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BioHPC

Image processing with Python

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Updated for 2022-10-12

A Python image processing workflow on BioHPC

1

webGUI/ webGPU

Create conda env and install python libraries

2

OnDemand Jupyter

Activate the conda env and run the workflow

https://portal.biohpc.swmed.edu/media/filer_public/33/6e/336ee943-7532-42ec-86a1-3931e562f988/2022_09_21_software_installation_on_biohpc.pdf

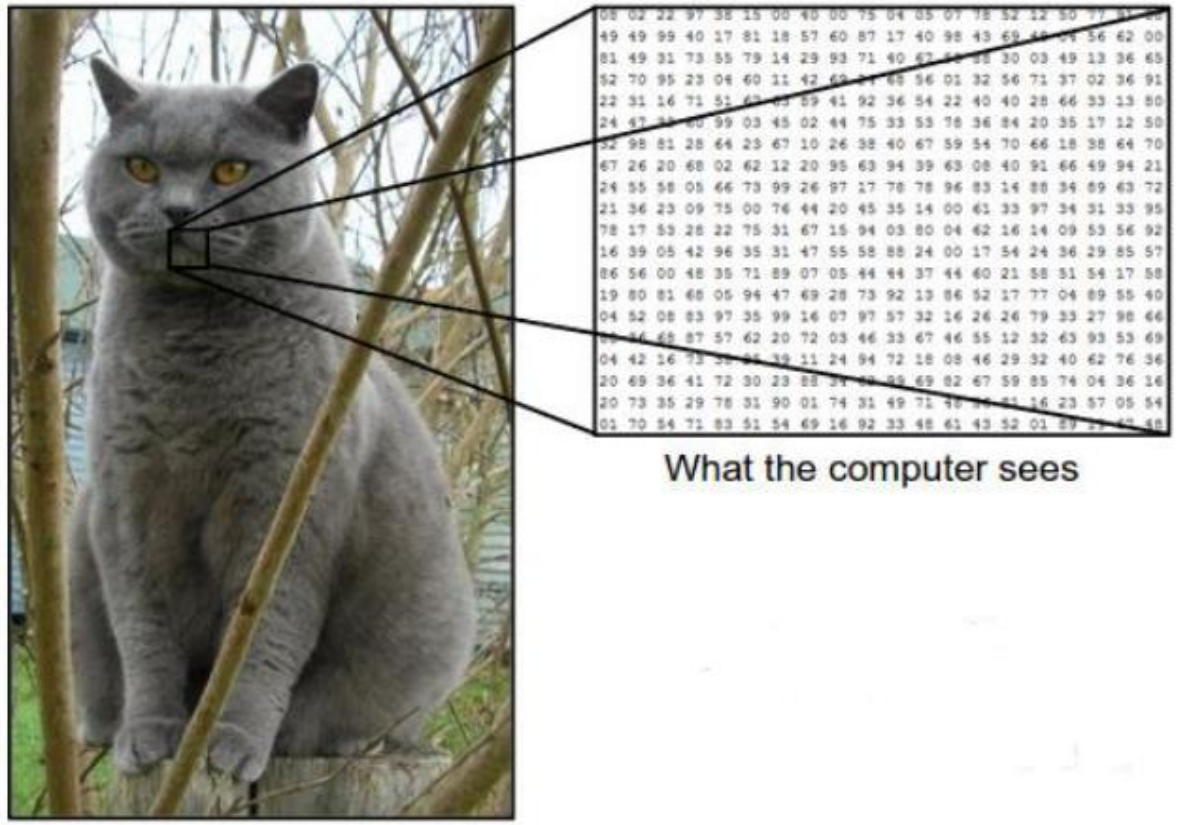
Python Libraries + Modules used in this training

All should already be installed with JupyterLab OnDemand.

