

## A simple hypothesis set - the 'perceptron'

For input  $\mathbf{x} = (x_1, \dots, x_d)$  'attributes of a customer'

Approve credit if  $\sum_{i=1}^d w_i x_i > \text{threshold}$ ,

Deny credit if  $\sum_{i=1}^d w_i x_i < \text{threshold}$ .

This linear formula  $h \in \mathcal{H}$  can be written as

$$h(\mathbf{x}) = \text{sign} \left( \left( \sum_{i=1}^d w_i x_i \right) - \text{threshold} \right)$$

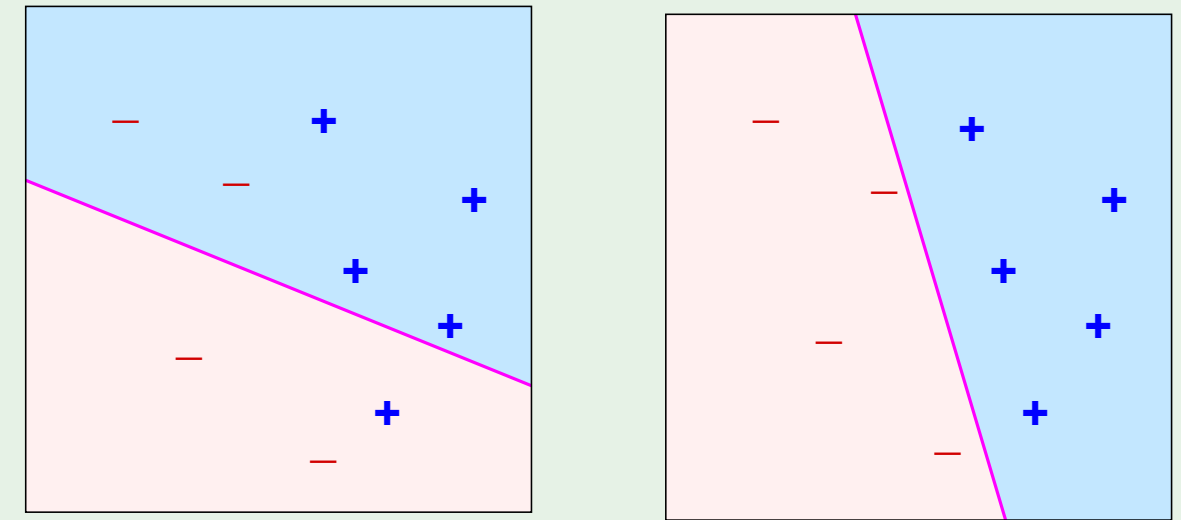
$$h(\mathbf{x}) = \text{sign} \left( \left( \sum_{i=1}^d w_i x_i \right) + w_0 \right)$$

Introduce an artificial coordinate  $x_0 = 1$ :

$$h(\mathbf{x}) = \text{sign} \left( \sum_{i=0}^d w_i x_i \right)$$

In vector form, the perceptron implements

$$h(\mathbf{x}) = \text{sign}(\mathbf{w}^T \mathbf{x})$$



'linearly separable' data

# A simple learning algorithm - PLA

The perceptron implements

$$h(\mathbf{x}) = \text{sign}(\mathbf{w}^T \mathbf{x})$$

Given the training set:

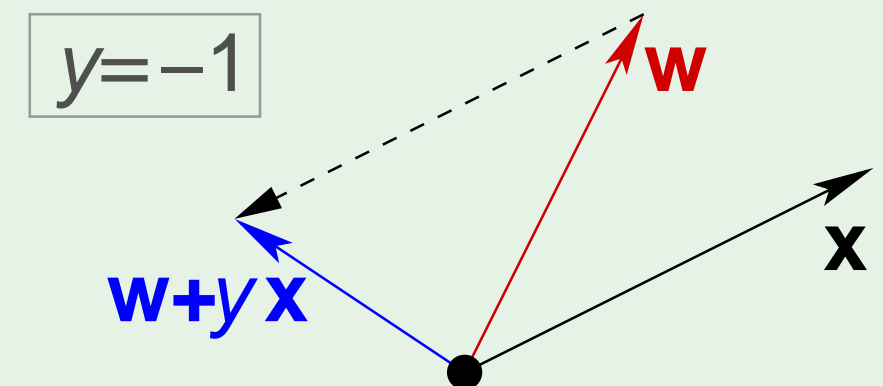
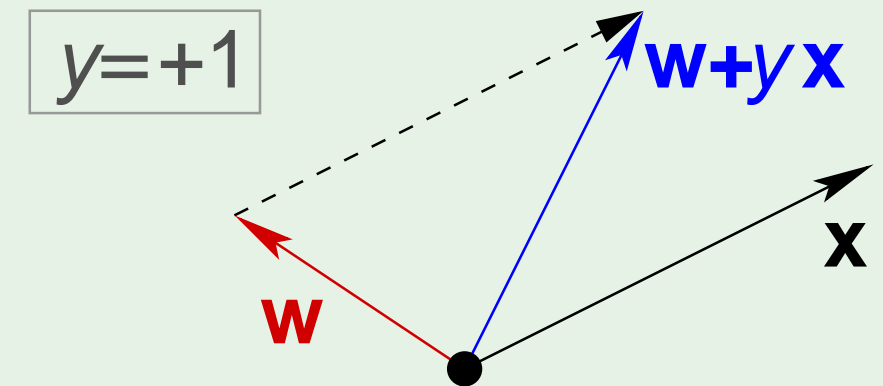
$$(\mathbf{x}_1, y_1), (\mathbf{x}_2, y_2), \dots, (\mathbf{x}_N, y_N)$$

pick a **misclassified** point:

$$\text{sign}(\mathbf{w}^T \mathbf{x}_n) \neq y_n$$

and update the weight vector:

$$\mathbf{w} \leftarrow \mathbf{w} + y_n \mathbf{x}_n$$



# Iterations of PLA

- One iteration of the PLA:

$$\mathbf{w} \leftarrow \mathbf{w} + y\mathbf{x}$$

where  $(\mathbf{x}, y)$  is a misclassified training point.

- At iteration  $t = 1, 2, 3, \dots$ , pick a misclassified point from

$$(\mathbf{x}_1, y_1), (\mathbf{x}_2, y_2), \dots, (\mathbf{x}_N, y_N)$$

and run a PLA iteration on it.

- That's it!

