

# MODELLING NEURAL SYSTEMS



COMPUTATIONAL MODELLING OF NEURONS AND MICROCIRCUITS

Prof. Ing. Michele GIUGLIANO, PhD

## Simplified models of excitability

### Plan for the day

- Simplification of AP generation: the Integrate-&-Fire model
- The Frequency-Current formula for the Integrate-&-Fire
- Electrodiffusion and ionic concentrations
- The Integrate-&-Fire with spike-frequency adaptation
- Assignment (optional, not compulsory!)
- How good is this model? Families of Integrate-and-Fire...

So far: adding biological **realism** and grounding mathematical descriptions into biophysics

Now: **simplify** the detailed models of neurons into reduced descriptions

Stripping down a complex model to its bare essential may provide an **explanatory model** (easier to *understand*)

Ultimate goal of studying emergence of non-intuitive phenomena in **large networks of neurons** where simpler neurons are *easier/faster* to simulate

*where to stop???*

**The data available.**  
it might (not) be possible to **constraint all parameters** of a complex model

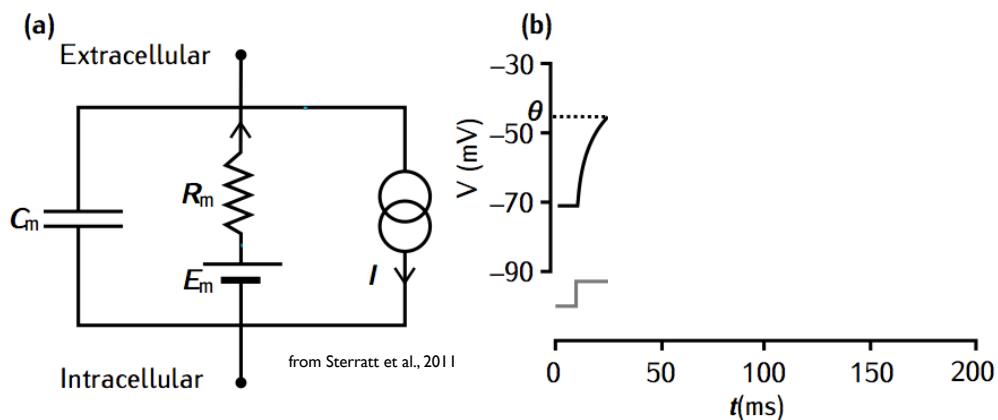
**The desired type of analysis.**  
one might want to investigate and analyse **collective properties** and not single-neuron phenomena

**Computational resources.**  
Simpler models are **faster to simulate** than complex ones.

**The level of explanation.**  
Correspondence between model parameters and physical elements (e.g. models of the appropriate types of ion channels are needed in order to predict what happens to a neuron when a particular neuromodulator is released or a particular type of channel is blocked).

## Integrate-and-Fire model(s)

$$G = 1 / R$$



$$C \frac{dV}{dt} = G (E - V) + I_{ext} \quad \text{if } V < \theta$$

$t^*$  spike,  $V(t^*) \leftarrow H$        $V(t)$  fixed at  $H$ ,  $t \in [t^*; t^* + \tau_{arp}]$

## Integrate-and-Fire model(s)

- there is a (fixed) explicit threshold ( $\theta$ )
- a spike *is said* to occur when there is a threshold crossing
- after a spike, the membrane potential is clamped to  $H$  a (hyperpolarized) level, for a time interval  $\tau_{arp}$

```

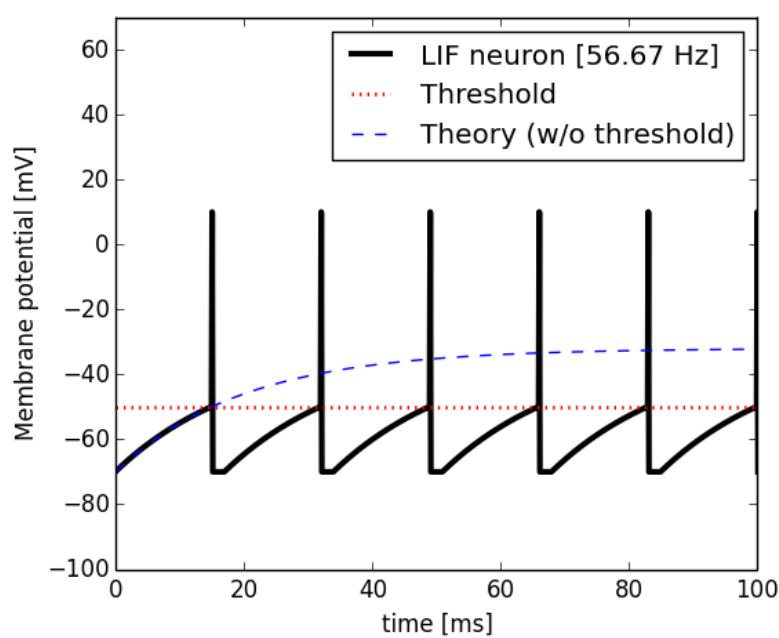
if V >= theta
    V = H
    to = t
elseif t < (to+Tarp)
    V = H
else
    V = V + dt*G/C * (E - V) + dt/C * I
end

```

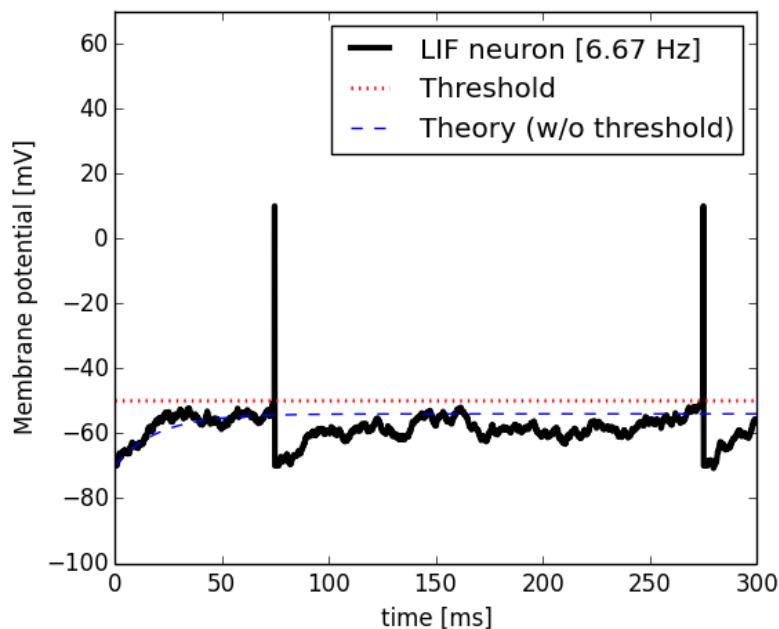


# Demo

## Leaky Integrate-and-Fire model



## Leaky Integrate-and-Fire model

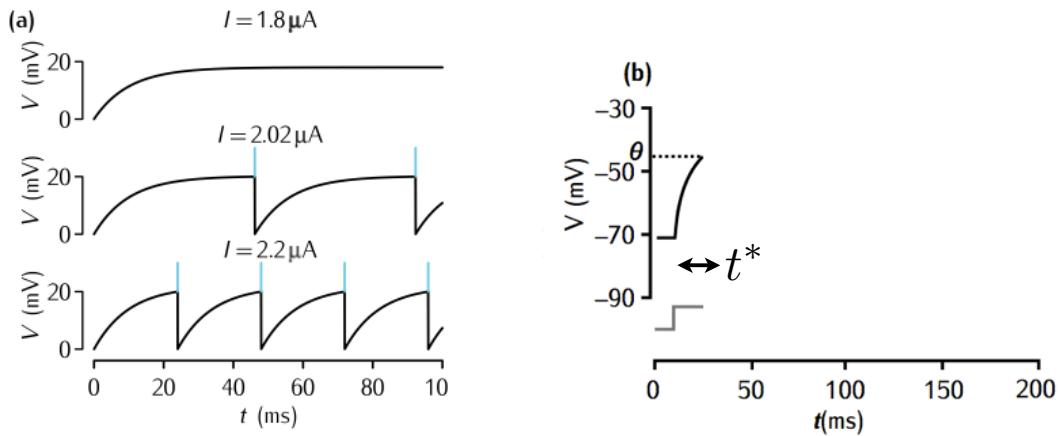


A network of **non-interacting**  
**(identical)** Leaky Integrate-and-Fire neurons

```
for i=1:10
if V[i] >= theta
V[i] = H
to[i] = t
elseif t < (to[i]+Tarp)
V[i] = H
else
V[i] = V[i] + dt/(R*C) * (E - V[i]) + dt/C *I[i]
end
end
```

The figure shows the firing patterns of 10 identical LIF neurons. The y-axis represents the cell number (Cell #) from 0 to 10, and the x-axis represents time in milliseconds (ms) from 0 to 400. Each neuron's firing is represented by vertical tick marks. The neurons exhibit different firing behaviors, including regular rhythmic spiking, irregular firing, and burst-like activity, demonstrating the non-interacting nature of the individual neurons.

