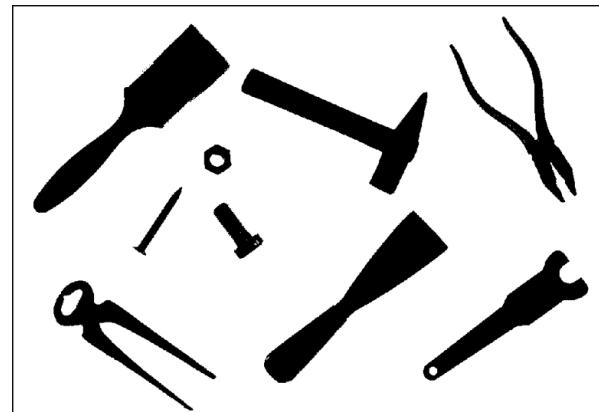


INF250

Regions and objects

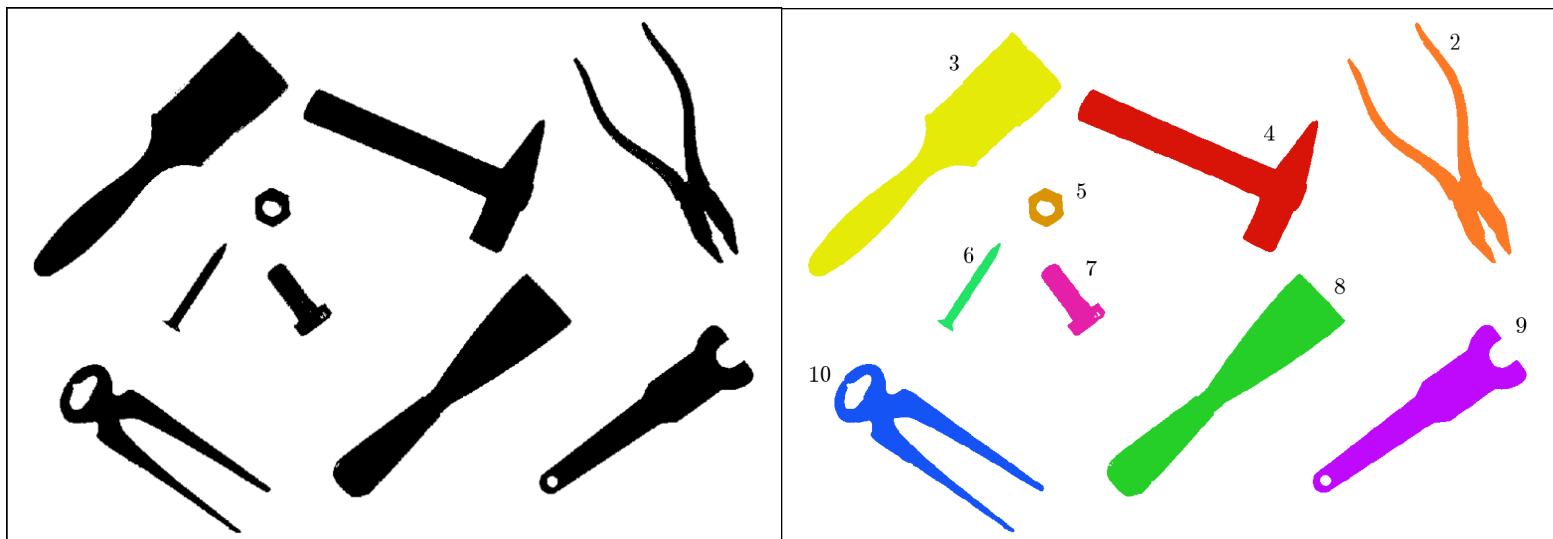
Foreground, Background

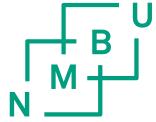
- In a binary image:
 - Pixels with values 1(255) belongs to the foreground and pixels with value 0 belongs to the background
 - Objects are part of neighboring structures or sample
- Our goal is to define objects and their shape
- Objects are defined by pixels that touch each other in coupled binary regions



Contours

1. Find object
2. Define contours of the objects





How to find regions ?

- An important task is to find which pixels defines a region and how many regions there are in the image
- Two methods are described in the book
 - «Flood Filling»
 - «Sequential region marking»

Regionlabeling



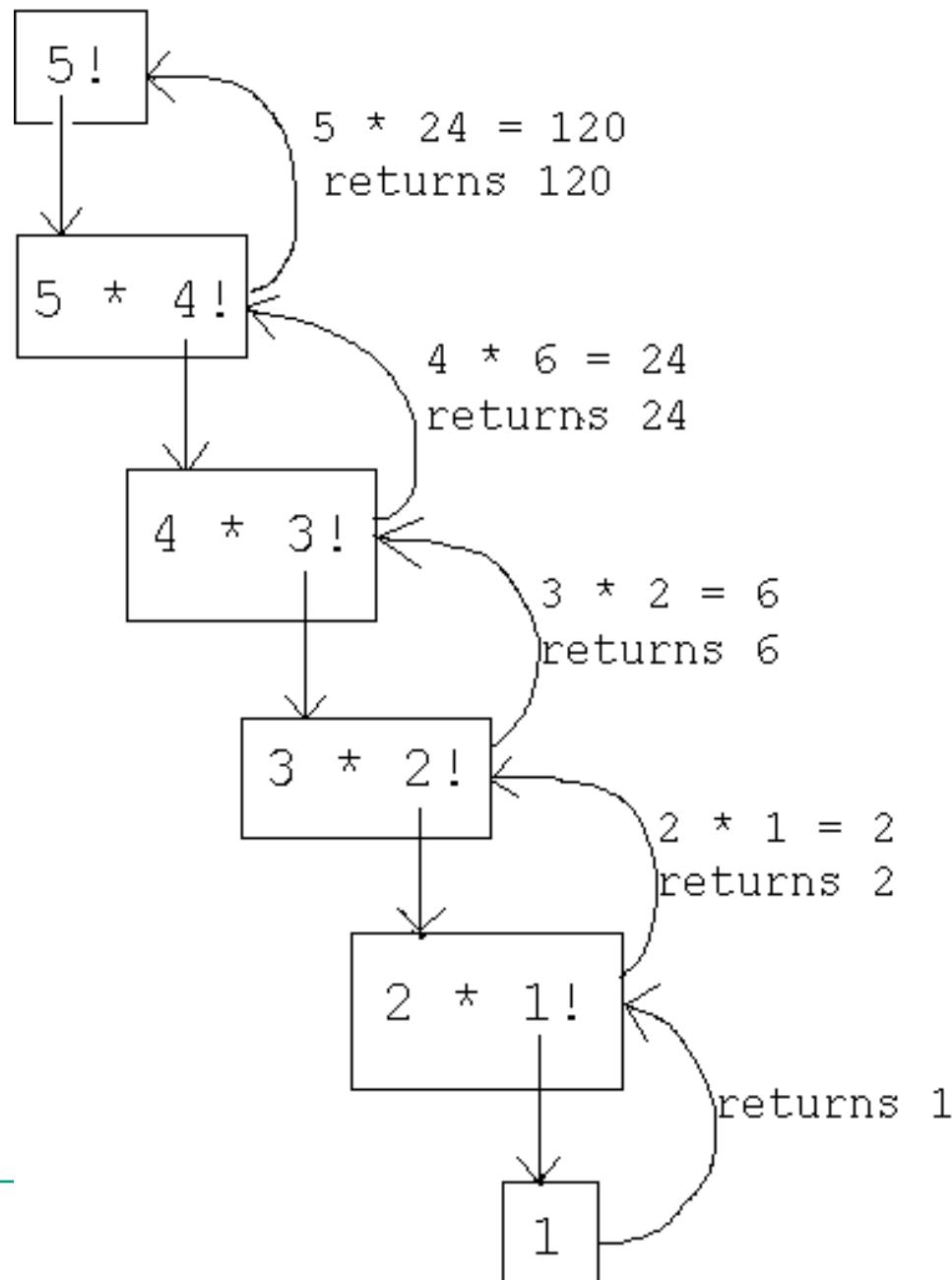
- The underlying algorithm «Flood Fill»:
 - Search for all non marked foreground pixels
 - Fill (mark) the rest of all the neighboring pixels in the region
 - Search for the next pixel outside the earlier found neighborhood
 - Analogy: Water flowing into all the neighboring pixels
 - Different algorithms exist for how the flow is defined
 - Depth-First
 - Breadth-First



