

# Perceptron

Foundations of Data Analysis

April 12, 2022

# History of Perceptron

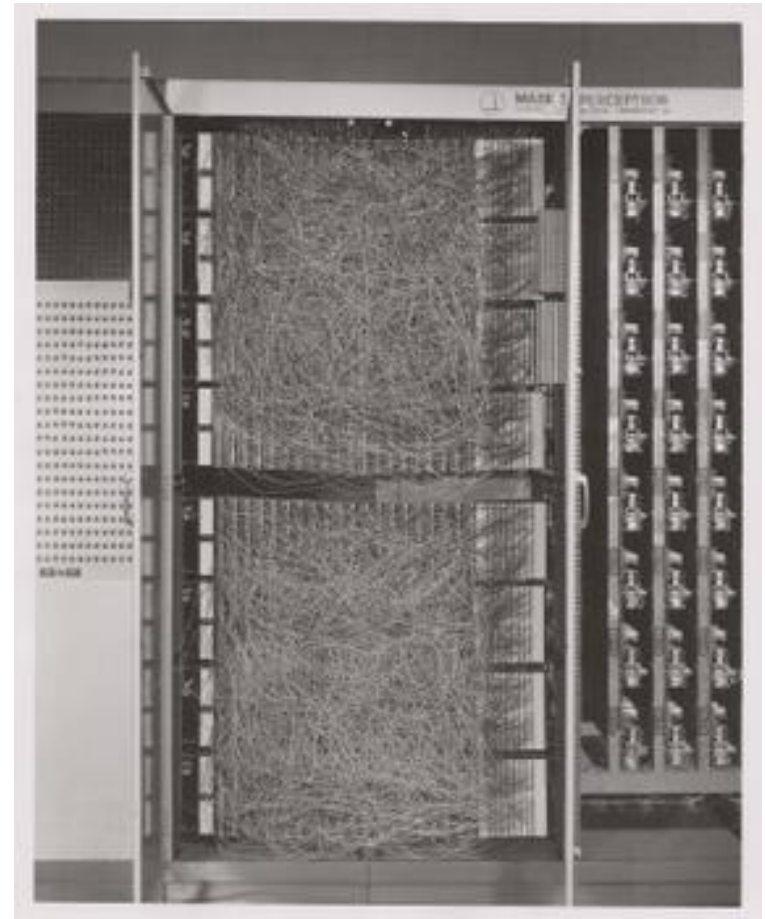
- Frank Rosenblatt
- 1928-1969



invented perceptron algorithm

# History of Perceptron

- Mark 1 Perceptron (1958)
- 20 x 20 pixel camera
- Hardware, not software!



"an electronic computer that [the Navy] expects will be able to walk, talk, see, write, reproduce itself and be conscious of its existence"

- NY Times, 1958

# Perceptron Learning Algorithm

- First neural network learning model in the 1960's

# Perceptron Learning Algorithm

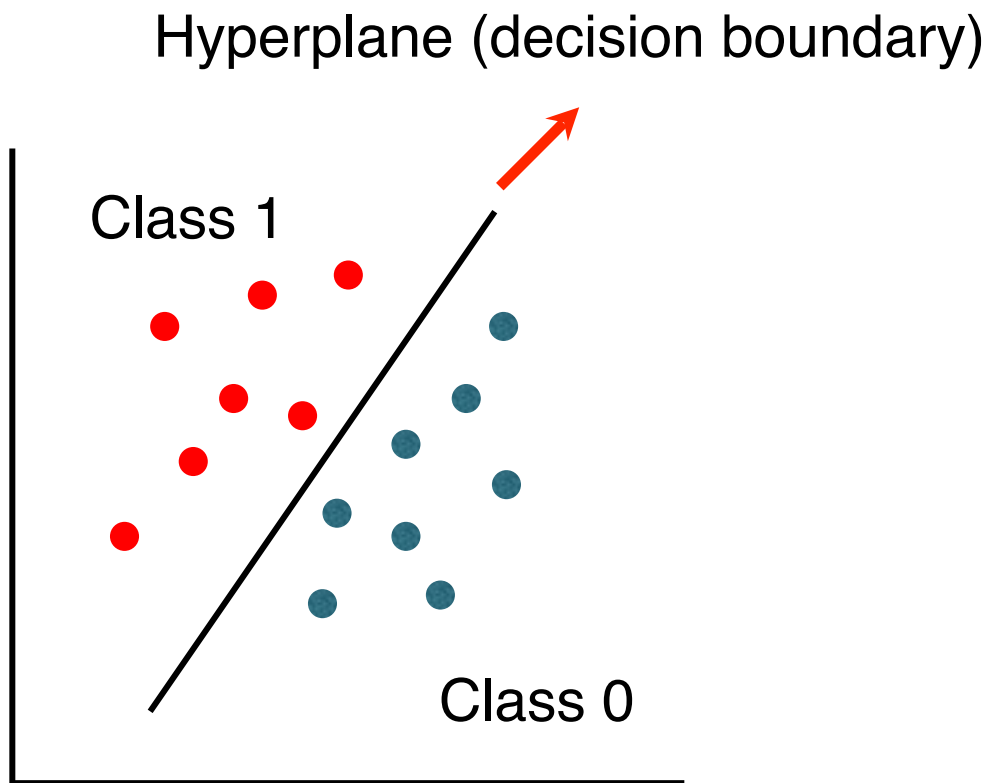
- First neural network learning model in the 1960's
- Simple and limited (single layer model)

