## CS6316 Homework 02: The Bias-variance Tradeoff, SVMs and Kernel Methods

## Gabriel Hanson

Due on Oct. 27, 2022 11:59 PM

## 1. Bias-Variance Tradeoff (12 points)

Please refer to the attached iPython notebook file.

2. KKT Conditions (4 points) The Lagrangian form of SVMs with slack variable  $\xi$  is formulated as

$$L(\boldsymbol{w}, b, \boldsymbol{\xi}, \boldsymbol{\alpha}, \boldsymbol{\beta}) = \frac{1}{2} \|\boldsymbol{w}\|_{2}^{2} + C \sum_{i=1}^{m} \xi_{i}$$

$$- \sum_{i=1}^{m} \alpha_{i} (y_{i}(\boldsymbol{w}^{\mathsf{T}}\boldsymbol{x}_{i} + b) - 1 + \xi_{i})$$

$$- \sum_{i=1}^{m} \beta_{i} \xi_{i}$$

$$(1)$$

Similar to SVMs in separable cases (page 19), to find the KKT conditions (as in page 29, we need to compute the derivative with respect to all parameters  $\{\boldsymbol{w}, b, \boldsymbol{\xi}, \boldsymbol{\alpha}, \boldsymbol{\beta}\}$ . Overall, please show that the KKT conditions can be represented to the following equations

$$\boldsymbol{w} = \sum_{i=1}^{m} \alpha_i y_i \boldsymbol{x}_i \tag{2}$$

$$\sum_{i=1}^{m} \alpha_i y_i = 0 \tag{3}$$

$$\alpha_i + \beta_i = C \tag{4}$$

$$\alpha_i = 0 \quad \text{or} \quad y_i(\boldsymbol{w}^\mathsf{T} \boldsymbol{x}_i + b) = 1 - \xi_i$$
 (5)

$$\beta_i = 0 \quad \text{or} \quad \xi_i = 0 \tag{6}$$

Solution:

$$\frac{\partial L}{\partial \boldsymbol{w}} = \boldsymbol{w} - \sum_{i=1}^{m} \alpha_i y_i x_i = 0 \qquad \longrightarrow \quad \boldsymbol{w} = \sum_{i=1}^{m} \alpha_i y_i x_i$$

$$\frac{\partial L}{\partial b} = -\sum_{i=1}^{m} \alpha_i y_i = 0 \qquad \longrightarrow \quad \alpha_i y_i = 0$$

$$\frac{\partial L}{\partial \xi} = C - \sum_{i=1}^{m} \alpha_i - \sum_{i=1}^{m} \beta_i = 0 \qquad \longrightarrow \quad C = \alpha_i + \beta_i$$

and

$$\forall i \quad \alpha_i (y_i(\boldsymbol{w}^{\mathsf{T}} \boldsymbol{x}_i + b) - 1 + \xi_i = 0 \qquad \longrightarrow \qquad \alpha_i = 0 \quad \text{or} \quad y_i(\boldsymbol{w}^{\mathsf{T}} \boldsymbol{x}_i + b) = 1 - \xi_i$$

$$\forall i \quad \beta_i \xi_i = 0 \qquad \longrightarrow \qquad \beta_i = 0 \quad \text{or} \quad \xi_i = 0$$

3. Kernel Methods (4 points) In our lecture, we show that a special case of the polynomial kernels

$$K(\boldsymbol{x}, \boldsymbol{x}') = (\langle \boldsymbol{x}, \boldsymbol{x}' \rangle + c)^d \tag{7}$$

with d=2 and  $x, x' \in \mathbb{R}^2$ . On page 53 of the slides, we should how this special case can be decomposed as a dot product with a nonlinear mapping  $\Phi(\cdot)$ 

$$K(\boldsymbol{x}, \boldsymbol{x}') = \langle \Phi(\boldsymbol{x}), \Phi(\boldsymbol{x}') \rangle. \tag{8}$$

In this problem, consider d=3 with  $x, x' \in \mathbb{R}^2$  and show how the  $\Phi(x)$  is defined in this case.