Recursive function

- Recursion is programming algorithm where the function calls itself one or several times
- Usually a value is returned from a function call
- A recursive function need a stop criteria (a base) in order to avoid an endless loop
- A common example is the factorial function
 - $N!$
 - $n! = n * (n-1)!$, if $n > 1$ and $f(1) = 1$

Final value = 120



```
1 def factorial(n):
2     print("factorial has been called with n = " + str(n))
3     if n == 1:
4         return 1
5     else:
6         res = n * factorial(n-1)
7         print("intermediate result for ", n, " * factorial(", n-1, "): ", res)
8         return res
9
10 print(factorial(5))
```

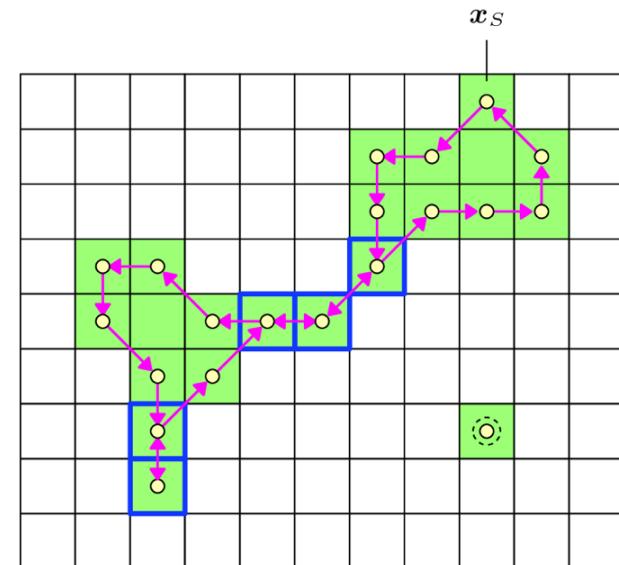
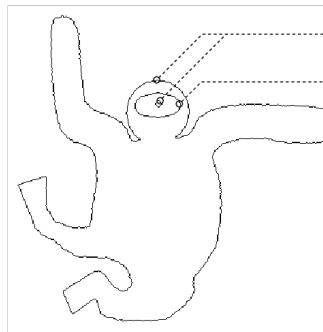
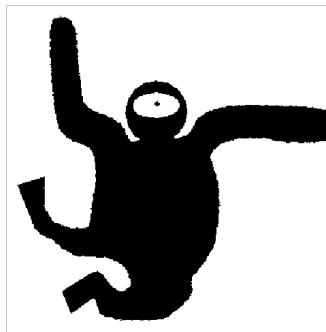
Python

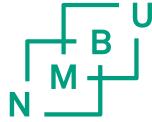
In [5]: %run /Users/kkvaal/CloudStation/ImageryPy/pyAMT/faktor.py

```
factorial has been called with n = 5
factorial has been called with n = 4
factorial has been called with n = 3
factorial has been called with n = 2
factorial has been called with n = 1
('intermediate result for ', 2, ' * factorial(', 1, '): ', 2)
('intermediate result for ', 3, ' * factorial(', 2, '): ', 6)
('intermediate result for ', 4, ' * factorial(', 3, '): ', 24)
('intermediate result for ', 5, ' * factorial(', 4, '): ', 120)
120
```

Region contours

- Inner and outer contour





Representation of areas

- Run Length Encoding
- Enkel og oversiktlig
- Benyttet i

datakompresjon

$$Run_i = \langle \text{row}_i, \text{column}_i, \text{length}_i \rangle$$

Bitmap

	0	1	2	3	4	5	6	7	8
0									
1			✗	✗	✗	✗	✗	✗	✗
2									
3					✗	✗	✗	✗	
4		✗	✗	✗		✗	✗	✗	
5	✗	✗	✗	✗	✗	✗	✗	✗	✗
6									



RLE

$$\langle \text{row}, \text{column}, \text{length} \rangle$$

$$\begin{aligned}\langle 1, 2, 6 \rangle \\ \langle 3, 4, 4 \rangle \\ \langle 4, 1, 3 \rangle \\ \langle 4, 5, 3 \rangle \\ \langle 5, 0, 9 \rangle\end{aligned}$$

Chain Code

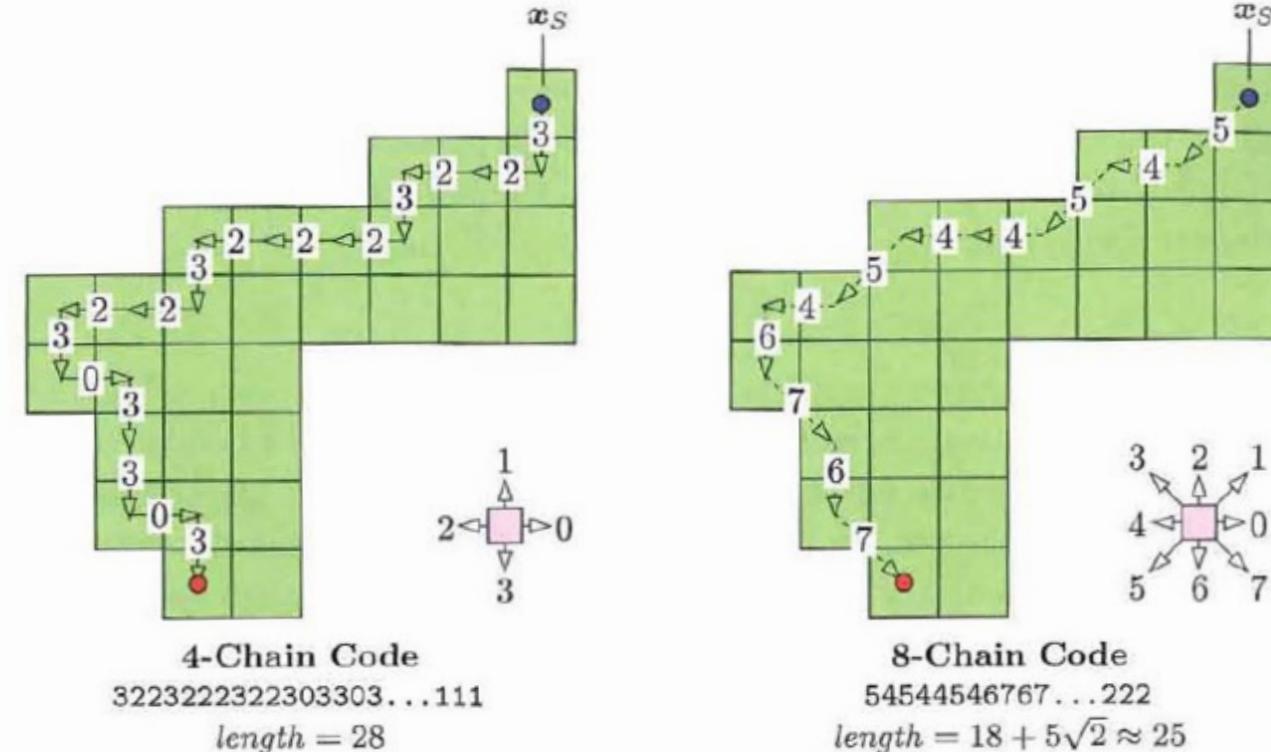
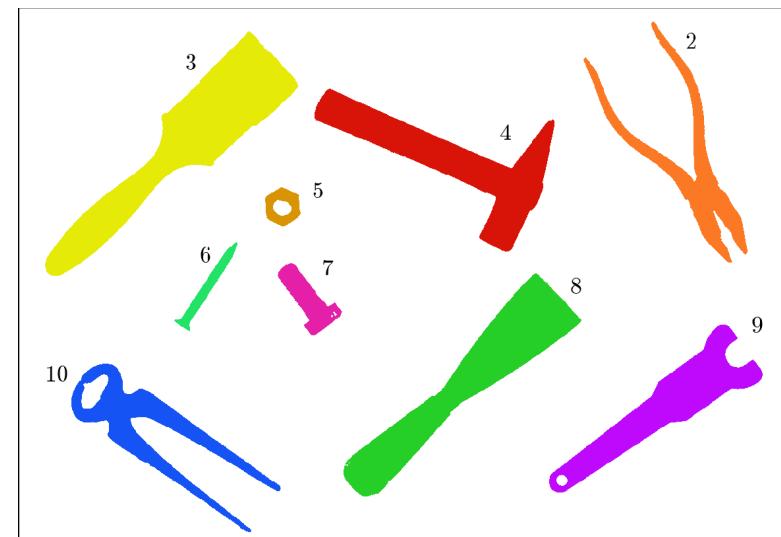


Figure 2.14 Chain codes with 4- and 8-connected neighborhoods. To compute a chain code, begin traversing the contour from a given starting point α_S . Encode the relative position between adjacent contour points using the directional code for either 4-connected (left) or 8-connected (right) neighborhoods. The length of the resulting path, calculated as the sum of the individual segments, can be used to approximate the true length of the contour.