# Perceptron Learning Algorithm

- First neural network learning model in the 1960's
- Simple and limited (single layer model)
- Basic concepts are similar to multi-layer models

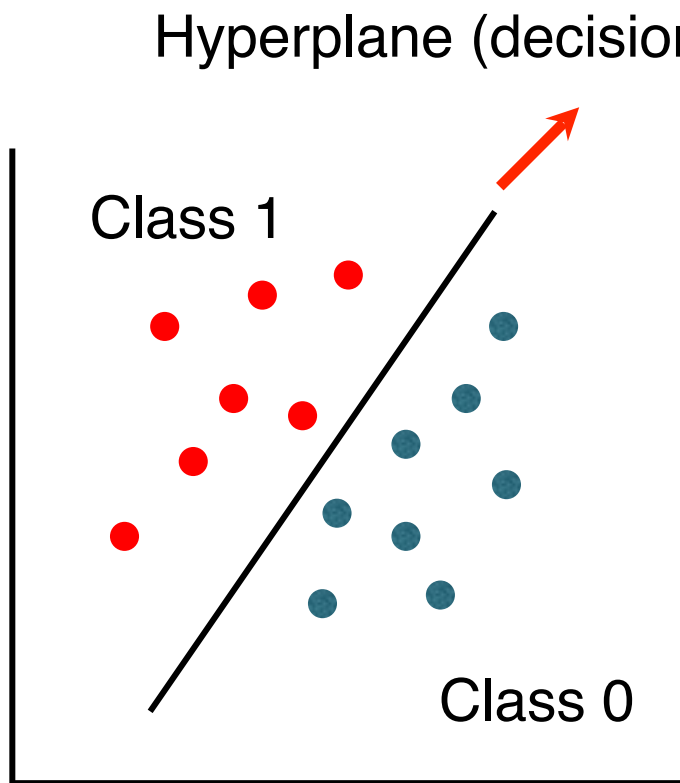
# What is Perceptron?

The goal of perceptron algorithm is to find a hyperplane that separates a set of data into two classes.



# What is Perceptron?

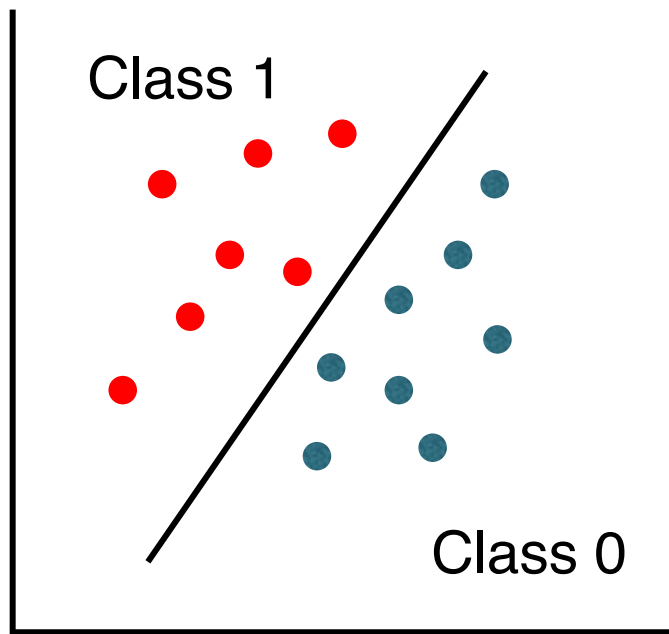
The goal of perceptron algorithm is to find a hyperplane that separates a set of data into two classes.



