## 2IMW30 - FOUNDATION OF DATA MINING Assignment 1c Report

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## EXERCISE 1

In this assignment, we implement the Random Projection Algorithm for reducing the number of dimensions in the MNIST dataset.

## EXERCISE 2

Since the nearest neighbor in F of a point P in  $R^d$  is the projection of P onto F, a linear mapping that maps a point to its nearest neighbor in a subspace can be simulated by a rotation followed by an orthogonal projection.

Let  $P_1$  and  $P_2$  be 2 points in  $\mathbb{R}^d$ , M be the rotation matrix, and P be the projection matrix.

Let  $Q_1 = M \cdot P_1$ ,  $Q_2 = M \cdot P_2$ ,  $P_1' = P \cdot Q_1$ , and  $P_2' = P \cdot Q_2$ . So  $P_1'$  and  $P_2'$  are the resulting points.

We know that Rotation preserves the Euclidean distance, so  $Q_1Q_2 = P_1P_2$ .

Because P contains only o and 1, multiplying P to  $Q_1$  and  $Q_2$  preserves several features of  $Q_1$  and  $Q_2$ , and set everything else to o. Thus,  $(P_1' - P_2') \leq (Q_1 - Q_2)$ , which gives  $(P_1' - P_2') \leq (P_1 - P_2)$ .

This completes the proof.

## EXERCISE 3

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Let d(x,y) be the distance from a point P(x,y) to the center of the circle. Then the unit circle equation is d(x,y) = 1.

- When p = 1, we have x + y = 1
- When p = 2, we have  $\sqrt{x^2 + y^2} = 1$
- When p = 10, we have  $(x^{10} + y^{10})^{\frac{1}{10}} = 1$
- When  $p = \infty$ , we have max(x, y) = 1

Plotting the graphs of those equations, we have the unit circles, shown in figure 1.

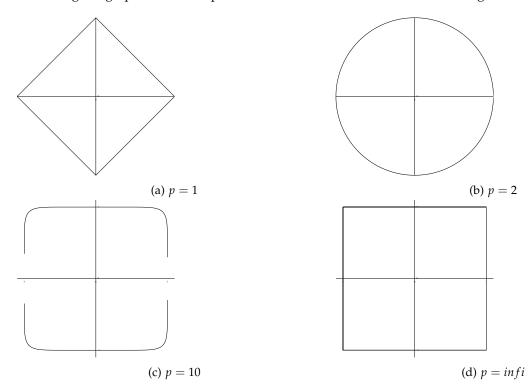


Figure 1: Unit Circles

We can see that the 2 unit circles (the 2 squares indeed) are quite similar. The differences are the angle (pi/4) and the length of the edge (2 and  $\sqrt{2}$ ). Thus, we can preserve the distance by doing a rotation by an angle of pi/4 followed by a scaling by a factor of  $\sqrt{2}$ .

The rotation matrix is:

$$\begin{bmatrix} \sqrt{2}/2 & -\sqrt{2}/2 \\ \sqrt{2}/2 & \sqrt{2}/2 \end{bmatrix}$$

So the mapping function can be done by multiplying the following matrix to the original data:

$$\begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$$

Now we prove that the mapping function actually preserves the distance. As we know, the difference between 2 vectors is also a vector. Let p = (a, b) be a difference between 2 vectors in space. After the transformation, p' = (a - b, a + b). We have:

$$||p||_1 = |a| + |b|$$
  
 $||p'||_{\infty} = max(|a - b|, |a + b|) = |a| + |b|$ 

So  $||p||_1 = ||p'||_{\infty}$ . This completes the proof.