Docs:

- os : https://docs.python.org/3/library/os.html
- matplotlib : https://matplotlib.org/
- scipy :
  - General : https://docs.scipy.org
  - ndimage : https://docs.scipy.org/doc/scipy/reference/ndimage.html
- skimage : https://scikit-image.org/
- sklearn : https://scikit-learn.org/stable/
- numpy :
  - General : https://numpy.org/doc/stable/index.html
  - ndarrays : https://numpy.org/doc/stable/reference/arrays.ndarray.html#id1

How a digital image is stored on a computer



Source - http://cs231n.github.io/

Images as Arrays

Different Python libraries have different array implementations

- array
- **numpy**
  - ndarray
- **openCV**
  - cv::Mat

Common data types for image pixels:

- **bool** (binary)- [0,1]
- **int8** (signed integer 8 bit) – numbers in the range: [-128 : 127]
- **float** (double-precision floating point) – Decimal numbers (e.g. 2.2251e-308, 0.4, 0.333333...)
- **uint8** (unsigned 8-bit) – [0,255]
- **uint16** (unsigned 16-bit) – [0,65535]

Python Array Indexing

- Python starts counting indexes from 0, and arranges coordinates like C does (row-major)
- Array elements can be access in two ways:
  - By forward indexing
  - By backward indexing

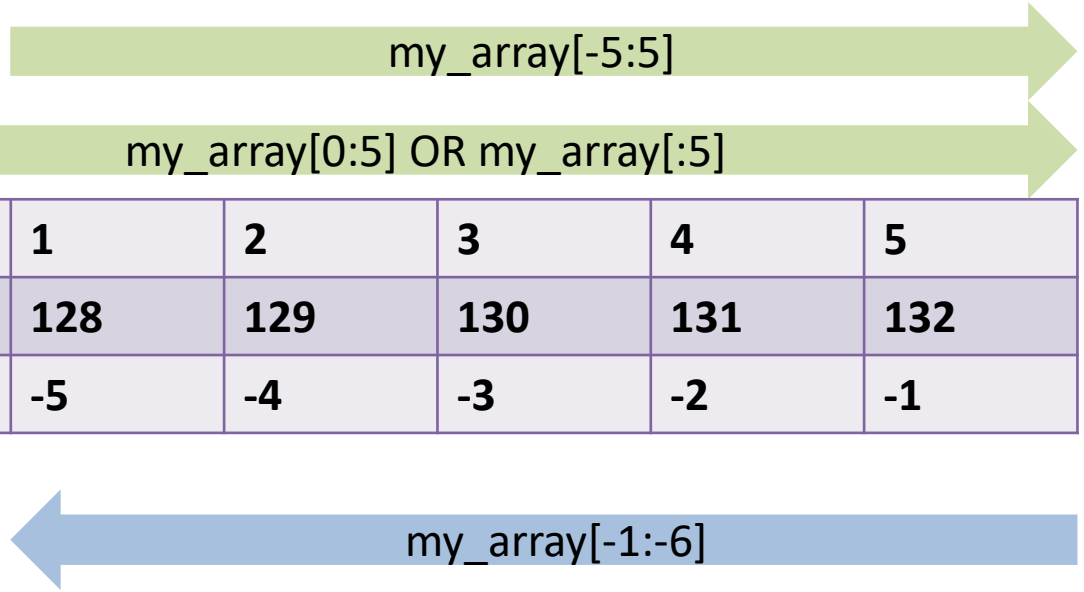
```
my_array = numpy.array([127, 128, 129, 130, 131, 132], dtype=np.int8)
```

+ index	0	1	2	3	4	5
Element	127	128	129	130	131	132
- index	-6	-5	-4	-3	-2	-1

Slice indexes are defined by [START:STOP] or [START:STOP:STRIDE]

Python Array Indexing

```
my_array = numpy.array([127, 128, 129, 130, 131, 132], dtype=np.int8)
```



