HOMEWORK 2

VIKRAM GUPTA

1. Prob 2b.

Given that:

(1)
$$\mathcal{L}(\mathbf{w}, S) = \frac{1}{2} \sum_{n=1}^{N} c_n (y_n - \mathbf{w}^t \mathbf{x}_n)^2 + \frac{1}{2} \lambda \mathbf{w}^t \mathbf{w}$$

To convert (1) to matrix form, each c_n should be multiplied to individual $e_n = (y_n - \mathbf{w}^t \mathbf{x}_n)^2$ elements, and then their sum should be computed.

Let $C = [c_1, c_2, \dots, c_N]^T$, and, for \mathcal{L} to be a scalar quantity as expected, the matrix form of the equation should have a $N \times N$ matrix in place of C. The elements c_n and e_n should still be multiplied individually. Hence the C should be used as a diagonal matrix $\mathbf{C_d}$ where

(2)
$$\mathbf{C_d} = \begin{bmatrix} c_1 & 0 & 0 & \cdots \\ 0 & c_2 & 0 & \cdots \\ \vdots & \vdots & \vdots & \ddots \\ 0 & 0 & \cdots & c_n \end{bmatrix}$$

The matrix form of \mathcal{L} can be written as:

$$\mathcal{L} = \frac{1}{2}[(\mathbf{Y} - \mathbf{X}\mathbf{W})^{\mathbf{T}}\mathbf{C}_{\mathbf{d}}(\mathbf{Y} - \mathbf{X}\mathbf{W})] + \frac{1}{2}\lambda\mathbf{W}^{\mathbf{T}}\mathbf{W}$$

$$= \frac{1}{2}[\mathbf{Y}^{\mathbf{T}}\mathbf{C}_{\mathbf{d}}\mathbf{Y} - \mathbf{Y}^{\mathbf{T}}\mathbf{C}_{\mathbf{d}}\mathbf{X}\mathbf{W} - (\mathbf{X}\mathbf{W})^{\mathbf{T}}\mathbf{C}_{\mathbf{d}}\mathbf{Y} + \mathbf{W}^{\mathbf{T}}\mathbf{X}^{\mathbf{T}}\mathbf{C}_{\mathbf{d}}\mathbf{X}\mathbf{W}] + \frac{1}{2}\lambda\mathbf{W}^{\mathbf{T}}\mathbf{W}$$

$$= \frac{1}{2}[\mathbf{Y}^{\mathbf{T}}\mathbf{C}_{\mathbf{d}}\mathbf{Y} - 2\mathbf{W}^{\mathbf{T}}\mathbf{X}^{\mathbf{T}}\mathbf{C}_{\mathbf{d}}\mathbf{Y} + \mathbf{W}^{\mathbf{T}}\mathbf{X}^{\mathbf{T}}\mathbf{C}_{\mathbf{d}}\mathbf{X}\mathbf{W} + \lambda\mathbf{W}^{\mathbf{T}}\mathbf{W}]$$

To find the normal equation we differentiate \mathcal{L}

$$\frac{d\mathcal{L}}{d\mathbf{W}} = 0$$

$$\frac{1}{2}[\phi - 2\mathbf{X}^{T}\mathbf{C_{d}}\mathbf{Y} + 2\mathbf{X}^{T}\mathbf{C_{d}}\mathbf{X}\mathbf{W} + 2\lambda\mathbf{W}] = 0$$

Hence,

$$\mathbf{X}^{\mathbf{T}}\mathbf{C_{d}}\mathbf{Y} = \mathbf{X}^{\mathbf{T}}\mathbf{C_{d}}\mathbf{X}\mathbf{W} + \lambda\mathbf{W}$$

And the normal equation is:

$$\mathbf{W} = (\mathbf{X^T}\mathbf{C_d}\mathbf{X} + \lambda \mathbf{I})^{-1}\mathbf{X^T}\mathbf{C_d}\mathbf{Y}$$

where I is identity matrix of size $D \times D$, and, D is the number of dimensions in the data.