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svm

using slack variable

$$\xi_i = 1 - y_i (w^T x_i + b)$$

$$\text{on } \min_{w, b} \sum_{i=1}^m \max(0, 1 - y_i (w^T x_i + b)) + \lambda \|w\|^2$$

SolnWe begin by taking the minimization function \mathcal{H} as

$$\Rightarrow \frac{1}{2} \|w\|^2 + C \sum_{i=1}^n \xi_i \quad \mathcal{H}_i, \quad i=1, \dots, n$$

which is subject to

SVM

using slack variable

$$\xi_i = 1 - y_i (w^T x_i - b)$$

$$\text{on } \min_{w, b} \sum_{i=1}^m \max(0, 1 - y_i (w^T x_i - b)) + \lambda \|w\|^2$$

Soln

We begin by taking the minimization function \mathcal{H} as

$$\Rightarrow \frac{1}{2} \|w\|^2 + C \sum_{i=1}^n \xi_i \quad \forall_i, i=1, \dots, n$$

which is subject to

$$\Rightarrow \xi_i = 1 - y_i (w^T x_i - b) \quad \forall_i, i=1, \dots, n$$

Thus, for $\xi_i \geq 0$, we have the above written as

$$1 - \xi_i \leq y_i (w^T x_i - b) \quad \forall_i, i=1, \dots, n$$

We begin by taking the minimization function V_i as

$$\Rightarrow \frac{1}{2} \|w\|^2 + C \sum_{i=1}^n \xi_i \quad V_i, \quad i=1, \dots, n$$

which is subject to

$$\Rightarrow \xi_i = 1 - y_i (w^T x_i - b) \quad V_i, \quad i=1, \dots, n$$

\Rightarrow Thus, for $\xi_i \geq 0$, we have the above written as

$$1 - \xi_i \leq y_i (w^T x_i - b) \quad V_i, \quad i=1, \dots, n$$

\Rightarrow In this case the opt variable are w, b, ξ_i

Next we look at all possible deviations for
Change in $w^T x$ and b as follows

$$w^T x + b = 1$$

$$w^T x + b = 0$$

$$\Rightarrow \xi_i = 1 - y_i (\omega^T x_i - b) \quad \forall i = 1, \dots, n$$

\Rightarrow Thus for $\xi_i \geq 0$, we have the above written as

$$1 - \xi_i \leq y_i (\omega^T x_i - b) \quad \forall i = 1, \dots, n$$

\Rightarrow In this case the opt variable are ω, b, ξ_i

\Rightarrow Next we look at all possible deviations for change in $\omega^T x$ and b as follows

$$\omega^T x + b = 1$$

$$\omega^T x + b = 0$$

$$\omega^T x + b = -1$$

\Rightarrow We let $\omega^T x + b = 0$, then the optimization problem will now become

$$\min_{\omega, b, \xi_i} \frac{1}{2} \|\omega\|^2 + C \sum_{i=1}^n \xi_i$$

$$\Rightarrow \xi_i = 1 - y_i (w^T x_i - b) \quad \forall i = 1, \dots, n$$

\Rightarrow Thus for $\xi_i \geq 0$, we have the above written as

$$1 - \xi_i \leq y_i (w^T x_i - b) \quad \forall i = 1, \dots, n$$

\Rightarrow In this case the opt variable are w, b, ξ_i

\Rightarrow Next we look at all possible deviations for change in $w^T x$ and b as follows

$$w^T x + b = 1$$

$$w^T x + b = 0$$

$$w^T x + b = -1$$

\Rightarrow We let $w^T x + b = 0$, then the optimization problem will now become

$$\min_{w, b, \xi_i} \frac{1}{2} \|w\|^2 + C \sum_{i=1}^n \xi_i$$

which is subject to

$$1 - y_i (\omega^T x_i - b) - \xi_i \leq 0 \quad \forall i = 1, \dots, n$$

with $-\xi_i \leq 0 \quad \forall i$

\Rightarrow Next we compute the associated Lagrangian L in relation to the optimization problem as follows

$$L(\alpha, \lambda, \xi, \omega, b) = \frac{1}{2} \|\omega\|^2 + C \sum_{i=1}^n \xi_i + \sum_{i=1}^n \mu_i (1 - \xi_i - y_i (\omega^T x_i - b)) - \sum_{i=1}^n \lambda_i \xi_i$$

where

μ_i are the Lagrange multipliers for the separability constraint

=> Next we compute the associated Lagrangian L in relation to the optimization problem as follows

$$L(\alpha, \lambda, \xi, w, b) = \frac{1}{2} \|w\|^2 + C \sum_{i=1}^n \xi_i + \sum_{i=1}^n \alpha_i (1 - \xi_i - y_i (w^T x_i - b)) - \sum_{i=1}^n \lambda_i \xi_i$$

where

=> α_i are the Lagrange multipliers for the separability constraint

=> λ_i are the Lagrange multipliers for the constraints $-\xi_i \leq 0$

=> We can now take the derivative of the above Lagrangian function w.r.t to both ξ_i and b

$$\Rightarrow \frac{\partial L}{\partial b} = 0 \Rightarrow \sum_{i=1}^n \alpha_i y_i = 0$$

$$- \sum_{i=1}^n \lambda_i \xi_i$$

where

$\Rightarrow \alpha_i$ over the language multipliers for the separability constraint