- Binary classifier
- Supervised learning



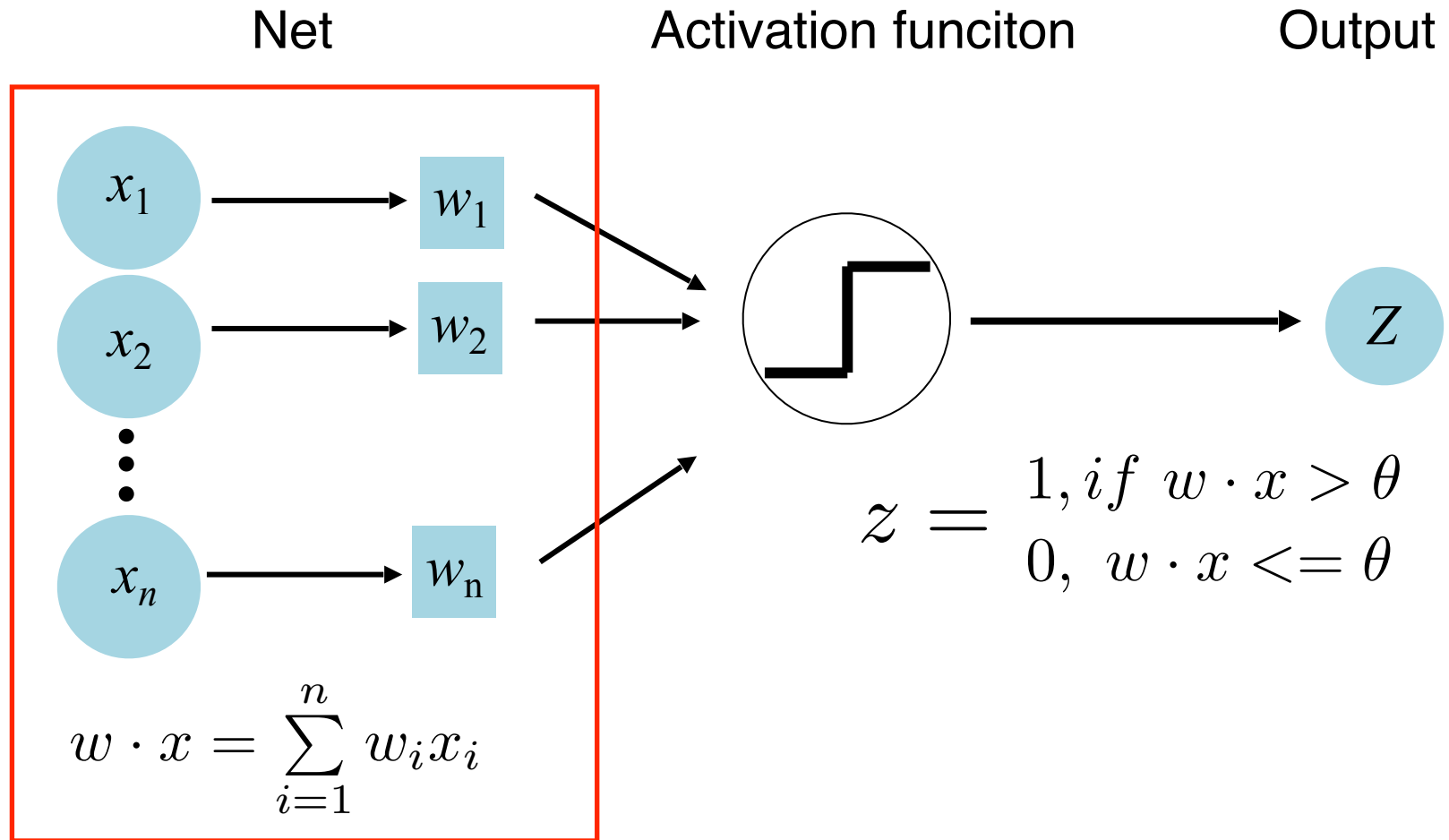
# Perceptron



bias term

$$f(x) = \begin{cases} 1, & \text{if } w \cdot x + \theta > 0 \\ 0, & \text{otherwise} \end{cases}$$
$$w \cdot x = \sum_{i=1}^n w_i x_i \quad (\text{dot product})$$

