

Lab 6

SVM Problem Set

Raymond Lin and Kenny Guo

$$1) \quad \vec{p} = \vec{x}_i - y_i y_i \cdot \frac{\vec{w}}{\|\vec{w}\|}$$

$$0 = \vec{w} \cdot \vec{p} + b$$

$$0 = \vec{w} \cdot (\vec{x}_i - y_i y_i \cdot \frac{\vec{w}}{\|\vec{w}\|}) + b$$

$$0 = \vec{w} \cdot \vec{x}_i - y_i y_i \cdot \frac{\vec{w} \cdot \vec{w}}{\|\vec{w}\|} + b$$

$$0 = \vec{w} \cdot \vec{x}_i - y_i y_i \cdot \frac{\|\vec{w}\| \|\vec{w}\| \cos \theta}{\|\vec{w}\|} + b$$

* $\theta = 0$ because the vector \vec{w} dotted with itself so the angle between them is 0

$$0 = \vec{w} \cdot \vec{x}_i - y_i y_i \|\vec{w}\| + b$$

$$y_i y_i \|\vec{w}\| = \vec{w} \cdot \vec{x}_i + b$$

$$y_i = \left(\frac{\vec{w} \cdot \vec{x}_i}{\|\vec{w}\|} + \frac{b}{\|\vec{w}\|} \right) \frac{1}{y_i}$$

* since y_i is just a sign, $\frac{1}{y_i} = y_i$

$$y_i = y_i \left(\frac{\vec{w} \cdot \vec{x}_i}{\|\vec{w}\|} + \frac{b}{\|\vec{w}\|} \right)$$

2) Corollary 1

We want to show $\|\vec{w}\|^2 = \sum_{i=1}^n \sum_{j=1}^n y_i y_j \alpha_i \alpha_j \vec{x}_i \cdot \vec{x}_j$

$$\|\vec{w}\|^2 = w_1^2 + w_2^2 + \dots + w_n^2$$

$$w_1^2 = \left(\sum_{i=1}^n \alpha_i y_i x_{i1} \right)^2 = \left(\alpha_1 y_1 x_{11} + \alpha_2 y_2 x_{21} + \dots \right)^2$$

$$= \sum_{i=1}^n \sum_{j=1}^n y_i y_j \alpha_i \alpha_j x_{i1} x_{j1}$$

and we do this for n weights so we have

$$\|\vec{w}\|^2 = \sum_{i=1}^n \sum_{j=1}^n y_i y_j \alpha_i \alpha_j \vec{x}_i \cdot \vec{x}_j$$

We want to show $L(\vec{w}, b, \vec{\alpha}) = \frac{1}{2} \|\vec{w}\|^2 - \sum \alpha_i [y_i (\vec{w} \cdot \vec{x}_i + b) - 1] = \sum \alpha_i - \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n y_i y_j \alpha_i \alpha_j \vec{x}_i \cdot \vec{x}_j$

$$L(\vec{w}, b, \vec{\alpha}) = \frac{1}{2} \|\vec{w}\|^2 - \sum \alpha_i [y_i (\vec{w} \cdot \vec{x}_i + b) - 1]$$

$$= \frac{1}{2} \|\vec{w}\|^2 - \sum (\alpha_i y_i \vec{w} \cdot \vec{x}_i + \alpha_i y_i b - \alpha_i)$$

$$= \frac{1}{2} \|\vec{w}\|^2 - \sum \alpha_i y_i \vec{w} \cdot \vec{x}_i - \sum \alpha_i y_i b + \sum \alpha_i$$

* $\sum \alpha_i y_i b = 0$

$$= \frac{1}{2} \|\vec{w}\|^2 - \sum \alpha_i y_i \vec{w} \cdot \vec{x}_i + \sum \alpha_i$$

$$= \frac{1}{2} \|\vec{w}\|^2 - \vec{w} \cdot \sum \alpha_i y_i \vec{x}_i + \sum \alpha_i$$

* $\sum \alpha_i y_i \vec{x}_i = \vec{w}$

$$\begin{aligned}
&= \frac{1}{2} \|\vec{w}\|^2 - \vec{w} \cdot \vec{w} + \sum \alpha_i \quad * \vec{w} \cdot \vec{w} = \|\vec{w}\|^2 \cos(0) = \|\vec{w}\|^2 \\
&= -\frac{1}{2} \|\vec{w}\|^2 + \sum \alpha_i \\
&= -\frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n y_i y_j \alpha_i \alpha_j \vec{x}_i \cdot \vec{x}_j + \sum_{i=1}^n \alpha_i \quad \text{by Corollary 1}
\end{aligned}$$

3) We want to show $\max_x \min_y f(x, y) \leq \min_y \max_x f(x, y)$

$$\text{Let } (x^*, y^*) = \max_x \min_y f(x, y)$$

Then we know $(x^*, y^*) \leq (x^*, y)$ for any y .

