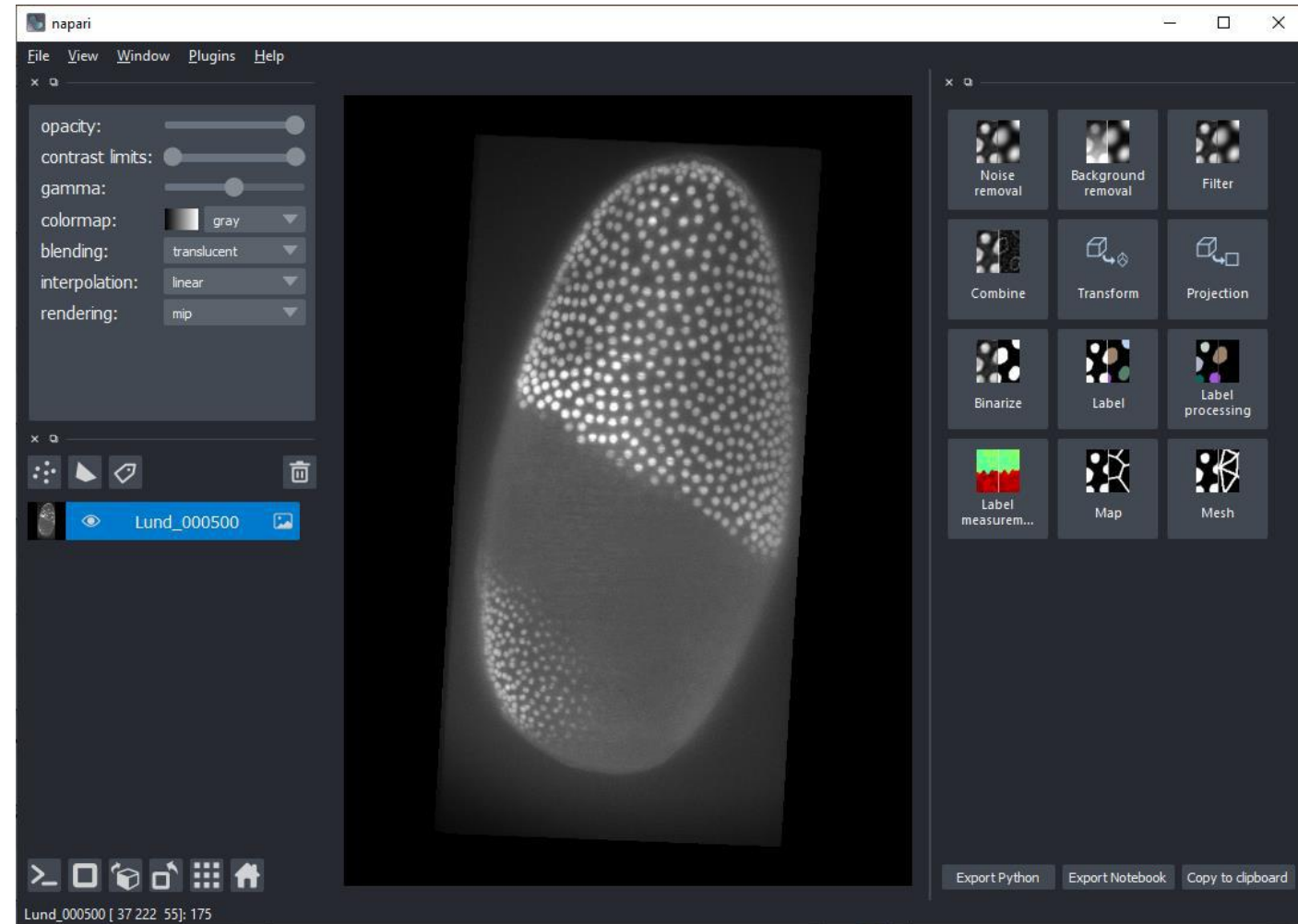
@haesleinhuepf
@PoLDresden

https://github.com/clesperanto/napari_pyclesperanto_assistant

Image data source: Mauricio Rocha Martins, Norden lab, MPI CBG (now at IGC Ciência)