+ index	0	1	2	3	4	5
Element	127	128	129	130	131	132
- index	-6	-5	-4	-3	-2	-1

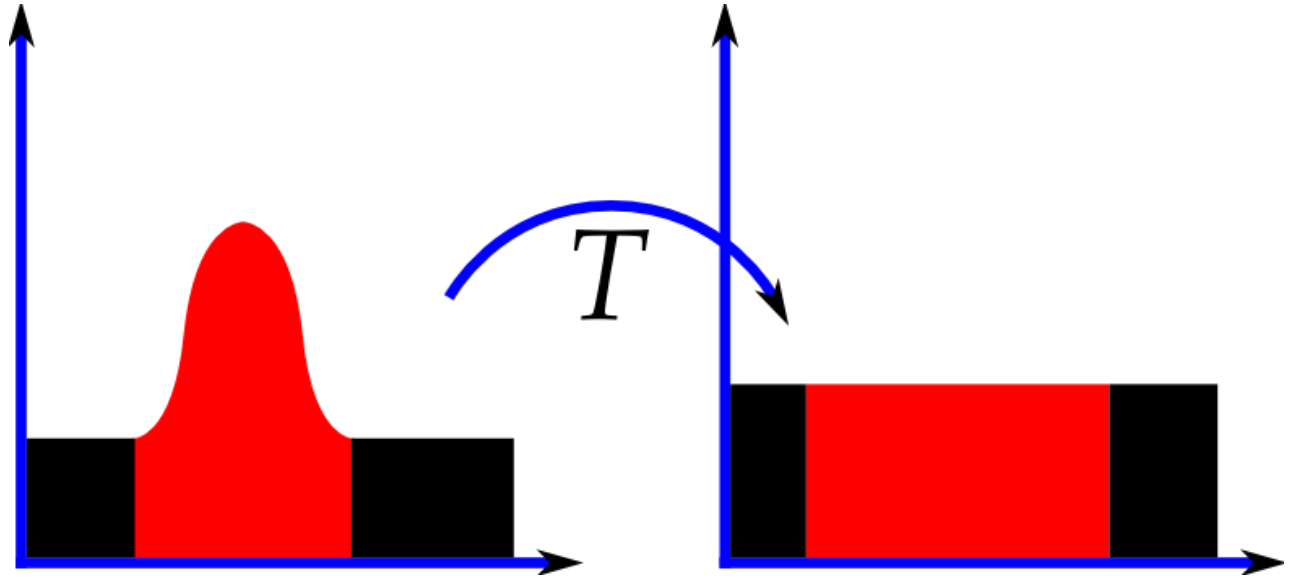
Python counts in ‘row-major’ ordering, and orders dimensions like C does.
- Multidimensional arrays are ‘lists of lists


```
my_2D_array = numpy.array([[127, 128, 129],
                           [130, 131, 132]
                           [133, 134, 135])
```


```
my_2D_array[1][0:1] = [133,134]
my_2D_array[-1][:] = [133, 134, 135]
```