$\Rightarrow \lambda_i$ over the language multipliers for the constraints $-\xi_i \leq 0$

\Rightarrow We can now take the derivative of the above Lagrangian function w.r.t to both ξ_i and b

$$\Rightarrow \frac{\partial L}{\partial b} = 0 \Rightarrow \sum_{i=1}^n \alpha_i y_i = 0$$

$$\Rightarrow \frac{\partial L}{\partial \xi_i} = 0 \Rightarrow \alpha_i \lambda_i = 0 \quad \forall i = 1, \dots, n$$

note that $-\xi_i - y_i (w^T x_i - b) + 1 \leq 0$; $\xi_i \geq 0 \quad \forall i$

\Rightarrow Using the complementary slackness in interval

$\Rightarrow \alpha_i$ are the Lagrange multipliers for the constraints

$\Rightarrow d_i$ are the Lagrange multipliers for the constraints

$$-\xi_i \leq 0$$

\Rightarrow We can now take the derivative of the above Lagrangian function w.r.t to both ξ_i and b

$$\Rightarrow \frac{\partial L}{\partial b} = 0 \Rightarrow \sum_{i=1}^n \alpha_i y_i = 0$$

$$\Rightarrow \frac{\partial L}{\partial \xi_i} = 0 \Rightarrow \alpha_i x_i = C \quad \forall i = 1, \dots, n$$

Note that $-\xi_i - y_i (w^T x_i - b) + 1 \leq 0$; $\xi_i \geq 0 \quad \forall i$

\Rightarrow Using the complementary slackness condition in interval $0 \leq \mu_i \leq C$

i.e

$1 - y_i (w^T x_i - b) = 0$, we need to find

the dual function for all u_i which is given by as

$$Q(u, \lambda) = \inf_{w, b, \xi_i} L(u, \lambda, \xi, w, b)$$

But from the Lagrangian function we have the term

$$\sum_i (C - u_i - \lambda_i) \xi_i$$

Taking the infimum w.r.t ξ_i and imposing the condition $C = u_i + \lambda_i \quad \forall_i$ on the m

the dual function for all μ_i which is given as

$$Q(\mu, \lambda) = \inf_{\omega, b, \xi_i} L(\mu, \lambda, \xi, \omega, b)$$

\Rightarrow But from the Lagrangian function we have the form

$$\sum_i (C - \mu_i - \lambda_i) \xi_i$$

\Rightarrow Taking the infimum w.r.t ξ_i and imposing the condition $C = \mu_i + \lambda_i \quad \forall_i$ on the interval

$$0 \leq \mu_i \leq C$$

the dual problem becomes,

$$\max_{\mu} Q(\mu) = \sum_{i=1}^n \mu_i - \frac{1}{2} \sum_{i,j=1}^n \mu_i \mu_j y_i y_j x_i^T x_j$$

$$\text{subject to } 0 \leq \mu_i \leq C \quad \forall_i \quad i=1, \dots, n$$

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$$\sum_i (C - \mu_i - \lambda_i) \xi_i$$

\Rightarrow Taking the infimum w.r.t ξ_i and imposing the condition $C = \mu_i + \lambda_i \quad \forall_i$ on the interval $0 \leq \mu_i \leq C$ the dual problem becomes

$$\max_{\mu} \quad \mathcal{L}(\mu) = \sum_{i=1}^n \mu_i - \frac{1}{2} \sum_{i,j=1}^n \mu_i \mu_j y_i y_j x_i^T x_j$$

Subject to $0 \leq \mu_i \leq C \quad \forall_i \quad i = 1, \dots, n$
and $\sum_{i=1}^n \mu_i y_i = 0$

\Rightarrow Associated primal problem is given by

$$\text{minimize} \quad \frac{1}{2} \|w\|^2 + C \sum_{i=1}^n \xi_i$$

Subject to

\Rightarrow Taking the infimum w.r.t q_i and imposing the condition $C = \mu_i + \lambda_i \quad \forall_i$ on the interval $0 \leq \mu_i \leq C$ the dual problem becomes,

$$\max_{\mu} \quad \mathcal{L}(\mu) = \sum_{i=1}^n \mu_i - \frac{1}{2} \sum_{i,j=1}^n \mu_i \mu_j y_i y_j x_i^T x_j$$

Subject to $0 \leq \mu_i \leq C \quad \forall_i \quad i=1, \dots, n$
 and $\sum_{i=1}^n \mu_i y_i = 0$

\Rightarrow Associated primal problem is given by

$$\text{minimize} \quad \frac{1}{2} \|w\|^2 + C \sum_{i=1}^n \xi_i$$

Subject to

$$1 - \xi_i \leq y_i (w^T x_i - b) \quad \forall_i \quad i=1, \dots, n$$