

# Advanced Machine Learning Homework 5

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### 1 Kernel Functions for Support Vector Machines

Given input dimension  $d$ , input data  $\mathbf{x}, \mathbf{y}$  can be expressed as  $\vec{\mathbf{x}} = (x_1, \dots, x_d), \vec{\mathbf{y}} = (y_1, \dots, y_d)$ , An SVM kernel function  $k(\vec{\mathbf{x}}, \vec{\mathbf{y}})$  can be expressed as the dot product of the transformation of  $\vec{\mathbf{x}}, \vec{\mathbf{y}}$  by a transformation function  $\gamma$ , that is to say,

$$k(\vec{\mathbf{x}}, \vec{\mathbf{y}}) = \gamma(\vec{\mathbf{x}}) \cdot \gamma(\vec{\mathbf{y}}) \quad (1)$$

#### Problems

Derive the transformation function  $\gamma$  for the following SVM kernels, also compute the VC Dimension for the SVM model based on these kernels.

1.  $k(\vec{\mathbf{x}}, \vec{\mathbf{y}}) = (\vec{\mathbf{x}} \cdot \vec{\mathbf{y}})^n$
2.  $k(\vec{\mathbf{x}}, \vec{\mathbf{y}}) = (\vec{\mathbf{x}} \cdot \vec{\mathbf{y}} + 1)^n$
3.  $k(\vec{\mathbf{x}}, \vec{\mathbf{y}}) = e^{-\sigma(\|\vec{\mathbf{x}}\|^2 + \|\vec{\mathbf{y}}\|^2)}$

## 1.  $k(\vec{x}, \vec{y}) = (\vec{x} \cdot \vec{y})^n$

展开:

$$(x_{\{1\}}y_{\{1\}} + x_{\{2\}}y_{\{2\}} + \dots + x_{\{d\}}y_{\{d\}})^n$$

根据多项式定理:

$$\sum \frac{n!}{n_1!n_2!\dots n_d!} (x_1y_1)^{n_1}(x_2y_2)^{n_2}\dots (x_dy_d)^{n_d}$$

,

其中  $n_1+n_2+\dots+n_d = n$

由此可得:

$$\gamma(\vec{x}) = \sqrt[n]{\sum \frac{n!}{n_1!n_2!\dots n_d!} (x_1)^{n_1}(x_2)^{n_2}\dots (x_d)^{n_d}}$$

VC dimension 取决于多项式的项数，因此为：  $C^{n-1}_{n+d-1} + 1$

2.  $k(\vec{x}, \vec{y}) = (\vec{x} \cdot \vec{y} + 1)^n$

和上面的类似，多一个常数项： 可以令  $\vec{x} = (\vec{x}, 1), \vec{y} = (\vec{y}, 1)$

$$\gamma(\vec{x}) = \sqrt{\sum \frac{n!}{n_1!n_2!\dots n_d!n_{d+1}!}}(x_1)^{n_1}(x_2)^{n_2}\dots(x_d)^{n_d}1^{n_{d+1}},$$

相应的，VC dimension =  $C^{n-1}_{n+d} + 1$

3.  $k(\vec{x}, \vec{y}) = e^{-\sigma(\|\vec{x}\|^2 + \|\vec{y}\|^2)}$

可得：  $\gamma(\vec{x}) = e^{-\sigma(\|\vec{x}\|^2)}$

dimension = 1, 因此对应的SVM VC dimension = 2.