

# Informatics 1 Cognitive Science

## Lecture 5: Neurons

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# Overview

The Action Potential

Anatomy of Neurons

Communication between Neurons

Two Simple Neuron Models

## The Action Potential

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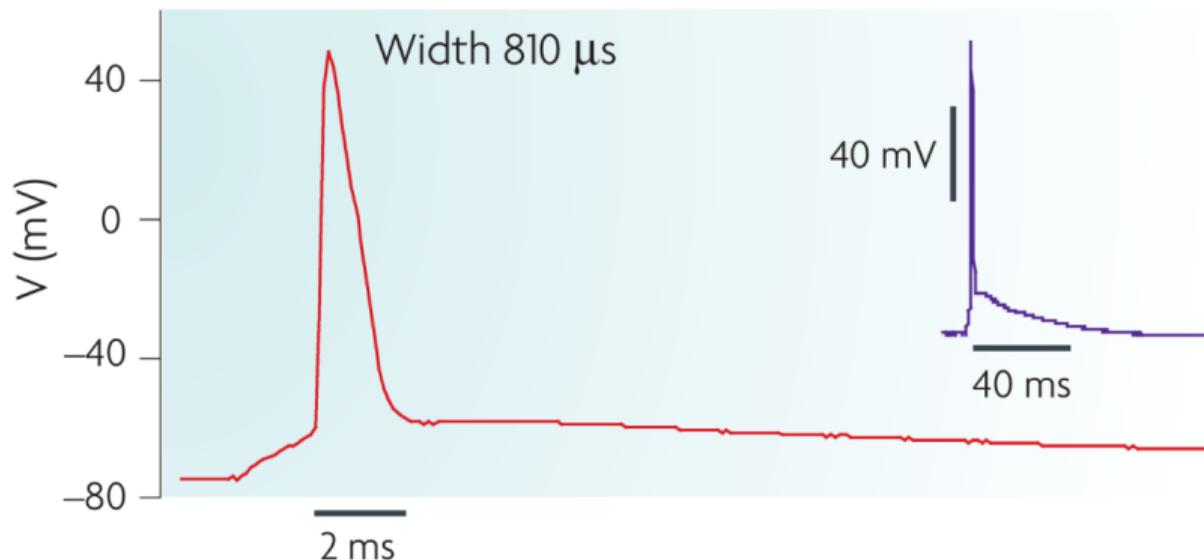
## Listening to a Neuron



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

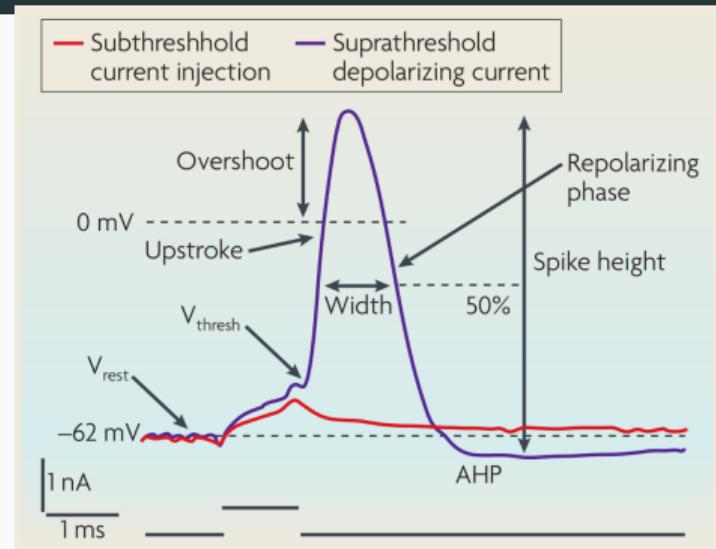
# The Action Potential (Spike)

