

Welcome!

#Pod_31



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INTRODUCTIONS

Hi! I am [NAME] working as [POSITION] at [AFFILIATION] and

- My research interests are [THIS]
- Or My hobbies are [THIS]

Agenda for the next TWO days


- **Day-#1**

- Leaky Integrate-and-Fire (LIF) Neurons
 - 14 Exercises

- **Day-#2**

- Spiking in LIF Neurons
 - 7 Exercises

Why?



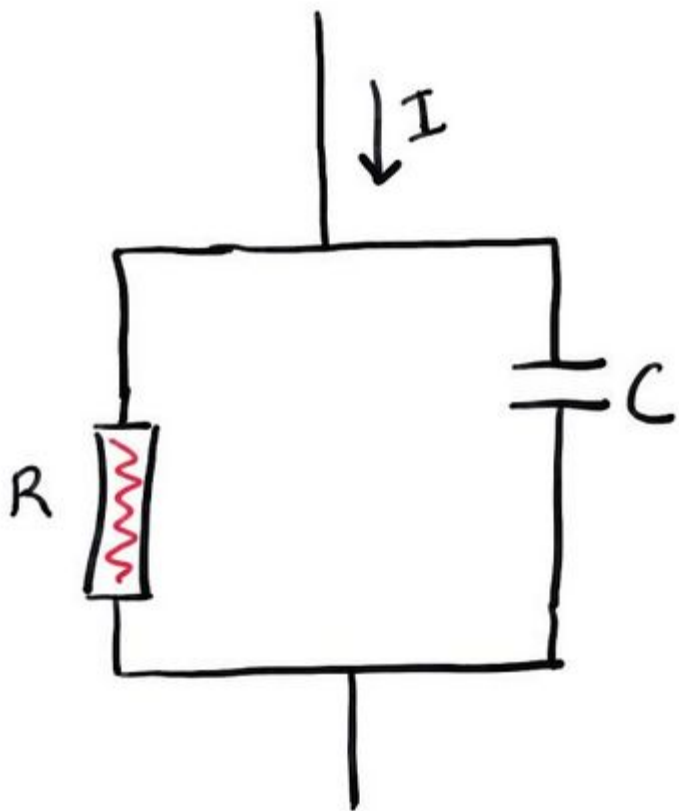
HackerRank

“Learning to write programs stretches your mind, and helps you think better, creates a way of thinking about things that I think is helpful in all domains.”

- Bill Gates

Day #1

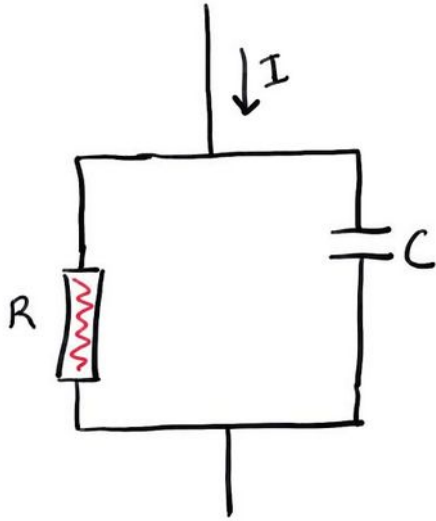
Concept of LIF Neurons



Leaky Integrate and Fire Neurons

Exercise#1

Membrane Equation and Reset Conditions



$$I(t) = IR + IC$$

$$IR = \frac{u}{R}$$

$$IC = C \frac{du}{dt} \quad \text{because} \quad C = \frac{q}{u}$$

$$I(t) = \frac{u(t)}{R} + C \frac{du}{dt}$$

Multiply the equation by R and substitute for TC

$$\tau_m \frac{du}{dt} = -u(t) + RI(t)$$

Spiking Neuron Equation

$$\tau_m \frac{d}{dt} V(t) = E_L - V(t) + R I(t) \quad \text{if } V(t) \leq V_{th}$$

$$V(t) = V_{reset} \quad \text{otherwise}$$

Spiking Neuron Equation

$$\tau_m \frac{d}{dt} \underline{V(t)} = \underline{E_L} - V_t + R \underline{I(t)} \quad \text{if } V(t) \leq \underline{V_{th}}$$

