## MSBD 5004 Mathematical Methods for Data Analysis Homework 3

Due date: 27 March, 9pm, Wednesday

1. Let V be a Hilbert space. Let  $S_1$  and  $S_2$  be two hyperplanes in V defined by

$$S_1 = \{ \boldsymbol{x} \in V \mid \langle \boldsymbol{a}_1, \boldsymbol{x} \rangle = b_1 \}, \quad S_2 = \{ \boldsymbol{x} \in V \mid \langle \boldsymbol{a}_2, \boldsymbol{x} \rangle = b_2 \}.$$

Let  $y \in V$  be given. We consider the projection of y onto  $S_1 \cap S_2$ , i.e., the solution of

$$\min_{\boldsymbol{x} \in S_1 \cap S_2} \|\boldsymbol{x} - \boldsymbol{y}\|. \tag{1}$$

- (a) Prove that  $S_1 \cap S_2$  is a plane, i.e., if  $x, z \in S_1 \cap S_2$ , then  $(1+t)z tx \in S_1 \cap S_2$  for any  $t \in \mathbb{R}$ .
- (b) Prove that z is a solution of (1) if and only if  $z \in S_1 \cap S_2$  and

$$\langle \boldsymbol{z} - \boldsymbol{y}, \boldsymbol{z} - \boldsymbol{x} \rangle = 0, \quad \forall \boldsymbol{x} \in S_1 \cap S_2.$$
 (2)

- (c) Find an explicit solution of (1).
- (d) Prove the solution found in part (c) is unique.
- 2. Let  $\{(\boldsymbol{x}_i, y_i)\}_{i=1}^N$  be given with  $\boldsymbol{x}_i \in \mathbb{R}^n$  and  $y_i \in \mathbb{R}$ . Assume N < n, and  $\boldsymbol{x}_i$ , i = 1, 2, ..., N, are linearly independent. Consider the ridge regression

$$\min_{oldsymbol{a} \in \mathbb{R}^n} \sum_{i=1}^N \left( \langle oldsymbol{a}, oldsymbol{x}_i 
angle - y_i 
ight)^2 + \lambda \|oldsymbol{a}\|_2^2,$$

where  $\lambda \in \mathbb{R}$  is a regularization parameter, and we set the bias b = 0 for simplicity.

- (a) Prove that the solution must be in the form of  $\boldsymbol{a} = \sum_{i=1}^{N} c_i \boldsymbol{x}_i$  for some  $\boldsymbol{c} = [c_1, c_2, \dots, c_N]^T \in \mathbb{R}^N$ . (Hint: Similar to the proof of the representer theorem.)
- (b) Re-express the minimization in terms of  $c \in \mathbb{R}^N$ , which has fewer unknowns than the original formulation.
- 3. Find the Frechet derivative of the following functions  $f:V\mapsto\mathbb{R}$ , where V is a Hilbert space.
  - (a)  $f(\mathbf{x}) = ||\mathbf{x} \mathbf{a}||$  for a given  $\mathbf{a} \in V$ , where  $\mathbf{x} \neq \mathbf{a}$ .
  - (b)  $f(\mathbf{x}) = ||2\mathbf{x} \mathbf{a}||^2$  for a given  $\mathbf{a} \in V$ .
  - (c)  $f(\boldsymbol{x}) = \frac{1}{\|\boldsymbol{x}\|}$ , where  $\boldsymbol{x} \neq \boldsymbol{0}$ .
  - (d)  $V = \mathbb{R}^n$  and  $f(\boldsymbol{x}) = \sum_{i=1}^n \sqrt{x_i^2 + c}$  for some c > 0.