

Kernel methods in machine learning — Homework 1

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Exercise 1. Kernels

1.1 K is symmetric since

$$\forall x, y \in \mathbb{R}^2 K(y, x) = \cos(-(x - y)) = \cos(x - y) = K(x, y)$$

and because for all sequences of (x_i) and (a_i) of \mathbb{R}

$$\begin{aligned} \sum_i \sum_j a_i a_j \cos(x_i - x_j) &= \sum_i \sum_j a_i a_j (\cos x_i \cos x_j + \sin x_i \sin x_j) \\ &= \left(\sum_i a_i \cos x_i \right)^2 + \left(\sum_i a_i \sin x_i \right)^2 \geq 0 \end{aligned}$$

K is p.d.

1.2 K is clearly symmetric and $\forall x, y \in \mathcal{X}$,

$$\begin{aligned} K(x, y) &= \frac{1}{1 - x^T y} \\ &= \lim_{n \rightarrow \infty} \sum_{k=0}^n (x^T y)^k \quad (\text{converges because by C-S } |x^T y| \leq \|x\|_2 \|y\|_2 < 1) \end{aligned}$$

Since all powers of the linear kernel are p.d. kernels, and a sum of p.d. kernels is a p.d. kernel, the above sum is a p.d. kernel for all n . Therefore, K is p.d. as a point-wise limit of a sequence of p.d. kernels.

1.3 The kernel K is symmetric because the intersection is a commutative operation.

$$\begin{aligned} \sum_{i=1}^n a_i a_j (P(A_i \cap A_j) - P(A_i)P(A_j)) &= \sum_{i=1}^n a_i a_j (\mathbb{E}[\mathbb{1}_{A_i \cap A_j}] - \mathbb{E}[\mathbb{1}_{A_i}]\mathbb{E}[\mathbb{1}_{A_j}]) \\ &= \mathbb{E} \left[\sum_{i=1}^n a_i a_j \mathbb{1}_{A_i} \mathbb{1}_{A_j} \right] - \sum_{i=1}^n \mathbb{E}[a_i \mathbb{1}_{A_i}] \mathbb{E}[a_j \mathbb{1}_{A_j}] \\ &= \mathbb{E} \left[\left(\sum_{i=1}^n a_i \mathbb{1}_{A_i} \right)^2 \right] - \left(\sum_{i=1}^n a_i \mathbb{E}[\mathbb{1}_{A_i}] \right)^2 \end{aligned}$$

By Jensen's inequality, the above quantity is positive because the function $\phi : (X_1, \dots, X_n) \mapsto (\sum_{i=1}^n a_i X_i)^2$ is convex. Therefore, K is **p.d.**

1.4

1.5

Exercise 2. RKHS

2.1 $\alpha K_1 + \beta K_2$ is p.d. as sum of p.d. kernels and because α and β are positive scalars. Let \mathcal{H}_1 and \mathcal{H}_2 be their respective RKHS.

Since, K_1 and K_2 are p.d., by Aronszajn's theorem there exist Φ_1, Φ_2 mappings from \mathcal{X} to \mathcal{H}_1 and \mathcal{H}_2 such that $\forall x, y \in \mathbb{R}^2$

$$\begin{aligned} K(x, y) &\triangleq \alpha K_1(x, y) + \beta K_2(x, y) \\ &= \langle \sqrt{\alpha} \Phi_1(x), \sqrt{\alpha} \Phi_1(y) \rangle_{\mathcal{H}_1} + \langle \sqrt{\beta} \Phi_2(x), \sqrt{\beta} \Phi_2(y) \rangle_{\mathcal{H}_2} \\ &= \langle \Phi(x), \Phi(y) \rangle_{\mathcal{H}} \end{aligned}$$

Where $\langle \cdot, \cdot \rangle_{\mathcal{H}}$ is defined above and

$$\begin{aligned} \Phi : \mathcal{X} &\rightarrow \mathcal{H} \triangleq \sqrt{\alpha} \mathcal{H}_1 + \sqrt{\beta} \mathcal{H}_2 \\ x &\mapsto \sqrt{\alpha} \Phi_1(x) + \sqrt{\beta} \Phi_2(x) \end{aligned}$$