membrane potential

input current

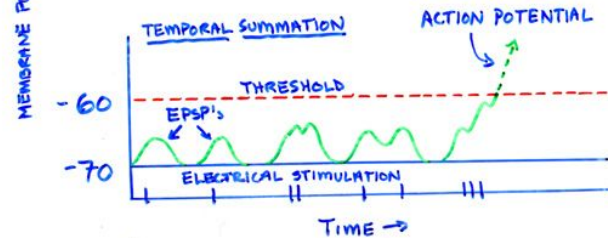
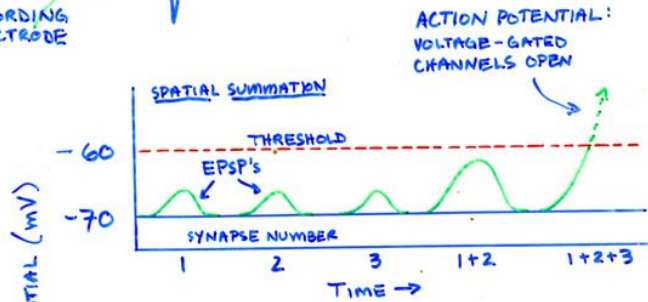
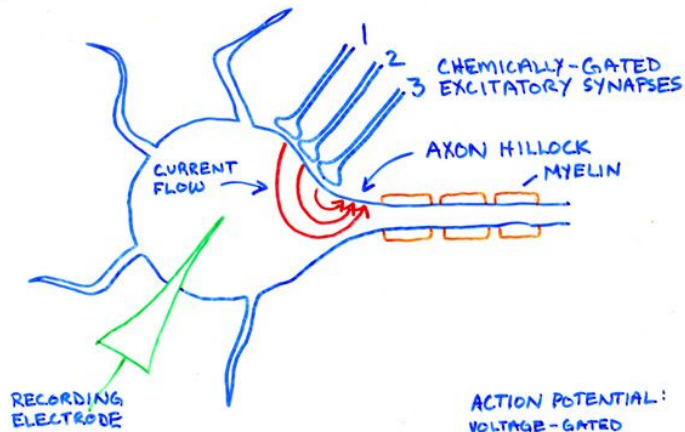
leak potential.

firing threshold

otherwise.

$$V(t) = \underline{V_{RESET}}$$

reset voltage.



Spiking Neuron

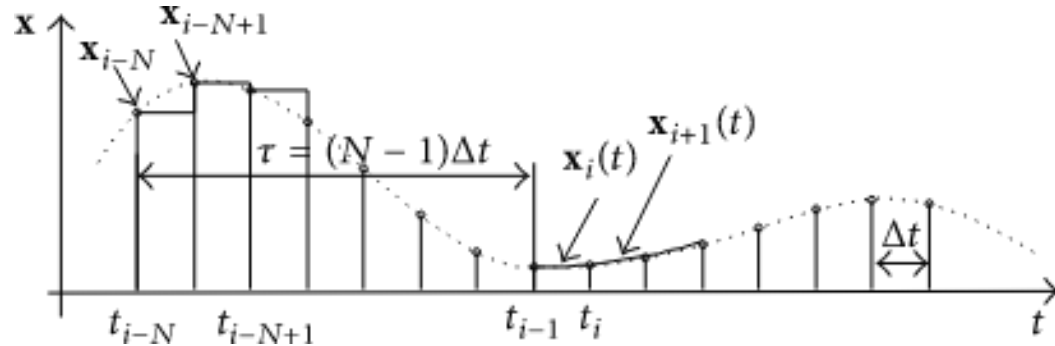
Exercise#2

Applying Sinusoidal Equation

$$V_{(t)} = V_m \sin(\omega t)$$

$$I(t) = I_{mean} \left(1 + \sin\left(\frac{2\pi}{0.01} t\right) \right)$$

Simulating evolution of equation in discrete time steps



Exercise#3, #4, #5, #6

Simulating evolution of equation in discrete time steps

$$\tau_m \frac{d}{dt} V(t) = E_L - V(t) + R I(t)$$

Time derivative of $V(t)$ in the membrane equation without taking the limit $\Delta t \rightarrow 0$:

$$\tau_m \frac{V(t + \Delta t) - V(t)}{\Delta t} = E_L - V(t) + R I(t) \quad (1)$$

Simulating evolution of equation in discrete time steps

$$\tau_m \frac{V(t + \Delta t) - V(t)}{\Delta t} = E_L - V(t) + R I(t) \quad (1)$$

Make $V(t + \Delta t)$ the subject!

$$-V(t) + V(t + \Delta t) = \frac{\Delta t}{T_m} (E_L - V(t) + R I(t))$$

$$V(t + \Delta t) = V(t) + \frac{\Delta t}{T_m} (E_L - V(t) + R I(t))$$

Exercise#7



Synaptic inputs are
random!