## 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)

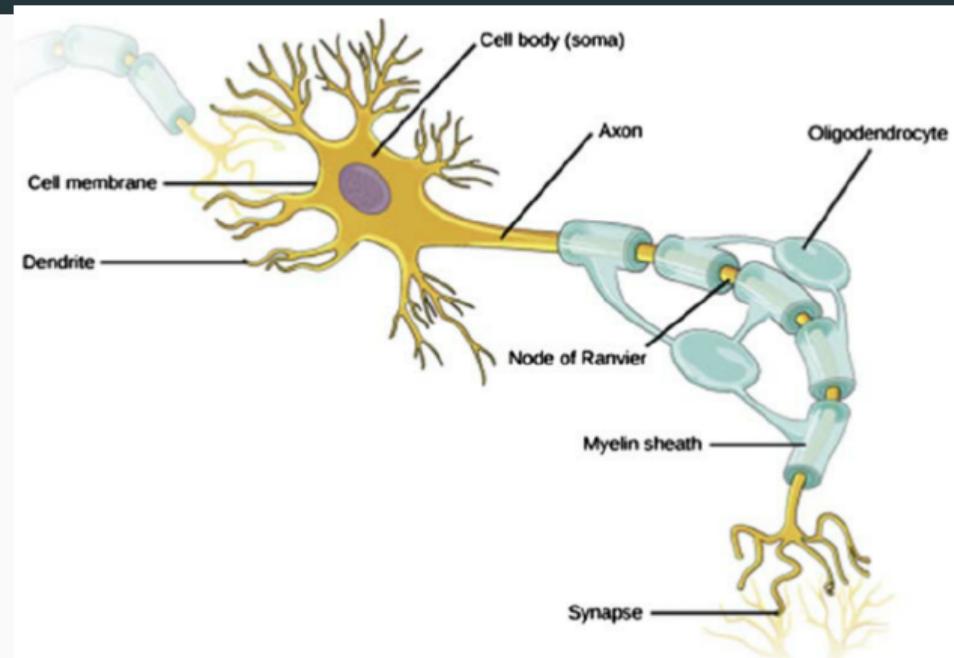


- An all-or-none electrical event in a neuron.
- A weak stimulus does not produce a weaker spike, but no spike.
- After a spike, the neuron is refractory for a short period and cannot spike again.
- The refractory period is a short hyperpolarisation of the membrane potential.

## Anatomy of Neurons

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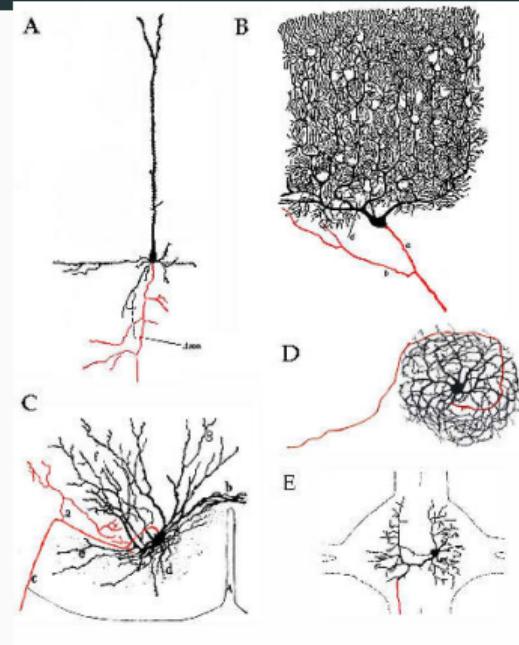
## A Neuron