$\langle \cdot, \cdot \rangle_{\mathcal{H}}$ is clearly an inner product. And for any $(x_n)_{n \in \mathbb{N}}$ sequence of \mathcal{H} , we have that $\forall n \in \mathbb{N}, x_n = \sqrt{\alpha} x_{1,n} + \sqrt{\beta} x_{2,n}$ and $\forall n, m \in \mathbb{N}$

$$\begin{aligned} \|x_n - x_m\|_{\mathcal{H}}^2 &= \|\sqrt{\alpha}(x_{1,n} - x_{1,m}) + \sqrt{\beta}(x_{2,n} - x_{2,m})\|_{\mathcal{H}}^2 \\ &= \langle \sqrt{\alpha}(x_{1,n} - x_{1,m}) + \sqrt{\beta}(x_{2,n} - x_{2,m}), \sqrt{\alpha}(x_{1,n} - x_{1,m}) + \sqrt{\beta}(x_{2,n} - x_{2,m}) \rangle_{\mathcal{H}} \\ &= \alpha \langle x_{1,n} - x_{1,m}, x_{1,n} - x_{1,m} \rangle_{\mathcal{H}_1} + \beta \langle x_{2,n} - x_{2,m}, x_{2,n} - x_{2,m} \rangle_{\mathcal{H}_2} \quad (\text{by definition of } \langle \cdot, \cdot \rangle_{\mathcal{H}}) \\ &= \alpha \|x_{1,n} - x_{1,m}\|_{\mathcal{H}_1}^2 + \beta \|x_{2,n} - x_{2,m}\|_{\mathcal{H}_2}^2 \end{aligned}$$

Therefore, if (x_n) is a Cauchy sequence, $(x_{1,n})$ and $(x_{2,n})$ are also Cauchy sequences. Because \mathcal{H}_1 and \mathcal{H}_2 are Hilbert spaces, the two sequences converge and (x_n) converges. **\mathcal{H} is a Hilbert space.**

Let's consider the functions $K_x : t \mapsto K(x, t) = \alpha K_1(x, t) + \beta K_2(x, t)$ for $x \in \mathcal{X}$

Let, $f \in \mathcal{H}, x \in \mathcal{X}$. f can be written $f = \sqrt{\alpha} f_1 + \sqrt{\beta} f_2$ with $f_1 \in \mathcal{H}_1$ and $f_2 \in \mathcal{H}_2$.

$$\begin{aligned} f(x) &= \sqrt{\alpha} f_1(x) + \sqrt{\beta} f_2(x) \\ &= \sqrt{\alpha} \langle f_1, K_{1,x} \rangle_{\mathcal{H}_1} + \sqrt{\beta} \langle f_2, K_{2,x} \rangle_{\mathcal{H}_2} \\ &= \langle f, K_x \rangle_{\mathcal{H}} \end{aligned}$$

2.2

Exercise 3. RKHS

3.1 \mathcal{H} is Hilbert:

\mathcal{H} is a vector space of functions. $\langle \cdot, \cdot \rangle_{\mathcal{H}}$ is a symmetric bilinear form verifying $\forall f, \langle f, f \rangle_{\mathcal{H}} \geq 0$.

Since f is absolutely continuous, it has a derivative almost everywhere and the following equality holds $\forall x \in [0, 1]$

$$\begin{aligned} |f(x)|^2 &= \left| f(0) + \int_0^x f'(x) dx \right|^2 \\ &= \left| \int_0^x f'(x) dx \right|^2 \quad (f(0) = 0) \\ &\leq x \cdot \int_0^x f'(u)^2 du = x \cdot \langle f, f \rangle_{\mathcal{H}} \end{aligned}$$

$\langle f, f \rangle_{\mathcal{H}} \implies f = 0$ and $\langle \cdot, \cdot \rangle_{\mathcal{H}}$ is therefore an inner product. **\mathcal{H} is a pre-Hilbert space with $\langle \cdot, \cdot \rangle_{\mathcal{H}}$ as inner product.**

Let (f_n) a Cauchy sequence of \mathcal{H} . (f'_n) is a Cauchy sequence of $L^2([0, 1])$ which is complete. Therefore it converges to a function $g \in L^2([0, 1])$.

Since for all (n, m) , $x \in [0, 1]$, $|f_n(x) - f_m(x)|^2 \leq x \cdot \|f_n - f_m\|_{\mathcal{H}}^2$, the sequence $f_n(x)$ is Cauchy for any x and converges to a real number $f(x)$. And since $f(x) = \lim_{n \rightarrow \infty} f_n(x) = \lim_{n \rightarrow \infty} \int_0^x f'_n(u) du = \int_0^x g(u) du$, f is absolutely continuous and $f' = g$ almost everywhere. Moreover, $f' \in L^2([0, 1])$ and $f(0) = \lim_{n \rightarrow \infty} f_n(0) = 0$.

Finally, $\|f_n - f\|_{\mathcal{H}} = \|f'_n - g\|_{L^2([0, 1])} \rightarrow 0$ and $f \in \mathcal{H}$. **\mathcal{H} is complete, therefore \mathcal{H} is a Hilbert space.**

Reproducing property:

We will now show that \mathcal{H} is the RKHS with corresponding kernel $K : (x, y) \rightarrow \min(x, y)$ on $[0, 1]$.

For $x \in [0, 1]$, the function $K_x = \min(x, \cdot)$ is differentiable except on the singleton x which has a null measure, it is absolutely continuous. Its derivative is square integrable and $\min(x, 0) = 0$, **therefore K_x is in \mathcal{H} for all x .**