

Introduction to Computational Neuroscience

Biol 698

Math 635

Biol 498

Math 430

Bibliography:

"Mathematical Foundations of Neuroscience", by G. B. Ermentrout & D. H. Terman - Springer (2010), 1st edition. ISBN 978-0-387-87707-5

* "Dynamical Systems in Neuroscience: The Geometry of Excitability and Bursting", by Eugene M. Izhikevich. The MIT Press, 2007. ISBN 0-262-09043-8

Overview

One-dimensional models

- Reduction of the Hodgkin-Huxley model to a one-dimensional model (review)
- One-dimensional models
- Equilibria and stability
- Bistability and hysteresis
- Topological equivalence
- Bifurcations

One-dimensional neural model

$$C\dot{V} = I - \overbrace{\bar{g}_K n^4 (V - E_K)}^{I_K} - \overbrace{\bar{g}_{Na} m^3 h (V - E_{Na})}^{I_{Na}} - \overbrace{g_L (V - E_L)}^{I_L}$$
$$\dot{n} = \overbrace{(n_\infty(V) - n) / \tau_n(V)}^{I_K}$$
$$\dot{m} = \overbrace{(m_\infty(V) - m) / \tau_m(V)}^{I_{Na}}$$
$$\dot{h} = \overbrace{(h_\infty(V) - h) / \tau_h(V)}^{I_L}$$

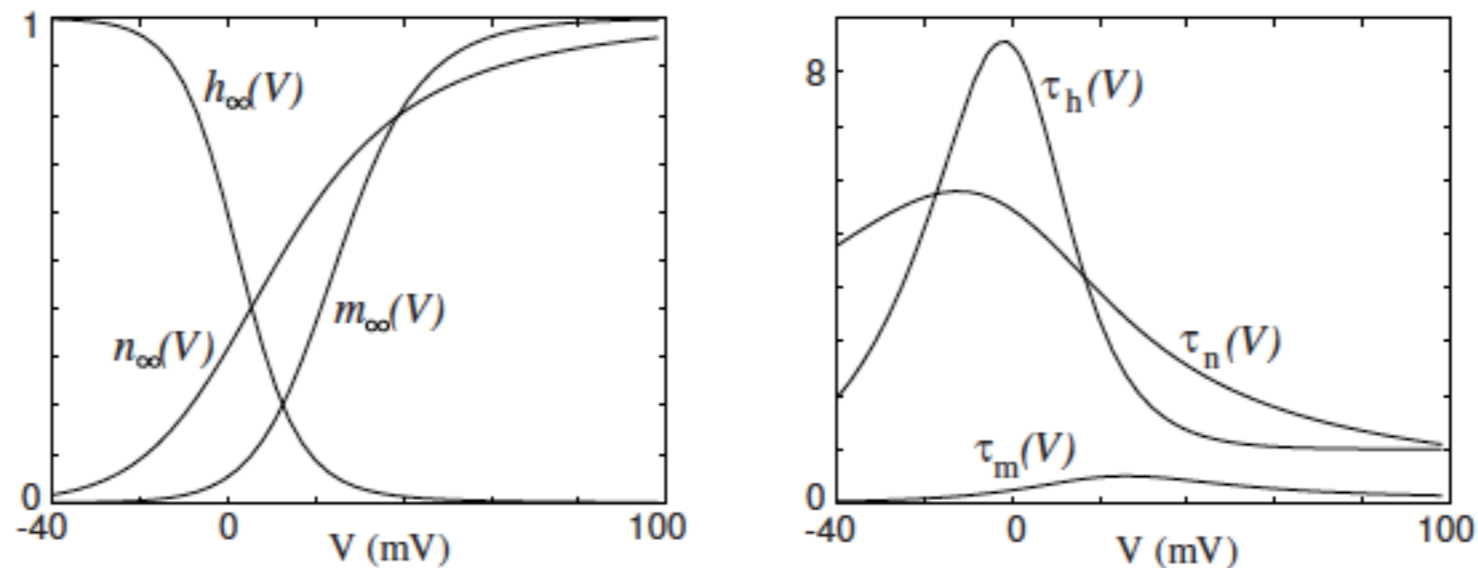


Figure 2.13: Steady-state (in)activation functions (left) and voltage-dependent time constants (right) in the Hodgkin-Huxley model.

One-dimensional neural model

$$C\dot{V} = I - \overbrace{\bar{g}_K n^4 (V - E_K)}^{I_K} - \overbrace{\bar{g}_{Na} m^3 h (V - E_{Na})}^{I_{Na}} - \overbrace{g_L (V - E_L)}^{I_L}$$

$$\dot{n} = (n_\infty(V) - n) / \tau_n(V),$$

$$\dot{m} = (m_\infty(V) - m) / \tau_m(V),$$

$$\dot{h} = (h_\infty(V) - h) / \tau_h(V),$$

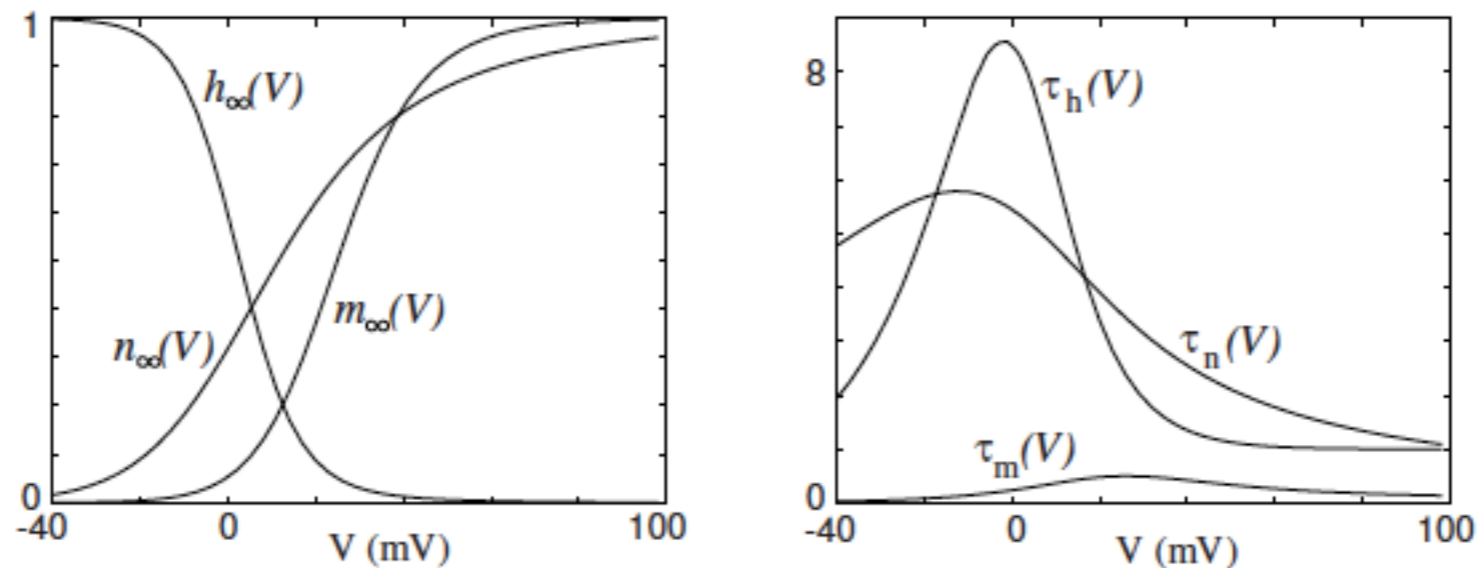


Figure 2.13: Steady-state (in)activation functions (left) and voltage-dependent time constants (right) in the Hodgkin-Huxley model.