## Frequency vs (DC) current curve for a Leaky I&F model



from Sterratt et al., 2011

## Frequency vs (DC) current curve $G = 1 / R$ for a Leaky I&F model

$$C \frac{dV}{dt} = G (E - V) + I_{ext}$$

$$V(t) = E + I_0/G \left( 1 - e^{-G/C t} \right)$$

$$V(t^*) = V_{th}$$

$$t^* = \frac{C}{G} \log \left( \frac{I_0}{G(E - V_{th}) + I_0} \right)$$

$$f(I) = \frac{1}{\tau_{arp} + t^*}$$

## Frequency vs (DC) current curve for a Leaky I&F model

- there is a minimal current (*rheobase*) for spikes to be fired (i.e. for the threshold to be crossed)

$$C \frac{dV}{dt} = G(E - V) + I_{ext}$$

...at the steady-state       $V = E + I_{ext}/G$

$$I_{rhe} = G(V_{th} - E)$$

(or if you like math, then look at where  $\ln(x)$  is defined...)

## Frequency vs (DC) current curve for a Leaky I&F model

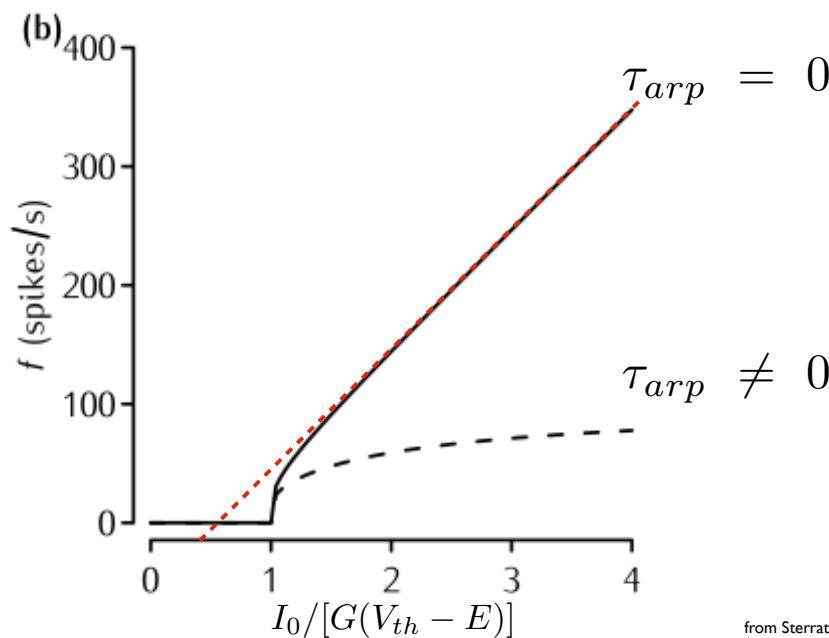
- if we neglect the refractoriness, for very large currents (i.e. far away the rheobase, where you have a non-linearity)...

$$f(I) \approx \frac{1}{t^*} \quad t^* = -\frac{C}{G} \log \left( 1 + \frac{G(E - V_{th})}{I_0} \right)$$

$$t^* \approx -C \frac{(E - V_{th})}{I_0} \quad f(I) \approx \frac{I_0}{C(V_{th} - E)}$$

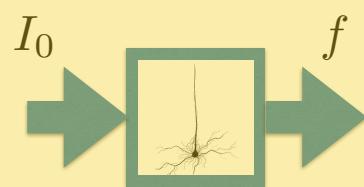
Then, the f-I curve is threshold-linear.

## Frequency vs (DC) current curve for a I&F model



Very rough (functional) approximation

$$t^* = -\frac{C}{G} \log \left( 1 + \frac{G(E - V_{th})}{I_0} \right)$$

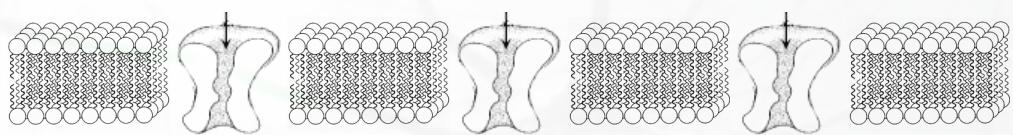


$$f(I) \approx \frac{I_0}{C(V_{th} - E)}$$

A neuron is a device that converts input current (**amplitudes**) into a train of action potential with a certain **frequency**.

# *Electrodiffusion and ionic concentrations*

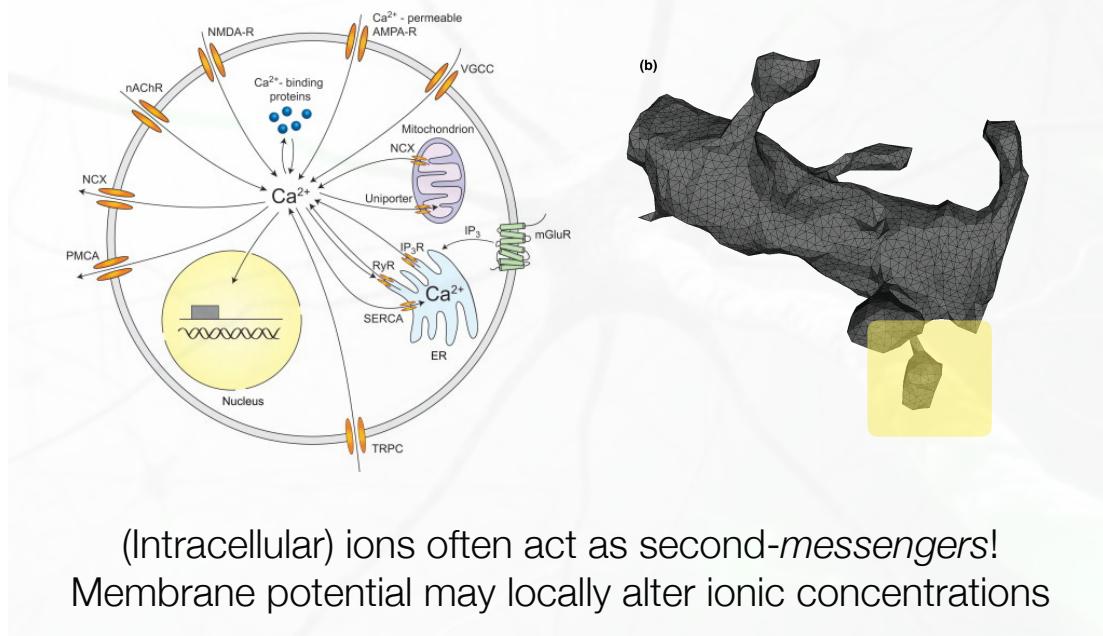
Given the existing concentration gradients, then the individual ionic membrane permeabilities determine  $V$



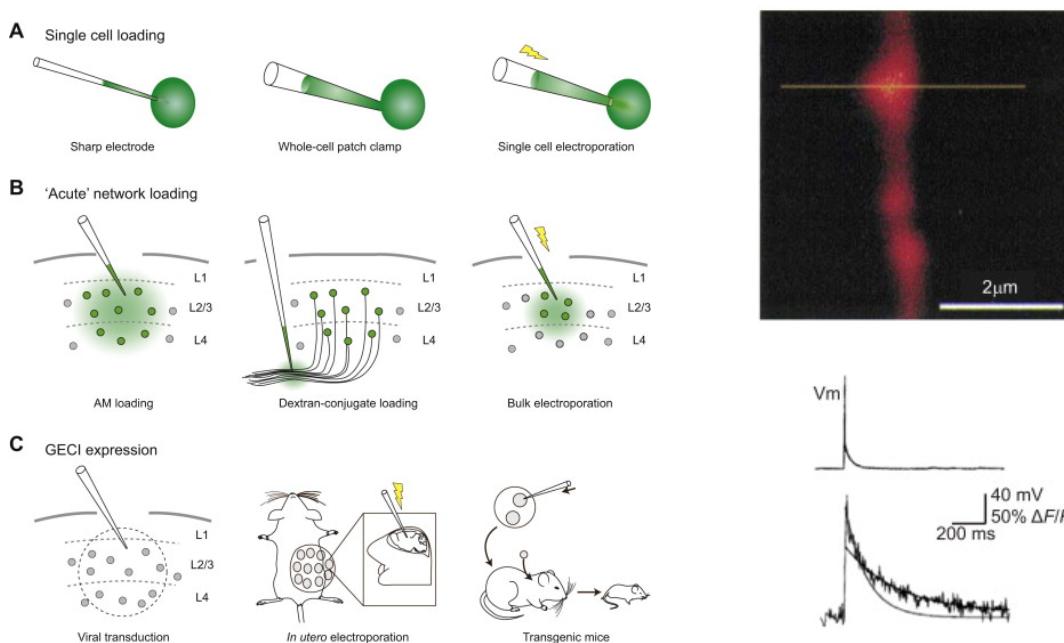
$$i_{tot} = G_1 (V - E_1) + G_2 (V - E_2) + G_3 (V - E_3) + \dots + G_N (V - E_N)$$

$$E_{Nernst} = V_{in} - V_{out} = \frac{R T}{z F} \ln \left( \frac{c_{out}}{c_{in}} \right)$$

Ion currents through the membrane may alter local ionic concentrations, thus alter Nernst potentials!