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

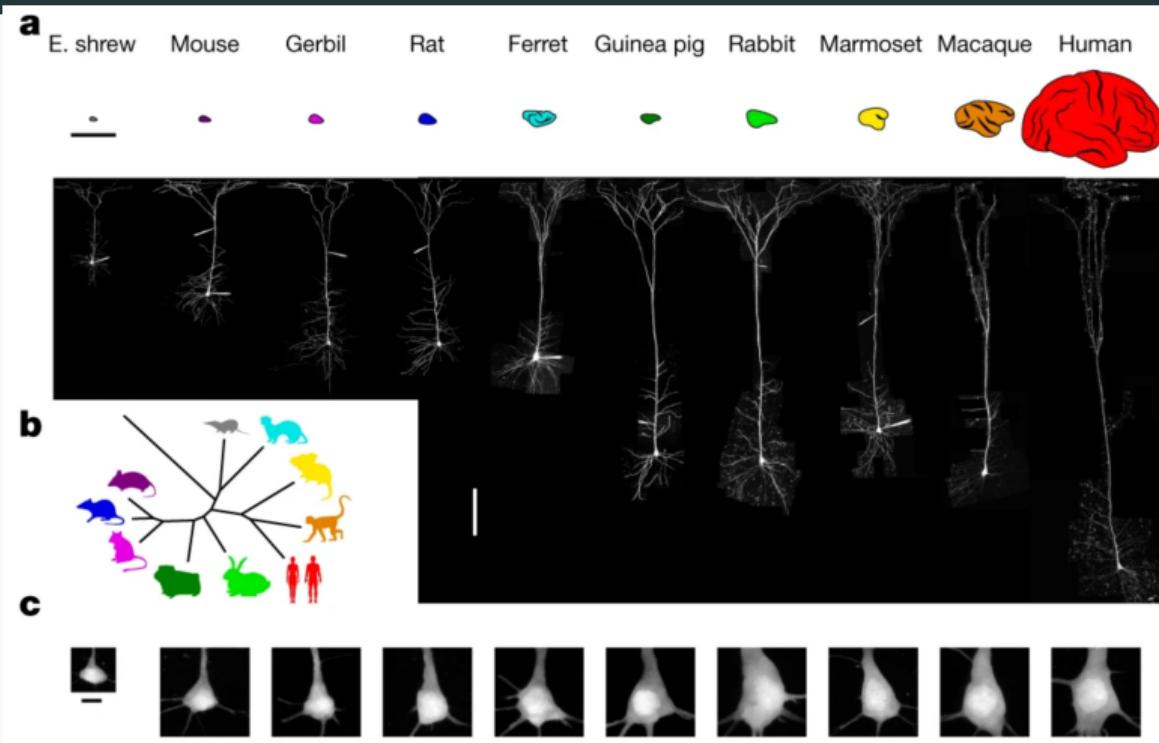
# Anatomical Diversity of Neurons

- 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 olfactory nucleus cell, E: Leech sensory neurons (red: axon, black: dendrites)

# Neuron anatomy across species (Cortex)

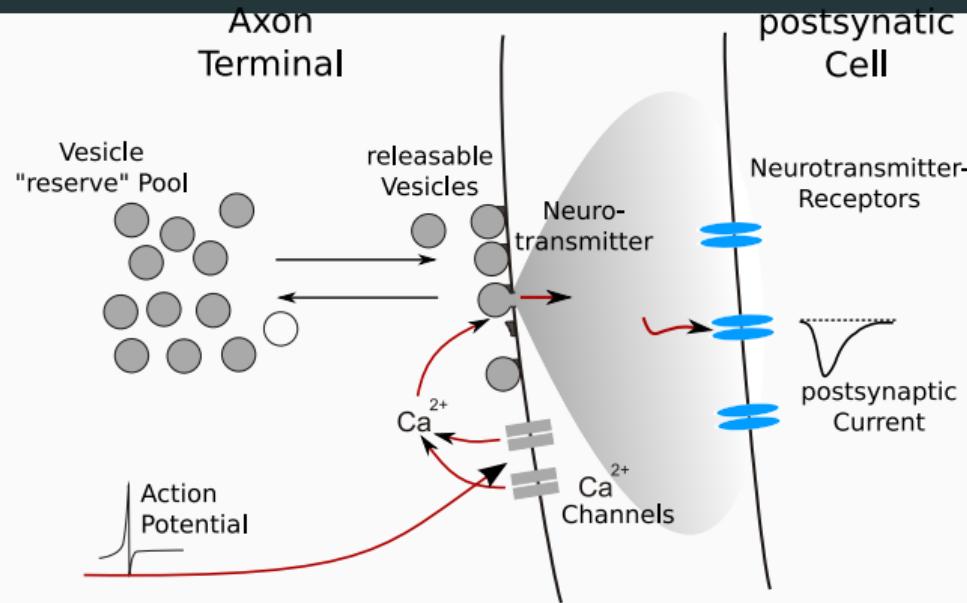


Scale bars: 5cm top, 200  $\mu$ m bottom. Beaulieu-Laroche et al. (2021). Nature, 600(7888), 274-278.

