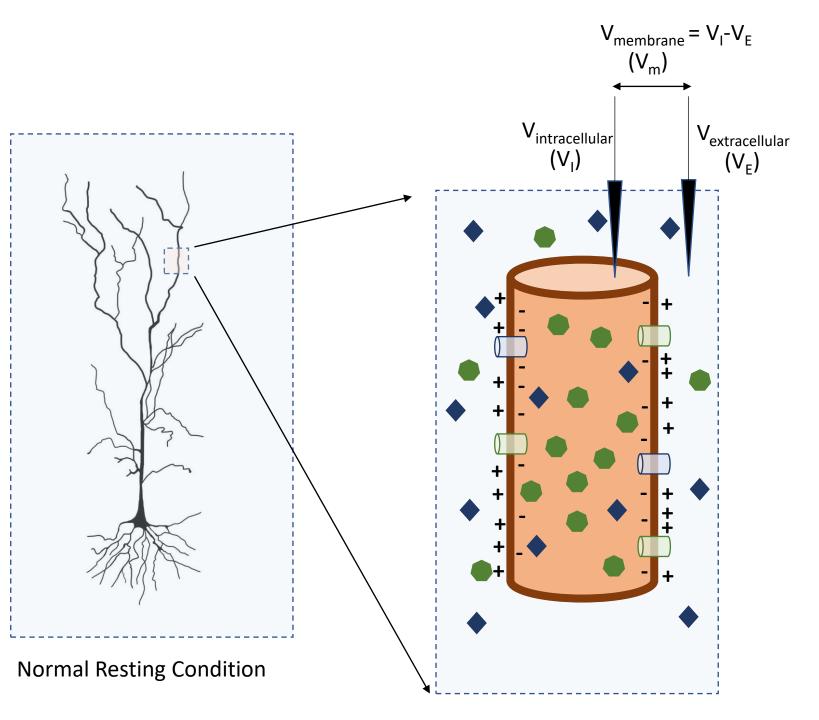
# COMS30017 COMPUTATIONAL NEUROSCIENCE

LECTURE: LEAKY INTEGRATE-AND-FIRE MODEL OF NEURON

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## **Intended learning outcomes**

- Understand the systematic construction of leaky Integrate-and-Fire (LIF) model of the neuron from electrophysiology of neuron
- The working details of the LIF model
- Frequency-current (F-I) response of the LIF model
- Low-pass filtering by the LIF model

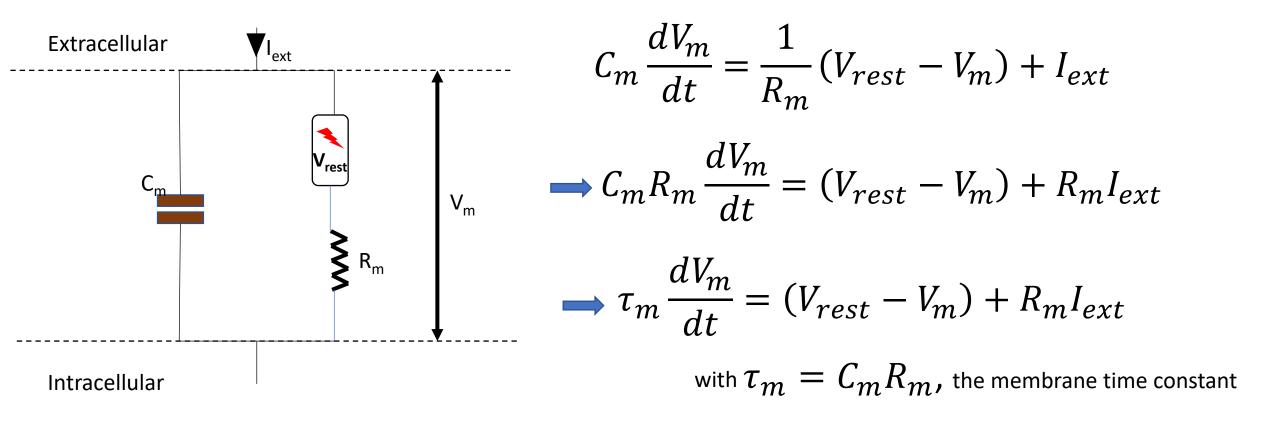


- More Na<sup>+</sup> ◆ outside than inside. Na<sup>+</sup>
  has a strong urge to go in
- More K<sup>+</sup> inside than outside. K<sup>+</sup> has a strong urge to go out
- Inside electric potential is lesser than outside potential, V<sub>m</sub> is –ve
- Membrane as such does not allow the ion flow. It acts as a Capacitor,  $C_m$
- Only ion channels (Na<sup>+</sup> channels and K<sup>+</sup> channels —) would allow flow of ions in an ion-specific manner.
- Channels act as passages for current flow and, thus, have a certain resistance
   R to the current flow.
- Both the channels' conductances are influenced by V<sub>m</sub>: Voltage-gated channels!!