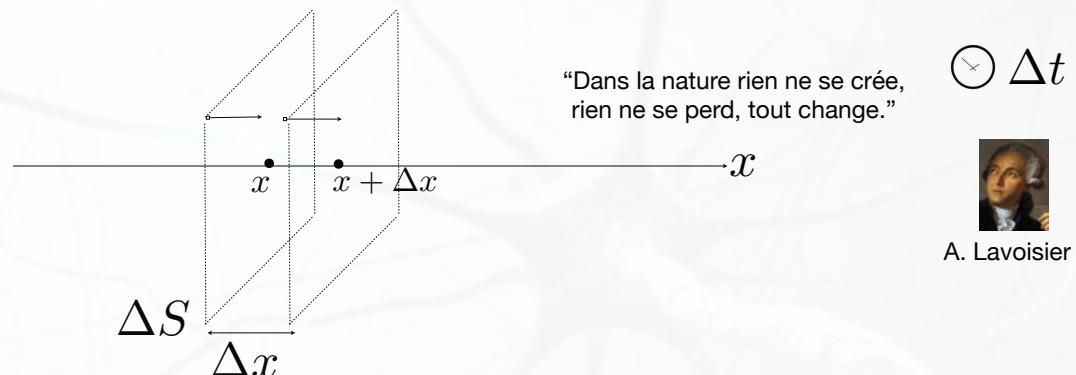
## Fluorophores and fluorescence-imaging of $[\text{Ca}^{++}]$



Grienberger & Konnerth (2012)

## Electro-diffusion equation

invoking conservation of mass for charged particles in aq. solution



$$c(x, t + \Delta t) (\Delta S \Delta x) = \\ c(x, t) (\Delta S \Delta x) + J(x, t) \Delta t \Delta S - J(x + \Delta x, t) \Delta t \Delta S$$

## Electro-diffusion equation

$$c(x, t + \Delta t) (\Delta S \Delta x) = c(x, t) (\Delta S \Delta x) + J(x, t) \Delta t \cancel{\Delta S} - J(x + \Delta x, t) \Delta t \cancel{\Delta S}$$

$$\frac{c(x, t + \Delta t) - c(x, t)}{\Delta t} = - \frac{J(x + \Delta x, t) - J(x, t)}{\Delta x}$$

$\downarrow$   
 $\Delta x \rightarrow 0$   
 $\Delta t \rightarrow 0$

$$\boxed{\frac{\partial c(x, t)}{\partial t} = - \frac{\partial J(x, t)}{\partial x}}$$

$$J = J_{diff} + J_{drift} \quad J = -D \frac{dc}{dx} - u c z F \frac{dV}{dx}$$

$$J_k \approx \frac{1}{z_k F} G_k (E_{Nernst, k} - V_m) \quad \begin{array}{c} \text{---} \\ \text{---} \\ \text{---} \end{array}$$

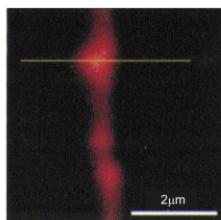
$$\frac{\partial c(x, t)}{\partial t} = - \frac{\partial J(x, t)}{\partial x}$$

↓

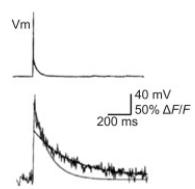
$$\frac{dc}{dt} = \frac{J_k}{vol} + \dots$$



$$\frac{dc}{dt} = \frac{J_k}{vol} - \beta (c - c_{min})$$



$$\frac{dc}{dt} = \frac{J_k}{vol} - \beta (c - c_{min})$$



Grienberger & Konnerth (2012)

$$\frac{dc}{dt} \approx - \beta (c - c_{min})$$

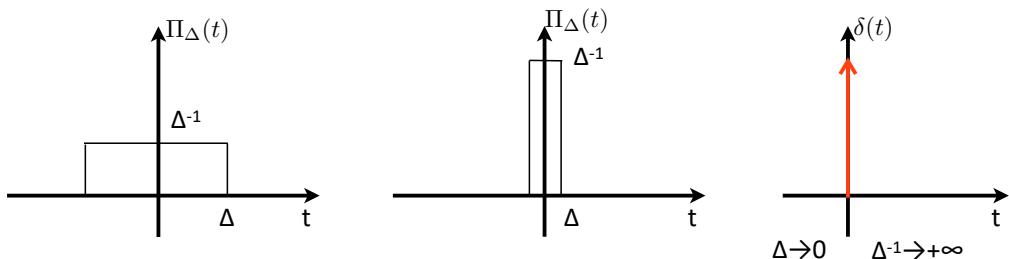
$$c \rightarrow c + \Delta$$

*Blackboard:*  
*prove that on average,*  
*“c” (e.g. [Ca++Ji) is proportional to ~f*  
(MG’s “bank account”)

### Module I: Mathematical preliminaries on the Dirac's Delta function $\delta(t)$

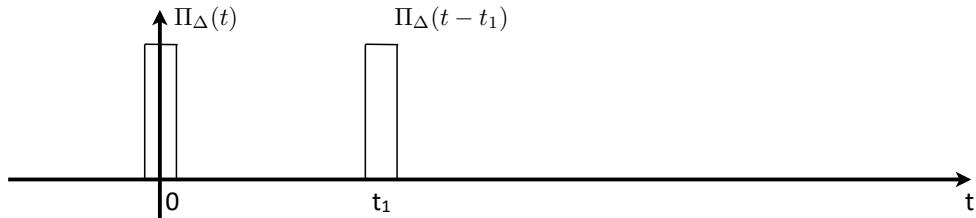
- It is **not** a *conventional* function
- It is defined through the “effect” on a test function  $g(t)$

$$\int_{-\infty}^{+\infty} g(t)\delta(t) dt = g(0) \quad \text{...it extracts the value in } t=0 \text{ of the test function}$$

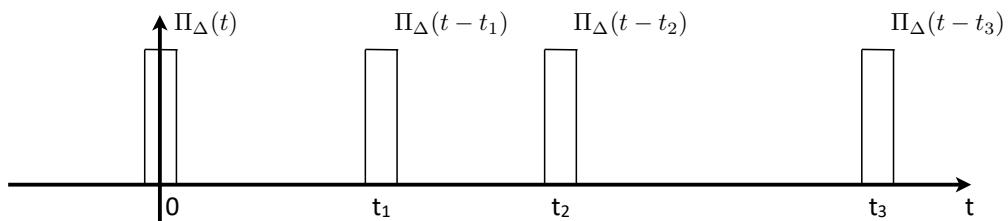


## Module I: Mathematical preliminaries on the Dirac's Delta function $\delta(t)$

- Remember how to “translate a function through time” ?



- and how to sum two functions (graphically) ?



## Module I: Mathematical preliminaries on the Dirac's Delta function $\delta(t)$

(it is like having a “train” of pulses... overlapping)

$$f(t) = \Pi_{\Delta}(t) + \Pi_{\Delta}(t - t_1) + \Pi_{\Delta}(t - t_2) + \Pi_{\Delta}(t - t_3)$$

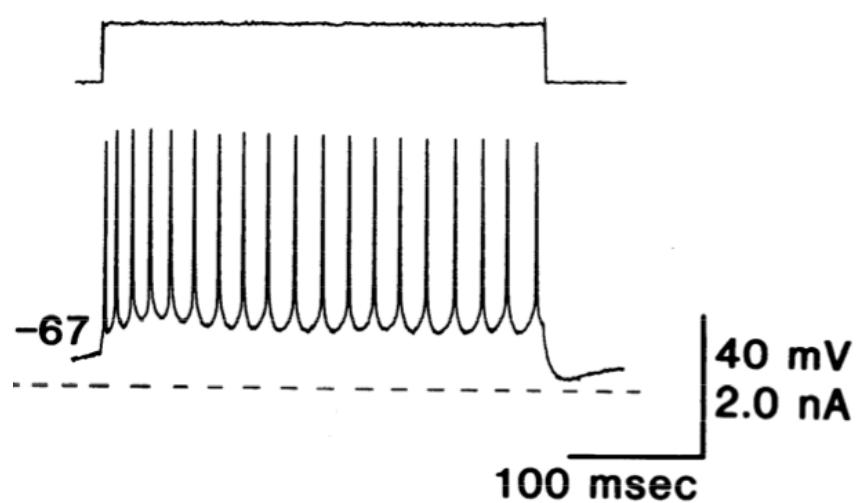
- Dirac's Deltas as the “input” to a first-order o.d.e.

$$\tau \frac{dx}{dt} = -x(t) + a \delta(t) \quad \int_{0^-}^{0^+} \tau \frac{dx(t)}{dt} dt =$$

$$\tau \frac{dx}{dt} = -x(t) \quad t \neq 0 \quad \int_{0^-}^{0^+} -x(t) dt + \int_{0^-}^{0^+} a \delta(t) dt$$

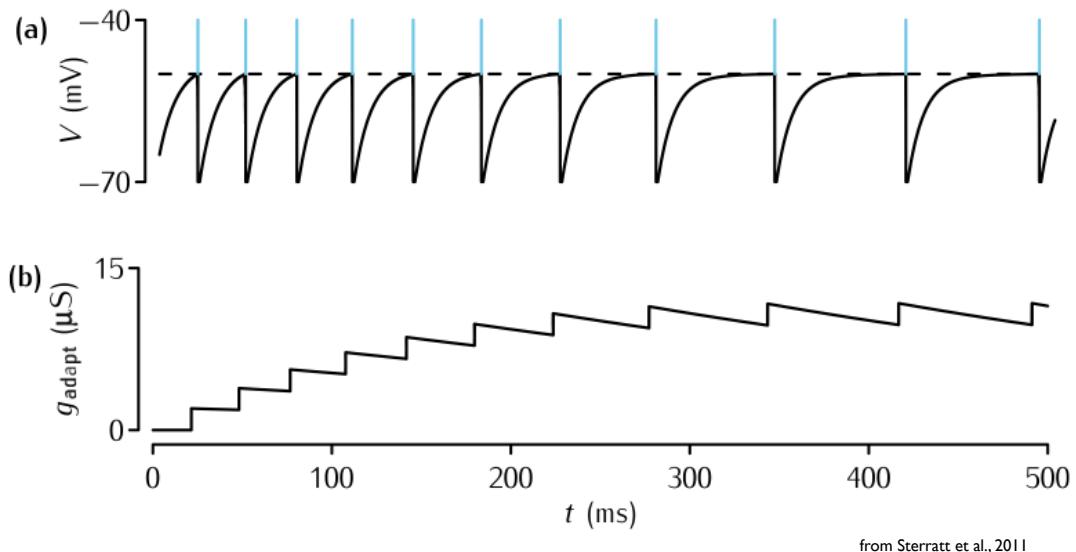
# Spike-frequency *adaptation /* *accommodation*

