### TRAINING DATA

for instance tuples of already labeled data: (genom, protein expression)

#### parameter tuning:

- optimal error rate
- but also good generalization to unknown data

### DATA TO BE CLASSIFIED

in our example of the form:

(genom1)

(genom2)

(genom3)

• • •

•••

### MACHINE LEARNING ALGORITHM

allows for certain parameters to be tuned according to training data

### **DATA CLASSIFICATION**

(genom1, expression1)

(genom2, expression2)

(genom3, expression3)

••

••

### **PREDICTION**

prediction of labels of the form

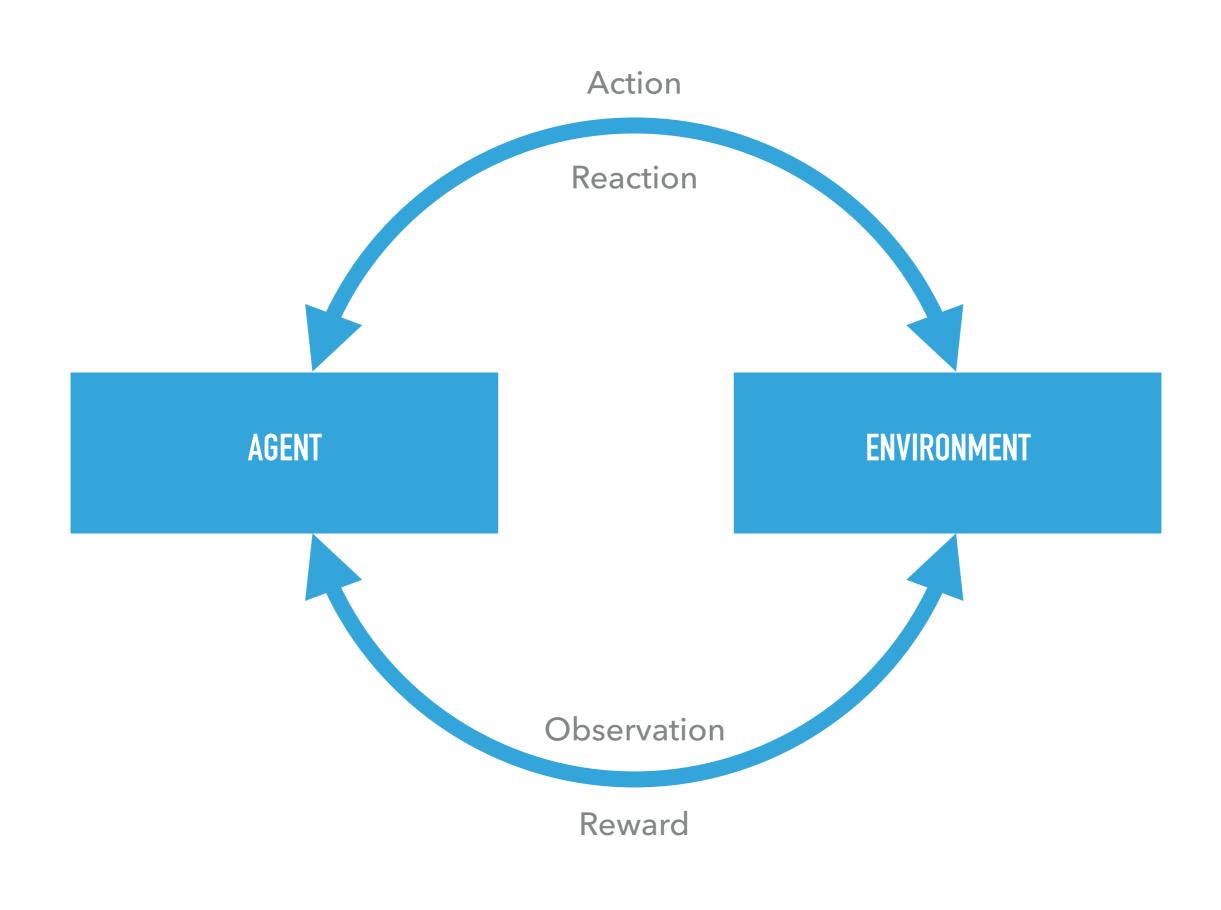
(expression1)

(expression2)

(expression3)

•••

. . .



# weighted sum

## activation function

$$\mathbb{R}^{n+1} \ni x = \begin{pmatrix} 1 \\ x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \longrightarrow \begin{pmatrix} \sigma(w \cdot x) \\ \vdots \\ x_n \end{pmatrix} \quad y \in \{-1, +1\}$$

### weight vector

$$w \in \mathbb{R}^{n+1}$$

## signum function

$$\sigma(z) := \begin{cases} +1 & \text{for } z \ge 0 \\ -1 & \text{for } z < 0 \end{cases}$$

Input signals

weighted sum

activation function

# feedback for update rule

## weight vector

$$w \in \mathbb{R}^{n+1}$$

## signum function

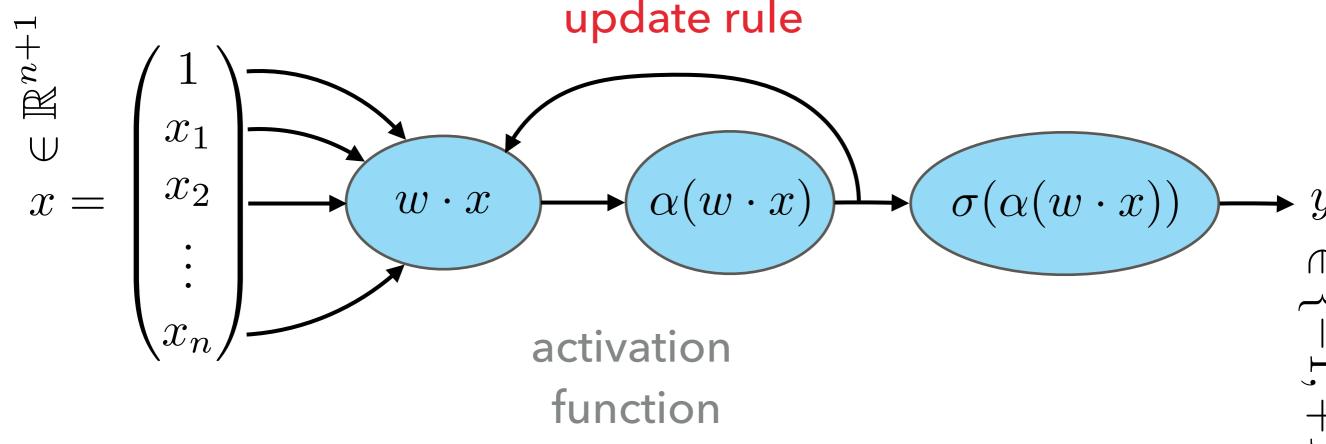
$$\sigma(z) := \begin{cases} +1 & \text{for } z \ge 0 \\ -1 & \text{for } z < 0 \end{cases}$$

Input signals

weighted sum

activation function





 $\alpha: \mathbb{R} \to \mathbb{R}$ 

weight vector

$$w \in \mathbb{R}^{n+1}$$

signum function

$$\sigma(z) := \begin{cases} +1 & \text{for } z \ge 0 \\ -1 & \text{for } z < 0 \end{cases}$$

