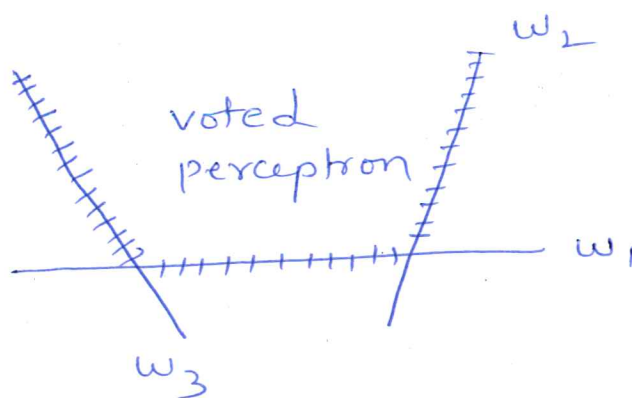


HW #3 Solutions

1. (a) The decision boundary of voted perceptron is non-linear. You cannot represent the decision function as a linear function. Take a simple two dimensional example, it is a piece-wise linear function.



- (b) Average perceptron is linear.

2. Scale learning rate using importance weight $h_i \rightarrow$ You want to move more aggressive with learning for examples which are important.

$$w_{\text{new}} = w_{\text{old}} + \eta_i y_i x_i$$

$$\text{where, } \eta_i = h_i * \eta$$

3. Balance the unbalanced by either

(a) up-sampling the minority class, i.e.,
create more copies of +ve examples

(or)

(b) Down-sampling the majority class, i.e.,
Remove negative examples appropriately

(Alternative)

We can use solution of Q_2 to solve

Q_3 .

H_+ will be the weight of all
+ve examples

H_- will be the weight of all
-ve examples

Make H_+ larger than H_- appropriately

Say $H_+ = 0.9$ & $H_- = 0.1$

Q4.

$$\text{SCORE}(i) = w \cdot f(x_i)$$

$$\text{SCORE}(j) = w \cdot f(x_j)$$

If candidate i is better than j , then

$$\text{SCORE}(i) > \text{SCORE}(j)$$

$$w \cdot f(x_i) > w \cdot f(x_j)$$

$$w \cdot (f(x_i) - f(x_j)) > 0$$

$$\Rightarrow \text{sign} \left(w \cdot \underbrace{f(x_i) - f(x_j)}_x \right) > 0$$

If mistake, then update weights

$$w_{\text{new}} = w_{\text{old}} + \eta \left(\underbrace{f(x_i) - f(x_j)}_x \right)$$

$$y = +1$$

Alternatively,

similar to multi-class classification

↑ increase score of i & ↓ decrease score of j

$$w = w + f(x_i)$$

$$w = w - f(x_j)$$

} same net result

Q5.

(a)

$$\|x - c_+\| \leq \|x - c_-\|$$

Square both sides will make life easier

$$\|x - c_+\|^2 \leq \|x - c_-\|^2$$

$$\|x\|^2 - 2x \cdot c_+ + \|c_+\|^2 \leq \|x\|^2 - 2x \cdot c_- + \|c_-\|^2$$

bring all the terms to one side.

$$2x \cdot c_+ - 2x \cdot c_- + \|c_-\|^2 - \|c_+\|^2 \geq 0$$

$$\underbrace{2(c_+ - c_-) \cdot x}_w + \underbrace{\|c_-\|^2 - \|c_+\|^2}_b \geq 0$$

$$w = 2(c_+ - c_-)$$

$$b = \|c_-\|^2 - \|c_+\|^2$$

(b)

$$w = 2 \frac{\sum_{i=1}^{n_+} x_i}{n_+} - 2 \frac{\sum_{j=1}^{n_-} x_j}{n_-}$$

$$\left. \begin{aligned} \alpha &= \frac{2}{n_+} \text{ for +ve examples} \\ \alpha &= \frac{2}{n_-} \text{ for -ve examples} \end{aligned} \right\} \text{(c) All are support vectors}$$