

# Informatics 1 Cognitive Science

## Lecture 5: Neurons

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The Action Potential

Anatomy of Neurons

Communication between Neurons

Two Simple Neuron Models

# The Action Potential

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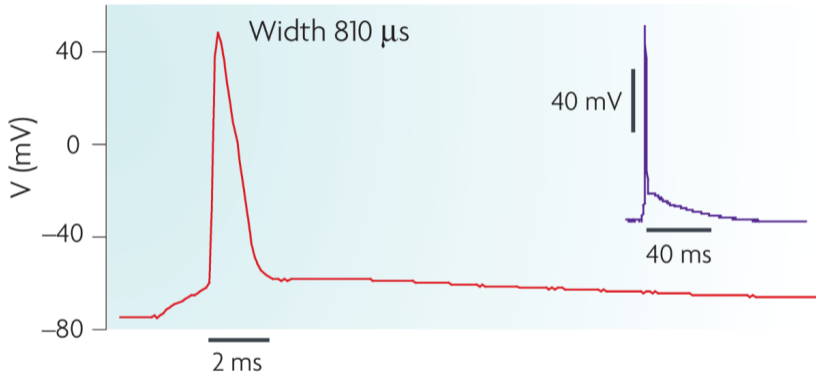
## Listening to a Neuron



With tiny electrodes (micropipettes filled with electrolyte and containing an electrode) we can record electrical activity in single neurons.

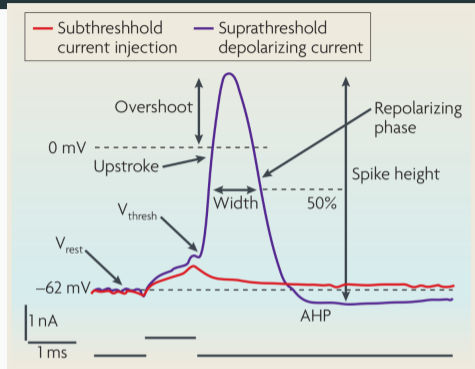
# The Action Potential (Spike)

## b CA1 pyramidal neuron



- An inward current pulse depolarises the cell membrane.
- When the depolarisation exceeds a threshold, the neuron fires a spike.

# The Action Potential (Spike)

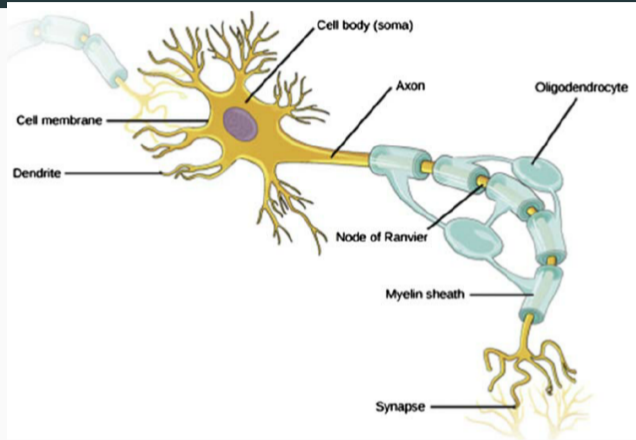


- An all-or-none electrical event in a neuron.
- A weak stimulus does not produce a weaker spike, but no spike.
- After a spike, the neuron is refractory for a short period and cannot spike again.
- The refractory period is a short hyperpolarisation of the membrane potential.

# Anatomy of Neurons

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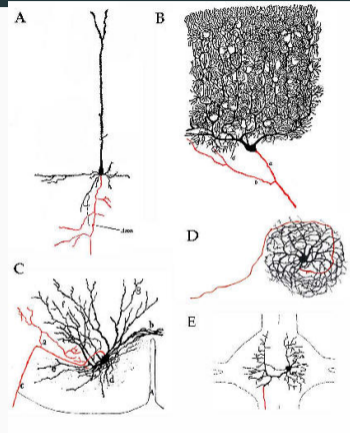
# A Neuron



Neurons collect inputs through *dendrites*, and send signals (spikes) to other neurons via their *axons*.