## Communication between Neurons

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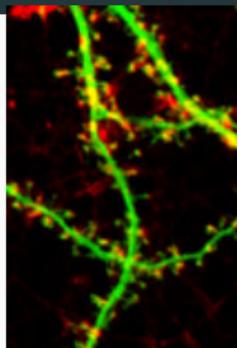
# Synaptic Transmission between Neurons



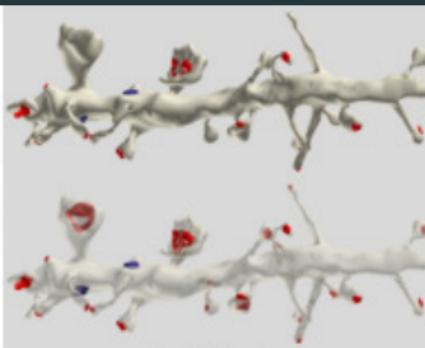
A *synapse* is a specialised contact between axon and dendrite of two neurons. A spike causes *neurotransmitter release* at the synapse, which in turn changes the potential of the receiving neuron.

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



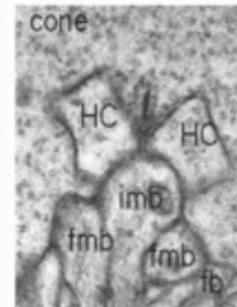
Dendritic spines (from Synapse Web)



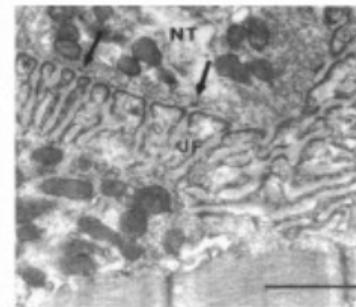
Dendritic Spine

0.25  $\mu$ m

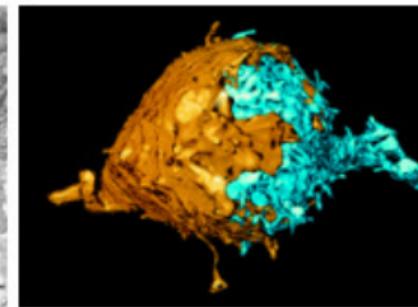
Axon and spine (from Synapse Web)



Human retinal cone terminal



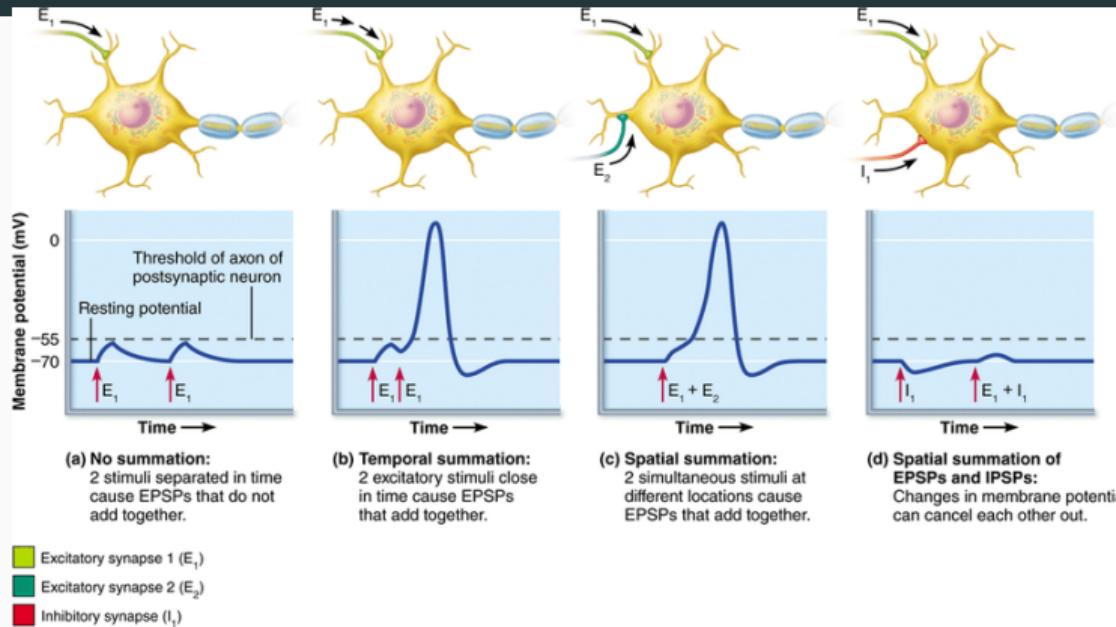
Mouse neuromuscular junction (Singer, 1987)



