

# The Exercise of SVM

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## 1 Kernel

(a) Determine the mapping  $\phi(x)$  such that the kernel

$$k(\mathbf{x}, \mathbf{z}) = (c + \mathbf{x}^T \mathbf{z})^2 = \phi(\mathbf{x}) \phi(\mathbf{z})$$

where  $\mathbf{x} = (x_1, x_2)^T$  and  $\mathbf{z} = (z_1, z_2)^T$ .

(b) Show, by a sketch, that an XOR is not linearly separable, but that after the mapping  $\phi(x)$  with  $c = 0$  it is linearly separable.

## 2 Steepest Gradient Descendent of SVM

Show that if the SVM cost function is written as

$$\mathcal{C}(\mathbf{w}) = \frac{1}{N} \sum_{i=1}^N \left( \frac{\lambda}{2} \|\mathbf{w}\|^2 + \max(0, 1 - y_i f(\mathbf{x}_i)) \right)$$

where  $f(\mathbf{x}_i) = \mathbf{w}^T \mathbf{x}_i$ , then using using steepest descent optimization,  $\mathbf{w}_{t+1}$  may be learnt from  $\mathbf{w}_t$  by cycling through the data with the following update rule,

$$\begin{aligned} \mathbf{w}_{t+1} &\leftarrow (1 - \eta\lambda) \mathbf{w}_t + \eta y_i \mathbf{x}_i \text{ if } y_i \mathbf{w}^T \mathbf{x}_i < 1 \\ &\leftarrow (1 - \eta\lambda) \mathbf{w}_t \text{ otherwise} \end{aligned}$$

where  $\eta$  is the learning rate.