

Exercise Nr. 6
(Deadline 30.05.2022)

1**a)**

Let k_1, k_2 be kernel functions with

$$\begin{aligned}\phi_1 : X &\rightarrow H_1, \phi_2 : X \rightarrow H_2 \\ k_1(x, y) &= \langle \Phi_1(x), \Phi_1(y) \rangle \\ k_2(x, y) &= \langle \Phi_2(x), \Phi_2(y) \rangle \\ k(x, y) &= \alpha_1 k_1 + k_2 \\ &= \alpha_1 k_1(x, y) + k_2(x, y) \\ &= \alpha_1 \langle \Phi_1(x), \Phi_1(y) \rangle + \langle \Phi_2(x), \Phi_2(y) \rangle \\ &= \langle \alpha_1 \Phi_1(x), \Phi_1(y) \rangle + \langle \Phi_2(x), \Phi_2(y) \rangle \\ &= \langle \alpha_1 \Phi_1(x) + \Phi_2(x), \Phi_1(y) + \Phi_2(y) \rangle \\ &= \alpha_1 \langle \Phi_1(x) + \Phi_2(x), \Phi_1(y) + \Phi_2(y) \rangle\end{aligned}$$

Let $\phi_3 : X \rightarrow H_3$ be a Hilbert space with

$$\begin{aligned}H_3 &= H_1 \cup H_2, \Phi_3(x) = \Phi_1(x) \cup \Phi_2(x) \\ &= \alpha_1 \langle \Phi_3(x), \Phi_3(y) \rangle\end{aligned}$$

Defining $k' = \langle \Phi_3(x), \Phi_3(y) \rangle$

Let K' be the kernel matrix of k' :

$$\begin{aligned}\Rightarrow c^T K c &\leq 0 \forall c \\ \Rightarrow \alpha_1 c^T K' c &= c^T \alpha_1 K' c \leq 0 \forall c \\ \Rightarrow \alpha_1 K' &= K \text{ is the kernel matrix of } k \\ \Rightarrow k &\text{ is kernel}\end{aligned}$$

b)

Proof by counterexample: Let $k_1(x, y) = x^T y, k_2(x, y) = 2x^T y$

$$\begin{aligned}k &= k_1(x, y) - k_2(x, y) = -x^T y \Rightarrow \text{not psd} \\ &\Rightarrow \text{Therefore } k \text{ is not a valid kernel}\end{aligned}$$

c)

1. Proof: $k = k_1 * k_2$ is a kernel with k_1, k_2 being any kernels

Let k_1, k_2 be kernel functions with

$$\phi_1 : X \rightarrow H_1, \phi_2 : X \rightarrow H_2$$

$$k_1(x, y) = \langle \Phi_1(x), \Phi_1(y) \rangle$$

$$k_2(x, y) = \langle \Phi_2(x), \Phi_2(y) \rangle$$

$$k = k_1 \cdot k_2$$

$$= \langle \Phi_1(x), \Phi_1(y) \rangle \cdot \langle \Phi_2(x), \Phi_2(y) \rangle$$

$$= \sum_i (\Phi_1)_i(x) \cdot (\Phi_1)_i(y) \cdot \sum_j (\Phi_2)_j(x) \cdot (\Phi_2)_j(y)$$

$$= \sum_i \sum_j (\Phi_1)_i(x) \cdot (\Phi_1)_i(y) \cdot (\Phi_2)_j(x) \cdot (\Phi_2)_j(y)$$

$$= \sum_i \sum_j (\Phi_1)_i(x) \cdot (\Phi_2)_j(x) \cdot (\Phi_1)_i(y) \cdot (\Phi_2)_j(y)$$

$$= \sum_i \sum_j (\Phi_3)_{ij}(x) \cdot (\Phi_3)_{ij}(y)$$

$$= \langle \Phi_3(x), \Phi_3(y) \rangle$$

With $\phi_3 : X \rightarrow H_3 : (\Phi_3)_{ij}(x) = (\Phi_1)_i(x) \cdot (\Phi_2)_j(x)$

$H_3 = H_1 \times H_2$ is a Hilbert space, the norm is given by: $\|(x, y)\| = \sqrt{\|x\|_{H_1}^2 + \|y\|_{H_2}^2}$

$\implies k$ is kernel

2. Proof: $\bar{k} = f(x)f(y)$ is a kernel with $f(x), f(y)$ being any functions

$$\bar{k} = f(x)f(y) = \langle f(x), f(y) \rangle$$

if f is one-dimensional, \bar{k} is a kernel as well. From above we know that the multiplication of two kernels is a kernel as well. Therefore $k = f(x) * k_1(x, y) * f(y) = k_1(x, y) * f(x)f(y) = k_1 * \bar{k}$ is a kernel.

