

## 58147 Machine Learning (Spring 2005)

### Exercise 9 (Wednesday 6 April)

- (a) Complete the proof of Theorem 4.7 by showing that if  $k_1$  and  $k_2$  are valid kernels, then so are  $k_1 + k_2$  and  $ak_1$  for  $a > 0$ .  
(b) Assume that  $k$  is a valid kernel with feature map  $\psi$ . Obtain  $\tilde{k}$  by *normalising*:

$$\tilde{k}(x, z) = \frac{k(x, z)}{(k(x, x)k(z, z))^{1/2}}.$$

Show that  $\tilde{k}$  is a kernel for the feature map given by  $\tilde{\psi}(x) = \psi(x) / \|\psi(x)\|$ .

- Prove Corollary 4.8.

*Hints:* For part 1, use induction. For part 2, notice that  $\exp(r)$  is a limit of polynomials  $p_n(r)$ . For part 3, normalise the kernel  $\exp(\langle x, z \rangle / \sigma^2)$ .

- Define  $f(x) = x^2$  and  $g(x) = (x - 2)^2$ . Consider the problem

$$\text{minimise } f(x) \quad \text{subject to } g(x) \leq 1.$$

Write out the Lagrangian  $L(x, \lambda)$ . Solve the problem using the KKT conditions. Write out also the dual  $g(\lambda)$  in a closed form.

Use some mathematical software (*e.g.*, Maple) to plot  $L(x, \lambda)$  (as a function of two variables, so you get a three-dimensional picture). Find from this plot the functions  $x \mapsto \max_{\lambda} L(x, \lambda)$  and  $\lambda \mapsto \min_x L(x, \lambda)$ . Notice the point where they coincide.

Repeat the exercise with the constraint  $g(x) \leq 8$ .

- In an online prediction setting, given an example  $(\mathbf{x}_t, y_t) \in \mathbf{R}^n \times \mathbf{R}$ , let  $W_t = \{\mathbf{w} \in \mathbf{R}^n \mid (\mathbf{w} \cdot \mathbf{x}_t - y_t)^2 \leq R^2\}$  where  $R$  is some constant. Then define the update as

$$\mathbf{w}_{t+1} = \arg \min_{\mathbf{w} \in W_t} \|\mathbf{w}_t - \mathbf{w}\|_2^2.$$

Write out the update in a closed form (*i.e.*, solve the minimisation).

- Recall the relative entropy

$$d_{\text{re}}(\mathbf{p}, \mathbf{q}) = \sum_{i=1}^n p_i \ln \frac{p_i}{q_i}$$

where we assume  $p_i, q_i \geq 0$  and  $\sum_{i=1}^n p_i = \sum_{i=1}^n q_i = 1$ . Fix  $\mathbf{q}$  such that  $q_i > 0$  for all  $i$ . Using the KKT conditions, show that subject to the constraints  $p_i \geq 0$ ,  $i = 1, \dots, n$ , and  $\sum_{i=1}^n p_i = 1$ , the relative entropy  $d_{\text{re}}(\mathbf{p}, \mathbf{q})$  is minimised at  $\mathbf{p} = \mathbf{q}$ . You may take it as known that  $d_{\text{re}}(\mathbf{p}, \mathbf{q})$  is convex in  $\mathbf{p}$  (and also  $\mathbf{q}$ , although that is not needed here.)

What is the dual?