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

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

# Excitatory and Inhibitory Synapses

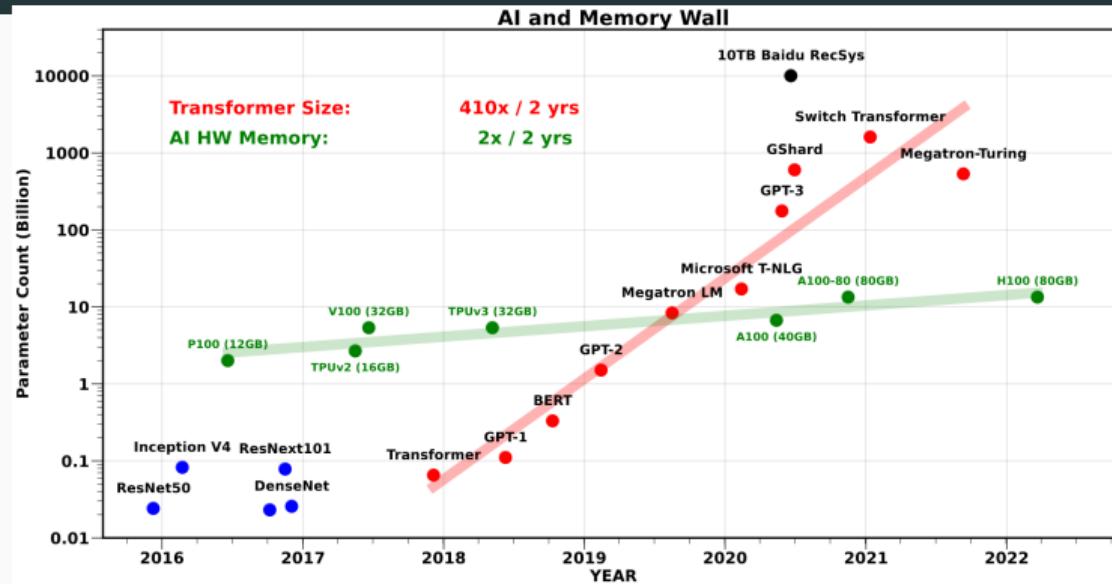


Synapses can *excite* (bring closer to spiking threshold), or *inhibit* the receiving neuron.

**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>

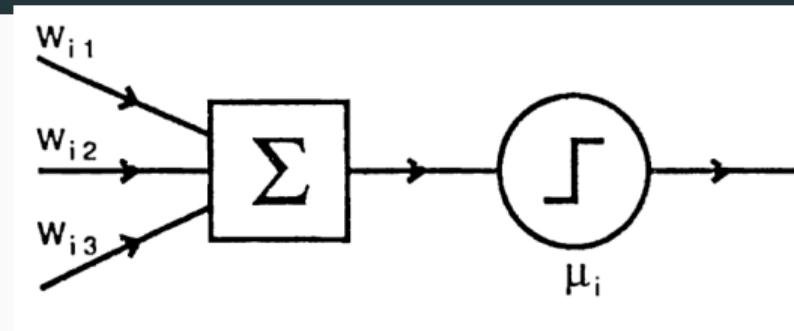
## The brain creates and eliminates synapses.

