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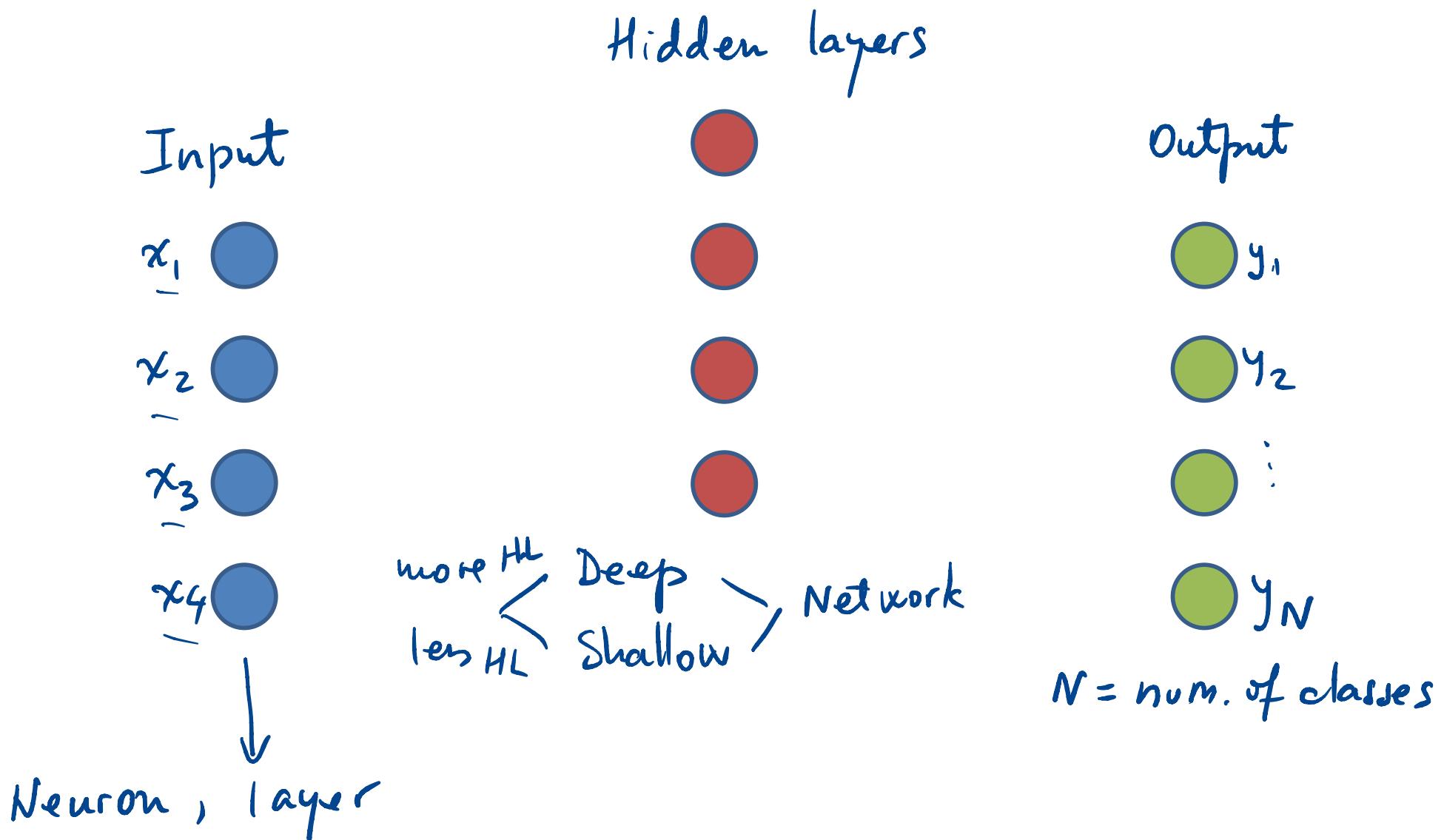
Deep Learning Applications for Computer Vision

Lecture 12: Neural Networks for Image
Classification



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Neural Network basics

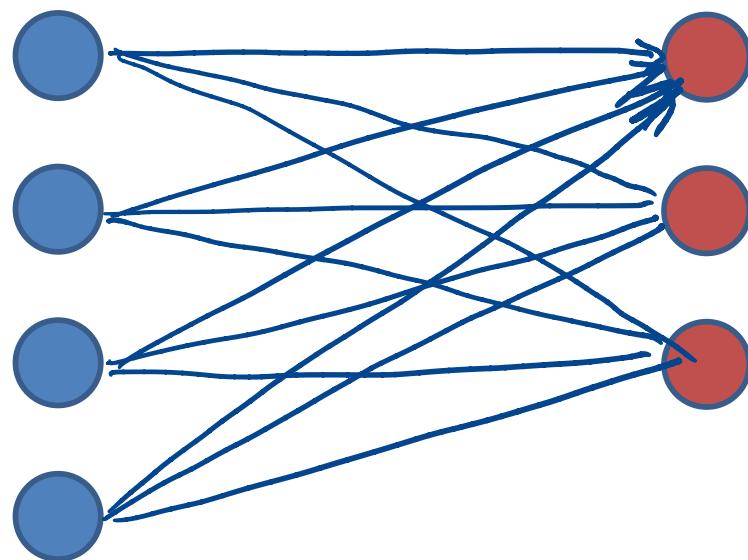


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Fully connected layers

- Also known as a **dense** layers
- Def: every output from every neuron in the previous layer is an input for every neuron from the dense layer