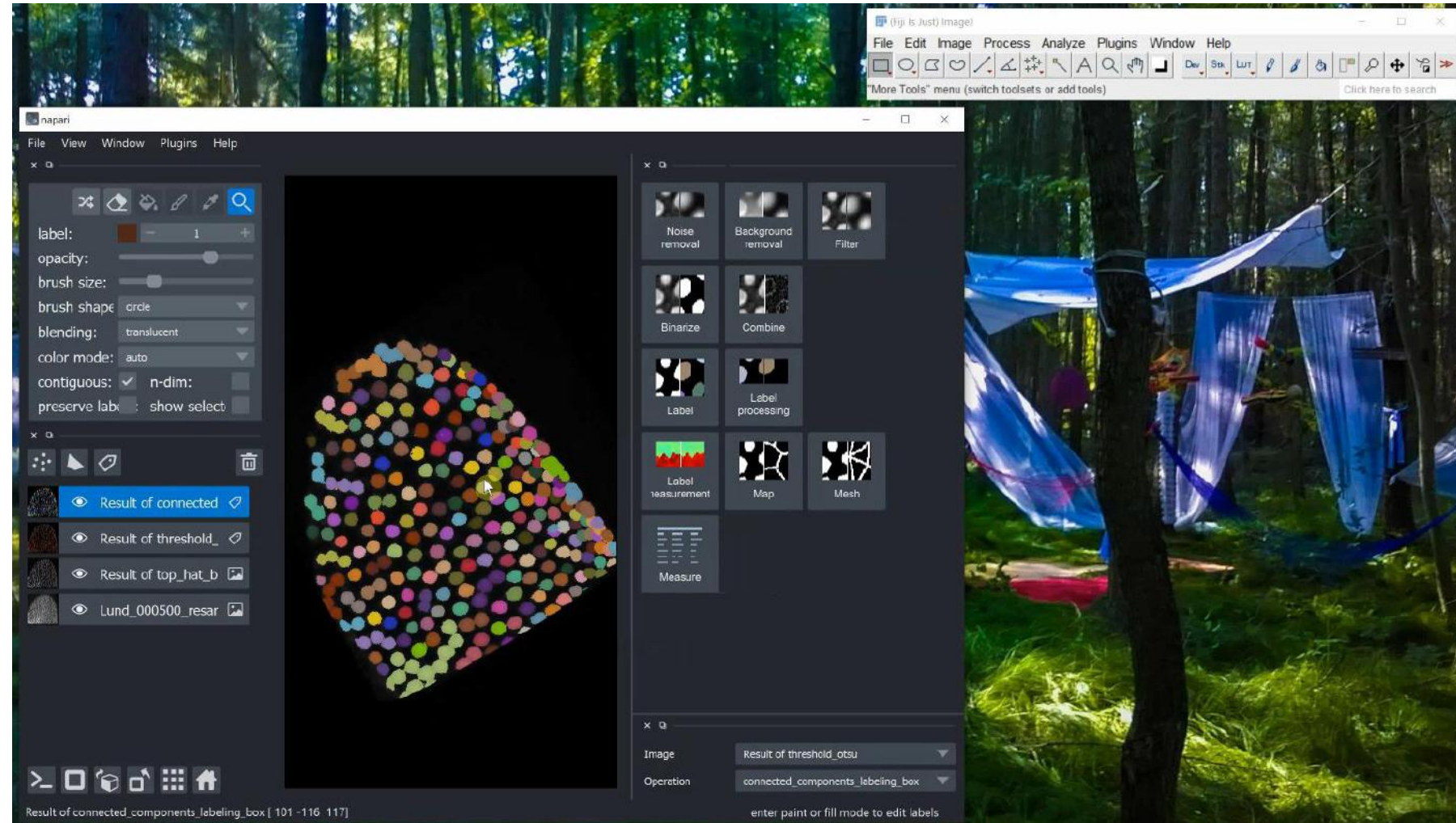
clEsperanto: Building image data flow graphs in napari

- The button order is intentional.
- Work from the top left to the bottom right



Code export

- Export code from napari and
 - run it in Fiji
 - run it in Python



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
```

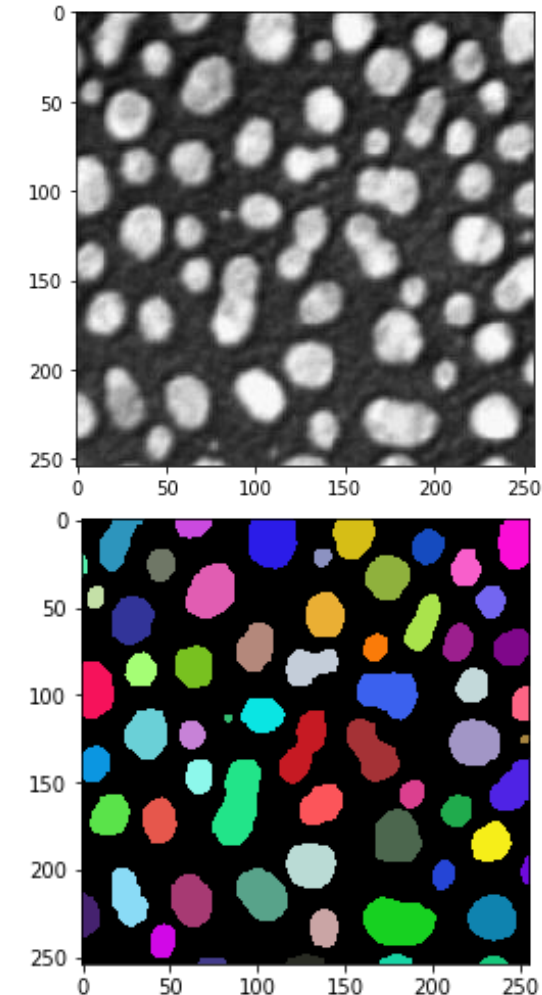
```
binary = cle.threshold_otsu(blurred)
```

```
# labeling
```

```
labels = cle.connected_components_labeling_box(binary)
```

```
# visualize results
```

```
cle.imshow(labels, labels=True)
```



Fantastic
support by



Juan Nunez-Iglesias
(Monash)
@jnuneziglesias



Talley J. Lambert
(HMS)
@TalleyJLambert

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()
```

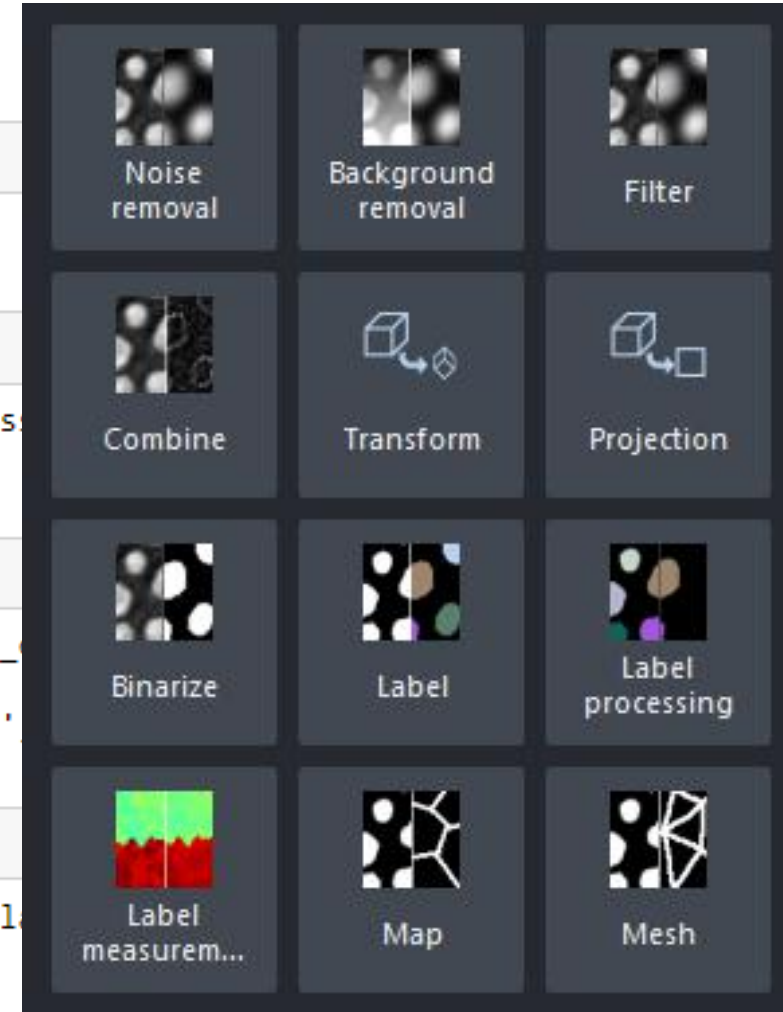
```
dict_keys(['bottom_hat_box', 'bottom_hat_sphere', 'difference_of_gaussian', 'divide_by_gaussian', 'background', 'top_hat_box', 'top_hat_sphere'])
```

```
cle.operations('binarize').keys()
```

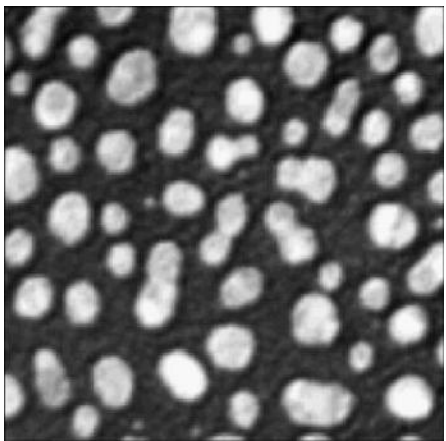
```
dict_keys(['detect_label_edges', 'detect_maxima_box', 'detect_minima_box', 'equal', 'equal_ant', 'greater_or_equal', 'greater_or_equal_constant', 'label_to_mask', 'local_threshold', 'smaller', 'smaller_constant', 'smaller_or_equal', 'smaller_or_equal_constant', 'threshold'])
```

```
cle.operations('label').keys()
```

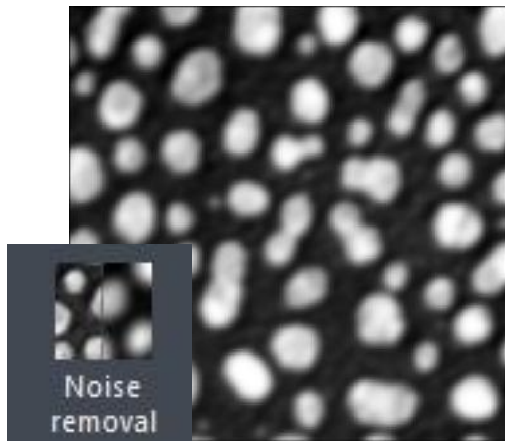
```
dict_keys(['connected_components_labeling_box', 'connected_components_labeling_diamond', 'label', 'g', 'voronoi_labeling', 'voronoi_otsu_labeling'])
```



