

# Feature extraction from images

we want to extract features from shapes by turning them into time series.

## importing the modules

```
In [ ]: import numpy as np
import matplotlib.image as mpimg # reading images
import matplotlib.pyplot as plt
import matplotlib.patches as mpatches
from skimage import measure # to find shape
import scipy.ndimage as ndi # image processing
from pylab import rcParams
```

```
In [ ]: %matplotlib inline
rcParams['figure.figsize'] = (3, 3) # setting default size of plots
```

## reading image file

```
In [ ]: img= mpimg.imread('images/53.jpg')
plt.imshow(img)
```

## using image processing module of scipy to find center of leaf

- scipy.ndimage is a multi-dimensional image processing tool.
- It is mainly used for image filtering, measurements, and morphology.
- It can be used for tasks such as smoothing, sharpening, edge detection, and noise reduction in images.

```
In [ ]: cy,cx = ndi.center_of_mass(img)
plt.imshow(img,cmap='Set3')
plt.scatter(cx, cy)
plt.show()
```

## finding edges of leaf

- from scikit-learn we use measure to find the contour
- we identify the boundaries of the leaf within the image, using a threshold value of 0.8 to determine which pixels belong to the leaf.

```
In [ ]: contours = measure.find_contours(img, .8)
contour = max(contours, key=len)
```

```
In [ ]: contour
```

```
In [ ]: plt.plot(contour[:,1], contour[:,0], linewidth=0.5)
```

```
In [ ]: plt.plot(contour[:,1], contour[:,0], linewidth=0.5)
plt.imshow(img, cmap='Set3')
plt.show()
```

## generating time-series

### project this contour (pairs of x,y coordinstes) into the polar coordinate system

- we represent each point using polar coordinates  $(r, \theta)$  instead of Cartesian coordinates.
- In polar coordinates, 'r' represents the distance from the origin to the point, and ' $\theta$ ' represents the angle measured from the positive x-axis to the line segment connecting the origin to the point.
- formulas-
  - $x = r \cos(\theta)$
  - $y = r \sin(\theta)$
  - $x^2 + y^2 = r^2$
- polar coordinates make it easy to identify features and charectitics of contour

converting cartesian to polar coordinates

```
In [ ]: def cart2pol(x,y):
        r=np.sqrt(x**2+y**2)
        theta=np.arctan2(y,x)
        return[r,theta]
```

```
In [ ]: polar_contour=np.array([cart2pol(x,y) for x,y in contour])
```

### plotting polar coordinates

```
In [ ]: plt.plot(polar_contour[:,1],polar_contour[:,0],linewidth=0.5)
plt.show()
```

we wanted a time series but got a leaf instead.

### trying again but move leaf to (0,0) this time

- we Demean the contour data (subtracting mean from each data point to make the overall mean to zero)
- Demean is needed becuase: the polar coordinate projection failed to yield what we want, because the shape is in the +,+ part of the Cartesian system, not around the center.

## demean

```
In [ ]: contour[:,1] -= cx # demean X
        contour[:,0] -= cy # demean Y
```

```
In [ ]: plt.plot(-contour[:,1], -contour[:,0], linewidth=0.5)
        plt.grid()
        plt.scatter(0, 0)
        plt.show()
```

## projecting new cartesian coordinates into polar space

```
In [ ]: # conversion
        polar_contour = np.array([cart2pol(x, y) for x, y in contour])
```

```
In [ ]: # visualisation
        rcParams['figure.figsize'] = (10,5)
        plt.subplot(121)
        plt.scatter(polar_contour[:,1], polar_contour[:,0], s=0.5, linewidth=0, c=polar_contour[:,0])
        plt.title('polar coordinates')
        plt.grid()

        plt.subplot(122)
        plt.scatter(contour[:,1], contour[:,0], linewidth=0, s=2, c=range(len(contour)))
        plt.scatter(0,0)
        plt.title('cartesian coordinates')
        plt.grid()
        plt.show()
```

