



University of Colorado  
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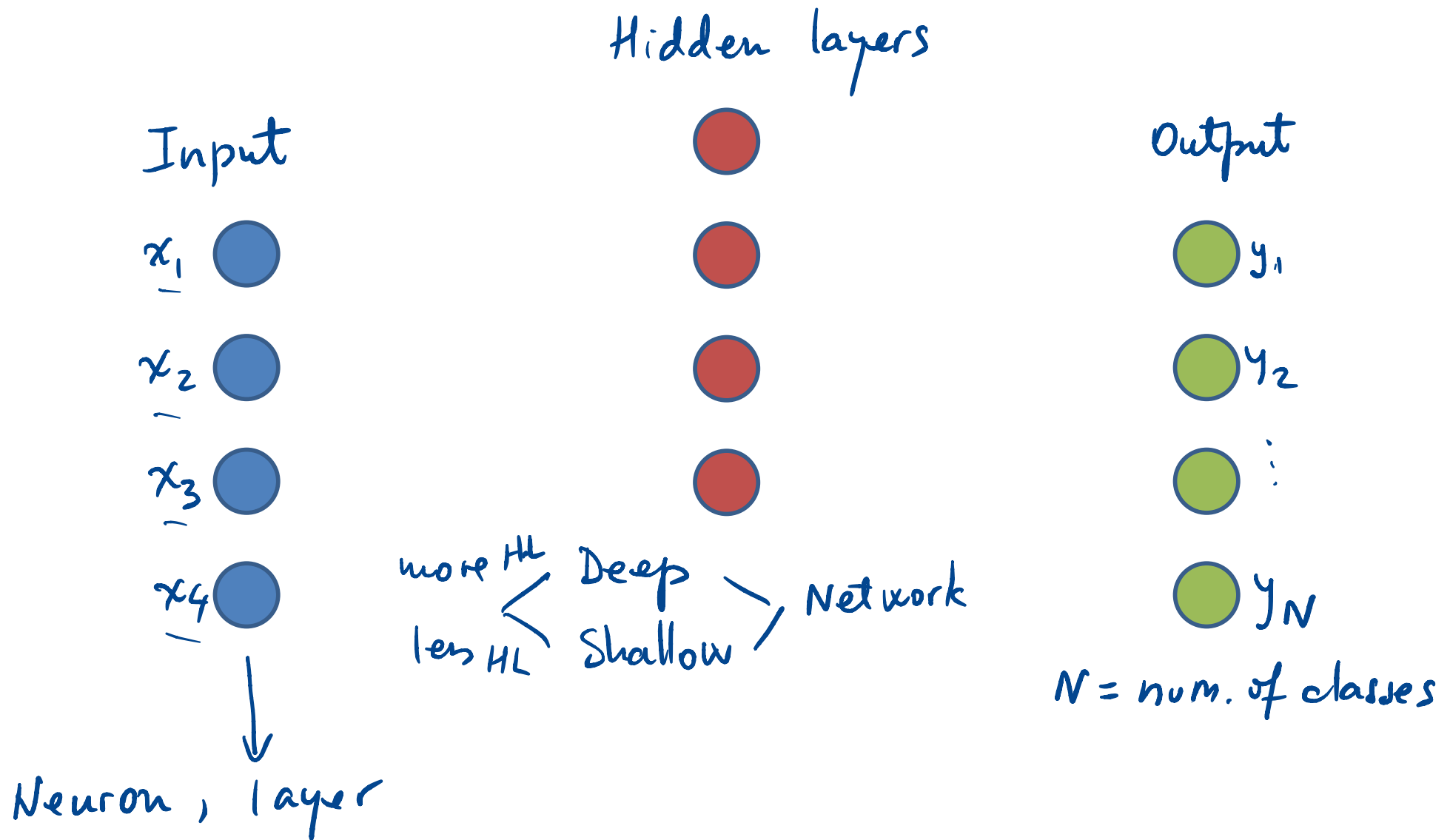