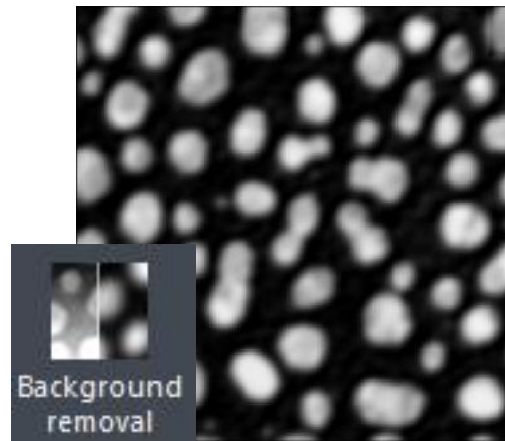
clEsperanto categories



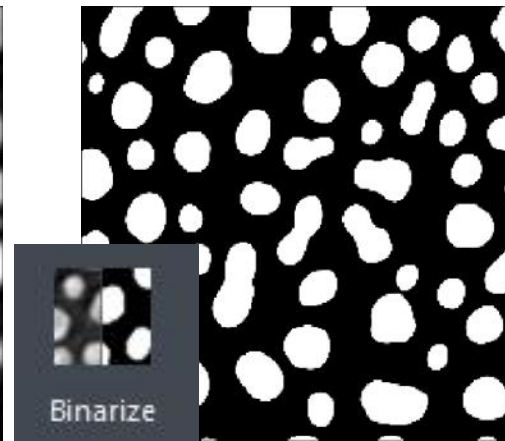
Original



Gaussian blur



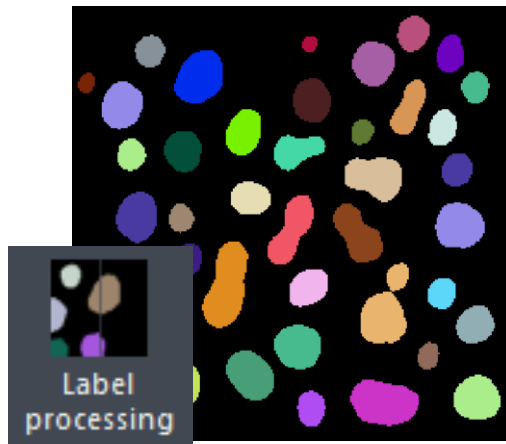
Top hat



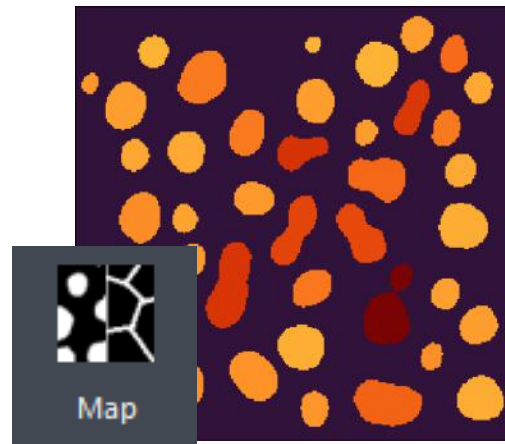
Threshold Otsu



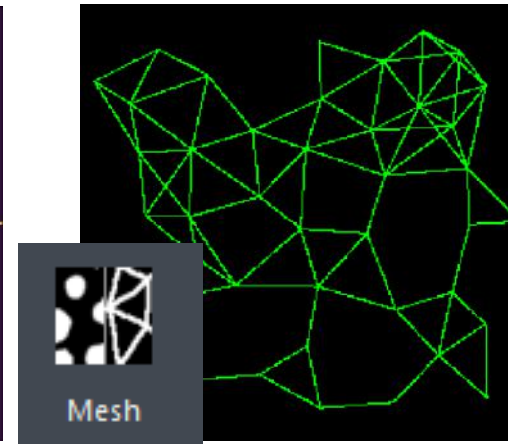
Connected Component



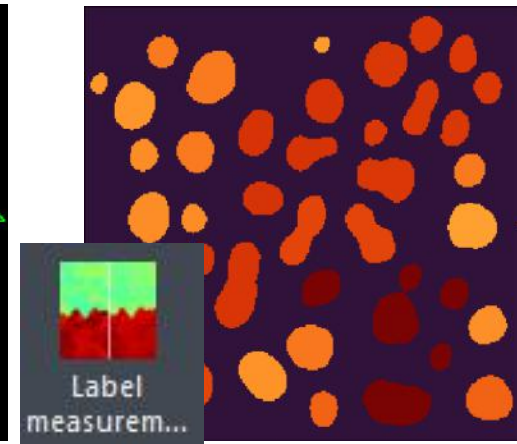
Exclude edges



Extension ratio



Mesh proximal neighbors



Maximum of n-nearest n.

- clEsperanto's images are numpy-compatible arrays.

```
from skimage.measure import regionprops
statistics = regionprops(labels)
```

```
import numpy as np
np.mean([s.area for s in statistics])
```

- Sometimes, they have to be converted to numpy arrays:

```
from skimage.io import imsave
imsave('output.tif', np.asarray(labels))
```

clEsperanto: GPU-accelerated image processing

- Why?
- Because operations can run orders of magnitude faster!

