

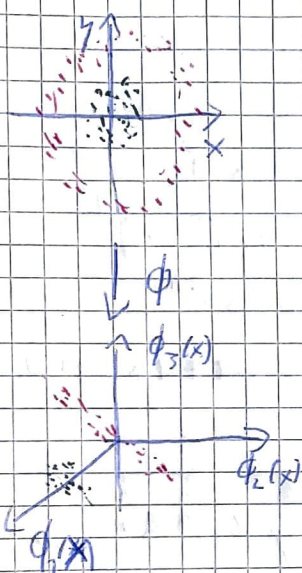
Sheet 5 Kernel Machine (structured kernel)

Linear Models

$X = x_1, \dots, x_N$ data
 $f(x) = y = y_1, \dots, y_N$ targets

$$f(x) = \beta^T x$$

often useful and good but do not solve non-linear problems



Solution: Apply non-linear feature map $\phi: \mathbb{R}^2 \rightarrow \mathbb{R}^3$

$$\phi(x_1, x_2) = (x_1^2, \sqrt{2}x_1x_2, x_2^2, x_1, x_2)$$

Summary

- structured kernels can be used for data that is not in \mathbb{R}^d : strings, trees
- structured kernels can be designed to incorporate prior domain-specific models
- other algorithms (LR, SVMs, PCA, CCA, ...) can be used with structured kernels

Kernel Models

$$f(x) = \beta^T \phi(x)$$

Tip

Express feature map as dot-product of nonlinear functions

$$x_1^2 y_1^2 + 2x_1 y_1 x_2 y_2 + x_2^2 y_2^2 = \left\langle \begin{pmatrix} x_1^2 \\ \sqrt{2}x_1x_2 \\ x_2^2 \end{pmatrix}, \begin{pmatrix} y_1^2 \\ \sqrt{2}y_1y_2 \\ y_2^2 \end{pmatrix} \right\rangle = \left\langle \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \right\rangle^2$$
$$\langle \phi(x), \phi(y) \rangle = k(x, y)$$

what is a kernel?

A (nonlinear) function $k: \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$

Which Kernels induce feature maps?

Mercer Theorem: All symmetric $k(x, y) = k(y, x)$ and PSD kernels!

Positive semi-definite (PSD):

$$\sum_{i=1}^N \sum_{j=1}^N c_i c_j k(x_i, x_j) \geq 0 \quad \forall x_1, \dots, x_N \in \mathbb{R}^d, c_1, \dots, c_N \in \mathbb{R}$$

Weighted Degree Kernel

$$k(x, y) = \sum_{d=0}^L \beta_d \sum_{i=1}^{L+1-d} S(x_{i:i+d-1}) = y_{i:i+d-1}$$

$L \in \{0, 1, 2, 3\}$

d : neighbors to check

AAACAAATAAG
TACCTAATTAT

$d=1$
 $d=2$
 $d=3$

$$\rightarrow k(x, y) = \phi(x)^T \phi(y)$$

$$\phi(x, y) \in \mathbb{R}^{\infty}$$

PSD Kernel Examples

Linear kernel

$$k(x, y) = x^T y$$

Polynomial

$$k(x, y) = (x^T y + a)^d$$

Gaussian

$$k(x, y) = e^{-\gamma \|x - y\|^2}$$

t-Student

$$k(x, y) = (a + \|x - y\|^2)^{-1}$$

String Kernels

Alphabet A : finite set of discrete symbols $A = \{s_1, \dots, s_d\}$

Examples: DNA: $A \in \{G, A, T, C\}$, $A = \{a, b, c, \dots\}$

String: concatenation of symbols of A

Task: compare two sequences $x = (a, b, c, c, e, a, b, d) \in A^L$
 $z = (a, b, a, c, c, c, b, e) \in A^L$

$$k(x, y) = \sum_{i=1}^L \mathbb{1}(x_i = z_i) \quad \mathbb{1}(x=z) \quad 1 \quad 1 \quad 0 \quad 1 \quad 0 \quad 1 \quad 0$$

$k: A^L \times A^L \rightarrow \mathbb{R}$ Kernel counts number of matching symbols

Feature map: $\phi(x) = [\mathbb{1}(x_i = a)]_{a \in A, i \in \{1, \dots, L\}}$

$$= \begin{pmatrix} a & 1 & 2 & \dots & L \\ b & 1 & 2 & \dots & L \\ c & 1 & 2 & \dots & L \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ z & 1 & 2 & \dots & L \end{pmatrix} \in \mathbb{R}^{\text{len}(A) \times \text{len}(A) \times L}$$

Variation: ^{some} shift tolerance $a < b$, full shift tolerance, subsequences, = Bag of words

Bag of Words

Implementation

sort words to save time when comparing!