Stochastic Differential Equations!

$$I(t) = I_{mean} \left(1 + 0.1 \sqrt{\frac{t_{max}}{\Delta t}} \xi(t) \right) \quad \text{with } \xi(t) \sim U(-1, 1)$$

Exercise#8 - #14

Ensemble Statistics

Multiple runs of the previous exercise may give the impression of periodic regularity in the evolution of $V(t)$

$$\langle V(t) \rangle = \frac{1}{N} \sum_{n=1}^N V_n(t) \quad \text{sample mean}$$

$$\langle (V(t) - \langle V(t) \rangle)^2 \rangle = \frac{1}{N-1} \sum_{n=1}^N (V_n(t) - \langle V(t) \rangle)^2 \quad \text{sample variance}$$

$$\langle (V(t) - \langle V(t) \rangle) (V(s) - \langle V(s) \rangle) \rangle = \frac{1}{N-1} \sum_{n=1}^N (V_n(t) - \langle V(t) \rangle) (V_n(s) - \langle V(s) \rangle) \quad \text{sample autocovariance}$$

Interpretation of Ensemble Statistics

What does “mean” mean when we talk about the brain?

The Average current could tell us about the activation in a certain region

What is the significance of variance and autocovariance?

VARIANCE: expectation of deviation from the mean

Linked with different variables like aging/development/mental degradation disorders.

AUTOCOVARIANCE: covariance at pairs of time points eg: within this range

Covariance tracks the changes in the brain structure over time (relationship of the variable with itself over time)

NOTE: **Variance** refers to the spread of a data set around its mean value, while a **covariance** refers to the measure of the directional relationship **between** two random variables.

Day #2

Agenda for the next TWO days

- **Day-#1**

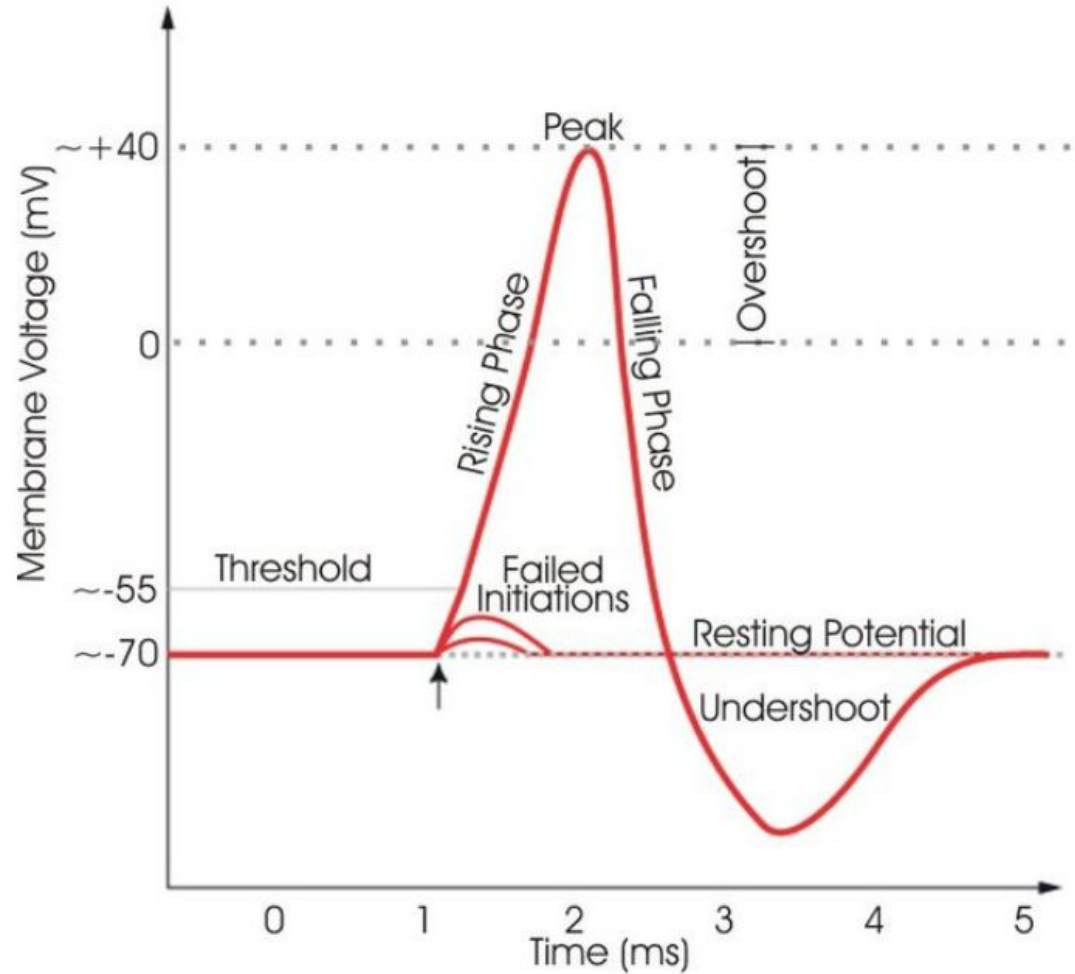
- Leaky Integrate-and-Fire (LIF) Neurons
 - 14 Exercises