Thus we know every value in $\max_x f(x, y) \geq (x^*, y)$

by the transitive property we know every value in $\max_x f(x, y) \geq (x^*, y) \geq (x^*, y^*)$

Thus $\min_y \max_x f(x, y) \geq (x^*, y^*)$ since all max values are greater than or equal to (x^*, y^*) .

So we have $(x^*, y^*) \leq \min_y \max_x f(x, y)$ or $\max_x \min_y f(x, y) \leq \min_y \max_x f(x, y)$

4) We know that $\hat{y}_i = y_i (\vec{w} \cdot \vec{x}_i + b)$

and on the margin $\hat{y}_i = 1$

$$\text{So } 1 \leq y_i (\vec{w} \cdot \vec{x}_i + b)$$

$$-y_i (\vec{w} \cdot \vec{x}_i + b) \leq 0$$

When $y_i = -1$

$$-1(-1)(\vec{w} \cdot \vec{x}_i + b) \leq 0$$

$$\vec{w} \cdot \vec{x}_i + b \leq 0$$

$$b \leq -\vec{w} \cdot \vec{x}_i$$



* and we want to be close to margin

so we want to maximize $-\vec{w} \cdot \vec{x}_i$

When $y_i = 1$

$$-1(\vec{w} \cdot \vec{x}_i + b) \leq 0$$

$$-b \leq \vec{w} \cdot \vec{x}_i$$

$$b \geq -\vec{w} \cdot \vec{x}_i$$

* so we want to minimize $-\vec{w} \cdot \vec{x}_i$

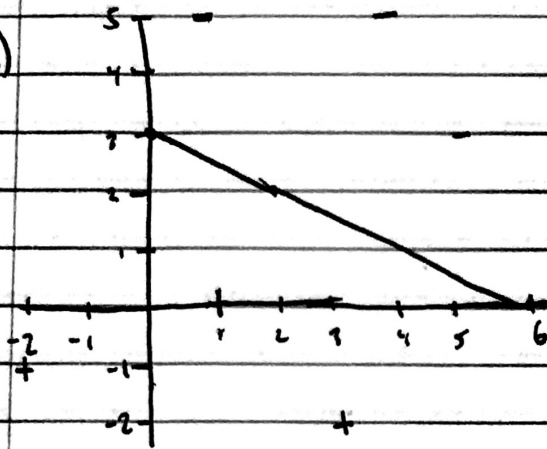
In order to find the optimal bias, we want to

find the midpoint of the support vectors.

$$b^* = \frac{1}{2} \left(\max_{i: y_i = -1} -\vec{w}^* \cdot \vec{x}_i + \min_{i: y_i = 1} -\vec{w}^* \cdot \vec{x}_i \right) \quad * \text{distribute the negative}$$

$$b^* = -\frac{1}{2} \left(\max_{i: y_i = -1} \vec{w}^* \cdot \vec{x}_i + \min_{i: y_i = 1} \vec{w}^* \cdot \vec{x}_i \right)$$

5) a)



b) $x_2 = -\frac{1}{2}x_1 + 3$

c) $\vec{x}_1, \vec{x}_2, \vec{x}_5$

d) $\vec{w} = a \begin{bmatrix} -1 \\ 2 \end{bmatrix}$

e) $l = -a + b$

$$l = -1(-a + 10a) + b$$

$$l = 11a + b$$

$$11a - b = -a + b$$

$$12a = 2b$$

$$b = 6a$$

$$l = 5a$$

$$a = 0.2 \quad b = 1.2$$

$$\vec{w}^* = \begin{bmatrix} -0.2 \\ 0.4 \end{bmatrix}$$

f) $\begin{bmatrix} -1/5 \\ -2/5 \end{bmatrix} = -\alpha_1 \begin{bmatrix} 1 \\ 6 \end{bmatrix} - \alpha_2 \begin{bmatrix} 5 \\ 3 \end{bmatrix} + \alpha_5 \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$$\alpha_1 + \alpha_2 = \alpha_5$$

$$2(-\alpha_1 - 5\alpha_2 + \alpha_5) = -5\alpha_1 - 3\alpha_2$$

$$-2\alpha_1 - 10\alpha_2 + 2\alpha_5 = -5\alpha_1 - 3\alpha_2$$

$$3\alpha_1 - 7\alpha_2 + 2\alpha_5 = 0$$

$$3\alpha_1 - 7\alpha_2 + 2\alpha_5 + 2\alpha_5 = 0$$

$$5\alpha_1 - 5\alpha_2 = 0$$

$$\alpha_1 = \alpha_2$$

$$\alpha_1 = \alpha_2 \quad -\frac{1}{5} = \alpha_1 - 5\alpha_1 + 2\alpha_5$$

$$\alpha_5 = 2\alpha_1 \quad -\frac{1}{5} = -4\alpha_1$$

$$\alpha_1 = \frac{1}{20} \quad \alpha_2 = 0$$

$$\alpha_3 = \frac{1}{20} \quad \alpha_4 = 0$$

$$\alpha_5 = \frac{1}{10} \quad \alpha_6 = 0$$