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# Exercise Sheet 5

A kernel function  $k : \mathbb{R}^d \times \mathbb{R}^d \to \mathbb{R}$  must satisfy the *Mercer's condition*, which verifies that for any sequence of data points  $x_1, \ldots, x_n \in \mathbb{R}^d$  and coefficients  $c_1, \ldots, c_n \in \mathbb{R}$  the inequality

$$\sum_{i=1}^{n} \sum_{j=1}^{n} c_i c_j k(x_i, x_j) \ge 0$$

is satisfied. If it is the case, the kernel is called a Mercer kernel.

Conversely, the representer theorem states that if k is a Mercer kernel on  $\mathbb{R}^d$ , then there exists a Hilbert space (i.e., a finite or infinite dimensional  $\mathbb{R}$ -vector space with norm and scalar product)  $\mathcal{F}$ , the so-called feature space, and a continuous map  $\varphi : \mathbb{R}^d \to \mathcal{F}$ , such that

$$k(x, x') = \langle \varphi(x), \varphi(x') \rangle_{\mathcal{F}}$$
 for all  $x, x' \in \mathbb{R}^d$ .

### Exercise 1: Mercer Kernels $(3 \times 20 \text{ P})$

- (a) Show that the following are Mercer kernels.
  - i.  $k(x, x') = \langle x, x' \rangle$
- ii.  $k(x,x') = f(x) \cdot f(x')$  where  $f: \mathbb{R}^d \to \mathbb{R}$  is an arbitrary continuous function
- (b) Let  $k_1, k_2$  be two Mercer kernels, for which we assume the existence of a finite-dimensional feature map associated to them. Show that the following are again Mercer kernels.
  - i.  $k(x, x') = k_1(x, x') + k_2(x, x')$
- ii.  $k(x, x') = k_1(x, x') \cdot k_2(x, x')$
- (c) Show using the results above that the polynomial kernel of degree d, where  $k(x, x') = (\langle x, x' \rangle + \vartheta)^d$  and  $\vartheta \in \mathbb{R}^+$ , is a Mercer kernel.

### Exercise 2: The Feature Map $(4 \times 10 P)$

Consider the homogenous polynomial kernel k of degree 2 which is  $k: \mathbb{R}^2 \times \mathbb{R}^2 \to \mathbb{R}$ , where

$$k(x,y) = \langle x, y \rangle^2 = \Big(\sum_{i=1}^2 x_i y_i\Big)^2.$$

- (a) Show that  $\mathcal{F} = \mathbb{R}^3$  and  $\varphi \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1^2 \\ \sqrt{2} x_1 x_2 \\ x_2^2 \end{pmatrix}$  are possible choices for feature space and feature map.
- (b) Consider the unit circle  $C = \left\{ \begin{pmatrix} \cos \theta \\ \sin \theta \end{pmatrix} ; \ 0 \le \theta < 2\pi \right\}$ . Show that the image  $\varphi(C)$  lies on a plane H in  $\mathbb{R}^3$ .
- (c) Consider the plane  $A = \left\{ \begin{pmatrix} t \\ s \end{pmatrix}; t, s \in \mathbb{R} \right\}$ . Find a point P in  $\mathcal{F}$  which is not contained in  $\varphi(A)$ .
- (d) Find a feature map associated to the kernel  $k : \mathbb{R}^d \times \mathbb{R}^d \to \mathbb{R}$  with  $k(x,y) = \langle x,y \rangle^2 = \Big(\sum_{i=1}^d x_i \, y_i\Big)^2$ .

A kernel function  $k: \mathbb{R}^d \times \mathbb{R}^d \to \mathbb{R}$  must satisfy the *Mercer's condition*, which verifies that for any sequence of data points  $x_1, \ldots, x_n \in \mathbb{R}^d$  and coefficients  $c_1, \ldots, c_n \in \mathbb{R}$  the inequality

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i. 
$$\sum_{i=A}^{n} \sum_{j=A}^{n} C_{i}C_{j} k(x_{i}, x_{j}) = \sum_{l=A}^{n} \sum_{j=A}^{n} C_{i}C_{j} \langle x_{i}, x_{j} \rangle$$

$$= \sum_{l=A}^{n} \sum_{l=A}^{n} C_{i}C_{l} \sum_{l=A}^{d} \langle x_{i}, x_{i} \rangle \langle x_{i} \rangle$$

$$= \sum_{l=A}^{n} \left( \sum_{l=A}^{n} \sum_{l=A}^{n} C_{i}x_{i} \rangle \langle x_{i} \rangle \langle x_{i} \rangle \langle x_{i} \rangle \rangle$$

$$= \sum_{l=A}^{n} \left( \sum_{l=A}^{n} \sum_{l=A}^{n} C_{i}x_{i} \rangle \langle x_{i} \rangle \langle x_{i} \rangle \langle x_{i} \rangle \rangle$$