- **Day-#2**

- Spiking in LIF Neurons and Refractory Periods
 - 7 Exercises

Introduction to Spiking Neural Networks

**Spiking with time to see
effect on membrane
potential**

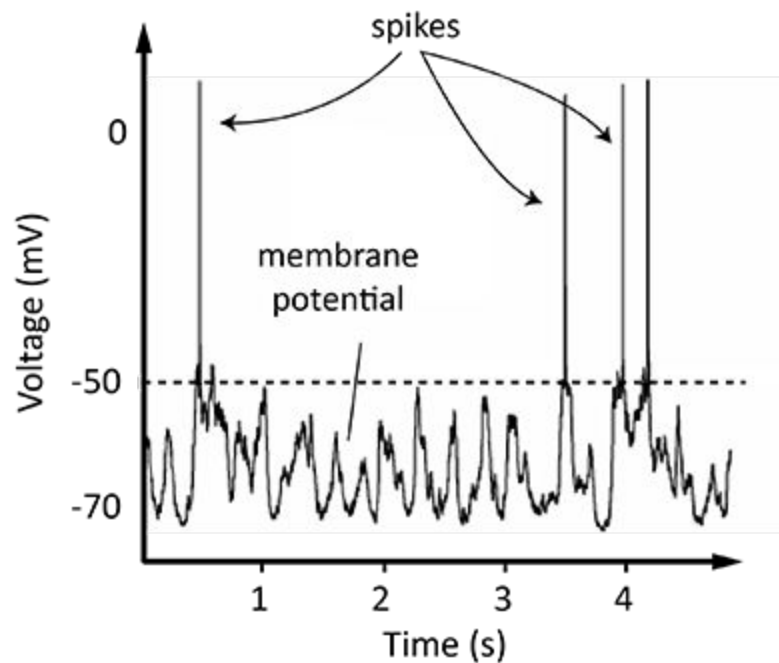
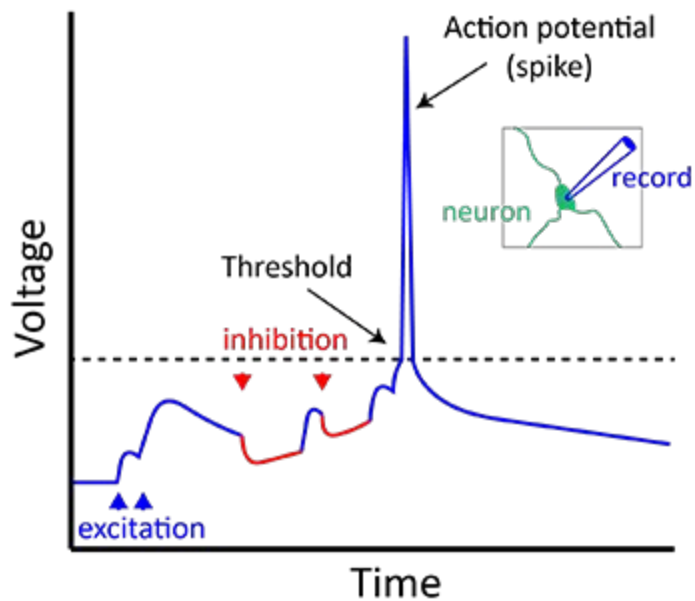


SNN in machines and brains

*Neurons do not fire at each propagation cycle (as it happens with typical multi-layer **perceptron networks**), but rather fire only when a **membrane potential** – an intrinsic quality of the neuron related to its membrane electrical charge – reaches a specific value. When a neuron fires, it generates a signal that travels to other neurons which, in turn, increase or decrease their potentials in accordance with this signal. They receive input from previous layer and signal to subsequent layer.*

