

# Neural networks

Feedforward neural network - activation function

# ARTIFICIAL NEURON

**Topics:** connection weights, bias, activation function

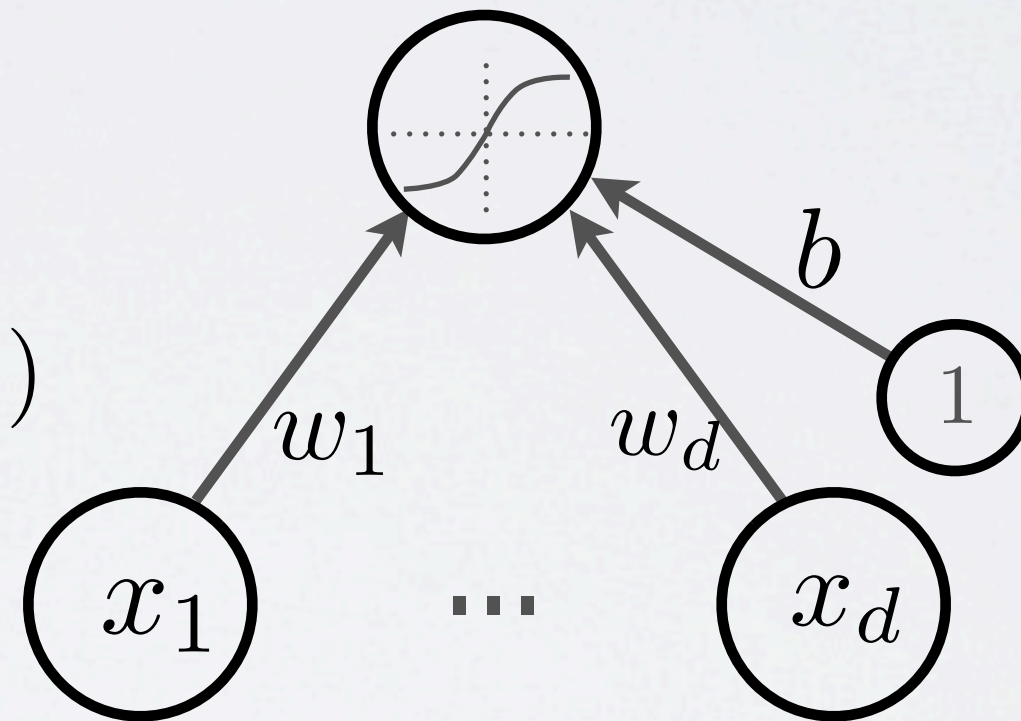
- Neuron pre-activation (or input activation):

$$a(\mathbf{x}) = b + \sum_i w_i x_i = b + \mathbf{w}^\top \mathbf{x}$$

- Neuron (output) activation

$$h(\mathbf{x}) = g(a(\mathbf{x})) = g(b + \sum_i w_i x_i)$$

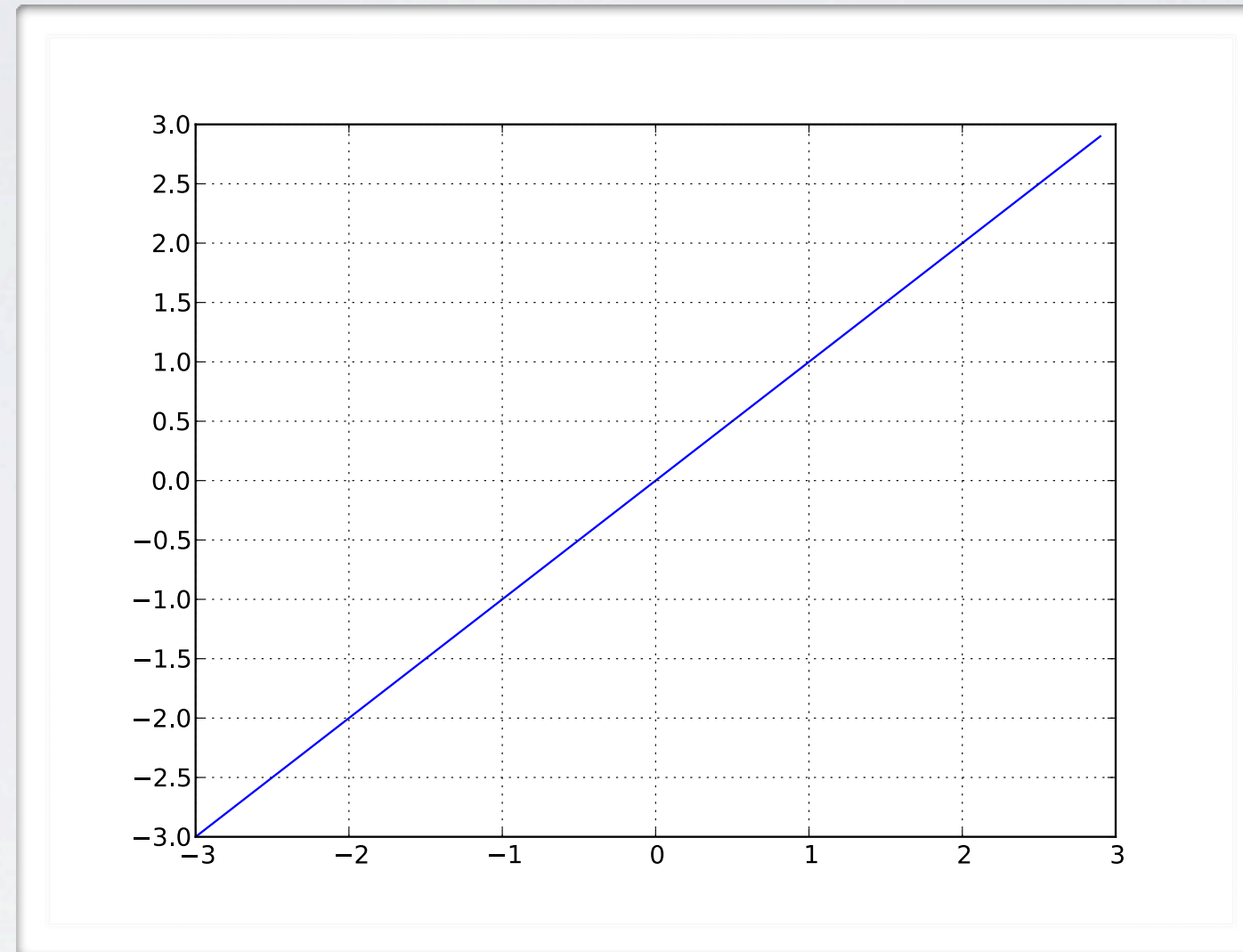
- $\mathbf{w}$  are the connection weights
- $b$  is the neuron bias
- $g(\cdot)$  is called the activation function



# ACTIVATION FUNCTION

**Topics:** linear activation function

- Performs no input squashing
- Not very interesting...



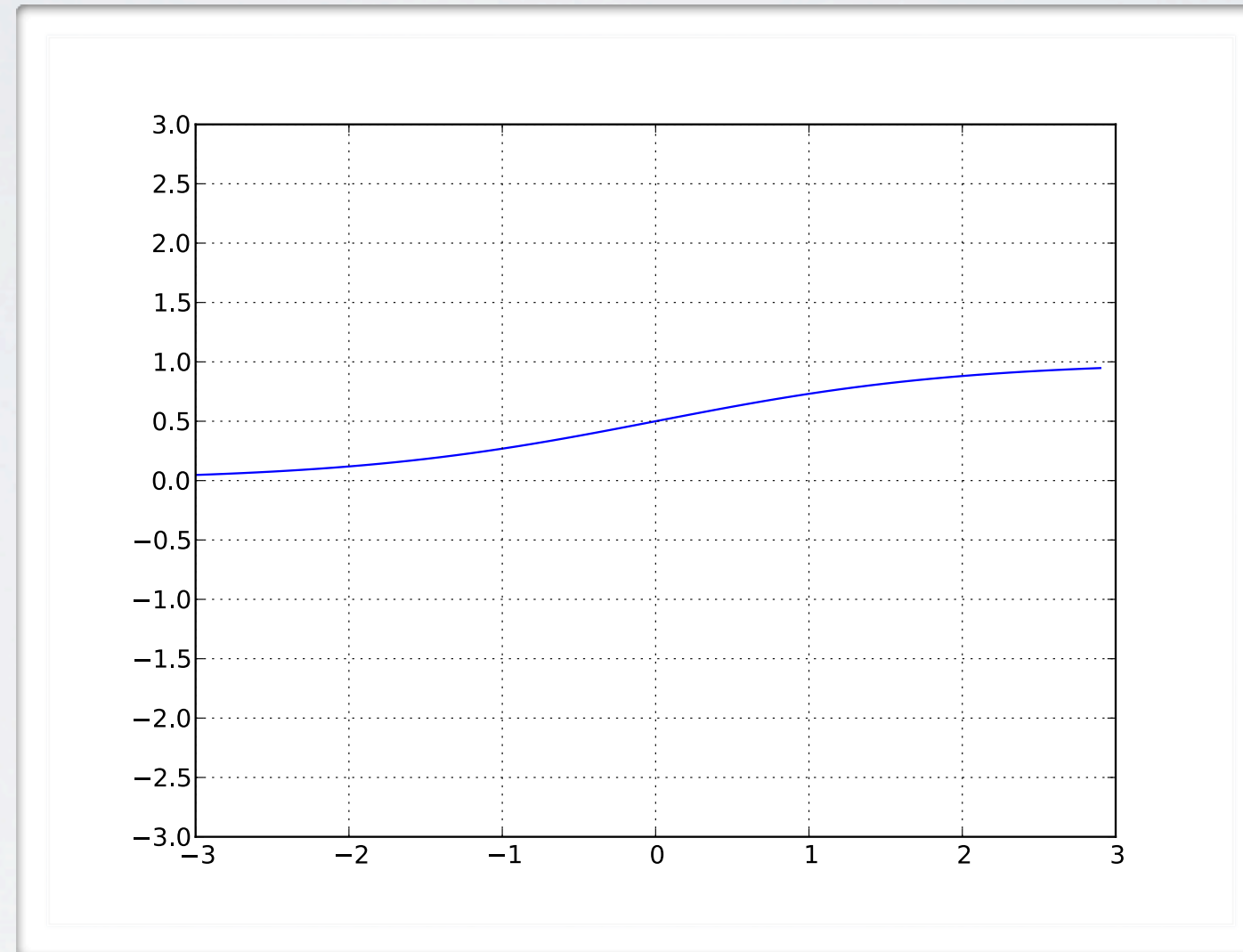
$$g(a) = a$$



# ACTIVATION FUNCTION

**Topics:** sigmoid activation function

- Squashes the neuron's pre-activation between 0 and 1
- Always positive
- Bounded
- Strictly increasing