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$$\sum_{i,j} c_{i}c_{j}k(x_{i},x_{j}) = \sum_{i,j} c_{i}c_{j}\langle x_{i},x_{j}\rangle$$

$$= \sum_{i,j} \langle c_{i} \cdot x_{i}, c_{j} \cdot x_{j}\rangle$$

$$= \langle \sum_{i} c_{i}x_{i} \mid \sum_{j} c_{j}x_{j}\rangle$$

$$= \| \sum_{i} c_{j}x_{i}\|^{2} \ge 0$$

$$\text{Bilinearity:}$$

$$\alpha \cdot \langle x_{i}y \rangle = \langle \alpha \cdot x_{i}y \rangle = \langle x_{i}\alpha \cdot y_{i}\rangle$$

$$\langle x_{i}y \rangle + \langle x_{i} \ge \rangle = \langle x_{i}y + \xi \rangle$$

$$\langle (x) = x \qquad K = Q(x) \Phi(x)^{T}$$

$$K_{ij} = \langle x_{i}, x_{j}\rangle \rightarrow c^{T} Kc > 0$$

$$= \langle \Phi(x_{i}), \Phi(x_{j})\rangle$$

ii. 
$$\sum_{i=1}^{N} \sum_{j=1}^{N} C_i C_j k(x_i, x_j) = \sum_{i=1}^{N} \sum_{j=1}^{N} C_i C_j f(x_i) f(x_j)$$

$$= \sum_{i=1}^{N} C_i f(x_i) \sum_{j=1}^{N} C_j f(x_j)$$

$$= \left(\sum_{i=1}^{N} C_i f(x_i)\right)^{\frac{1}{2}} \geqslant 0$$

- (b) Let  $k_1, k_2$  be two Mercer kernels, for which we assume the existence of a finite-dimensional feature map associated to them. Show that the following are again Mercer kernels.
  - i.  $k(x, x') = k_1(x, x') + k_2(x, x')$
- ii.  $k(x, x') = k_1(x, x') \cdot k_2(x, x')$

i. 
$$\sum_{i=A}^{\infty} \sum_{j=A}^{\infty} C_{i}C_{j} k(x_{i}, x_{j}) = \sum_{i=A}^{\infty} \sum_{j=A}^{\infty} C_{i}C_{j} \left(k_{A}(x_{1}, x_{j}) + k_{2}(x_{i}, x_{j})\right)$$

$$= \sum_{i=A}^{\infty} \sum_{j=A}^{\infty} C_{i}C_{j} k_{A}(x_{1}, x_{j}) + \sum_{i=A}^{\infty} \sum_{j=A}^{\infty} C_{i}C_{j} k_{A}(x_{1}, x_{j})$$

$$\geqslant 0$$

$$\sum_{i=1}^{n} \sum_{j=1}^{n} C_{i}C_{j} k(x_{i}, x_{j}) = \sum_{i=1}^{n} \sum_{j=1}^{n} C_{i}C_{j} (k_{1}(x_{i}, x_{j}) k_{2}(x_{i}, x_{j}))$$

$$= \sum_{i=1}^{n} \sum_{j=1}^{n} C_{i}C_{j} \langle \varphi_{1}(x_{i}), \varphi_{2}(x_{j}) \rangle \langle \varphi_{2}(x_{i}), \varphi_{2}(x_{j}) \rangle$$

We assume Un, Us project the data into the same dimension.

If they project into different dimensions, we extend the lower-dimensional vector with zeros until the two projected dimensions are equal

Note that if there is a  $\varphi: \mathbb{R}^d \to \mathbb{R}^a$  but that  $k(x,x') = \langle \varphi(x), \varphi(x') \rangle$  for all  $x, x' \in \mathbb{R}^d$  then there will also be a  $\varphi: \mathbb{R}^d \to \mathbb{R}^{ab}$  such that  $k(x,x') = \langle \varphi(x), \varphi(x') \rangle$  for all  $x, x' \in \mathbb{R}^d$ .  $\varphi'(x) = \begin{pmatrix} \varphi_{\alpha}(x) \\ \varphi_{\alpha}(x) \\ \vdots \end{pmatrix}_{b}^{a}$ because  $\langle \varphi'(x), \varphi'(x) \rangle = \langle \varphi(x), \varphi(x) \rangle$ .