this is not a time series yet

## using scikitlearn image feature extractions

- corner\_harris - Detects corners in an image
- corner\_subpix - Refines corner positions detected by an initial method and improves accuracy
- corner\_peaks - Identifies strongest corners after corner detection.
- CENSURE - Identifies points of interest based on local extremas in intensity

```
In [ ]: from skimage.feature import corner_harris, corner_subpix, corner_peaks, CENSURE
```

```
from skimage.feature import corner_harris, corner_subpix, corner_peaks, CENSURE
```

```
CENSURE().detect(img)
```

```
coords = corner_peaks(corner_harris(img), min_distance=5)
coords_subpix = corner_subpix(img, coords, window_size=13)
```

```
plt.subplot(121) plt.title('CENSURE feature detection') plt.imshow(img, cmap='Set3')
plt.scatter(detector.keypoints[:, 1], detector.keypoints[:, 0], 2 ** detector.scales,
            facecolors='none', edgecolors='r')
```

```
plt.subplot(122) plt.title('Harris Corner Detection') plt.imshow(img, cmap='Set3') # show me the leaf
plt.plot(coords[:, 1], coords[:, 0], '.b', markersize=5) plt.show()
```

### finding local maxima and minima

argrextrema finds indices of local extrema (peaks or valleys) in an array.

```
In [ ]: from scipy.signal import argrextrema
```

```
In [ ]: c_max_index= argrextrema(polar_contour[:,0],np.greater,order=50)
c_min_index= argrextrema(polar_contour[:,0],np.less,order=50)
```

```
In [ ]: plt.subplot(121)
plt.scatter(polar_contour[:,1],polar_contour[:,0],linewidth=0,s=2,c='k')
plt.scatter(polar_contour[:,1][c_max_index],polar_contour[:,0][c_max_index],linewidth=0,s=2,c='r')
plt.scatter(polar_contour[:,1][c_min_index],polar_contour[:,0][c_min_index],linewidth=0,s=2,c='b')

plt.subplot(122)
plt.scatter(contour[:,1], contour[:,0],linewidth=0, s=2, c='k')
plt.scatter(contour[:,1][c_max_index],contour[:,0][c_max_index],linewidth=0,s=2,c='r')
plt.scatter(contour[:,1][c_min_index],contour[:,0][c_min_index],linewidth=0,s=2,c='b')

plt.show()
```

finds the points with the greatest distance from the center but not the tip of leaf

### mathematical morphology functions

- Erosion: Shrinks the boundaries of objects in an image by removing pixels near the object edges.
- Dilation: Expands the boundaries of objects in an image by adding pixels near the object edges.
- Opening: Removes small objects and smooths boundaries by performing an erosion followed by a dilation.
- Closing: Fills small gaps and smooths object boundaries by performing a dilation followed by an erosion.

```
In [ ]: def cont(img):
        return max(measure.find_contours(img, .8), key=len)

# let us set the brush to a 6x6 circle
struct = [[ 0., 0., 1., 1., 0., 0.],
          [ 0., 1., 1., 1., 1., 0.],
          [ 1., 1., 1., 1., 1., 1.],
          [ 1., 1., 1., 1., 1., 1.],
          [ 1., 1., 1., 1., 1., 1.],
          [ 0., 1., 1., 1., 1., 0.],
          [ 0., 0., 1., 1., 0., 0.]]
```

```
In [ ]: from scipy.ndimage import binary_erosion, binary_closing, binary_opening, b

erosion = cont(binary_erosion(img, structure=struct).astype(img.dtype))
closing = cont(binary_closing(img, structure=struct).astype(img.dtype))
opening = cont(binary_opening(img, structure=struct).astype(img.dtype))
dilation = cont(binary_dilation(img, structure=struct).astype(img.dtype))
```

```
In [ ]: plt.imshow(img.T, cmap='Greys', alpha=.2)
plt.plot(erosion[:,0], erosion[:,1], c='b')
plt.plot(opening[:,0], opening[:,1], c='g')
plt.plot(closing[:,0], closing[:,1], c='r')
plt.plot(dilation[:,0], dilation[:,1], c='k')
plt.xlim([0, 400])
plt.ylim([400, 800])
plt.show()
```