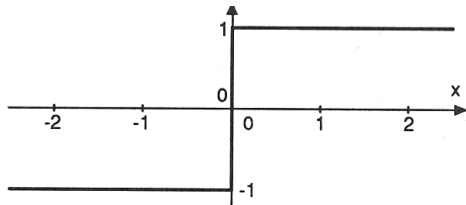
# Perceptron



- Learning weights such that an objective function is minimized

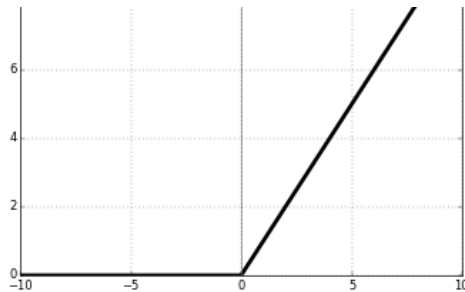
# Activation Function

Outputs the label given an input or a set of inputs.



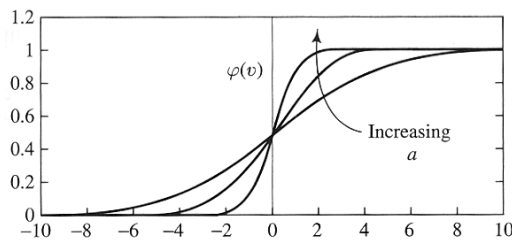
Step function

$$f(x) := \text{sgn}(x) = \begin{cases} 1 & \text{if } x \geq 0 \\ -1 & \text{if } x < 0 \end{cases}$$



ReLU (rectified linear unit)

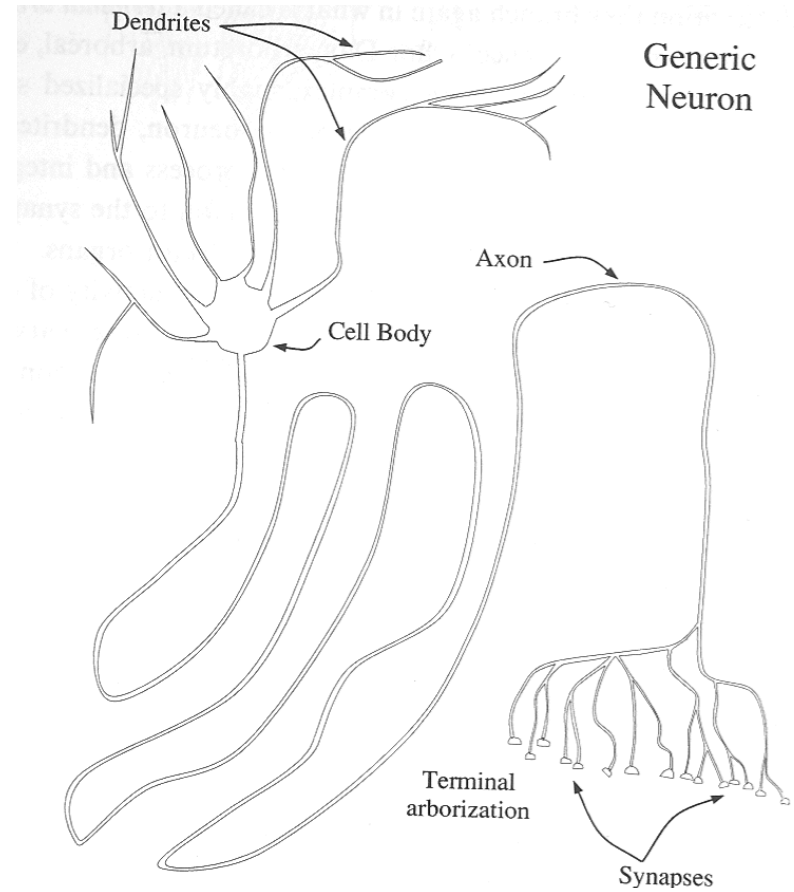
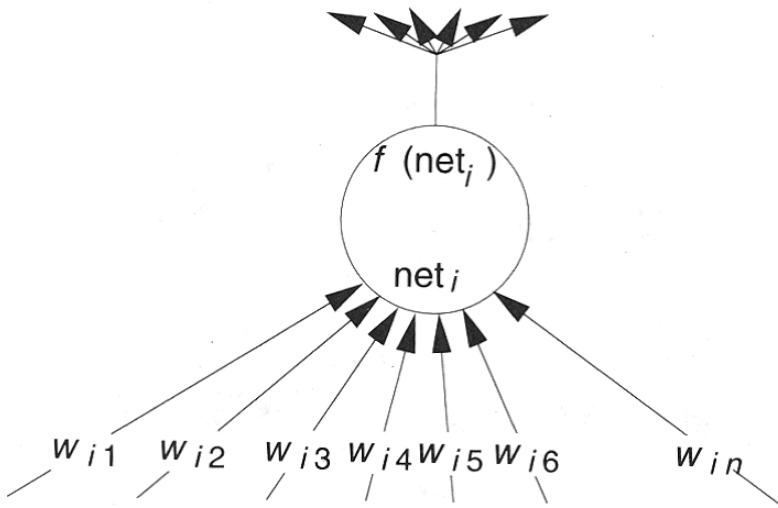
$$f(x) = \max(0, x)$$



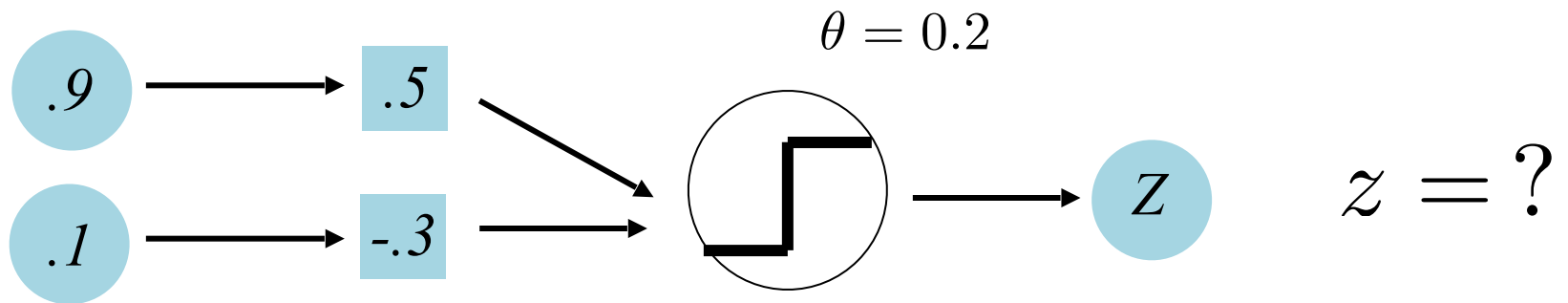
Sigmoid function

$$f(x) := \sigma(x) = \frac{1}{1 + e^{(-ax)}}$$

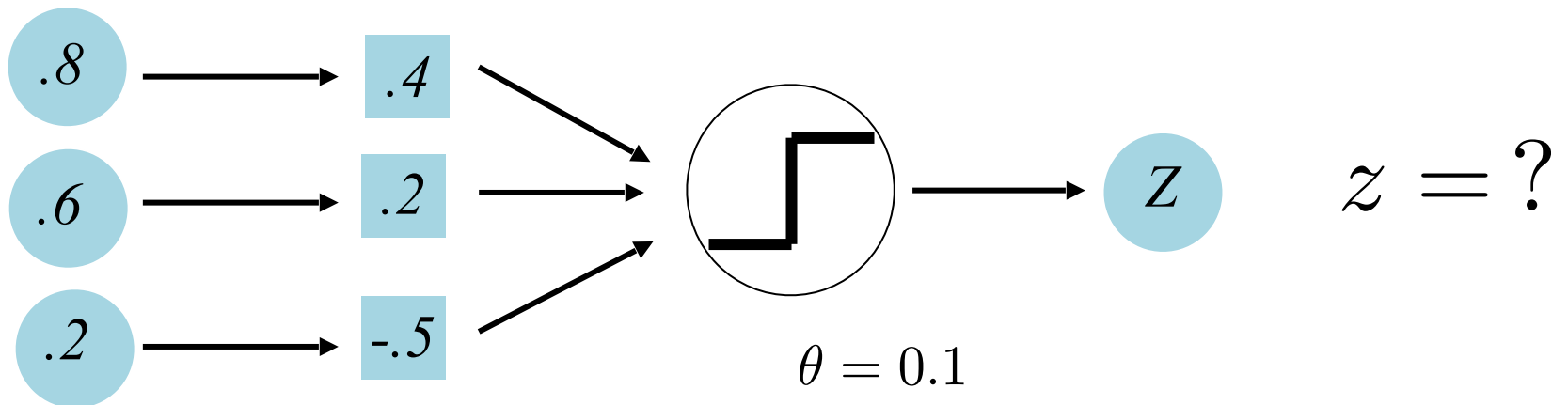
# Perceptron as a Single Layer Neuron



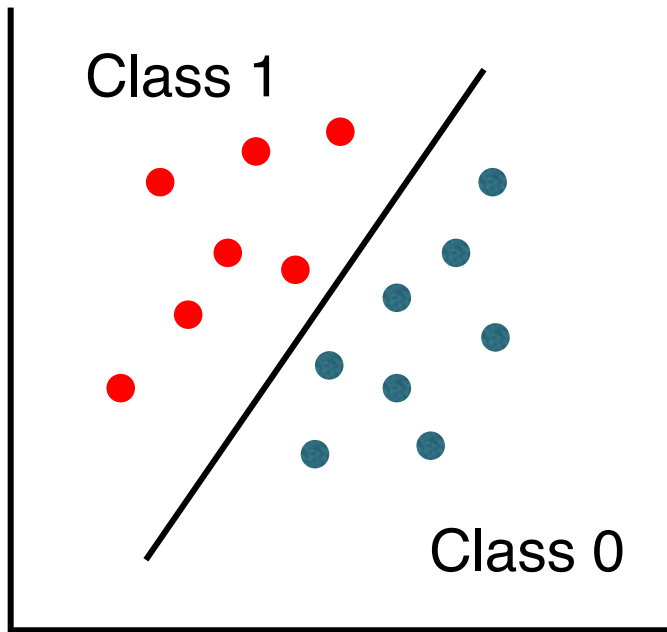
# Examples



$$z = \begin{cases} 1, & \text{if } w \cdot x > \theta \\ 0, & \text{if } w \cdot x \leq \theta \end{cases}$$



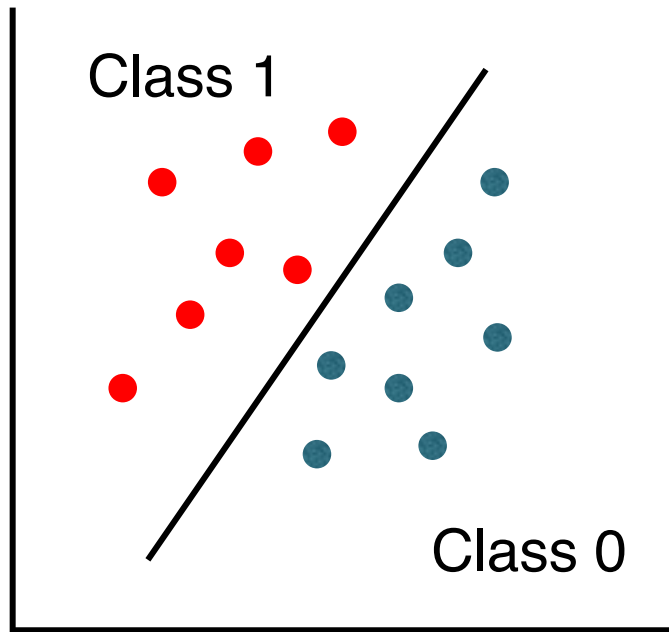
# How to Learn Perceptron?



$$f(x) = \begin{cases} 1, & \text{if } w \cdot x + \theta > 0 \\ 0, & \text{otherwise} \end{cases}$$

