# fitting-circles

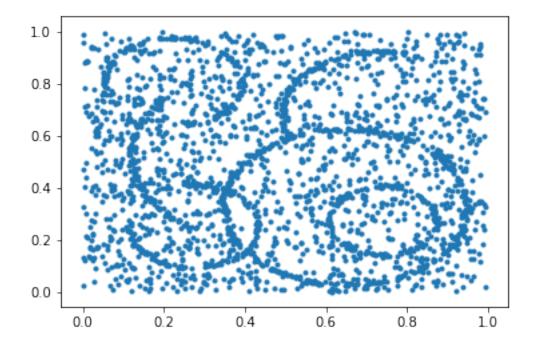
December 19, 2017

## 0.1 3.1

```
In [1]: import numpy as np
    import matplotlib.pyplot as plt

data = np.load("circles.npy")

plt.scatter(data[:,0], data[:,1], marker=".")
    plt.show()
```

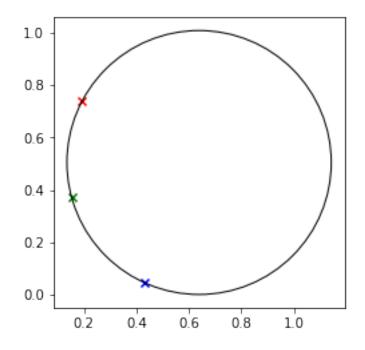


In this picture I can see four clearly recognizable circles

```
In [2]: # RANSAC Test if the parameters are correct

xdata = data[:,0]
ydata = data[:,1]
```

```
dim = np.shape(xdata)
N = 1
for i in range(N):
               rands = np.random.choice(dim[0], 3, replace=False)
              X = xdata[rands]
              Y = ydata[rands]
              cy = (Y[2]**2+X[2]**2-X[0]**2-Y[0]**2+((Y[2]**2-Y[1]**2+X[2]**2-X[1]**2)/(X[1]-X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2+X[2]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]**2-X[1]*
               cy = cy/(-2*Y[0]+2*Y[2]+2*Y[2]*(X[2]-X[0])/(X[1]-X[2])-2*Y[1]*(X[2]-X[0])/(X[1]-X[2])
               cx = ((Y[2]-cy)**2-(Y[1]-cy)**2+X[2]**2-X[1]**2)/(-2*X[1]+2*X[2])
               r = np.sqrt((Y[2]-cy)**2+(X[2]-cx)**2)
plt.scatter(X[0],Y[0],marker="x", color = 'r') # First point, color red
plt.scatter(X[1],Y[1],marker="x", color = 'g') # Second Point
plt.scatter(X[2],Y[2],marker="x", color = 'b') # Third Point
circle = plt.Circle((cx, cy), radius=r, fill=False) # Create a circle
plt.gca().add_patch(circle) # Add it to the plot
plt.axis('scaled')
```



```
In [18]: # RANSAC function
          def ransac(data, e, N):
```