But... cortical pyramidal neurons do display spike-frequency adaptation!



from McCormick et al., 1985

**Integrate-&-Fire with extra “adaptation mechanism”  
(i.e. spike-frequency adaptation current)**



**Integrate-&-Fire with extra “adaptation mechanism”  
(i.e. spike-frequency adaptation current)**

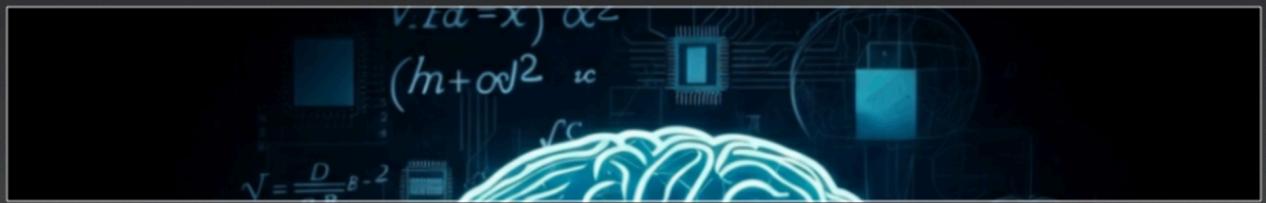
$$I_{\text{adapt}} = \bar{g}_{\text{adapt}} x (E - V)$$

$$\frac{dx}{dt} = -\frac{x}{\tau_{\text{adapt}}} \quad \text{below threshold, if } V < \theta$$

$$x \rightarrow x + \Delta_{\text{adapt}} \quad \text{during a “spike”}$$

$I_{\text{adapt}} \approx -\bar{g}_{\text{adapt}} x \quad \text{approx. equivalent}$

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## Simplified models of excitability

Demo  
(adaptive) IF

**Exercise:**  
implement, simulate, and  
explore the I&F model

(extensive narrative) text/  
discussion/comments  
+  
code and figure(s)

as a Google Colab Notebook

**Exercise:**  
implement, simulate, and  
explore the following points

- Simulate the IF model to plot its F-I curve
- Test whether the analytical formula for  $F(I)$  is accurate
- Discuss possible differences between theory and simulation
- Simulate the IF model + adaptation and plot its F-I curve
- Discuss the effects of the adaptation near the rheobase
- Discuss the effects of the adaptation far from the rheobase

**Exercise:**  
implement, simulate, and  
explore the following points

Parameters - given

$$1/G_m = R_m = 10 \text{ } k\Omega$$

$$C_m = 1 \text{ } \mu F$$

$$\theta = -50 \text{ mV}$$

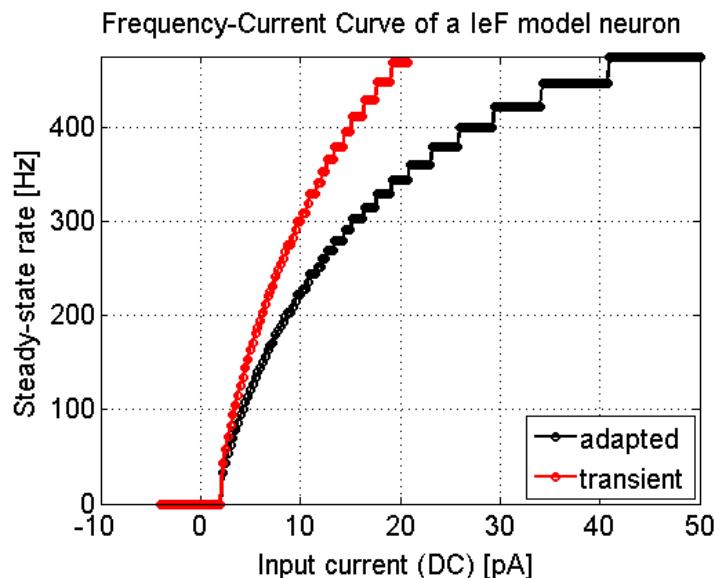
$$E = -70 \text{ mV}$$

$$\tau_{arp} = 1 \text{ ms}$$

Parameters - to explore:

$$g_{adapt}, \Delta_{adapt}, \tau_{adapt}$$

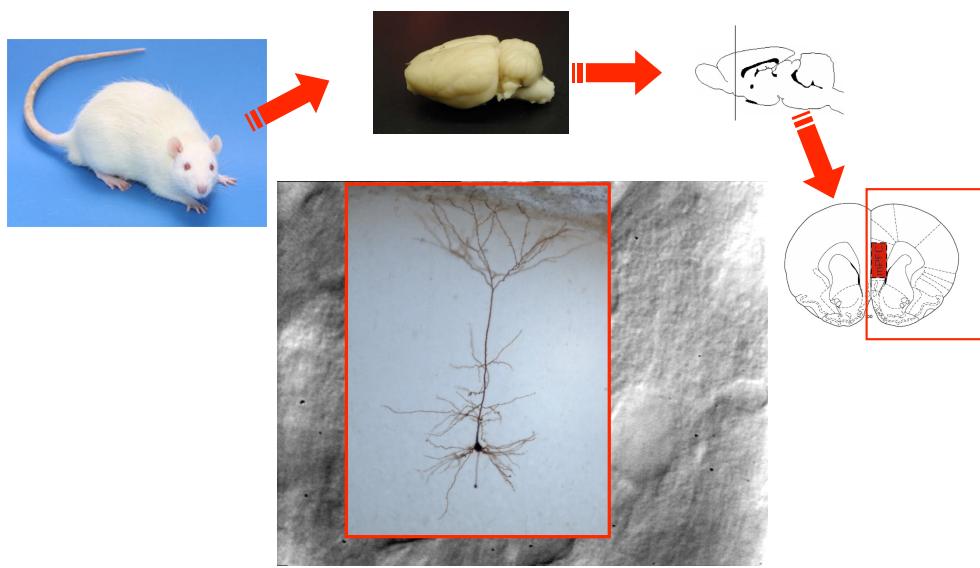
## Leaky-Integrate-and-Fire model: f-I curve (with spike-frequency adaptation)





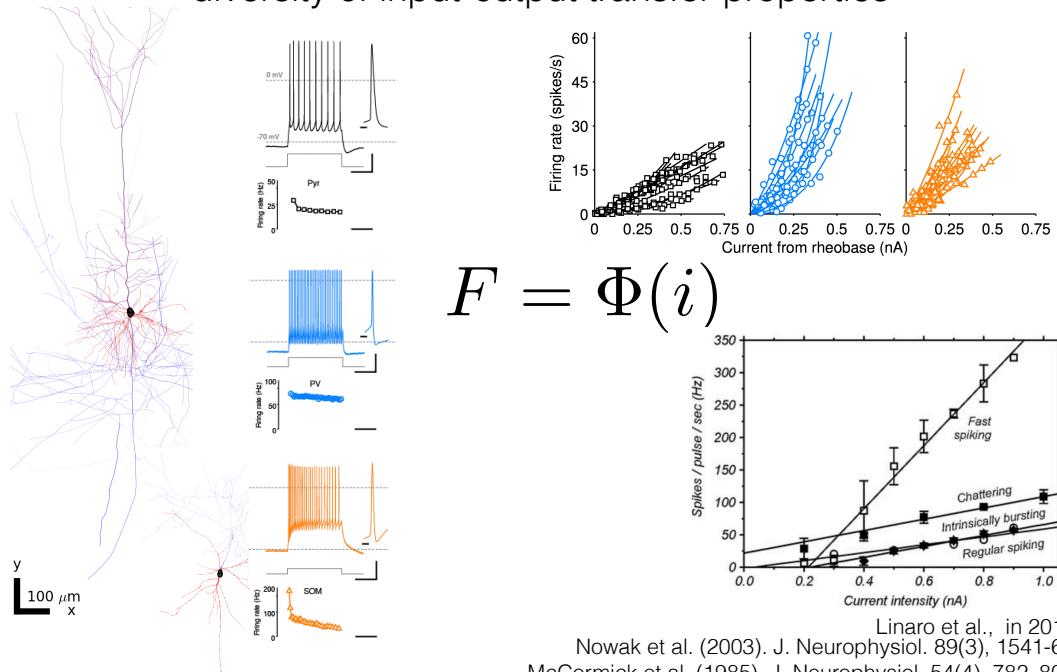
# Are $I\&F$ models good?