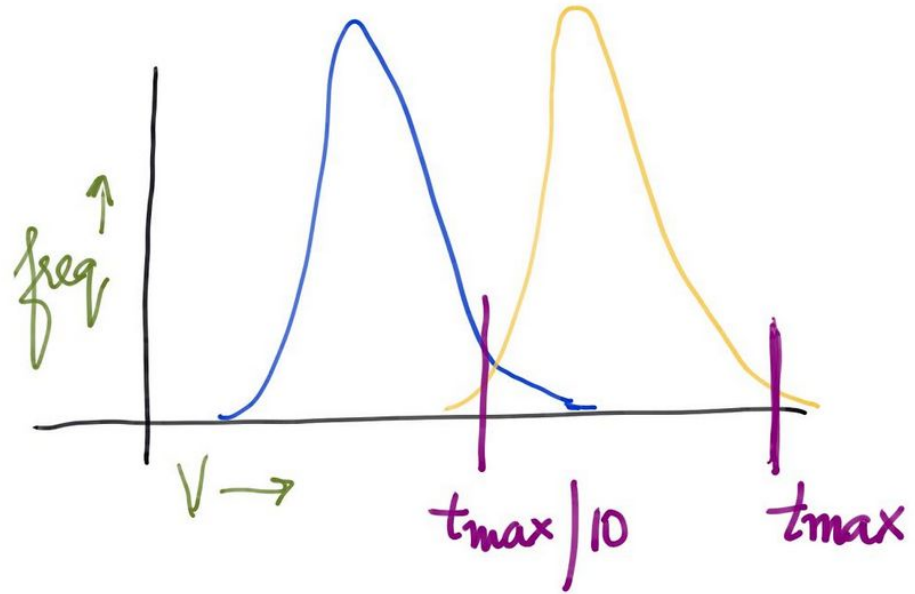
Exercise#1

Spiking in the brain



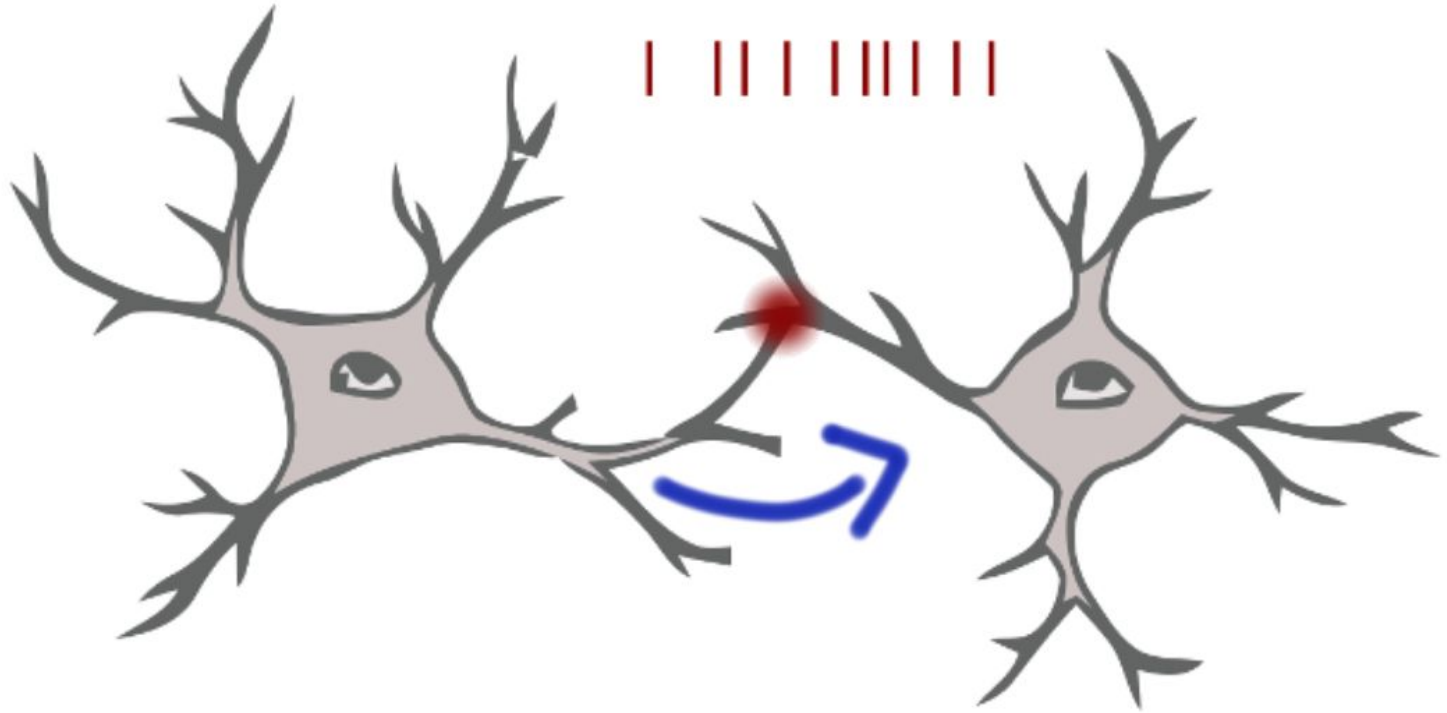
Expected Output:
(With output explanation)

NOTE: these images are only to communicate ideas and hence are not drawn to scale.