(c) Show using the results above that the polynomial kernel of degree d, where  $k(x, x') = (\langle x, x' \rangle + \vartheta)^d$  and  $\vartheta \in \mathbb{R}^+$ , is a Mercer kernel.

First, we prove that k(x,x') = V, VGR+ is a Mercer Kernel

$$\sum_{i=A}^{\infty} \sum_{j=A}^{n} C_{i}C_{j} k(x_{i}, x_{j}) = \sum_{i=A}^{\infty} \sum_{j=A}^{n} C_{i}C_{j} \cdot \emptyset$$

$$= \emptyset \cdot \sum_{i=A}^{\infty} \sum_{j=A}^{n} C_{i}C_{j}$$

$$= \emptyset \cdot \sum_{i=A}^{\infty} C_{i} \cdot \sum_{j=A}^{n} C_{j}$$

$$= \emptyset \cdot \left(\sum_{i=A}^{n} C_{i}\right)^{2} \geq 0$$

=> k(x,x') = V, V GR+ is a Mercer Kernel

Now we give a proof by induction

• Base Case: We show that  $k(X_1X') = (\langle X_1X' \rangle + U)^d$  is a Mercer Kernel for d=1i.e.  $k(X_1X') = \langle X_1X' \rangle + U$ 

k(x,x') = (x,x')+10 is clearly line became <x,x'> and v are Mercer terms. And the Sum of

two Mercer Kernel is also a Mercer Kernel. (from b.i)

· induction step: We show that for every  $d \ge 1$ , if  $k(x_1x') = (\langle x_1x' \rangle + U)^d$  is Mercer Kernel,

then  $k(x_1x') = (\langle x_1x' \rangle + U)^{d+1}$  also is a Mercer Kernel

$$k(x_1x') = (\langle x_1x' \rangle + U)^{d+1} = \frac{(\langle x_1x' \rangle + U)^d \cdot (\langle x_1x' \rangle + U)}{\text{We alkered}}$$

· Conolusion: Since holy the base cure and he induction step have been proved as true,

## Exercise 2: The Feature Map $(4 \times 10 \text{ P})$

Consider the homogenous polynomial kernel k of degree 2 which is  $k: \mathbb{R}^2 \times \mathbb{R}^2 \to \mathbb{R}$ , where

$$k(x,y) = \langle x, y \rangle^2 = \left(\sum_{i=1}^2 x_i y_i\right)^2.$$

(a) Show that  $\mathcal{F} = \mathbb{R}^3$  and  $\varphi \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1^2 \\ \sqrt{2} x_1 x_2 \\ x_2^2 \end{pmatrix}$  are possible choices for feature space and feature map.

$$R(x,y) = \angle \varphi(x), \varphi(y) > = \angle \varphi(x_1), \varphi(y_1) >$$

$$= \angle (x_1^2 + x_2^2), (x_2^2 + x_3^2 + x_4^2) >$$

$$= (x_1^2 + x_1^2 + x_2^2), (x_2^2 + x_3^2 + x_3^2) >$$

$$= (x_1^2 + x_1^2 + x_2^2), (x_2^2 + x_3^2 + x_3^2 + x_3^2) >$$

$$= (x_1^2 + x_1^2 + x_2^2), (x_2^2 + x_3^2 + x_3^2 + x_3^2 + x_3^2) >$$

$$= (x_1^2 + x_1^2 + x_2^2), (x_2^2 + x_3^2 + x_3$$

(b) Consider the unit circle  $C = \left\{ \begin{pmatrix} \cos \theta \\ \sin \theta \end{pmatrix} \; ; \; 0 \leq \theta < 2\pi \right\}$ . Show that the image  $\varphi(C)$  lies on a plane H in  $\mathbb{R}^3$ .

