## Computational Linear Algegbra For Large Scale Problems

## Nenna Giulio, Ornella Elena Grassi

## Spectral Clustering Homework

The aim of this homework is to implement and apply **Spectral Clustering** to two different sets of datapoints in  $\mathbb{R}^2$ . The two sets are shown in Figure 1

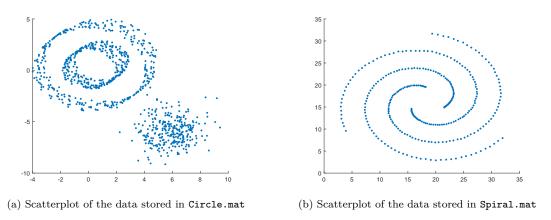


Figure 1: Scatterplot of the two Datasets

As it is clearly visible through visual inspection, both datasets contain 3 different shapes that can be classified as different clusters. In the Circle dataset there are two concentrical circles and a cloud of points in the bottom right while in the Spiral dataset there are 3 spirals. Traditional clustering algorithms, that mainly rely on euclidean distance, may fail in recognizing the presence of shapes in our data hence our need to rely on a different technique called **Spectral clustering**.

## 1 K-Nearest Neighborhood Graph

First, we need to define a similarity function that measures "how much our points are similar to each other". Let  $X_i$  and  $X_j$  be two points in our data, then we will use a similarity measure defined as:

$$s_{i,j} = \exp\left(-\frac{\|X_i - X_j\|^2}{2\sigma^2}\right)$$
 (1.1)

Then, a K-Nearest Neighborhood similarity graph is a Graph G = (V, E) where each vertex  $v_1, \ldots v_n$  represents a point and two vertices  $v_i$  and  $v_j$  are connected by an undirected edge  $e_{i,j}$  if the similarity between  $v_i$  and  $v_j$  is among the K-th highest similarities between  $v_i$  and other vertices in V. For such graph we can define the relative adjacency matrix as  $W_{i,j} = s_{i,j}$  where each entry  $W_{i,j}$  is nonzero only if there exists an edge between  $v_i$  and  $v_j$ . W has zero-values on diagonal by definition.

The following MATLAB code was use to generate the K-NN similarity graph of our data: