

## CSE 404: Perceptron Pop-Quiz Solution

### 1. Why does PLA Work?

We can show that the weight update rule above has the nice interpretation that it moves in the direction of classifying  $\mathbf{x}_*$  correctly.

- Show that  $y_* \mathbf{w}(t)^T \mathbf{x}_* < 0$ ;
  - **Solution:** Since data point  $(y_*, \mathbf{x}_*)$  is misclassified by  $\mathbf{w}(t)$ , then the prediction  $\text{sign}(\mathbf{w}(t)^T \mathbf{x}_*)$  is different sign from  $y_*$ , i.e., when  $y_* = 1$ ,  $\text{sign}(\mathbf{w}(t)^T \mathbf{x}_*) = -1$ , and  $\mathbf{w}(t)^T \mathbf{x}_* < 0$ , and therefore  $y_* \mathbf{w}(t)^T \mathbf{x}_* < 0$ . We can draw the same conclusion when  $y_* = -1$ .
- Show that  $y_* \mathbf{w}(t+1)^T \mathbf{x}_* > y_* \mathbf{w}(t)^T \mathbf{x}_*$ ;
  - **Solution:** Note that  $\mathbf{w}(t+1) = \mathbf{w}(t) + y_* \mathbf{x}_*$ , therefore

$$\begin{aligned} y_* \mathbf{w}(t+1)^T \mathbf{x}_* &= y_* (\mathbf{w}(t) + y_* \mathbf{x}_*)^T \mathbf{x}_* \\ &= y_* (\mathbf{w}(t))^T \mathbf{x}_* + y_* (y_* \mathbf{x}_*)^T \mathbf{x}_* \\ &= y_* (\mathbf{w}(t))^T \mathbf{x}_* + y_*^2 \mathbf{x}_*^T \mathbf{x}_* > y_* (\mathbf{w}(t))^T \mathbf{x}_* \end{aligned}$$

where the last inequality is because  $y_*^2 \mathbf{x}_*^T \mathbf{x}_* > 0$  (given that  $\|\mathbf{x}_*\| \neq 0$ , i.e.,  $\mathbf{x}_*$  is not an empty data point).

- Move from  $\mathbf{w}(t)$  to  $\mathbf{w}(t+1)$  is a move ‘in the right direction’ in terms of classifying  $\mathbf{x}_*$ 
  - **Solution:** Given  $y_* \mathbf{w}(t+1)^T \mathbf{x}_* > y_* \mathbf{w}(t)^T \mathbf{x}_*$ , we are pushing forward the classifier  $\mathbf{w}(t+1)$  in a way that has a higher chance to make  $y_* \mathbf{w}(t+1)^T \mathbf{x}_*$  positive, and thus a higher chance to classify  $(\mathbf{x}_*, y_*)$  correctly.