# Deep Learning Applications for Computer Vision

## Lecture 12: Neural Networks for Image Classification

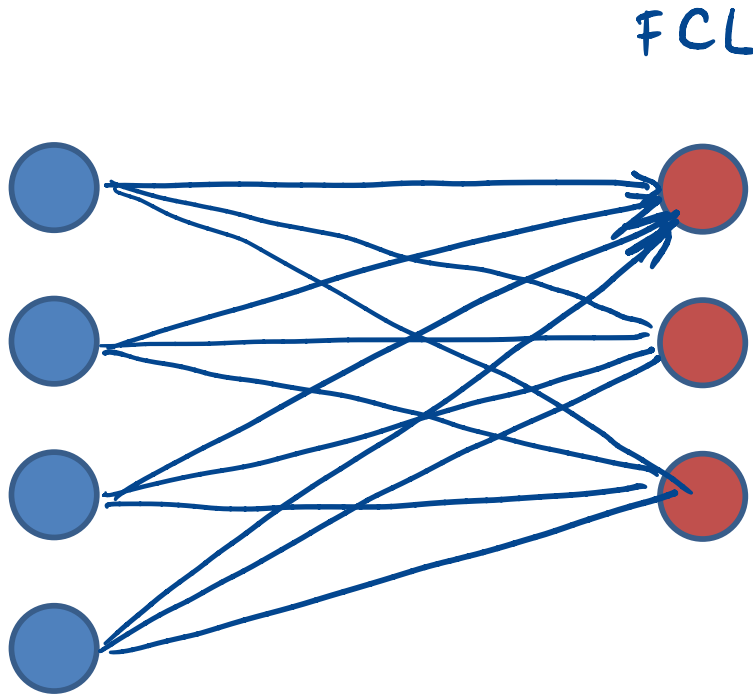


# Neural Network basics



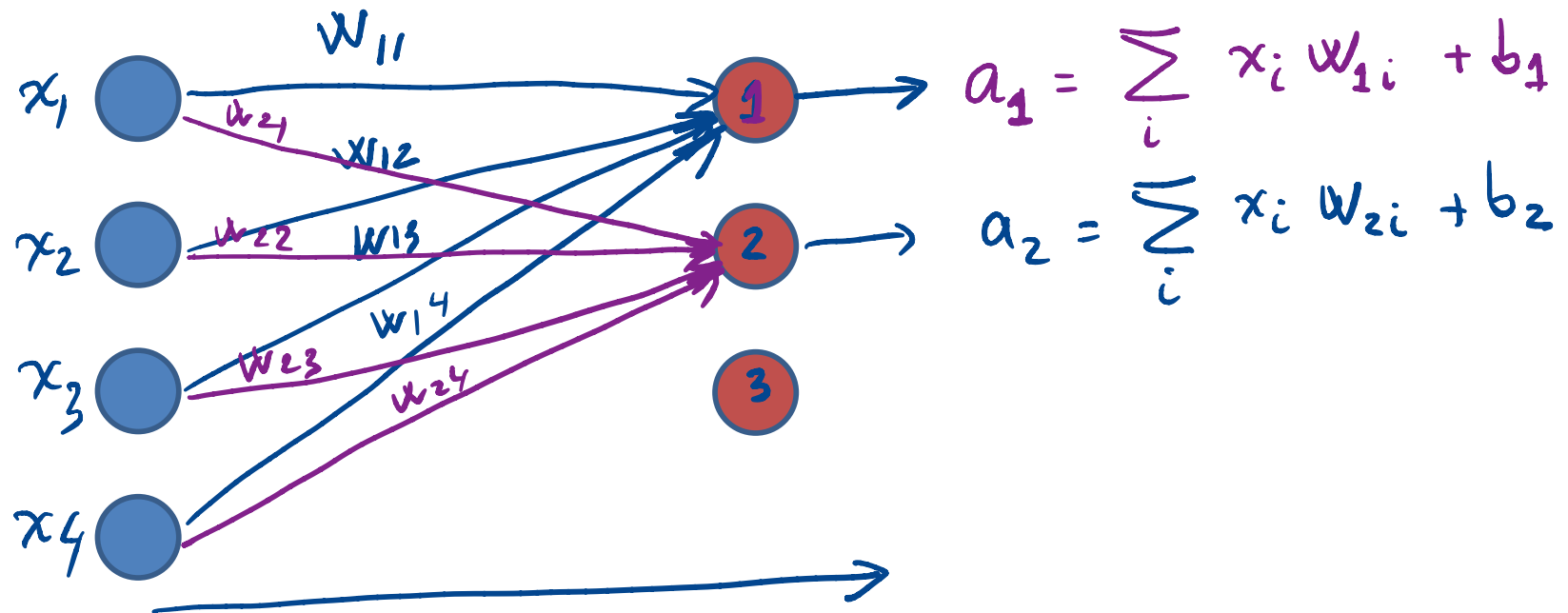
# Fully connected layers

- 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



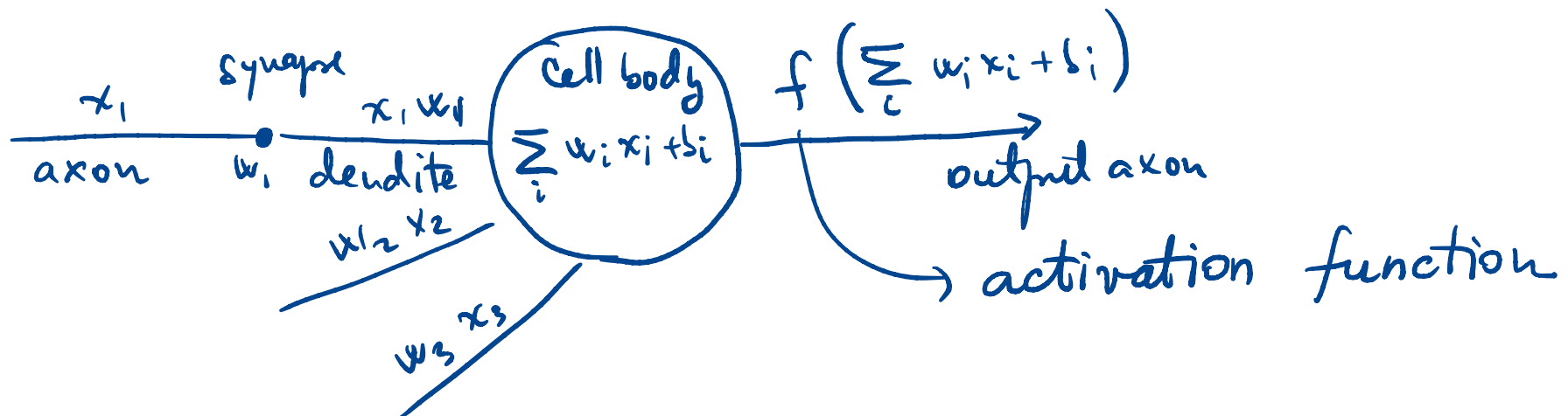
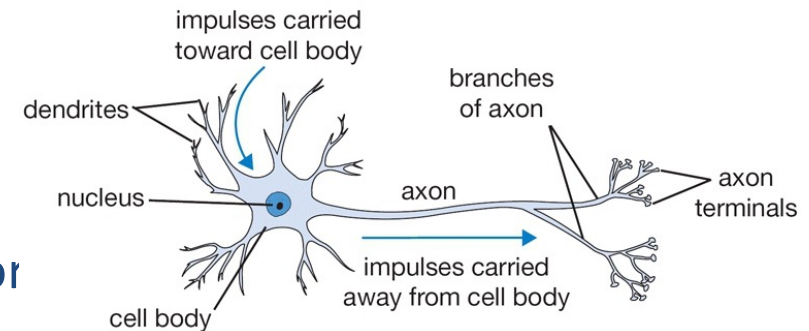
# Input, Output, Weights, Biases

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

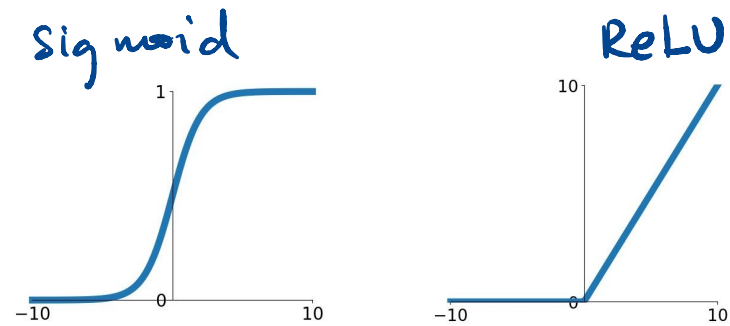


# Inspiration: biological neural networks

- Neuron cells
  - accept information from multiple inputs
  - transmit information to other neurons
  - connect a 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 ( $\geq 0$ )

