

# Homework 6

ESE 4020/5420

Due December 4, 2024 at 11:59pm

**Problem 1.** *k-means is suboptimal.* Consider the following 4 points on the 2-D plane (think of  $t$  as a positive number larger than 3):

$$\{(-1, -t), (+1, -t), (-1, t), (+1, t)\}.$$

We want to use  $k$ -means to cluster these points into 2 clusters.

- (a) What are the optimal centers for the 2-Means problem? (Here, by optimal we mean the two centers that minimize the 2-Means objective over the 4 data points.)
- (b) Let us now see what the K-Means algorithm would give us (here,  $K = 2$ ). Find an initial set of centers so that if the K-Means algorithm starts with those centers then it will not find the optimal centers found in part (a).

**Problem 2.** K-means clustering can be viewed as an optimization problem that attempts to minimize some objective function. For the given objectives, determine the update rule for the centroid,  $c_k$  of the  $k$ -th cluster  $C_k$ . In other words, find the optimal  $c_k$  that minimizes the objective function. The data  $x$  contains  $p$  features.

- (a) Show that setting the objective to the sum of the squared Euclidean distances of points from the center of their clusters,

$$\sum_{k=1}^K \sum_{x \in C_k} \sum_{i=1}^p (c_{ki} - x_i)^2$$

results in an update rule where the optimal centroid is the mean of the points in the cluster. [Hint: Look at the derivatives of the objective with respect to each  $c_{ki}$ , and set them to zero. Solving these equations should determine  $c_k$  for each  $k$ .]

- (b) Show that setting the objective to the sum of the Manhattan distances of points from the center of their clusters,

$$\sum_{k=1}^K \sum_{x \in C_k} \sum_{i=1}^p |c_{ki} - x_i|$$

results in an update rule where the optimal centroid is the median of the points in the cluster.

**Problem 3.** Load the Labeled Faces in the Wild dataset from sklearn. You can load this data as follows:

```
from sklearn.datasets import fetch_lfw_people
faces = fetch_lfw_people(min_faces_per_person=60)
```

For this exercise, we will use PCA on image data, in particular pictures of faces, to extract features.

- (a) Perform PCA on the dataset to find the first 150 components. Since this is a large dataset, you should use randomized PCA instead, which can also be found on sklearn. Show the eigenfaces associated with the first 1 through 25 principal components.
- (b) Using the first 150 components you found, reconstruct a few faces of your choice and compare them with the original input images.