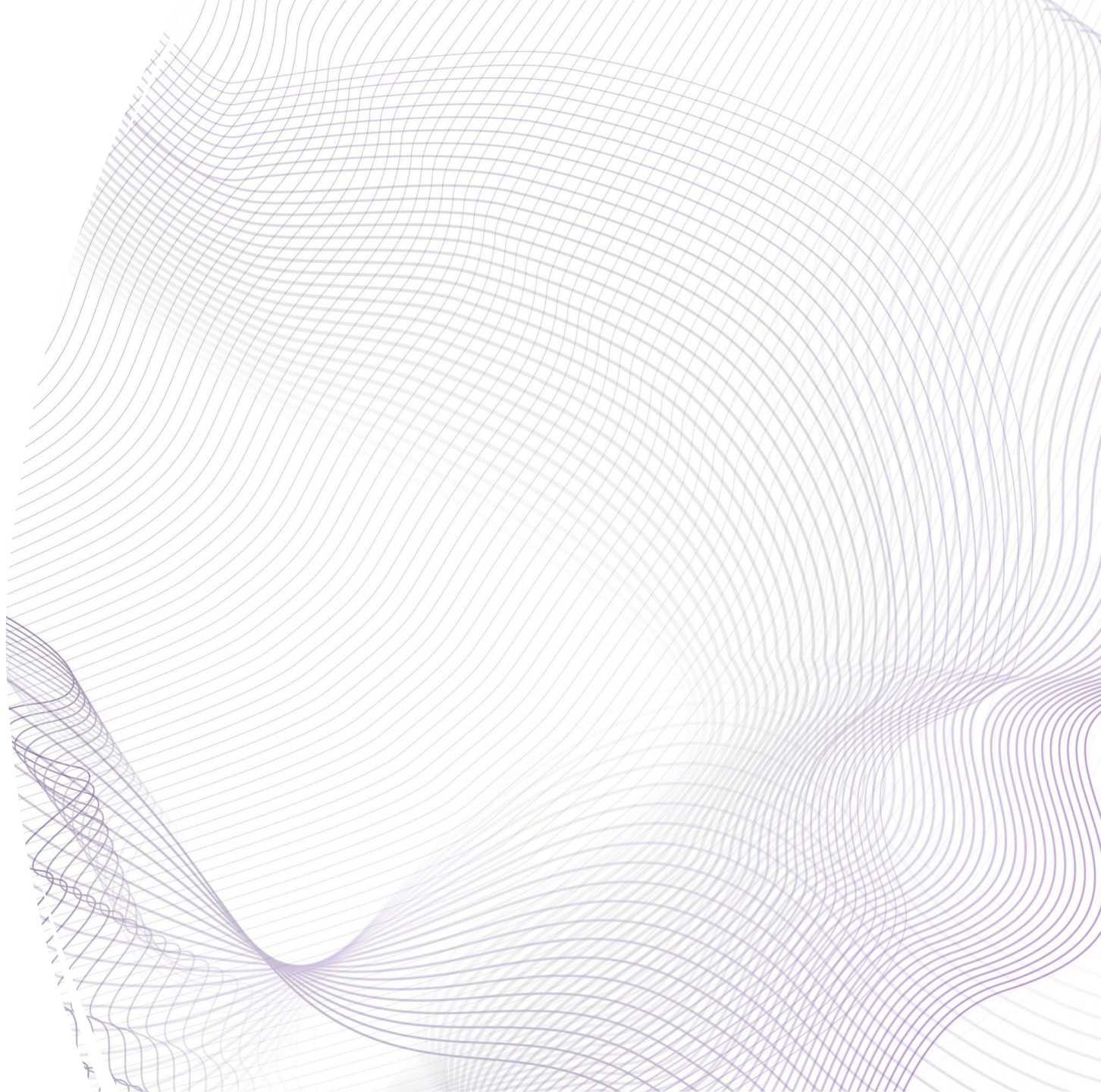


*SPIKING NEURONS*  
*Week 10*

Dr. Luca Manneschi



# So far...

Differential Equations and initial condition. How to find a unique solution?

Solutions to

Look at the exercises

$$\frac{dx(t)}{dt} = \alpha x(t) \quad \begin{array}{ll} \alpha > 0 & \text{Exponential growth} \\ \alpha < 0 & \text{Exponential decay} \end{array}$$

$$\frac{dx(t)}{dt} = \alpha x(t)(1 - x(t))$$

Saturating

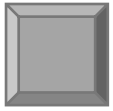
Euler's method and its error  $\mathcal{O}(\delta t^2)$   $\mathcal{O}(\delta t)$

Definition of equilibrium points in one and two dimensions. Phase portraits...

