

# Real reproducing kernel Hilbert spaces

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## 1 Reproducing kernels

We shall often speak about functions  $F : X \times X \rightarrow \mathbb{R}$ , where  $X$  is a nonempty set. For  $x \in X$ , we define  $F_x : X \rightarrow \mathbb{R}$  by  $F_x(y) = F(x, y)$  and for  $y \in X$  we define  $F^y : X \rightarrow \mathbb{R}$  by  $F^y(x) = F(x, y)$ .  $F$  is said to be **symmetric** if  $F(x, y) = F(y, x)$  for all  $x, y \in X$  and **positive-definite** if for any  $x_1, \dots, x_n \in X$  and  $c_1, \dots, c_n \in \mathbb{R}$  it holds that

$$\sum_{1 \leq i, j \leq n} c_i c_j F(x_i, x_j) \geq 0.$$

**Lemma 1.** *If  $F : X \times X \rightarrow \mathbb{R}$  is symmetric and positive-definite then*

$$F(x, y)^2 \leq F(x, x)F(y, y), \quad x, y \in X.$$

*Proof.* For  $\alpha, \beta \in \mathbb{R}$  define<sup>1</sup>

$$\begin{aligned} C(\alpha, \beta) &= \alpha^2 F(x, x) + \alpha\beta F(x, y) + \beta\alpha F(y, x) + \beta^2 F(y, y) \\ &= \alpha^2 F(x, x) + 2\alpha\beta F(x, y) + \beta^2 F(y, y), \end{aligned}$$

which is  $\geq 0$ . Let

$$\begin{aligned} P(\alpha) &= C(\alpha, F(x, y)) \\ &= \alpha^2 F(x, x) + 2\alpha F(x, y)^2 + F(x, y)^2 F(y, y), \end{aligned}$$

which is  $\geq 0$ . In the case  $F(x, x) = 0$ , the fact that  $P \geq 0$  implies that  $F(x, y) = 0$ . In the case  $F(x, x) \neq 0$ ,  $P(\alpha)$  is a quadratic polynomial and because  $P \geq 0$  it follows that the discriminant of  $P$  is  $\leq 0$ :

$$4F(x, y)^4 - 4 \cdot F(x, x) \cdot F(x, y)^2 F(y, y) \leq 0.$$

That is,  $F(x, y)^4 \leq F(x, x)F(y, y)F(x, y)^2$ , and this implies that  $F(x, y)^2 \leq F(x, x)F(y, y)$ .  $\square$

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<sup>1</sup>See Alain Berline and Christine Thomas-Agnan, *Reproducing Kernel Hilbert Spaces in Probability and Statistics*, p. 13, Lemma 3.

A **real reproducing kernel Hilbert space** is a Hilbert space  $H$  contained in  $\mathbb{R}^X$ , where  $X$  is a nonempty set, such that for each  $x \in X$  the map  $\Lambda_x f = f(x)$  is continuous  $H \rightarrow \mathbb{R}$ . In this note we speak always about real Hilbert spaces.

Let  $H \subset \mathbb{R}^X$  be a reproducing kernel Hilbert space. Because  $H$  is a Hilbert space, the Riesz representation theorem states that  $\Phi : H \rightarrow H^*$  defined by

$$(\Phi g)(f) = \langle f, g \rangle_H, \quad g, f \in H$$

is an isometric isomorphism. Because  $H$  is a reproducing kernel Hilbert space,  $\Lambda_x \in H^*$  for each  $x \in X$  and we define  $T_x = \Phi^{-1}\Lambda_x \in H$ , which satisfies

$$f(x) = \Lambda_x(f) = \langle f, T_x \rangle_H, \quad f \in H.$$

In particular, because  $T_x \in H$ , for  $y \in X$  it holds that

$$T_x(y) = \Lambda_y(T_x) = \langle T_x, T_y \rangle_H.$$

Define  $K : X \times X \rightarrow \mathbb{R}$  by

$$K(x, y) = \langle T_x, T_y \rangle_H,$$

called **the reproducing kernel of  $H$** . For  $x, y \in X$ ,

$$T_x(y) = \langle T_x, T_y \rangle_H = K(x, y) = K_x(y),$$

which means that  $T_x = K_x$ .

