# Spiking neural network simulation with Brian

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### Why study networks of neurons?

- Studying neurons in isolation has its limits
  - the brain is highly recurrent, the output of a neuron affects the network and therefore its input
- Everything we perceive, think, or do results from the activity of many neurons
- **Memories** (short and long-term) are stored on the network level, not in individual neurons
- Dynamics of the network (and not just of individual neurons) are important for healthy brains ("brain rhythms") as well as in disease (e.g. epilepsy)

### Modelling networks of neurons

#### Three main components:

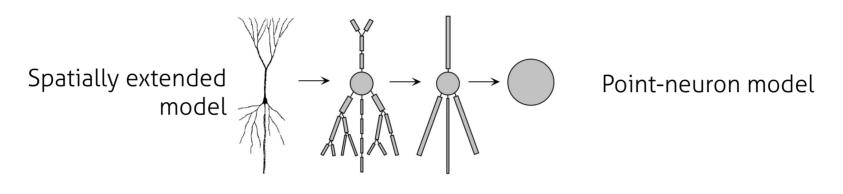
- 1) Individual neurons

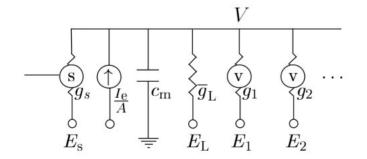
  How is a single neuron modeled?
- 2) Synaptic connectivity Which neurons connect to each other?
- 3) Synaptic models

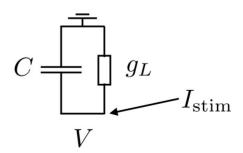
What is the effect of a spike arriving at the synapse?

Also: how does this effect change over time (plasticity)

### Individual neurons







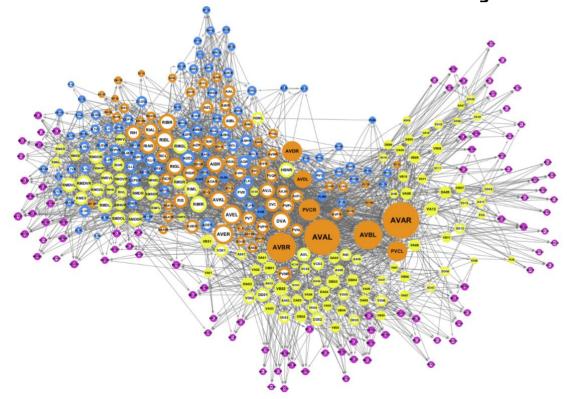
Hodgkin-Huxley formalism

integrate-and-fire model

→ see lecture/tutorial 7

## Synaptic connectivity

Structured connectivity

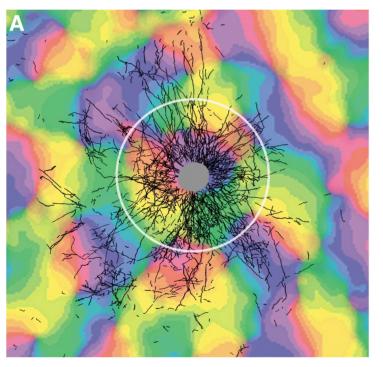


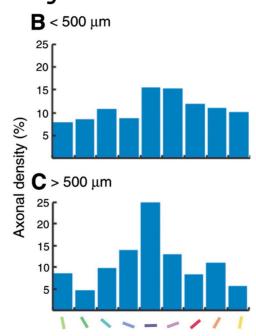
C. elegans connectome

Yan et al., Nature (2017)

### Synaptic connectivity

Structured connectivity

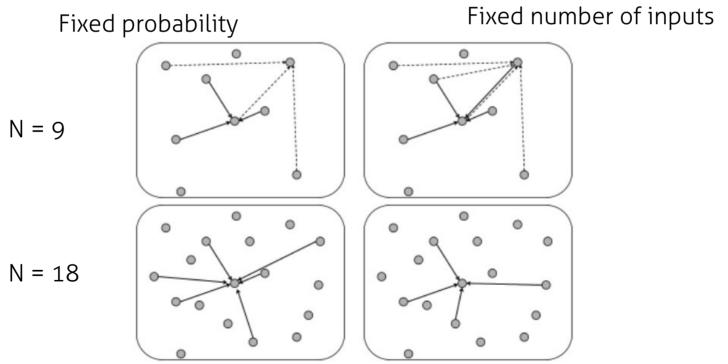




Superficial layers of V1 in Macague monkeys

### Synaptic connectivity

Unstructured (random) connectivity



Gerstner et al., Neuronal Dynamics , chapter 12

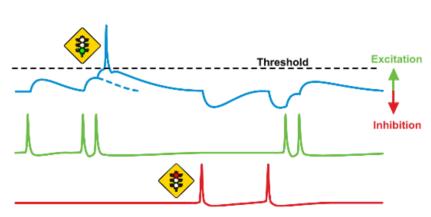
**Excitatory neuron** 

Inhibitory neuron

#### Synapses

Why can we talk about excitatory/inhibitory *neurons* and not just synapses?