Contours and statistics

- After all regions and contours are found the statistics of the regions can be computed

Label	Area (pixels)	Bounding Box (left, top, right, bottom)	Center (x_c, y_c)
2	14978	(887, 21, 1144, 399)	(1049.7, 242.8)
3	36156	(40, 37, 438, 419)	(261.9, 209.5)
4	25904	(464, 126, 841, 382)	(680.6, 240.6)
5	2024	(387, 281, 442, 341)	(414.2, 310.6)
6	2293	(244, 367, 342, 506)	(294.4, 439.0)
7	4394	(406, 400, 507, 512)	(454.1, 457.3)
8	29777	(510, 416, 883, 765)	(704.9, 583.9)
9	20724	(833, 497, 1168, 759)	(1016.0, 624.1)
10	16566	(82, 558, 411, 821)	(208.7, 661.6)





Geometrical features

- Perimeter: $P(R) = \sum_{i=0}^{M-1} \text{length}(C_i)$



$$\begin{aligned} &= 1 \text{ when } c=0,2,4,6 \\ &= \sqrt{2} \text{ when } c=1,3,5,7 \end{aligned}$$

- Area: $A(R) = N$ (number of pixels)

- Circularity $C(R) = 4\pi \cdot \frac{A(R)}{P^2(R)}$

Circularity(sirkularitet)

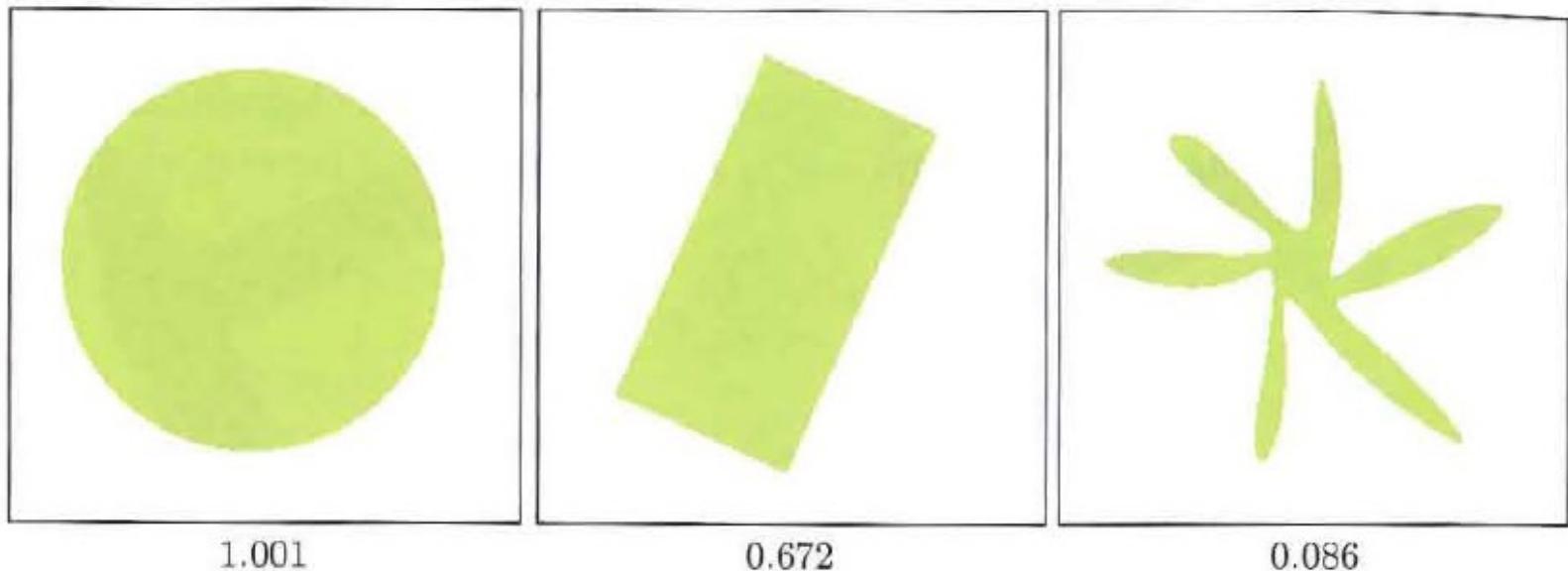
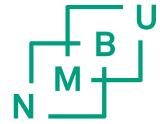
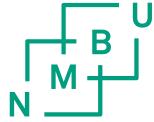


Figure 2.15 Circularity values for different shapes. Shown are the corresponding estimates for $\text{Circularity}(\mathcal{R})$ as defined in Eqn. (2.12).



Circularity vs Roundness

There is a difference between the terms «Circularity» and «Roundness» (not described in the book)

«Circularity» is computed from the perimeter whereas
«Roundness» is computed from the major axis

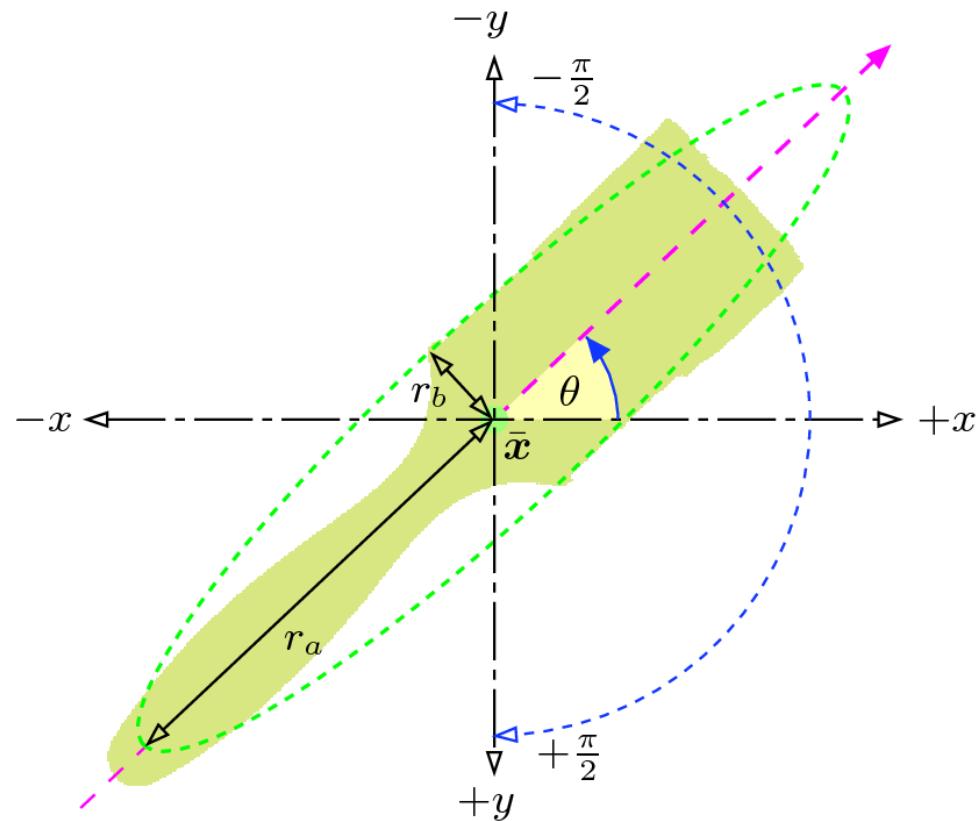
$$Circularity = 4\pi \cdot \frac{Area}{Perimeter^2}$$