Frequency vs (DC) current curve  
for a real pyramidal neuron



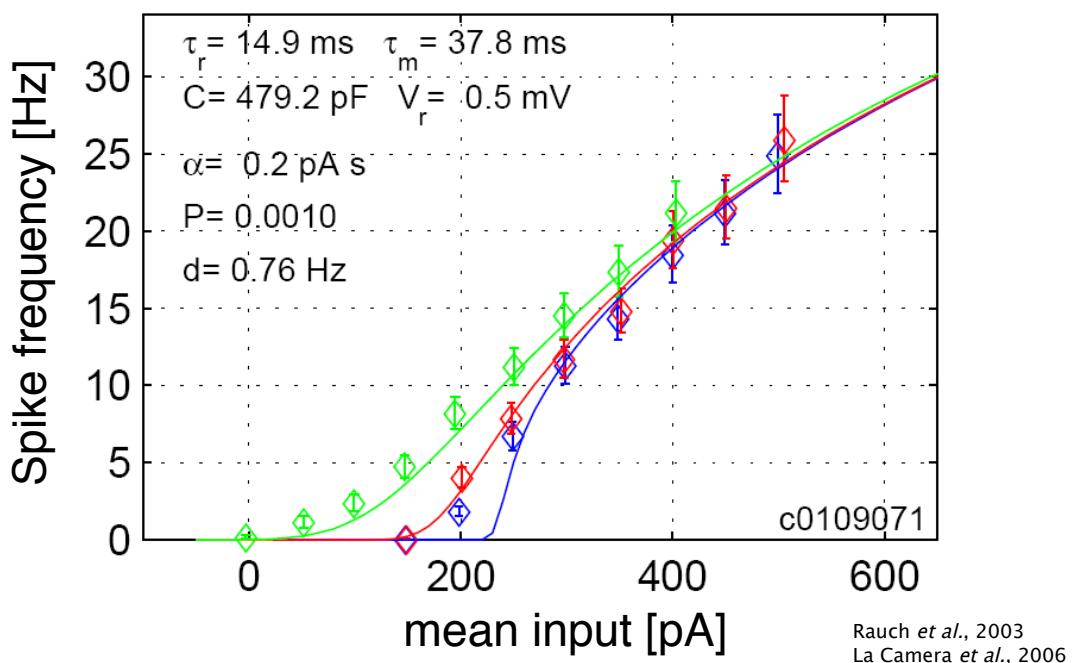
# Cellular electrophysiology of neocortex

## diversity of input-output transfer properties



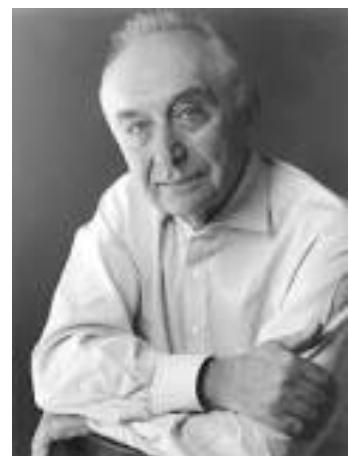
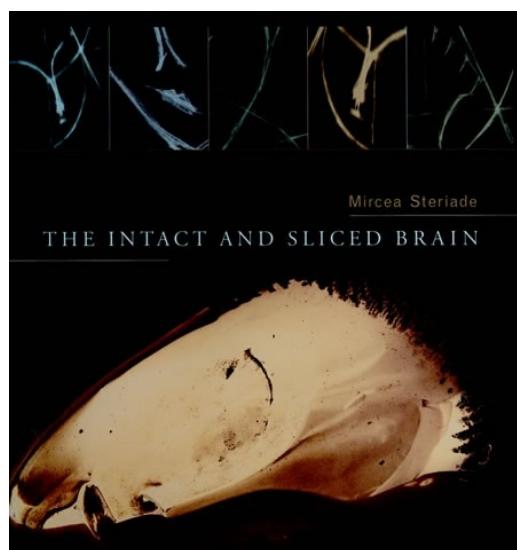
Linaro et al., in 2019  
 Nowak et al. (2003). J. Neurophysiol. 89(3), 1541-66  
 McCormick et al. (1985). J. Neurophysiol. 54(4), 782-806

**Frequency vs current curve**  
**for a pyramidal neuron: I&F are accurate enough!!**

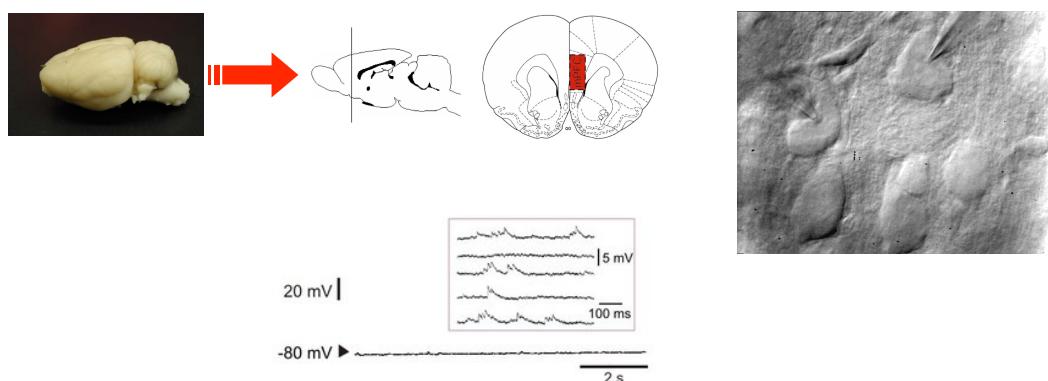


# But what are those “colors” for?

(Cortical) *in vitro* vs *in vivo* physiology

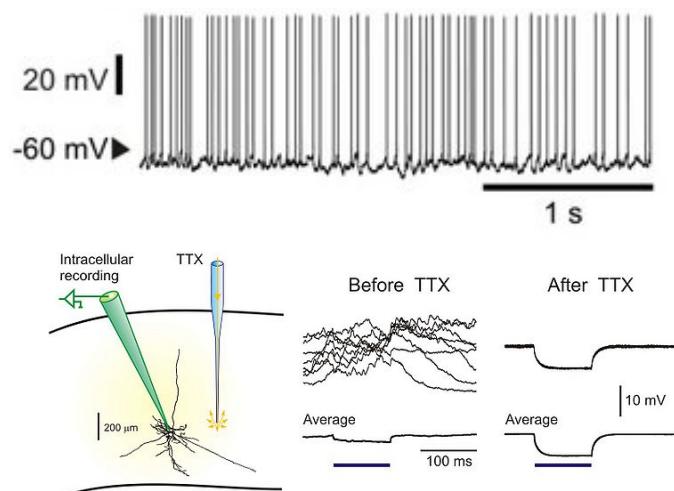


## “Sliced” brain: acute brain tissue slices



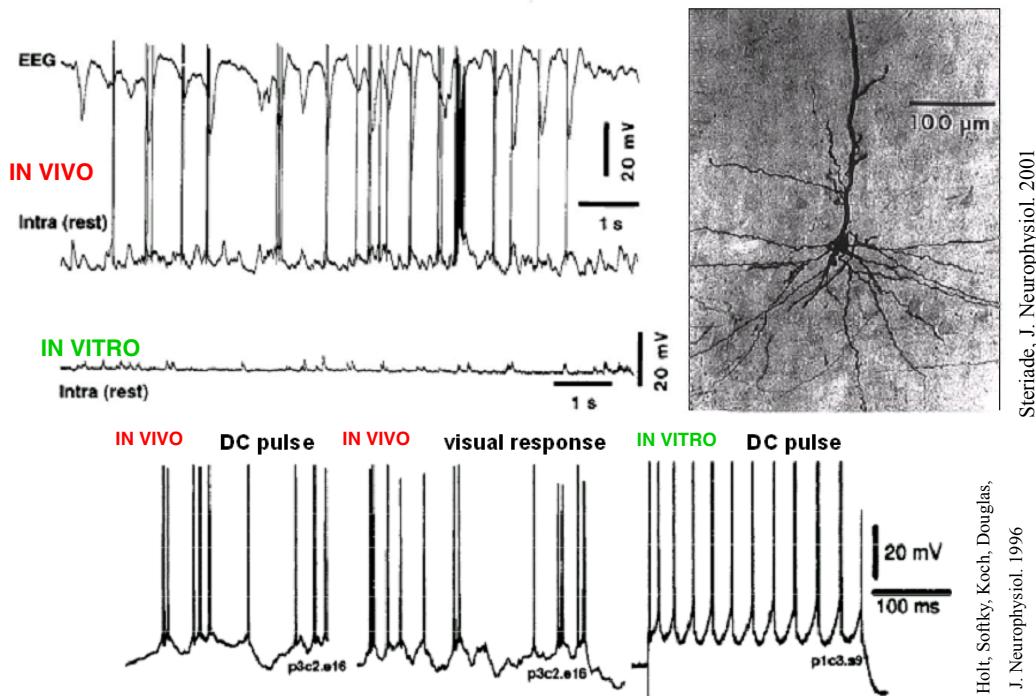
- Lack of any spontaneous firing
- The membrane potential “sits” at resting membrane potential
- Episodic, small synaptic potentials