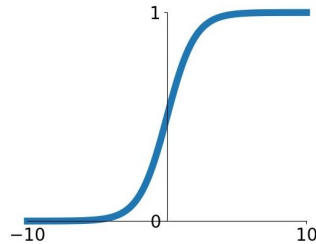
0 . -  $\infty$



# Activation functions

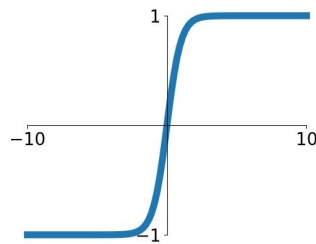
## Sigmoid

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



## tanh

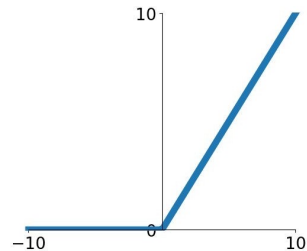
$$\tanh(x)$$



## ReLU

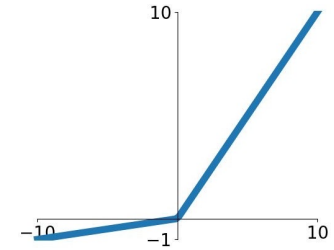
$$\max(0, x)$$

Rectified Linear Unit



## Leaky ReLU

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

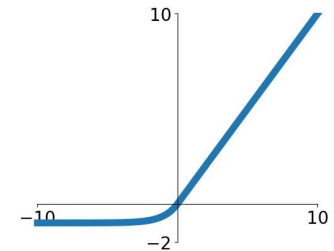


## Maxout

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

## ELU

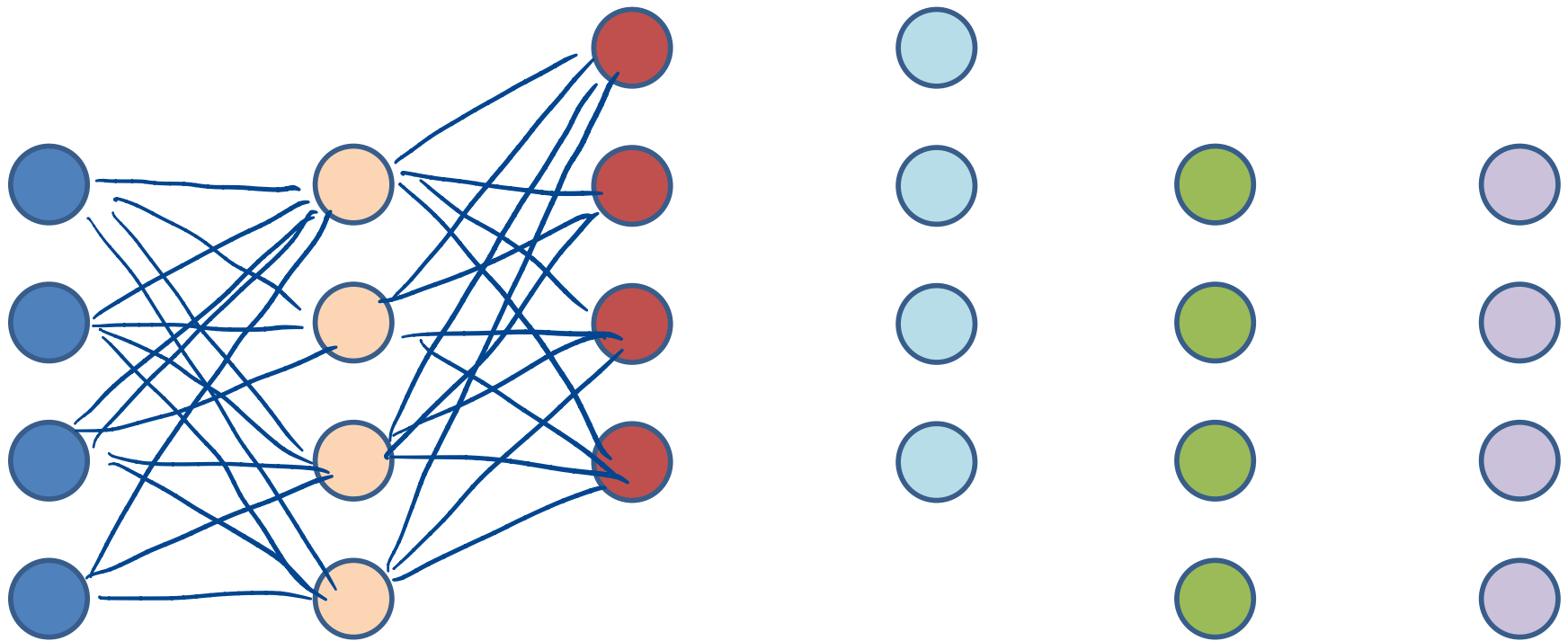
$$\begin{cases} x & x \geq 0 \\ \alpha(e^x - 1) & x < 0 \end{cases}$$





# Deep neural networks

- Many hidden layers
- Many weights to learn



# 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.