$$Roundness = 4 \cdot \frac{Area}{\pi * (Major_axis)^2}$$

Other statistical shape descriptors



- Centroid
- Moments
- Central moments
- Directions
-



Bounding box og Convex Hull

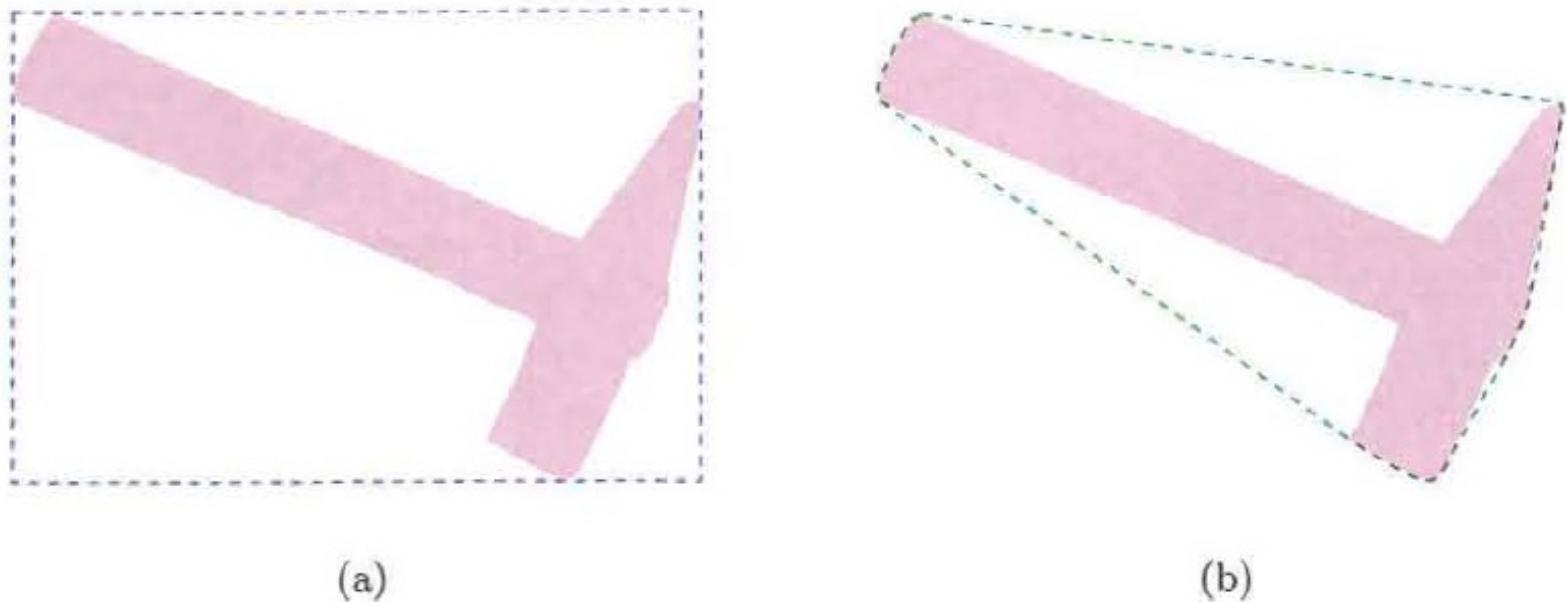


Figure 2.16 Example bounding box (a) and convex hull (b) of a binary image region.

Projections

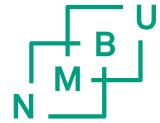


Figure 2.20 Example of the horizontal projection $P_{\text{hor}}(v)$ (right) and vertical projection $P_{\text{ver}}(u)$ (bottom) of a binary image.

Image moments

Moments summarize a shape given image $I(x, y)$:

$$M_{ij} = \sum_x \sum_y x^i y^j I(x, y)$$

Central moments are translation invariant:

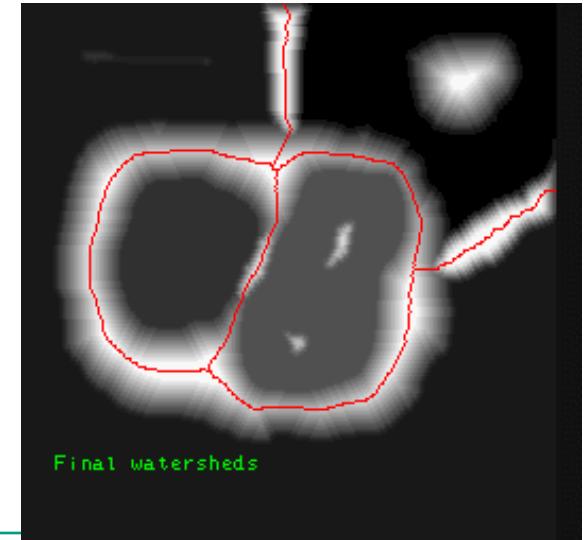
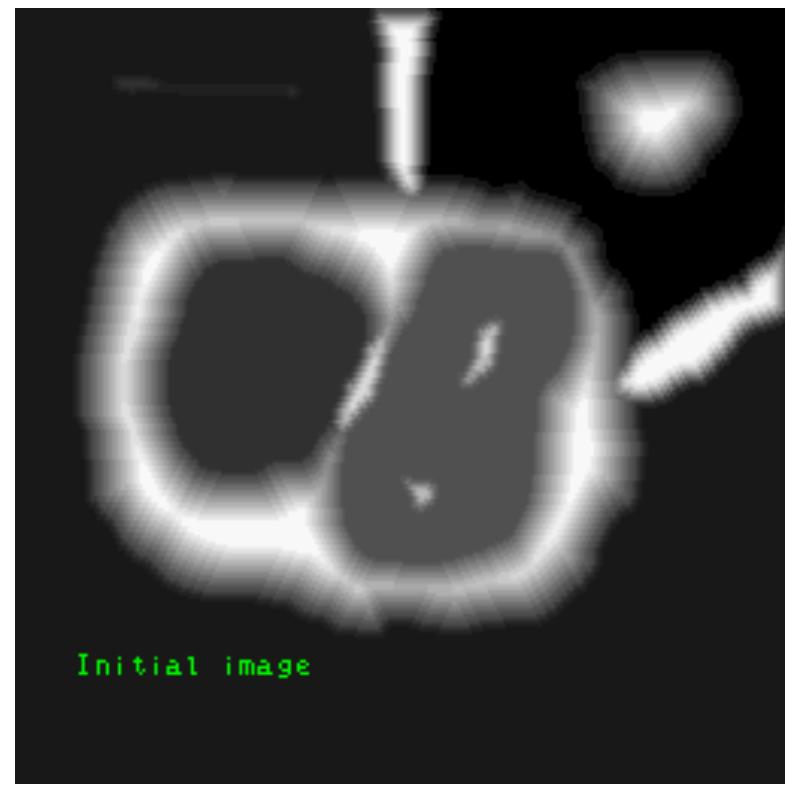
$$\mu_{pq} = \sum_x \sum_y (x - \bar{x})^p (y - \bar{y})^q I(x, y)$$