Autonomous and non-autonomous: Definition

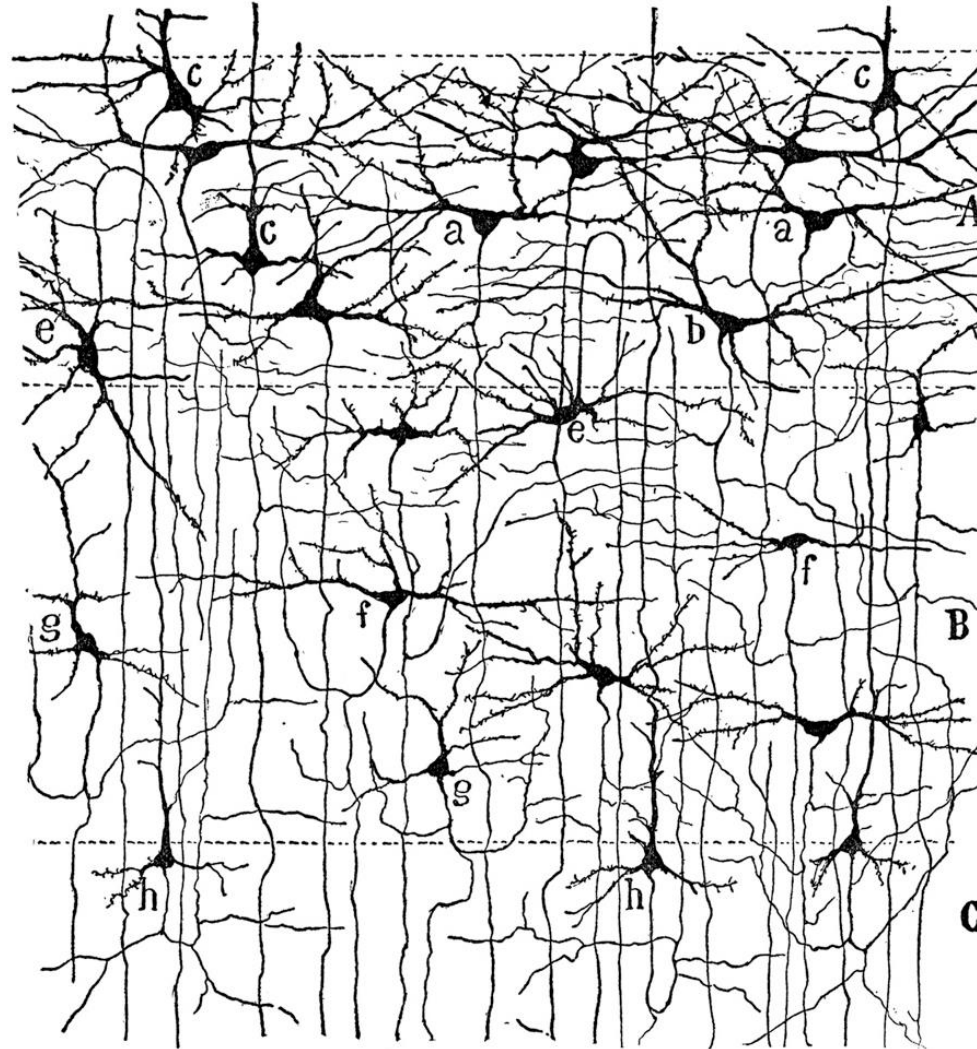
MidPoint Method, Runge-Kutta 2 and 4

# Today...

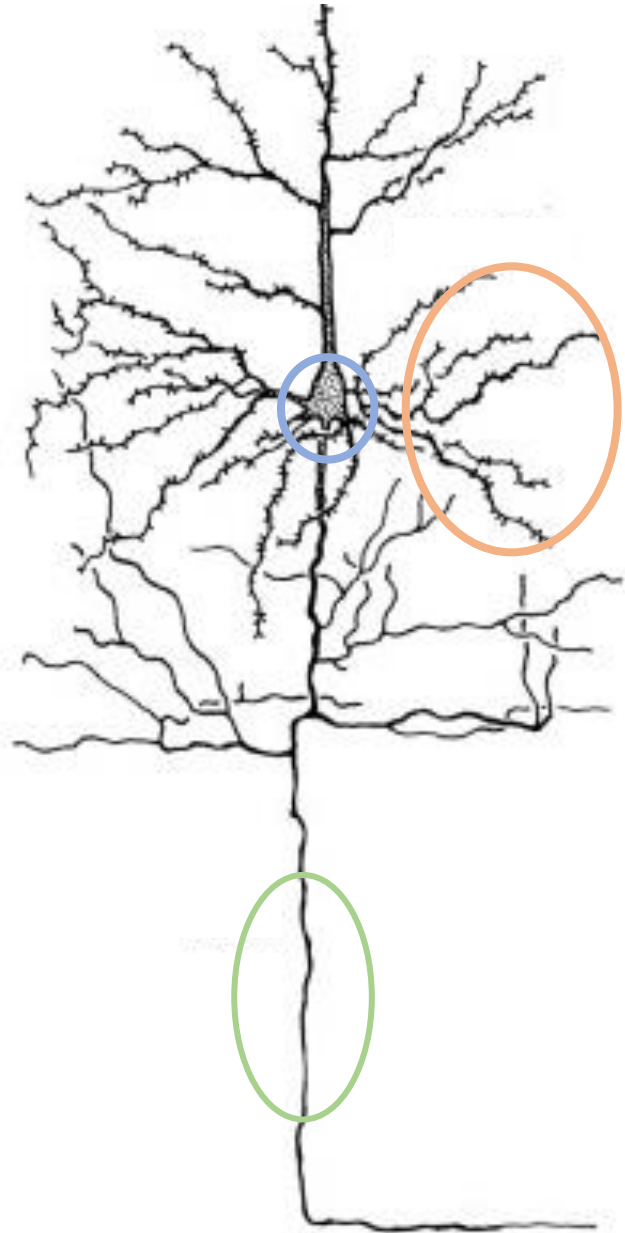


1 mm

~10 000 Neurons  
3 Km of wires



# Neuron: the main parts

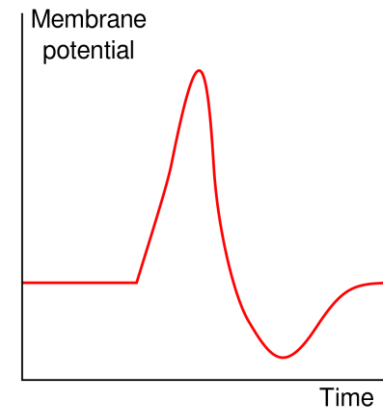


Dendrites: Input device

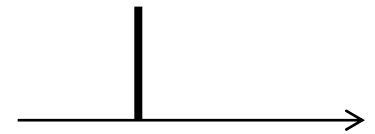
Soma: Central processing unit

Axon: Output