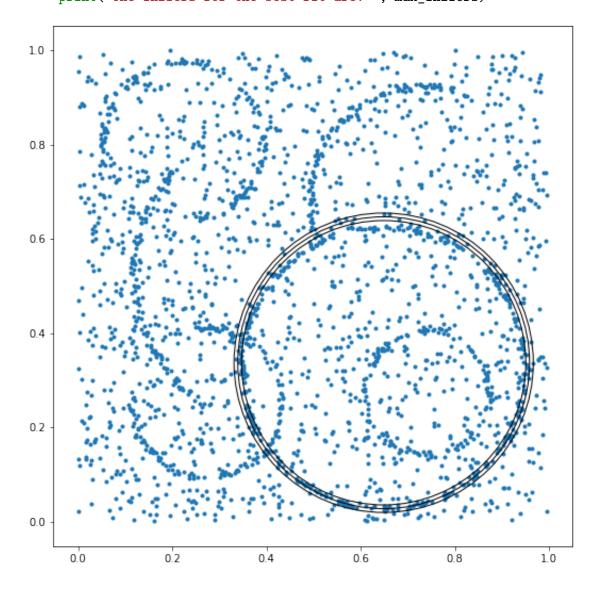
plt.show()

```
xdata = data[:,0]
    ydata = data[:,1]
    dim = np.shape(xdata)
    max_inliers = 0
    for i in range(N):
        rands = np.random.choice(dim[0], 3, replace=False)
        X = xdata[rands]
        Y = ydata[rands]
        cy = (Y[2]**2+X[2]**2-X[0]**2-Y[0]**2+((Y[2]**2-Y[1]**2+X[2]**2-X[1]**2)/(X[1]**2+X[2]**2-X[1]**2)
        cy = cy/(-2*Y[0]+2*Y[2]+2*Y[2]*(X[2]-X[0])/(X[1]-X[2])-2*Y[1]*(X[2]-X[0])/(X[1]-X[2])
        cx = ((Y[2]-cy)**2-(Y[1]-cy)**2+X[2]**2-X[1]**2)/(-2*X[1]+2*X[2])
        r = np.sqrt((Y[2]-cy)**2+(X[2]-cx)**2)
        index = np.arange(dim[0])
        X_rest = xdata
        Y_rest = ydata
        inliers = 0
        inlier_points = []
        inlier_index = []
        for j in range(np.shape(index)[0]):
            dist = np.sqrt((X_rest[j]-cx)**2+(Y_rest[j]-cy)**2)
            if (dist < r+e) and (dist > r-e):
                inliers = inliers + 1
                inlier_points.append([X_rest[j], Y_rest[j]])
                inlier_index.append(j)
        if inliers > max_inliers:
            parameters = [cx, cy, r]
            max_inliers = inliers
            max_inlier_points = inlier_points
            max_inlier_index = inlier_index
    return parameters, max_inliers, max_inlier_points, max_inlier_index
fig = plt.figure(figsize=(9, 9))
plt.scatter(xdata,ydata,marker=".")
e = 0.008 \#accuracy
parameters, max_inliers, max_inlier_points, max_inlier_index = ransac(data, e, 400)
circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2], fill=False)
plt.gca().add_patch(circle) # Add it to the plot
```

```
circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2]-e, fill=False
plt.gca().add_patch(circle) # Add it to the plot

circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2]+e, fill=False
plt.gca().add_patch(circle) # Add it to the plot

plt.axis('scaled')
plt.show()
print("The Inliers for the best fit are: ", max_inliers)
```



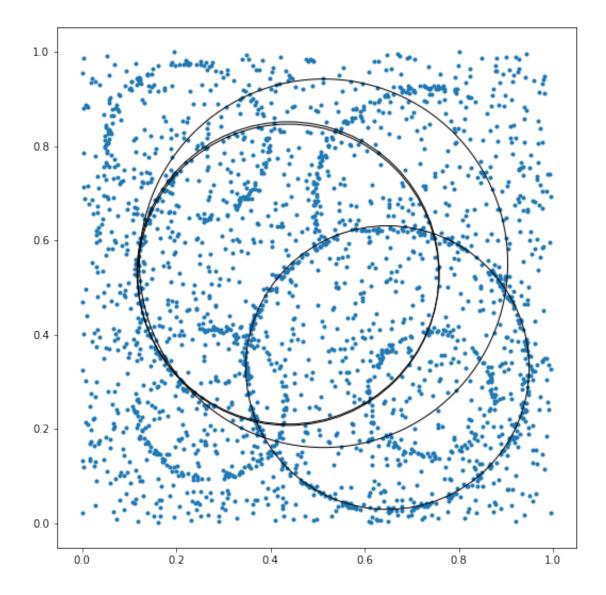
The Inliers for the best fit are: 218

In [43]: data = np.load("circles.npy")

```
xdata = np.array(data[:,0])
ydata = np.array(data[:,1])
#FIRST CIRCLE
e = 0.008 \#accuracy
parameters, max_inliers, max_inlier_points, max_inlier_index = ransac(data, e, 400)
max_inlier_index = np.array(max_inlier_index)
max_inlier_points = np.array(max_inlier_points)
N = np.shape(xdata)[0]
red_index = np.setdiff1d(np.arange(N), max_inlier_index)
red_x = xdata[red_index]
red_y = ydata[red_index]
red_data = np.array([red_x,red_y]).T
print(np.shape(red_data))
fig = plt.figure(figsize=(9, 9))
plt.scatter(xdata,ydata,marker=".")
circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2], fill=False)
plt.gca().add_patch(circle) # Add it to the plot
print("The Inliers for the best fit are: ", max_inliers)
#SECOND CIRCLE
parameters, max_inliers, max_inlier_points, max_inlier_index = ransac(red_data, e, 40
circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2], fill=False)
plt.gca().add_patch(circle) # Add it to the plot
print("The Inliers for the best fit are: ", max_inliers)
max_inlier_index = np.array(max_inlier_index)
red_index = np.setdiff1d(red_index, max_inlier_index)
red_x = xdata[red_index]
red_y = ydata[red_index]
red_data = np.array([red_x,red_y]).T
print(np.shape(red_data))
```