Scipy

```
import time
from scipy import ndimage as ndi

scaled = np.ndarray(output_shape)
for i in range(0, 10):
    start_time = time.time()
    ndi.affine_transform(image, matrix, output=scaled, out
    print("scipy affine transform duration: " + str(time.ti
    imshow(scaled[120])

scipy affine transform duration: 9.005393266677856
scipy affine transform duration: 9.04822826385498
scipy affine transform duration: 9.089797019958496
scipy affine transform duration: 8.864994764328003
scipy affine transform duration: 9.063815832138062
scipy affine transform duration: 8.975829362869263
scipy affine transform duration: 8.582808256149292
scipy affine transform duration: 9.115242958068848
scipy affine transform duration: 9.170432567596436
scipy affine transform duration: 9.086284637451172
```

clEsperanto

```
ocl_image = cle.push(image)

ocl_scaled = cle.create(output_shape)
for i in range(0, 10):
    start_time = time.time()
    cle.affine_transform(ocl_image, ocl_scaled, transform=np.lina
    print("clEsperanto affine transform duration: " + str(time.ti

result = cle.pull(ocl_scaled)
imshow(result[120])

clEsperanto affine transform duration: 0.03690147399902344
clEsperanto affine transform duration: 0.003954648971557617
clEsperanto affine transform duration: 0.0029942989349365234
clEsperanto affine transform duration: 0.00398564338684082
clEsperanto affine transform duration: 0.0030252933502197266
clEsperanto affine transform duration: 0.002960205078125
clEsperanto affine transform duration: 0.003988981246948242
clEsperanto affine transform duration: 0.003989696502685547
clEsperanto affine transform duration: 0.0030260086059570312
clEsperanto affine transform duration: 0.003954887390136719
```

CUDA based GPU-acceleration:

<https://cupy.dev/>
<https://github.com/rapidsai/cucim>

cupy

```
import time
import cupy
from cupyx.scipy import ndimage as ndi

cuda_image = cupy.asarray(image)

cuda_scaled = cupy.ndarray(output_shape)
for i in range(0, 10):
    start_time = time.time()
    ndi.affine_transform(cuda_image, cupy.asarray(matrix), output=cuda_scaled, output_shape=output_shape)
    cupy.cuda.stream.get_current_stream().synchronize() # we need to wait here to measure time properly
    print("cupy affine transform duration: " + str(time.time() - start_time))

result = cupy.asnumpy(cuda_scaled)
imshow(result[120])

C:\Programs\miniconda3\envs\beetlesafari\lib\site-packages\cupyx\scipy\ndimage\interpolation.py:15: UserWarning: The default order of affine_transform is 1. It is different from scipy.ndimage and can change in the future
warnings.warn('In the current feature the default order of {} is 1. '

cupy affine transform duration: 0.4258244037628174
cupy affine transform duration: 0.01000523567199707
cupy affine transform duration: 0.010988473892211914
cupy affine transform duration: 0.010972261428833008
cupy affine transform duration: 0.010921955108642578
cupy affine transform duration: 0.011012792587280273
cupy affine transform duration: 0.010975122451782227
cupy affine transform duration: 0.009923934936523438
cupy affine transform duration: 0.011012554168701172
cupy affine transform duration: 0.010970354080200195
```

Note: This is just a snapshot! GPU-acceleration speedup / performance depends on many aspects such as image size, operation, parameters, used GPU, repetitions, ...

https://github.com/clEsperanto/pyclesperanto_prototype/blob/master/benchmarks/affine_transforms.ipynb

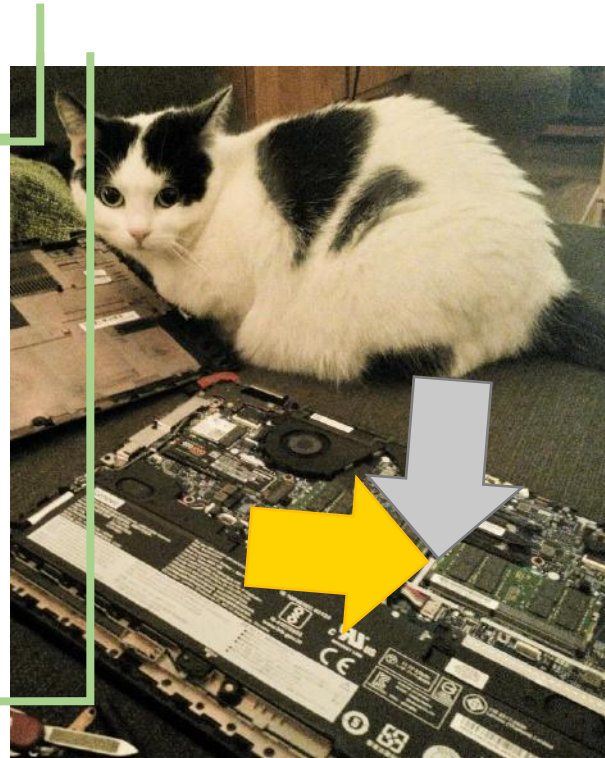
Graphics Processing Units (GPUs)

- ... don't have to be expensive.