$$\bar{x} = \frac{M_{10}}{M_{00}} \quad \bar{y} = \frac{M_{01}}{M_{00}}$$

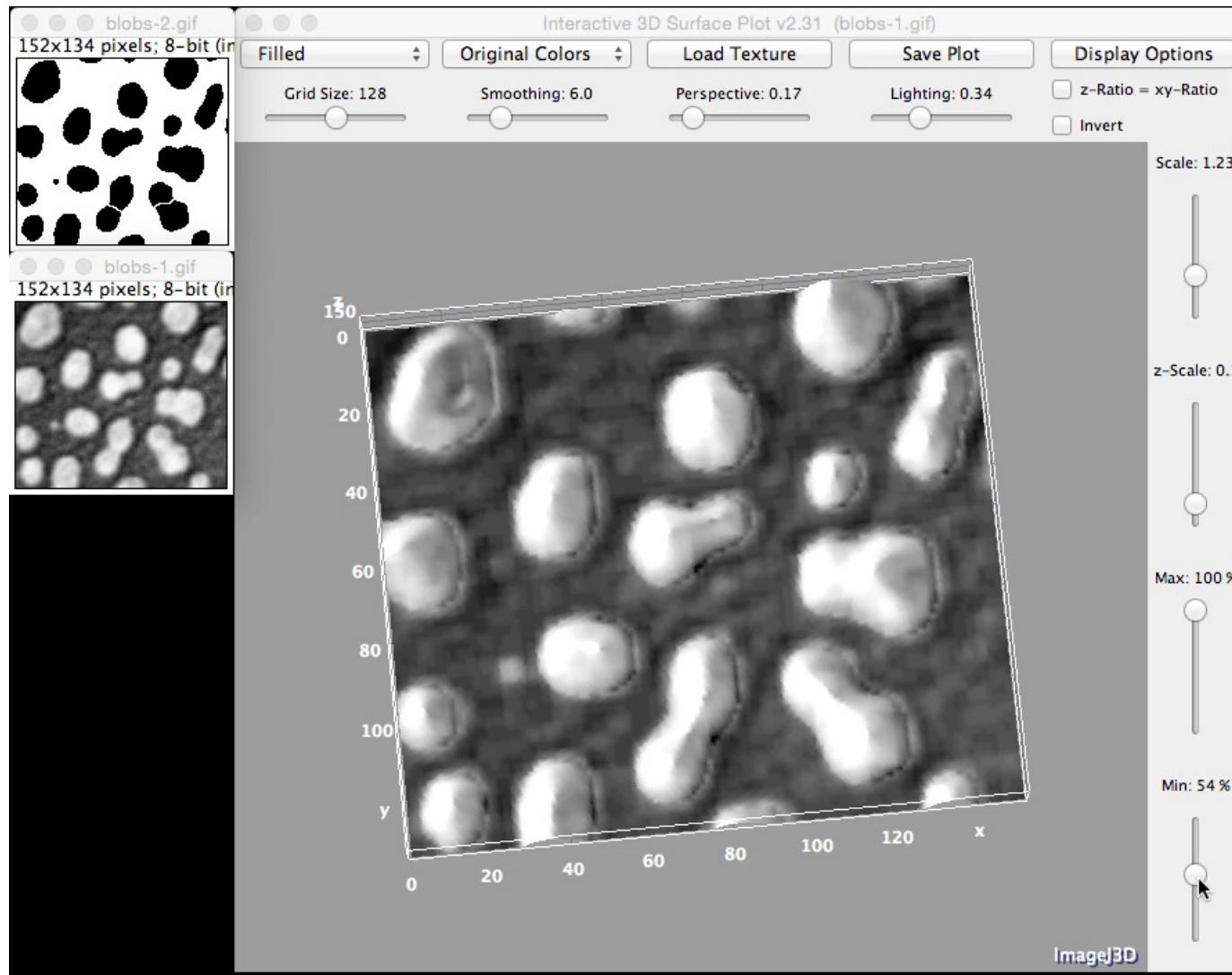
https://www.youtube.com/watch?v=AAbUfZD_09s

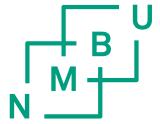
Watershed - segmentation

- Grey level images can be viewed as a topologic map
- Binary images can be transformed to a topologic map by using the distance from the perimeter to the center as a weight
- Morphologic operations
 - Fill water from a minimum point and build a shed so that the water doesn't flood inn from other areas
- This gives us two dataset:
 - Water basins
 - Floodlines



Watershed animation



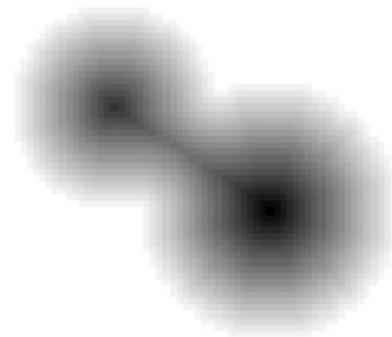


Watershed - Python

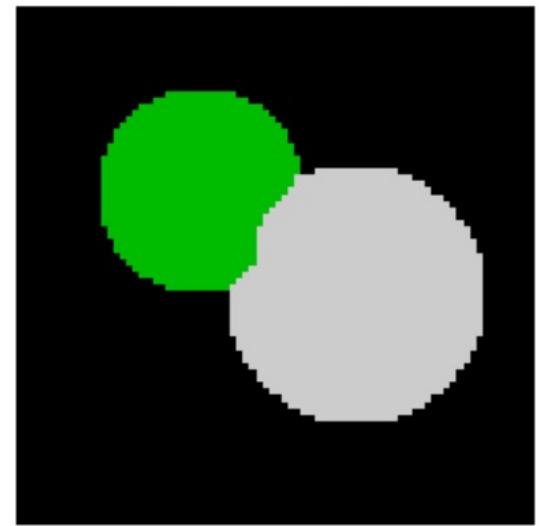
Overlapping objects



Distances



Separated objects



http://www.scipy-lectures.org/advanced/image_processing/auto_examples/plot_watershed_segmentation.html

