

# MVA "Kernel methods"

## Homework 4

Jean-Philippe Vert

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### 1 Splines

Let  $H = C_2([0, 1])$  be the set of twice continuously differentiable functions  $f : [0, 1] \rightarrow \mathbb{R}$ , and  $H_1 \subset H$  be the set of functions  $f \in H$  that satisfy:

$$f(0) = f'(0) = 0.$$

**1.1.** Show that  $H_1$  endowed with the norm:

$$\|f\|_{H_1}^2 = \int_0^1 f''(t)^2 dt$$

is a reproducing kernel Hilbert space (RKHS), and compute the reproducing kernel  $K_1$ .

**1.2.** Let  $H_2$  be the set of affine functions  $f : [0, 1] \rightarrow \mathbb{R}$  (i.e., the functions that can be written as  $f(x) = ax + b$ , with  $a, b \in \mathbb{R}$ ). Show that  $H_2$  endowed with the norm:

$$\|f\|_{H_2}^2 = f(0)^2 + f'(0)^2$$

is a RKHS and compute the corresponding kernel  $K_2$ .

**1.3.** Deduce that  $H$  endowed with the norm:

$$\|f\|_H^2 = \int_0^1 f''(t)^2 dt + f(0)^2 + f'(0)^2$$

is a RKHS and compute the reproducing kernel  $K$ .

**1.4.** Let  $0 < x_1 < \dots < x_n < 1$  and  $(y_1, \dots, y_n) \in \mathbb{R}^n$ . In order to estimate a regression function  $f : [0, 1] \rightarrow \mathbb{R}$ , we consider the following optimization problem:

$$\min_{f \in H} \frac{1}{n} \sum_{i=1}^n (f(x_i) - y_i)^2 + \lambda \int_0^1 f''(t)^2 dt. \quad (1)$$

Show that any solution of (1) can be expanded as:

$$\hat{f}(x) = \sum_{i=1}^n \alpha_i K_1(x_i, x) + \beta_1 x + \beta_2,$$

with  $\alpha = (\alpha_1, \dots, \alpha_n)' \in \mathbb{R}^n$  et  $\beta = (\beta_0, \beta_1)' \in \mathbb{R}^2$ .

**1.5.** Let  $I$  be the  $n \times n$  identity matrix,  $M$  be the square  $n \times n$  matrix defined by:

$$M_{i,j} = \begin{cases} K_1(x_i, x_j) & \text{si } i \neq j, \\ K_1(x_i, x_j) + n\lambda & \text{si } i = j, \end{cases}$$

$T$  be the  $n \times 2$  matrix:

$$T = \begin{pmatrix} 1 & x_1 \\ \vdots & \vdots \\ 1 & x_n \end{pmatrix},$$

and  $\mathbf{y} = (y_1, \dots, y_n)'$ .

Show that  $\alpha$  and  $\beta$  satisfy:

$$\begin{cases} T' \alpha = 0, \\ M \alpha + T \beta = \mathbf{y}. \end{cases}$$

**1.6.** Deduce that  $\alpha$  and  $\beta$  are given by:

$$\begin{cases} \alpha = M^{-1} \left( I - T (T' M^{-1} T)^{-1} T' M^{-1} \right) \mathbf{y}, \\ \beta = (T' M^{-1} T)^{-1} T' M^{-1} \mathbf{y}. \end{cases}$$

**1.7.** Show that

- $\hat{f} \in C_2([0, 1])$ ;
- $\hat{f}$  is a polynomial of degree 3 on each interval  $[x_i, x_{i+1}]$  for  $i = 1, \dots, n-1$ ;
- $\hat{f}$  is an affine function on both intervals  $[0, x_1]$  and  $[x_n, 1]$ .

$\hat{f}$  is called a *spline*.

## 2 More kernels...

Are the following functions positive definite kernels?

$$\forall x, y \in \mathbb{R}^p, \quad K_2(x, y) = \frac{1}{2 - e^{-\|x-y\|^2}}$$

$$\forall x, y \in \mathbb{R}, \quad K_3(x, y) = \max(0, 1 - |x - y|)$$