

CSEP 576  
Machine Learning  
Neural Networks  
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# Today

- Act I
  - Perceptrons
  - Multilayered Perceptrons
- Act II
  - Backpropagation
- Act III
  - Convolutional Neural Networks

# Neural Networks

Inspired by the human brain:

- Neuron switching time  $\sim .001$  second
- Number of neurons  $\sim 10^{11}$
- Connections per neuron  $\sim 10^4$
- Total synapses  $\sim 10^{15} \rightarrow 10^{14}$
- Scene recognition time  $\sim .1$  second
- All in parallel
- Total power  $\sim 13$  Watts

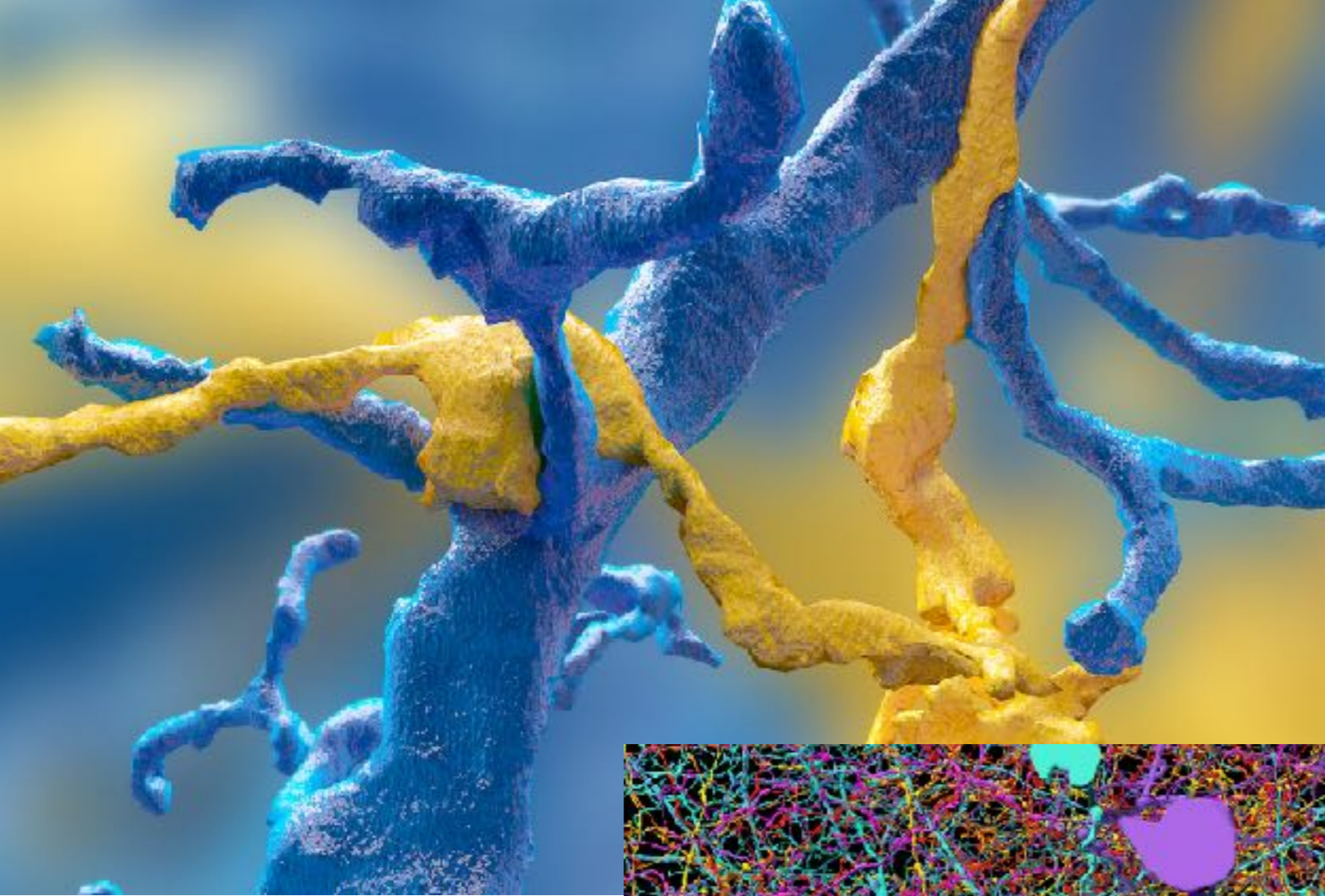
**GV100 Volta**

$10^{-9}$  second

$10^{10}$  transistors

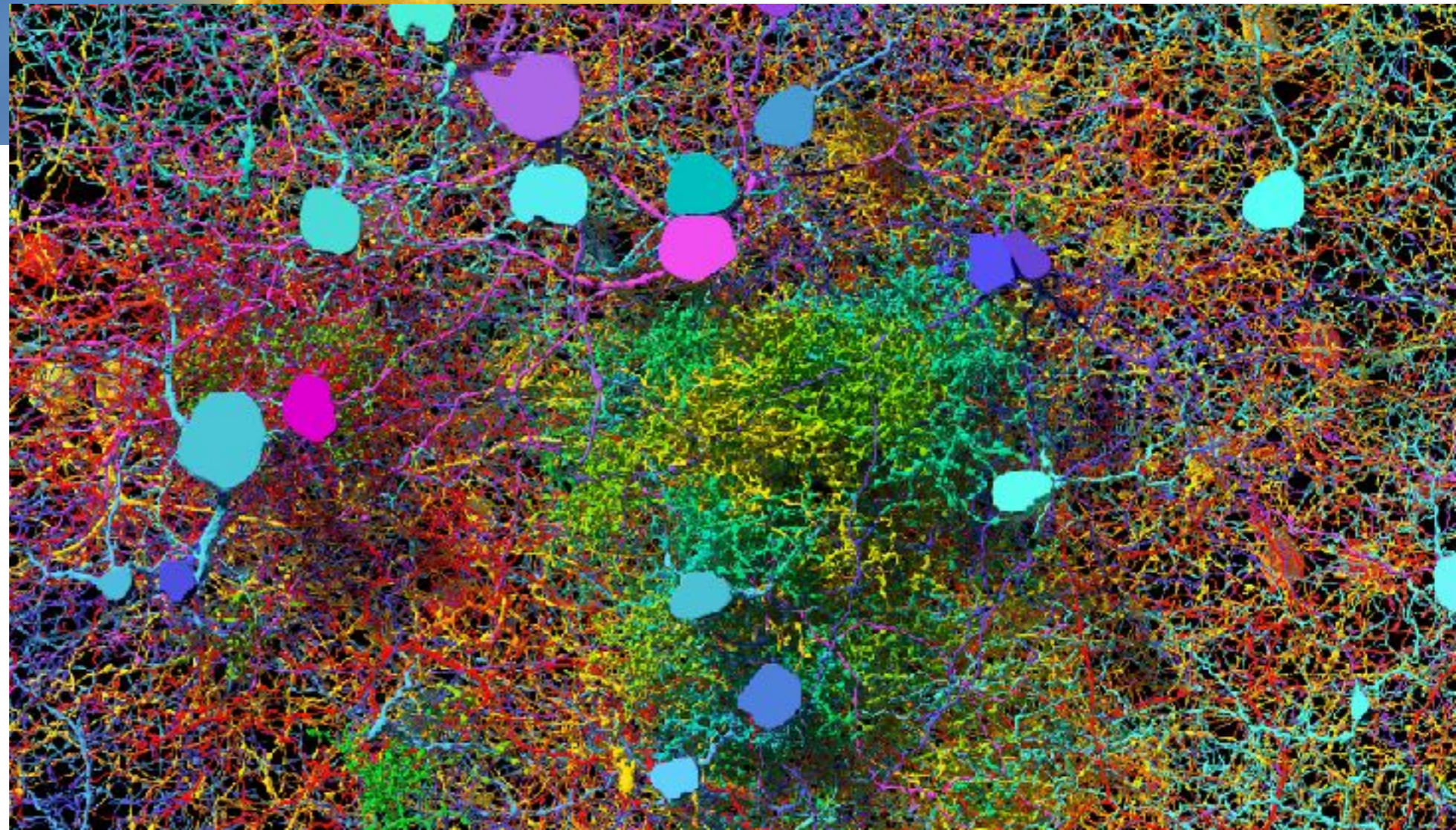
2

300 Watts



Dense Carpet of Neurons by Seung Lab, licensed by CC 2.0

Synapse by Seung Lab, licensed by CC 2.0



NEO

# NEURON ANATOMY

## DENDRITES

Signals come in through dendrites. These vast, tree-like branches grow up and out from the soma. Dendrites are thicker than axons and covered in synapses.

## AXON

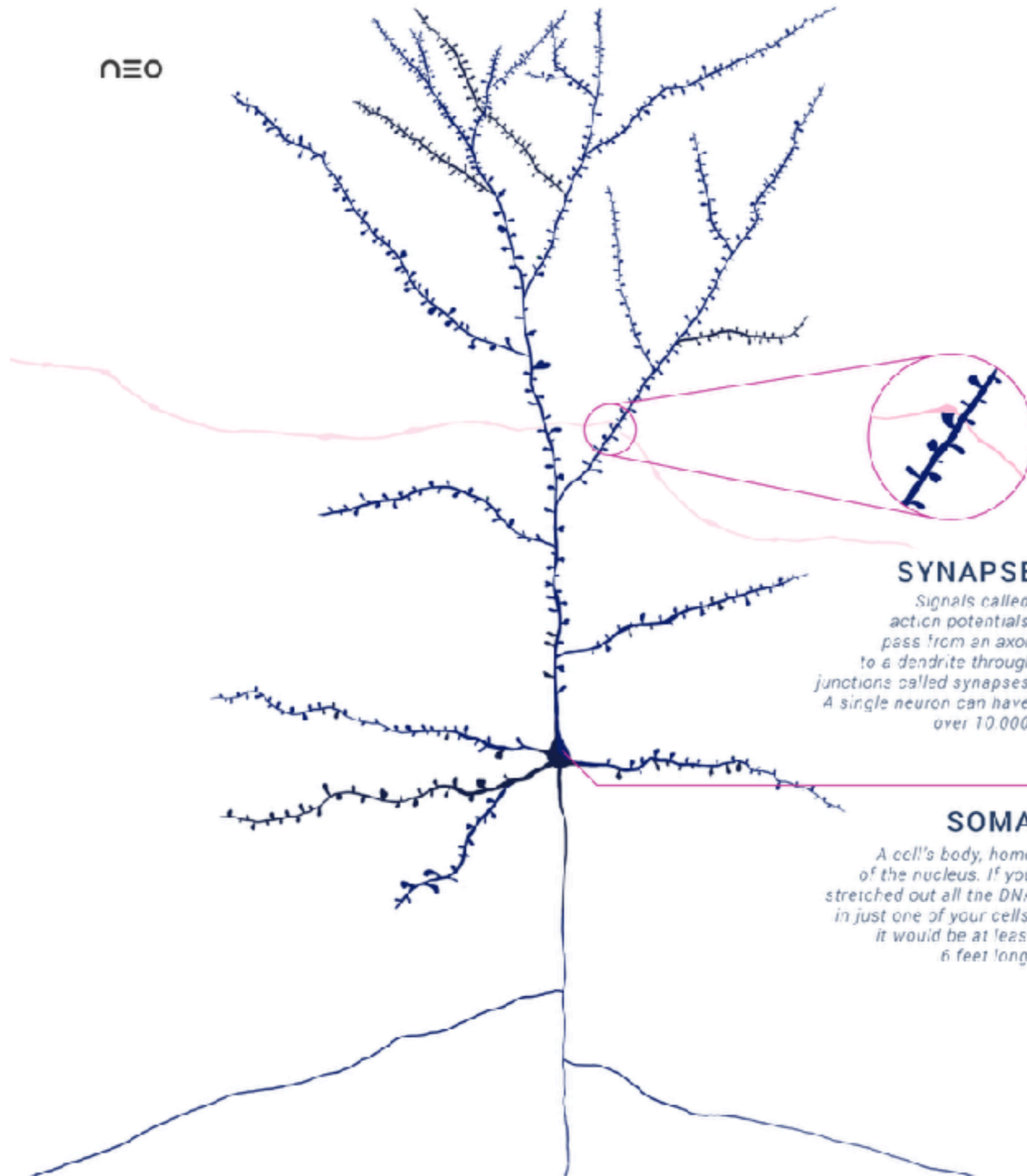
Signals go out through axons, which branch many times and stretch vast distances. Neurons send action potentials down their axons and through synapses they've formed to communicate with other cells. The longest axons in your body reach from your toes to your spine.