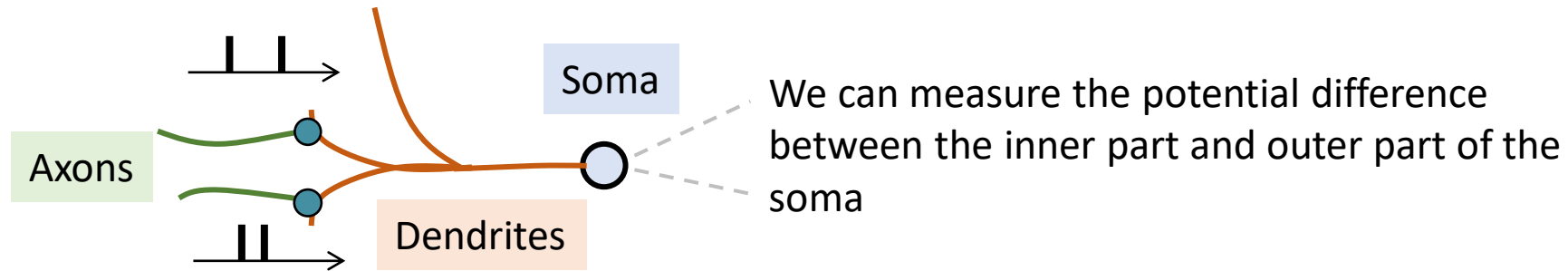
Communication? In a 'stereotype' manner...  
Action potentials



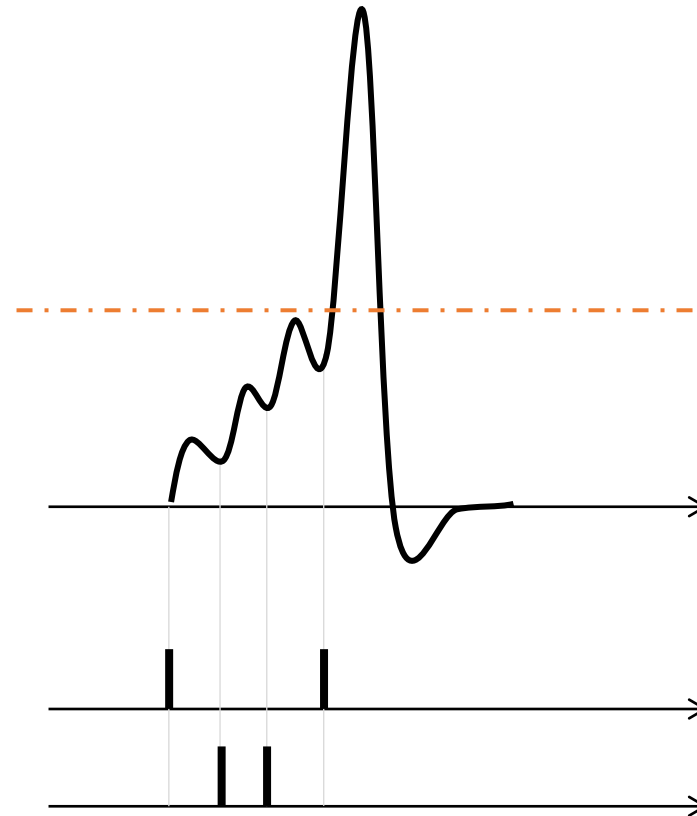
Often we will treat this event as instantaneous



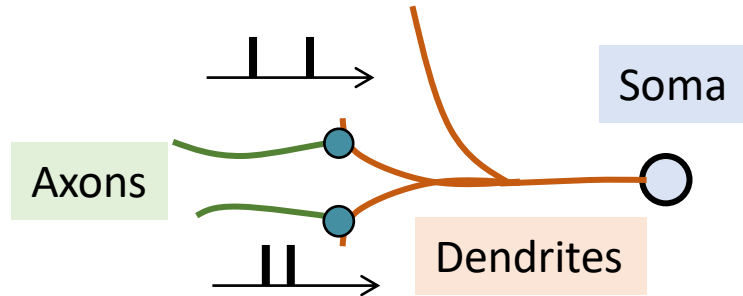
# Neuron: the main parts



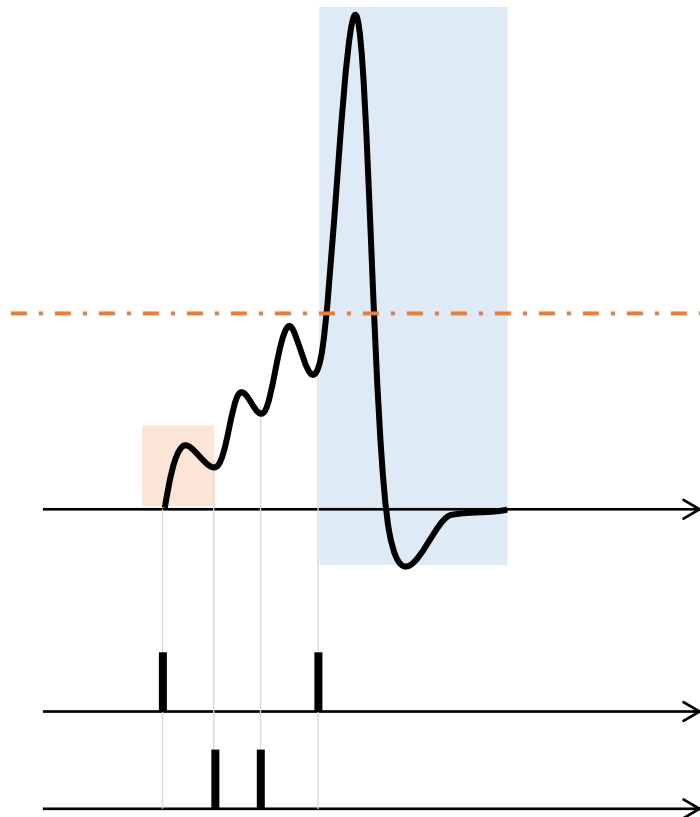
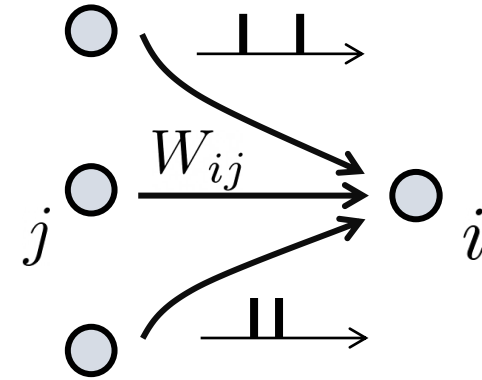
- The junction between neurons is called the synapse