there is noise around the edge of original image

### check if noise is present

```
In [ ]: plt.imshow(img.astype(bool).astype(float), cmap='hot')
plt.show()
```

### removing noise

```
In [ ]: # pixels with value greater than 254 are set to true(foreground) and < 254
erosion = cont(binary_erosion(img > 254, structure=struct).astype(img.dtype))
closing = cont(binary_closing(img > 254, structure=struct).astype(img.dtype))
opening = cont(binary_opening(img > 254, structure=struct).astype(img.dtype))
dilation = cont(binary_dilation(img > 254, structure=struct).astype(img.dty
```

```
In [ ]: plt.imshow(img.T, cmap='Greys', alpha=.2)
plt.plot(erosion[:,0], erosion[:,1], c='b')
plt.plot(opening[:,0], opening[:,1], c='g')
plt.plot(closing[:,0], closing[:,1], c='r')
plt.plot(dilation[:,0], dilation[:,1], c='k')
plt.xlim([0, 400])
plt.ylim([400, 800])
plt.show()
```

from 2 morphology tests, it is clear:

- the leaf has debris around its edge
- there are no 100% white pixels at the edge

we use the red contour line as the base

### finding core shape of leaf and edge texture

```
In [ ]: # calculates the distance of each non-zero (foreground) pixel to the nearest
dist_2d = ndi.distance_transform_edt(img)
plt.imshow(img, cmap='Greys', alpha=.2)
plt.imshow(dist_2d, cmap='plasma', alpha=.2)
plt.contour(dist_2d, cmap='plasma')
plt.show()
```

## Reading image file - 2

```
In [ ]: rcParams['figure.figsize'] = (3, 3)
img2= mpimg.imread('images/1.jpg')
plt.imshow(img2)
```

### finding center of image

```
In [ ]: cy,cx = ndi.center_of_mass(img2)
plt.imshow(img2,cmap='Set3')
plt.scatter(cx, cy)
plt.show()
```

### finding edge of leaf

```
In [ ]: contours = measure.find_contours(img2, .8)
contour = max(contours, key=len)
plt.plot(contour[:,1], contour[:,0], linewidth=0.5)
plt.imshow(img2, cmap='Set3')
plt.show()
```

### moving centre to (0,0)

```
In [ ]: contour[:,1] -= cx # demean X
        contour[:,0] -= cy # demean Y
```

```
In [ ]: plt.plot(-contour[:,1], -contour[:,0], linewidth=0.5)
        plt.grid()
        plt.scatter(0, 0)
        plt.show()
```

## performing mathematical morphology

### checking for noise

```
In [ ]: rcParams['figure.figsize'] = (5, 5)
```

```
In [ ]: plt.imshow(img2.astype(bool).astype(float), cmap='hot')
        plt.show()
```

noise exists

### removing noise and applying functions

```
In [ ]: def cont(img2):
        return max(measure.find_contours(img2, .8), key=len)
```

```
struct = [[ 0., 0., 1., 1., 0., 0.],
           [ 0., 1., 1., 1., 1., 0.],
           [ 1., 1., 1., 1., 1., 1.],
           [ 1., 1., 1., 1., 1., 1.],
           [ 1., 1., 1., 1., 1., 1.],
           [ 0., 1., 1., 1., 1., 0.],
           [ 0., 0., 1., 1., 0., 0.]]
```

```
In [ ]: erosion = cont(binary_erosion(img2 > 254, structure=struct).astype(img.dtype)
        closing = cont(binary_closing(img2 > 254, structure=struct).astype(img.dtype)
        opening = cont(binary_opening(img2 > 254, structure=struct).astype(img.dtype)
        dilation = cont(binary_dilation(img2 > 254, structure=struct).astype(img.dtype)

        plt.imshow(img2.T, cmap='Greys', alpha=.2)
        plt.plot(erosion[:,0], erosion[:,1], c='b')
        plt.plot(opening[:,0], opening[:,1], c='g')
        plt.plot(closing[:,0], closing[:,1], c='r')
        plt.plot(dilation[:,0], dilation[:,1], c='k')
        plt.xlim([0, 500])
        plt.ylim([300, 800])
        plt.show()
```

### finding core shape if leaf and edge texture

```
In [ ]: dist_2d = ndi.distance_transform_edt(img2)
plt.imshow(img2, cmap='Greys', alpha=.2)
plt.imshow(dist_2d, cmap='plasma', alpha=.2)
plt.contour(dist_2d, cmap='plasma')
plt.show()
```

```
In [ ]:
```

```
In [ ]:
```