A reproducing kernel is symmetric and positive-definite:

$$K(x, y) = \langle T_x, T_y \rangle_H = \langle T_y, T_x \rangle_H = K(y, x)$$

and

$$\begin{aligned} \sum_{1 \leq i, j \leq n} c_i c_j K(x_i, x_j) &= \sum_{1 \leq i, j \leq n} \langle c_i T_{x_i}, c_j T_{x_j} \rangle_H \\ &= \left\langle \sum_{1 \leq i \leq n} c_i T_{x_i}, \sum_{1 \leq j \leq n} c_j T_{x_j} \right\rangle_H \\ &\geq 0. \end{aligned}$$

**Lemma 2.** *If  $E$  is an orthonormal basis for a reproducing kernel Hilbert space  $H \subset \mathbb{R}^X$  with reproducing kernel  $K : X \times X \rightarrow \mathbb{R}$ , then*

$$K(x, y) = \sum_{e \in E} e(x)e(y), \quad x, y \in X.$$

*Proof.* Because  $E$  is an orthonormal basis for  $H$ , Parseval's identity tell us

$$\langle T_x, T_y \rangle_H = \sum_{e \in E} \langle T_x, e \rangle \langle T_y, e \rangle = \sum_{e \in E} \langle e, T_x \rangle \langle e, T_y \rangle = \sum_{e \in E} e(x)e(y).$$

□

If  $H \subset \mathbb{R}^X$  is a reproducing kernel Hilbert space with reproducing kernel  $K : X \times X \rightarrow \mathbb{R}$  and  $V$  is a closed linear subspace of  $H$ , then  $V$  is itself a reproducing kernel Hilbert space, with some reproducing kernel  $G : X \times X \rightarrow \mathbb{R}$ . The following theorem expresses  $G$  in terms of  $K$ .<sup>2</sup>

**Theorem 3.** *Let  $H \subset \mathbb{R}^X$  be a reproducing kernel Hilbert space with reproducing kernel  $K : X \times X \rightarrow \mathbb{R}$ , let  $V$  be a closed linear subspace of  $H$  with reproducing kernel  $G : X \times X \rightarrow \mathbb{R}$ , and let  $P_V : H \rightarrow V$  be the projection onto  $V$ . Then*

$$G_x = P_V K_x, \quad x \in X.$$

*Proof.*  $H = V \oplus V^\perp$ , thus for  $f \in H$  there are unique  $g \in V, h \in V^\perp$  such that  $f = g + h$ , and  $P_V f = g$ .<sup>3</sup> Then  $f - P_V f \in V^\perp$ . Therefore for  $y \in X$ , as  $G_y \in V$  it holds that

$$\langle f, G_y \rangle_H = \langle f - P_V f + P_V f, G_y \rangle_H = \langle P_V f, G_y \rangle_H = (P_V f)(y).$$

In particular, for  $x, y \in X$  and  $f = K_x$ ,

$$(P_V K_x)(y) = \langle K_x, G_y \rangle_H = \langle G_y, T_x \rangle_H = G_y(x) = G(y, x) = G(x, y) = G_x(y),$$

which means that  $P_V K_x = G_x$ , proving the claim.  $\square$

The **Moore-Aronszajn theorem** states that if  $X$  is a nonempty set and  $K : X \times X \rightarrow \mathbb{R}$  is a symmetric and positive-definite function, then there is a unique reproducing kernel Hilbert space  $H \subset \mathbb{R}^X$  for which  $K$  is the reproducing kernel.