## SYNAPSE

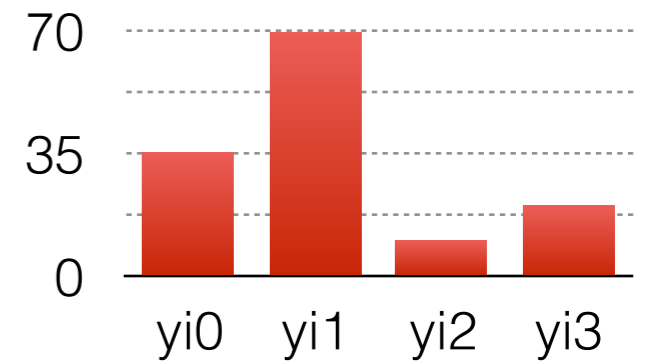
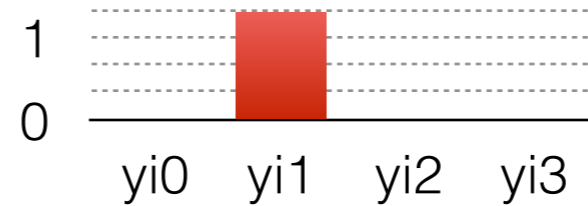
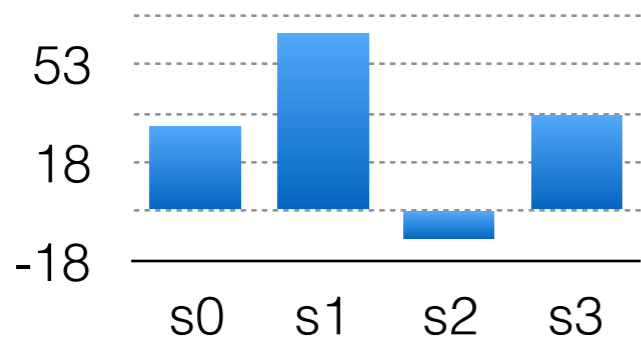
Signals called action potentials pass from an axon to a dendrite through junctions called synapses. A single neuron can have over 10,000.

## SOMA

A cell's body, home of the nucleus. If you stretched out all the DNA in just one of your cells, it would be at least 6 feet long.



