Exercise Sheet 11

Machine Learning 2, SS16

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Exercise 1

The Lagrangian of the given primal problem is:

$$\mathcal{L}(R, c, \xi, \alpha, \lambda) = R^2 + \frac{1}{n\nu} \sum_{i} \xi_i - \sum_{i} \alpha_i (R^2 + \xi_i - \|\phi(x_i) - c\|) - \sum_{i} \lambda_i \xi_i$$

$$= R^2 + \frac{1}{n\nu} \sum_{i} \xi_i - \sum_{i} \alpha_i (R^2 + \xi_i - (\phi(x_i)^\top \phi(x_i) + c^\top c - 2c^\top \phi(x_i))) - \sum_{i} \lambda_i \xi_i$$

$$= R^2 + \frac{1}{n\nu} \sum_{i} \xi_i - \sum_{i} \alpha_i (R^2 + \xi_i - \phi(x_i)^\top \phi(x_i) - c^\top c + 2c^\top \phi(x_i)) - \sum_{i} \lambda_i \xi_i$$

We now differentiate w.r.t. primal variables R, c, ξ :

$$\frac{\partial}{\partial R} \mathcal{L}(R, c, \xi, \alpha, \lambda) = 2R - 2R \sum_{i} \alpha_{i} \stackrel{!}{=} 0 \qquad \Longrightarrow \qquad \sum_{i} \alpha_{i} = 1$$

$$\frac{\partial}{\partial c} \mathcal{L}(R, c, \xi, \alpha, \lambda) = 2c \sum_{i} \alpha_{i} - 2 \sum_{i} \alpha_{i} \phi(x_{i}) \stackrel{!}{=} 0 \qquad \Longrightarrow \qquad c = \sum_{i} \alpha_{i} \phi(x_{i})$$

$$\frac{\partial}{\partial \xi_{i}} \mathcal{L}(R, c, \xi, \alpha, \lambda) = \frac{1}{n\nu} - \alpha_{i} - \lambda_{i} \stackrel{!}{=} 0 \qquad \Longrightarrow \qquad \frac{1}{n\nu} = \alpha_{i} + \lambda_{i}$$

The Lagrangian of the dual problem then can be obtained by plugging in the derived results:

$$\mathcal{L}(\alpha, \lambda) = R^{2} + \frac{1}{n\nu} \sum_{i} \xi_{i} - \sum_{i} \alpha_{i} (R^{2} + \xi_{i} - \phi(x_{i})^{\top} \phi(x_{i}) - c^{\top} c + 2c^{\top} \phi(x_{i})) - \sum_{i} \lambda_{i} \xi_{i}$$

$$= R^{2} + \frac{1}{n\nu} \sum_{i} \xi_{i} - R^{2} \sum_{i} \alpha_{i} - \sum_{i} \alpha_{i} \xi_{i} + \sum_{i} \alpha_{i} k(x_{i}, x_{i}) + c^{\top} c \sum_{i} \alpha_{i} - 2c^{\top} \sum_{i} \alpha_{i} \phi(x_{i}) - \sum_{i} \lambda_{i} \xi_{i}$$

$$= \frac{1}{n\nu} \sum_{i} \xi_{i} - \sum_{i} (\alpha_{i} + \lambda_{i}) \xi_{i} + \sum_{i} \alpha_{i} k(x_{i}, x_{i}) + c^{\top} c - 2c^{\top} \sum_{i} \alpha_{i} \phi(x_{i})$$

$$= \sum_{i} \alpha_{i} k(x_{i}, x_{i}) - c^{\top} c$$

Lastly, by using the definition of c, we obtain the dual program:

$$\max_{\alpha} \sum_{i} \alpha_{i} k(x_{i}, x_{i}) - \sum_{i} \sum_{j} \alpha_{i} \alpha_{j} k(x_{i}, x_{j})$$
s.t.
$$\sum_{i} \alpha_{i} = 1 \text{ and } \alpha_{i} \geq 0 \text{ and } \lambda_{i} \geq 0 \text{ and } \alpha_{i} + \lambda_{i} = \frac{1}{n\nu}$$

$$\implies \max_{\alpha} \sum_{i} \alpha_{i} k(x_{i}, x_{i}) - \sum_{i} \sum_{j} \alpha_{i} \alpha_{j} k(x_{i}, x_{j}) \quad \text{s.t.} \quad \sum_{i} \alpha_{i} = 1 \text{ and } \frac{1}{n\nu} \geq \alpha_{i} \geq 0$$

The primal variable c is determined by $c = \sum_i \alpha_i \phi(x_i)$. R, the radius, then can be found by using the constraint of the primal problem (i.e. using the support vectors on the gutter to solve $\|\phi(x_i) - c\| = R$).

Exercise 2

Form of the QP problem:

$$\min_{\alpha} \alpha^{\top} P \alpha + q^{\top} \alpha$$
 s.t. $G \alpha \leq h$ and $A \alpha = b$

The objective of the dual problem can be further derived like this:

$$\max_{\alpha} \mathcal{L}(\alpha) = \max_{\alpha} \sum_{i} \alpha_{i} k(x_{i}, x_{i}) - \sum_{i} \sum_{j} \alpha_{i} \alpha_{j} k(x_{i}, x_{j}) \quad \text{s.t.} \quad \sum_{i} \alpha_{i} = 1 \quad \text{and} \quad \frac{1}{n\nu} \geq \alpha_{i} \geq 0$$

$$\implies \min_{\alpha} -\mathcal{L}(\alpha) = \min_{\alpha} \sum_{i} \sum_{j} \alpha_{i} \alpha_{j} k(x_{i}, x_{j}) - \sum_{i} \alpha_{i} k(x_{i}, x_{i}) \quad \text{s.t.} \quad \sum_{i} \alpha_{i} = 1 \quad \text{and} \quad \frac{1}{n\nu} \geq \alpha_{i} \geq 0$$

The variables can now be identified as follows:

$$P = (k(x_i, x_j))_{i,j} \in \mathbb{R}^{n \times n} \qquad q = -(k(x_i, x_i))_i \in \mathbb{R}^n \qquad G = \begin{bmatrix} \mathbf{I}_n \\ -\mathbf{I}_n \end{bmatrix} \in \mathbb{R}^{2n \times n}$$

$$h = \frac{1}{n\nu} \begin{bmatrix} \vec{\mathbb{I}}_n \\ \vec{0}_n \end{bmatrix} \in \mathbb{R}^{2n} \qquad A = \begin{bmatrix} 1 & \dots & 1 \\ \phi(x_1) & \dots & \phi(x_n) \end{bmatrix} \in \mathbb{R}^{(d+1) \times n} \qquad b = \begin{bmatrix} 1 \\ \vec{c} \end{bmatrix} \in \mathbb{R}^{d+1}$$