### Voltage-dependent behaviours of Na<sup>+</sup> and K<sup>+</sup> channels: Nonlinear switching

- At normal resting condition with  $V_m$  -60 to -70mV, more K<sup>+</sup> channels are open than Na+ channels: **Resting** membrane potential  $V_{rest}$
- External supply of positive current I<sub>ext</sub> to the inside of neuron makes Vm to reach a threshold potential Vth -50 to -55mV: Reaching the threshold potential
- A sudden nonlinear switching in situation happens with many open Na<sup>+</sup> channels at threshold. Strong flow of Na<sup>+</sup> happen towards inside, making inside more positive than outside: Generation of Action potential.
- After a while (1 ms), Na<sup>+</sup> channels start closing again, plugging the inward Na<sup>+</sup> flow. Open K<sup>+</sup> channels boost loss of +ve ions from inside towards outside, bringing back the V<sub>m</sub> to the resting condition: Reset of the membrane potential.
- The event of action potential generation involves nonlinear switching dynamics in the channels' conductance.
- The famous Hodgkin-Huxley Model deals with the entire dynamics using a complex set of nonlinear differential equations
- Everything at least upto the V<sub>th</sub> follows a very simple linear dynamics: Linear Subthreshold Dynamics

#### Electrical RC circuit model of the Subthreshold Dynamics



$$\frac{dV_m}{dt} = \frac{1}{\tau_m} (V_{rest} - V_m) + \frac{R_m}{\tau_m} I_{ext}$$

**Leak:** making  $V_m$  always to relax back to  $V_{rest}$ 

**Excitation:** making  $V_m$  to rise

from V<sub>rest</sub>

# Some typical magnitudes and units to remember

 $V_m$ : in millivolts (mV)

 $V_{th}$ : -55mV

 $V_{rest}$ : -70mV

 $I_{ext}$ : in pA (pico-Ampere)

Can you check whether  $\tau_m = R_m C_m$  turns out to be 10ms? Particularly, the unit of ms?

C<sub>m</sub>: 100pF (pico-Farad)

 $R_m: 100M\Omega$  (Mega-Ohm)

 $\tau_m$ :10ms

### Leaky Integrate-and-Fire (LIF) Model (1907)

- The leaky integrate-and-fire neuron model has two key components:
- 1. An equation describing the voltage dynamics.

