

# Homework Set 6, CPSC 8420, Fall 2024

Last Name, First Name

**Due 12/05/2024, 11:59PM EST**

## Problem 1

Frequently, the affinity matrix is constructed as:

$$A_{ij} = e^{-d(x_i, x_j)^2 / \sigma} \quad (1)$$

where  $\sigma$  is some user-specified parameter. The best that we can hope for in practice is a near block-diagonal affinity matrix. It can be shown in this case, that after projecting to the space spanned by the top  $k$  eigenvectors, points which belong to the same block are close to each other in a euclidean sense. The steps are as follows:

- Construct an affinity matrix  $A$  using the above equation.
  - Symmetrically ‘normalize’ the rows and columns of  $A$  to get a matrix  $N$  such that  $N(i, j) = \frac{A(i, j)}{\sqrt{d(i)d(j)}}$ , where  $d(i) = \sum_k A(i, k)$ .
  - Construct a matrix  $Y$  whose columns are the first  $k$  eigenvectors of  $N$ .
  - Normalize each row of  $Y$  such that it is of unit length.
  - Cluster the dataset by running  $k$ -means on the set of embedded points, where each row of  $Y$  is a data-point.
1. Run  $k$ -means on the datasets provided in the .zip file. For text.mat, take  $k = 6$ . For all others use  $k = 2$ .
  2. Implement the above spectral clustering algorithm and run it on the four provided datasets using the same  $k$ . Plot your clustering results using  $\sigma = .025, .05, .2, .5$ . Hints: You may find the MATLAB functions `pdist` and `eig` to be helpful. A function `plotClusters.m` has been provided to help visualize clustering results.
  3. Plot the first 10 eigenvalues for the rectangles.mat and text.mat datasets when  $\sigma = .05$ . What do you notice?
  4. How do  $k$ -means and spectral clustering compare?

