# DSC 140A - Discussion 03

## Problem 1.

Recall that the regularized least squares risk is

$$\tilde{R}(\vec{w}) = \frac{1}{n} \sum_{i=1}^{n} (\vec{w} \cdot \vec{\phi}(\vec{x}^{(i)}) - y_i)^2 + \lambda \|\vec{w}\|^2$$

Show that

$$\tilde{R}(\vec{w}) = \frac{1}{n} \left( \vec{w}^T \Phi^T \Phi \vec{w} - 2\vec{w}^T \Phi^T \vec{y} + \vec{y}^T \vec{y} \right) + \lambda \vec{w}^T \vec{w},$$

where  $\Phi$  is the matrix whose *i*th row is  $\vec{\phi}(\vec{x}^{(i)})$ , and where  $\vec{y} = (y_1, \dots, y_n)^T$ .

# Solution:

$$\begin{split} \tilde{R}(\vec{w}) &= \frac{1}{n} \sum_{i=1}^{n} (\vec{w} \cdot \vec{\phi}(\vec{x}^{(i)}) - y_i)^2 + \lambda \|\vec{w}\|^2 \\ &= \frac{1}{n} \sum_{i=1}^{n} ((\vec{w} \cdot \vec{\phi}(\vec{x}^{(i)}))^2 - 2y_i \vec{w} \cdot \vec{\phi}(\vec{x}^{(i)}) + y_i^2) + \lambda \|\vec{w}\|^2 \\ &= \frac{1}{n} (\sum_{i=1}^{n} (\vec{w} \cdot \vec{\phi}(\vec{x}^{(i)}))^2 - 2\sum_{i=1}^{n} y_i \vec{w} \cdot \vec{\phi}(\vec{x}^{(i)}) + \sum_{i=1}^{n} y_i^2) + \lambda \vec{w}^T \vec{w} \\ &= \frac{1}{n} \left( \vec{w}^T \Phi^T \Phi \vec{w} - 2 \vec{w}^T \Phi^T \vec{y} + \vec{y}^T \vec{y} \right) + \lambda \vec{w}^T \vec{w} \end{split}$$

## Problem 2.

In lecture, we defined a kernel function to be a function k which computes the dot product of vectors after they are mapped to some high-dimensional space. The useful thing about kernel functions is that they allow us to compute these dot products without actually mapping vectors them to the high-dimensional space, which can be costly. In this problem, we will consider the the 2nd-order *polynomial kernel*, defined to be

$$k(\vec{x}, \vec{x}') = (1 + \vec{x} \cdot \vec{x}')^2.$$

Let  $\vec{\phi}(\vec{x}) : \mathbb{R}^3 \to \mathbb{R}^{10}$  be the mapping:

$$\vec{\phi}(\vec{x}) = (1, x_1^2, x_2^2, x_3^2, \sqrt{2} x_1, \sqrt{2} x_2, \sqrt{2} x_3, \sqrt{2} x_1 x_2, \sqrt{2} x_1 x_3, \sqrt{2} x_2 x_3)^T,$$

where  $x_1, x_2, x_3$  are the components of the input vector,  $\vec{x}$ . That is,  $\vec{\phi}$  is a feature map which maps a vector into a higher-dimensional space.

Show that  $k(\vec{x}, \vec{y}) = \vec{\phi}(\vec{x}) \cdot \vec{\phi}(\vec{y})$ . That is, that k indeed computes the inner product of vectors in the higher-dimensional space.

#### Solution:

$$k(\vec{x},\vec{y}) = (1 + x_1y_1 + x_2y_2 + x_3y_3)^2$$
  
=  $1 + x_1^2y_1^2 + x_2^2y_2^2 + x_3^2y_3^2 + 2x_1y_1 + 2x_2y_2 + 2x_3y_3 + 2x_1x_2y_2y_1 + 2x_1x_3y_3y_1 + 2x_2x_3y_2y_3$ 

Now,  $\vec{\phi}(\vec{x}) \cdot \vec{\phi}(\vec{y}) = (1, x_1^2, x_2^2, x_3^2, \sqrt{2} x_1, \sqrt{2} x_2, \sqrt{2} x_3, \sqrt{2} x_1 x_2, \sqrt{2} x_1 x_3, \sqrt{2} x_2 x_3) \begin{pmatrix} 1 \\ y_1^2 \\ y_2^2 \\ y_3^2 \\ \sqrt{2} y_1 \\ \sqrt{2} y_2 \\ \sqrt{2} y_3 \\ \sqrt{2} y_1 y_2 \\ \sqrt{2} y_1 y_3 \\ \sqrt{2} y_2 y_3 \end{pmatrix}$   $= 1 + x_1^2 y_1^2 + x_2^2 y_2^2 + x_3^2 y_3^2 + 2x_1 y_1 + 2x_2 y_2 + 2x_3 y_3 + 2x_1 x_2 y_2 y_1 + 2x_1 x_3 y_3 y_1 + 2x_2 x_3 y_2 y_3$ Therefore, they are equivalent.