

## Exercise Sheet 9

### Exercise 1: Dual of One-Class SVM (60 P)

The spherical version of the one-class SVM (also called “support vector data description”) is given by the minimization problem:

$$\min_{R, c, (\xi_i)_{i=1}^n} R^2 + \frac{1}{n\nu} \sum_{i=1}^n \xi_i$$

subject to

$$\forall_{i=1}^n : \|\phi(x_i) - c\|^2 \leq R^2 + \xi_i \quad \text{and} \quad \xi_i \geq 0$$

where  $x_1, \dots, x_n$  are the training data and  $\phi(x_i) \in \mathbb{R}^d$  is a feature space representation.

(a) *Derive* the dual program for the one-class SVM.

(b) *Show* that the kernelized dual has the form

$$\begin{aligned} & \max_{\alpha} \quad \sum_{i=1}^n \alpha_i k(x_i, x_i) - \sum_{i=1}^n \sum_{j=1}^n \alpha_i \alpha_j k(x_i, x_j) \\ & \text{subject to} \quad \sum_{i=1}^n \alpha_i = 1 \quad \text{and} \quad \forall_{i=1}^n : 0 \leq \alpha_i \leq \frac{1}{n\nu} \\ & \text{and with center} \quad c = \sum_{i=1}^n \alpha_i \phi(x_i) \end{aligned}$$

where  $k$  is the kernel associated to the feature map  $\phi$ .

### Exercise 2: Quadratic Programming (40 P)

*Show* that the dual program derived in Exercise 1 is a linearly constrained quadratic program, by writing it in the matrix form

$$\begin{aligned} & \min_{\alpha} \quad \alpha^\top P \alpha + q^\top \alpha \\ & \text{subject to} \quad G\alpha \leq h \quad \text{and} \quad A\alpha = b \end{aligned}$$

with matrices  $P, G, A$  and vectors  $q, h, b$ . That is, *express* the matrices  $P, G, A$  and vectors  $q, h, b$  in terms of the solution of Exercise 1, and *specify* their dimensions.