$$\tau_m \frac{dV_m}{dt} = (V_{rest} - V_m) + R_m I_{ext}$$

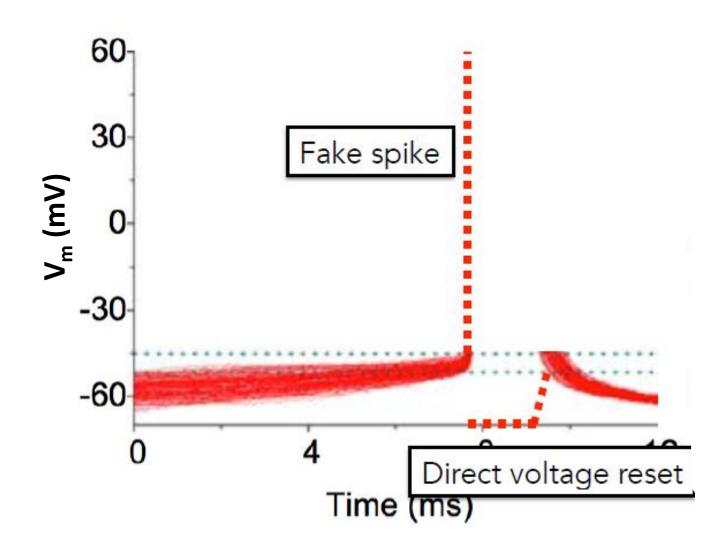


**Louis Lapique** 

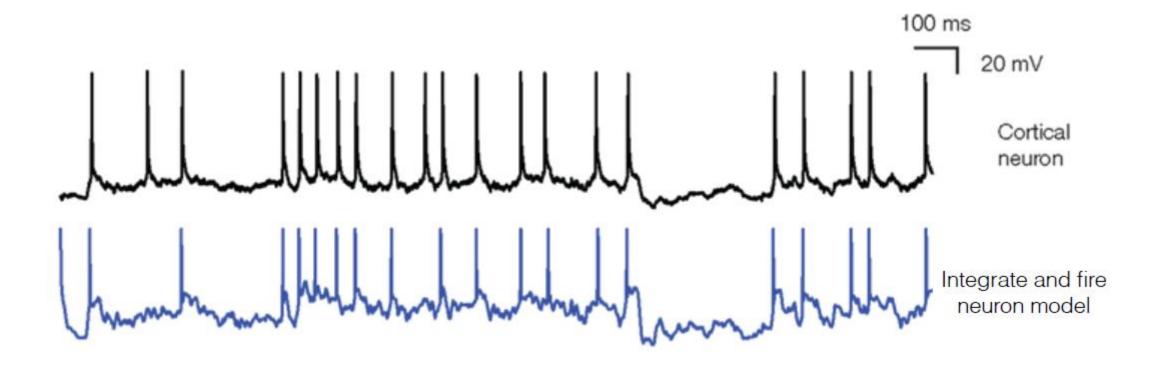
2. A voltage-reset mechanism, mimicking a spike.

If 
$$V_m \geq V_{th} \colon V_m o V_{rest}$$
 , and a spike is realized in an ad hoc manner

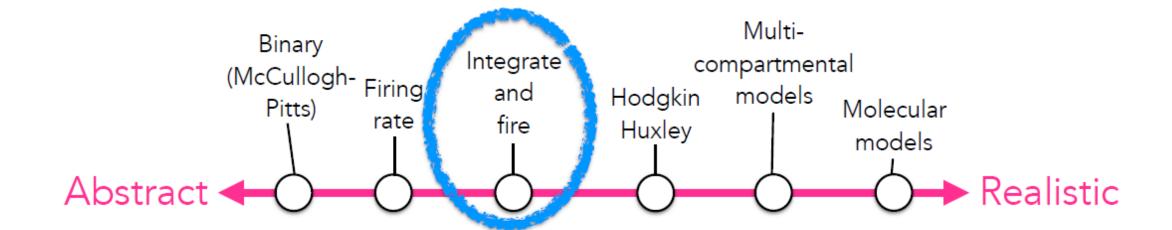
- The name is a bit misleading, the LIF model doesn't actually generate any spikes.
- The LIF is heavily used in computational neuroscience because of its simplicity and analytical tractability.



## Success of Leaky Integrate-and-Fire (LIF) Model



Rossant et al., Frontiers in Neurosci (2011)



#### Abstract models Realistic models

Simple vs Detailed

Hard to relate to biology vs Contains stuff you could measure

Few parameters vs Lots of parameters

Fast simulation vs Slow simulation

Mathematical analysis vs Intractable

Generic vs Specific

# Remember the differential equation lecture for analytical solution !!!

$$\tau_m \frac{dV_m}{dt} = (V_{rest} - V_m) + R_m I_{ext}, \qquad V_m(t = 0) = V_{rest}$$

Where  $I_{ext}$  is a constant input

Analytical Solution 
$$\rightarrow$$
  $V_m(t) = V_{rest} + R_m I_{ext} \left(1 - e^{-\frac{t}{\tau_m}}\right)$ 

Asymptotic Steady State 
$$\rightarrow$$
  $V_m(t) = V_{rest} + R_m I_{ext}$ 

Time to reach the steady state is governed by  $\tau_m$ : Larger is the time constant, slower will be the act of reaching to steady state

# Remember the numerical method lecture for solving ODEs!!!

$$\tau_m \frac{dV_m}{dt} = (V_{rest} - V_m) + R_m I_{ext}, \qquad V_m(t=0) = V_{rest}$$

#### **Euler Method:**

$$V_m(t_0 + \Delta t) = V_m(t_0) + \Delta t \cdot \left(\tau_m \left( \left( V_{rest} - V_m(t_0) \right) + R_m I_{ext} \right) \right)$$

$$V_m(t_0 + n\Delta t) = V_m(t_0 + (n-1)\Delta t) + \Delta t \cdot \left(\tau_m \left( \left( V_{rest} - V_m(t_0 + (n-1)\Delta t) \right) + R_m I_{ext} \right) \right)$$