Exercise#2

Introducing Spikes



Raster Plots

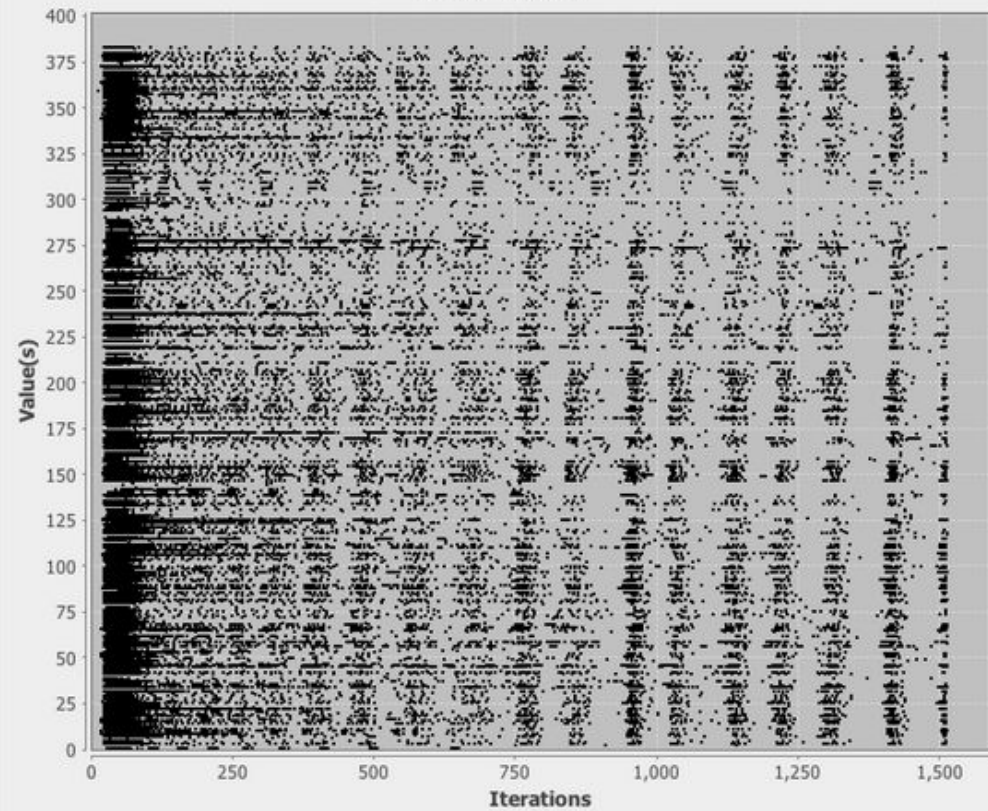
A spike **raster plot** displays the spiking activity of a group of neurons over time (for more information see [this page](#)). In a **raster plot** each row (y-axis) corresponds to the index of a neuron in a neuron group. The columns (x-axis) corresponds to the current time in the simulation.