The plane can be written in the form

$$\begin{aligned}
H: \chi &= \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + r \cdot \begin{bmatrix} 0 \\ 0 \\ 1 \end{pmatrix} - \begin{pmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + s \cdot \begin{bmatrix} 1/2 \\ 5/2 \\ 1/2 \end{pmatrix} - \begin{pmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \\
&= \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} + r \cdot \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} + s \cdot \begin{pmatrix} -1/2 \\ 5/2 \\ 1/2 \end{pmatrix}
\end{aligned}$$

We can rewrite this into the normal-vector representation

$$H : \left( \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} X \begin{pmatrix} -1/2 \\ 52/2 \\ 1/2 \end{pmatrix} \right)^{T} \left( X - \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \right) = \begin{pmatrix} -52/2 \\ -52/2 \end{pmatrix}^{T} \left( X - \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \right)$$

( the cross product of two vector is and i results in a vector which is obtagoned to it and i)
The formula returns 0 for all points on the plane.

$$\begin{pmatrix}
-\sqrt{5}/2 \\
0 \\
-\sqrt{5}/2
\end{pmatrix} \cdot \left( \varphi(x) - \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \right) = \begin{pmatrix} -\sqrt{5}/2 \\ -\sqrt{5}/2 \end{pmatrix}^{T} \cdot \left( \begin{pmatrix} \cos(\theta)^{2} \\ \sin(\theta)^{2} \\ \sin(\theta)^{2} \end{pmatrix} - \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \right)$$

$$= \begin{pmatrix} -\sqrt{5}/2 \\ 0 \\ -\sqrt{5}/2 \end{pmatrix}^{T} \cdot \begin{pmatrix} \cos(\theta)^{2} - 1 \\ \sqrt{5} \cos(\theta) \sin(\theta) \\ \sin(\theta)^{2} \end{pmatrix}$$

$$= -\sqrt{5}/2 \cdot \left( \cos(\theta)^{2} - 1 \right) - \sqrt{2}/2 \sin(\theta)^{2}$$

$$= -\sqrt{5}/2 \cdot \left( \cos(\theta)^{2} + \sin(\theta)^{2} - 1 \right)$$

$$= -\sqrt{5}/2 \cdot \left( \cos(\theta)^{2} + \sin(\theta)^{2} - 1 \right)$$

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$$= -\sqrt{5}/2 \cdot \left( \cos(\theta)^{2} + \sin(\theta)^{2} - 1 \right)$$

Therefore the image &(C) lies on the place Hin R3

(c) Consider the plane  $A = \left\{ \begin{pmatrix} t \\ s \end{pmatrix} ; t, s \in \mathbb{R} \right\}$ . Find a point P in  $\mathcal{F}$  which is not contained in  $\varphi(A)$ .

$$\begin{array}{l} (1) = Q \begin{pmatrix} t \\ s \end{pmatrix} = \begin{pmatrix} t^2 \\ 5 t \\ s^2 \end{pmatrix} \\ \text{if We choose } t = 1 \text{ and } s = 1 \text{ , clearly we have } Q \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \\ \text{Then alle the Point } \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \text{ with } R \in \mathbb{R} \setminus \{12\} \text{ are not contained in } Q \setminus \{13\} \cdot \ell \cdot \ell \cdot \ell = 2 \text{ or } \ell \cdot \ell + 2 \text{ or } \ell \cdot \ell = 2 \text{ or } \ell = 2 \text{ or }$$

(d) Find a feature map associated to the kernel  $k : \mathbb{R}^d \times \mathbb{R}^d \to \mathbb{R}$  with  $k(x,y) = \langle x,y \rangle^2 = \Big(\sum_{i=1}^d x_i y_i\Big)^2$ .

$$k = \langle x, y \rangle^{2} = (X^{T}, Y)^{2} = (\langle x_{\Lambda} x_{2} \dots x_{d} \rangle) \begin{pmatrix} y_{\Lambda} \\ y_{\Lambda} \end{pmatrix}^{2}$$

$$= (\langle x_{\Lambda} x_{2} \dots x_{d} \rangle) \begin{pmatrix} y_{\Lambda} \\ y_{\Lambda} \end{pmatrix}^{2}$$

$$\Rightarrow \varphi(x) = \varphi \qquad \vdots \qquad = \begin{pmatrix} x_4 \\ \vdots \\ x_4 \\ \vdots \\ \vdots \\ x_4 \\ \vdots \\ x_4 \\ x_4 \end{pmatrix}$$

$$\phi(x) = \left[ (x_i^2)_i, (\sqrt{2} \times_i x_j)_{i < j} \right] \in \mathbb{R}^{d \cdot (d+\Lambda)/2} \rightarrow \phi(x) = \left[ (\times_i x_j)_{ij} \right] \in \mathbb{R}^{d^2} \rightarrow \langle \phi(x), \phi(y) \rangle \sim \mathcal{O}(d^2) \rightarrow \langle x_i, y_i \rangle^2 \sim \mathcal{O}(d)$$