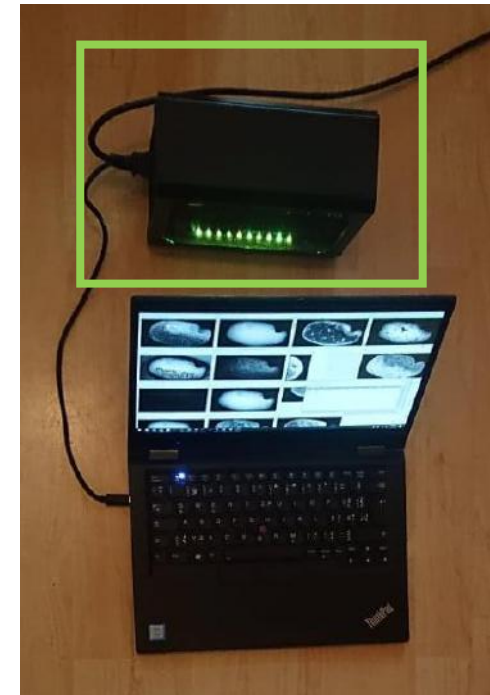
Central Processing Unit (CPU) Graphics Processing Unit (GPU)



dedicated GPUs



integrated GPUs



external “eGPUs”



Checklist

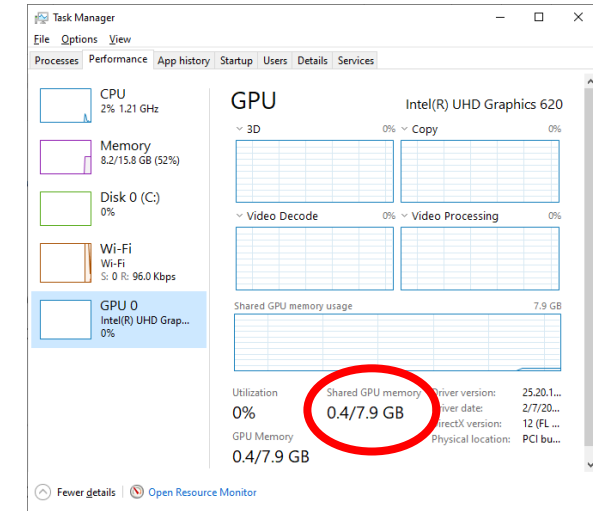
When does GPU-accelerated image processing make sense?

You need

- a workflow that is slow, ideally:
 - image processing time **10 : 1** image loading time,
- Enough **graphics card memory**,
 - > 5 times image size
- Basic coding experience
- A recent Intel, AMD or Nvidia GPU (integrated, dedicated, external)

If you really want to get the most out of it:

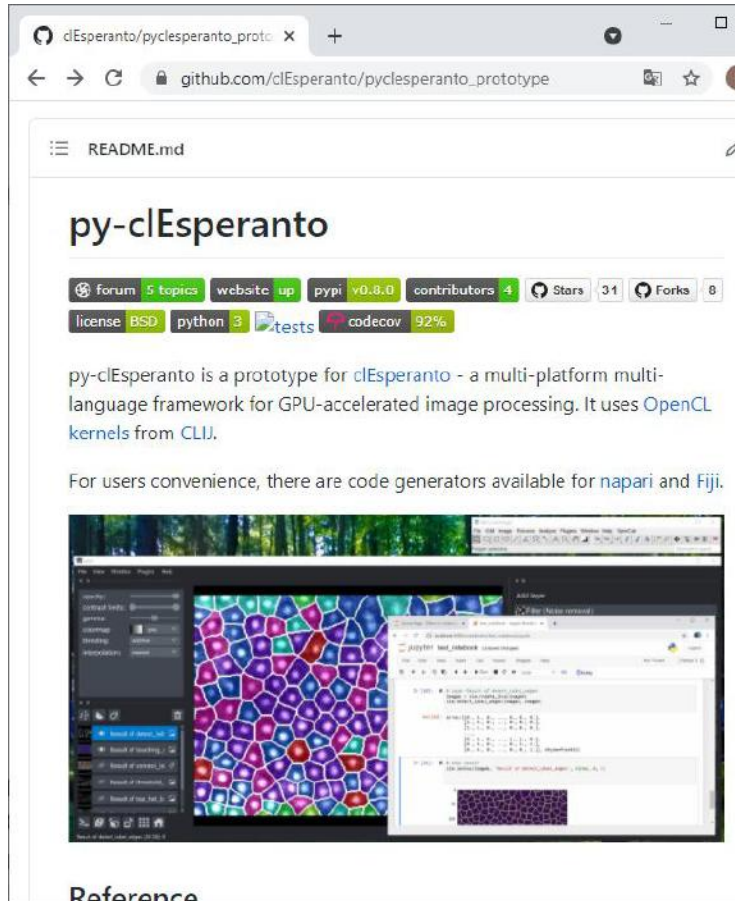
- a graphics card with **GDDR6** memory (memory bandwidth > 400 Gb/s).

