

# Machine Learning: Homework Assignment 7

E4525 Spring 2018,  
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## 1. Separating Hyperplane for two points

Assume we have two points  $x_+$  and  $x_-$  both in  $\mathcal{R}^D$ . We have that the class of  $x_+$  is  $y_+ = +1$  and the class of  $x_-$  is  $y_- = -1$ .

Give an explicit expression for the parameters  $w$  and  $b$  of the separating hyper-plane with maximum margin. Set the normalization such that  $w^T x_+ + b = 1$ .

## 2. Dual coefficients and Margin

Prove that, for the hard margin case, the optimal margin  $\rho$  satisfies

$$\frac{1}{\rho^2} = \sum_i \alpha_i \quad (1)$$

## 3. Kernel Construction and Mercer's Condition

Using Mercer's Condition that a kernel  $K(x, x')$  generates a scalar product if, and only if, the  $N \times N$  matrix,

$$K_{i,j} = K(x_i, x_j) \quad (2)$$

is positive semi-definite for all  $x_i, x_j$  in  $(x_1, \dots, x_N) \in \mathcal{X}$ .

And that a matrix is positive semi-definite if and only if

$$c^T K c \geq 0 \quad (3)$$

for all vectors  $c \in \mathbb{R}^N$ .

Show that

(a) The sum of two kernels is a kernel

$$K(x, x') = k_1(x, x') + k_2(x, x') \quad (4)$$

(b) The product of a kernel  $k_1$  and a scalar  $0 \leq \lambda \in \mathbb{R}$  is a kernel

$$K(x, x') = \lambda k_1(x, x') \quad (5)$$

(c) The scaling of a kernel  $k$  by a function is a kernel

$$K(x, x') = f(x)k_1(x, x')f(x') \quad (6)$$

#### 4. Product of two Kernels and Tensor Product

Let's assume that we have two kernels  $k^1$  and  $k^2$  that can be written as the scalar product in feature spaces  $\phi^1$  and  $\phi^2$  of dimension  $D_1$  and  $D_2$ :

$$\begin{aligned} k^1(x, x') &= \sum_i^{D_1} \phi_i^1(x) \phi_i^1(x') \\ k^2(x, x') &= \sum_j^{D_2} \phi_j^2(x) \phi_j^2(x') \end{aligned} \quad (7)$$

Show that the product kernel

$$K(x, x') = k^1(x, x')k^2(x, x') \quad (8)$$

can be represented by the *tensor product* feature space  $\phi$

$$\phi_k(x) = \phi_{i+D_1j}(x) = \phi_i^1(x) \phi_j^2(x) \quad (9)$$

for  $k = 1, \dots, D_1D_2$ , and

$$K(x, x') = \sum_{k=1}^{D_1D_2} \phi_k(x) \phi_k(x') \quad (10)$$

#### 5. Off Diagonal terms of Gram Matrix

- (a) Show that the diagonal terms of a kernel  $k(x, x)$  must be positive or zero.
- (b) Using the fact that the Gram matrix of two points  $x$  and  $x'$  must be positive semi-definite show that

$$k(x, x') \leq \sqrt{k(x, x)k(x', x')} \quad (11)$$

- (c) Give an example of a symmetric  $2 \times 2$  matrix where all entries are positive and yet the matrix is not positive semi-definite.
- (d) Give an example of a symmetric  $2 \times 2$  matrix where some elements are negative but the matrix is still positive semi-definite.