# Response neuron model



We simplify the left diagram considering that there are presynaptic neurons and postsynaptic neurons connected (right diagram)



$$\epsilon(t - t_j^f)$$

Postsynaptic potential: the effect that an input spike has on the potential

$$\eta(t - t_i^l)$$

Spike of the neuron i considered

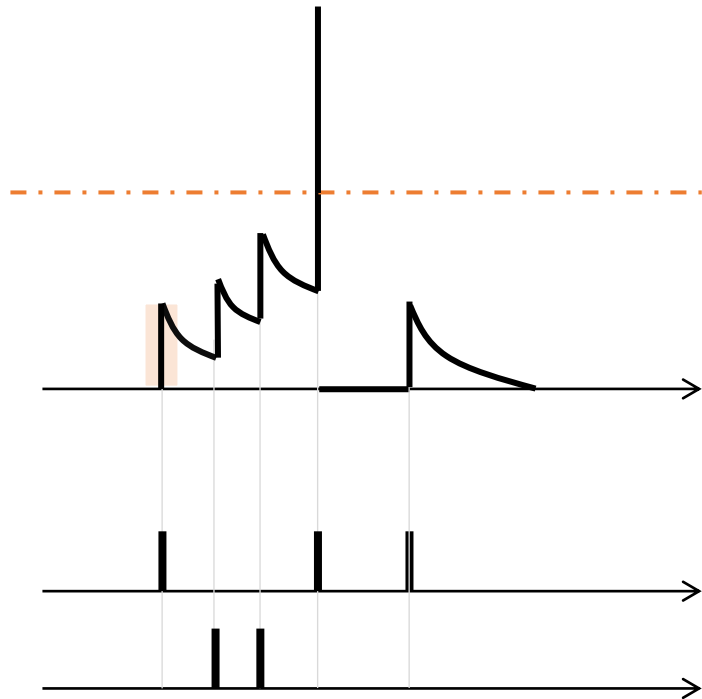
Sum across all presynaptic neurons (j) and all spikes (f)

$$u_i(t) = \eta(t - t_i^l) + \sum_j \sum_f W_{ij} \epsilon(t - t_j^f)$$

$$u_i(t) = \theta, \rightarrow t_i^l = t$$

# Leaky integrate and fire

If we consider that the postsynaptic potential is linear we can define the leaky-integrate and fire model



$$\tau \frac{du_i(t)}{dt} = -u_i(t) + \overset{\text{Input}}{V(t)}$$

$$u_i(t) = \theta \rightarrow \text{Fire + Reset}$$

Does it sound familiar?

Autonomous or not?



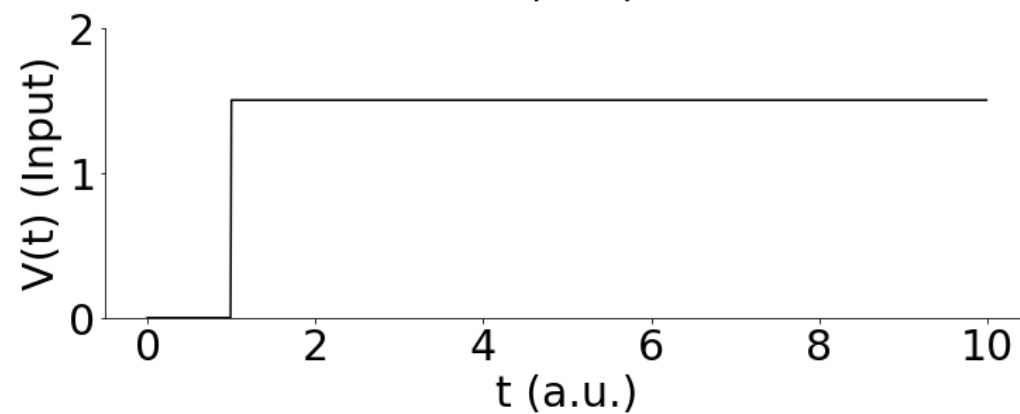
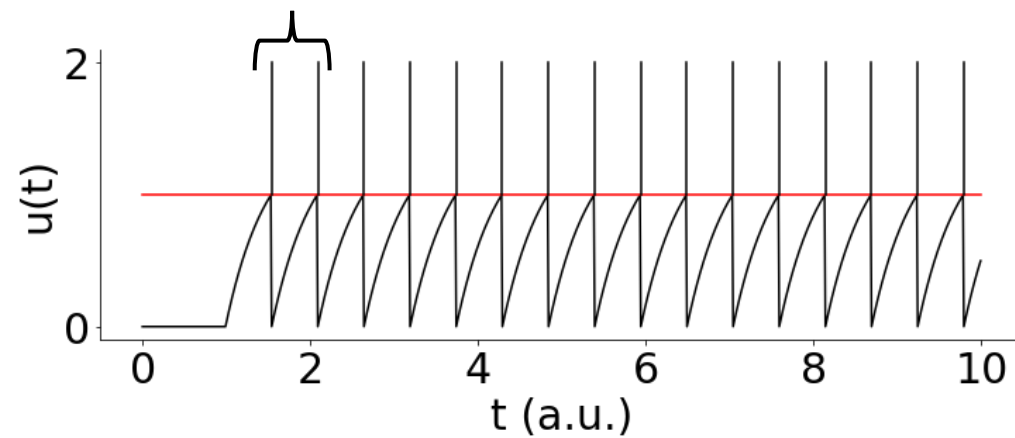
# Response Examples

