

1. Let $\vec{x}_n^{\text{orig}} = [x_{n1} \ x_{n2} \ \dots \ x_{nd}] \in \mathbb{R}^d$ for $n \in \{1, \dots, N\}$
 then $\vec{x}_n^{\downarrow} = [1 \ x_{n1} \ x_{n2} \ \dots \ x_{nd}] \in \mathbb{R}^{d+1}$ for $n \in \{1, \dots, N\}$
 and $\vec{x}_n^{\uparrow} = [2 \ x_{n1} \ x_{n2} \ \dots \ x_{nd}] \in \mathbb{R}^{d+1}$ for $n \in \{1, \dots, N\}$

Suppose for the the \vec{x}_n s that are formed conventionally,
 the first $p-1$ inputs are predicted correctly, then:

$$\text{sign}(\vec{w}_0^T \vec{x}_i) = y_i \quad \text{for } i = 0 \sim p-1$$

If we assume $\text{sign}(0) = -1$ for simplicity, and $\vec{w}_0 = \vec{0}$, then this means that:

$$\text{sign}(\vec{0}^T \vec{x}_i) = \text{sign}(0) = -1 = y_i \quad \text{for } i = 0 \sim p-1$$

If we further assume \vec{x}_p receives incorrect prediction, then

$$\text{sign}(\vec{0}^T \vec{x}_p) \neq y_p$$

$$\Rightarrow \text{sign}(0) = -1 \neq y_p \quad \therefore y_p = 1$$

Thus we update \vec{w}_1 as:

$$\vec{w}_1 = \vec{w}_0 + y_p \vec{x}_p = \vec{0} + 1 \cdot \begin{bmatrix} 1 \\ x_{p1} \\ x_{p2} \\ \vdots \\ x_{pd} \end{bmatrix} = \begin{bmatrix} 1 \\ x_{p1} \\ x_{p2} \\ \vdots \\ x_{pd} \end{bmatrix}$$

Since $\vec{w}_0 = \vec{0}$, and y_i remain unchanged no matter we use \vec{x}_i or \vec{x}_i' , so:

$$\text{sign}(\vec{w}_0^T \vec{x}_i') = \text{sign}(\vec{0}^T \vec{x}_i') = 1 = y_i \quad \text{for } i=1, \dots, p-1$$

We next want to know if \vec{x}_p' will get correct prediction or not:

$$\text{sign}(\vec{w}_0^T \vec{x}_p') = \text{sign}(\vec{0}^T \vec{x}_p') = 1 \neq -1 = y_p$$

Thus \vec{x}_p' is also mispredicted, and we update the weight vector as:

$$\vec{w}_1 = \vec{w}_0 + y_p \vec{x}_p' = \vec{0} + 1 \cdot \begin{bmatrix} x_{p1} \\ x_{p2} \\ \vdots \\ x_{pd} \end{bmatrix} = \begin{bmatrix} x_{p1} \\ x_{p2} \\ \vdots \\ x_{pd} \end{bmatrix}$$

Consider the next input \vec{x}_{p+1} and \vec{x}_{p+1}' :

$$\boxed{\text{sign}(\vec{w}_1^T \vec{x}_{p+1})} = \text{sign} \left([1 \ x_{p1} \ x_{p2} \ \dots \ x_{pd}] \begin{bmatrix} x_{p+1,1} \\ x_{p+1,2} \\ \vdots \\ x_{p+1,d} \end{bmatrix} \right)$$

$$= \text{sign} \left(1 + \sum_{i=1}^d x_{pi} x_{p+1,i} \right)$$

$$\boxed{\text{sign}(\vec{w}_1^T \vec{x}_{p+1}')} = \text{sign} \left([-x_{p1} \ -x_{p2} \ \dots \ -x_{pd}] \begin{bmatrix} x_{p+1,1} \\ x_{p+1,2} \\ \vdots \\ x_{p+1,d} \end{bmatrix} \right)$$

$$= \text{sign} \left(-1 - \sum_{i=1}^d x_{pi} x_{p+1,i} \right)$$

Therefore, consider the situation that if $\sum_{i=1}^d \pi_i x_{pi,i} = -3$

then:

$$\text{sign}(\vec{w}_1^T \vec{x}_{pi}) = \text{sign}(1 + (-3)) = \text{sign}(-2) = -1$$

$$\text{sign}(\vec{w}_1'^T \vec{x}_{pi}) = \text{sign}(4 + (-3)) = \text{sign}(1) = 1$$

\Rightarrow This will cause \vec{w}_1 to update if $y_{pi} = 1$, \vec{w}_1' to update if $y_{pi} = -1$.

If the case is $y_{pi} = -1$, and there are only $p+1$ points in the dataset,
(i.e. $\{\vec{x}_n, y_n\}_{n=1}^{p+1}$ / $\{\vec{x}_n', y_n'\}_{n=1}^{p+1}$)

then $\vec{w}_{p+1} = \vec{w}_1$

and we continue to update \vec{w}_1' and check for \vec{x}_1' :

$$\vec{w}_2' = \vec{w}_1' + y_{pi} \cdot \vec{x}_{pi}' = \begin{bmatrix} x_{p1} \\ x_{p2} \\ \vdots \\ x_{pd} \end{bmatrix} + (-1) \cdot \begin{bmatrix} x_{p+1,1} \\ x_{p+1,2} \\ \vdots \\ x_{p+1,d} \end{bmatrix} = \begin{bmatrix} 0 \\ x_{p1} - x_{p+1,1} \\ x_{p2} - x_{p+1,2} \\ \vdots \\ x_{pd} - x_{p+1,d} \end{bmatrix}$$

$$\text{sign}(\vec{w}_2'^T \vec{x}_1') = \text{sign} \left(\begin{bmatrix} 0 & x_{p1} - x_{p+1,1} & \dots & x_{pd} - x_{p+1,d} \end{bmatrix} \begin{bmatrix} x_{11} \\ x_{12} \\ \vdots \\ x_{1d} \end{bmatrix} \right)$$

$$= \text{sign} \left(\sum_{i=1}^d (x_{pi} - x_{p+1,i}) x_{1i} \right)$$

If $\sum_{i=1}^d (x_{p,i} - x_{pn,i}) x_{i,i} \leq 0$, and since $y_i = 1$ as we assumed earlier,
we have correct prediction using \vec{w}_2^1 .

We can imagine that we construct the inputs to make:

$$\text{sign}(\vec{w}_2^{1T} \vec{x}_i) = y_i \quad \text{for } i = 1, \dots, p+1$$

$$\text{then } \vec{w}_{\text{PLA}}^1 = \vec{w}_2^1 = \begin{bmatrix} 0 \\ x_{p1} - x_{pn,1} \\ \vdots \\ x_{p1} - x_{pn,d} \end{bmatrix}$$

or we can see it is trivial that each time we update \vec{w}_i^1 ,
we add/subtract 2 to w_{i0} , thus

$$\vec{w}_{\text{PLA}}^1 = \begin{bmatrix} x^k \\ \vdots \\ 1 \end{bmatrix} \quad \text{for } k \in \mathbb{Z}$$

$$\text{and } \vec{w}_{\text{PLA}}^1 = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}, \text{ so } \vec{w}_{\text{PLA}}^1 \neq \vec{w}_{\text{PLA}}. \quad \square$$