## Awake, *in vivo* cortical recordings



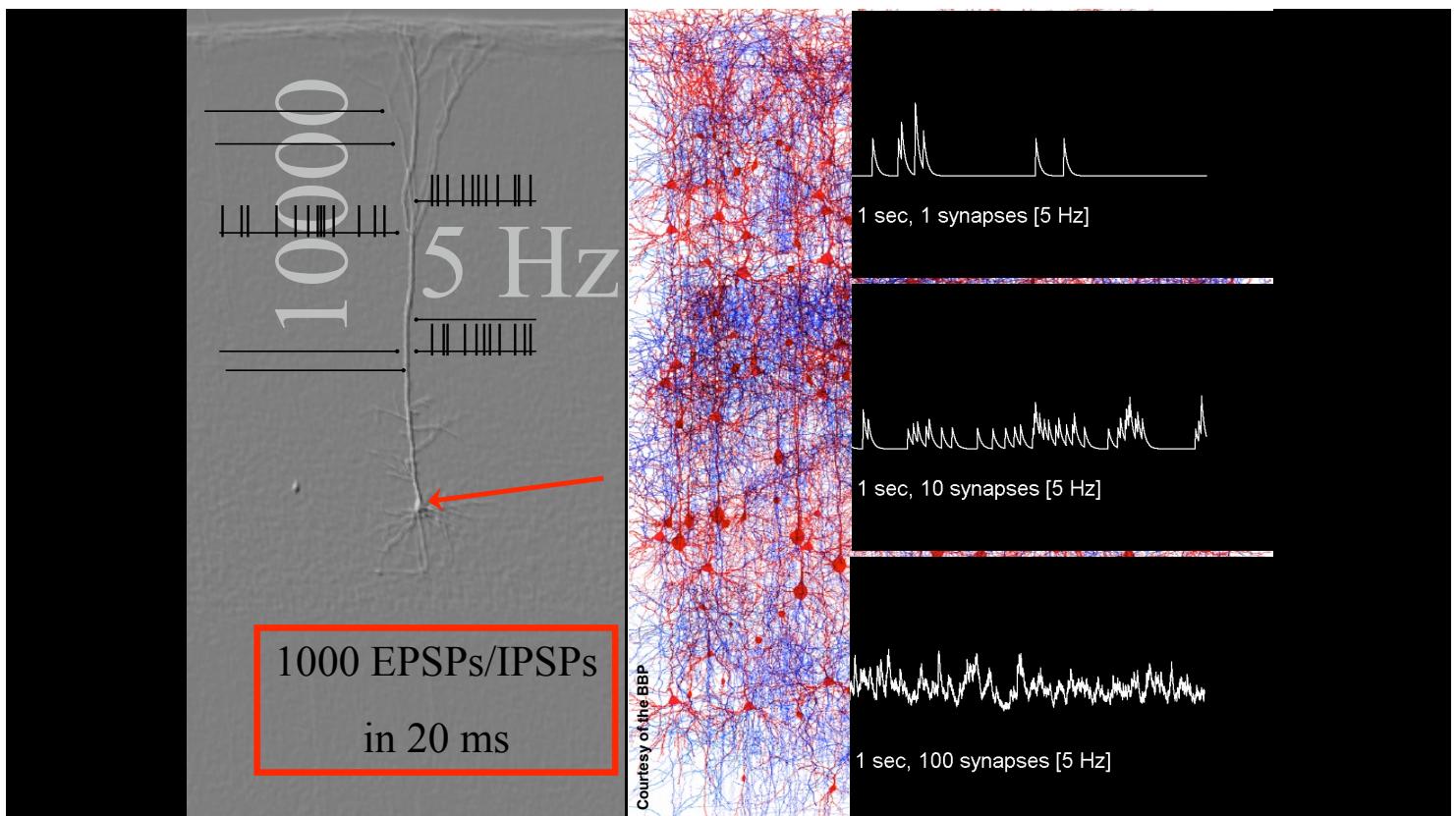
- Spontaneous irregular firing
- Random fluctuations of the membrane potential, subthreshold
- Reduction of the apparent “membrane input resistance”

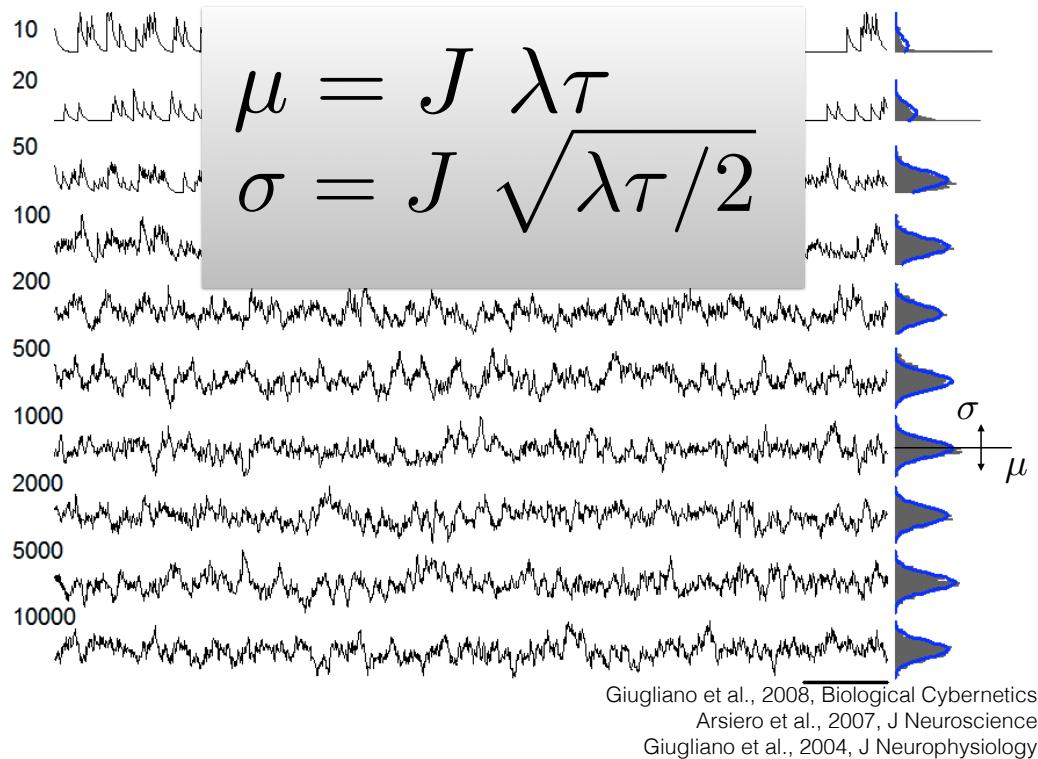
## (Cortical) *in vitro* versus *in vivo* physiology



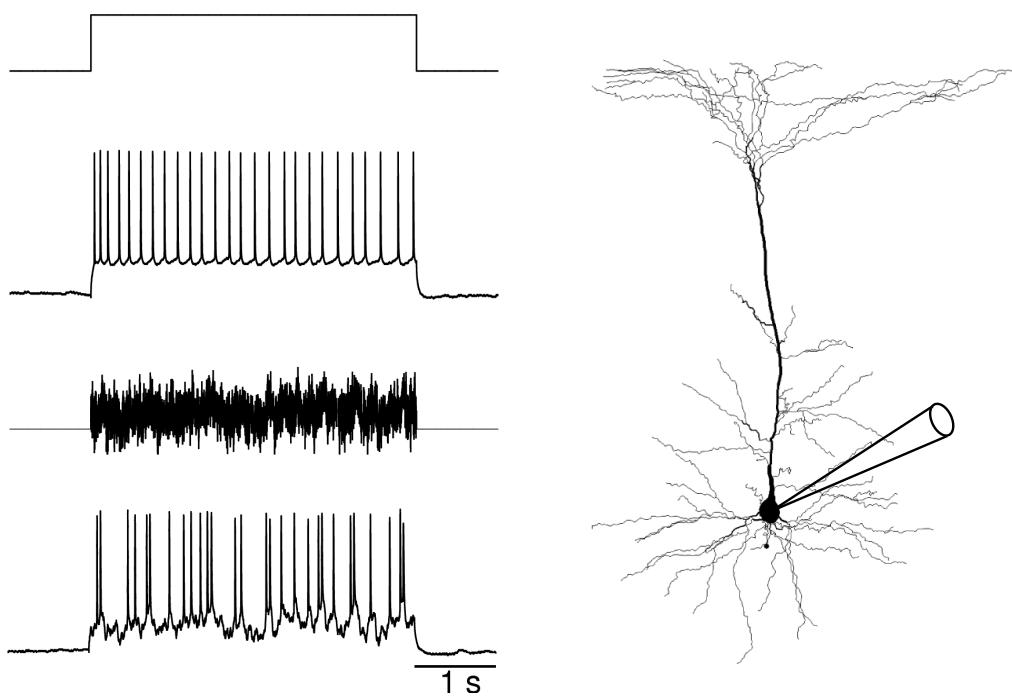
Steriade, J. Neurophysiol. 2001

Holt, Sofsky, Koch, Douglas,  
J. Neurophysiol. 1996

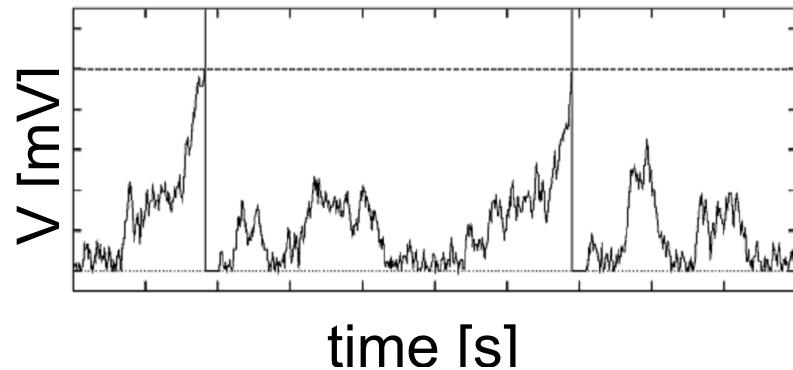




Restore *in vivo*-like activity, *in vitro*??