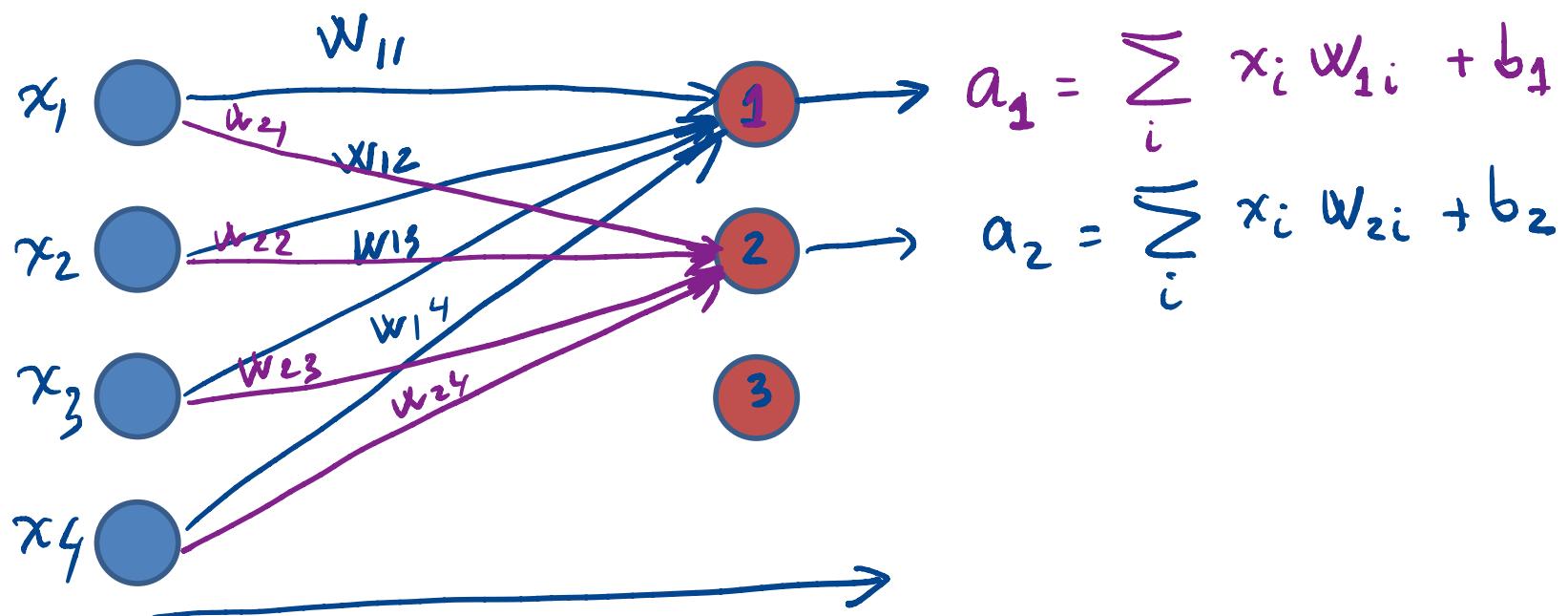
FCL



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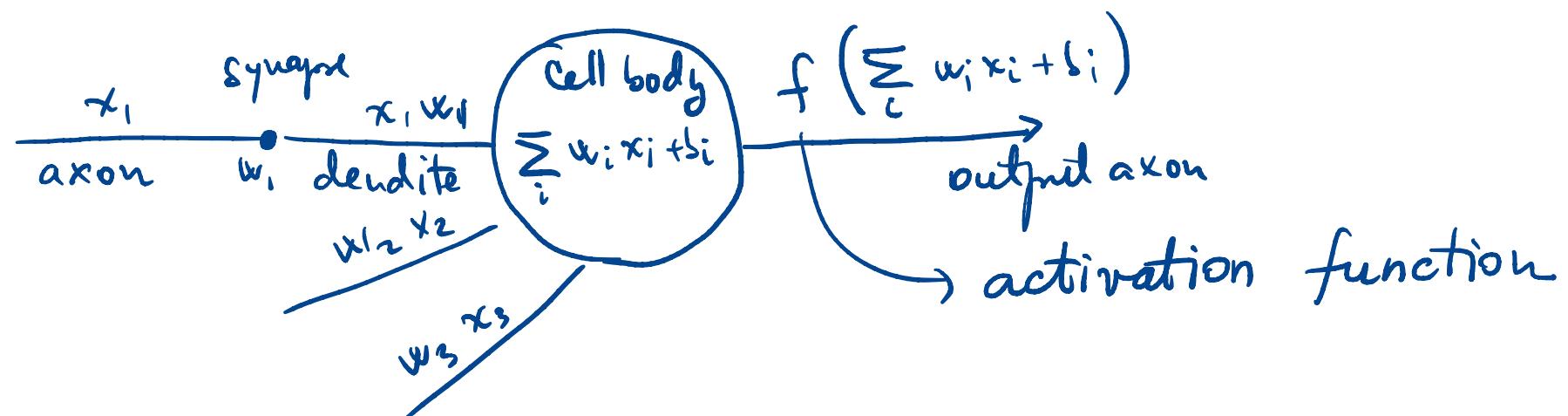
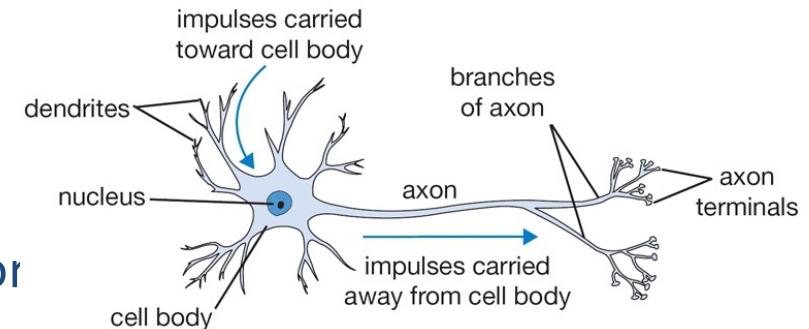
Input, Output, Weights, Biases

- Every connection between two neurons has a **weight** associated with it
- Output = (weighted sum of inputs + bias)
activation function

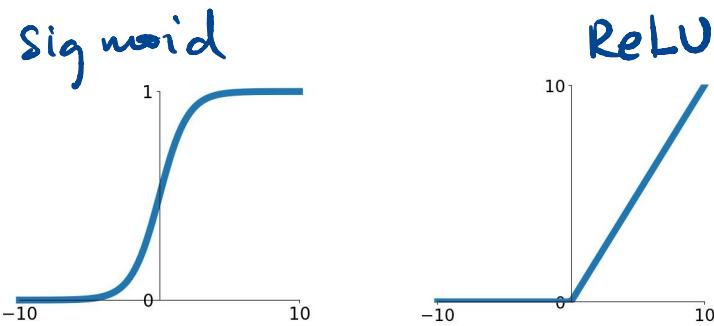


Inspiration: biological neural networks

- Neuron cells
 - accept information from multiple inputs
 - transmit information to other neurons
 - connect at synapses
 - have thresholds that must be attained for the response to take place – the neuron “fires”



Activation functions



- Reasoning
 - if we only have linear mappings at every level/layer, the result will also be a linear mapping
 - image classification is more complex; linear mapping would not be a good fit
 - we need a non-linear function – done by activation functions
 - similar to the thresholding at synapses, in biological NN
- What happens?
 - sometimes output needs to be in a certain range (0-1) or
 - sometimes output needs to always be positive ($>=0$)
 $0 \dots \infty$

