

# CS 4758/6758: Robot Learning: Homework 5

Due: May 3, 2012 (In class)

## 1 Support Vector Machines(40 pts.)

### 1.1 RBF Kernel (20 pts)

Which of the following are valid kernels? If yes, use the given properties to prove that they are kernels.

- (a)  $K(x, z) = xz \exp(x + z)$
- (b)  $K(x, z) = \cos^2(x - z)$
- (c)  $K(x, z) = (x - z)^2$
- (d)  $K(x, z) = \exp(-\|x - y\|_2^2)$

**Properties that can be used:** Let  $K_1(\mathbf{x}, \mathbf{z})$  and  $K_2(\mathbf{x}, \mathbf{z})$  be kernels. Then all of the following are also kernels:

- (a)  $aK_1 + bK_2, \quad \forall a, b \in \mathbb{R}, a > 0, b > 0$
- (b)  $f(\mathbf{x})^T f(\mathbf{z}), \quad f : \mathbb{R}^n \rightarrow \mathbb{R}^m$
- (c)  $K_1 K_2$
- (d)  $e^{K_1(\mathbf{x}, \mathbf{z})}$
- (e)  $K_1(x, z) = x^T z$  (linear kernel).

### 1.2 Kernels and SVMs (20 pts)

A robot has collected data with its sensor. We want to use the data to build a classifier. In order to make our classifier robust, we would like to find the max-margin hyperplane (the hyperplane that has the largest margin). Currently, the feature space is a one dimensional space  $X \in \mathbb{R}$ . The desired classification output is  $Y = \{+, -\}$ , as shown in Figure 1. The training set contains three positive examples,  $x_1 = 0$ ,  $x_3 = 3$ , and  $x_4 = 4$ , and one negative example  $x_2 = 1$ ,

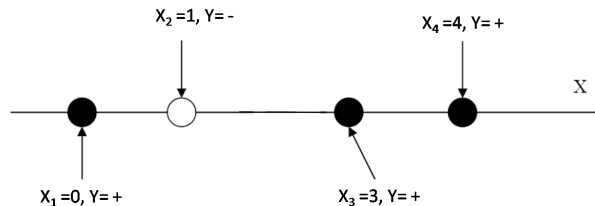


Figure 1: Training set of four examples with their desired classifications

Currently, the data points are unseparable, and so therefore we must define a transformation (i.e. a feature mapping function) that maps the data into a projected space in  $\mathbb{R}^2$ . Consider the mapping  $\Phi(X) = (X, (X - 1)^2)$ .

- (a) First, draw the data points after the transformation to the 2D domain. Draw the separation plane and indicate the margin. Also write which examples (out of  $x_1, x_2, x_3, x_4$ ) are support vectors.
- (b) If we get one more datapoint  $x_5 = 2$ , would it affect the margin? Justify. (Don't draw for this.)

## 2 Particle Filters (10 pts.)

Consider the following proposition:

*There is no point in using a particle filter if the number of distinct states at any time is no more than the number of particles one plans to allocate to the particle filter.*

Justify or refute this statement.

## 3 Particle Filters (50 pts.)

In this question, you will implement a particle filter for the non-linear system defined over three state variables, and given by a deterministic state transition:

$$\begin{pmatrix} x' \\ y' \\ \theta' \end{pmatrix} = \begin{pmatrix} x + \cos \theta \\ y + \sin \theta \\ \theta \end{pmatrix}$$

The initial state estimate was:

$$\mu = \begin{pmatrix} 0 & 0 & 0 \\ 0.01 & 0 & 0 \\ 0 & 0.01 & 0 \\ 0 & 0 & 10000 \end{pmatrix}$$

**A.** Give a suitable initial estimate for the particle prior, which reflects the state of knowledge in the gaussian prior.

**B.** Implement a particle filter and run its prediction step. Compare the resulting prior with the one from your intuitive analysis. What can be said about the resolution of the  $x - y$  co-ordinates and the orientation  $\theta$  in your particle filter?

**C.** Now let us add a measurement to our estimate. The measurement is a noisy projection of the  $x$ -coordinate of the robot, with covariance  $Q=0.01$ . Implement the step, compute the result and plot it. Compare with the result of your intuitive drawing.