We now prove that given a symmetric positive-definite kernel there is a unique reproducing Hilbert space for which it is the reproducing kernel.<sup>4</sup>

## 2 Sobolev spaces on $[0, T]$

Let  $f \in \mathbb{R}^{[0, T]}$ . The following are equivalent:<sup>5</sup>

1.  $f$  is absolutely continuous.
2.  $f$  is differentiable at almost all  $t \in [0, T]$ ,  $f' \in L^1$ , and

$$f(t) = f(0) + \int_0^t f'(s) ds, \quad t \in [0, T].$$

3. There is some  $g \in L^1$  such that

$$f(t) = f(0) + \int_0^t g(s) ds, \quad t \in [0, T].$$

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<sup>2</sup>Ward Cheney and Will Light, *A Course in Approximation Theory*, p. 234, Chapter 31, Theorem 4.

<sup>3</sup><http://individual.utoronto.ca/jordanbell/notes/pvm.pdf>

<sup>4</sup>Alain Berlinet and Christine Thomas-Agnan, *Reproducing Kernel Hilbert Spaces in Probability and Statistics*, p. 19, Theorem 3.

<sup>5</sup>Elias M. Stein and Rami Shakarchi, *Real Analysis*, p. 130, Theorem 3.11.

In particular, if  $f$  is absolutely continuous and  $f' = 0$  almost everywhere then  $\int_0^t f'(s)ds = 0$  and so  $f(t) = f(0)$  for all  $t \in [0, T]$ . That is, if  $f$  is absolutely continuous and  $f' = 0$  almost everywhere then  $f$  is constant.

Let  $H$  be the set of those absolutely continuous functions  $f \in \mathbb{R}^{[0, T]}$  such that  $f(0) = 0$  and  $f' \in L^2$ . For  $f, g \in H$  define

$$\langle f, g \rangle_H = \int_0^T f'(s)g'(s)ds.$$

If  $\|f\|_H = 0$  then  $\int_0^T f'(s)^2 ds = 0$ , which implies that  $f' = 0$  almost everywhere and hence that  $f$  is constant, and therefore  $f = 0$ . Thus  $\langle \cdot, \cdot \rangle_H$  is indeed an inner product on  $H$ .

If  $f_n$  is a Cauchy sequence in  $H$  then  $f'_n$  is a Cauchy sequence in  $L^2$  and hence converges to some  $g \in L^2$ . Then the function  $f \in \mathbb{R}^{[0, T]}$  defined by

$$f(t) = \int_0^t g(s)ds, \quad t \in [0, T],$$

is absolutely continuous,  $f(0) = 0$ , and satisfies  $f' = g$  almost everywhere, which shows that  $f \in H$ . Then  $f_n \rightarrow f$  in  $H$ , which proves that  $H$  is a Hilbert space. For  $t \in [0, T]$ , by the Cauchy-Schwarz inequality,

$$|f(t)|^2 = \left| \int_0^t f'(s)ds \right|^2 \leq \left| \int_0^T f'(s)ds \right|^2 \leq T \int_0^T f'(s)^2 ds = T \|f\|_H^2,$$

i.e.  $|L_t f| \leq T^{1/2} \|f\|_H$ , which shows that  $L_t \in H^*$ . Therefore  $H$  is a reproducing kernel Hilbert space.

For  $a \in [0, T]$  define  $h_a : [0, T] \rightarrow \mathbb{R}$  by  $h_a(s) = 1_{[0, a]}(s)$ , which belongs to  $L^2$ , and define  $g_a : [0, T] \rightarrow \mathbb{R}$  by

$$g_a(t) = \int_0^t h_a(s)ds = \min(t, a),$$

which belongs to  $H$ . For  $f \in H$ ,

$$\langle f, g_a \rangle_H = \int_0^T f'(s)g'_a(s)ds = \int_0^T f'(s)1_{[0, a]}(s)ds = \int_0^a f'(s)ds = f(a).$$

This means that  $K_a = g_a$ . For  $a, b \in [0, T]$ ,

$$\langle K_a, K_b \rangle_H = \int_0^T g'_a(s)g'_b(s)ds = \int_0^T 1_{[0, a]}(s)1_{[0, b]}(s)ds = \int_0^T 1_{[0, \min(a, b)]}(s)ds.$$

