

KERNEL METHODS IN ML

Homework 1 12 16 17 2 3

Exercise 1. Kernels

2. $X = \mathbb{N}$ $K(x, x') = 2^{x+x'}$

• Let $x, x' \in \mathbb{N}$, $K(x, x') = 2^{x+x'} = 2^{x'+x} = K(x', x)$

• Let $N \in \mathbb{N}$, $(x_1, \dots, x_N) \in \mathbb{N}^N$ and $(a_1, \dots, a_N) \in \mathbb{R}^N$

$$\sum_{i=1}^N \sum_{j=1}^N a_i a_j K(x_i, x_j) = \sum_{i=1}^N \sum_{j=1}^N a_i a_j 2^{x_i+x_j} = \left(\sum_{i=1}^N a_i 2^{x_i} \right)^2 \geq 0$$

So K is positive definite

6. $X = \mathbb{R}$ $K(x, x') = \cos(x+x')$

• Take $N=2$, $x_1 = x_2 = \pi/2$ and $a_1 = a_2 = 1$

One has $\sum_{i=1}^2 \sum_{j=1}^2 a_i a_j K(x_i, x_j) = 4 \cdot \cos(\pi) = -4 < 0$

So K is not positive definite

7. $X = \mathbb{R}$ $K(x, x') = \cos(x-x')$

• Let $x, x' \in \mathbb{R}$, $K(x, x') = \cos(x-x') = \cos(x'-x) = K(x', x)$

• Let $N \in \mathbb{N}$, $(x_1, \dots, x_N) \in \mathbb{R}^N$ and $(a_1, \dots, a_N) \in \mathbb{R}^N$

$$\sum_{i=1}^N \sum_{j=1}^N a_i a_j K(x_i, x_j) = \sum_{i=1}^N \sum_{j=1}^N a_i a_j \cos(x_i - x_j)$$

$$= \sum_{i=1}^N \sum_{j=1}^N a_i a_j (\cos x_i \cos x_j + \sin x_i \sin x_j)$$

$$= \left(\sum_{i=1}^N a_i \cos(x_i) \right)^2 + \left(\sum_{i=1}^N a_i \sin(x_i) \right)^2 \geq 0$$

So K is positive definite

Exercise 2. Function and kernel boundedness

Let $K: X \times X \rightarrow \mathbb{R}$ a pd kernel s.t. $\forall x, z \in X, K(x, z) \leq b^2$

Let $f \in \mathcal{H}$ such that $\|f\|_{\mathcal{H}} \leq 1$

Let $x \in X$.

$$\begin{aligned} |f(x)| &= |\langle f, K_x \rangle_{\mathcal{H}}| \leq \|f\|_{\mathcal{H}} \cdot \|K_x\|_{\mathcal{H}} \quad (\text{Cauchy-Schwarz}) \\ &\leq 1 \cdot \sqrt{\langle K_x, K_x \rangle_{\mathcal{H}}} = \sqrt{K(x, x)} \\ &\leq b \end{aligned}$$

Therefore, $\sup_{x \in X} |f(x)| \leq b \Rightarrow \underline{\|f\|_{\infty} \leq b}$

Exercise 3. Non-expansiveness of the Gaussian kernel

$K: \mathbb{R}^p \times \mathbb{R}^p \rightarrow \mathbb{R}$

$\varphi: \mathbb{R}^p \rightarrow \mathcal{H}$ s.t. $K(x, x') = \langle \varphi(x), \varphi(x') \rangle_{\mathcal{H}}$

$$x, x' \mapsto e^{-\frac{\alpha}{2} \|x - x'\|^2}$$

Let $x \in \mathbb{R}^p$.

$$\begin{aligned} \|\varphi(x) - \varphi(x')\|_{\mathcal{H}} &= \sqrt{\langle \varphi(x) - \varphi(x'), \varphi(x) - \varphi(x') \rangle_{\mathcal{H}}} \\ &= \sqrt{K(x, x) + K(x', x') - 2K(x, x')} \\ &= \sqrt{2 \left(1 - e^{-\frac{\alpha}{2} \|x - x'\|^2} \right)} \end{aligned}$$

Define $g: \mathbb{R}^+ \rightarrow \mathbb{R}$

$$t \mapsto \sqrt{2 \left(1 - e^{-\frac{t^2}{2}} \right)}$$

One has $\forall t \in \mathbb{R}^+, 0 \leq 1 - e^{-\frac{t^2}{2}} \leq \frac{t^2}{2} \Rightarrow 0 \leq 2 \left(1 - e^{-\frac{t^2}{2}} \right) \leq t^2$
 $\Rightarrow g(t) \leq t$

Therefore $\underline{\|\varphi(x) - \varphi(x')\|_{\mathcal{H}} = g(\sqrt{\alpha} \|x - x'\|) \leq \sqrt{\alpha} \|x - x'\|}$