

# MSBD 5004 Mathematical Methods for Data Analysis

## Homework 2

Due date: 27 March, 11:59pm, Friday

1. Let  $(V, \|\cdot\|)$  be a normed vector space.

(a) Prove that, for all  $\mathbf{x}, \mathbf{y} \in V$ ,

$$||\mathbf{x}\| - \|\mathbf{y}\|| \leq \|\mathbf{x} - \mathbf{y}\|.$$

(b) Let  $\{\mathbf{x}_k\}_{k \in \mathbb{N}}$  be a convergent sequence in  $V$  with limit  $\mathbf{x} \in V$ . Prove that

$$\lim_{k \rightarrow \infty} \|\mathbf{x}_k\| = \|\mathbf{x}\|.$$

(Hint: Use part (a).)

(c) Let  $\{\mathbf{x}^{(k)}\}_{k \in \mathbb{N}}$  be a sequence in  $V$  and  $\mathbf{x}, \mathbf{y} \in V$ . Prove that, if

$$\mathbf{x}^{(k)} \rightarrow \mathbf{x}, \quad \text{and} \quad \mathbf{x}^{(k)} \rightarrow \mathbf{y},$$

then  $\mathbf{x} = \mathbf{y}$ . (In other words, the limit of the same sequence in a normed vector space is unique.)

2. Let  $V$  be a vector space, and  $\langle \cdot, \cdot \rangle$  be an inner product on  $V$ . Use the definition of inner product to prove the following.

(a) Prove that  $\langle \mathbf{0}, \mathbf{x} \rangle = \langle \mathbf{x}, \mathbf{0} \rangle = 0$  for any  $\mathbf{x} \in V$ . Here  $\mathbf{0}$  is the zero vector in  $V$ .

(b) Prove that the second condition

$$\langle \alpha \mathbf{x}_1 + \beta \mathbf{x}_2, \mathbf{y} \rangle = \alpha \langle \mathbf{x}_1, \mathbf{y} \rangle + \beta \langle \mathbf{x}_2, \mathbf{y} \rangle, \quad \forall \mathbf{x}_1, \mathbf{x}_2, \mathbf{y} \in V, \alpha, \beta \in \mathbb{R}$$

is equivalent to

$$\langle \mathbf{x}_1 + \mathbf{x}_2, \mathbf{y} \rangle = \langle \mathbf{x}_1, \mathbf{y} \rangle + \langle \mathbf{x}_2, \mathbf{y} \rangle \quad \text{and} \quad \langle \alpha \mathbf{x}, \mathbf{y} \rangle = \alpha \langle \mathbf{x}, \mathbf{y} \rangle, \quad \forall \mathbf{x}_1, \mathbf{x}_2, \mathbf{x}, \mathbf{y} \in V, \alpha \in \mathbb{R}.$$

3.  $\mathbb{R}^{m \times n}$  is a vector space over  $\mathbb{R}$ . Show that  $\langle \mathbf{A}, \mathbf{B} \rangle = \text{trace}(\mathbf{A}^T \mathbf{B})$  for  $\mathbf{A}, \mathbf{B} \in \mathbb{R}^{m \times n}$  is an inner product on  $\mathbb{R}^{m \times n}$ . Here  $\text{trace}(\cdot)$  is the trace of a matrix, i.e., the sum of all diagonal entries.

4. Consider the polynomial kernel  $K(\mathbf{x}, \mathbf{y}) = (\mathbf{x}^T \mathbf{y} + 1)^2$  for  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^2$ . Find an explicit feature map  $\phi : \mathbb{R}^2 \rightarrow \mathbb{R}^6$  satisfying  $\langle \phi(\mathbf{x}), \phi(\mathbf{y}) \rangle = K(\mathbf{x}, \mathbf{y})$ , where the inner product is the standard inner product in  $\mathbb{R}^6$ .

5. (You don't need to do anything for this question.) A good Matlab code and demonstration of kernel K-means can be found at <http://www.dcs.gla.ac.uk/~srogers/firstcourseml/matlab/chapter6/kernelkmeans.html>. Read the code. Run the code in Matlab, if possible, to see how kernel K-means works for nonlinear data.