segmentationExample

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1 Segmentation Example

A really nice example from sci-py lecture notes.

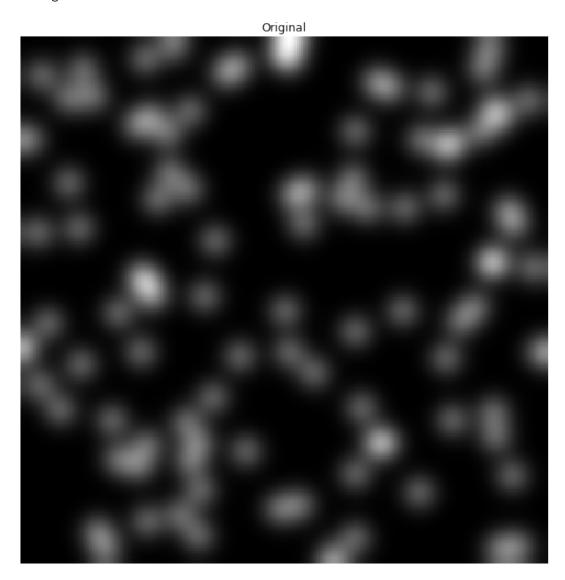
First, let's set up the imports and define a couple of custom display functions.

```
In [1]: %matplotlib inline
        import numpy as np
        from scipy import ndimage
        import matplotlib.pyplot as plt
        from sklearn.feature_extraction import image
        from sklearn.cluster import spectral_clustering
        import warnings
        warnings.filterwarnings('ignore')
        def plotOneImage(dat, label, sFont=12, size=(10,10), colormap=plt.cm.gray):
            """plotOneImage(dat, label, sizeFont=12, size=(10,10), colormap=plt.cm.gray)
            Use matplotlib to plot the image
            Input:
            dat
                     - a numpy array with the data
                     - a title
            l.a.b.e.l.
                    - the size of the font, default 12
                    - a tuple, default (10,10) with the figure size
            colormap - a pyplot color map, default plt.cm.qray
            Return:
            a matplotlib figure"""
            fig, axes = plt.subplots(ncols=1, nrows=1, figsize=size)
            ax0.imshow(dat, cmap=colormap, interpolation='nearest')
            ax0.set_title(label, fontsize=sFont)
            ax0.axis('off')
            fig.show()
            return(fig)
```

1.1 First, generate the image.

```
In [2]: np.random.seed(42) # Hey, I like the Hitchhiker's Guide to the Universe...
    n = 10
    1 = 256
    im = np.zeros((1, 1))
    points = l*np.random.random((2, n**2))
    im[(points[0]).astype(np.int), (points[1]).astype(np.int)] = 1
    im = ndimage.gaussian_filter(im, sigma=l/(4.*n))
    mask = (im > im.mean()).astype(np.float)
```

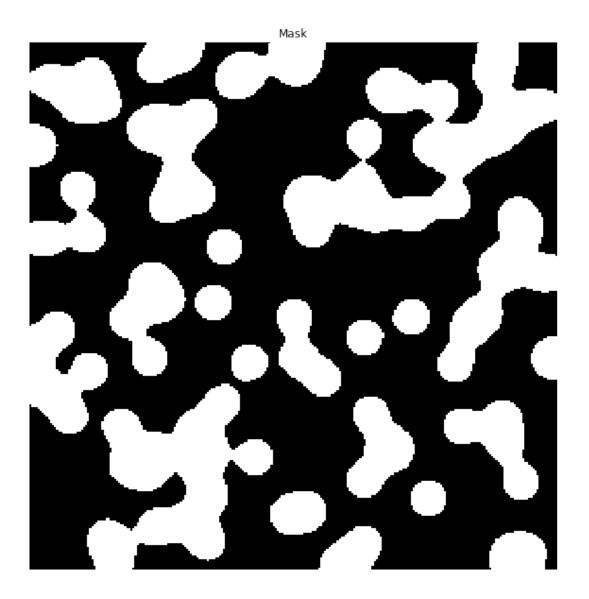
```
mask += 0.1 * im
img = mask + 0.2*np.random.randn(*mask.shape)
```