$$\tau \frac{du_i(t)}{dt} = -u_i(t) + V(t)$$

$$u_i(t) = \theta \rightarrow \text{Fire + Reset}$$

'Constant' Input

Interval between successive spikes



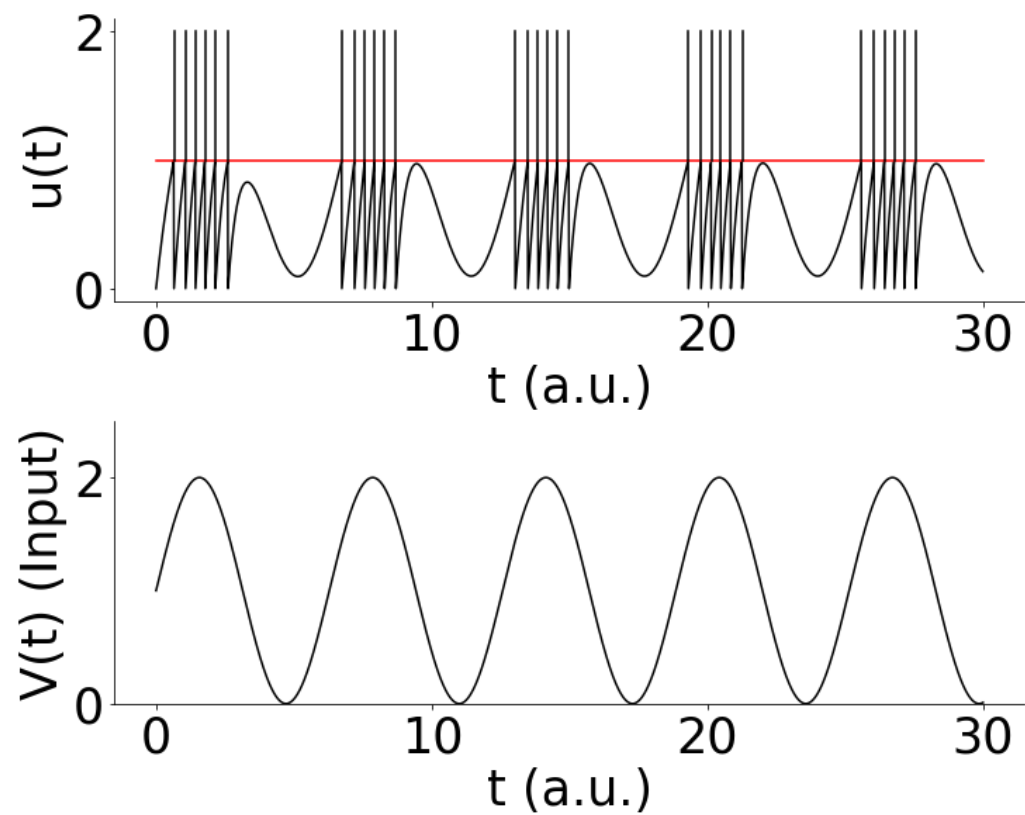


# Response Examples

$$\tau \frac{du_i(t)}{dt} = -u_i(t) + V(t)$$

$$u_i(t) = \theta \rightarrow \text{Fire + Reset}$$

Sinusoidal Input



# Response Examples

$$\tau \frac{du_i(t)}{dt} = -u_i(t) + V(t)$$

$$u_i(t) = \theta \rightarrow \text{Fire + Reset}$$

Poisson spike model

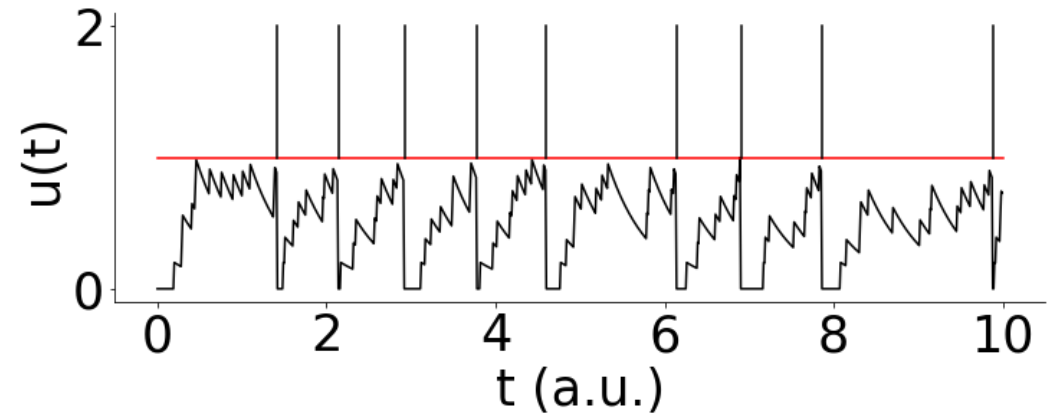
Rate

$$p_i(t) \approx r_i(t) \delta t$$

Probability of firing

Practically, you compute the probability from the rate. Then, generate a random number between  $[0, 1]$ . If the random number is less than the probability the neuron fires, otherwise it does not.