- 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

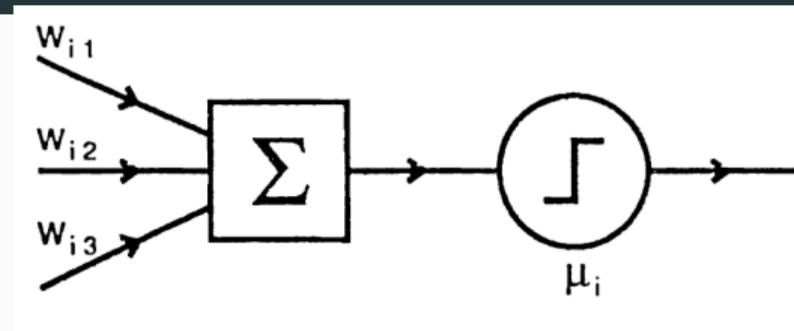
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# The McCulloch-Pitts Neuron



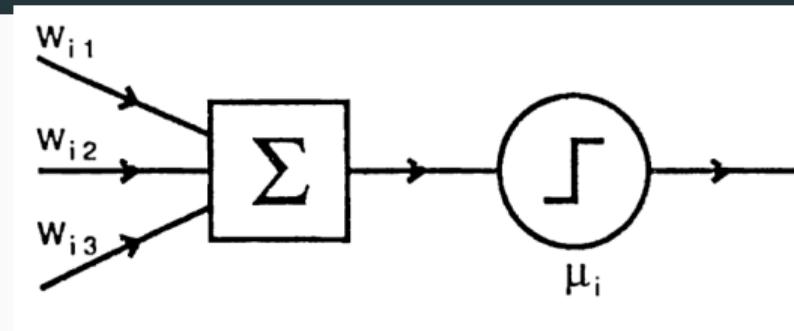
- Activity is computed as:  $n_i(t + 1) = \Theta \left( \sum_j W_{ij} n_j(t) - \mu_i \right)$
- Threshold function:  $\Theta(x) = (1 \text{ if } x \leq 0; 0 \text{ otherwise})$

# The McCulloch-Pitts Neuron



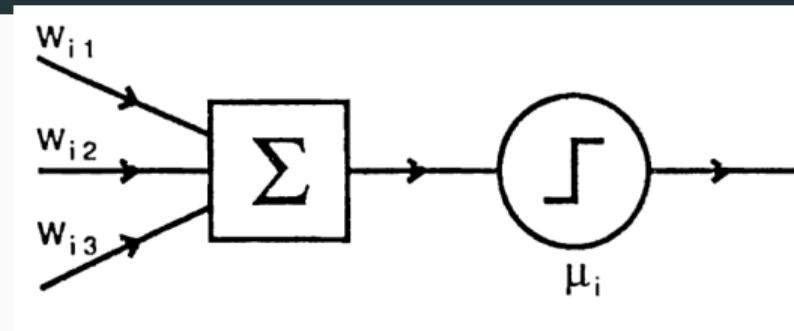
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- The weight  $w_{ih}$  determines the type/strength of each synapse. Dale's principle is not respected.

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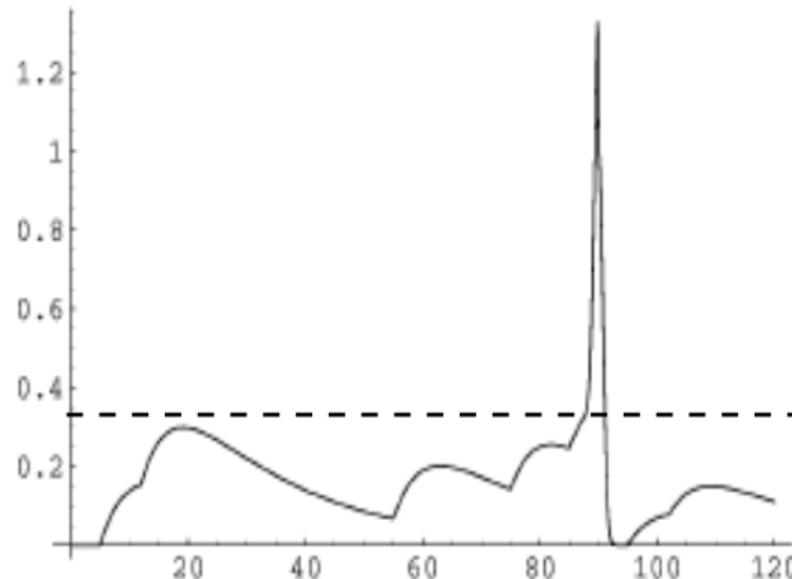
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- The neuron fires if the weighted sum  $\sum_j W_{ij} n_j$  of the inputs reaches or exceeds the threshold  $\mu_i$ .

