# CSC411 Fall 2018 – Homework 7 Solution

## 1 Representer Theorem

### Part a)

Let  $\mathbf{w} = \mathbf{w}_{\Psi} + \mathbf{w}_{\perp}$ , in which  $\mathbf{w}_{\Psi}$  is the projection onto  $\Psi$ . Note that  $z = \mathbf{w}_{\Psi}^T \Psi + \mathbf{w}_{\perp}^T \Psi = \mathbf{w}_{\Psi}^T \Psi$  due to orthogonality. Therefore, the minimizer of expected loss  $\sum_{i=1}^{N} \mathcal{L}(y^{(i)}, t^{(i)})$  does not depend on  $\mathbf{w}_{\perp}$ . (+1pt)

On the other hand,  $\|\mathbf{w}\|^2$  is minimized when  $\mathbf{w}_{\perp} = 0$ , since  $\|\mathbf{w}_{\Psi} + \mathbf{w}_{\perp}\|^2 = \|\mathbf{w}_{\Psi}\|^2 + \|\mathbf{w}_{\perp}\|^2 \ge \|\mathbf{w}_{\Psi}\|^2$ , where the first equality follows from orthogonality. (+1pt) Thus the optimal  $\mathbf{w}$  lies in the row space of  $\Psi$ .

$$\begin{split} \mathbf{Part\;b)} & \qquad \mathcal{J}(\mathbf{w}) = \frac{1}{2N}\|t - \Psi\mathbf{w}\|^2 + \frac{\lambda}{2}\|\mathbf{w}\|^2 \\ &= \frac{1}{2N}(t - \Psi\mathbf{w})^T(t - \Psi\mathbf{w}) + \frac{\lambda}{2}\mathbf{w}^T\mathbf{w} \\ &= \frac{1}{2}\alpha(\frac{1}{N}\mathbf{K}^T\mathbf{K} + \lambda\mathbf{K})\alpha - \frac{1}{N}t^T\mathbf{K}\alpha + \frac{1}{2N}t^Tt. \; \textbf{(+2pt)} \end{split}$$

Plug in the minimum of this quadratic and simplify:  $\alpha = (\mathbf{K} + \lambda N\mathbf{I})^{-1}t$ . (The assumption that  $\mathbf{K}$  is positive definite allows us to cancel the  $\mathbf{K}$ s in this step.) (+1pt)

### 2 Compositional Kernels

Here we use superscripts to denote coordinates of the feature vectors.

### Part a)

$$\Psi_S(\boldsymbol{x}) = [\Psi_1(\boldsymbol{x}); \Psi_2(\boldsymbol{x})] = [\Psi_1^{(1)}(\boldsymbol{x}); ... \Psi_1^{(m)}(\boldsymbol{x}); \Psi_2^{(1)}(\boldsymbol{x}); ... \Psi_2^{(n)}(\boldsymbol{x})]. \quad \textbf{(+1pt)}$$

### Part b)

$$\Psi_S(\boldsymbol{x}) = vec[\Psi_1(\boldsymbol{x}) \otimes \Psi_2(\boldsymbol{x})] = [\Psi_1^{(1)}(\boldsymbol{x})\Psi_2^{(1)}(\boldsymbol{x}); ...\Psi_1^{(m)}(\boldsymbol{x})\Psi_2^{(1)}(\boldsymbol{x}); ...\Psi_1^{(m)}(\boldsymbol{x})\Psi_2^{(n)}(\boldsymbol{x})].$$

- (+1pt) for writing the product of two kernels in terms of two feature maps
- (+1pt) for noting  $\Psi_S$  should contain products of elements in the two feature maps
- (+1pt) for taking the outer product of two feature maps and flatten.