One-dimensional neural model

- The passive membrane equation

$$C \frac{dV}{dt} = - G_L (V - E_L) + I_{app}$$

One-dimensional neural model

$$\begin{aligned}
 C\dot{V} &= I - \overbrace{\bar{g}_K n^4 (V - E_K)}^{I_K} - \overbrace{\bar{g}_{Na} m^3 h (V - E_{Na})}^{I_{Na}} - \overbrace{g_L (V - E_L)}^{I_L} \\
 \dot{n} &= (n_\infty(V) - n) / \tau_n(V), \\
 \dot{m} &= (m_\infty(V) - m) / \tau_m(V), \\
 \dot{h} &= (h_\infty(V) - h) / \tau_h(V),
 \end{aligned}$$

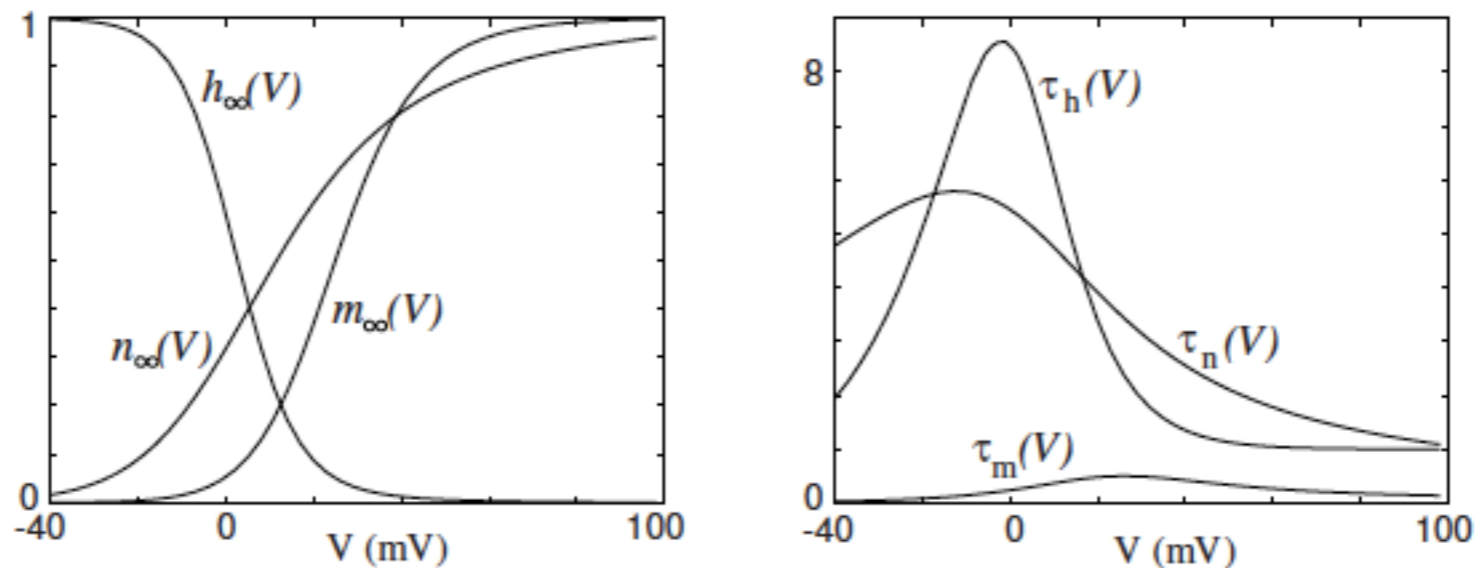


Figure 2.13: Steady-state (in)activation functions (left) and voltage-dependent time constants (right) in the Hodgkin-Huxley model.

One-dimensional neural model

$$C\dot{V} = I - \overbrace{\bar{g}_K n^4 (V - E_K)}^{I_K} - \overbrace{\bar{g}_{Na} m^3 h (V - E_{Na})}^{I_{Na}} - \overbrace{g_L (V - E_L)}^{I_L}$$

$$\dot{n} = \overbrace{(n_\infty(V) - n) / \tau_n(V)}^{\text{activation}} ,$$

$$\dot{m} = (m_\infty(V) - m) / \tau_m(V) ,$$

$$\dot{h} = \overbrace{(h_\infty(V) - h) / \tau_h(V)}^{\text{deactivation}} ,$$

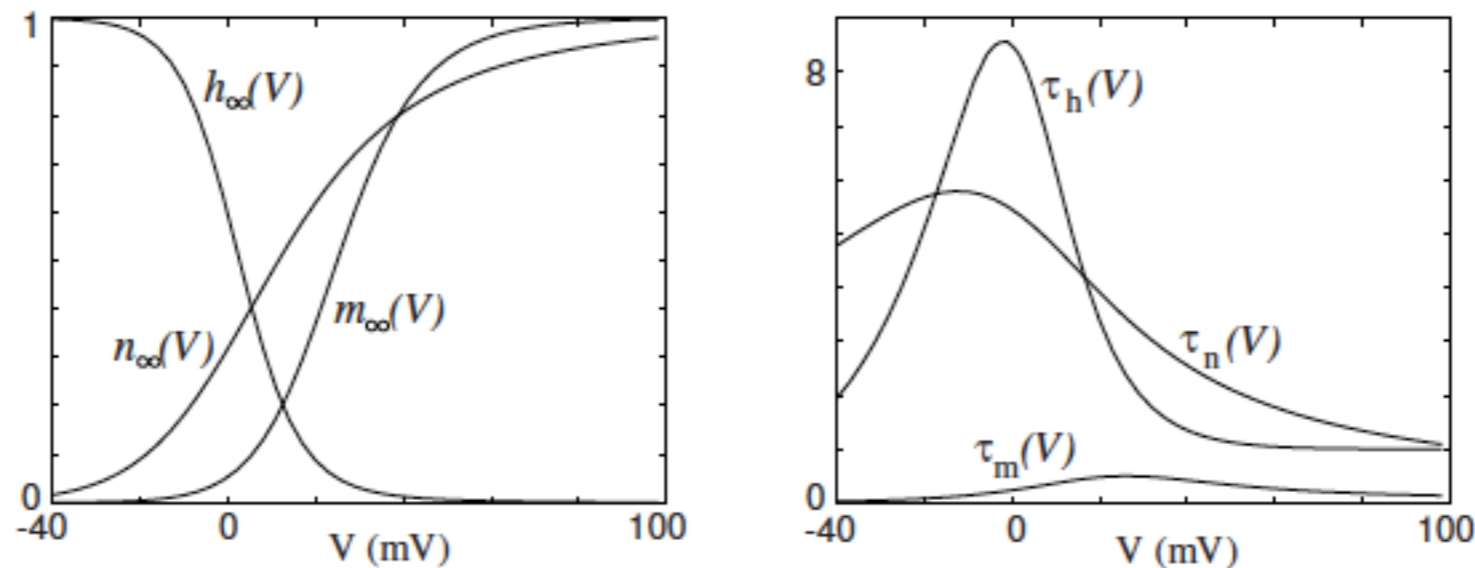


Figure 2.13: Steady-state (in)activation functions (left) and voltage-dependent time constants (right) in the Hodgkin-Huxley model.

One-dimensional neural model

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$$\dot{n} = (n_\infty(V) - n) / \tau_n(V),$$

