# Kernel Matrix Calculation

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### The Kernel Matrix

Consider a data matrix X consisting of n vectors  $x_1, x_2, \ldots, x_n$  each of size m. The kernel matrix  $\mathbf{K}$  is symmetric and calculated from kernel functions of the form  $K(x_i, x_j)$  which forms element  $K_{i,j}$  of the kernel matrix [1].

# **Kernel Functions**

The following kernel functions are used in the article [2, 3, 4]:

• Linear Kernel (dot product)

$$K(x_i, x_j) = x_i^T x_j \tag{1}$$

Gaussian

$$K(x_i, x_j) = \exp\left(\frac{-\|x_i - x_j\|^2}{\theta}\right)$$
 (2)

• Polynomial

$$K(x_i, x_j) = (x_i^T x_j + \text{Offset})^d$$
(3)

Exponential

$$K(x_i, x_j) = \exp\left(\frac{-\|x_i - x_j\|}{\theta}\right) \tag{4}$$

• Log Kernel

$$K(x_i, x_j) = -\log(1 + ||x_i - x_j||^{\beta})$$
 (5)

• Cauchy

$$K(x_i, x_j) = \frac{1}{1 + \frac{\|x_i - x_j\|^2}{a}}$$
 (6)

• Power

$$K(x_i, x_j) = -\|x_i - x_j\|^{\beta} \tag{7}$$

• Wave

$$K(x_i, x_j) = \frac{\theta}{\|x_i - x_j\|} \sin\left(\frac{\|x_i - x_j\|}{\theta}\right)$$
 (8)

• Sigmoid

$$K(x_i, x_j) = \tanh\left(\beta_0 x_i^T x_j + \beta_1\right) \tag{9}$$

The kernel matrix calculation scales  $O(mn^2)$  with the dimensions of the data matrix X.

## References

- [1] An introduction to support vector machines and other kernel-based learning methods, Nello Cristianini & John Shawe-Taylor
- [2] Data Variant Kernel Analysis, Yuichi Motai
- [3] Classes of Kernels for Machine Learning: A Statistics Perspective, Marc G. Genton.
- [4] Conditionally Positive Definite Kernels for SVM Based Image Recognition, S. Boughorbel, J. -. Tarel and N. Boujemaa, 2005 IEEE International Conference on Multimedia and Expo, Amsterdam, 2005, pp. 113-116