$w, \theta$  are unknown parameters

# How to Learn Perceptron?

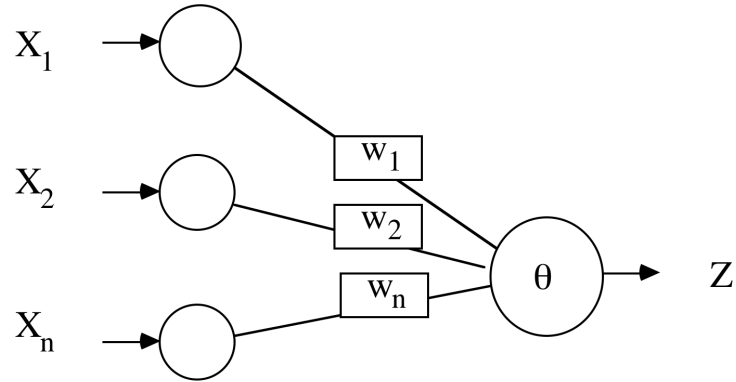


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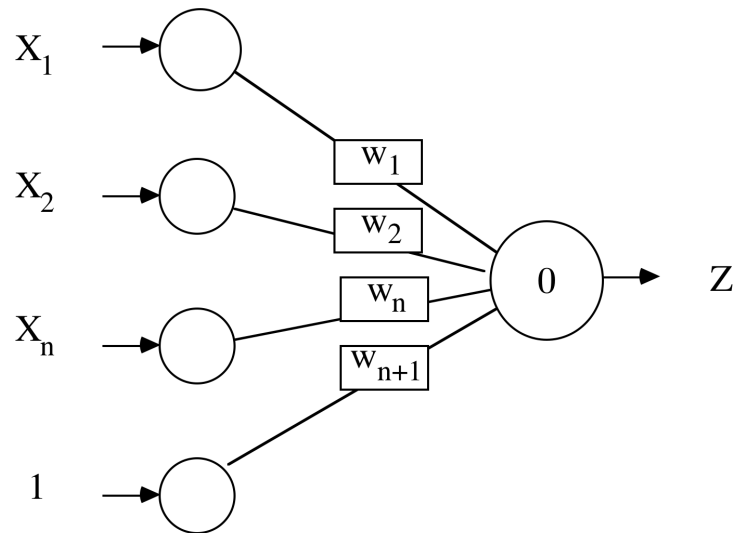
$w, \theta$  are unknown parameters

- In supervised learning the network has its output compared with known correct answers
  - Supervised learning
  - Learning with a teacher

## Weight Versus Threshold



Do you need to adjust Theta? Yes, in most cases



where  $w_{n+1} = -\theta$



# Perceptron Learning Rules

- Consider linearly separable problems
- How to find appropriate weights
- Look if the output result  $o$  belongs to the desired class has the desired value  $d$  (give labels)

$$w^{new} = w^{old} + \Delta w \quad \Delta w = \eta \sum_i (d - o)x_i$$

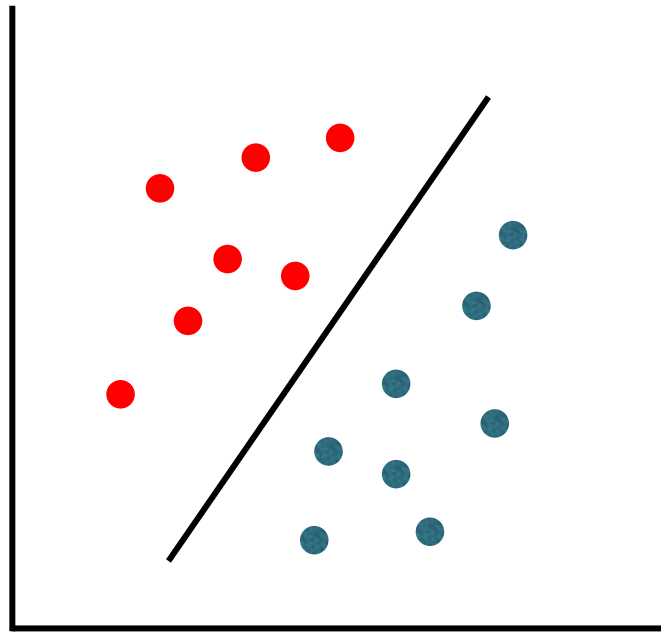
$\eta$  is called the **learning rate**, with  $0 < \eta \leq 1$

