MATH3431 Machine Learning and Neural Networks III

Epiphany term 2023

## Homework 3: SVM

Lecturer: Georgios P. Karagiannis georgios.karagiannis@durham.ac.uk

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  $\alpha_i \in (0, C)$  but  $\alpha_i \neq C$ , and why the *i*-th point  $x_i$  lies inside the margin when  $\alpha_i \neq C$ .

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  $b^{*}$ .