Intensity enhancement

- Contrast stretching
- Histogram equalization
- Adaptive equalization

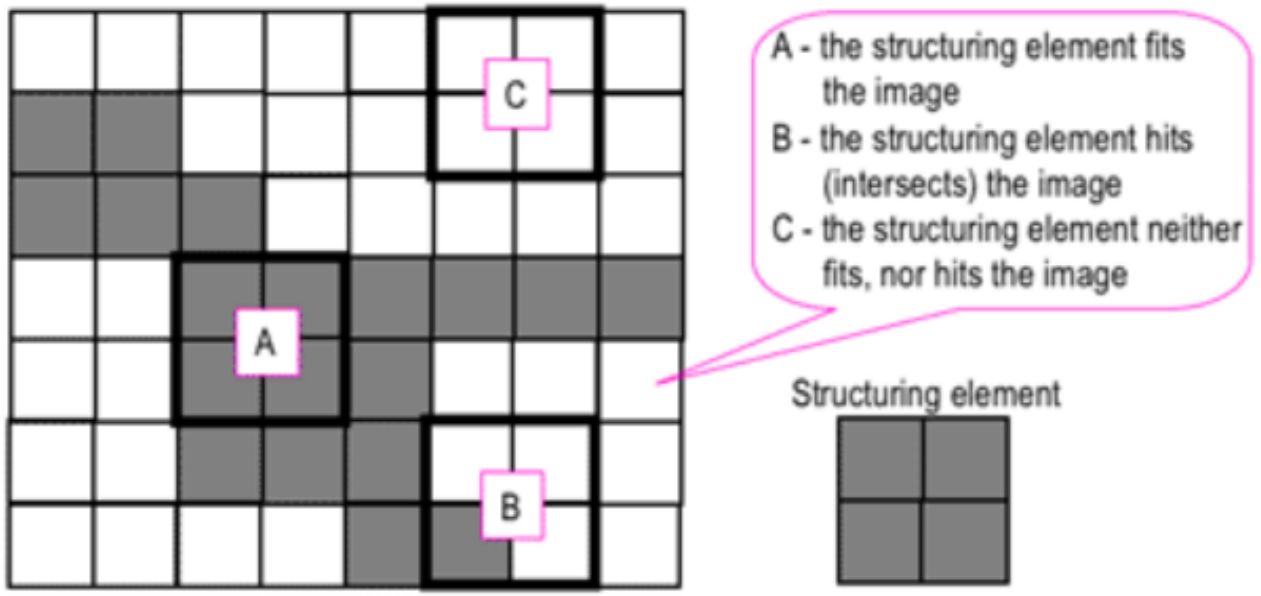


https://en.wikipedia.org/wiki/Histogram_equalization

Morphological operations: Structuring element

The structuring element is a small binary image or matrix such that:

- The matrix dimensions specify the size of the structuring element.
- The pattern of ones and zeros specifies the shape of the structuring element.

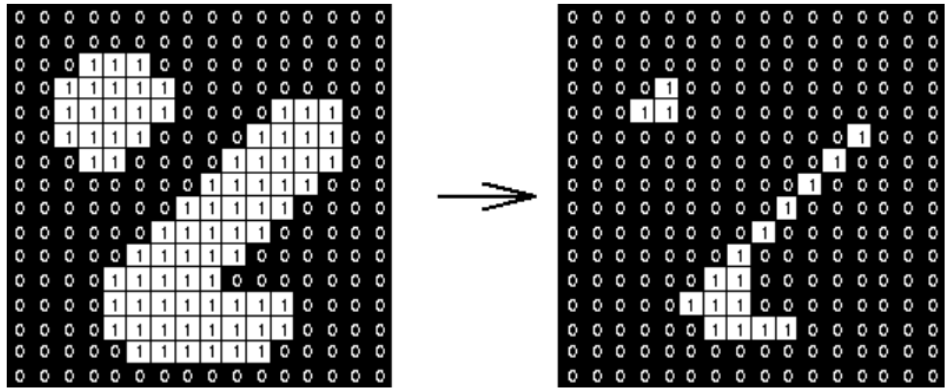


Probing of an image with a structuring element
(white and grey pixels have zero and non-zero values, respectively).

https://www.cs.auckland.ac.nz/courses/compsci773s1c/lectures/ImageProcessing-html/topic4.htm

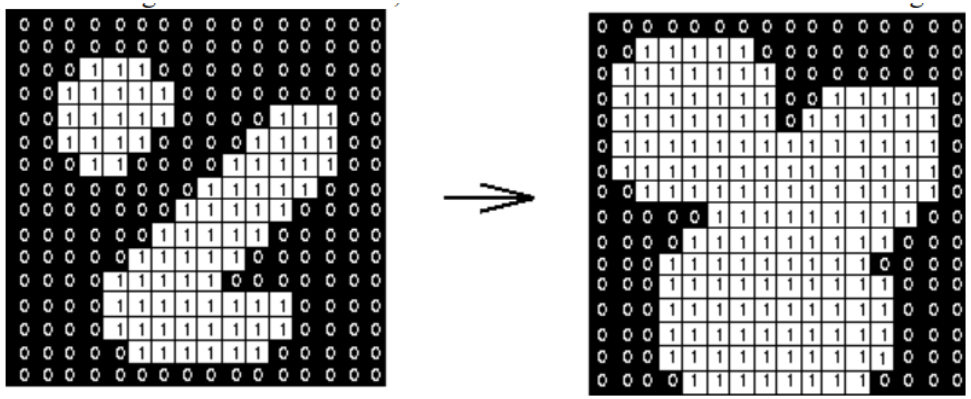
Morphological operations: Dilation and Erosion

Erosion:



Erosion: a 3×3 square structuring element
(www.cs.princeton.edu/~pshilane/class/mosaic/).

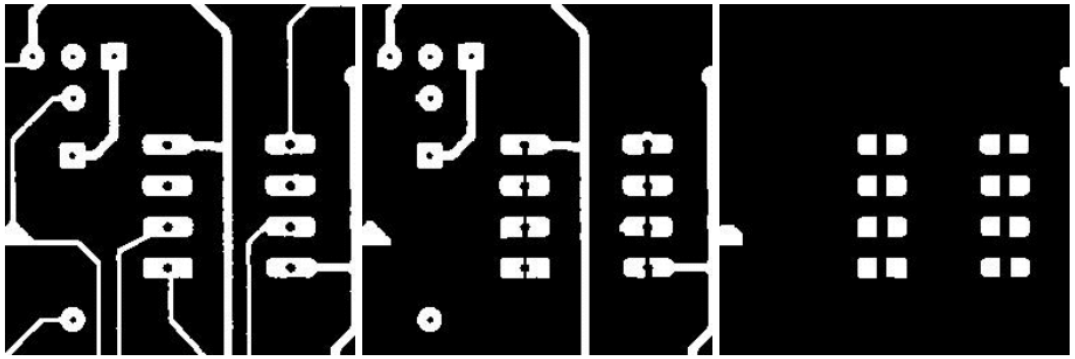
Dilation:



Dilation: a 3×3 square structuring element
(www.cs.princeton.edu/~pshilane/class/mosaic/).

Morphological operations: Open and Close

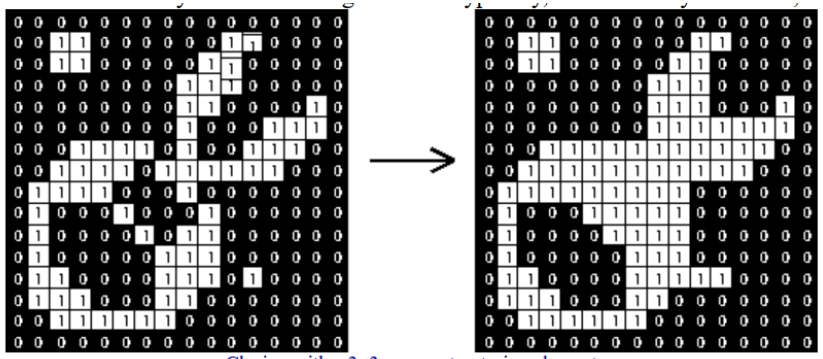
Opening: erosion followed by a dilation



Binary image f $f \circ s$ (5×5 square) $f \circ s$ (9×9 square)

Results of opening with a square structuring element (www.mmorph.com/html/morph/mmopen.html/).

Closing: dilation followed by a erosion



Closing with a 3×3 square structuring element
(www.cs.princeton.edu/~pshilane/class/mosaic/).

Mathematical Morphology - Grayscale

- Grayscale images can be treated similarly, but with a slightly modified interpretation of ‘hit or miss’
- Dilation will result in a pixel taking on the max value defined by the moving window of the strel.
- Erosion will result in a pixel taking on the min value defined by the moving window of the strel.