**Perceptron Convergence Theorem:** Guaranteed to find a solution in finite time if a solution exists

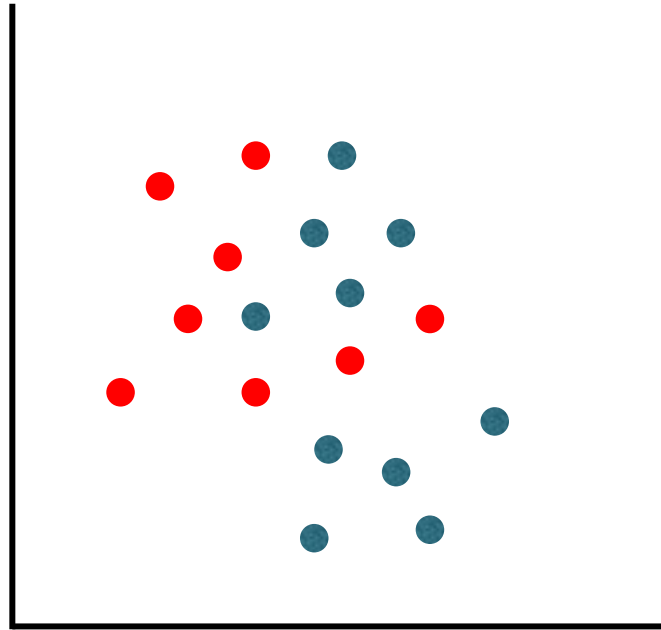
# Perceptron Learning Rules

- The algorithm converges to the correct classification *if and only if* the training data is linearly separable
- When assigning a value to  $\eta$  we must keep in mind two conflicting requirements
  - Averaging of past inputs to provide stable weights estimates, which **requires small  $\eta$**
  - Fast adaptation with respect to real changes in the underlying distribution, which **requires large  $\eta$**

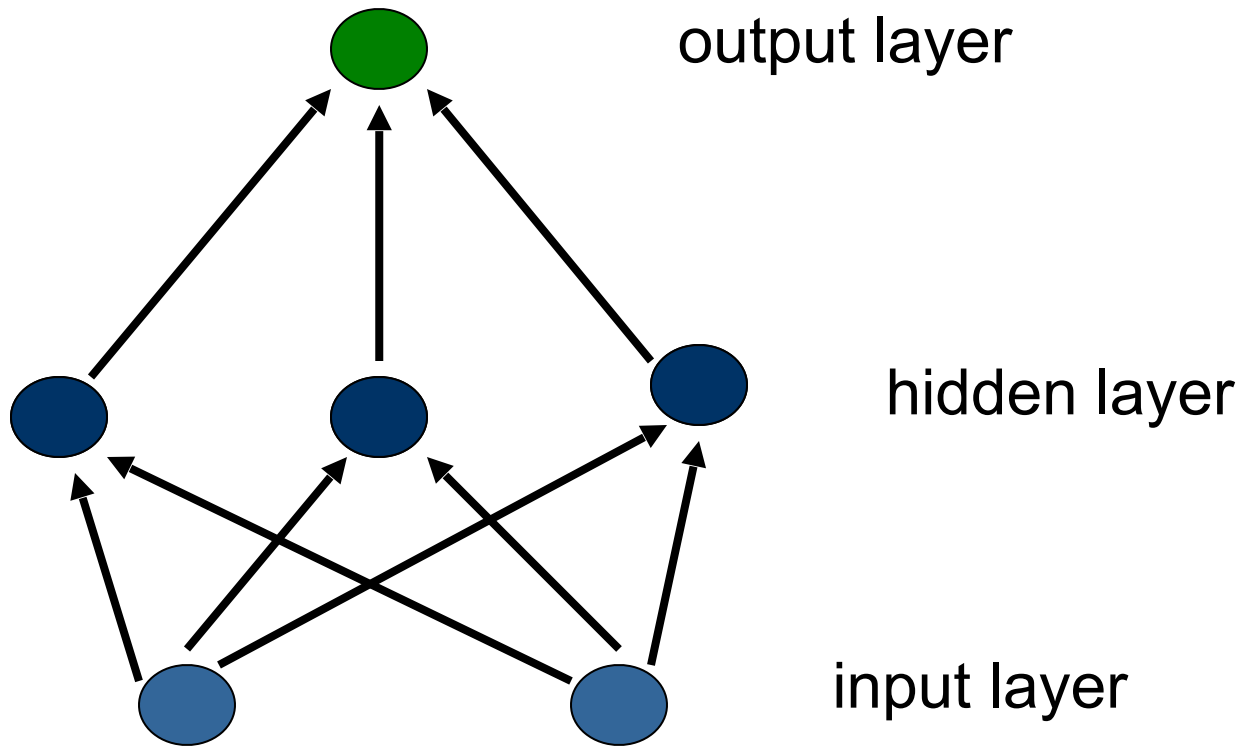
# Linear Separability



# Limited Functionality of Hyperplane



# Multilayer Network



$$o_1 = \text{sgn}\left(\sum_{i=0}^n w_{1i} x_i\right)$$

$$o_2 = \text{sgn}\left(\sum_{i=0}^n w_{2i} x_i\right)$$