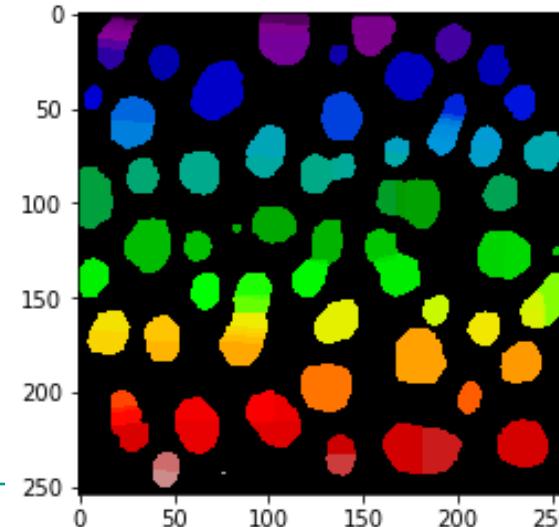
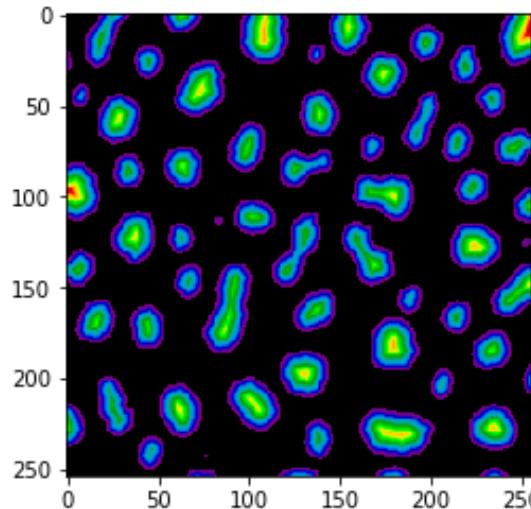
```
from skimage.morphology import watershed
import matplotlib.pyplot as plt
import numpy as np
from skimage.feature import peak_local_max
from scipy import ndimage as ndi
from skimage import io

filename = 'blobs.tif'

blob = io.imread(filename)
plt.imshow(blob, 'gray')

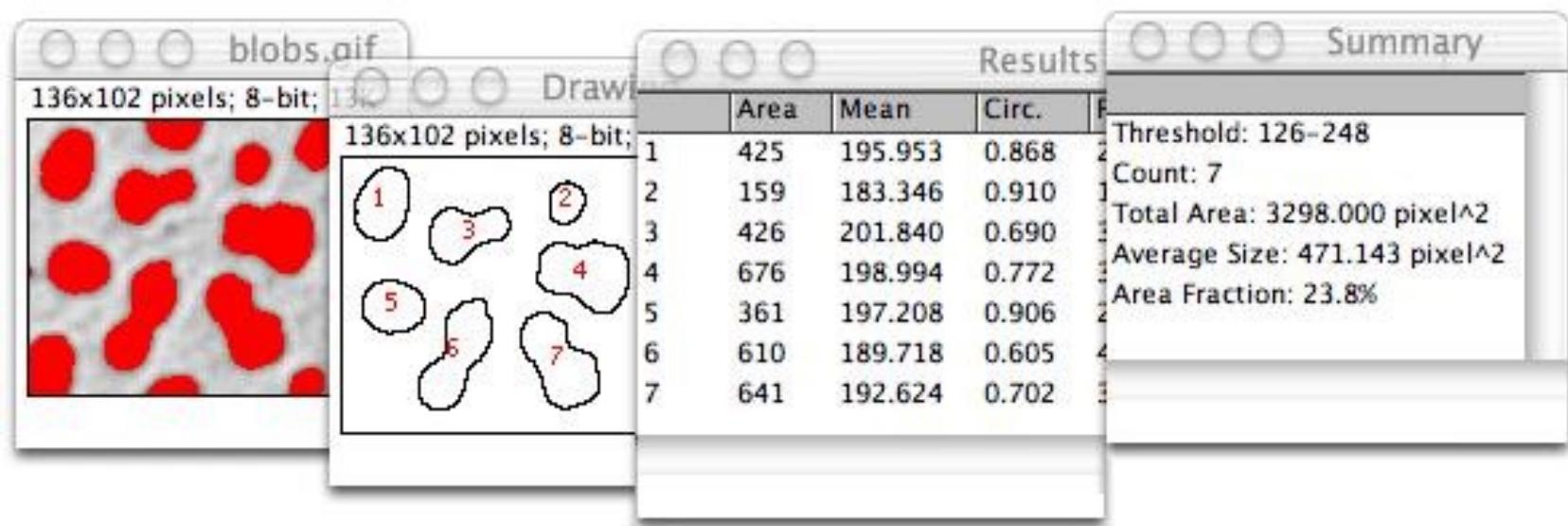
distance = ndi.distance_transform_edt(blob)
local_maxi = peak_local_max(distance, indices=False, footprint=np.ones((3, 3)),
                             labels=blob)
markers = ndi.label(local_maxi)[0]
labels = watershed(-distance, markers, mask=blob)

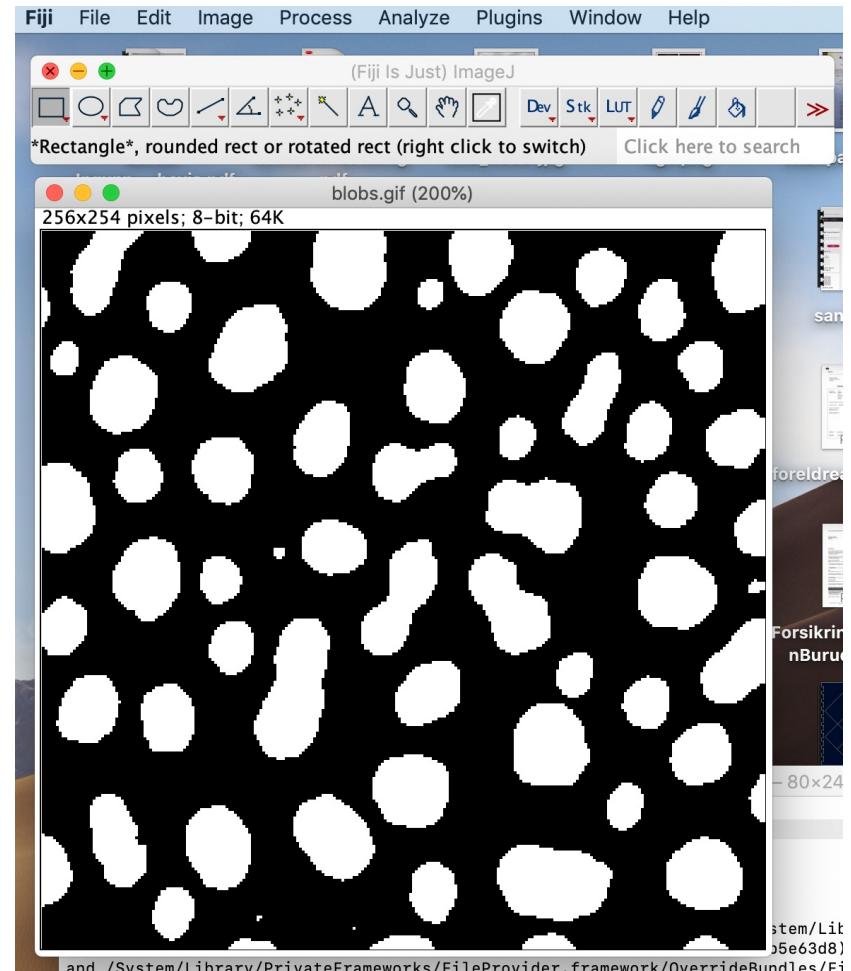
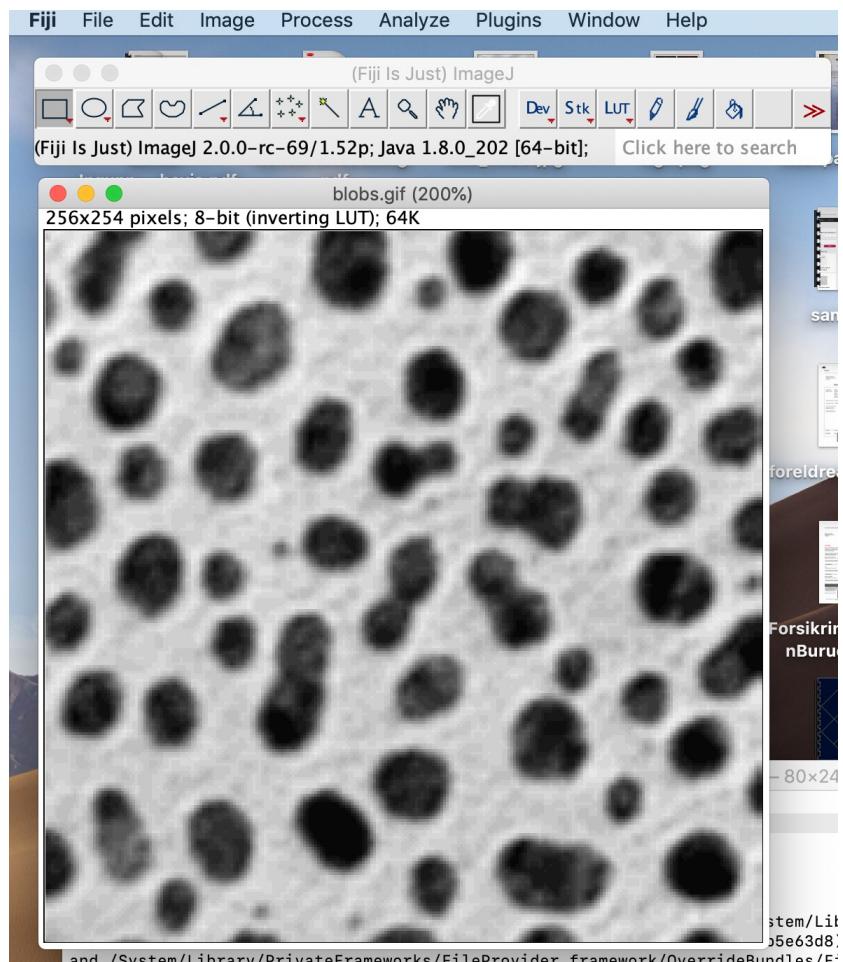
plt.imshow(labels, 'spectral')
```

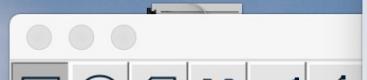


Analyzis of particles

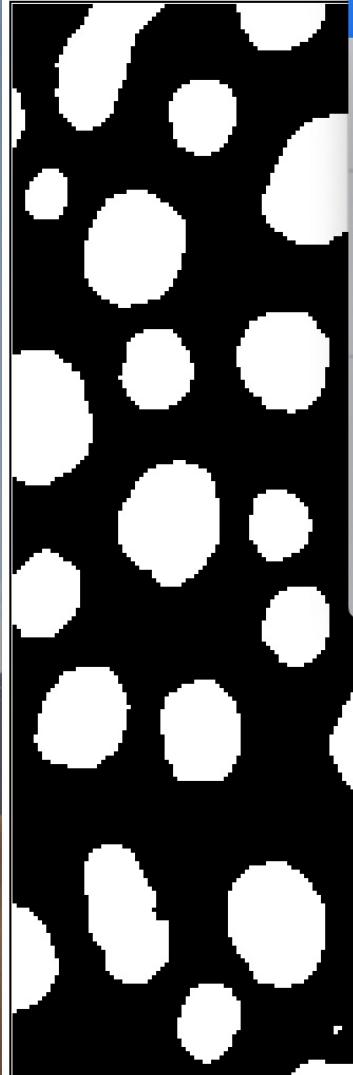
http://imagejdocu.tudor.lu/doku.php?id=gui:analyze:analyze_particles