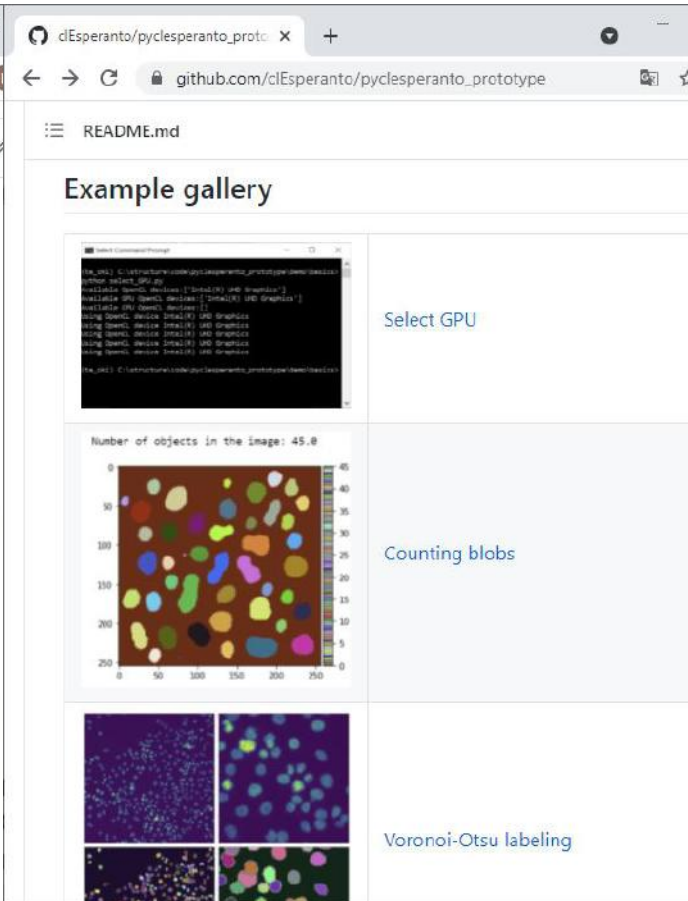


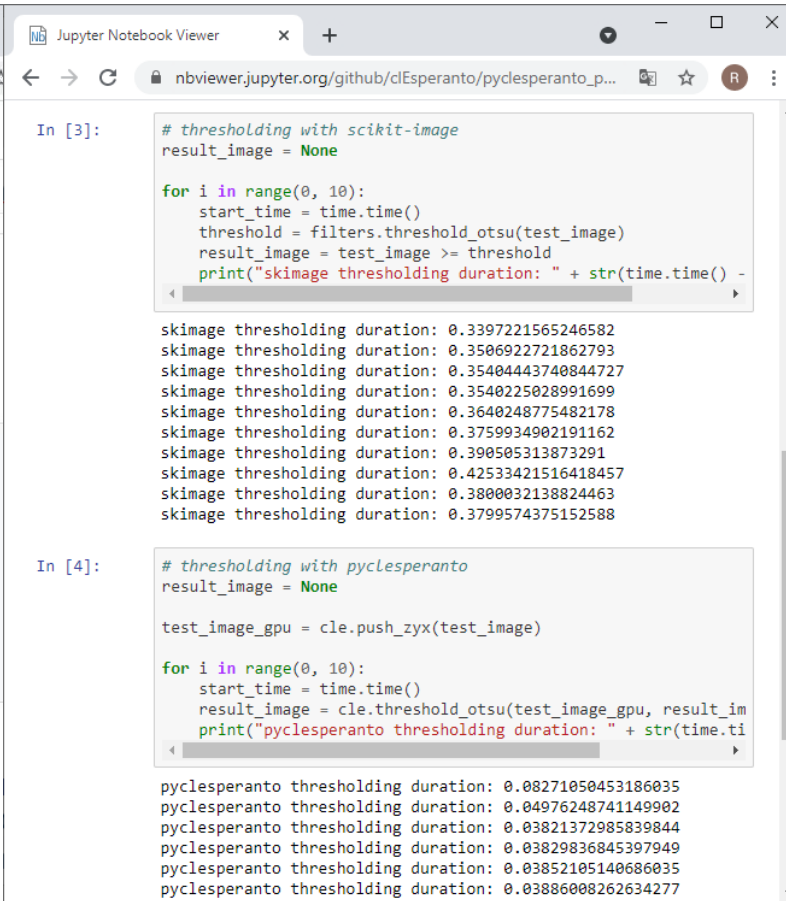
Online tutorials: clEsperanto

- The py-clEsperanto website offers examples and simple benchmarking for snake county.

