

# Homework Set 4, CPSC 8420, Fall 2020

Last Name, First Name

**Due 11/18/2020, Wednesday, 11:59PM EST**

## Problem 1

Given  $X, Y, Z$ , now please follow the idea/method used in LDA/PCA to find the best solution to:

$$\begin{aligned} & \underbrace{\arg \max}_{a,b} a^T Z b \\ & \text{s.t. } a^T X a = 1, b^T Y b = 1 \end{aligned} \quad (1)$$

- Let  $a = X^{-1/2}u, b = Y^{-1/2}v$ , and we have:

$$\begin{aligned} a^T X a = 1 & \Rightarrow u^T X^{-1/2} X X^{-1/2} u = 1 \Rightarrow u^T u = 1 \\ b^T Y b = 1 & \Rightarrow v^T Y^{-1/2} Y Y^{-1/2} v = 1 \Rightarrow v^T v = 1 \end{aligned} \quad (2)$$

therefore it is equivalent to solve:

$$\begin{aligned} & \underbrace{\arg \max}_{u,v} u^T X^{-1/2} Z Y^{-1/2} v \\ & \text{s.t. } u^T u = 1, v^T v = 1 \end{aligned} \quad (3)$$

Apparently, if we set  $[U, \Sigma, V] = \text{svd}(X^{-1/2} Z Y^{-1/2})$ , then  $u = U[:, 1]$  and  $v = V[:, 1]$ ,  $u^T X^{-1/2} Z Y^{-1/2} v = u^T U \Sigma V^T v = \sigma_1$ ,  $a = X^{-1/2}u, b = S_Y^{-1/2}v$ .

- By utilizing Lagrangian Multipliers, we have:

$$J(a, b) = a^T Z b - \frac{\lambda}{2}(a^T X a - 1) - \frac{\theta}{2}(b^T Y b - 1) \quad (4)$$

taking the derivative w.r.t.  $a, b$  and set to be 0, we obtain:

$$\begin{aligned} Z b - \lambda X a &= 0 \\ Z a - \theta Y b &= 0 \end{aligned} \quad (5)$$

Multiplying  $a^T$  and  $b^T$  on the left of above equations and due to the fact that  $a^T X a = 1$ ,  $b^T Y b = 1$ , we have  $\lambda = \theta = a^T Z b$ .

Again by multiplying  $X^{-1}$  and  $Y^{-1}$  on the left of above equations, we get:

$$\begin{aligned} X^{-1}Zb &= \lambda a \\ Y^{-1}Za &= \lambda b \end{aligned} \tag{6}$$

Plugin the second equation to the first we get  $X^{-1}ZY^{-1}Za = \lambda^2 a$ , and plugin the first to second we get  $Y^{-1}ZX^{-1}Zb = \lambda^2 b$ , where we can get the optimal  $a, b$  respectively.

## Problem 2

Recall vanilla SVM objective:

$$L(w, b, \alpha) = \frac{1}{2} \|w\|_2^2 - \sum_{i=1}^m \alpha_i [y_i(w^T x_i + b) - 1] \quad s.t. \quad \alpha_i \geq 0 \tag{7}$$

If we denote the margin as  $\gamma$ , and vector  $\alpha = [\alpha_1, \alpha_2, \dots, \alpha_m]$ , now please show  $\gamma^2 * \|\alpha\|_1 = 1$ .

$$\begin{aligned} L(w, b, \alpha) &= \frac{1}{2} \|w\|_2^2 - \sum_{i=1}^m \alpha_i [y_i(w^T x_i + b) - 1] \quad s.t. \quad \alpha_i \geq 0 \\ \frac{\partial L}{\partial w} &= 0 \Rightarrow w = \sum_{i=1}^m \alpha_i y_i x_i \\ \frac{\partial L}{\partial b} &= 0 \Rightarrow \sum_{i=1}^m \alpha_i y_i = 0 \end{aligned} \tag{8}$$

For any support vector  $x_i$ , since  $\langle w, x_i \rangle + b = y_i$ , thus  $b$  can be obtained via  $b = y_i - \sum_{j=1}^m \alpha_j y_j \langle x_i, x_j \rangle$ .

Multiplying both sides by  $\alpha_i y_i$  and taking the sum will give:

$$\sum_{i=1}^m \alpha_i y_i b = \sum_{i=1}^m \alpha_i y_i^2 - \sum_{i,j=1}^m \alpha_i \alpha_j y_i y_j \langle x_i, x_j \rangle$$

By the fact  $\sum_{i=1}^m \alpha_i y_i = 0, y_i^2 = 1, w = \sum_{i=1}^m \alpha_i y_i x_i$ , the above equation is equivalent to:

$$0 = \sum_{i=1}^m \alpha_i - \|w\|^2$$

, which says  $\|\alpha\|_1 = \|w\|^2 = \frac{1}{\gamma^2} \implies \gamma^2 * \|\alpha\|_1 = 1$ .