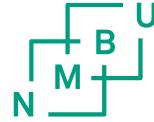
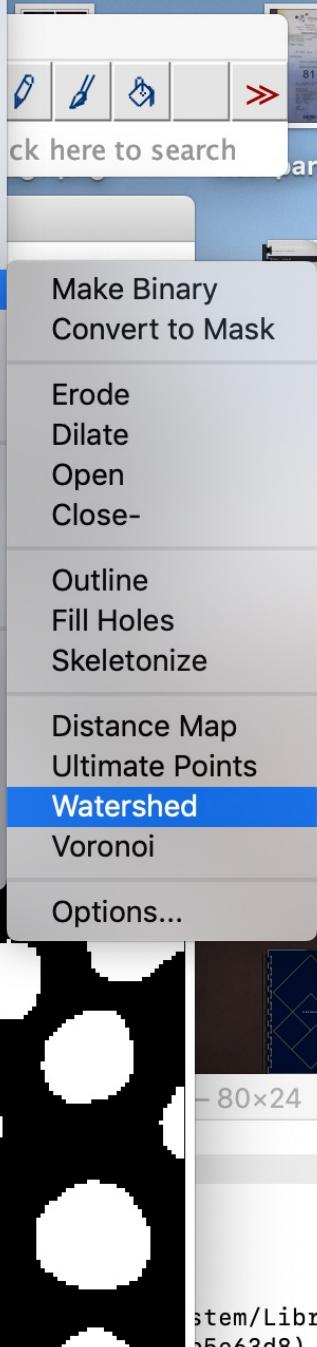


