

# Assignment Eight

## ECE 4200

- Provide credit to **any sources** other than the course staff that helped you solve the problems. This includes **all students** you talked to regarding the problems.
- You can look up definitions/basics online (e.g., wikipedia, stack-exchange, etc)
- **The due date is 5/03/2019, 23.59.59 eastern time.**
- Submission rules are the same as previous assignments.

**Problem 1. (10 points).** Suppose  $W$  is a  $k \times d$  matrix, where each entry of  $W$  is picked independently from the set  $\{-\frac{1}{\sqrt{k}}, \frac{1}{\sqrt{k}}\}$ . In other words, for each  $i, j$ ,

$$\Pr\left(W_{ij} = -\frac{1}{\sqrt{k}}\right) = \Pr\left(W_{ij} = \frac{1}{\sqrt{k}}\right) = \frac{1}{2}.$$

1. Let  $\vec{x} \in \mathbb{R}^d$ . If we pick  $W$  with this distribution, show that

$$\mathbb{E} [\|W\vec{x}\|_2^2] = \|\vec{x}\|_2^2.$$

2. Just like the Gaussian matrix we considered in the class, we might as well take a random matrix  $W$  designed like this for JL transform. What is an advantage of this matrix over the Gaussian matrix?

**Problem 2. (10 points).** Suppose  $d = 1$ . Come up with a set of  $n$  real numbers, and an initial set of  $k$  distinct cluster centers such that the  $k$ -means algorithm **does not converge** to the best solution of the  $k$ -means clustering problem. You can choose any value of  $n$ , and  $k$  that you want! (Hint: small  $n, k$  are easier to think about.)

**Problem 3. (15 points).** Let  $C = \{\bar{x}_1 \cdots \bar{x}_{|C|}\}$  be a cluster where  $\bar{x}_i \in \mathbb{R}^d$ . Let

$$c_{av} = \frac{1}{|C|} \sum_{\bar{x}_i \in C} \bar{x}_i$$

Prove that for any  $c \in \mathbb{R}^d$ ,

$$\sum_{\bar{x}_i \in C} \|\bar{x}_i - c\|_2^2 \geq \sum_{\bar{x}_i \in C} \|\bar{x}_i - c_{av}\|_2^2$$

(Hint:  $\bar{x}_i - c = \bar{x}_i - c_{av} + c_{av} - c$ )

**Problem 4. (30 points).** Please see attached jupyter notebook.