Poisson spike model as Input



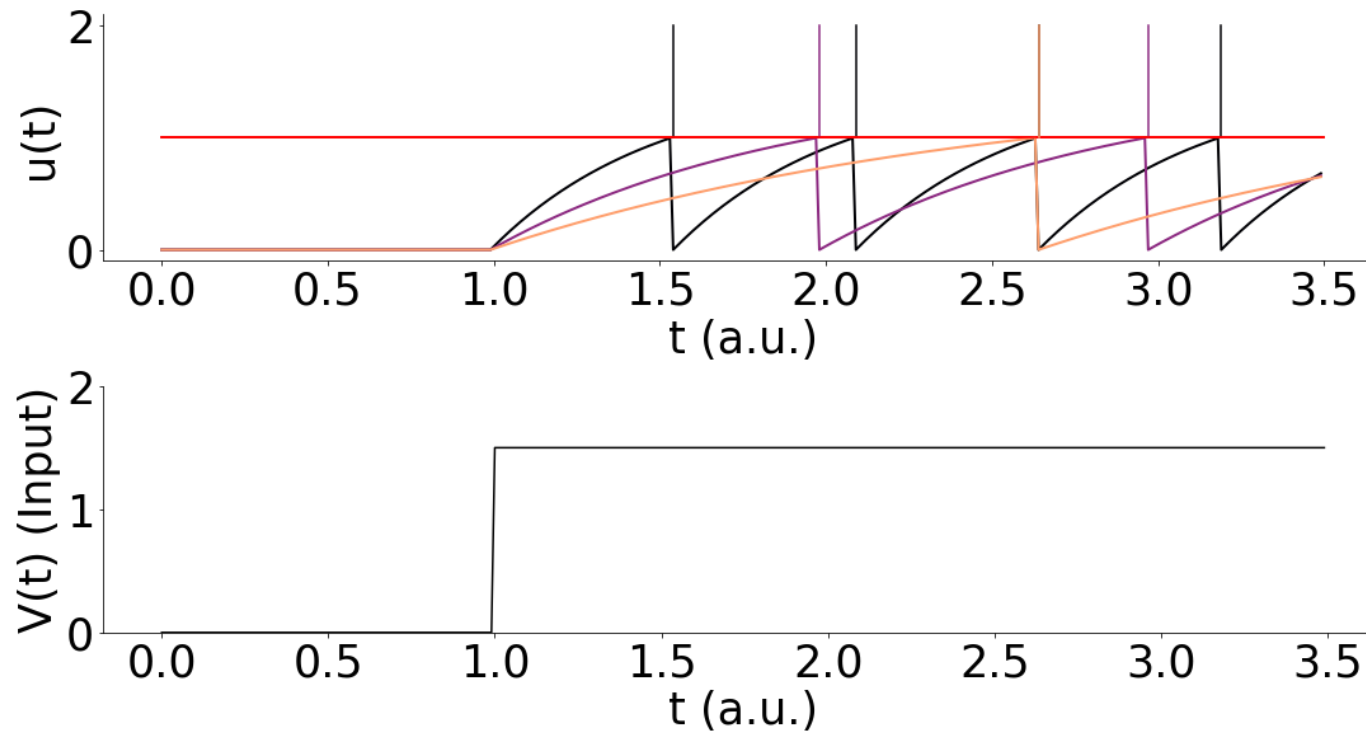
# The meaning of tau

$$\tau \frac{du_i(t)}{dt} = -u_i(t) + V(t)$$

It describes the rate at which information is integrated

$$u_i(t) = \theta \rightarrow \text{Fire + Reset}$$

It increases from darker to brighter colours



## Constant Input, Solution

$$\tau \frac{du_i(t)}{dt} = -u_i(t) + C$$

$$\int \frac{du_i(t)}{-u_i(t) + C} = \int \frac{dt}{\tau}$$

$$-\log(-u_i(t) + C) = \frac{t}{\tau} + k$$

$$-u_i(t) + C = e^{-t/\tau + k}$$

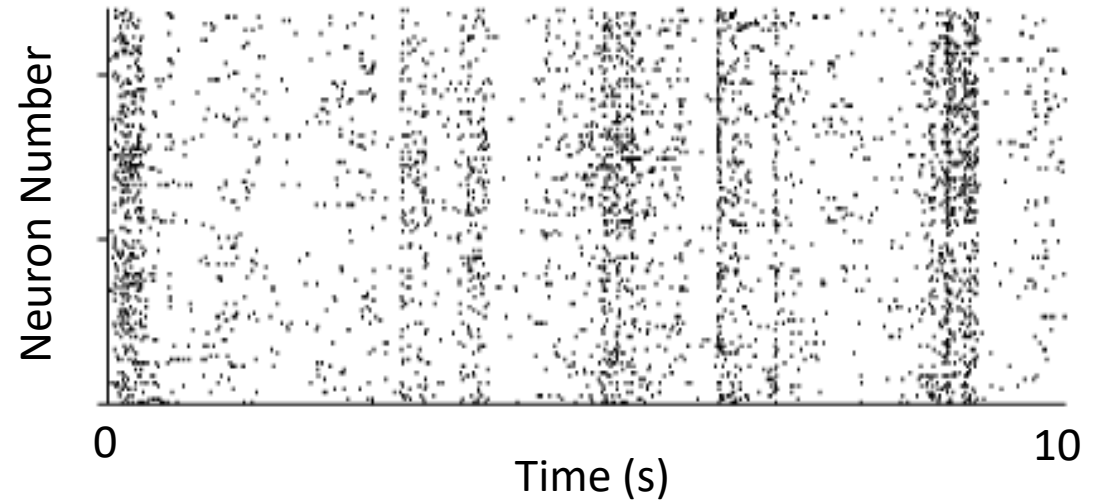
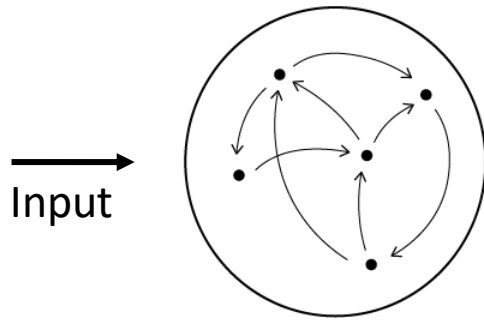
$$-u_i(t) = e^{-t/\tau} e^k - C$$

$$u_i(t) = C - e^{-t/\tau} k_1$$

Can you compute the time at which it reaches the threshold?