$$\dot{m} = (m_\infty(V) - m) / \tau_m(V),$$

$$\dot{h} = (h_\infty(V) - h) / \tau_h(V),$$

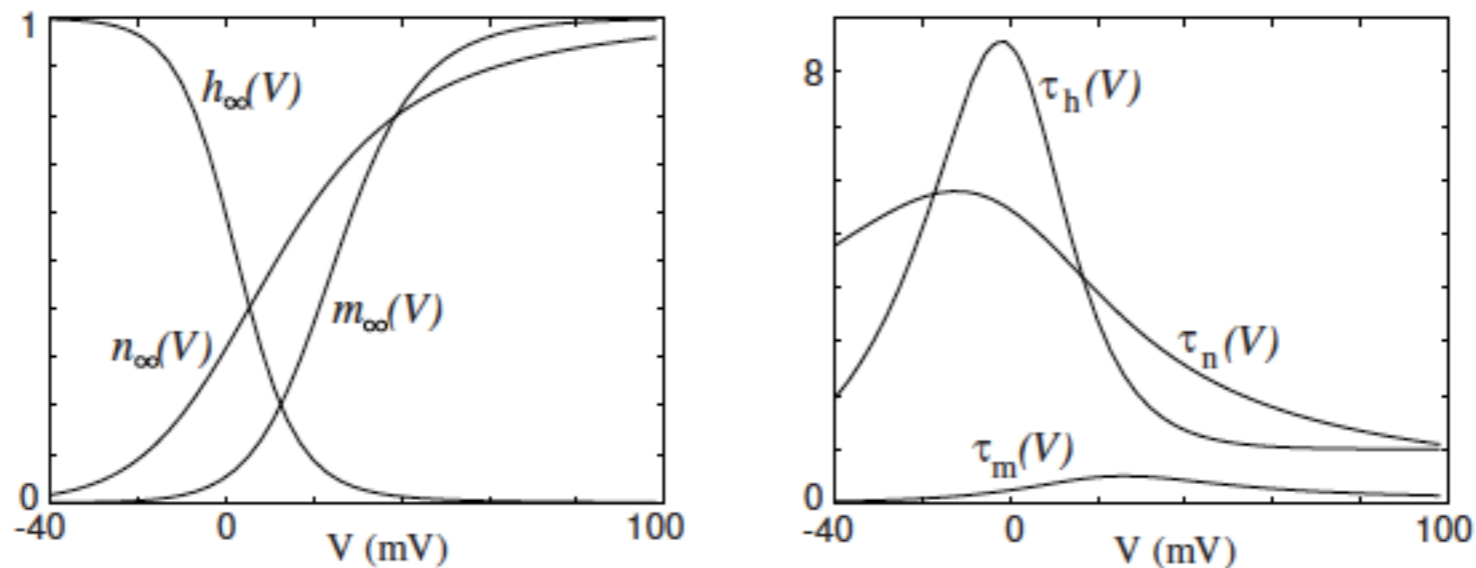
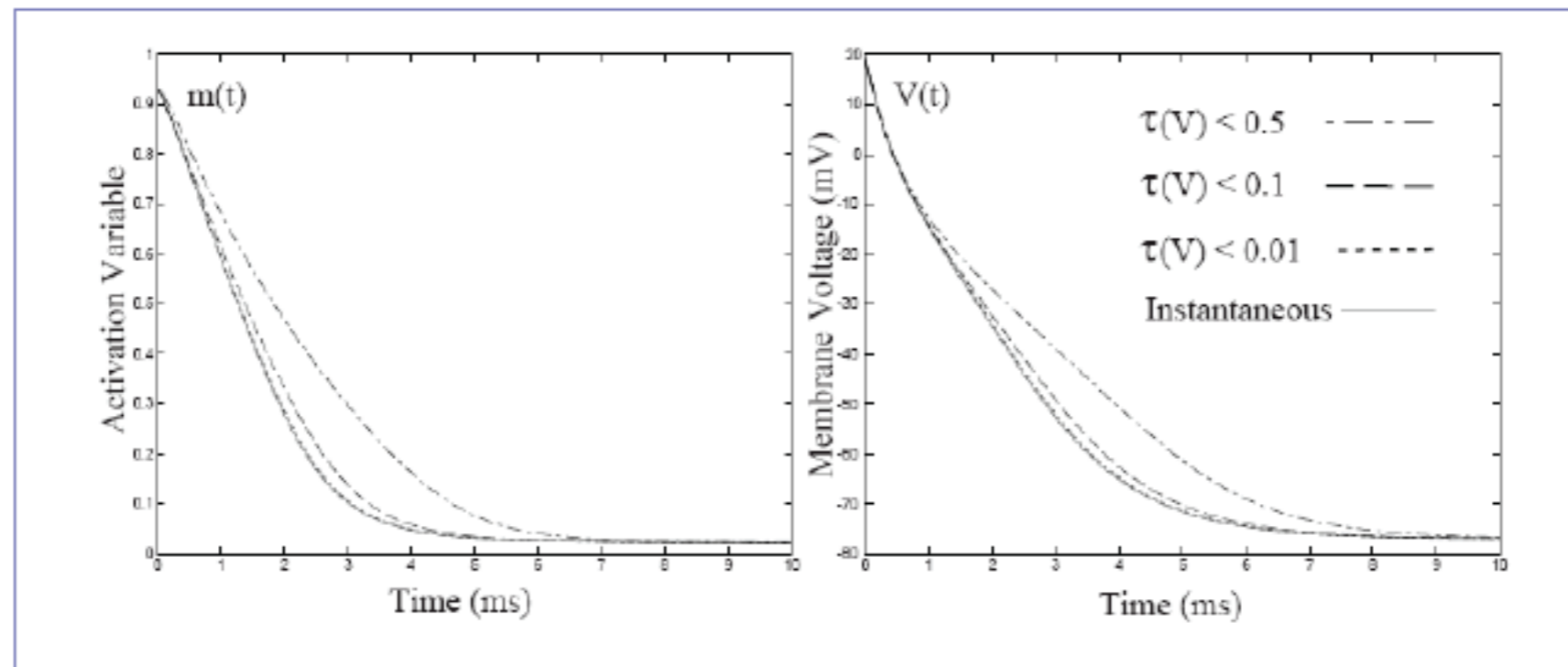


Figure 2.13: Steady-state (in)activation functions (left) and voltage-dependent time constants (right) in the Hodgkin-Huxley model.

One-dimensional neural model

$$C \frac{dV}{dt} = - G_{Na} m_{\infty}^3(V) (V - E_{Na}) - G_L (V - E_L) + I_{app}$$

$$\frac{dm}{dt} = \frac{m_{\infty}(V) - m}{\tau(V)}$$



One-dimensional neural model

- Persistent sodium model

$$C \frac{dV}{dt} = - G_{Na} m_{\infty}^3 (V) (V - E_{Na}) - G_L (V - E_L) + I_{app}$$

One-dimensional neural model

- Persistent sodium model

$$C \frac{dV}{dt} = - G_{Na} p_{\infty}^3(V) (V - E_{Na}) - G_L (V - E_L) + I_{app}$$

One-dimensional neural model

- Models with one gating variable

$$C \frac{dV}{dt} = - G_X x_\infty(V) (V - E_X) - G_L (V - E_L) + I_{app}$$

One-dimensional neural model

- Models with one gating variable

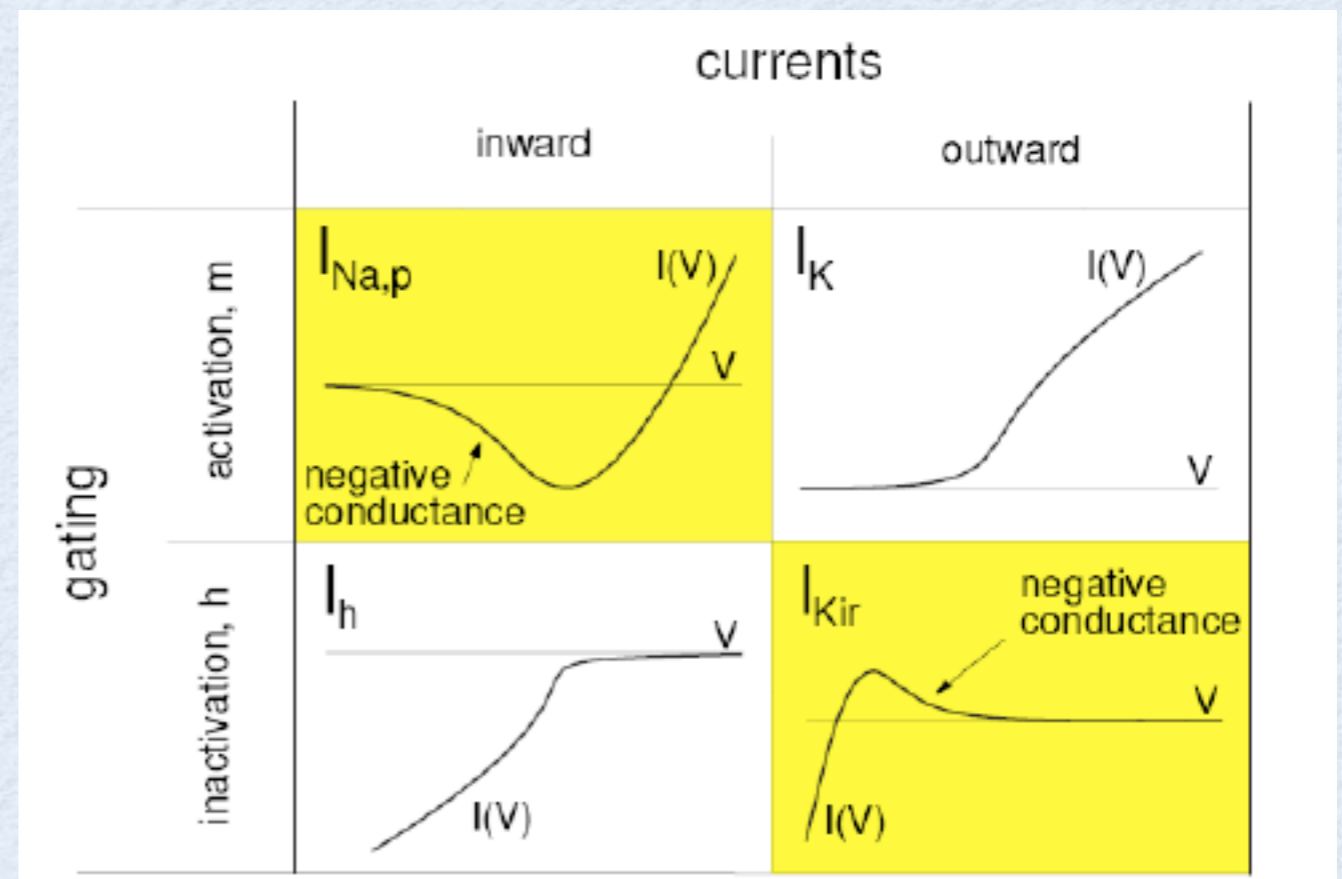
$$C \frac{dV}{dt} = - G_X x_\infty(V) (V - E_X) - G_L (V - E_L) + I_{app}$$

