

Advanced Machine Learning

Problem Sheet 1 due Friday, Oct-30 2015

Problem 1 (Kernel Trick for polynomial SVM) Consider the polynomial Kernel $K(\mathbf{u}, \mathbf{v}) = (1 + \mathbf{u} \cdot \mathbf{v})^2$ where $\mathbf{u}, \mathbf{v} \in [-1, 1]^d$.

- show that for $d=2$ that $K(\mathbf{u}, \mathbf{v})$ can be written as a scalar product $\phi(\mathbf{u}) \cdot \phi(\mathbf{v})$ in a *six-dimensional state space* and provide explicit expressions for the feature transform $\phi(\mathbf{u})$.
- generalize your result for general (integer) $d > 1$.
- demonstrate for this case explicitly the validity of Mercer's condition, if the integration range is chosen as the hypercube $[-1, 1]^d$.
- use your results to explain why the "Kernel Machine"

$$y(\mathbf{x}) = \sum_{l=1}^L c_l K(\mathbf{x}_l, \mathbf{x})$$

with $\mathbf{x}, \mathbf{x}_1, \dots, \mathbf{x}_L \in [-1, 1]^d$ is equivalent to a perceptron

$$y(\mathbf{x}) = \sum_{k=1}^D w_k \phi_k(\mathbf{x})$$

with D linearly independent features $\phi_k(\mathbf{x}), k = 1..D$, and specify D

Problem 2 (Linear filter features for time series prediction) Consider the learning a predictor for the simple time sequence

$$x_t = \sin(\omega \cdot t).$$

- show that for a linear predictor that uses only the previous time step value as its only feature, i.e.,

$$x_t = a \cdot x_{t-1} + b$$

no accurate prediction becomes possible (Hint: consider how much b has to minimally vary about a constant value for a given value of a and show from this that no choice of a can make this variation zero).

- show that using the values x_{t-1} and x_{t-2} of *two* consecutive previous time steps as features allows to find a linear predictor

$$x_t = a_0 + a_1 \cdot x_{t-1} + a_2 \cdot x_{t-2}$$

that predicts the time sequence *perfectly*. What are the values for the coefficients? (Hint: express x_{t-1} and x_{t-2} in terms of $x_t = \sin(\omega \cdot t)$ and $\cos(\omega \cdot t)$ and solve for x_t).