

Exercise Sheet 8.

Hand in due to: Friday, 12/05/2023, 14:00

**Exercise 1** (Orthonormal basis of  $H_X \mid$  4 points).

Let  $\Omega \in \mathbb{R}^d$  be a nonempty open set and  $H_K(\Omega)$  be a RKHS with the reproducing kernel  $K : \Omega \times \Omega \to \mathbb{R}$ . Let  $\emptyset \neq X_0 \subset X_n = \{x_1, \dots, x_n\} \subset X_N \subset \Omega$  be sets of pairwise distinct points. Consider  $\{v_i(x)\}_{i=1}^N$  as the Gram-Schmidt orthonormalization of  $\{K(\cdot, x_i)\}_{i=1}^N$ , i.e.,

$$v_1(x) := \frac{K(x, x_1)}{\|K(\cdot, x_1)\|_{H_K(\Omega)}}$$

and for all  $1 < n \le N$ 

$$v_n(x) = \frac{\widetilde{v}_n(x)}{\|\widetilde{v}_n\|_{H_K(\Omega)}}, \text{ where } \widetilde{v}_n(x) := K(x, x_n) - \sum_{i=1}^{n-1} \left( K(\cdot, x_n), v_i \right)_{H_K(\Omega)} v_i(x).$$

Show that the power function is

$$P_{H_{X_{n-1}}}(x_n) = v_n(x_n)$$

Hint. Show the identity  $P_{H_{X_n}}(x)^2 = K(x, x) - \sum_{i=1}^n v_i(x)^2$ .

Exercise 2 (Iterative interpolation | 4 points).

Let  $\Omega \in \mathbb{R}^d$  be a nonempty, open set and  $H_K(\Omega)$  be a RKHS with the reproducing kernel  $K : \Omega \times \Omega \to \mathbb{R}$ . Let  $\emptyset \neq X_0 \subset X_n = \{x_1, \dots, x_n\} \subset X_N \subset \Omega$  sets of pairwise distinct points and  $\{v_i(x)\}_{i=1}^N$  be defined as in Exercise 1. For  $f \in H_K(\Omega)$ , we set  $r_{X_0} := f$  and  $r_{X_n} := f - f_{X_n}$ . Show that for all  $n \ge 1$ :

(a) The interpolant can be computed as  $f_{X_N}(x) = \sum_{i=1}^N (f, v_i)_{H_K(\Omega)} v_i(x)$  and there holds

$$f_{X_n}(x) = f_{X_{n-1}}(x) + (f, v_n)_{H_K(\Omega)}v_n(x),$$
  

$$P^2_{H_{X_n}}(x) = P^2_{H_{X_{n-1}}}(x) - v^2_n(x),$$
  

$$r_{X_n}(x) = r_{X_{n-1}}(x) - (f, v_n)_{H_K(\Omega)}v_n(x).$$

(b) For the coefficients of the basis, there holds  $(f, v_n)_{H_K(\Omega)} = r_{X_{n-1}}(x_n)/P_{H_{X_{n-1}}}(x_n)$ .

Exercise 3 (Kernel characterization by Fourier transform | 4 points).

Let  $0 \neq \Phi \in L^1(\mathbb{R}^d) \cap C(\mathbb{R}^d)$  be nonnegative, i.e.,  $\widehat{\Phi}(x) \geq 0$  for all  $x \in \mathbb{R}^d$ . Show that  $K(x, y) := \Phi(x - y)$  is a reproducing kernel by verifying the following steps:

(1) First, for  $g \in L^1(\mathbb{R}^d)$ , show that

$$\int_{\mathbb{R}^d} \Phi(x)\widehat{g}(x)dx = \int_{\mathbb{R}^d} \widehat{\Phi}(x)g(x)dx.$$

(2) By using the first step, show that

$$\Phi(0) = \lim_{m \to \infty} \int_{\mathbb{R}^d} \widehat{\Phi}(y) \widehat{g}_m(y) dy,$$

where  $g_m$  is defined in Exercise 2 from Sheet 7.

- (3) Next, show  $\widehat{\Phi} \in L^1(\mathbb{R}^d)$  by applying the Monotone Convergence Theorem.
- (4) By previous step and fact that  $\widehat{\Phi} \in L^1(\mathbb{R}^D)$  implies  $\Phi = F^{-1}(\widehat{\Phi})$  conclude that K(x, y) is positive definite.

**Exercise 4** (Representer Theorem | 4 points).

Let  $\Omega \in \mathbb{R}^d$  be a nonempty, open set and  $H_K(\Omega)$  be a RKHS with a reproducing kernel  $K : \Omega \times \Omega \to \mathbb{R}$ . Let  $X_N = \{x_1, \dots, x_n\} \subset \Omega$  be a set of pairwise distinct points and  $\{f_i = f(x_i)\}_{i=1}^N \subset \mathbb{R}$  the associated point evaluations of a given function  $f \in H_K(\Omega)$ . For some regularization parameter  $\lambda > 0$ , let us define the regularized interpolant  $f_X^{\lambda}$  as

$$f_X^{\lambda} := \underset{g \in H_K(\Omega)}{\operatorname{argmin}} J_{\lambda}(g),$$

where

$$J_{\lambda}(g) = \sum_{i=1}^{N} \left(f_i - g(x_i)\right)^2 + \lambda \|g\|_{H_K(\Omega)}^2.$$

Show that for every  $f \in H_K(\Omega)$  there exists a regularized interpolant

$$f_X^{\lambda}(\cdot) = \sum_{i=1}^N c_i K(x_i, \cdot),$$

where the vector of coefficients  $c \in \mathbb{R}^N$  is the solution of the linear system

$$(A + \lambda I)c = b, \quad A = [K(x_i, x_j)]_{i,j=1}^N, \quad b = [f(x_1), \dots, f(x_N)]^\top$$

Moreover, if K is strictly positive definite, this is the unique solution of the minimization problem.

Hint. Prove first that for every  $h \in H_K(\Omega)$  there exists  $g \in H_X := span\{K(\cdot, x_1), \dots, K(\cdot, x_N)\}$  such that  $J_{\lambda}(h) \leq J_{\lambda}(g)$ . Then, restrict the minimization over  $H_X$ , i.e., consider only functions of the form

$$g := \sum_{i=1}^N c_i K(\cdot, x_i).$$