#### #THIRD CIRCLE

```
parameters, max_inliers, max_inlier_points, max_inlier_index = ransac(red_data, e, 40
         circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2], fill=False)
        plt.gca().add_patch(circle) # Add it to the plot
        print("The Inliers for the best fit are: ", max_inliers)
        max_inlier_index = np.array(max_inlier_index)
        red_index = np.setdiff1d(red_index, max_inlier_index)
        red_x = xdata[red_index]
        red_y = ydata[red_index]
        red_data = np.array([red_x,red_y]).T
        print(np.shape(red_data))
         #FOURTH CIRCLE
        parameters, max_inliers, max_inlier_points, max_inlier_index = ransac(red_data, e, 40
        circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2], fill=False)
        plt.gca().add_patch(circle) # Add it to the plot
        plt.axis('scaled')
        print("The Inliers for the best fit are: ", max_inliers)
        plt.show()
(2081, 2)
The Inliers for the best fit are: 350
The Inliers for the best fit are: 150
(1947, 2)
The Inliers for the best fit are: 133
(1833, 2)
The Inliers for the best fit are: 119
```

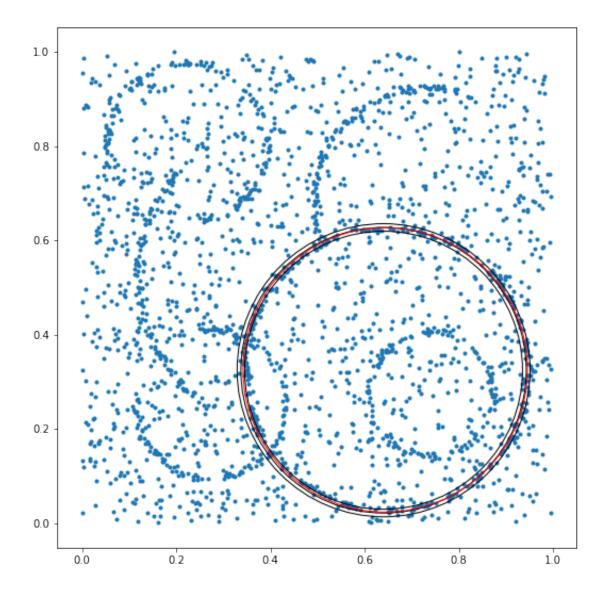


The result is sensitive to e because if you choose it to big, then there is no difference if the circle is shifted in an arbitrary direction. If it's too small none of the points get detected.

## 0.2 3.2

```
parameters, max_inliers, max_inlier_points = ransac(data, e, 400)
        circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2], fill=False)
        plt.gca().add_patch(circle) # Add it to the plot
        circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2]-e, fill=False
        plt.gca().add_patch(circle) # Add it to the plot
        circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2]+e, fill=False
       plt.gca().add_patch(circle) # Add it to the plot
       plt.axis('scaled')
        print("The Inliers for the best fit are: ", max_inliers)
       max_inlier_points = np.array(max_inlier_points)
        Y = \Gamma
        for i in range(np.shape(max_inlier_points)[0]):
            Y.append(max_inlier_points[i,:].T.dot(max_inlier_points[i,:]))
        inlier_x = max_inlier_points[:,0]
        inlier_y = max_inlier_points[:,1]
       X = np.ones((inlier x.shape[0],3))
       X[:,0] = inlier_x
       X[:,1] = inlier_y
       b = np.linalg.lstsq(X,Y)[0]
        cx = b[0]/2
        cy = b[1]/2
       r = np.sqrt(b[2]+cx**2+cy**2)
       parameters = [cx, cy, r]
       print(cx,cy,r)
        circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2], fill=False,
       plt.gca().add_patch(circle) # Add it to the plot
       plt.show()
The Inliers for the best fit are: 259
```

0.643612382035 0.32526563289 0.301740050641



As one can see in the plot, the algebraic-distance-method (red circle) gives a similar result as the RANSAC-algorithm

## 0.3 3.3

```
In [48]: import matplotlib.pyplot as plt
    import matplotlib.image as mpimg
    fig = plt.figure(figsize=(12, 12))
    img=mpimg.imread('derivation.png')
    imgplot = plt.imshow(img)
    plt.show()
```

$$\frac{\min \sum_{i} (\|x_{i} - c\|_{2} - r)^{2}}{\min \sum_{i} (\|x_{i} - c\|_{2} - r)^{2}}$$

$$= \sum_{i} \min \sum_{i} (\|x_{i} - c\|_{2} - r)^{2} = R$$

$$= \sum_{i} \min \sum_{i} (D_{i} - r)^{2} = R$$

$$= \sum_{i} \sum_{i} (D_{i} - r)^{2} = 0$$

$$= \sum_{i} \sum_{i} ($$

In [13]: import scipy.optimize as opt

```
data = np.load("circles.npy")
                        xdata = data[:,0]
                        ydata = data[:,1]
                        fig = plt.figure(figsize=(9, 9))
                        plt.scatter(xdata,ydata,marker=".")
                        e = 0.008 \#accuracy
                        parameters, max_inliers, max_inlier_points = ransac(data, e, 400)
                        circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2], fill=False)
                        plt.gca().add_patch(circle) # Add it to the plot
                        circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2]-e, fill=False
                        plt.gca().add_patch(circle) # Add it to the plot
                        circle = plt.Circle((parameters[0], parameters[1]), radius=parameters[2]+e, fill=False
                        plt.gca().add_patch(circle) # Add it to the plot
                        plt.axis('scaled')
                        print("The Inliers for the best fit are: ", max_inliers)
                        max_inlier_points = np.array(max_inlier_points)
                        def f(x, c):
                                   xdata = x[:,0]
                                   ydata = x[:,1]
                                   m = xdata.shape[0]
                                   cx = 1/m*np.sum(xdata) + (1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(xdata) + (1/m*np.sum(xdata) + (1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(xdata) + (1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(xdata) + (1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0]))**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0]))**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0]))**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0]))**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0]))**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0]))**2+(ydata-c[1])**2))*1/m*np.sum(np.sqrt((xdata-c[0]))**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1])**2+(ydata-c[1
                                   cy = 1/m*np.sum(ydata) + (1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*1/n
                                   return [cx,cy]
                        def residual(p, data):
                                   return f(data, *p)
                        p0 = [1]
                        popt, pcov = opt.leastsq(residual, p0, args=(data))
                        print(popt)
The Inliers for the best fit are: 350
```

```
IndexError
                                              Traceback (most recent call last)
    <ipython-input-13-42afd07d6355> in <module>()
     43
---> 44 popt, pcov = opt.leastsq(residual, p0, args=(data))
     46 print(popt)
    ~\Anaconda3\lib\site-packages\scipy\optimize\minpack.py in leastsq(func, x0, args, Dfu
            if not isinstance(args, tuple):
    375
    376
                args = (args,)
--> 377
            shape, dtype = _check_func('leastsq', 'func', func, x0, args, n)
    378
           m = shape[0]
    379
           if n > m:
    ~\Anaconda3\lib\site-packages\scipy\optimize\minpack.py in _check_func(checker, argnametric)
    24 def _check_func(checker, argname, thefunc, x0, args, numinputs,
                        output_shape=None):
---> 26
          res = atleast_1d(thefunc(*((x0[:numinputs],) + args)))
            if (output_shape is not None) and (shape(res) != output_shape):
    27
                if (output_shape[0] != 1):
     28
    <ipython-input-13-42afd07d6355> in residual(p, data)
     39 def residual(p, data):
           return f(data, *p)
---> 40
    42 p0 = [1]
    <ipython-input-13-42afd07d6355> in f(x, c)
           ydata = x[:,1]
    33
           m = xdata.shape[0]
     34
           cx = 1/m*np.sum(xdata) + (1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))
---> 35
            cy = 1/m*np.sum(ydata) + (1/m*np.sum(np.sqrt((xdata-c[0])**2+(ydata-c[1])**2))*
     36
     37
           return [cx,cy]
```

IndexError: invalid index to scalar variable.