

# SVM for Regression

# SVM Recall

Two-class classification problem using linear model:

$$y(x) = w^T \phi(x) + b$$

# Regularized Error Function

In linear regression, we minimize the error function:

$$\frac{1}{2} \sum_{n=1}^N \{y_n - t_n\}^2 + \frac{\lambda}{2} \|w\|^2$$

Replace the quadratic error function by  $\epsilon$ -insensitive error function:

$$C \sum_{n=1}^N E_{\epsilon}(y(x_n) - t_n) + \frac{1}{2} \|w\|^2$$

An example of  $\epsilon$ -insensitive error function:

$$L_{\epsilon}(y) = \begin{cases} 0 & \text{for } |f(\mathbf{x}) - y| < \epsilon \\ |f(\mathbf{x}) - y| - \epsilon & \text{otherwise} \end{cases}$$

# Slack Variables

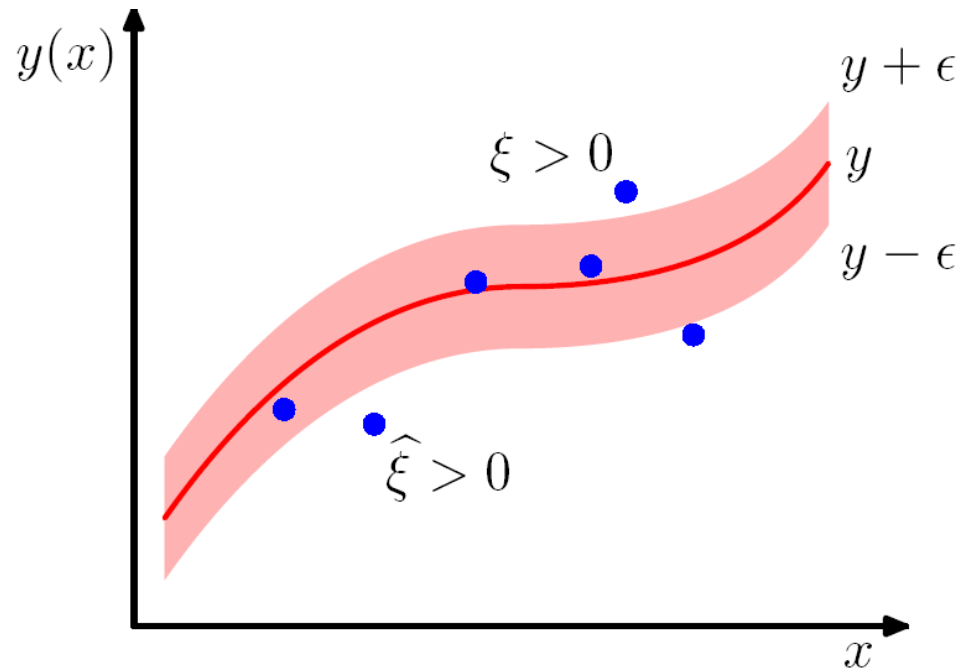
For a target point to lie inside the tube:

$$y_n - \epsilon \leq t_n \leq y_n + \epsilon$$

Introduce slack variables to allow points to lie outside the tube:

$$t_n \leq y(x_n) + \epsilon + \xi_n$$

$$t_n \geq y(x_n) - \epsilon - \xi_n^-$$



# Error Function for Support Vector Regression

Minimize:

$$C \sum_{n=1}^N (\xi_n + \xi_n^-) + \frac{1}{2} \|w\|^2$$

Subject to:

$$\xi_n \geq 0 \quad \text{and} \quad t_n \leq y(x_n) + \epsilon + \xi_n$$

$$\xi_n^- \geq 0 \quad t_n \geq y(x_n) - \epsilon - \xi_n^-$$

# Lagrangian

Minimize:

$$L = C \sum_{n=1}^N (\xi_n + \xi_n^-) + \frac{1}{2} \|w\|^2 - \sum_{n=1}^N (\mu_n \xi_n + \mu_n^- \xi_n^-) - \sum_{n=1}^N a_n (\epsilon + \xi_n + y_n - t_n) - \sum_{n=1}^N a_n^- (\epsilon + \xi_n^- - y_n + t_n)$$

$$\frac{\partial L}{\partial w} = 0 \Rightarrow w = \sum_{n=1}^N (a_n - a_n^-) \phi(x_n)$$

$$\frac{\partial L}{\partial b} = 0 \Rightarrow \sum_{n=1}^N (a_n - a_n^-) = 0$$

$$\frac{\partial L}{\partial \xi_n} = 0 \Rightarrow a_n + \mu_n = C$$

$$\frac{\partial L}{\partial \xi_n^-} = 0 \Rightarrow a_n^- + \mu_n^- = C$$

# Dual Form of Lagrangian

Maximize:

$$W(a, a^-) = -\frac{1}{2} \sum_{n=1}^N \sum_{m=1}^N (a_n - a_n^-)(a_m - a_m^-) k(x_n, x_m) - \epsilon \sum_{n=1}^N (a_n + a_n^-) + \sum_{n=1}^N (a_n - a_n^-) t_n$$

$$0 \leq a_n \leq C$$

$$0 \leq a_n^- \leq C$$

Prediction can be made using:

$$y(x) = \sum_{n=1}^N (a_n - a_n^-) k(x, x_n) + b$$

# How to determine b?

Karush-Kuhn-Tucker (KKT) conditions:

$$a_n (\epsilon + \xi_n + y_n - t_n) = 0$$

$$a_n^- (\epsilon + \xi_n^- - y_n + t_n) = 0$$

$$(C - a_n) \xi_n = 0$$

$$(C - a_n^-) \xi_n^- = 0$$

Support vectors are points that lie on the boundary or outside the tube

$$b = t_n - \epsilon - w^T \phi(x_n) = t_n - \epsilon - \sum_{m=1}^N (a_m - a_m^-) k(x_n, x_m)$$