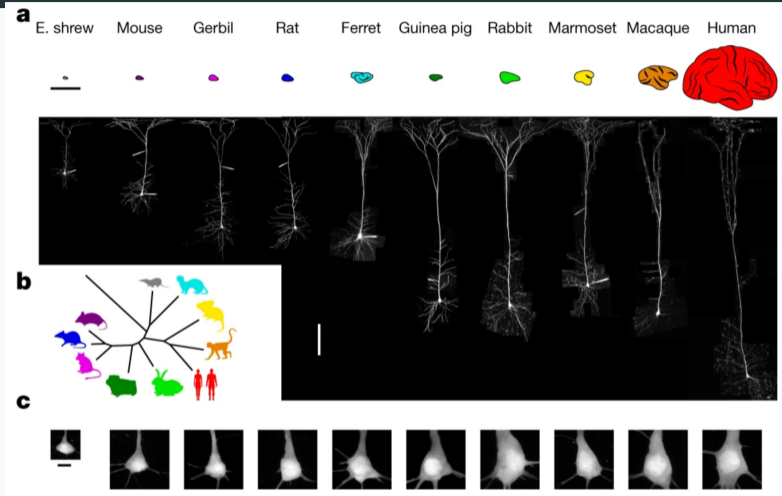
# Anatomical Diversity of Neurons

- Neurons differ markedly in anatomy and physiology.
- Dendrites can be extensive, but are usually confined to 100s of micrometers ( $\mu\text{m}$ ).
- Axons may transmit signals over long distances (up to meters), and to multiple targets.
- Communication in axons is fast with around 100 m/s.



A: Pyramidal cell, cortex; B: Purkinje cell, cerebellum; C: Motorneuron, spinal cord; D: Inferior olivary nucleus cell, E: Leech sensory neurons (red: axon, black: dendrites)

# Neuron anatomy across species (Cortex)

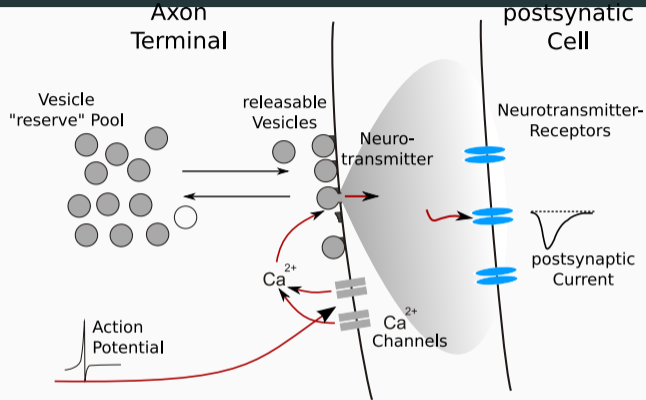


Scale bars: 5cm top, 200  $\mu$ m bottom. Beaulieu-Laroche et al. (2021). Nature, 600(7888), 274-278.

# Communication between Neurons

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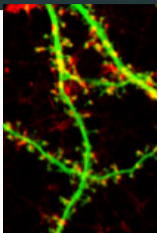
# Synaptic Transmission between Neurons



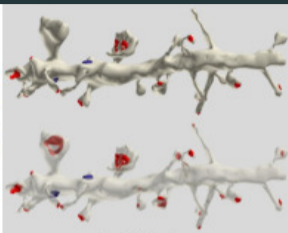
A *synapse* is a specialised contact between axon and dendrite of two neurons. A spike causes *neurotransmitter release* at the synapse, which in turn changes the potential of the receiving neuron.

The strength of a synapse can vary: it depends on amount of transmitter released and number of postsynaptic receptors.

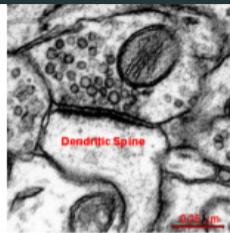
# Anatomical Diversity of Synapses



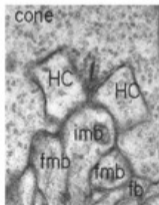
Cultured hippocampal neurons



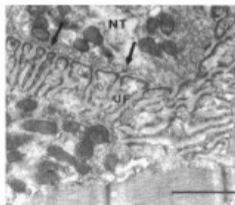
Dendritic spines  
(from Synapse Web)



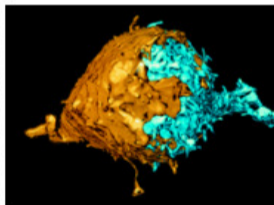
Axon and spine  
(from Synapse Web)



Human retinal cone terminal



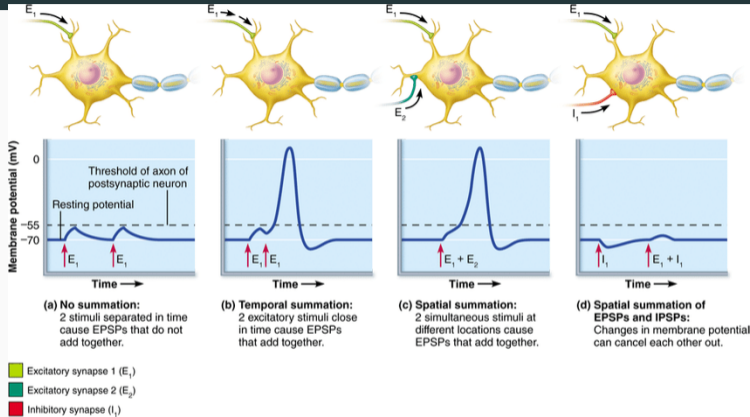
Mouse neuromuscular junction  
(Seligeter, 1987)



Calyx of Held in rat auditory brainstem  
(Smetzik et al, 2002)

A neuron may receive anywhere between 1 and 100,000 synaptic inputs.

# Excitatory and Inhibitory Synapses

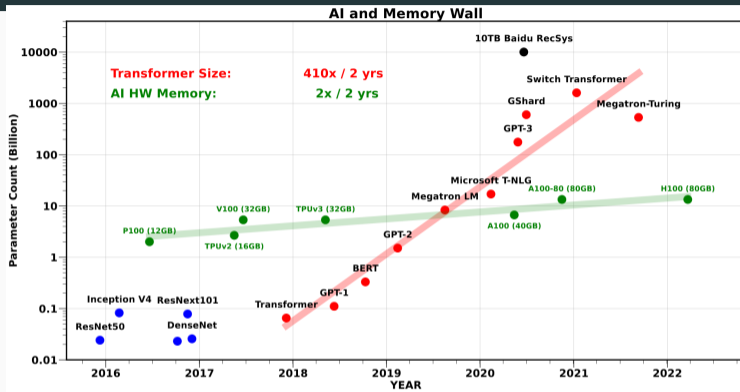


Synapses can *excite* (bring closer to spiking threshold), or *inhibit* the receiving neuron.

**Dale's principle:** Each neuron can *make* only either excitatory or inhibitory synapses.

A neuron will usually receive *both* excitatory and inhibitory inputs.

# More synapses, better brains?



- Deep learning models (esp. LLMs): 1+ trillion parameters (weights).
- The human brain: 100 trillion synapses (weights).

image from <https://medium.com/riselab/ai-and-memory-wall-2cb4265cb0b8>

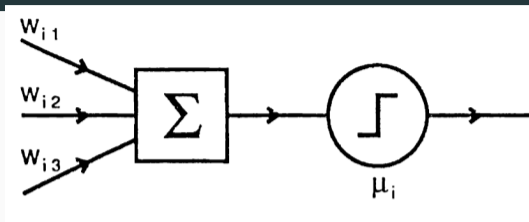
## The brain creates and eliminates synapses.

- A typical adult cortical neuron has around 10,000 synapses.
- During early development (synaptogenesis) and beyond, neurons and synapses are overproduced and eliminated.
- Cell loss: 25-40% or more (Finlay & Pallas, 1989).
- Collateral/synapse loss: ubiquitous (Purves & Lichtmann, 1980); in primate visual cortex alone around 5000 synapses/second are eliminated during adolescence (Bourgeois & Rakic, 1993).
- This is called pruning and has been linked to circuit refinement and specification.

## Two Simple Neuron Models

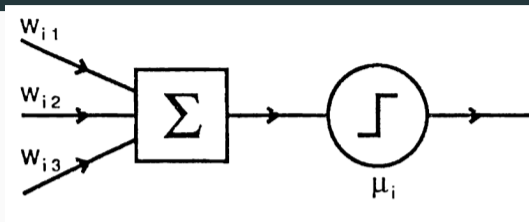
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# The McCulloch-Pitts Neuron



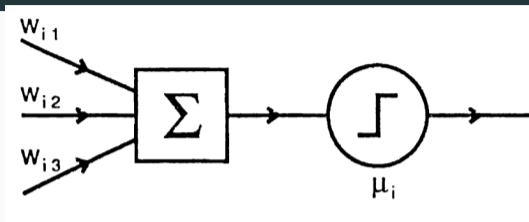
- Activity is computed as:  $n_i(t+1) = \Theta \left( \sum_j W_{ij} n_j(t) - \mu_i \right)$
- Threshold function:  $\Theta(x) = (1 \text{ if } x \leq 0; 0 \text{ otherwise})$

# The McCulloch-Pitts Neuron



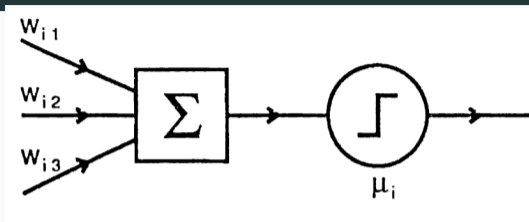
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- The weight  $w_{ih}$  determines the type/strength of each synapse. Dale's principle is not respected.

# The McCulloch-Pitts Neuron



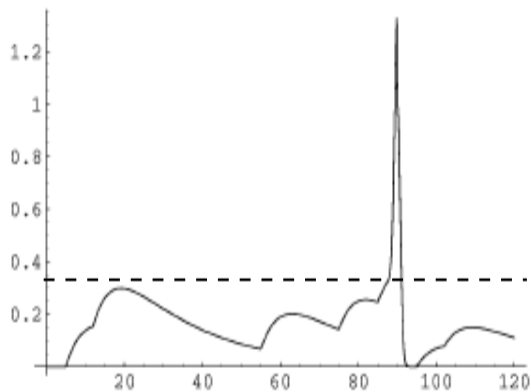
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- The neuron fires if the weighted sum  $\sum_j W_{ij} n_j$  of the inputs reaches or exceeds the threshold  $\mu_i$ .

# The McCulloch-Pitts Neuron



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- Threshold function:  $\Theta(x) = (1 \text{ if } x \leq 0; 0 \text{ otherwise})$
- The weight  $w_{ih}$  determines the type/strength of each synapse. Dale's principle is not respected.
- The neuron fires if the weighted sum  $\sum_j W_{ij} n_j$  of the inputs reaches or exceeds the threshold  $\mu_i$ .
- This neuron has no memory of its past activity.