Solution:

$$K(\boldsymbol{x}, \boldsymbol{x}') = (\langle \boldsymbol{x}, \boldsymbol{x}' \rangle + c)^3 \ \forall \ \boldsymbol{x}, \boldsymbol{x}' \in \mathbb{R}^2$$
$$K(\boldsymbol{x}, \boldsymbol{x}') = (\langle \boldsymbol{x}, \boldsymbol{x}' \rangle + c)^2 * (\langle \boldsymbol{x}, \boldsymbol{x}' \rangle + c)$$
$$= (x_1 x_1' + x_2 x_2' + c)^2 * (x_1 x_1' + x_2 x_2' + c)$$

This first term was solved for in slide 56 of the SVM lecture, which I will substitute in here

$$= (x_1^2 x_1'^2 + x_2^2 x_2'^2 + 2x_1 x_1' x_2 x_2' + 2cx_1 x_1' + 2cx_2 x_2' + c^2) * (x_1 x_1' + x_2 x_2' + c)$$

$$= (x_1^3 x_1'^3 + x_1 x_1' x_2^2 x_2'^2 + 2x_1^2 x_1'^2 x_2 x_2' + 2cx_1^2 x_1'^2 + 2cx_1 x_1' x_2 x_2' + c^2 x_1 x_1'$$

$$+ x_1^2 x_1'^2 x_2 x_2' + x_2^3 x_2'^3 + 2x_1 x_1' x_2^2 x_2'^2 + 2cx_1 x_1' x_2 x_2' + 2cx_2^2 x_2'^2 + c^2 x_2 x_2'$$

$$+ cx_1^2 x_1'^2 + cx_2^2 x_2'^2 + 2cx_1 x_1' x_2 x_2' + 2c^2 x_1 x_1' + 2c^2 x_2 x_2' + c^3)$$

Combining terms here leaves us with

$$= (x_1^3 x_1'^3 + x_2^3 x_2'^3 + 3x_1^2 x_1'^2 x_2 x_2' + 3x_1 x_1' x_2^2 x_2'^2 + 3cx_1^2 x_1'^2 + 3cx_2^2 x_2'^2 + 6cx_1 x_1' x_2 x_2' + 3c^2 x_1 x_1' + 3c^2 x_2 x_2' + c^3)$$
 which can be rewritten as

$$\left[x_{1}^{3}, x_{2}^{3}, \sqrt{3}x_{1}^{2}x_{2}, \sqrt{3}x_{1}x_{2}^{2}, \sqrt{3}cx_{1}^{2}, \sqrt{3}cx_{2}^{2}, \sqrt{6}cx_{1}x_{2}, \sqrt{3}cx_{1}, \sqrt{3}cx_{2}, c^{3/2}\right] \begin{bmatrix} x_{1}^{\prime 3} \\ x_{2}^{\prime 3} \\ \sqrt{3}x_{1}^{\prime 2}x_{2}^{\prime} \\ \sqrt{3}cx_{1}^{\prime 2}x_{2}^{\prime} \\ \sqrt{3}cx_{1}^{\prime 2} \\ \sqrt{3}cx_{1}^{\prime 2} \\ \sqrt{6}cx_{1}^{\prime}x_{2}^{\prime} \\ \sqrt{3}cx_{2}^{\prime} \\ \sqrt{3}cx_{2}^{\prime} \\ c^{3/2} \end{bmatrix}$$

Let 
$$K(\boldsymbol{x}, \boldsymbol{x}') = \langle \phi(\boldsymbol{x}), \phi(\boldsymbol{x}') \rangle$$
, then

$$\phi(\boldsymbol{x}) = [x_1^3, x_2^3, \sqrt{3}x_1^2x_2, \sqrt{3}x_1x_2^2, \sqrt{3}cx_1^2, \sqrt{3}cx_2^2, \sqrt{6}cx_1x_2, \sqrt{3}cx_1, \sqrt{3}cx_2, c^{3/2}]^\mathsf{T}$$