# Leaky Integrate-and-Fire neuron model



$$r_0 = f(\mu, \sigma, \tau)$$

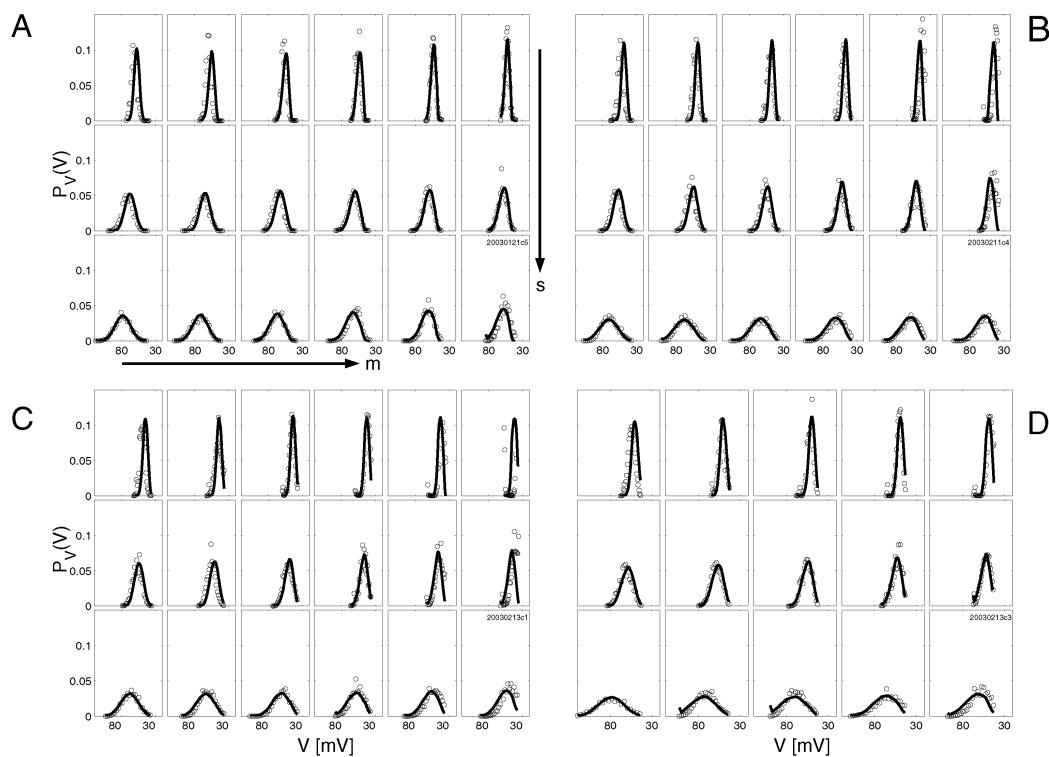
Giugliano et al., 2008, Biological Cybernetics

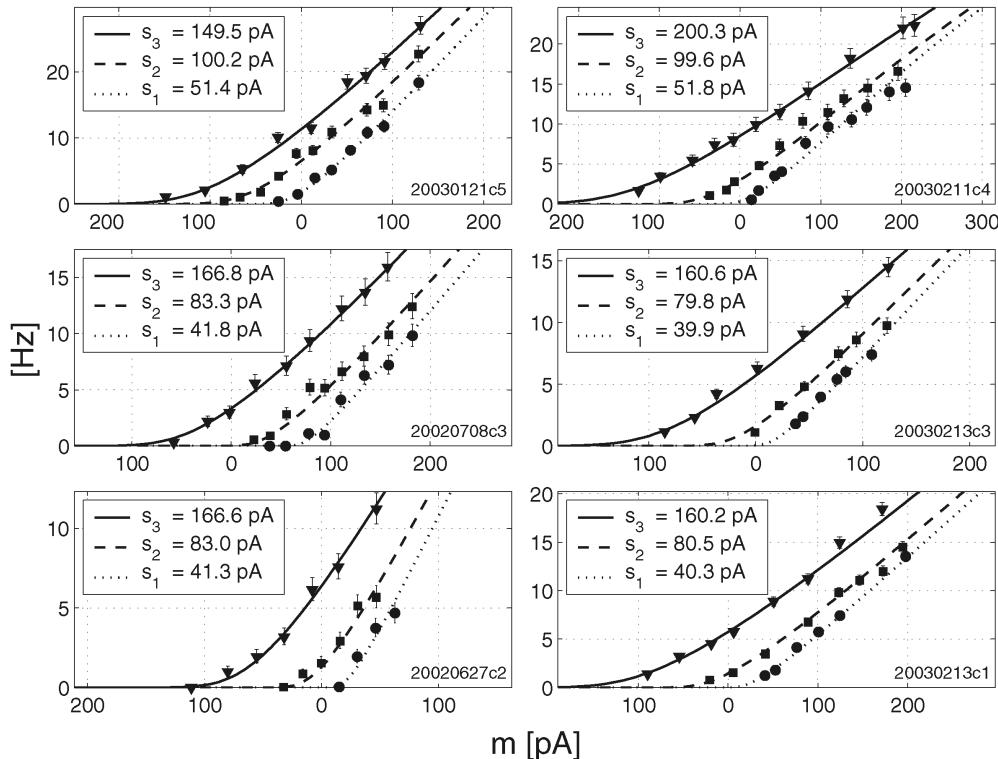
Arsiero et al., 2007, J Neuroscience

Giugliano et al., 2004, J Neurophysiology

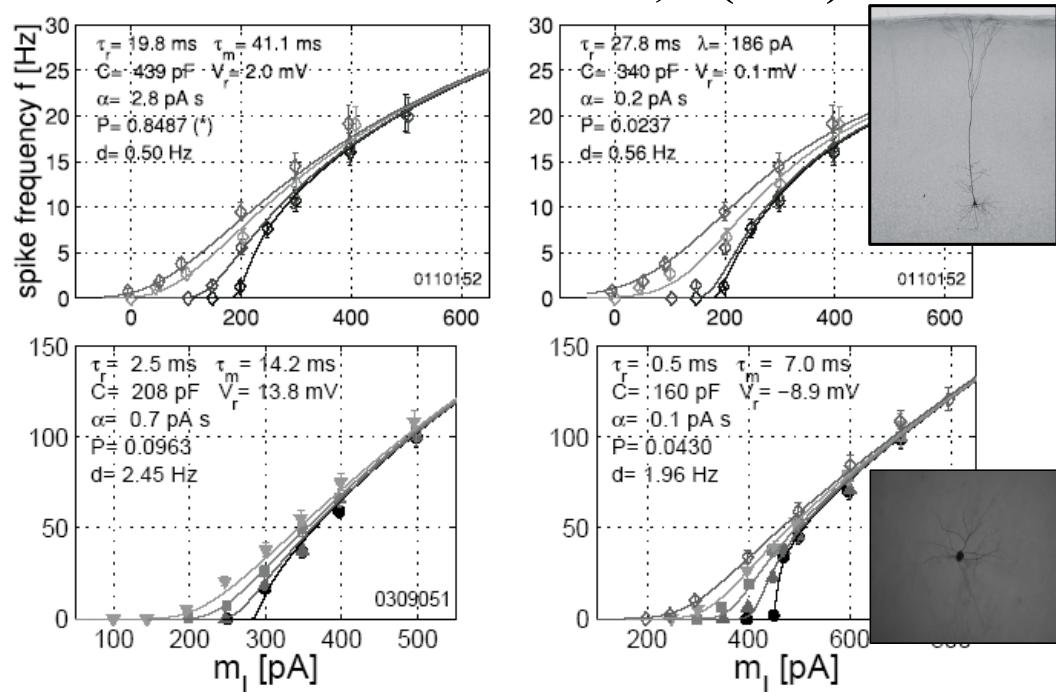
Destexhe & co. ; Gerstner & co. ; and many others

McCormick et al., 1985; Powers and Binder, 2001



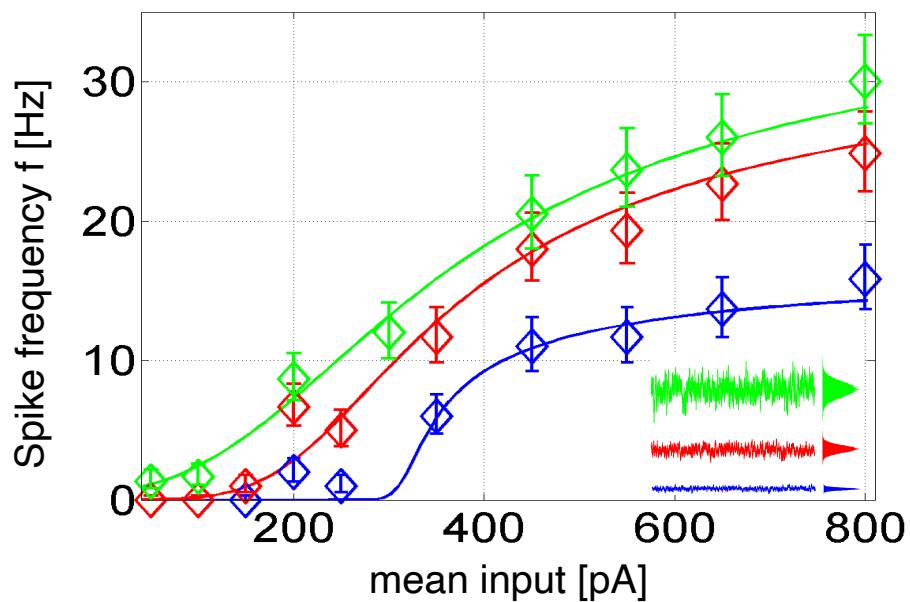
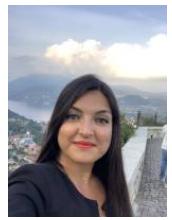