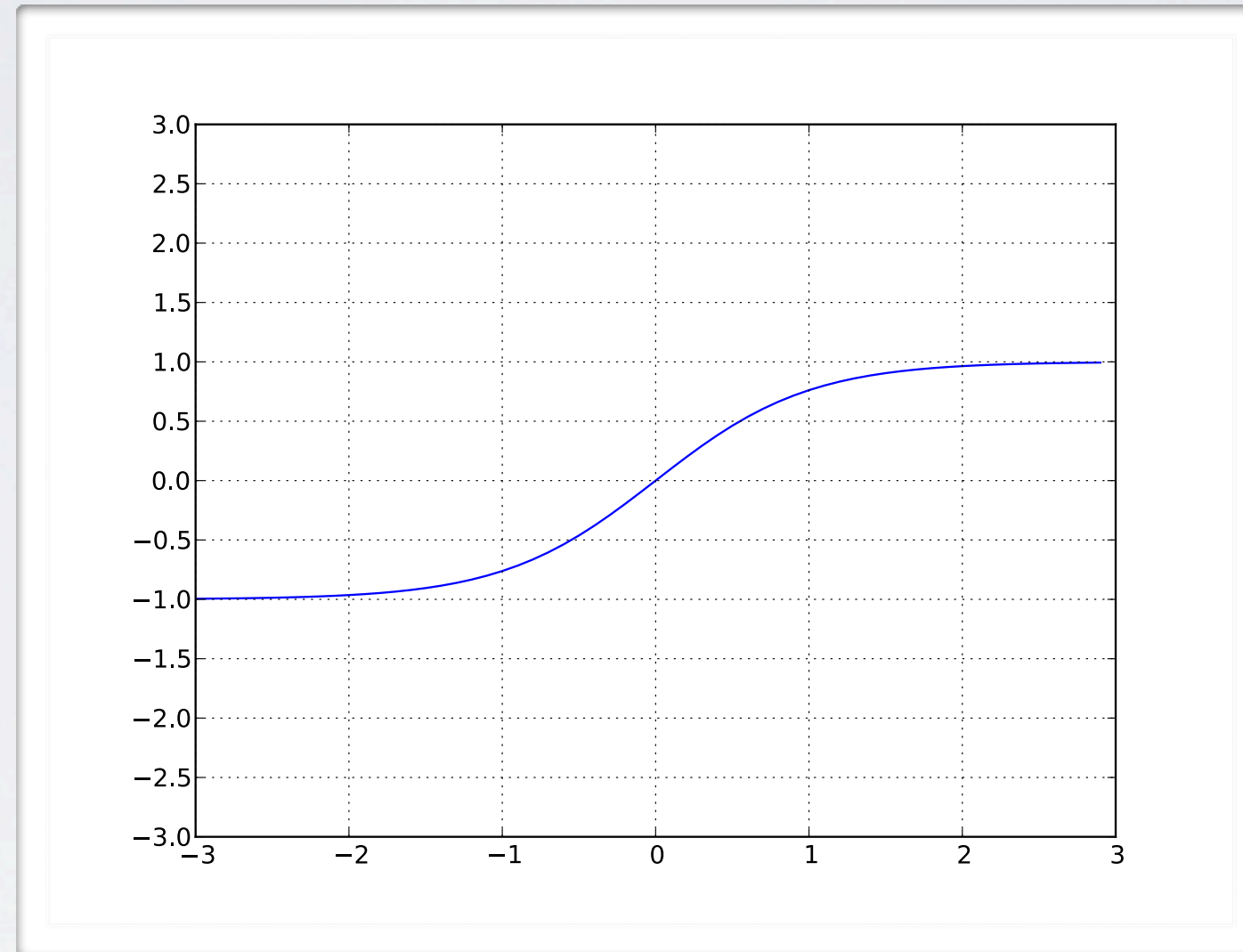


$$g(a) = \text{sigm}(a) = \frac{1}{1 + \exp(-a)}$$

# ACTIVATION FUNCTION

**Topics:** hyperbolic tangent (“tanh”) activation function

- Squashes the neuron’s pre-activation between -1 and 1
- Can be positive or negative
- Bounded
- Strictly increasing