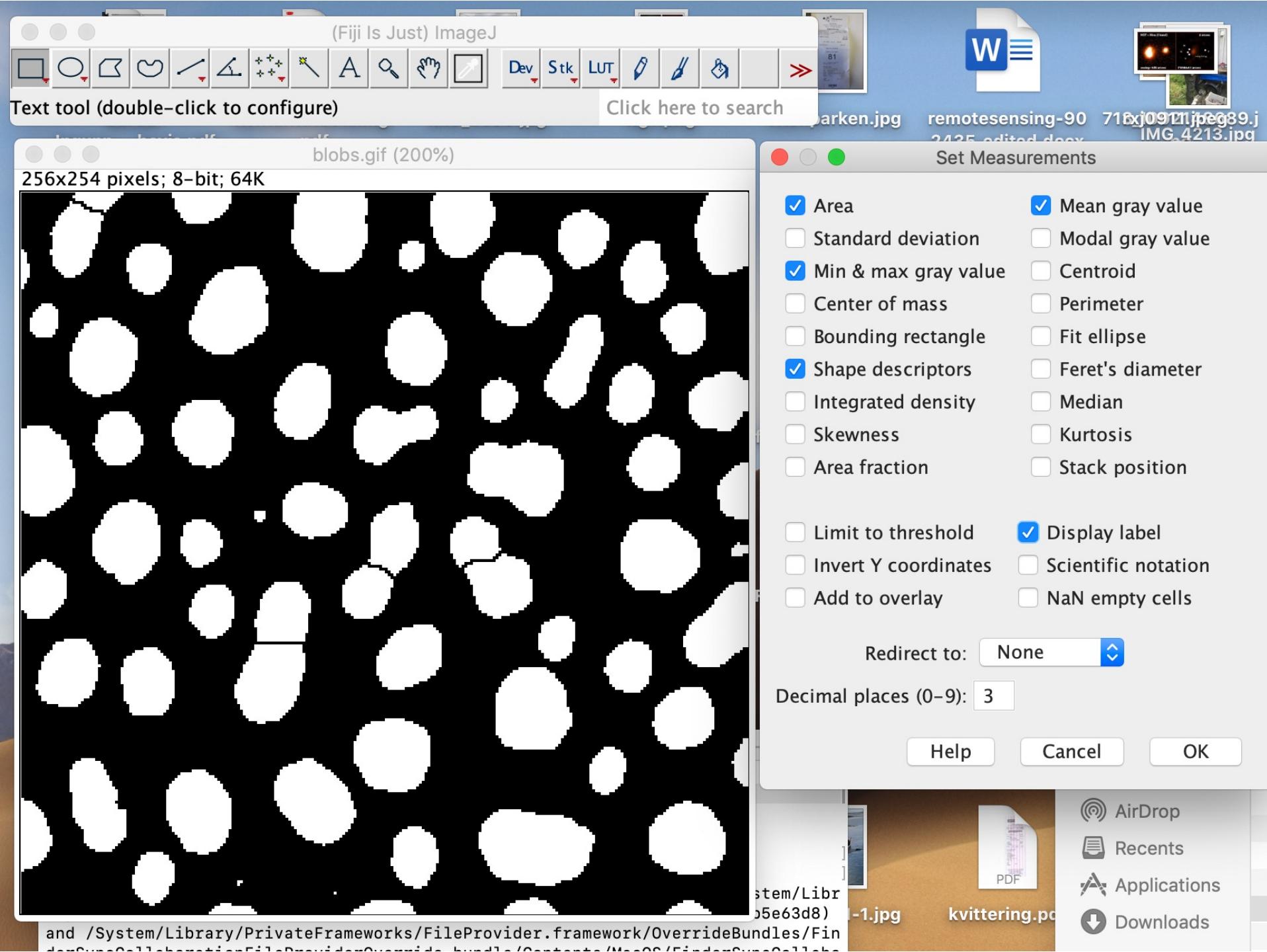


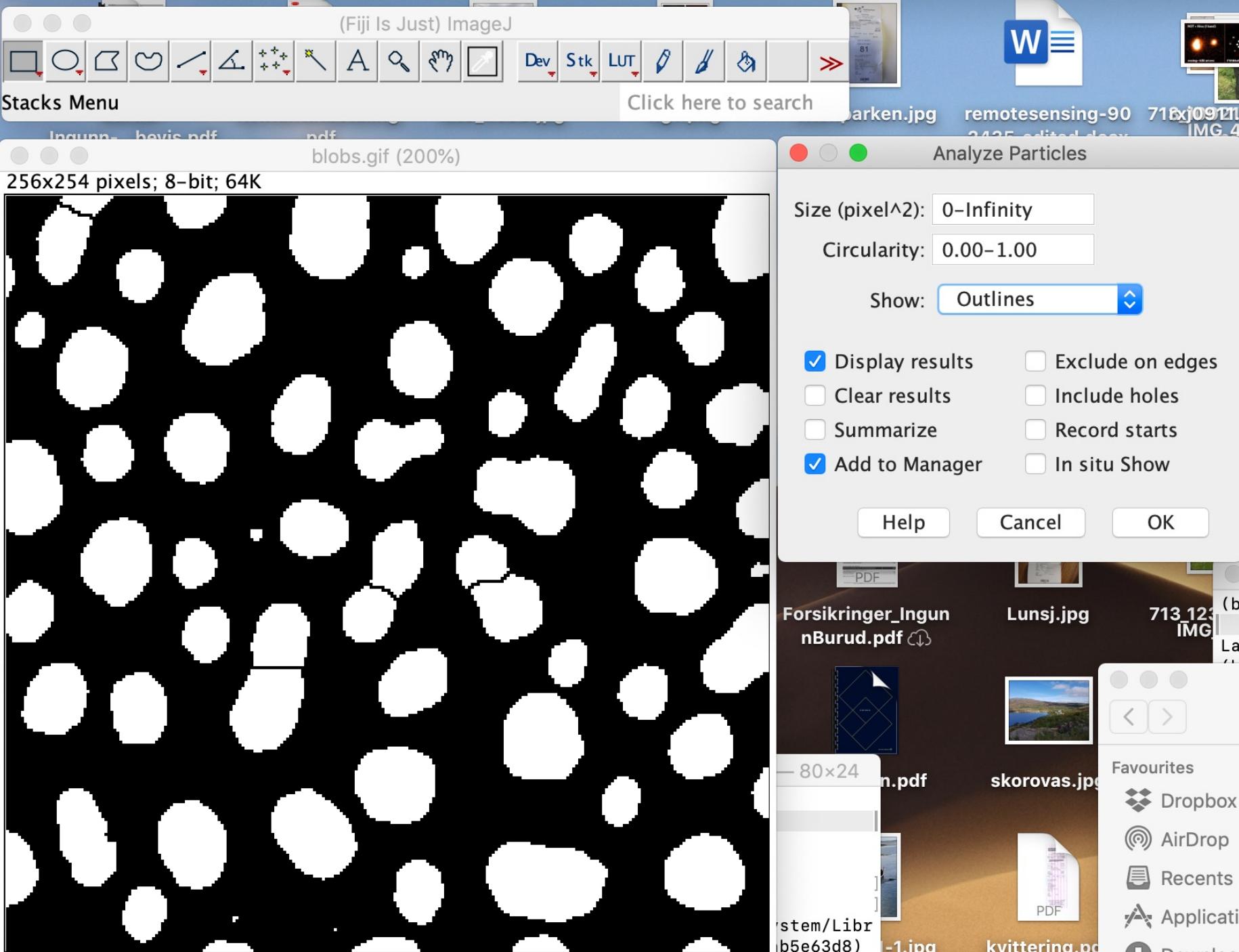
Text tool (double-click to



- Smooth ⌘⌘S
- Sharpen
- Find Edges
- Find Maxima...
- Enhance Contrast...
- Noise
- Shadows
- Binary**
- Math
- FFT
- Filters
- Batch
- Image Calculator...
- Subtract Background...
- Repeat Command ⌘⌘R
- Calculator Plus
- Morphology
- Image Expression Parser
- Image Expression Parser (Macro)
- Multiple Image Processor
- Enhance Local Contrast (CLAHE)









(Fiji Is Just) ImageJ 2.0.0-rc-69/1.52p; Java 1.8.0_202 [64-bit];

Click here to search

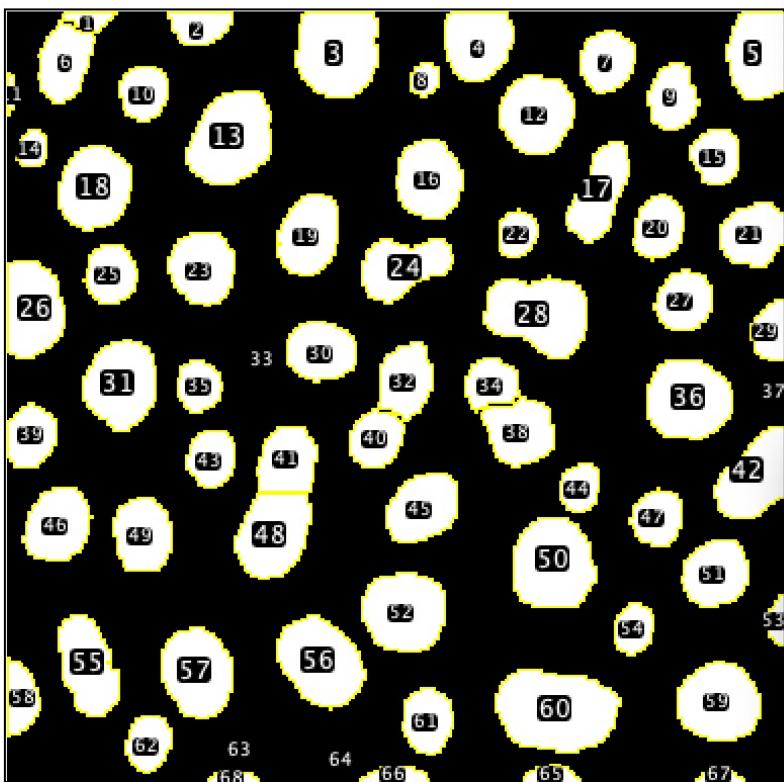
Burud-

Kontantutbetaling.

IMG 2590.jpg

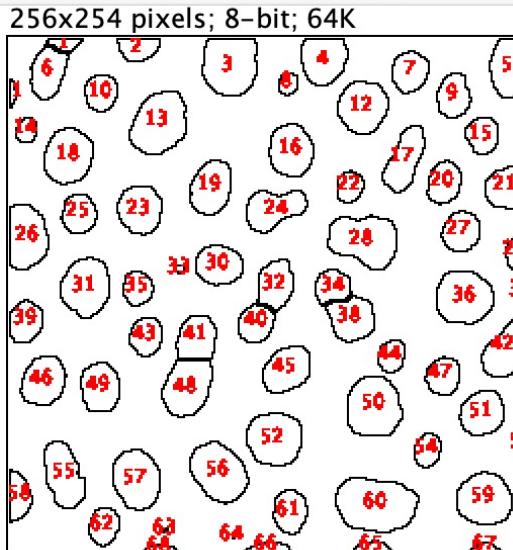
fig

Results

blobs.gif (150%)
256x254 pixels; 8-bit; 64K

	Area	Mean	Min	Max	Circ.	AR	Round	Solidity
1	76	255	255	255	0.562	2.522	0.396	0.884
2	185	255	255	255	0.763	1.778	0.562	0.951
3	658	255	255	255	0.872	1.068	0.936	0.967
4	434	255	255	255	0.884	1.064	0.940	0.959
5	477	255	255	255	0.821	1.570	0.637	0.968
6	341	255	255	255	0.757	1.603	0.624	0.905
7	285	255	255	255	0.926	1.153	0.867	0.938
8	81	255	255	255	0.971	1.204	0.830	0.926
9	278	255	255	255	0.854	1.389	0.720	0.919
10	231	255	255	255	0.936	1.141	0.877	0.941
11	30	255	255	255	0.406	4.338	0.230	0.870

Drawing of blobs.gif



ROI Manager

- 0001-0003
 - 0002-0005
 - 0003-0014
 - 0004-0011
 - 0005-0014
 - 0006-0016
 - 0007-0016
 - 0008-0022
 - 0009-0028
 - 0010-0027
 - 0011-0027
 - 0012-0034
 - 0013-0041
 - 0014-0045
 - 0015-0048
 - 0016-0055
 - 0017-0059
 - 0018-0058
 - 0019-0073
- Show All
 Labels

timeliste_signed.p

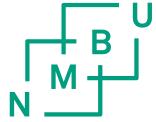
df

Ingr



iCloud Drive

Record



Python - measure

```
from skimage import measure
properties = measure.regionprops(labels)
for prop in properties:
    print(prop.perimeter, prop.area, prop.centroid, prop.eccentricity)
```

<http://scikit-image.org/docs/dev/api/skimage.measure.html#skimage.measure.regionprops>

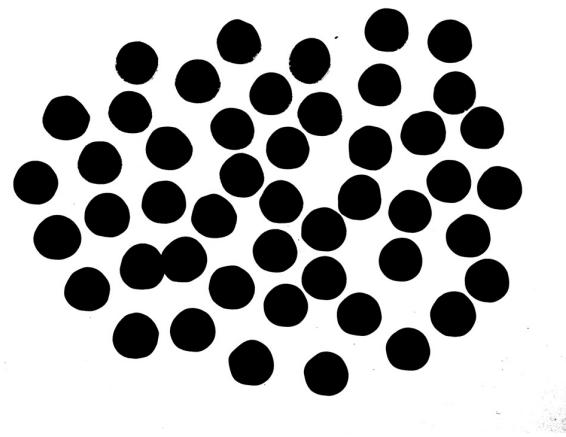
Watershed - Imagej



- Watershed on gray level images
 - <http://bigwww.epfl.ch/sage/soft/watershed/>



Masks



Making a binary
(holes=0, background=1)

Masking original image by
combining original OR mask