1.2 Generate touching blobs.

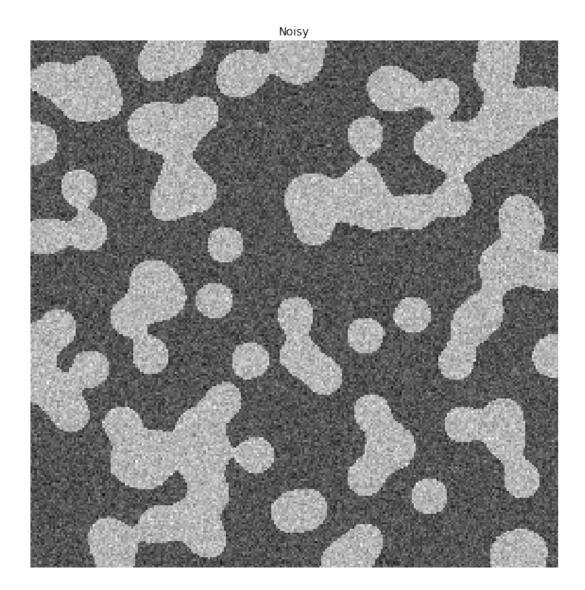
Select the mean gray level as a threshold and create a mask image with white blobs.

```
In [4]: mask = (im > im.mean()).astype(np.float)
    mask += 0.1 * im
    mFig = plotOneImage(mask, "Mask", sFont=12, size=(10,10), colormap=plt.cm.gray)
    mFig.show()
```



Now create a noisy image.

```
In [5]: img = mask + 0.2*np.random.randn(*mask.shape)
    nFig = plotOneImage(img, "Noisy", sFont=12, size=(10,10), colormap=plt.cm.gray)
    nFig.show()
```

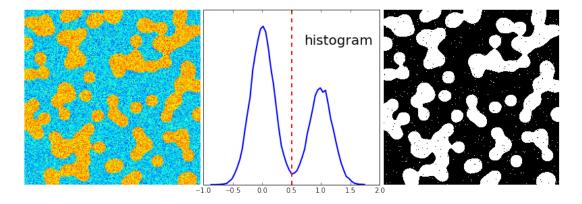


Now compute the histogram and binarize the image.

Finally make a nice side by side plot.

```
In [7]: plt.figure(figsize=(11,4))
        plt.subplot(131)
        plt.imshow(img)
        plt.axis('off')
        plt.subplot(132)
        plt.plot(bin_centers, hist, lw=2)
        plt.axvline(0.5, color='r', ls='--', lw=2)
        plt.text(0.57, 0.8, 'histogram', fontsize=20, transform = plt.gca().transAxes)
```

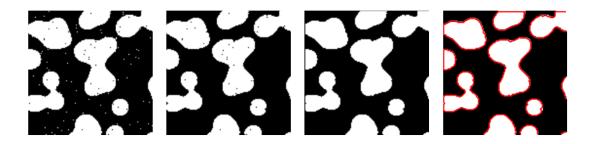
```
plt.yticks([])
plt.subplot(133)
plt.imshow(binary_img, cmap=plt.cm.gray, interpolation='nearest')
plt.axis('off')
plt.subplots_adjust(wspace=0.02, hspace=0.3, top=1, bottom=0.1, left=0, right=1)
plt.show()
```



Note that we have a noisy, binarized image. Use mathematical morphology to clean it up.