RasterPlot2

File Edit Help

Time series



• 1 • 2

Remove

Add



Exercise#3

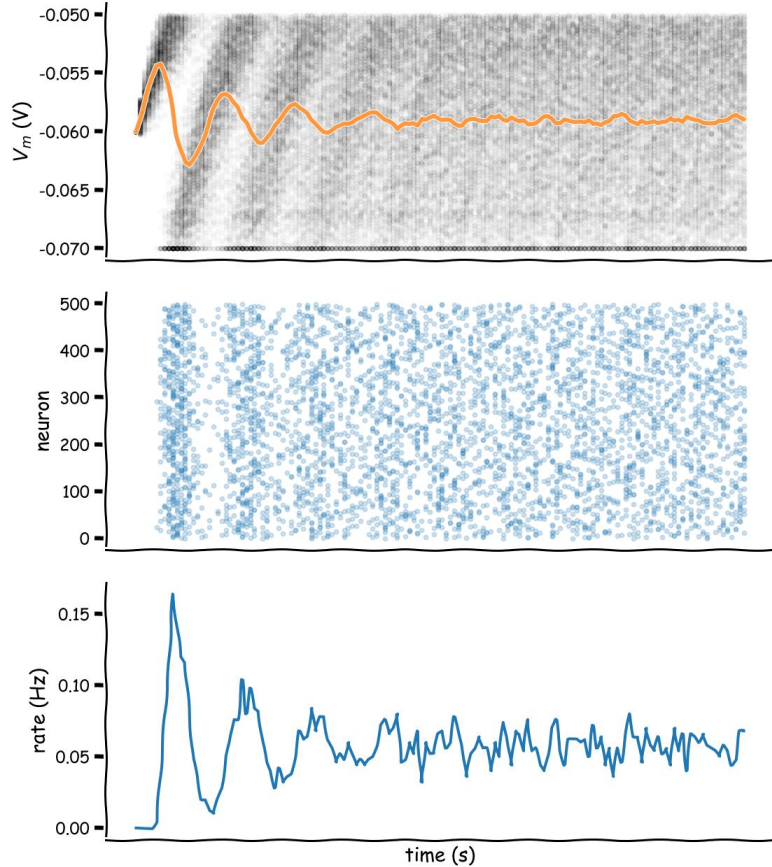
Boolean Indexes

In boolean indexing, we will select subsets of data based on the actual values of the data and not on their row/column labels or integer locations. In boolean indexing, we use a boolean vector to filter the data.

Input:

You're given a dictionary:
{Neuron_no:
[list-of-spike-times]}

Expected Output:



OPTIONAL Fun programming exercise #1

(Impress your TA session)

Try to find out if and how we can get boolean indexes from Bloom Filters?

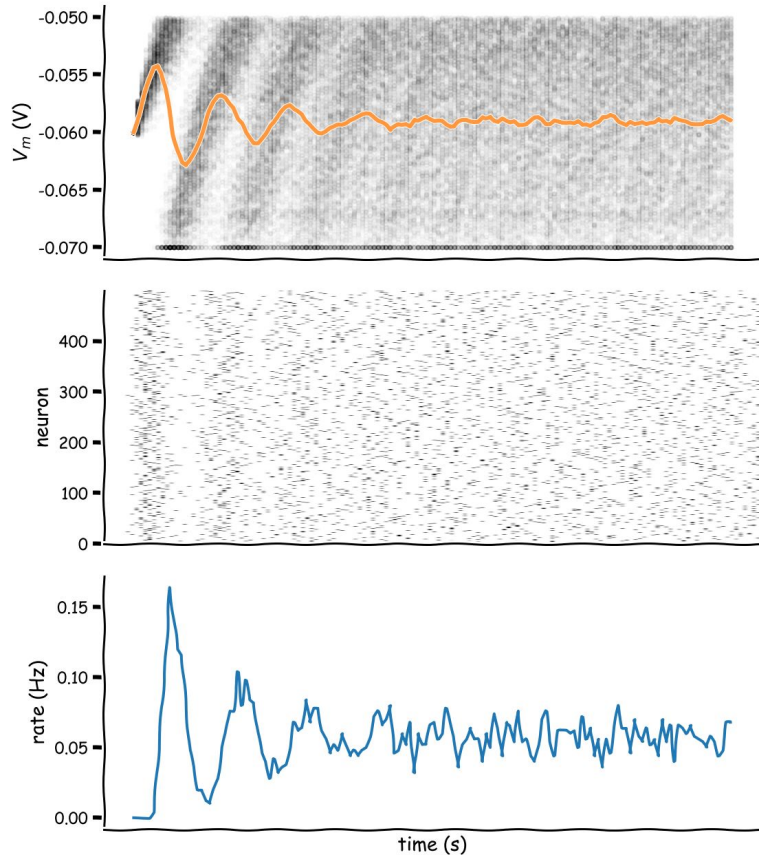
For help on bloom filters: <https://lmlib.github.io/bloomfilter-tutorial/>

Exercise#4

#TASK:

Convert raster plot of
Exercise 3 to binary raster
plot.