# The problem of neuronal coding

We consider a population of neurons



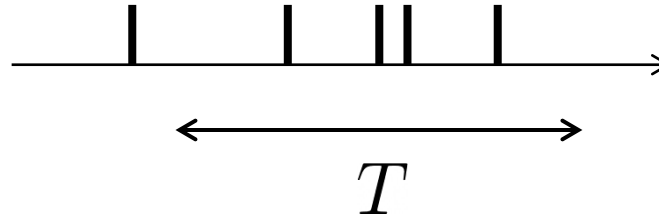
Where is the information?

# The problem of neuronal coding

Rate coding. The information is in the rate at which neurons spike

How can we find the rate? We have spikes (binary information), we need to average...

For a single neuron



Temporal average

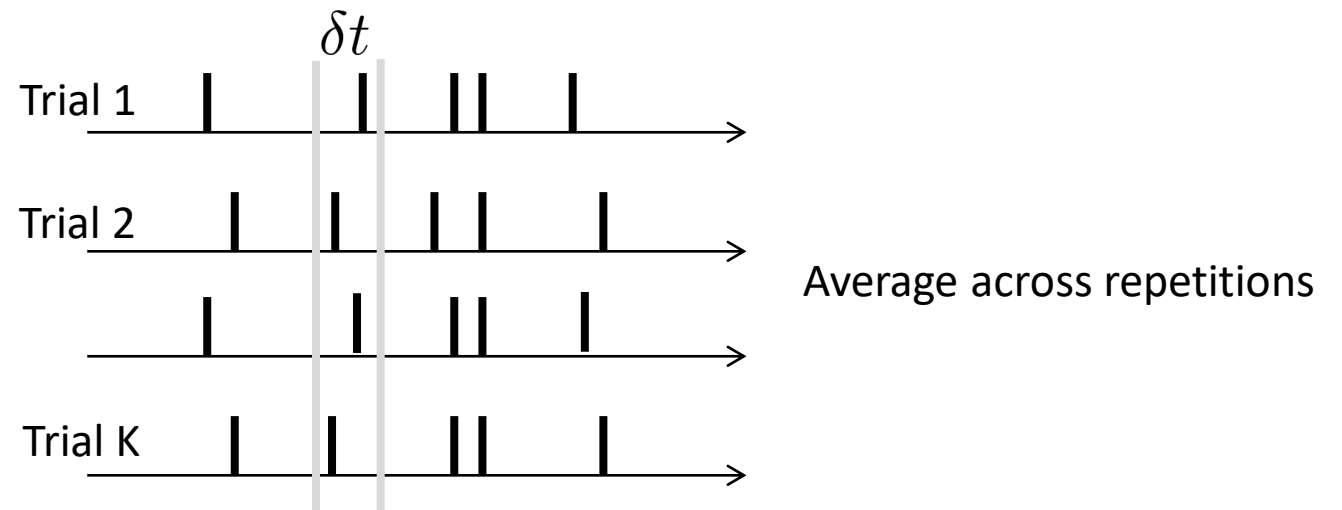
$$\nu = \frac{n(t, t + T)}{T}$$

# The problem of neuronal coding

Rate coding. The information is in the rate at which neurons spike

How can we find the rate? We have spikes (binary information), we need to average...

For a single neuron



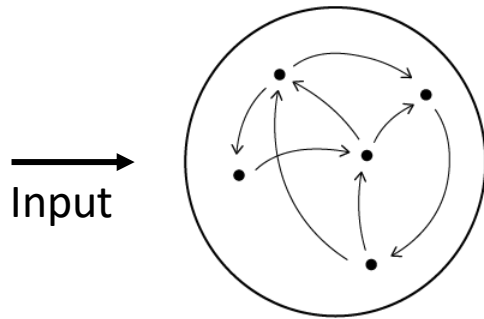
$$PSTH(t) = \frac{n(t, t + \delta t)}{K \delta t}$$

$K$  Number of repetitions



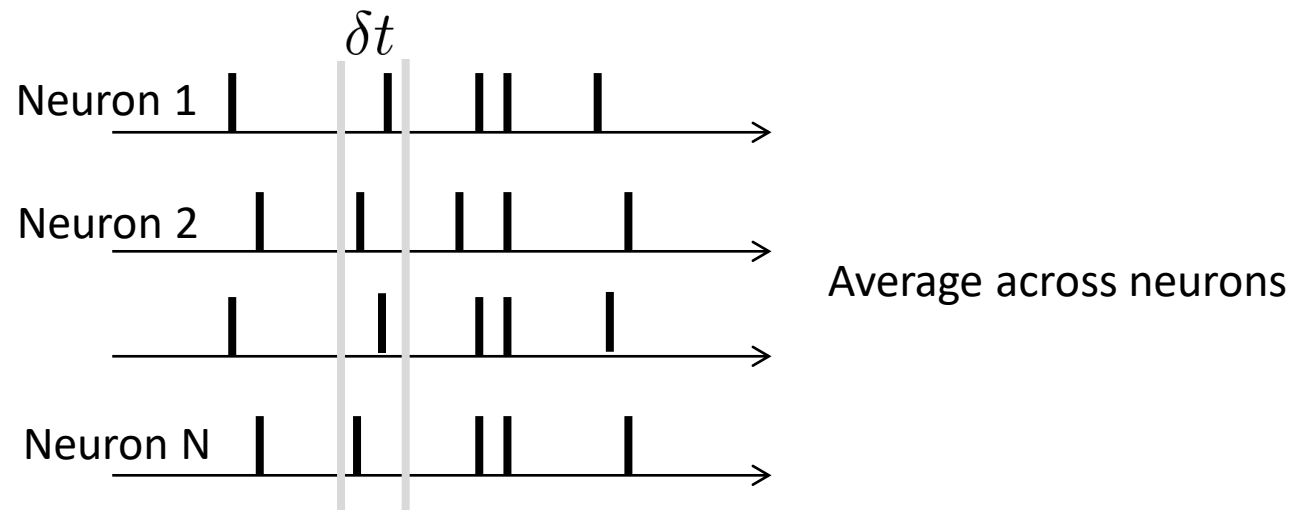
# The problem of neuronal coding

Rate coding. The information is in the rate at which neurons spike



How can we find the rate? We have spikes (binary information), we need to average...