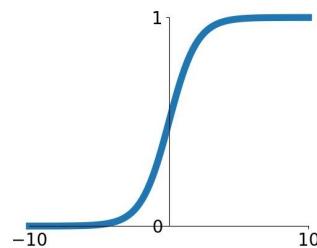


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Activation functions

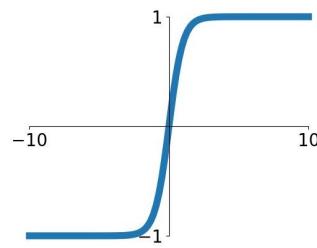
Sigmoid

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



tanh

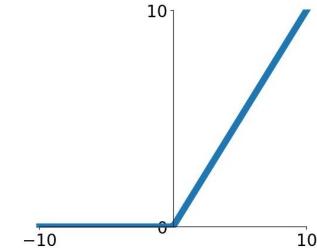
$$\tanh(x)$$



ReLU

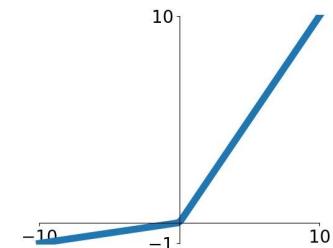
$$\max(0, x)$$

Rectified Linear Unit



Leaky ReLU

$$\max(0.1x, x)$$

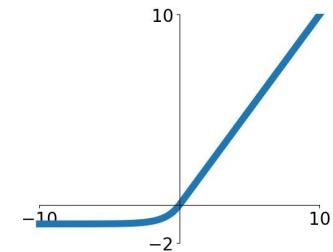


Maxout

$$\max(w_1^T x + b_1, w_2^T x + b_2)$$

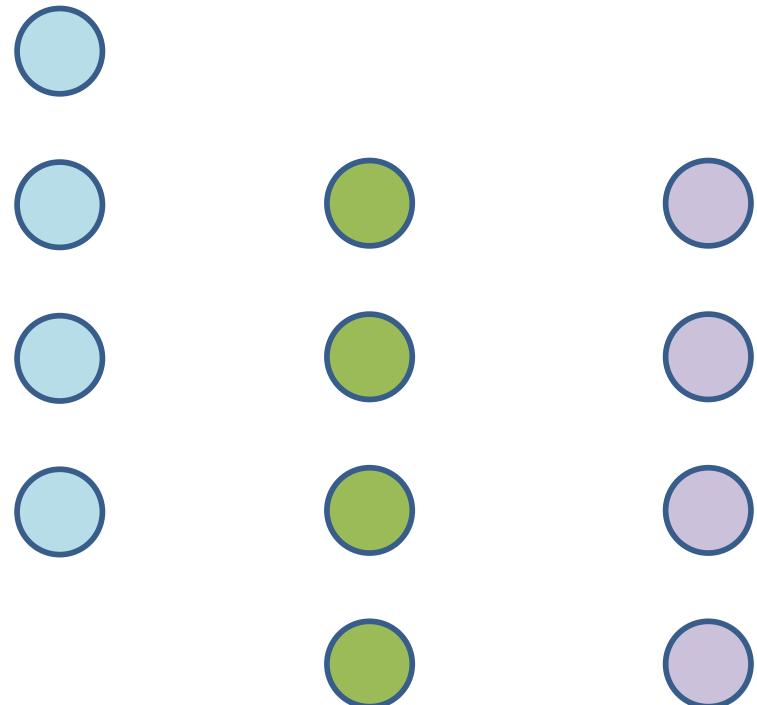
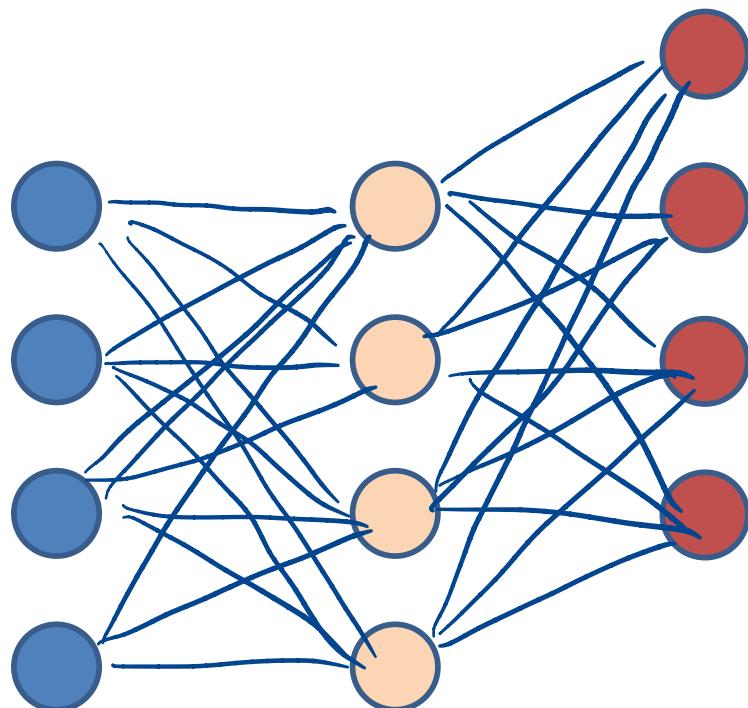
ELU

$$\begin{cases} x & x \geq 0 \\ \alpha(e^x - 1) & x < 0 \end{cases}$$



Deep neural networks

- Many hidden layers
- Many weights to learn



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How do we train the network?

- The goal is to iteratively find such a set of weights that allow the activations/outputs to match the desired output (the labels)
- We want to minimize a loss function; the loss function is a function of the weights and biases in the network

Next time: classification example when we have only a single layer of weights, and a 10-class image classification problem.



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