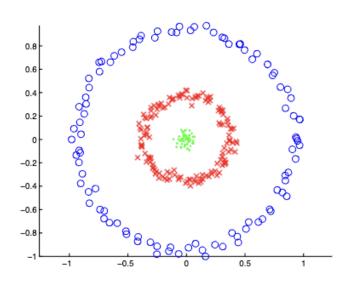
# Lesson 12

**Final Project, Python** 

# guidelines

- Submission Date: 24/11/2024
- NO EXTENSION!!!
- Automatic tests, manual test

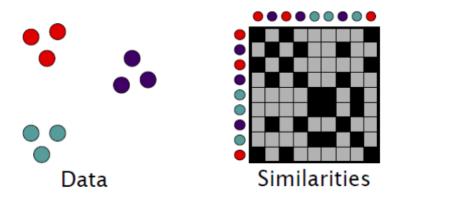
#### **Motivation - Non Linear Data**

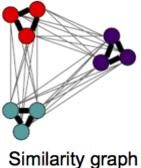


[Kuang et al.(2012), Simon et al.(2005)]

# From Data points to Similarity

- Given: d-dimensional points: x\_1, x\_2, ..., x\_n.
- Transform them into a graph (Similarity Graph).
  - $\circ$  G = (V, E; W) undirected with no self loops.





#### **Weighted Adjacency Matrix**

Gaussian RBF

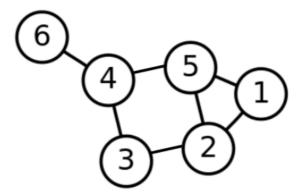
$$a_{ij} = \exp(\frac{-\|x_i - x_j\|^2}{2})$$

A is symmetric, non-negative and no self loops

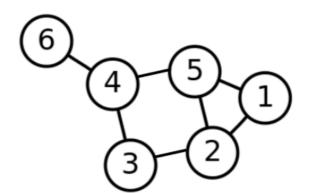
$$a_{ii} = 0$$

## **Example**

- For simplicity, in the next example we will show a non fully connected graph, with all weights set to 1
- We are given d-dimensional data points: x\_1 ... x\_n
- Choose random points and connect them, and we get:



## notations



$$A = \begin{bmatrix} 0 & 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix}$$

$$D = \begin{bmatrix} 2 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 3 & 0 & 0 \\ 0 & 0 & 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

$$W = \begin{bmatrix} \frac{\sqrt{6}}{6} & 0 & \frac{\sqrt{6}}{6} & 0 & \frac{1}{3} & 0\\ 0 & \frac{\sqrt{6}}{6} & 0 & \frac{\sqrt{6}}{6} & 0 & 0\\ 0 & 0 & \frac{\sqrt{6}}{6} & 0 & \frac{1}{3} & \frac{\sqrt{3}}{3}\\ \frac{\sqrt{6}}{6} & \frac{1}{3} & 0 & \frac{1}{3} & 0 & 0\\ 0 & 0 & 0 & \frac{\sqrt{3}}{3} & 0 & 0 \end{bmatrix}$$

$$d_i = \sum_{j=1}^n a_{ij}$$

$$W = D^{-1/2} A D^{-1/2}$$

# **SymNMF Clustering**

- Form the similarity matrix A from X
- Compute the Diagonal Degree Matrix
- Compute the normalized similarity W
- Factorize:

Find 
$$H_{n \times k}$$
 that solves:  $\min_{H > 0} \lVert W - HH^T \rVert_F^2$ 

### **Matrix factorization**

Objective:

$$\min_{H\geq 0} \lVert W - HH^T 
Vert_F^2$$

 $H_{n \times k}, \ k \ll n,$ 

#### **Matrix factorization**

• Initialize H randomly, from interval [0,2\*sqrt(m/k)], where m is average of W.

Update H

$$H_{ij}^{(t+1)} \leftarrow H_{ij}^{(t)} \left( 1 - \beta + \beta \frac{(WH^{(t)})_{ij}}{(H^{(t)}(H^{(t)})^T H^{(t)})_{ij}} \right)$$

where  $\beta = 0.5$ 

Convergence

$$||H^{(t+1)} - H^{(t)}||_F^2 < \epsilon.$$

# **Assign clusters**

Hard clustering, we choose for each element the cluster with the highest association score.

$$H = \begin{bmatrix} 0.06 & 0.01 \\ 0.01 & 0.05 \\ 0.01 & 0.04 \\ 0.02 & 0.04 \\ \hline 0.05 & 0.02 \end{bmatrix}$$