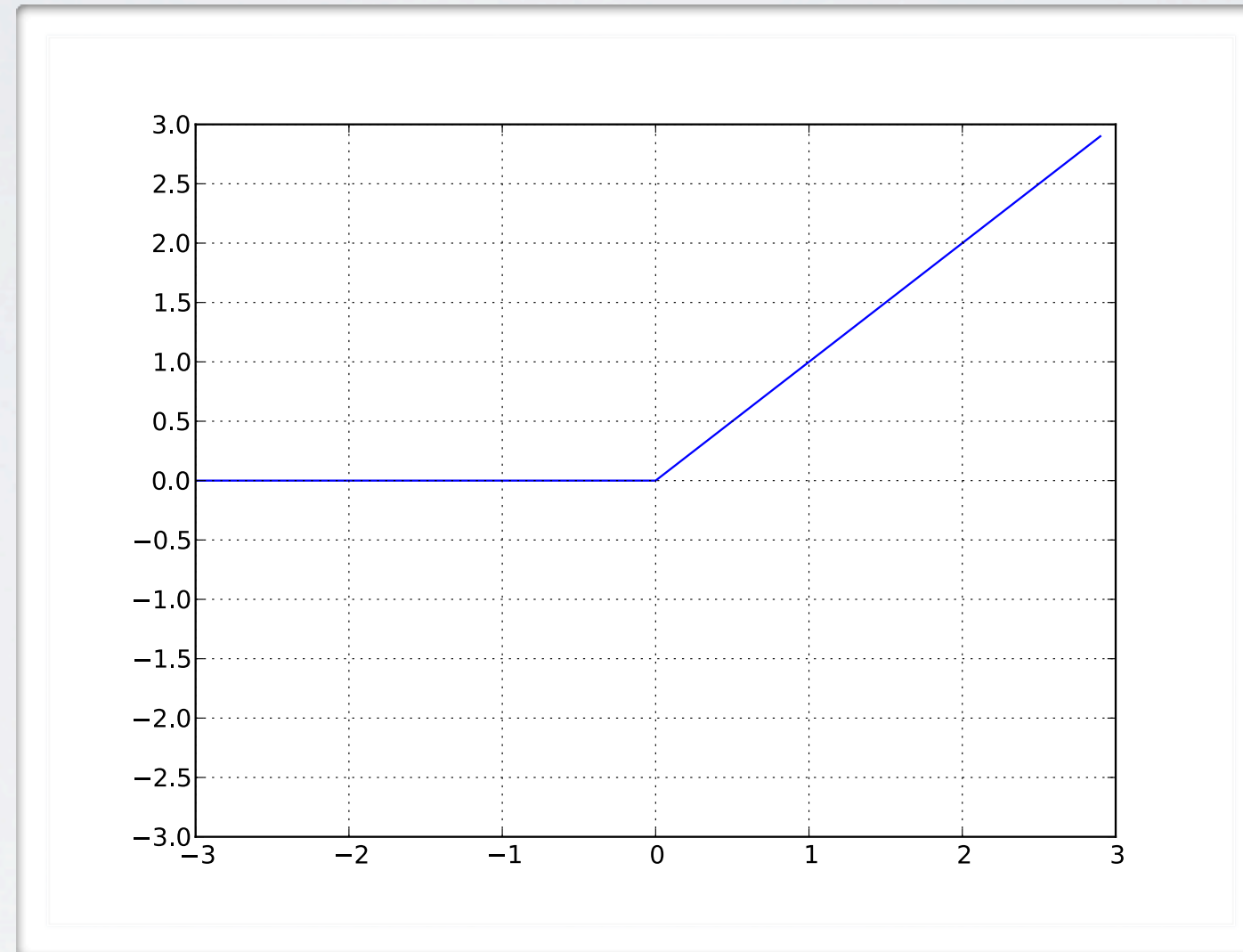


$$g(a) = \tanh(a) = \frac{\exp(a) - \exp(-a)}{\exp(a) + \exp(-a)} = \frac{\exp(2a) - 1}{\exp(2a) + 1}$$

# ACTIVATION FUNCTION

**Topics:** rectified linear activation function

- Bounded below by 0 (always non-negative)
- Not upper bounded
- Strictly increasing
- Tends to give neurons with sparse activities



$$g(a) = \text{reclin}(a) = \max(0, a)$$