

## Recall Linear Regression with Regularization

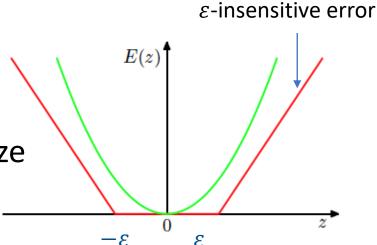
• Objective function of linear regression: 
$$\frac{1}{2}\sum_{i=1}^{n}(y_{i}-\hat{y}_{i})^{2}+\frac{\lambda}{2}||w||^{2}$$

- Let's replace the loss  $\frac{1}{2}\sum_{i=1}^{n}(y_i-\hat{y}_i)^2$ with  $\varepsilon$ -insensitive error function E(z)
  - Consider error only if they are more than  $\varepsilon$

• The objective now becomes to minimize

$$C\sum_{i=1}^{n} E(y_{i-}\hat{y}_{i}) + \frac{1}{2}||w||^{2}$$

C is the inverse regularization parameter

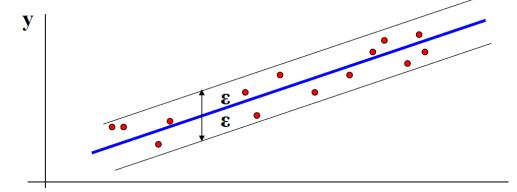


## Regression using SVC

$$f(\mathbf{x}) = \mathbf{w}^{\mathsf{T}} \mathbf{x} + b = \langle \mathbf{w}, \mathbf{x} \rangle + b$$

- Linear function for regression
- If all points assumed to be with  $\varepsilon$  neighborhood
  - With  $\varepsilon$ -insensitive error function
- The objective will be

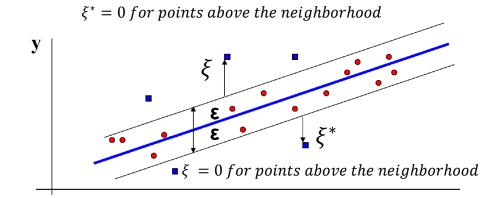
minimize 
$$\frac{1}{2} \|w\|^{2}$$
subject to 
$$\begin{cases} y_{i} - \langle w_{i}, x_{i} \rangle - b \leq \varepsilon \\ \langle w_{i}, x_{i} \rangle + b - y_{i} \leq \varepsilon \end{cases}$$



## Get it more general

- Not all data could satisfy neighborhood
- Allow some errors but add constraint to limit the error
- Objective function

minimize 
$$\frac{1}{2} \|w\|^2 + C \sum_{i=1}^{l} (\xi_i + \xi_i^*)$$
subject to 
$$\begin{cases} y_i - \langle w_i, x_i \rangle - b \le \varepsilon + \xi_i \\ \langle w_i, x_i \rangle + b - y_i \le \varepsilon + \xi_i^* \\ \xi_i, \xi_i^* \ge 0 \end{cases}$$



$$< w, x > = w^T x$$

Solve using Lagrange as before

## Solution

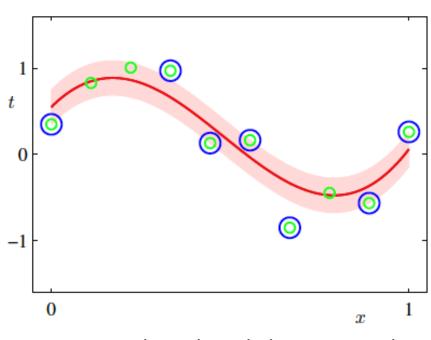
• Use Lagrange, we get

$$f(x) = \sum_{i=1}^{n} (C_i) < x_i, x > +b$$

• where  $C_i$  depends on Lagrange multipliers

- We can apply Kernel as before
  - Replace dot product with kernel <sup>-1</sup> to get Get non linear function

$$f(x) = \sum_{i=1}^{n} (C_i) K(x_i, x) + b$$



example with Radial Basis Kernel