		Current	
		inward	outward
Gating	activation	$I_{Na,p}$	I_K
	inactivation	I_h	I_{Kir}

One-dimensional neural model

$$C \frac{dV}{dt} = -G_X x_\infty(V) (V - E_X) - G_L (V - E_L) + I_{app}$$

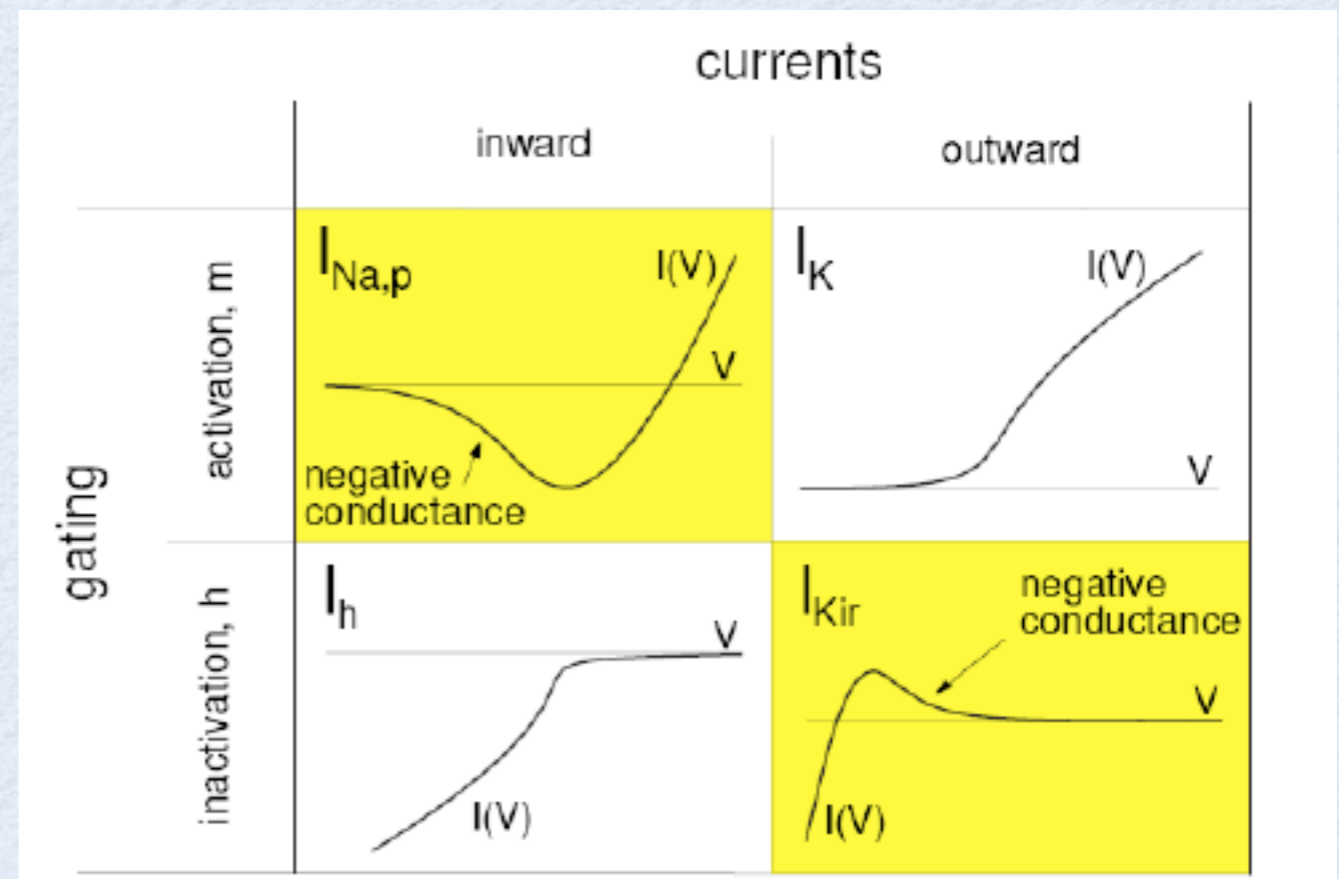
- Negative conductance in the I-V relation creates positive feedback between V and x
- Amplifying role in neural dynamics
- Amplifying currents



One-dimensional neural model

$$C \frac{dV}{dt} = - G_X x_\infty(V) (V - E_X) - G_L (V - E_L) + I_{app}$$

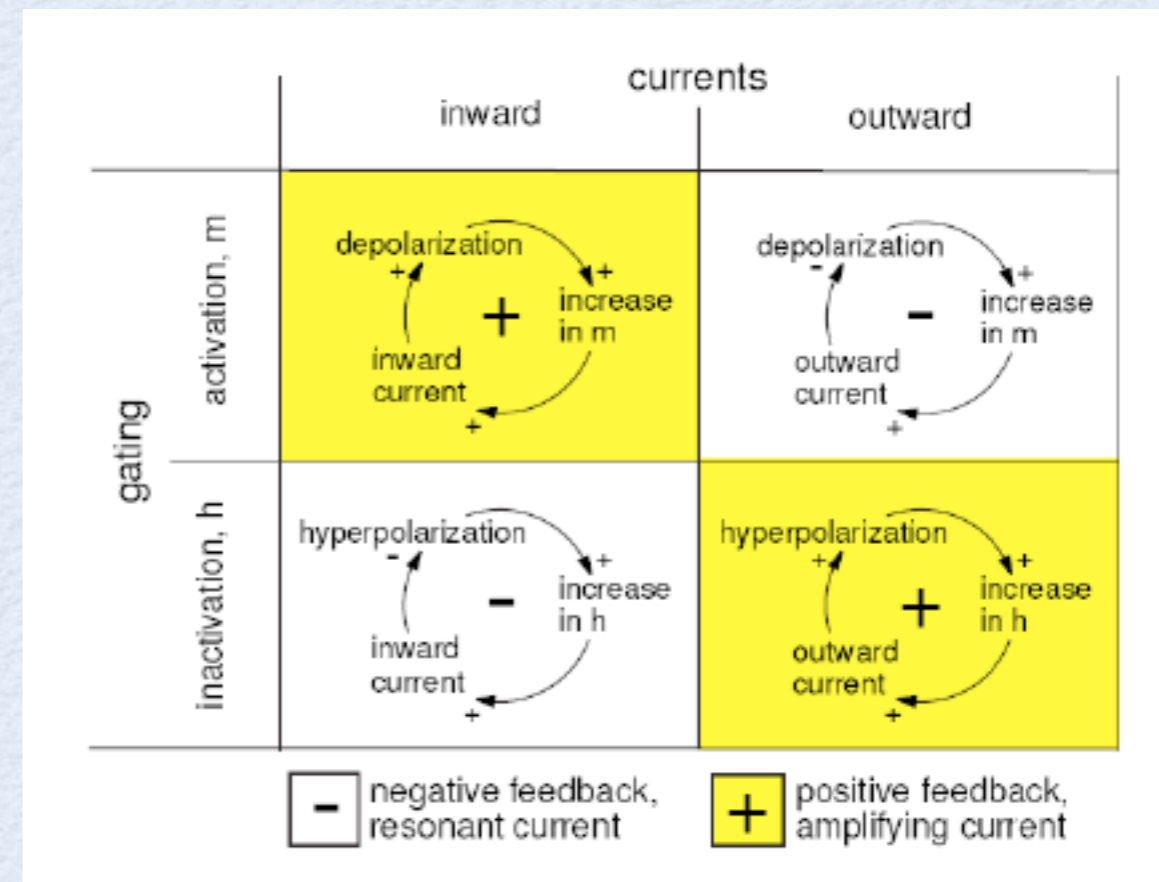
- The currents in the other group have negative feedback between V and x
- Damped oscillations of the membrane potential
- Resonant currents



One-dimensional neural model

$$C \frac{dV}{dt} = -G_X x_\infty(V) (V - E_X) - G_L (V - E_L) + I_{app}$$

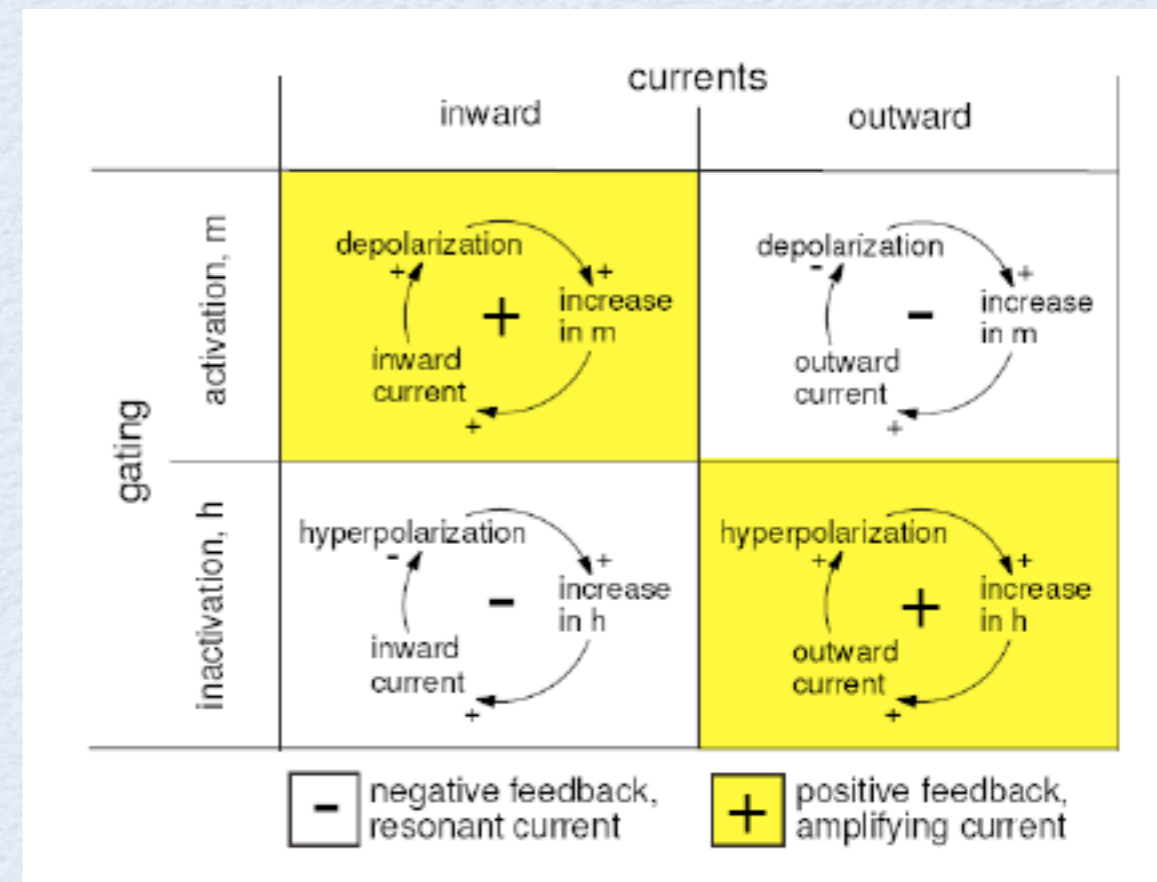
- Negative conductance in the I-V relation creates positive feedback between V and x
- Amplifying role in neural dynamics
- Amplifying currents



One-dimensional neural model

$$C \frac{dV}{dt} = -G_X x_\infty(V) (V - E_X) - G_L (V - E_L) + I_{app}$$

- The currents in the other group have negative feedback between V and x
- Damped oscillations of the membrane potential
- Resonant currents



One-dimensional neural model

- $I_{Na,p}$ model

$$C \frac{dV}{dt} = - G_{Na} p_{\infty}^3(V) (V - E_{Na}) - G_L (V - E_L) + I$$

$$p_{\infty}(V) = \frac{1}{1 + e^{\frac{V_{1/2} - V}{V_{sl}}}}$$

$$C = 10 \mu F$$

$$I = 0 \text{ pA}$$

$$G_L = 19 \text{ mS}$$

$$E_L = -67 \text{ mV}$$

$$G_{Na} = 74 \text{ mS}$$

$$V_{1/2} = 1.5 \text{ mV}$$

$$V_{sl} = 16 \text{ mV}$$

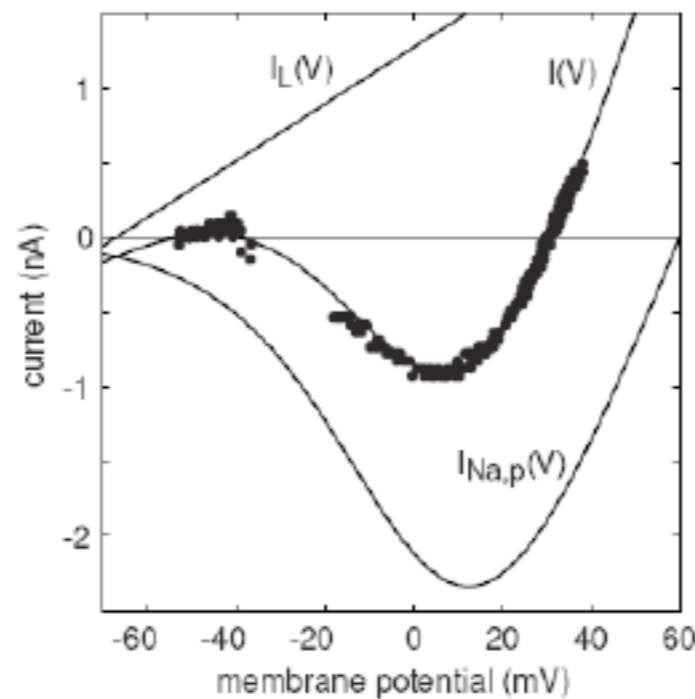
$$E_{Na} = 60 \text{ mV}$$

- Experimental parameter values obtained using whole-patch clamp recordings of a layer V pyramidal neuron in the visual cortex of a rat at room temperature.