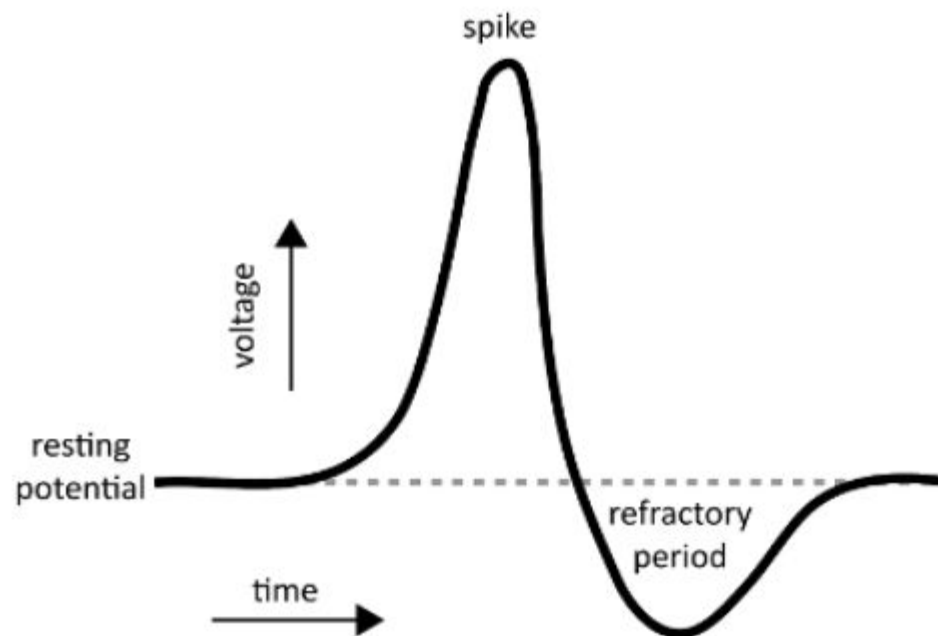
Expected Output:



Exercise#5, #6

Refractory Periods

The absolute refractory period is a time interval in the order of a few milliseconds during which synaptic input will not lead to a 2nd spike, no matter how strong.

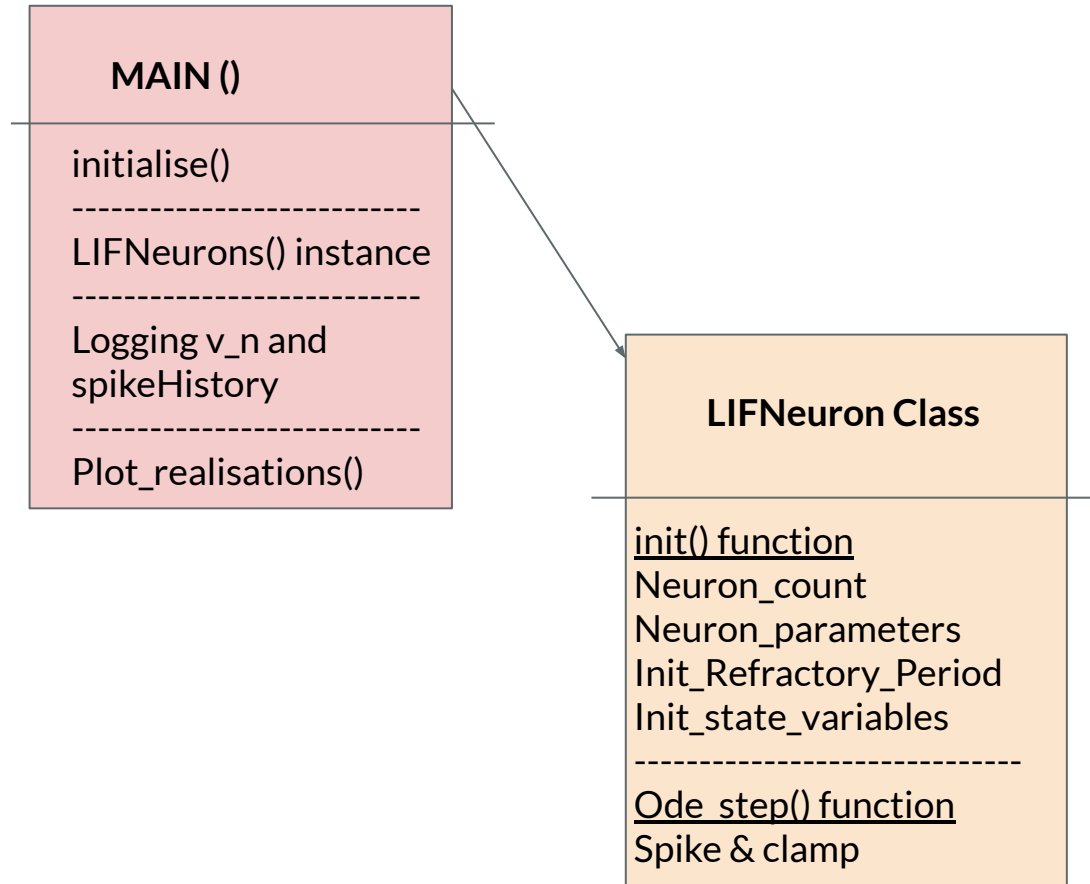


Psychological Refractory Periods in the brain

The term psychological **refractory period** (PRP) refers to the **period** of **time** during which the response to a second stimulus is significantly slowed because a first stimulus is still being processed.

Exercise#7

**Informal class diagram to
help with
implementation:**



NOTE

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