

Support Vector machine

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Motivation:

Soft margin SVC(C – SVC) (w, b, ξ)

$$C - SVC \left\{ \begin{array}{ll} \text{Optimize} & f(w, \xi) = (f_1(w), f_2(\xi)) \\ \text{s.t} & y_i(w^T x_i + b) \geq 1 - \xi_i, i = 1, \dots, n \\ & \xi_i \geq 0, w \in \mathbb{R}^d, b \in \mathbb{R} \end{array} \right.$$

Linear Soft ε -band SVR

$$\text{LS} - \varepsilon\text{-band} - \text{SVR:} \left\{ \begin{array}{ll} \text{Optimize} & f(w, \xi) = (f_1(w), f_2(\xi^1), f_3(\xi^2)) \\ \text{s.t} & y_i - (w^T x_i - b) \leq \varepsilon + \xi_i^1, \quad i = 1, \dots, n \\ & w^T x_i + b - y_i \leq \varepsilon + \xi_i^2, \quad i = 1, \dots, n \\ & \xi_i^1 \geq 0, \xi_i^2 \geq 0, (w, b) \in \mathbb{R}^d \times \mathbb{R} \end{array} \right.$$

Motivation:

Soft margin SVC(C – SVC): Two Objectives Programming Problem

To optimize $f(w, \xi) = (f_1(w), f_2(\xi))$ we have to Minimize $f_1(w)$ and $f_2(\xi)$, where

- $f_1(w) = \frac{1}{\text{Margin}}$
- $f_2(\xi)$ is a noise

Linear Soft ϵ -band SVR: Three Objectives Programming Problem

To optimize $f(w, \xi) = (f_1(w), f_2(\xi^1), f_3(\xi^2))$ we have to Minimize $f_1(w)$, $f_2(\xi^1)$, and $f_3(\xi^2)$ where

- $f_1(w) = \frac{1}{\text{Margin}}$
- $f_2(\xi^1)$, and $f_3(\xi^2)$ are noises

Motivation

It's Multi Objective Programming Problem

- How to make it easy to solve?
 - it has several variance
- Under what conditions on its to ensure the solution?
 - Convex Programming Problem
 - Convex Quadratic Programming Problem
- How to fix the Big Data problem?
 - Dual programming problem
- How to solve the NonLinearity problem ?
 - kernel function
- What are methods kind used for their solution?
- How to solve them if their constraints are satisfied with a probability?
- How to solve them if their Labels are Random Variables ?

Motivation: NonLinearity problem

Soft margin SVC($C - SVC$)

$$C - SVC \left\{ \begin{array}{l} \text{Optimize} \\ \text{s.t} \end{array} \right. \quad \begin{array}{l} f(w, \xi) = (f_1(w), f_2(\xi)) \\ y_i(w^T \varphi(x_i) + b) \geq 1 - \xi_i, i = 1, \dots, n \\ \xi_i \geq 0, w \in \mathbb{R}^d, b \in \mathbb{R} \end{array}$$

Linear Soft ε -band SVR

$$\text{LS} - \varepsilon\text{-band} - \text{SVR:} \left\{ \begin{array}{l} \text{Optimize} \quad f(w, \xi) = (f_1(w), f_2(\xi^1), f_3(\xi^2)) \\ \text{s.t.} \quad y_i - (w^T \varphi(x_i) - b) \leq \varepsilon + \xi_i^1, \quad i = 1, \dots, n \\ \quad \quad w^T \varphi(x_i) + b - y_i \leq \varepsilon + \xi_i^2, \quad i = 1, \dots, n \\ \quad \quad \xi_i^1 \geq 0, \xi_i^2 \geq 0, (w, b) \in \mathbb{R}^d \times \mathbb{R} \end{array} \right.$$

Motivation: Probabilistic Approach

Soft margin SVC($C - SVC$)

$$C - SVC \left\{ \begin{array}{l} \text{Min} \\ \text{s.t} \end{array} \right. \begin{array}{l} f(w, \xi) = (f_1(w), f_2(\xi)) \\ P(y_i(w^T x_i + b) \geq 1 - \xi_i) \geq p_i, \quad i = 1, \dots, n \\ \xi_i \geq 0, w \in \mathbb{R}^d, b \in \mathbb{R} \end{array}$$

Motivation:

Linear Soft ε -band SVR

$$\text{LS} - \varepsilon\text{-band} - \text{SVR:} \left\{ \begin{array}{l} \text{min} \\ \text{s.à} \end{array} \right. \begin{array}{l} f(w, \xi) = (f_1(w), f_2(\xi^1), f_3(\xi^2)) \\ P(y_i - (w^T x_i - b) \leq \varepsilon + \xi_i^1) \geq p_i^1, \quad i = 1, \dots, n \\ P(w^T x_i + b - y_i \leq \varepsilon + \xi_i^2) \geq p_i^2, \quad i = 1, \dots, n \\ \xi_i^1 \geq 0, \xi_i^2 \geq 0, (w, b) \in \mathbb{R}^d \times \mathbb{R} \end{array}$$

Motivation: Random Variable $Y_i \sim D_i$

Soft margin SVC($C - SVC$)

$$C - SVC \begin{cases} \text{Min} & f(w, \xi) = (f_1(w), f_2(\xi)) \\ \text{s.t} & Y_i(w^T x_i + b) \geq 1 - \xi_i, i = 1, \dots, n \\ & \xi_i \geq 0, w \in \mathbb{R}^d, b \in \mathbb{R} \end{cases}$$

Linear Soft ε -band SVR

$$\text{LS} - \varepsilon\text{-band} - \text{SVR}: \begin{cases} \min & f(w, \xi) = (f_1(w), f_2(\xi^1), f_3(\xi^2)) \\ \text{s.t} & Y_i - (w^T x_i - b) \leq \varepsilon + \xi_i^1, \quad i = 1, \dots, n \\ & w^T x_i + b - Y_i \leq \varepsilon + \xi_i^2, \quad i = 1, \dots, n \\ & \xi_i^1 \geq 0, \xi_i^2 \geq 0, (w, b) \in \mathbb{R}^d \times \mathbb{R} \end{cases}$$

Plan

- Part I: **Support Vector machine(50%)**
 - Duality Theory in the Convex Programming Problem
 - Linear Classification
 - Linear Regression
 - Theory of kernel function
 - NonLinear Classification
 - NonLinear Regression
 - Résolution Method
- Part II: **Support Vector machine with Random variables (50%)**
 - Robust Support Vector Classification
 - Robust Support Vector Regression
 - Probabilistic Constraints Support Vector Classification
 - Least Squares Probabilistic Support Vector Classification
 - Probabilistic Constraints Support Vector Regression