cle._







```
In [3]: # thresholding with skimage-image
result_image = None

for i in range(0, 10):
    start_time = time.time()
    threshold = filters.threshold_otsu(test_image)
    result_image = test_image >= threshold
    print("skimage thresholding duration: " + str(time.time() - start_time))

skimage thresholding duration: 0.3397221565246582
skimage thresholding duration: 0.3506922721862793
skimage thresholding duration: 0.35404443740844727
skimage thresholding duration: 0.3540225028991699
skimage thresholding duration: 0.3640248775482178
skimage thresholding duration: 0.3759934902191162
skimage thresholding duration: 0.390505313873291
skimage thresholding duration: 0.42533421516418457
skimage thresholding duration: 0.3800032138824463
skimage thresholding duration: 0.3799574375152588

In [4]: # thresholding with pyclesperanto
result_image = None

test_image_gpu = cle.push_zyx(test_image)

for i in range(0, 10):
    start_time = time.time()
    result_image = cle.threshold_otsu(test_image_gpu, result_image)
    print("pyclesperanto thresholding duration: " + str(time.time() - start_time))

pyclesperanto thresholding duration: 0.08271050453186035
pyclesperanto thresholding duration: 0.04976248741149902
pyclesperanto thresholding duration: 0.03821372985839844
pyclesperanto thresholding duration: 0.03829836845397949
pyclesperanto thresholding duration: 0.03852105140686035
pyclesperanto thresholding duration: 0.03886008262634277
```

Online tutorials: napari

The image displays three overlapping browser windows showing napari tutorials. The top window shows the 'viewer.html' tutorial. The middle window shows the 'image.html' tutorial, which includes a 3D rendering of a brain and a sidebar with a search bar and a list of tutorials. The bottom window shows the 'surface.html' tutorial, which includes a 3D rendering of a brain and a sidebar with a search bar and a list of tutorials. The tutorials are titled 'viewer.html', 'image.html', and 'surface.html'.

viewer.html

image.html

surface.html

napari

3D rendering of images

All our layers can be rendered in both 2D and 3D mode, and one of our viewer buttons can toggle between each mode. The number of dimensions sliders will be 2 or 3 less than the total number of dimensions of the layer. See for example these brain surfaces rendered in 3D:

napari

opacity: contrast limits: colormap: mymap blending: translucent

gradient *

base *

canvas

dimension sliders

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@pr4deepr



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(Institut Pasteur)
@StRigaud



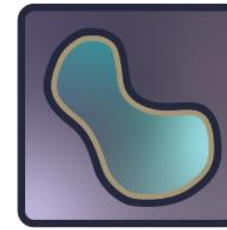
Juan Nunez-
Iglesias
(Monash)
@jnuneziglesias



Talley J. Lambert
(HMS)
@TalleyJLambert



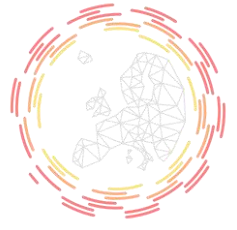
Nicholas
Sofroniew (CZI)
@sofroniewn



<https://napari.dev>

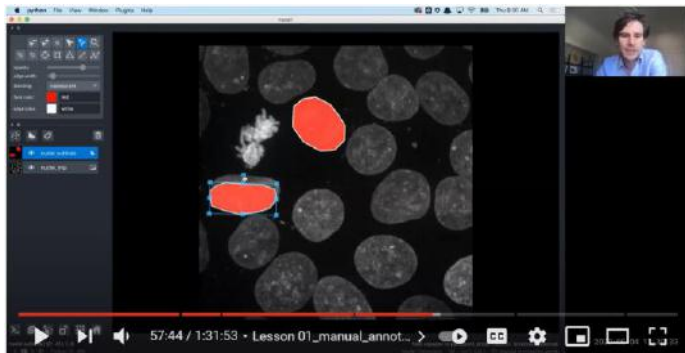


<https://image.sc>



<http://eubias.org/NEUBIAS/>

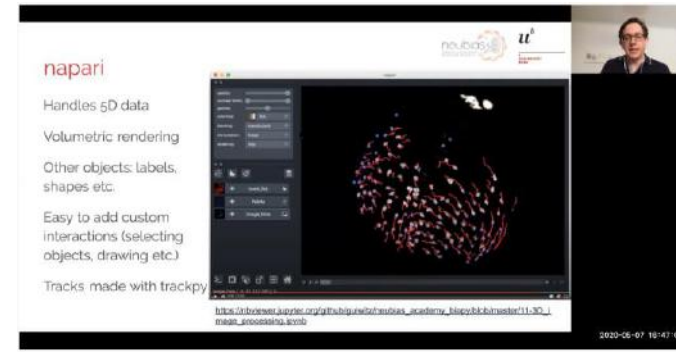
#OpenKnowledge



<https://www.youtube.com/watch?v=VgvDSq5aCDQ>



<https://www.youtube.com/watch?v=rOO6tK7z6yk>



<https://www.youtube.com/watch?v=2KF8vBrp3Zw>

Funding:



Let's take a break

```
git clone https://github.com/BiAPoL/on_the_fly_image_processing_napari.git
cd on_the_fly_image_processing_napari
pip install -r requirements.txt
```

```
conda install -c conda-forge pyopencl
pip install napari-pyclesperanto-assistant
```

```
napari --with clEsperanto Assistant
```



Linux users: you may need to install OpenCL, e.g.
`apt-get install ocl-icd-devel`