→ "Dale's law"
Neurons release the same neurotransmitter(s) on every synapse



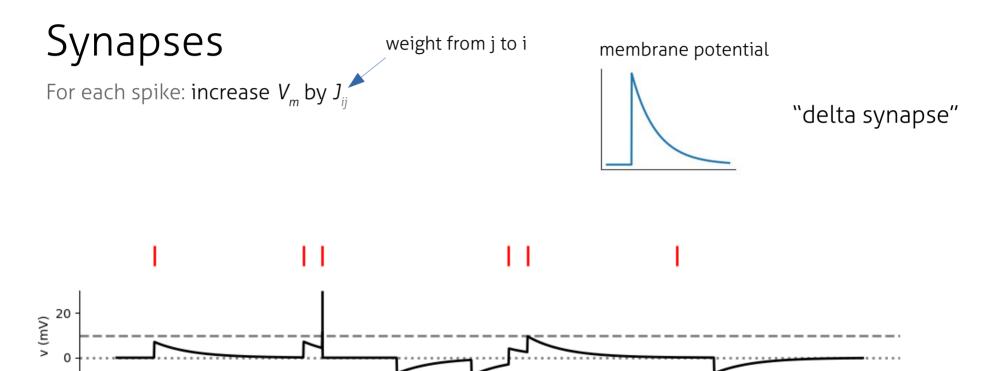
Postsynaptic neuron

**Excitatory neuron** 

Inhibitory neuron

© 2000 UTHSCH

Postsynaptic neuron

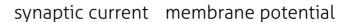


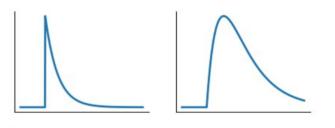
t (ms)

#### Synapses

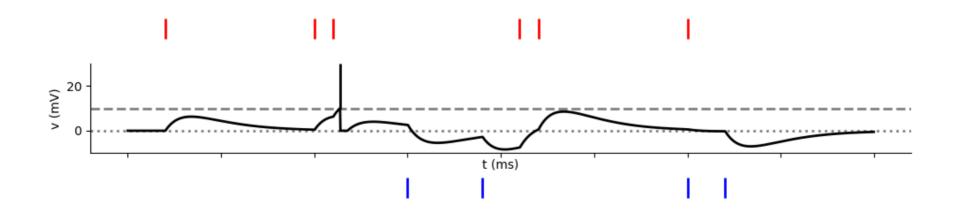
For each spike: increase  $I_{syn}$  by  $J_{ij}$ 

Between spikes:  $I_{syn}$  exponentially decays to 0





Current-based synapse



### Synapses

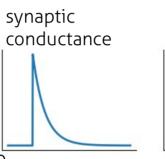
 $I_{syn} = g_{syn} (E_{syn} - V_m)$ 

For each spike:

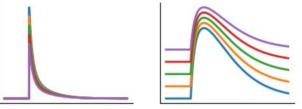
increase  $g_{syn}$  by  $J_{ij}$ 

Between spikes:

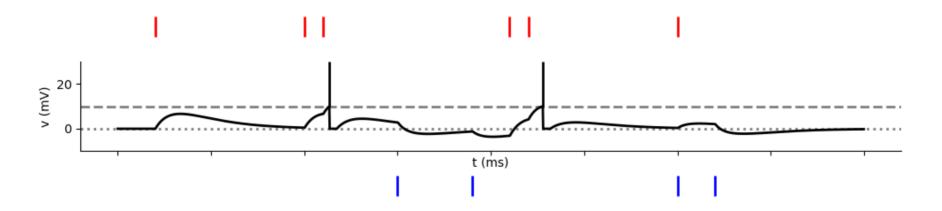
 $q_{\perp}$  exponentially decays to 0



synaptic current membrane potential



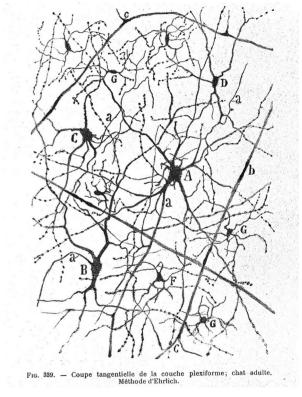
exponential conductance-based



#### Neurons in a network

#### In the mouse cortex

- Neurons: ~90000/mm³
   Synapses: ~8000 per neuron ~2700000000 synapses/mm³
- Fewer inhibitory neurons (~20% of all neurons)



Wellcome Collection gallery (2018-03-27): https://wellcomecollection.org/works/xq4qbgyd CC-BY-4.0

### Network models: approaches

With thousands of neurons and millions of synapses (even in a very small part of the brain), how can we decide on the parameters? How can we get closer to an understanding of the system?

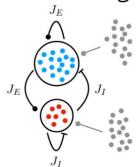
Roughly, two opposing approaches:

Try to gather as much data as possible Use complex models and supercomputers



© Blue Brain Project/EPFL

Simplify everything as much as possible Use mathematics to gain insights



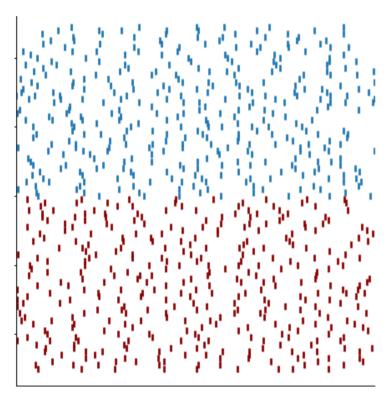
### Network models: approaches

Today, we'll simplify things a lot

- Leaky integrate-and-fire point neuron model (no Hodgkin-Huxley dynamics, no neuronal morphology, ...)
- Homogeneous population, all neurons are "the same" (no neuron types, no heterogeneity, ...)
- Simple synapse model ("delta synapses") (no synaptic dynamics, no adaptation, no plasticity...)
- Random connectivity between neurons (no structure, no position in space, ...)

### Network dynamics

#### How can we characterize network activity?



#### Individual neurons:

Weakly or strongly active? Firing regularly or irregularly?

#### **Network:**

Synchronized or unsynchronized? Oscillating?

Firing rate, ISI distribution → CV

e.g. spectral analysis (auto)correlation