Let's plot the result. For visiblility, we will just plot the top left quadrant of the image.

```
In [9]: plt.figure(figsize=(12, 3))
        1 = 128
       plt.subplot(141)
       plt.imshow(binary_img[:1, :1], cmap=plt.cm.gray)
       plt.axis('off')
       plt.subplot(142)
       plt.imshow(open_img[:1, :1], cmap=plt.cm.gray)
       plt.axis('off')
       plt.subplot(143)
       plt.imshow(close_img[:1, :1], cmap=plt.cm.gray)
        plt.axis('off')
       plt.subplot(144)
       plt.imshow(mask[:1, :1], cmap=plt.cm.gray)
       plt.contour(close_img[:1, :1], [0.5], linewidths=2, colors='r')
        plt.axis('off')
       plt.subplots_adjust(wspace=0.02, hspace=0.3, top=1, bottom=0.1, left=0, right=1)
        plt.show()
```

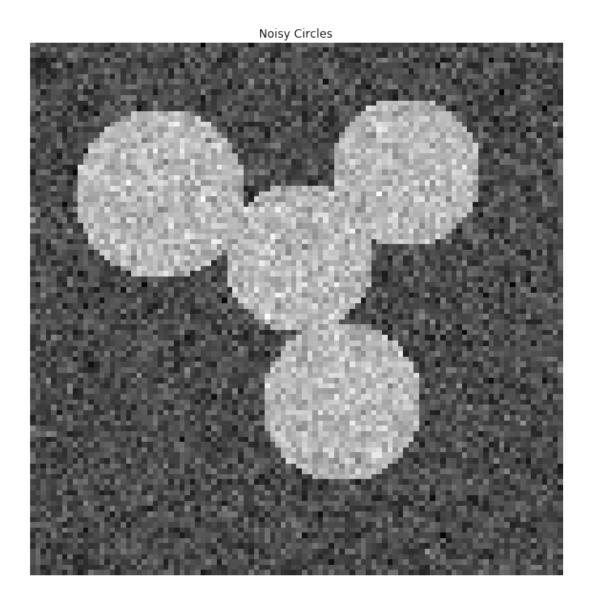


Show that reconstruction operations (erosion + propagation) produce a better result than opening/closing.

Other Scientific Packages provide algorithms that can be useful for image processing. In this example, we use the spectral clustering function of the scikit-learn in order to segment touching objects.

1.3 Generate some noisy, touching circles

```
In [11]: 1 = 100
         x, y = np.indices((1, 1))
         center1 = (28, 24)
         center2 = (40, 50)
         center3 = (67, 58)
         center4 = (24, 70)
         radius1, radius2, radius3, radius4 = 16, 14, 15, 14
         circle1 = (x - center1[0])**2 + (y - center1[1])**2 < radius1**2
         circle2 = (x - center2[0])**2 + (y - center2[1])**2 < radius2**2
         circle3 = (x - center3[0])**2 + (y - center3[1])**2 < radius3**2
         circle4 = (x - center4[0])**2 + (y - center4[1])**2 < radius4**2
         # 4 circles
         img = circle1 + circle2 + circle3 + circle4
         mask = img.astype(bool)
         img = img.astype(float)
         img += 1 + 0.2*np.random.randn(*img.shape)
         cFig = plotOneImage(img, "Noisy Circles", sFont=12, size=(10,10), colormap=plt.cm.gray)
         cFig.show()
```



Convert the image into a graph with the value of the gradient on the edges.

```
In [12]: graph = image.img_to_graph(img, mask=mask)
```

Next, take a decreasing function of the gradient: we take it weakly dependant from the gradient the segmentation is close to a Voronoi.

```
ax = plt.gca()
ax.set_title("Noisy", fontsize=sFont)
plt.axis('off')
plt.subplot(122)
plt.imshow(label_im)
ax = plt.gca()
ax.set_title("Labeled", fontsize=sFont)
plt.axis('off')
plt.subplots_adjust(wspace=0.02, hspace=0.3, top=1, bottom=0.1, left=0, right=1)
plt.show()
```