For a population



$$A(t) = \frac{n(t, t + \delta t)}{N \delta t}$$

Number of neurons

# The problem of neuronal coding

The problem with rate coding on a single neuronal level.

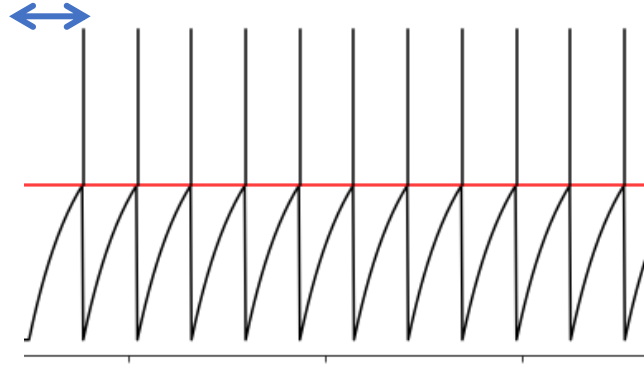
Imagine a frog that catches a fly. The frog reacts in approximately 70 ms. In rate coding we need to average, and averaging takes time. In this activity, there is no time for the considered neuron of the frog to encode the rate across time.



# The problem of neuronal coding

Alternative to rate coding: Temporal coding. Examples:

Time-to-first-spike after input



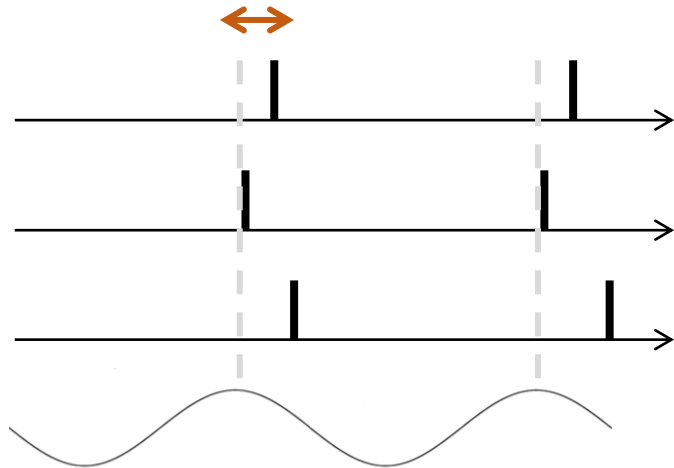
It would need only a single spike to encode this information

# The problem of neuronal coding

Alternative to rate coding: Temporal coding. Examples:

## Phase with respect to oscillation

We know that parts of the brain as the hippocampus are characterized by oscillations. We can then measure the 'first spike' relative to a background oscillation



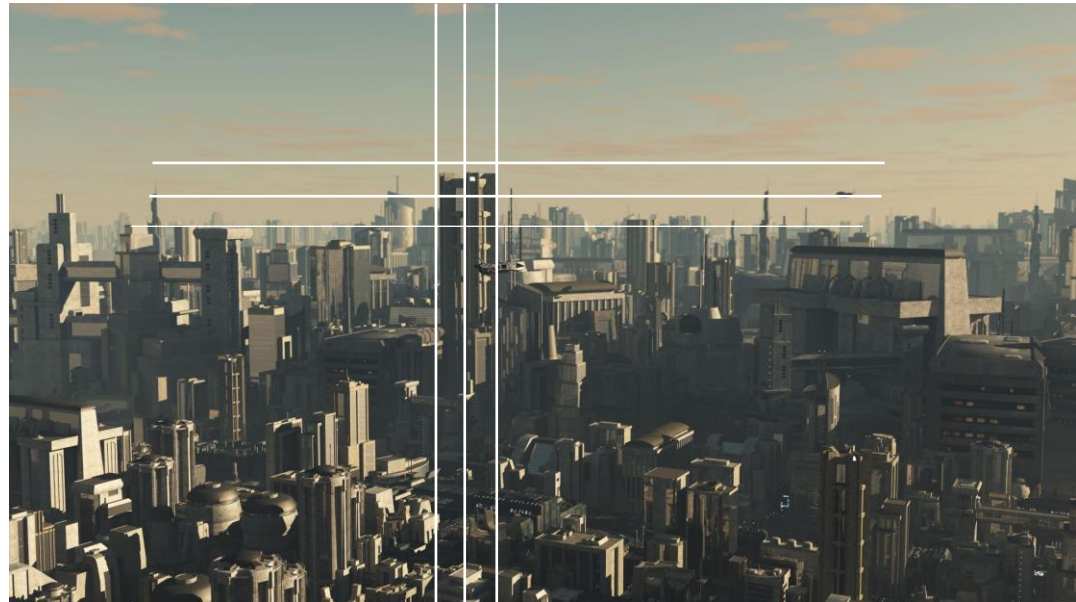
# The problem of neuronal coding

Alternative to rate coding: Temporal coding. Examples:

## Correlations and synchrony



The fact that a pair or an ensemble of neurons have synchronous activity could mean something. For instance, it could mean that they 'belong together'...Imagine that neurons encode different locations in an image. Neurons that have synchronous activity could mean that they are encoding the same object



Thank you