SVM and Nearest Neighbor Classifiers

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K-Nearest Neighbor Classifier

- 对于一个新的点, 从训练数据中找出 k 个最接近的点
- 对 k 个点的标签进行分类
- 数据量大
- 合适的距离函数

距离函数

L1 distance

$$D(h_1,h_2) = \sum_{i=1}^{N} |h_1(i) - h_2(i)|$$
 (1)

 χ^2 distance

$$D(h_1, h_2) = \sum_{i=1}^{N} \frac{(h_1(i) - h_2(i))^2}{h_1(i) + h_2(i)}$$
 (2)

Quadratic distance

$$D(h_1,h_2) = \sum_{i,j} A_{ij} (h_1(i) - h_2(i))^2$$
 (3)

Earth Mover's Distance

$$EMD(S_1, S_2) = \sum_{i,j} rac{f_{ij}d(m_{1i}, m_{2j})}{f_{ij}} \hspace{1.5cm} (4)$$

SVM

Linear SVM

找到w和b使得 $\Phi(w) = ||w||^2 = w^T w$ 最小,对于 $\forall (x_i, y_i), i = 1...n, y_i(w^T x_i + b) \geq 1$

二次优化问题:构造对偶问题

寻找 $\alpha_1 \dots \alpha_n$ 使得

$$Q(\alpha) = \sum \alpha_i - \frac{\sum \sum \alpha_i \alpha_j y_i y_j x_i^T x_j}{2}$$
 (5)

最大化

其中

$$\sum_{\alpha_i y_i = 0} \alpha_i y_i = 0$$

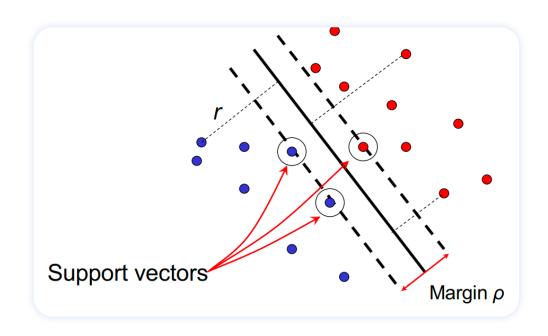
$$\alpha_i \ge 0, \forall \alpha_i$$

$$(6)$$

解得

$$w = \sum_{i} \alpha_i y_i x_i$$
 (7) $b = y_i - w x_i$

分类函数的决策边界



$$f(x) = wx + b = \sum_{i} \alpha_i y_i x_i x + b \tag{8}$$

Nonlinear SVMs

寻找 $\alpha_1 \dots \alpha_n$ 使得

$$Q(\alpha) = \sum \alpha_i - \frac{\sum \sum \alpha_i \alpha_j y_i y_j K(x_i, x_j)}{2}$$
 (9)

最大化

其中

$$\sum_{\alpha_i y_i = 0} \alpha_i y_i = 0 \tag{10}$$

$$\alpha_i \ge 0, \forall \alpha_i$$

解得

$$f(x) = \sum_{i} \alpha_i y_i K(x_i, x_j) + b \tag{11}$$

Kernel Functions

Linear:

$$K(x_i, x_j) = x_i^T x_j \tag{12}$$

Polynomial of power p:

$$K(x_i, x_j) = (1 + x_i^T x_j)^p (13)$$

Gaussian:

$$K(x_i,x_j) = e^{-rac{||x_i-x_j||}{2\sigma^2}}$$
 (14)

Histogram intersection kernel:

$$I(h_1, h_2) = \sum_{i=1}^{N} min(h_1(i), h_2(i))$$
 (15)

Generalized Gaussian kernel:

$$K(h_1, h_2) = exp(-\frac{D(h_1, h_2)^2}{A})$$
(16)