

Machine Learning

3. classificatie en logistische regressie



ml: classificatie

binaire classificatie

- kanker kwaadaardig?
- e-mail spam?
- online transactie betrouwbaar?
- bloemwilg virginica?
- is dit Henk Tattje?



multi-class classificatie



- wat voor soort insect?
- welk cijfer staat op dit plaatje?
- wie staat er op deze foto?
- wat voor soort muziekstuk is dit?

classificatie van classificatie-algoritmen

- Linear classifiers

 - Fisher's linear discriminant

 - Logistic regression

 - Naive Bayes classifier

 - Perceptron

- Support vector machines

 - Least squares support vector machines

- Quadratic classifiers

- Kernel estimation

 - k-nearest neighbor

- Boosting (meta-algorithm)

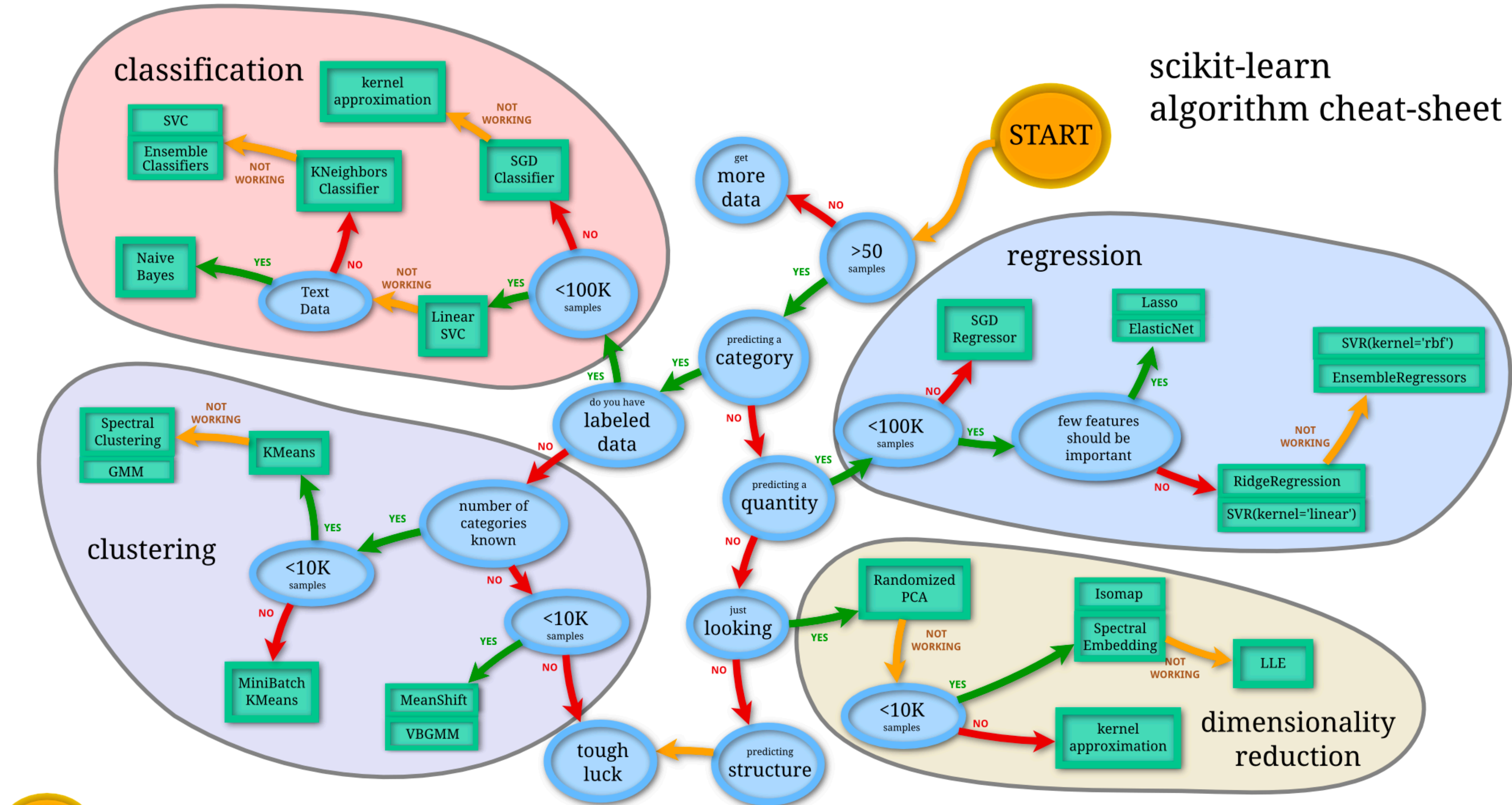
- Decision trees

 - Random forests

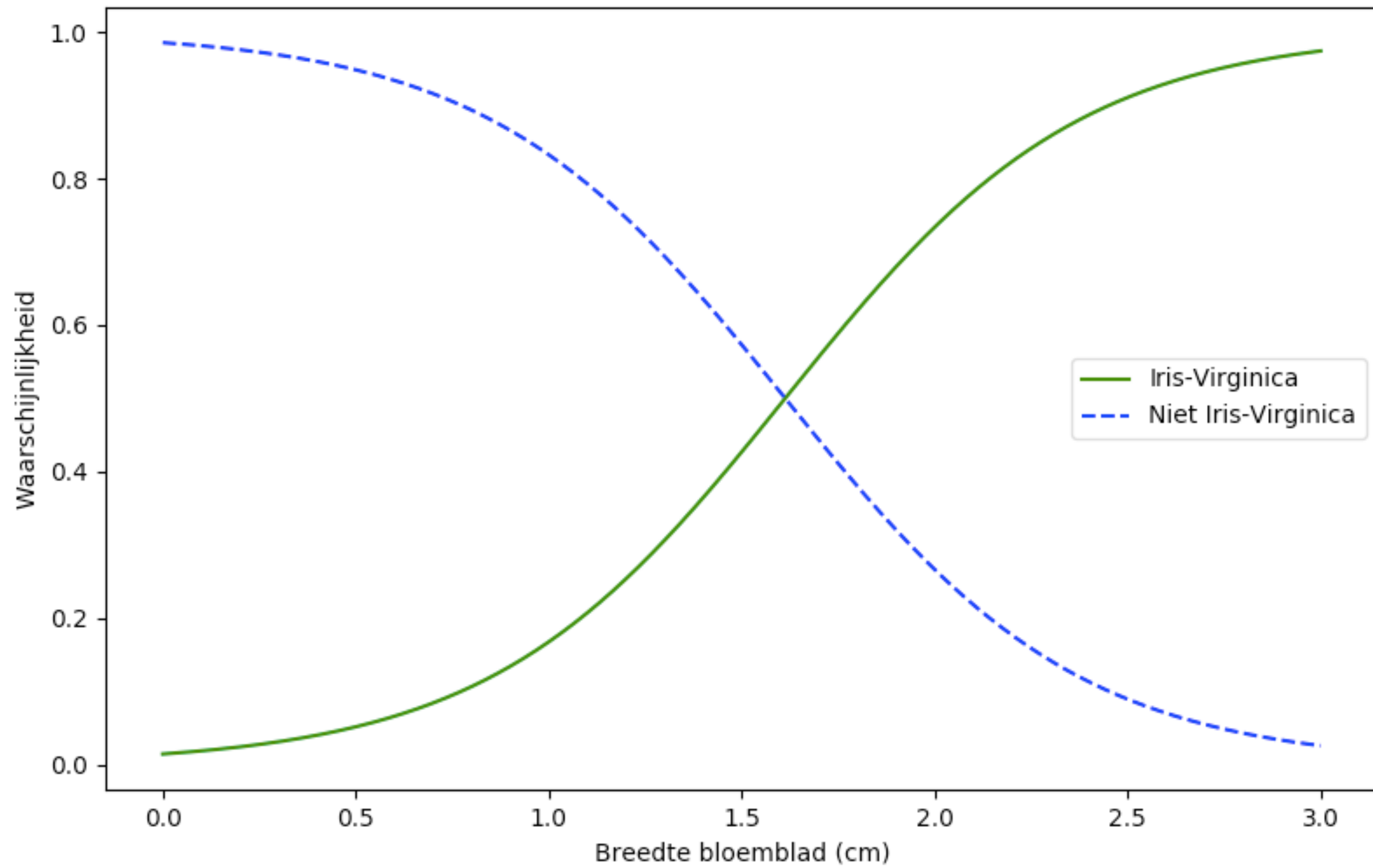
- Neural networks

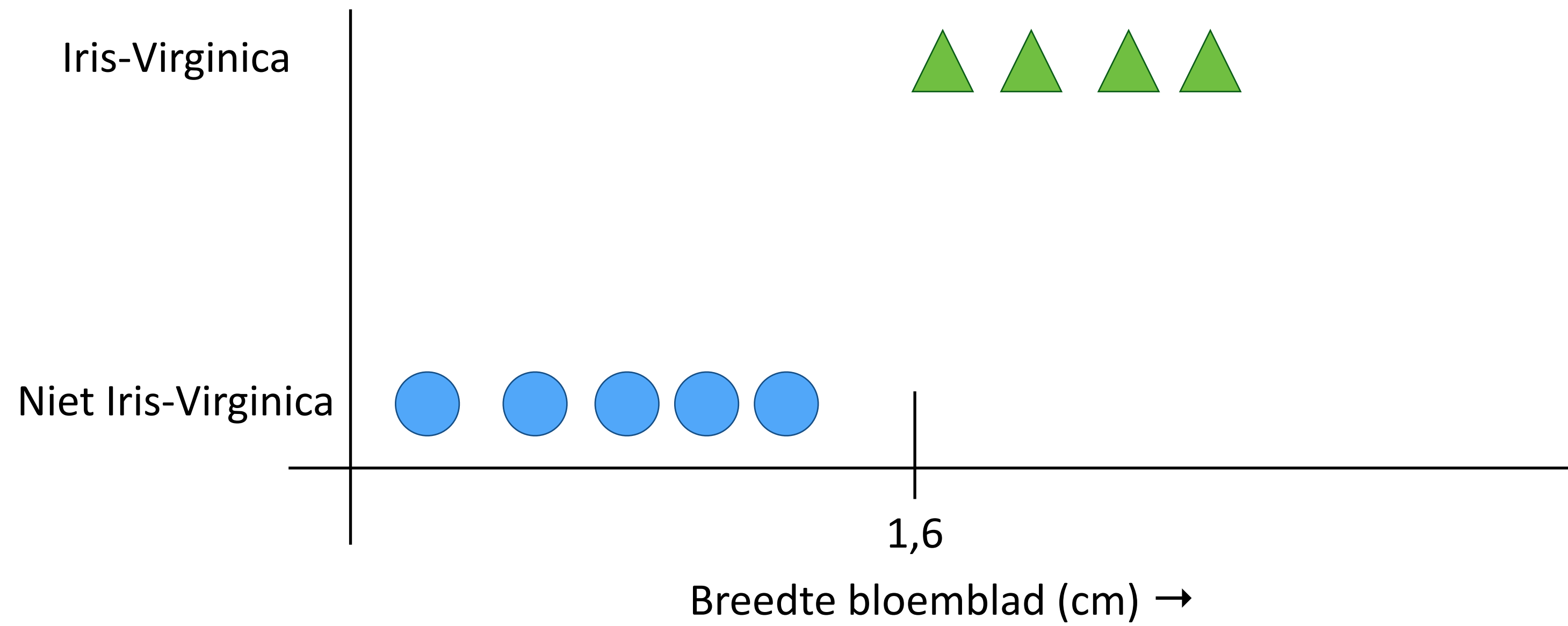
- Learning vector quantization

scikit-learn
algorithm cheat-sheet









Binaire classificatie

$$y \in \{0,1\}$$

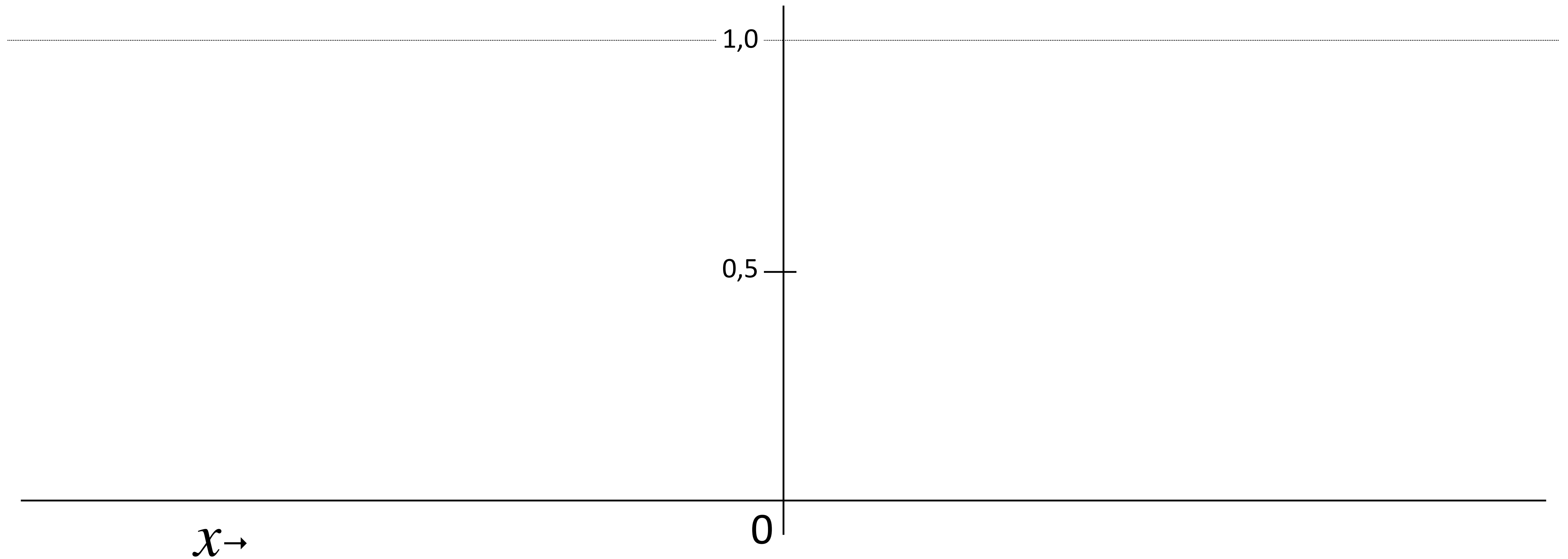
$$0 \leq h_{\theta}(x) \leq 1$$

Lineaire regressie

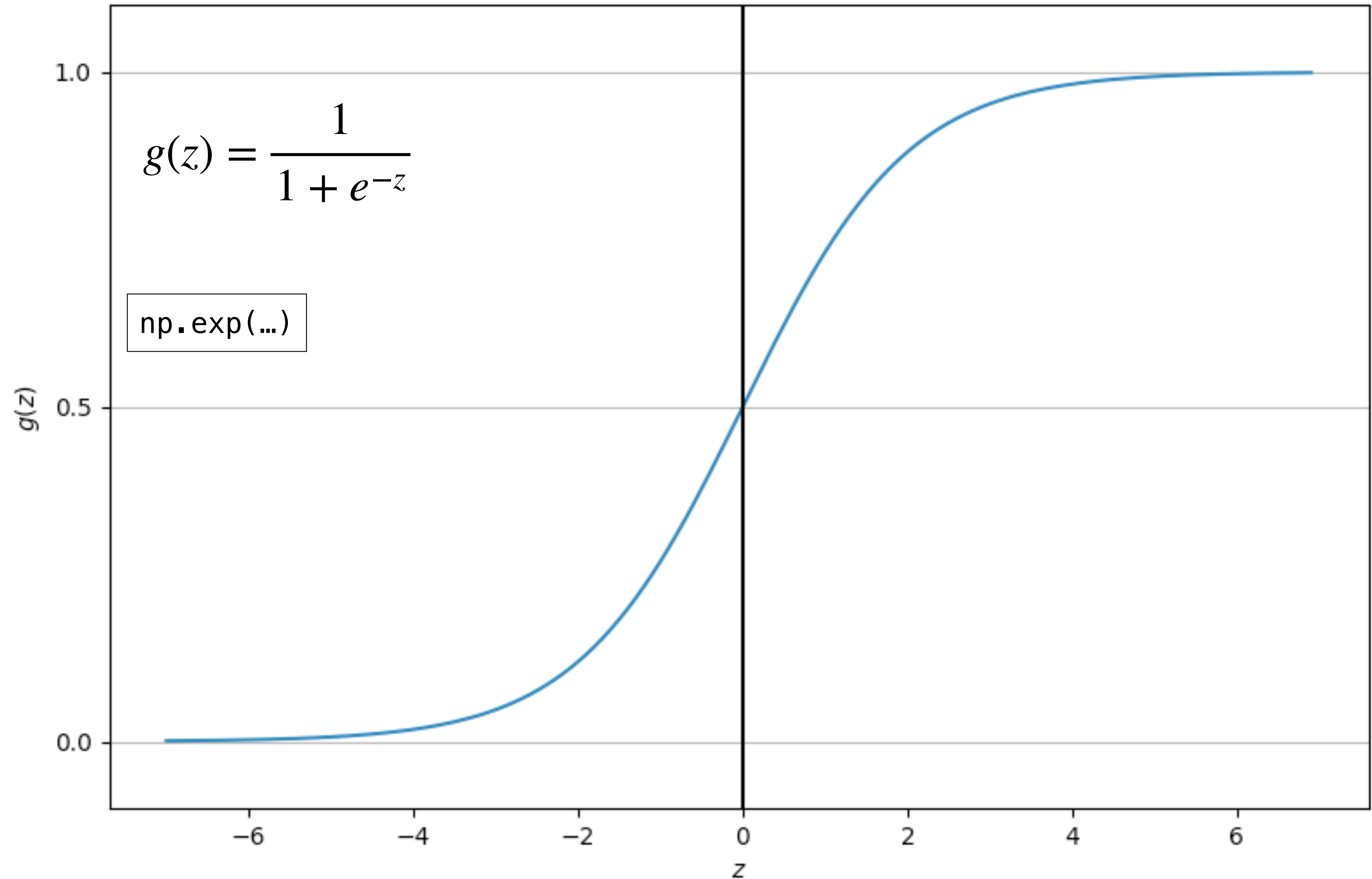
$$h_{\theta}(x) = \theta^T x$$

logistische regressie

$$0 \leq h_{\theta}(x) \leq 1$$



sigmoïdefunctie

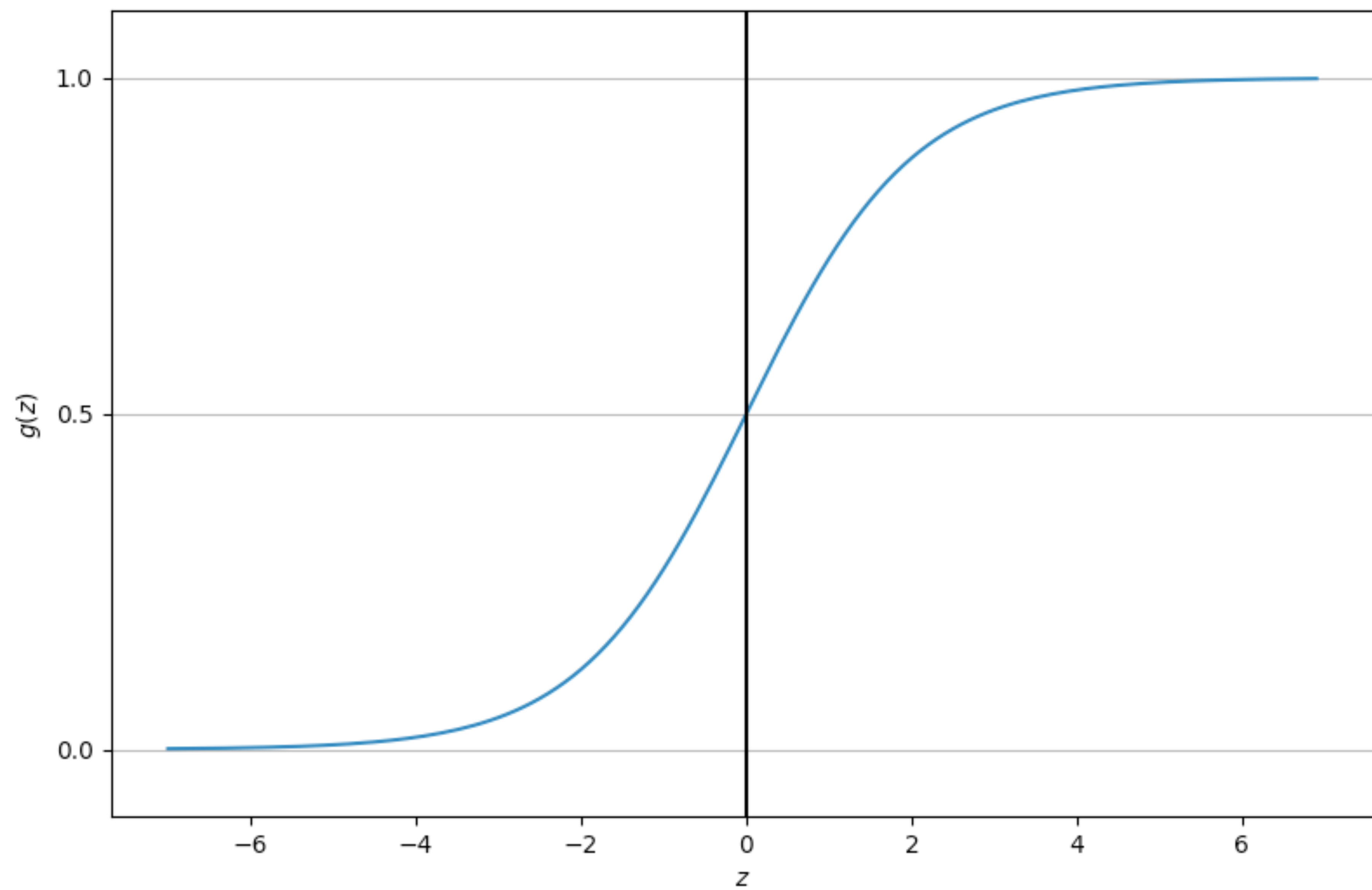


Wat betekent die voorspelling?

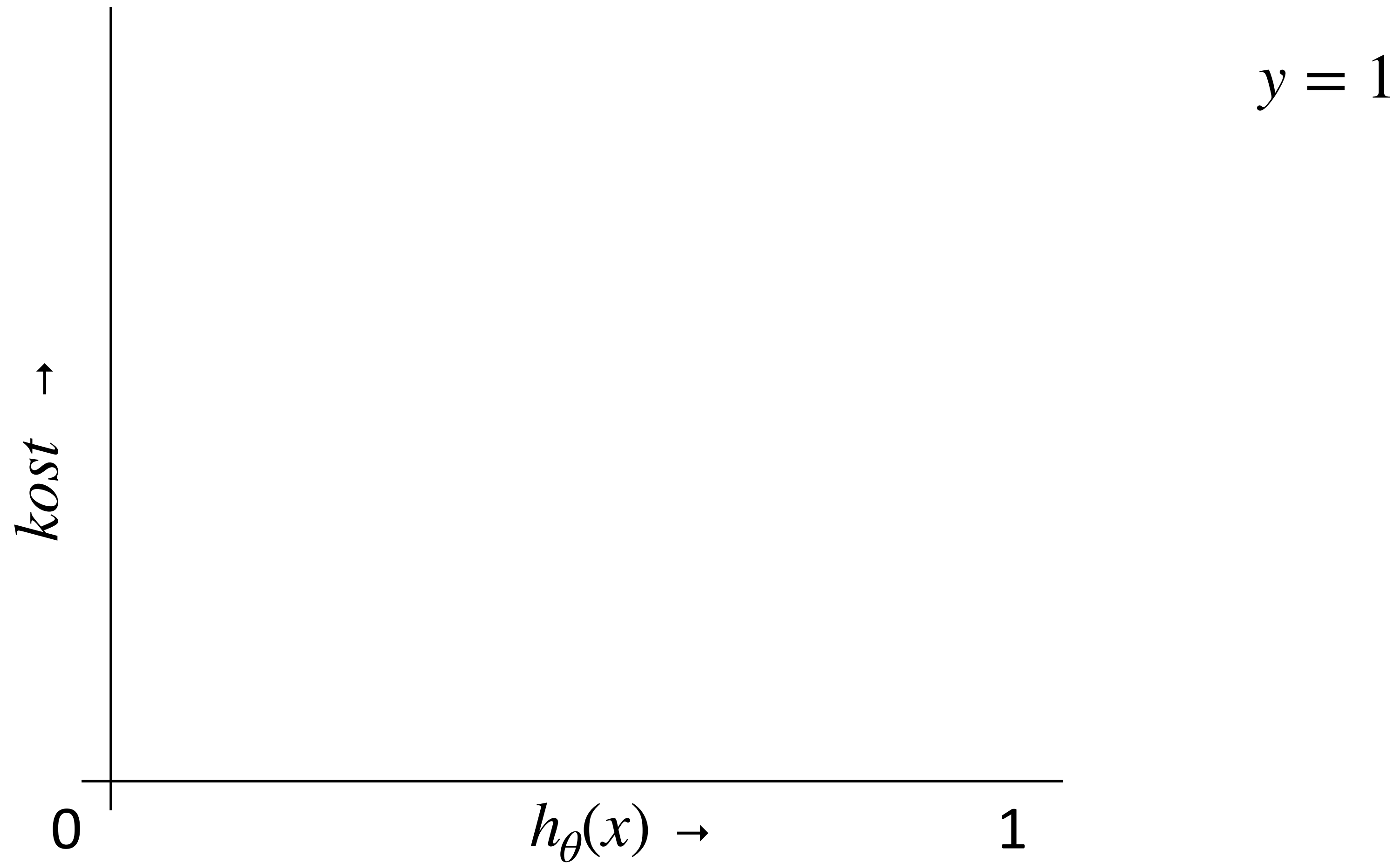
$$X = \begin{bmatrix} 1 & 1,6 \\ 1 & 2,5 \end{bmatrix}$$

$$\theta = \begin{bmatrix} 0,3 \\ 0,125 \end{bmatrix}$$

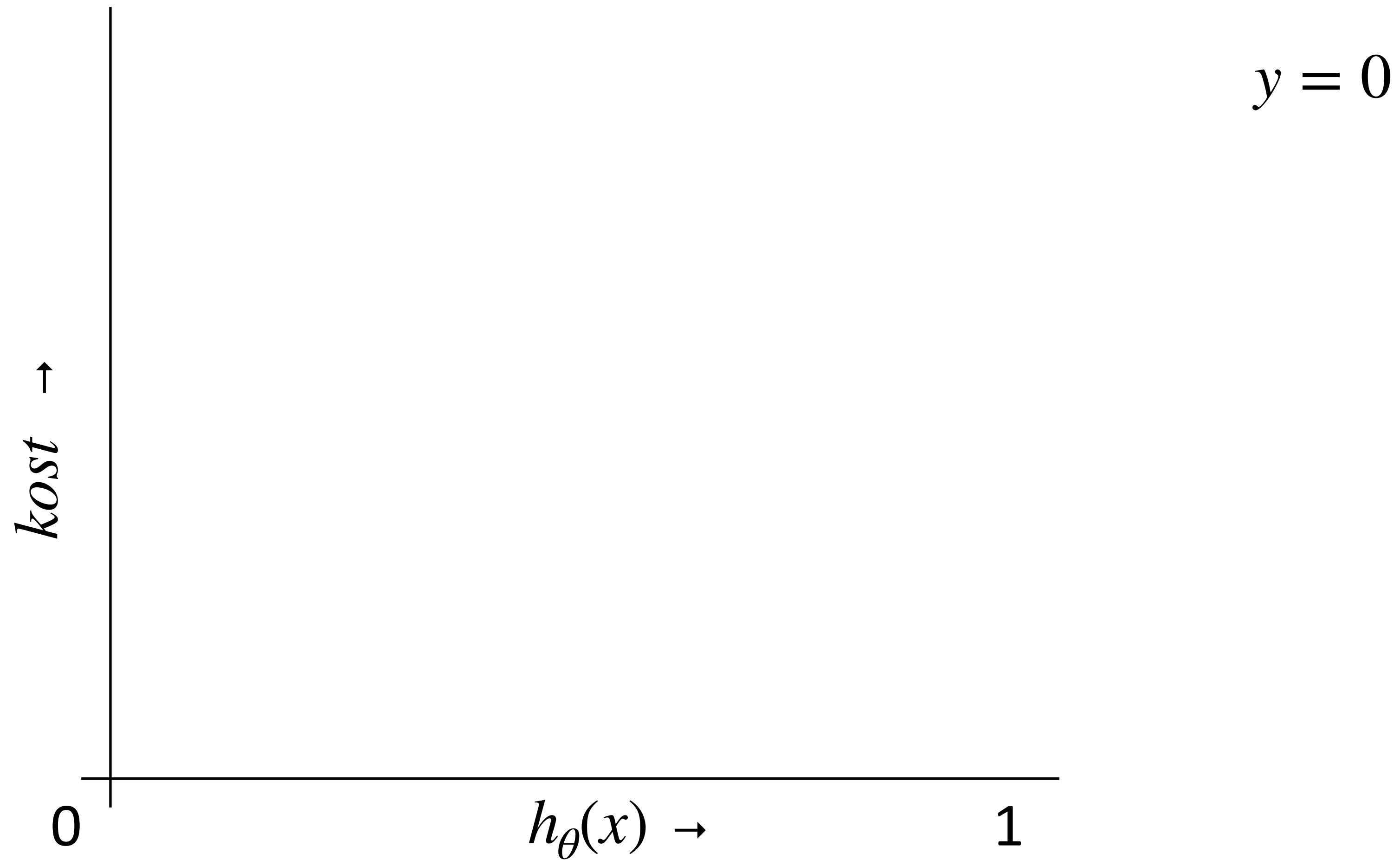
ml: classificatie kostenfunctie



$$Kost = \begin{cases} -\log(h_{\theta}(x)) & \text{als } y = 1 \\ -\log(1 - h_{\theta}(x)) & \text{als } y = 0 \end{cases}$$



$$Kost = \begin{cases} -\log(h_{\theta}(x)) & \text{als } y = 1 \\ -\log(1 - h_{\theta}(x)) & \text{als } y = 0 \end{cases}$$



$$\begin{aligned}
Kost &= \begin{cases} -\log(h_{\theta}(x)) & \text{als } y = 1 \\ -\log(1 - h_{\theta}(x)) & \text{als } y = 0 \end{cases} \\
&= -y\log(h_{\theta}(x)) - (1 - y)\log(1 - h_{\theta}(x)) \\
&= y\log(h_{\theta}(x)) + (1 - y)\log(1 - h_{\theta}(x))
\end{aligned}$$

Kostenfunctie voor logistische regressie

$$J(\theta) = \frac{1}{m} \sum_{i=1}^m \left[y^{(i)} \log(h_{\theta}(x^{(i)})) + (1 - y^{(i)}) \log(1 - h_{\theta}(x^{(i)})) \right]$$

Dat ziet er indrukwekkend uit!

Maar de partiële afgeleide ervan ziet er juist weer heel vertrouwd uit:

$$\frac{\partial}{\partial \theta_j} J(\theta) = (h_{\theta}(x^{(i)}) - y^{(i)}) x_j^{(i)}$$

Gradient Descent bij logistische regressie

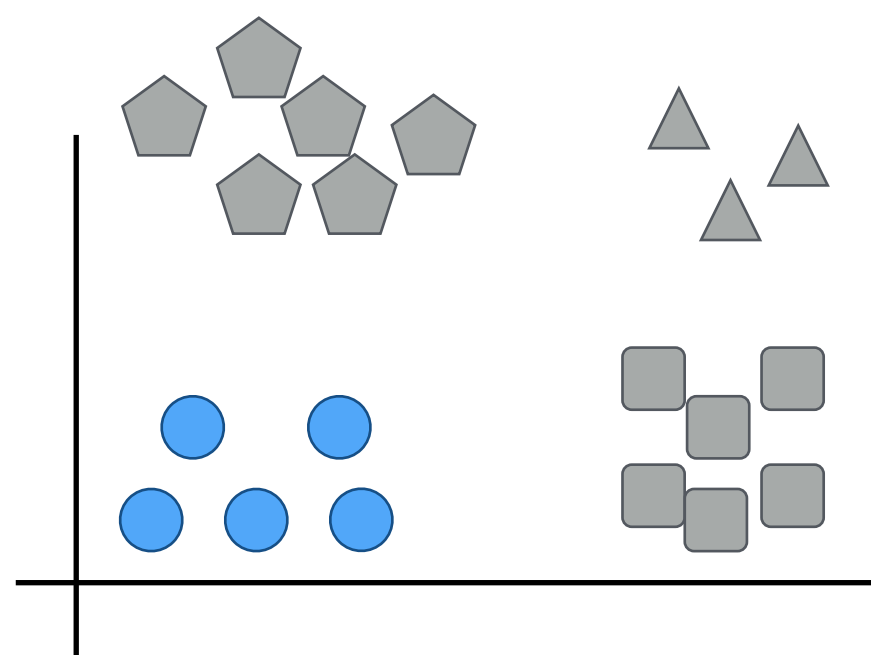
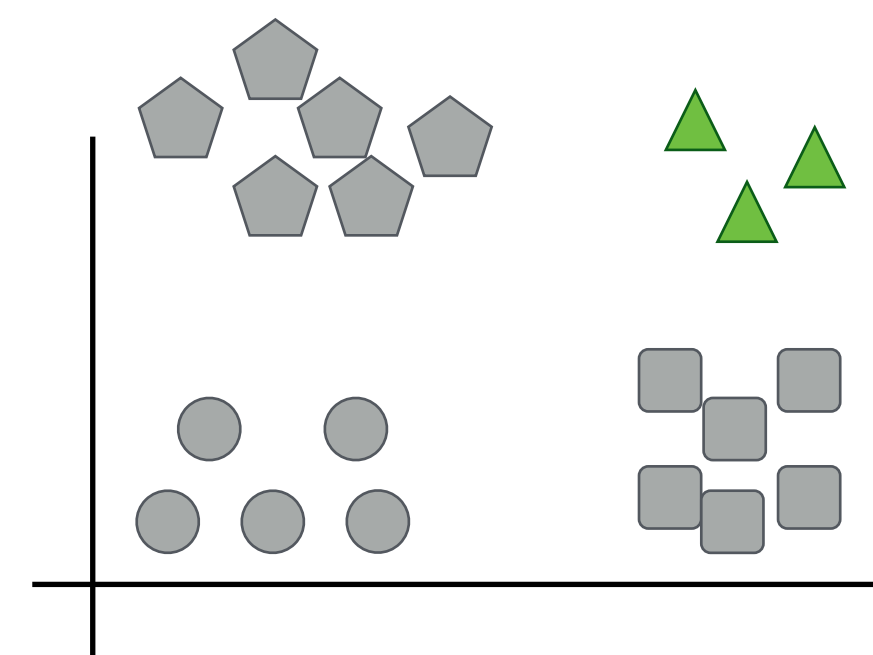
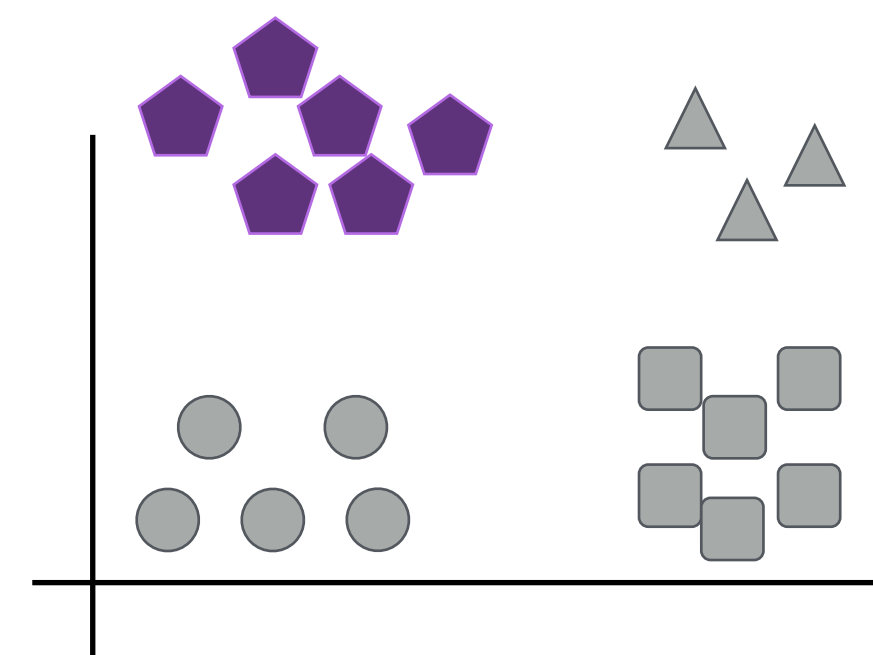
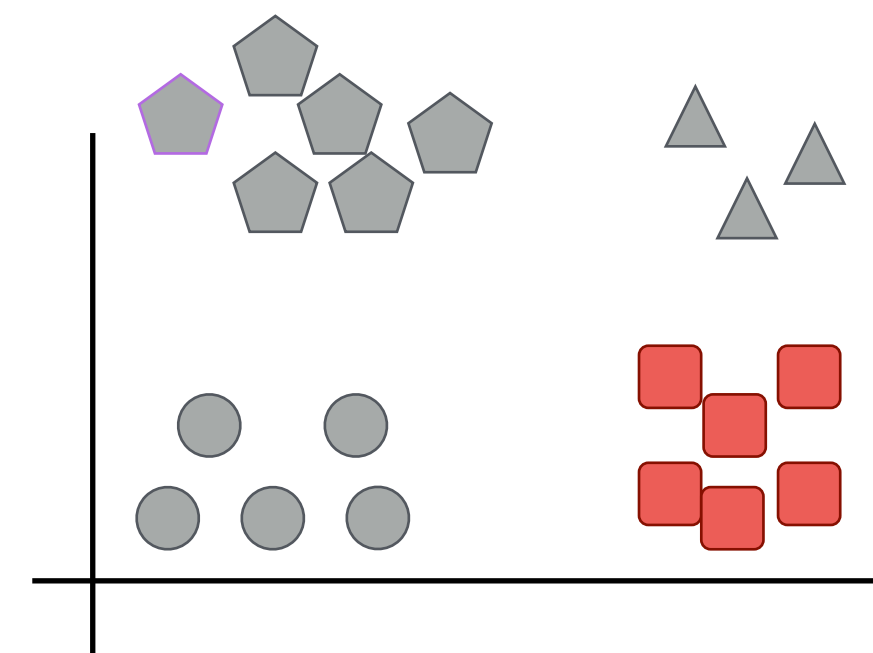
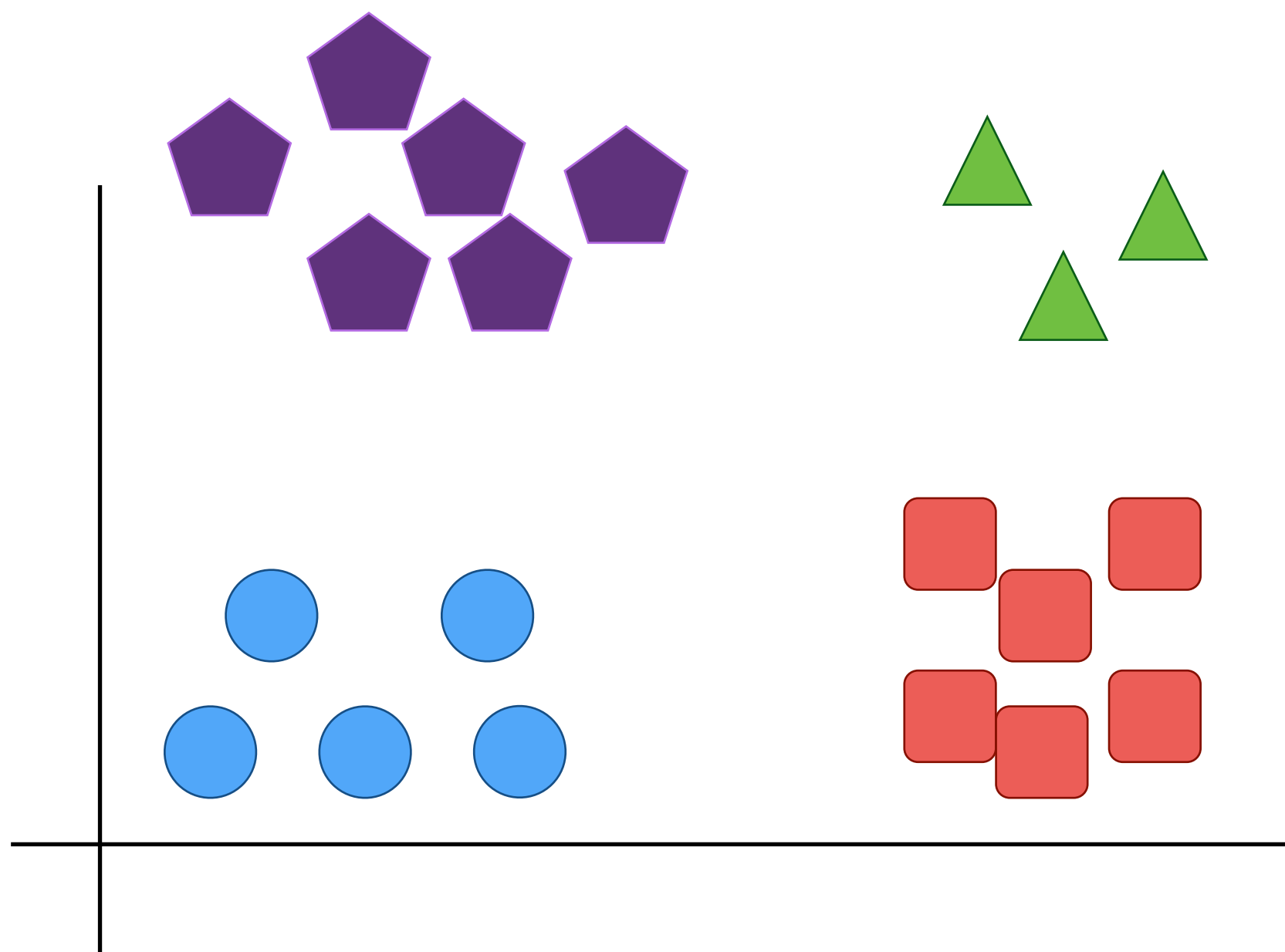
update alle $\theta_j, j = 1, j = 2, \dots, j = n$

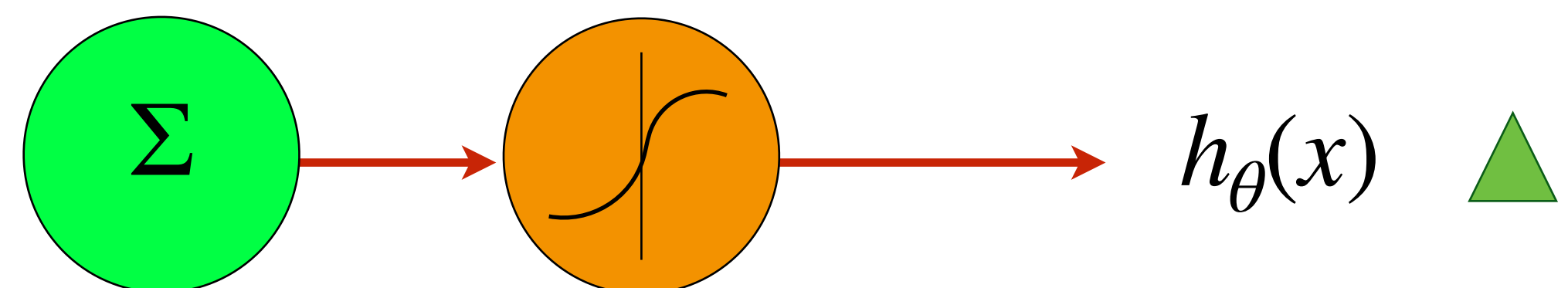
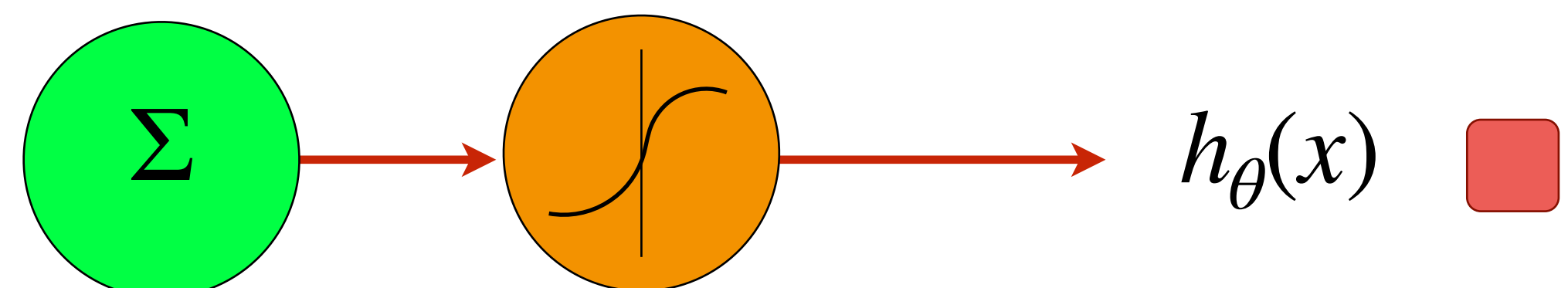
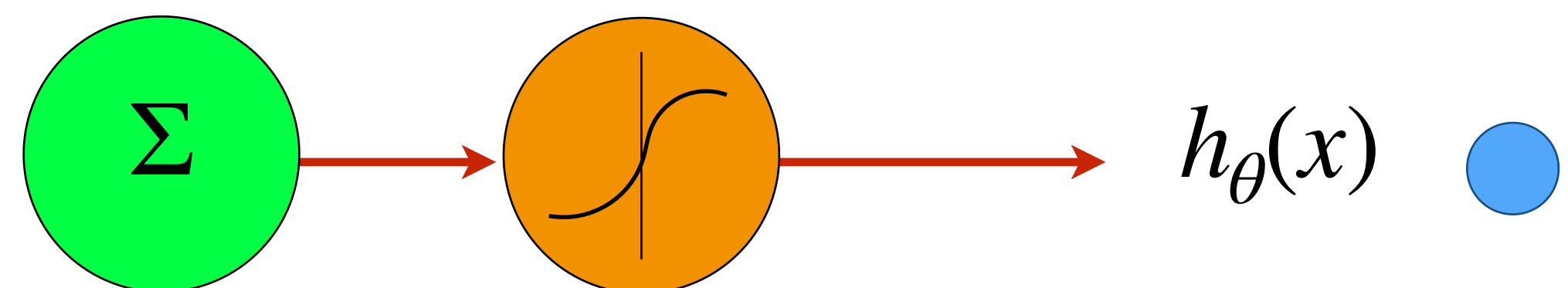
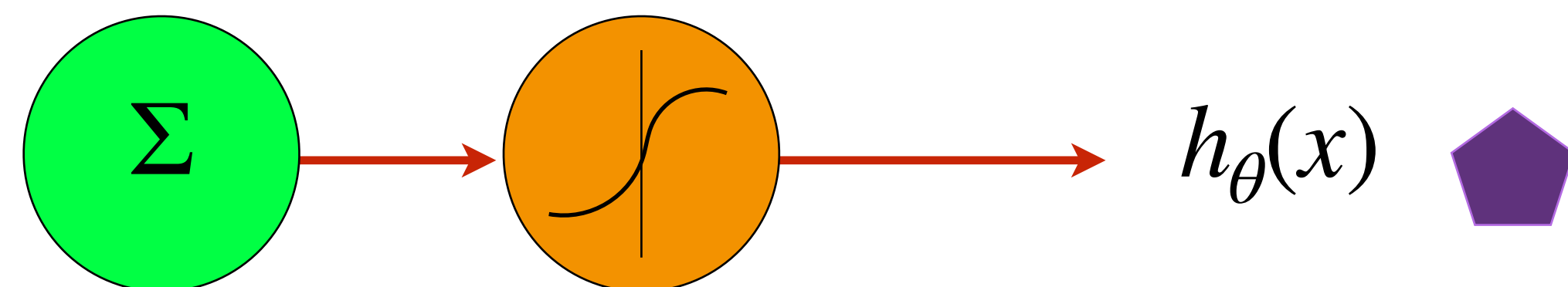
herhaal totdat een minimum bereikt is:

$$\theta_j := \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta)$$

$$:= \theta_j - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) x_j^{(i)}$$

`ml:multiclass classification`





$$\begin{bmatrix} 0.213 \\ 0.423 \\ 0.786 \\ 0.143 \end{bmatrix}$$

$$\begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$$