## The Integrate and Fire Neuron



$$V(t) = V(t-1) + \frac{\Delta}{\tau} [-V_m(t-1) + I_{\text{ext}}(t)]$$

## Integrate-and-Fire: step by step

- (1) **Leak:**  $V(t) \leftarrow V(t-1) + \frac{\Delta}{\tau} [-V(t-1)]$
- (2) **Add input:**  $V(t) \leftarrow V(t) + \frac{\Delta}{\tau} I_{\text{ext}}(t)$
- (3) **Threshold:** if  $V(t) > V_{\text{thr}}$  then emit a spike
- (4) **Reset:** if spike then set  $V(t+1) = V_{\text{reset}}$  (else keep integrating)

- The term  $-V(t-1)$  makes the voltage decay back toward baseline (a *leaky* integrator).
- $\Delta/\tau$  controls how much the voltage changes per step (Euler update).
- Intuition: input pushes  $V$  up/down; the leak pulls  $V$  back; spikes happen when  $V$  crosses  $V_{\text{thr}}$ .

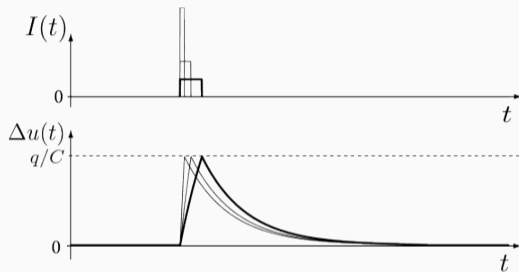
# The Integrate and Fire Neuron

$$V(t) = V(t-1) + \frac{\Delta}{\tau} [-V_m(t-1) + I_{ext}(t)]$$

if  $V(t) > V_{thr}$  then Spike and  $V(t+1) = V_{reset}$

- $I(t)$  is the external input from synapses.
- $V(t)$  without input tends towards zero, the *resting potential*.
- $\tau$  is the *membrane time constant*, determines how fast the potential changes (10-40ms).
- $V_{reset}$  is the reset potential after a spike, usually below the resting potential.
- $\Delta$  is a parameter that determines the simulation time step (as small as possible).

# The Integrate and Fire Neuron

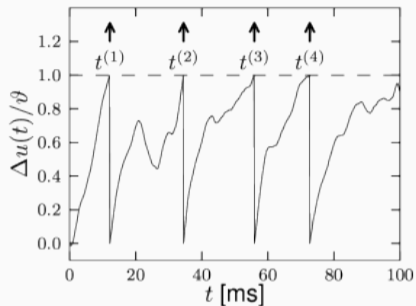
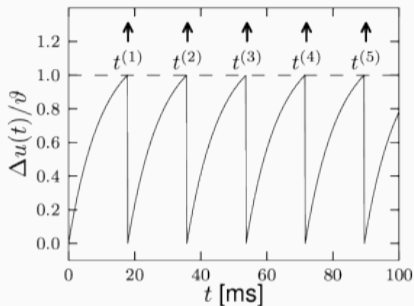


$$V(t) = V(t-1) + \frac{\Delta}{\tau} [-V_m(t-1) + I_{ext}(t)]$$

if  $V(t) > V_{thr}$  then Spike and  $V(t+1) = V_{reset}$

This neuron integrates its past activation and hence has some memory ( $\tau=10$ - $20$  ms).

# The Integrate and Fire Neuron



$$V(t) = V(t-1) + \frac{\Delta}{\tau} [-V_m(t-1) + I_{ext}(t)]$$

if  $V(t) > V_{thr}$  then Spike and  $V(t+1) = V_{reset}$

Fluctuations in the input current  $I_{ext}(t)$  lead to irregular spiking.

# Summary

- Neurons transmit information through spikes: electrical all-or-non events.
- Synapses mediate communication between neurons through chemical neurotransmitters.
- Synapses have either an excitatory or inhibitory effect.
- Synapses have different strength or weight, which quantifies their influence on the receiving neuron.
- We can use highly simplified neuron models to investigate neural computations.