That is, the reproducing kernel of  $H$  is  $K : [0, T] \times [0, T] \rightarrow \mathbb{R}$ ,

$$K(a, b) = \langle K_a, K_b \rangle_H = \min(a, b).$$

### 3 Sobolev spaces on $\mathbb{R}$

Let  $\lambda$  be Lebesgue measure on  $\mathbb{R}$ . Let  $\mathcal{L}^2(\lambda)$  be the collection of Borel measurable functions  $f : \mathbb{R} \rightarrow \mathbb{R}$  such that  $|f|^2$  is integrable, and let  $L^2(\lambda)$  be the Hilbert space of equivalence classes of elements of  $\mathcal{L}^2(\lambda)$  where  $f \sim g$  when  $f = g$  almost everywhere, with

$$\langle f, g \rangle_{L^2} = \int_{\mathbb{R}} fg d\lambda.$$

Let  $H^1(\mathbb{R})$  be the set of locally absolutely continuous functions  $f : \mathbb{R} \rightarrow \mathbb{R}$  such that  $f, f' \in L^2(\lambda)$ . This is a Hilbert space with the inner product<sup>6</sup>

$$\langle f, g \rangle_{H^1} = \langle f, g \rangle_{L^2} + \langle f', g' \rangle_{L^2}.$$

Define  $K : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$  by

$$K(x, y) = \frac{1}{2} \exp(-|x - y|), \quad x, y \in \mathbb{R}.$$

Let  $x \in \mathbb{R}$ . For  $y < x$ ,  $K'_x(y) = K_x(y)$  and for  $y > x$ ,  $K'_x(y) = -K_x(y)$ , which shows that  $K_x \in H^1(\mathbb{R})$ . For  $f \in H^1(\mathbb{R})$ , doing integration by parts,

$$\begin{aligned} \langle f, K_x \rangle_{H^1} &= \int_{-\infty}^{\infty} f K_x d\lambda + \int_{-\infty}^x f'(y) K_x(y) d\lambda(y) - \int_x^{\infty} f'(y) K_x(y) d\lambda(y) \\ &= \int_{-\infty}^{\infty} f K_x d\lambda + f(x) K_x(x) - \int_{-\infty}^x f(y) K'_x(y) d\lambda(y) \\ &\quad + f(x) K_x(x) + \int_x^{\infty} f(y) K'_x(y) d\lambda(y) \\ &= \int_{-\infty}^{\infty} f K_x d\lambda + \frac{1}{2} f(x) - \int_{-\infty}^x f(y) K_x(y) d\lambda(y) \\ &\quad + \frac{1}{2} f(x) - \int_x^{\infty} f(y) K_x(y) d\lambda(y) \\ &= f(x) \\ &= T_x f. \end{aligned}$$

This shows that  $H^1(\mathbb{R})$  is a reproducing kernel Hilbert space. We calculate, for  $x < y$ ,

$$\begin{aligned} \langle T_x, T_y \rangle_{H^1} &= \int_{-\infty}^x K_x K_y d\lambda + \int_x^y K_x K_y d\lambda + \int_y^{\infty} K_x K_y d\lambda \\ &\quad + \int_{-\infty}^x K_x K_y d\lambda - \int_x^y K_x K_y d\lambda + \int_y^{\infty} K_x K_y d\lambda \\ &= 4 \cdot \frac{1}{8} \exp(x - y) \\ &= K(x, y). \end{aligned}$$

<sup>6</sup><http://individual.utoronto.ca/jordanbell/notes/sobolev1d.pdf>

This shows that  $K(x, y) = \frac{1}{2} \exp(-|x - y|)$  is the reproducing kernel of  $H^1(\mathbb{R})$ .<sup>7</sup>

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<sup>7</sup>cf. Alain Berlinet and Christine Thomas-Agnan, *Reproducing Kernel Hilbert Spaces in Probability and Statistics*, pp. 8–9, Example 5.