## PYR/FS neurons L2-3, 5 (SSC)

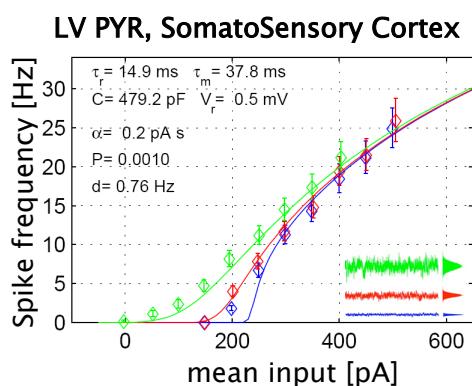


La Camera et al., 2006; Rauch et al., 2003

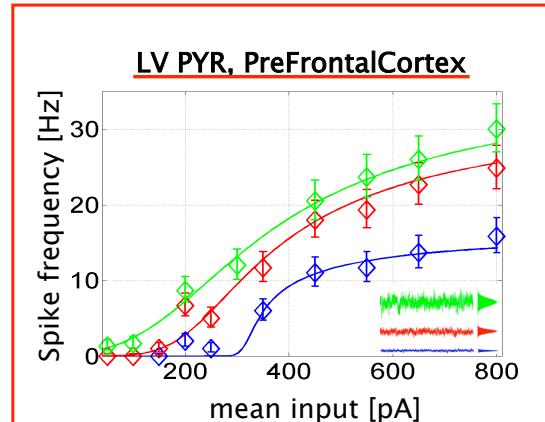
# Medial prefrontal cortical L5 pyramids



## Frequency vs (DC) current curve



Sustained neuronal responses  
increase monotonically  
and become progressively  
insensitive to the amplitude of  
input fluctuations  
(convergent)



Sustained neuronal responses  
saturate at relatively low  
frequency (sigmoidal f-/I curve)  
and remain sensitive to the  
amplitude of input fluctuations  
(divergent)



# MODELLING NEURAL SYSTEMS

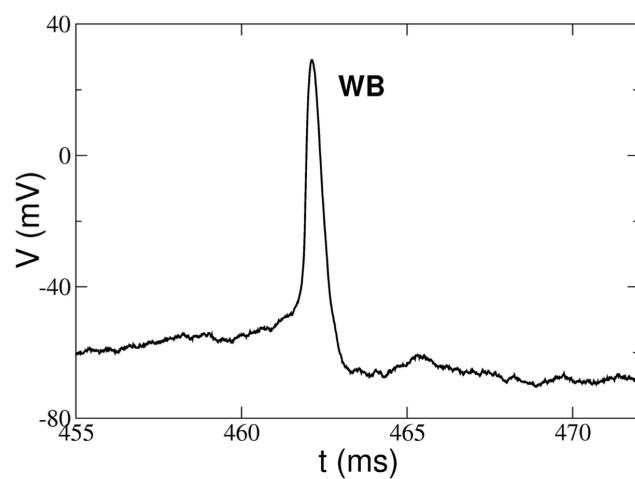


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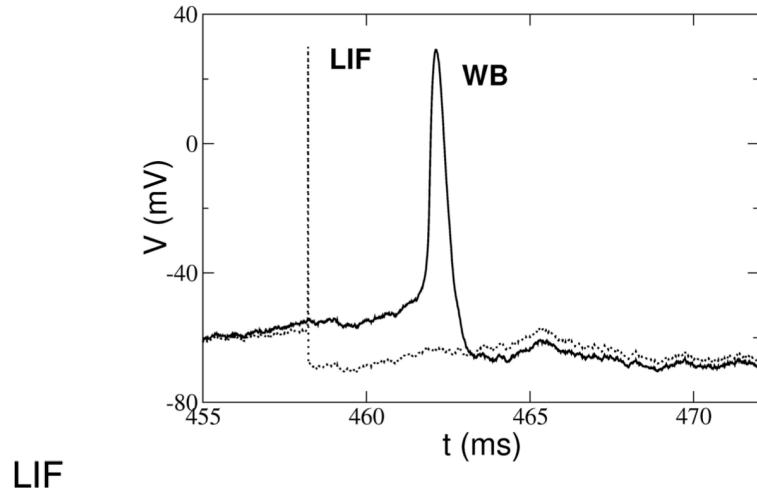
Prof. Ing. Michele GIUGLIANO, PhD

## Simplified models of excitability

What about predicting AP times?



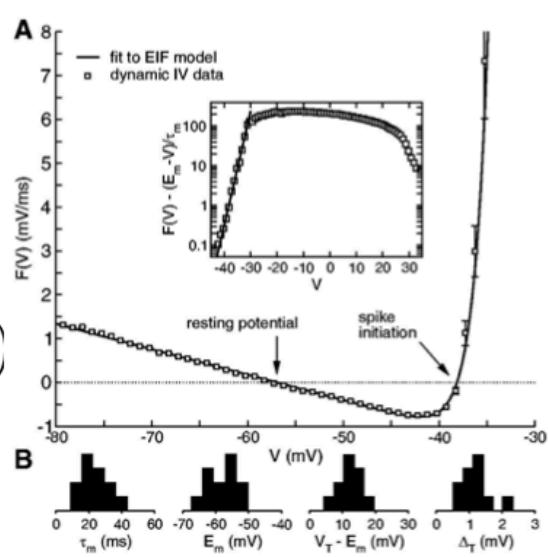
# The LIF fails...



$$C \frac{dV}{dt} = -g_L(V - V_L) + I_{syn}(t)$$

## Extending the LIF model

$$\begin{aligned}\frac{dV}{dt} &= F(V) + \frac{I_{in}(t)}{C} \\ F(V) &= \frac{1}{\tau_m} \left( E_m - V + \Delta_T \exp \left( \frac{V - V_T}{\Delta_T} \right) \right)\end{aligned}$$



Badel et al 2008

# Extending the LIF model

$$C \frac{dV}{dt} = -g_L(V - V_L) + \psi(V) + I_{syn}(t)$$

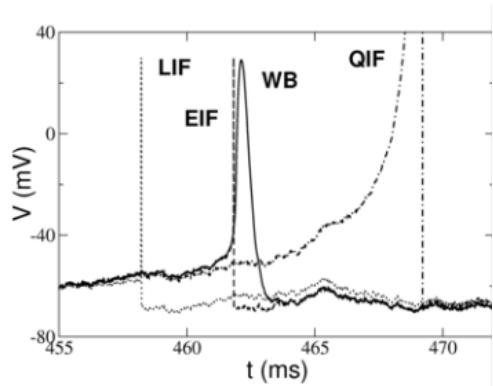
QIF: quadratic integrate-and-fire neuron

$$\begin{aligned} C \frac{dV}{dt} &= -g_L(V - V_L) + \psi(V) + I_{syn}(t) \\ \psi(V) &= \frac{g_L}{2\Delta_T} (V - V_T)^2 + g_L(V - V_L) - I_T \end{aligned}$$

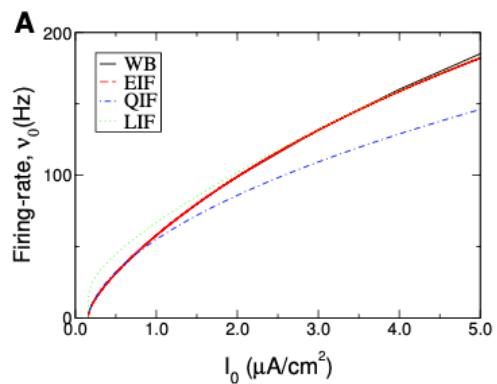
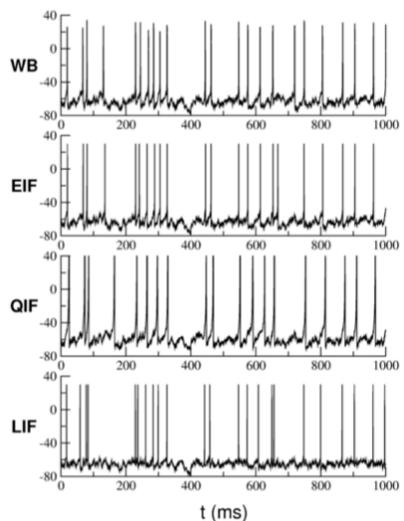
Ermentrout and Kopell 1986

$$\begin{aligned} C \frac{dV}{dt} &= -g_L(V - V_L) + \psi(V) + I_{syn}(t) \\ \psi(V) &= g_L \Delta_T \exp\left(\frac{V - V_T}{\Delta_T}\right) \end{aligned}$$

(Fourcaud-Trocme et al 2003)



# Extending the LIF model



- INCF competition:

