MATH3431 Machine Learning and Neural Networks III

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Homework 3: SVM

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Instructions: For Formative assessment, submit the solutions to all the parts of the Exercise.

Exercise 1. $(\star\star)$ Consider a training data set $\mathcal{D} = \{z_i = (x_i, y_i)\}_{i=1}^m$. Consider the Soft-SVM Algorithm that requires the solution of the following quadratic minimization problem (in a slightly modified but equivalent form to what we have discussed)

Primal problem

$$(w^*, b^*, \xi^*) = \underset{(w,b,\xi)}{\arg\min} \left(\|w\|_2^2 + C \sum_{i=1}^m \xi_i \right)$$
 (1)

subject to:
$$y_i(\langle w, x_i \rangle + b) \ge 1 - \xi_i, \ \forall i = 1, ..., m$$
 (2)

$$\xi_i \ge 0, \ \forall i = 1, ..., m \tag{3}$$

for some user-specified fixed parameter C > 0. We seek to find the dual problem of 1-2.

- 1. Specify the Lagrangian function L associated to the above primal quadratic minimization problem, where $\{\alpha_i\}$ are the Lagrange coefficients wrt (2), and $\{\beta_i\}$ are the Lagrange coefficients wrt (3). Write down any possible restrictions on the Lagrange coefficients.
- 2. Compute the dual Lagrangian function denoted as \tilde{L} as a function of the Lagrange coefficients and the data points \mathcal{D} .
- 3. Apply the KKT conditions to the above problem, and write them down.
- 4. Derive and write down the dual Lagrangian quadratic maximization problem, along with the inequality and equality constraints, where you seek to find $\{\alpha_i\}$.
- 5. Justify why the *i*-th point x_i lies on the margin boundary when $\alpha_i \in (0, C)$ (beware it is $\alpha_i \neq C$), and why the *i*-th point x_i lies inside the margin when $\alpha_i = C$.

6. Given optimal values $\{\alpha_i^*\}$ for Lagrangian coefficients $\{\alpha_i\}$ as they are derived by solving the dual Lagrangian maximization problem in part 4, derive the optimal values w^* and b^* for the parameters w and b as function of the support vectors. Regarding parameter b it should be in the derived in the form

$$b^* = \frac{1}{|\mathcal{M}|} \sum_{i \in \mathcal{M}} \left(y_i - \sum_{j \in \mathcal{S}} \alpha_j^* y_j \langle x_j, x_i \rangle \right)$$

where you determine the sets \mathcal{M} and \mathcal{S} .

7. Report the halfspace predictive rule $h_{w,b}\left(x\right)$ of the above problem as a function of w^{*} and b^{*} .