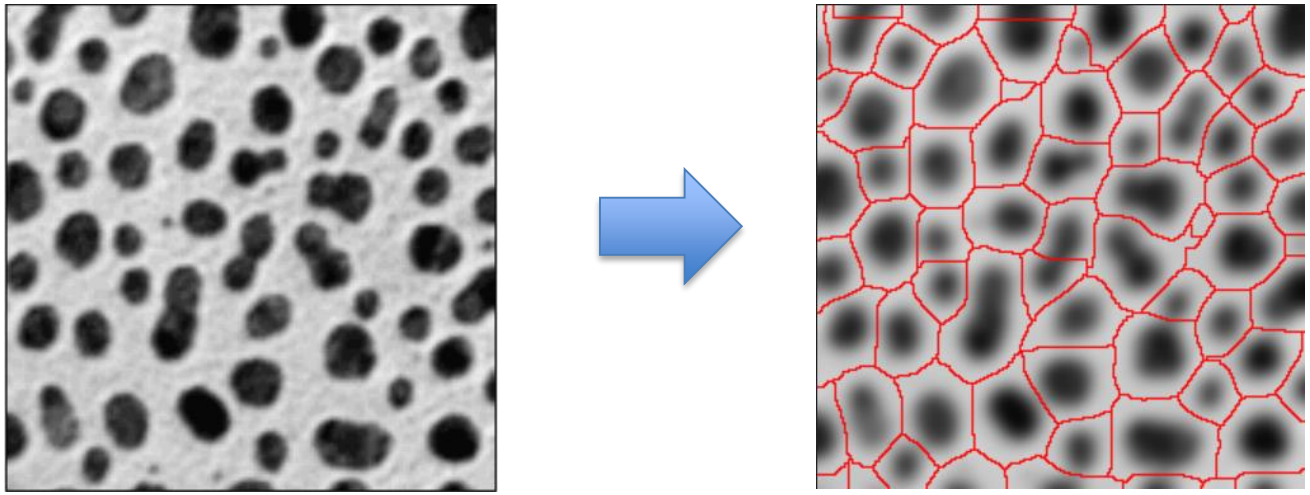
Segmentation – Separating an image into parts

Most basic: Manual thresholding

- Bright or dark background with a dark or bright foreground, respectively.
- Choose a cutoff value, threshold.
- Global thresholds can work, but can miss important elements

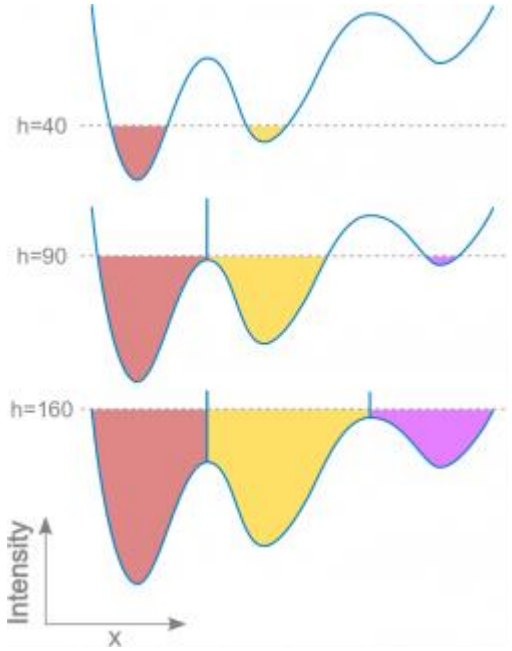
More complex:

- Adaptive thresholds
- Morphological segmentation
- Clustering
- Machine learning methods



Watershed Segmentation

- Consider grey levels as altitudes
- Identify local minima
- Flood basins starting from minima
- Separate the basins by a “dam” → the watershed



Steps for performing the watershed method:

1. Segment objects of interest
2. Convert the mask into an intensity profile using the distance transform
3. Run the watershed algorithm