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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



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

## Integrate-and-Fire: step by step

(1) **Leak:**  $V(t) \leftarrow V(t-1) + \frac{\Delta}{\tau}[-V(t-1)]$

(2) **Add input:**  $V(t) \leftarrow V(t) + \frac{\Delta}{\tau} I_{ext}(t)$

(3) **Threshold:** if  $V(t) > V_{thr}$  then emit a spike

(4) **Reset:** if spike then set  $V(t+1) = V_{reset}$  (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  $V_{thr}$ .

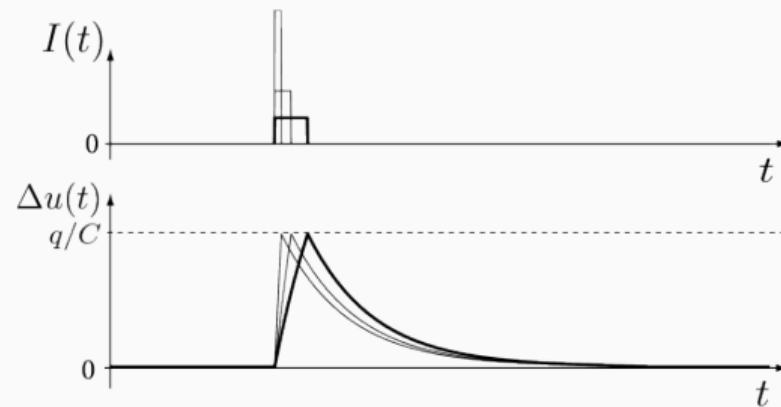
## The Integrate and Fire Neuron

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

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

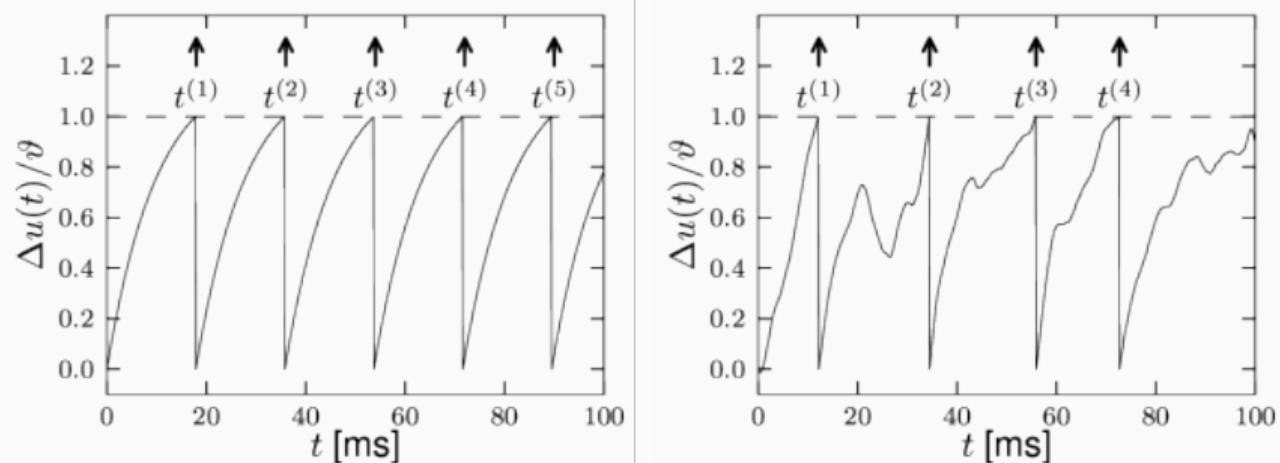


$$V(t) = V(t-1) + \frac{\Delta}{\tau} [-V_m(t-1) + I_{ext}(t)]$$

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-none 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.