d)

$$\begin{aligned} k(x, y) &= \frac{k_1(x, y)}{\sqrt{k_1(x, x)k_1(y, y)}} \\ &= \frac{1}{\sqrt{k_1(x, x)}} k_1(x, y) \frac{1}{\sqrt{k_1(y, y)}} \\ &= f(x)k_1(x, y)f(y) \end{aligned}$$

$$\text{with } f(x) = \frac{1}{\sqrt{k_1(x, x)}}$$

$$f(y) = \frac{1}{\sqrt{k_1(y, y)}}$$

see c) for the proof that a k of this form is a kernel

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⇒ next weeks assignment

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a)

The setting can be applied to friendship connections within communities or to the travelling salesman problem with different locations on a map.

b)

$$\begin{aligned}
 f^T L f &= \sum_{i,j} f_i f_j L_{ij} \\
 &= \sum_{i,j} f_i f_j (D_{ij} - A_{ij}) \\
 &= \sum_{i,j} f_i f_j D_{ij} - \sum_{i,j} f_i f_j A_{ij} \\
 &= \sum_i D_{ii} f_i^2 - \sum_{i \sim j} f_i f_j \\
 &= \frac{1}{2} \sum_i D_{ii} f_i^2 - \sum_{i \sim j} f_i f_j + \frac{1}{2} \sum_j D_{jj} f_j^2 \\
 &\quad \left(= \frac{1}{2} \sum_i \left[\sum_{j \text{ with: } i \sim j} f_i^2 \right] - \sum_{i \sim j} f_i f_j + \frac{1}{2} \sum_j \left[\sum_{i \text{ with: } i \sim j} f_i^2 \right] \right) \\
 &= \frac{1}{2} \sum_{i \sim j} f_i^2 - \sum_{i \sim j} f_i f_j + \frac{1}{2} \sum_{i \sim j} f_j^2 \\
 &= \frac{1}{2} x \sum_{i \sim j} (f_i^2 - 2f_i f_j + f_j^2) \\
 &= \frac{1}{2} \sum_{i \sim j} (f_i - f_j)^2 \\
 &= \Omega(f)^2
 \end{aligned}$$

c)

Well Defined

$$f, g \in \mathcal{H} \implies f, g \in \mathbb{R}^n$$

thus, they are unique and $\langle f, g \rangle_{\mathcal{H}} = f^T L g$ is well-defined

Linearity

show:

$$\langle x_1 + x_2, y \rangle = \langle x_1, y \rangle + \langle x_2, y \rangle$$

$$\begin{aligned} \langle f_1 + f_2, g \rangle_{\mathcal{H}} &= (f_1 + f_2)^T Lg \\ &= \sum_{ij} (f_{1i} + f_{2i}) L_{ij} g_j \\ &= \sum_{ij} f_{1i} L_{ij} g_j + \sum_{ij} f_{2i} L_{ij} g_j \\ &= f_1^T Lg + f_2^T Lg \\ &= \langle f_1, g \rangle_{\mathcal{H}} + \langle f_2, g \rangle_{\mathcal{H}} \end{aligned}$$

show:

$$\langle \lambda x, y \rangle = \lambda \langle x, y \rangle \quad \forall \lambda \in \mathbb{R}$$

$$\begin{aligned} \langle \lambda f, g \rangle_{\mathcal{H}} &= (\lambda f^T) Lg \\ &= \lambda (f^T Lg) \\ &= \lambda \langle f, g \rangle_{\mathcal{H}} \end{aligned}$$

Symmetry

show:

$$\langle x, y \rangle = \langle y, x \rangle$$

$$\begin{aligned} \langle f, g \rangle_{\mathcal{H}} &= f^T Lg \\ &= \sum_{ij} f_i L_{ij} g_j \\ &= \sum_{ij} g_j L_{ij} f_i \\ &= g^T Lf \\ &= \langle g, f \rangle_{\mathcal{H}} \end{aligned}$$

Pos. Def.

show:

$$\langle x, x \rangle \geq 0$$

$$\begin{aligned} \langle f, f \rangle_{\mathcal{H}} &= f^T Lg \\ &= \Omega(f)^2 \geq 0 \end{aligned}$$

show:

$$\langle x, x \rangle = 0 \implies x = 0$$

$$\begin{aligned} \langle f, f \rangle_{\mathcal{H}} &= 0 \\ \implies \Omega(f)^2 &= 0 \\ \implies \sum_{i \sim j} (f_i - f_j)^2 &= 0 \\ \implies f_i &= f_j \quad \forall i, j : i \sim j \end{aligned}$$

as graph is connected:

$$\begin{aligned} \implies f_i &= f_j \quad \forall i, j \\ \text{as } \sum_{i=1}^n f_i &= 0 : \\ \implies f_i &= 0 \quad \forall i \\ \iff f &= 0 \end{aligned}$$

$\implies \langle f, g \rangle_{\mathcal{H}}$ is a scalar product on \mathcal{H}

d)

Let $k(i, j) \in \mathbb{R}$ be the weight of the edge between the nodes i and j .
Then $\Phi(i)$ is the vector of edge weights to all other nodes.