# Loss functions



Classification

Regression

- Classification ( $y_i$  is label)

- Softmax cross-entropy

$$L_i = -\ln\left(\frac{e^{s_{y_i}}}{\sum_j e^{s_j}}\right)$$

- Multiclass hinge loss

$$L_i = \sum_{j \neq y_i} \max(0, s_j - s_{y_i} + 1)$$

- Regression ( $y_i$  is vector)

- L1/L2 distance

$$L_i = \sum_d |s_d - y_{id}|^2$$

- Robust norms

$$L_i = \sum_d p(s_d - y_{id})$$

# Backpropagation Algorithm

Initialize all weights to small random numbers

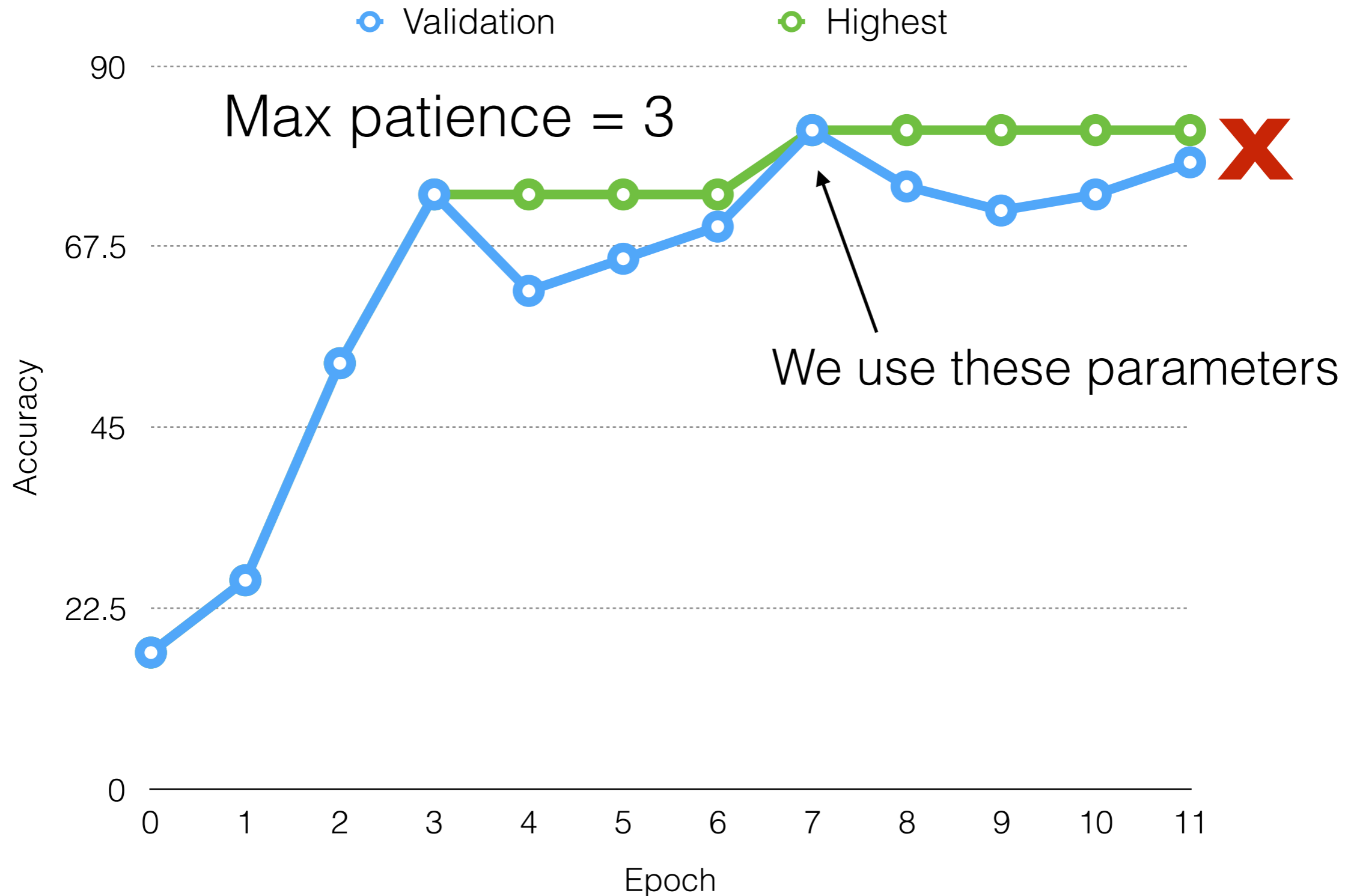
Until convergence:

For each training example:

1. Input example to network
2. For each unit from bottom to top (**forward pass**):  
Compute output from inputs
3. For each unit from top to bottom (**backward pass**):  
Compute gradient of inputs from gradient of outputs
4. Update each network weight

$$w_i \leftarrow w_i - \eta \frac{\partial E}{\partial w_i}$$

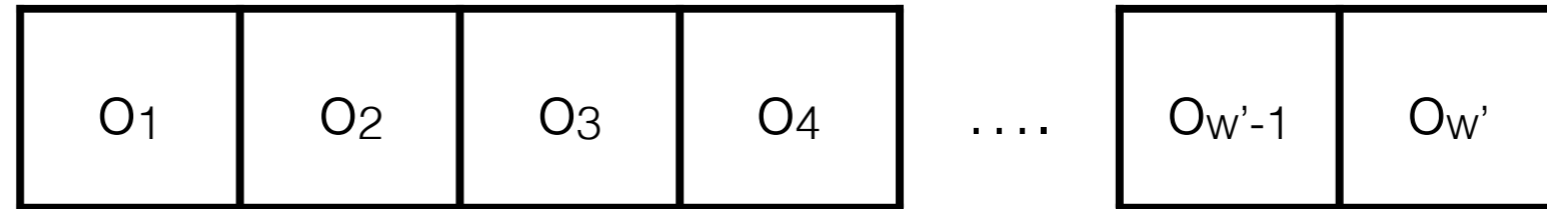
# Early Stopping





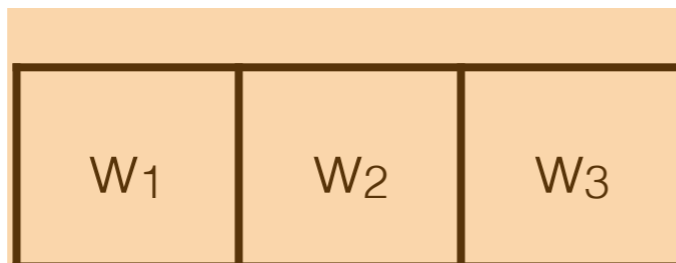
# Gradient of 1D Conv

Outputs



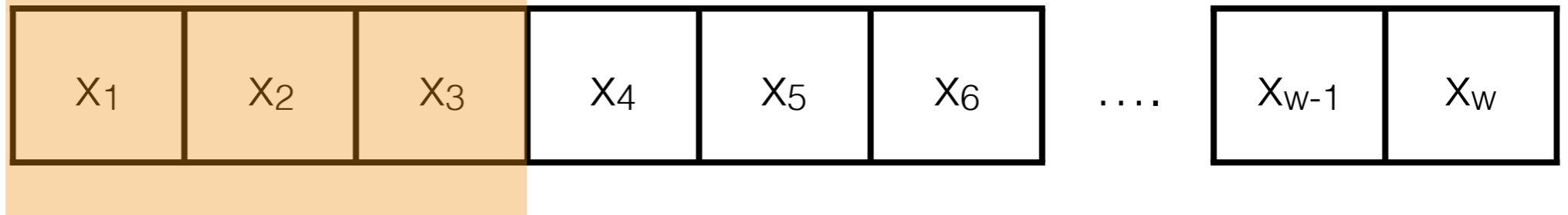
$$W_1X_1 + W_2X_2 + W_3X_3$$

Filter



\* \* \*

Inputs



**Given**  $\frac{\partial L}{\partial o_i}$ , **what are**  $\frac{\partial L}{\partial w_i}$  **and**  $\frac{\partial L}{\partial x_i}$  **?**