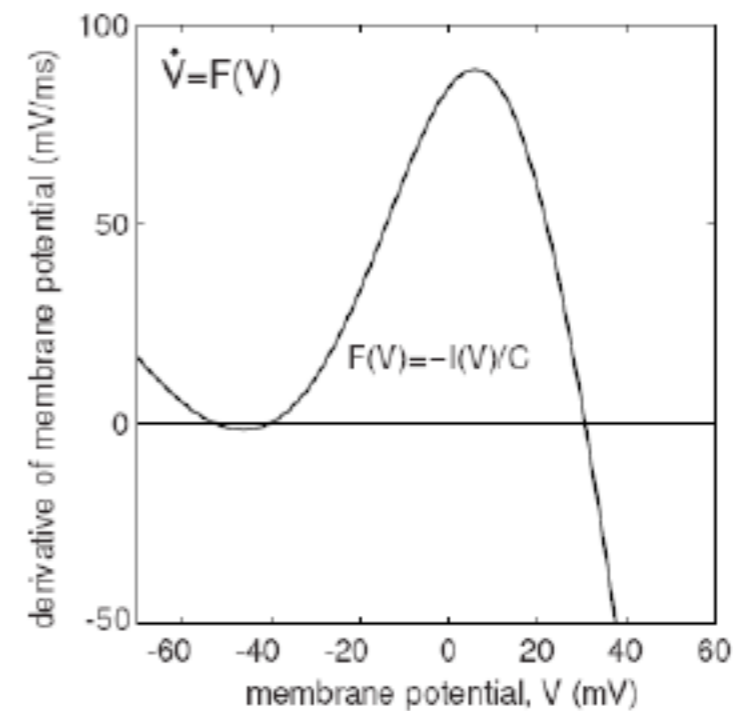
One-dimensional neural model

- $I_{Na,p}$ model

$$C \frac{dV}{dt} = - G_{Na} p_{\infty}(V) (V - E_{Na}) - G_L (V - E_L) + I$$



a

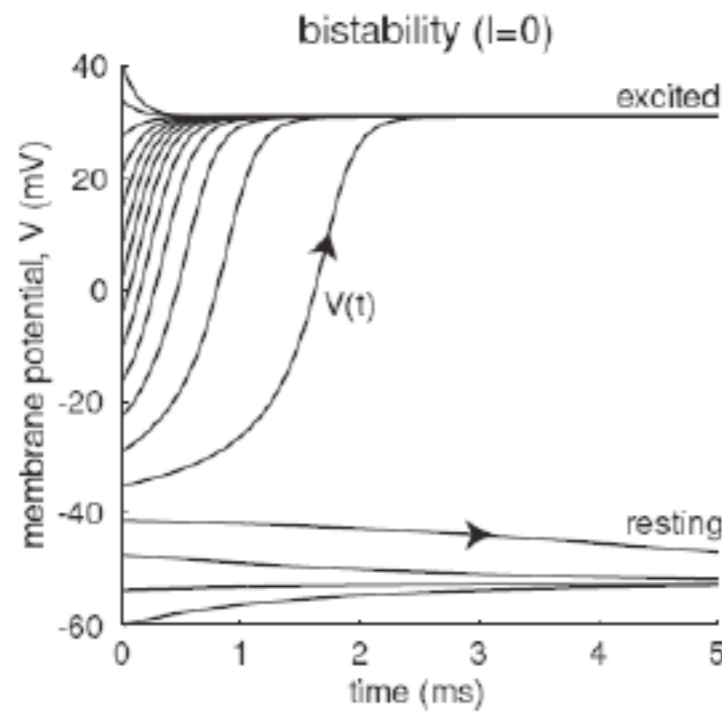


b

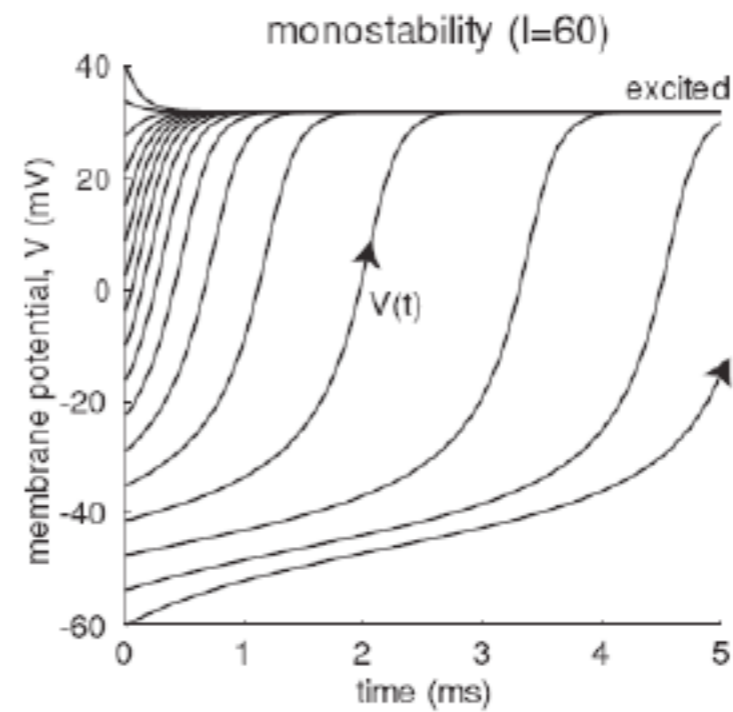
One-dimensional neural model

- $I_{Na,p}$ model

$$C \frac{dV}{dt} = - G_{Na} p_{\infty}(V) (V - E_{Na}) - G_L (V - E_L) + I$$



a



b

Typical voltage trajectories of the $I_{Na,p}$ -model

One-dimensional neural model

- Equilibria

$$C \frac{dV}{dt} = - G_{Na} p_{\infty}(V) (V - E_{Na}) - G_L (V - E_L) + I$$

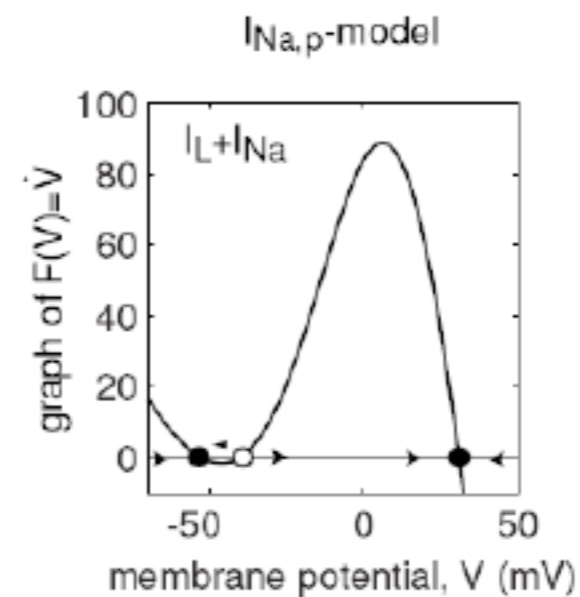
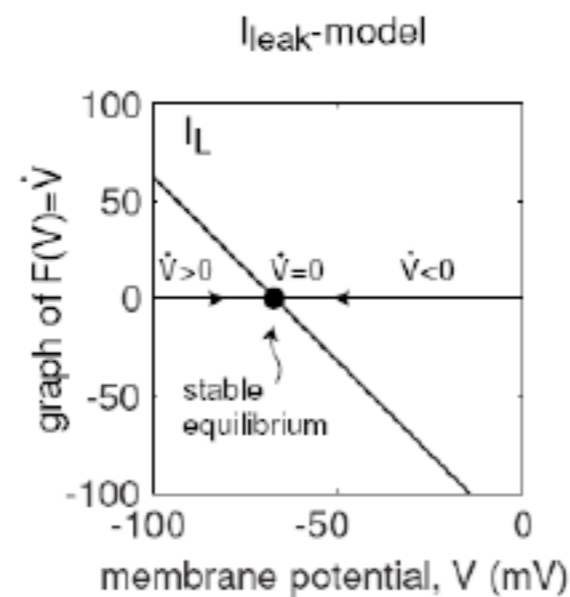
$$\frac{dV}{dt} = 0 \Rightarrow V = V_{eq}$$

$$\frac{dV}{dt} = F(V) \quad \rightarrow \quad F(V_{eq}) = 0$$

One-dimensional neural model

- Equilibria

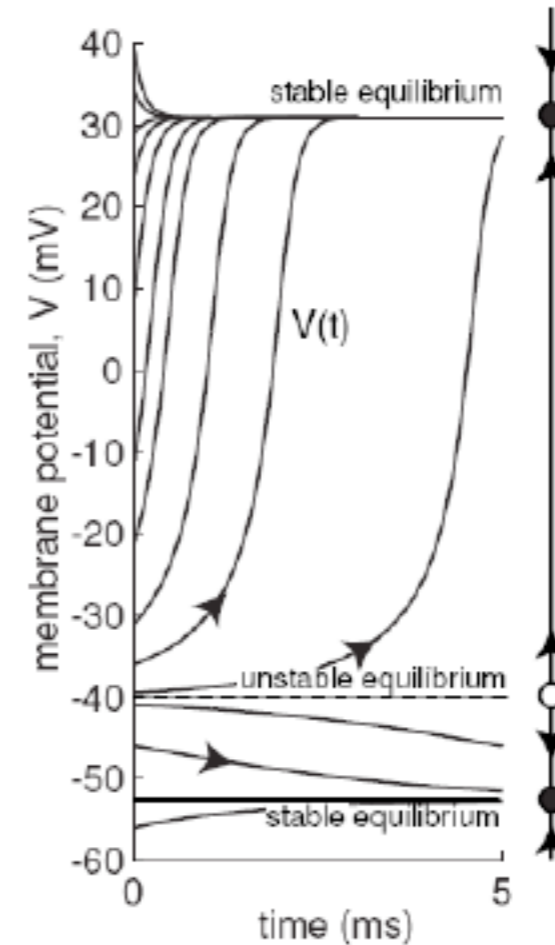
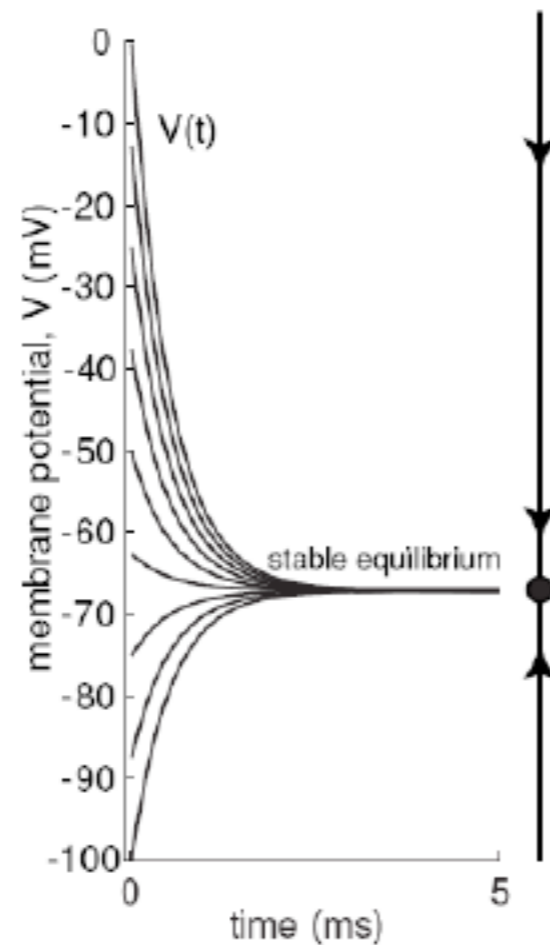
$$C \frac{dV}{dt} = -G_{Na} p_{\infty}(V) (V - E_{Na}) - G_L (V - E_L) + I$$



One-dimensional neural model

- Equilibria

$$C \frac{dV}{dt} = -G_{Na} p_{\infty}(V) (V - E_{Na}) - G_L (V - E_L) + I$$



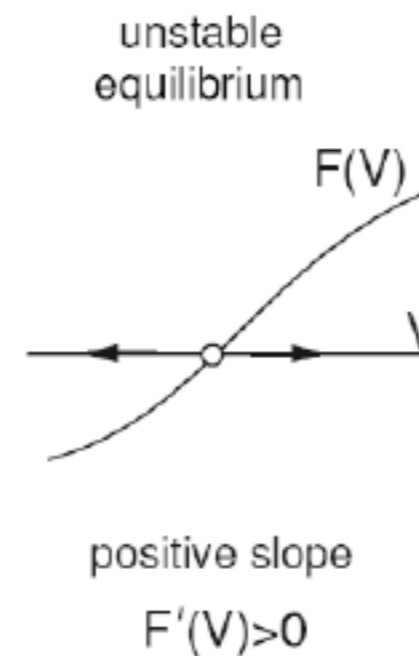
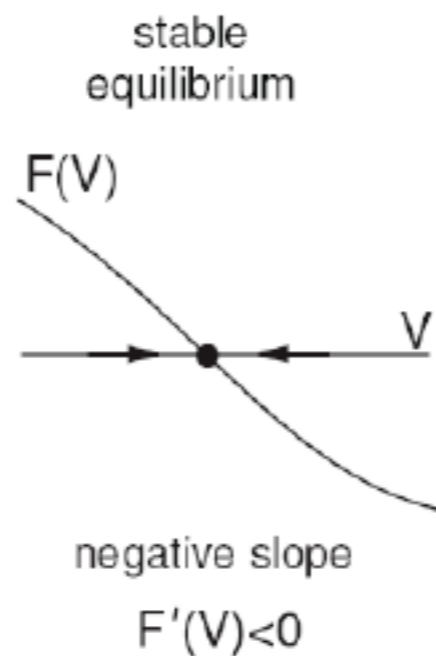
One-dimensional neural model

- Stability

$$\frac{dV}{dt} = F(V) \quad V(0) = V_0$$

$$\lambda = F'(V_{eq}) \text{ (eigenvalue)}$$

$$F(V_{eq}) = 0$$



Stability of an equilibrium is determined by the signs of $F(V)$ around it

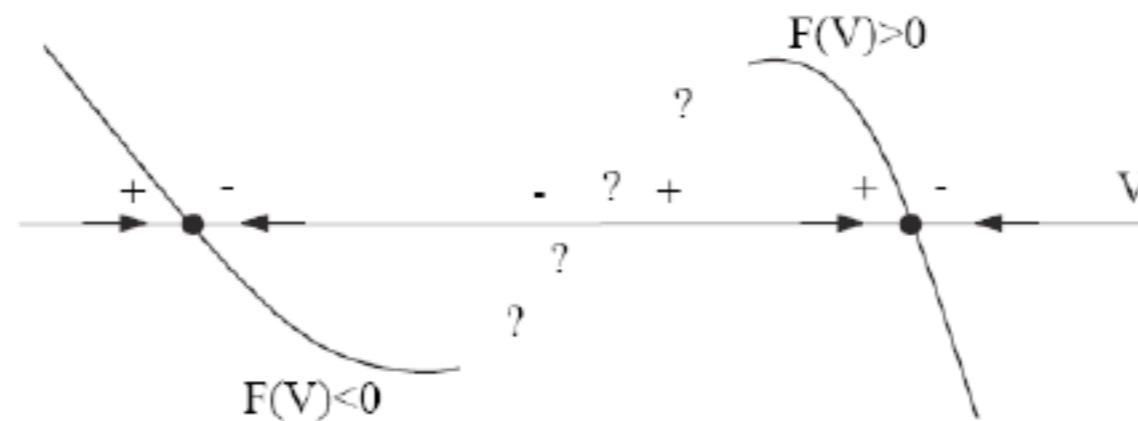
One-dimensional neural model

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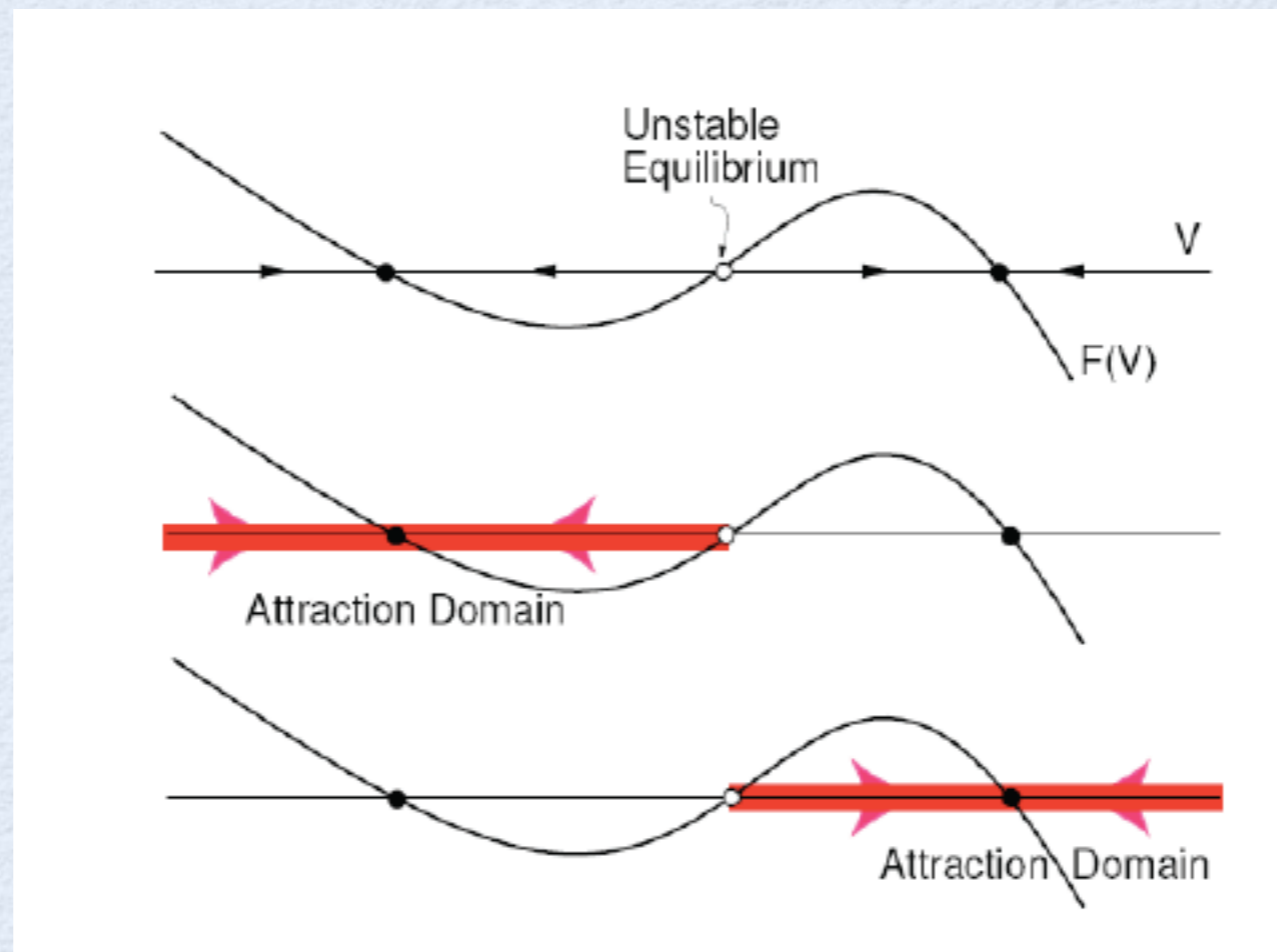
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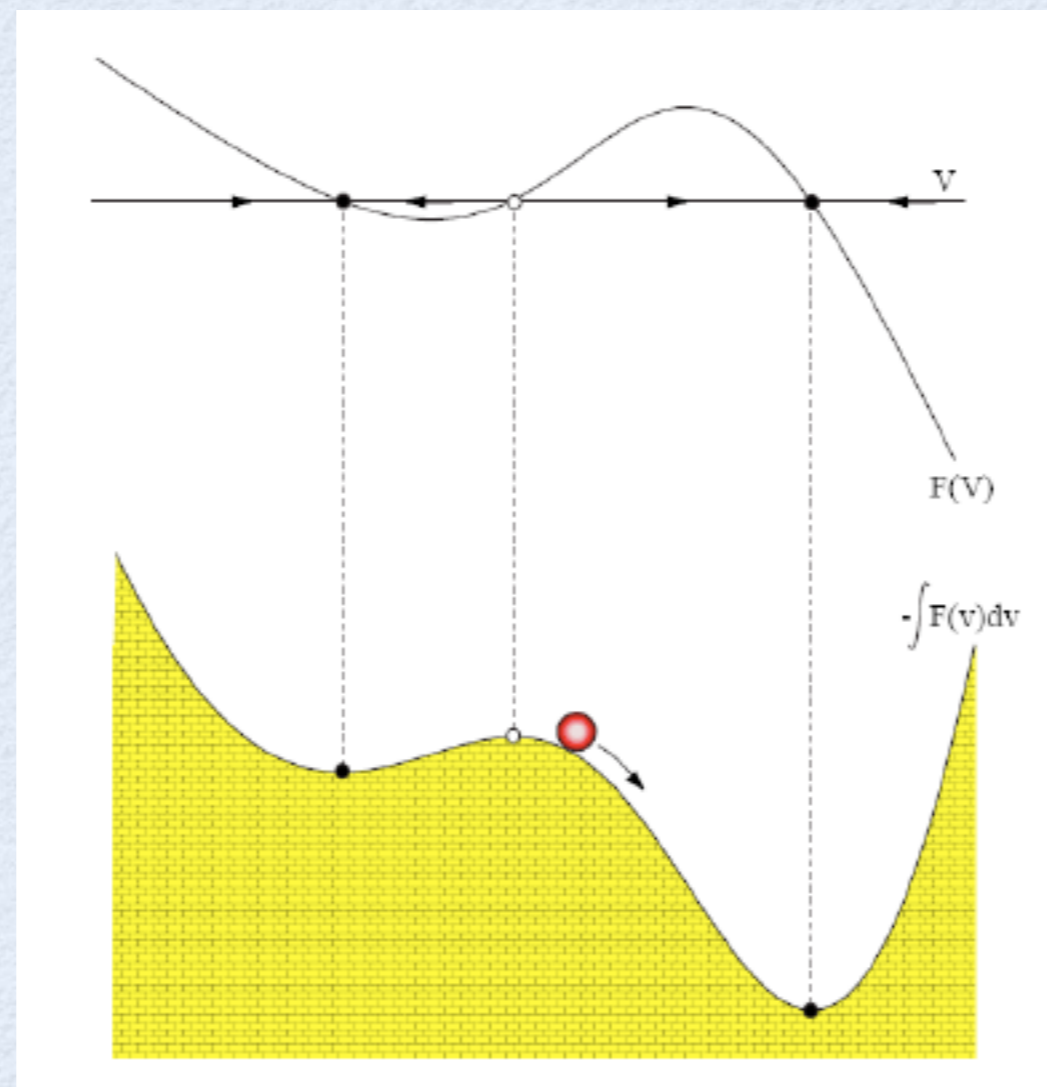
One-dimensional neural model

- Threshold and action potential



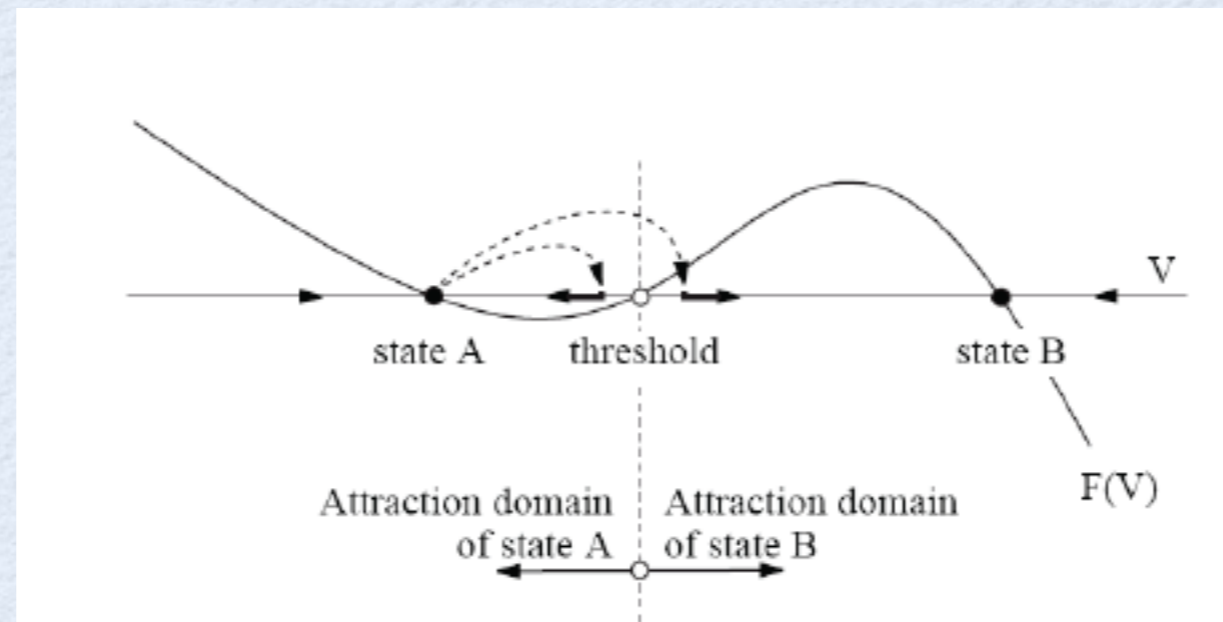
One-dimensional neural model

- Threshold and action potential - mechanistic (energy) interpretation



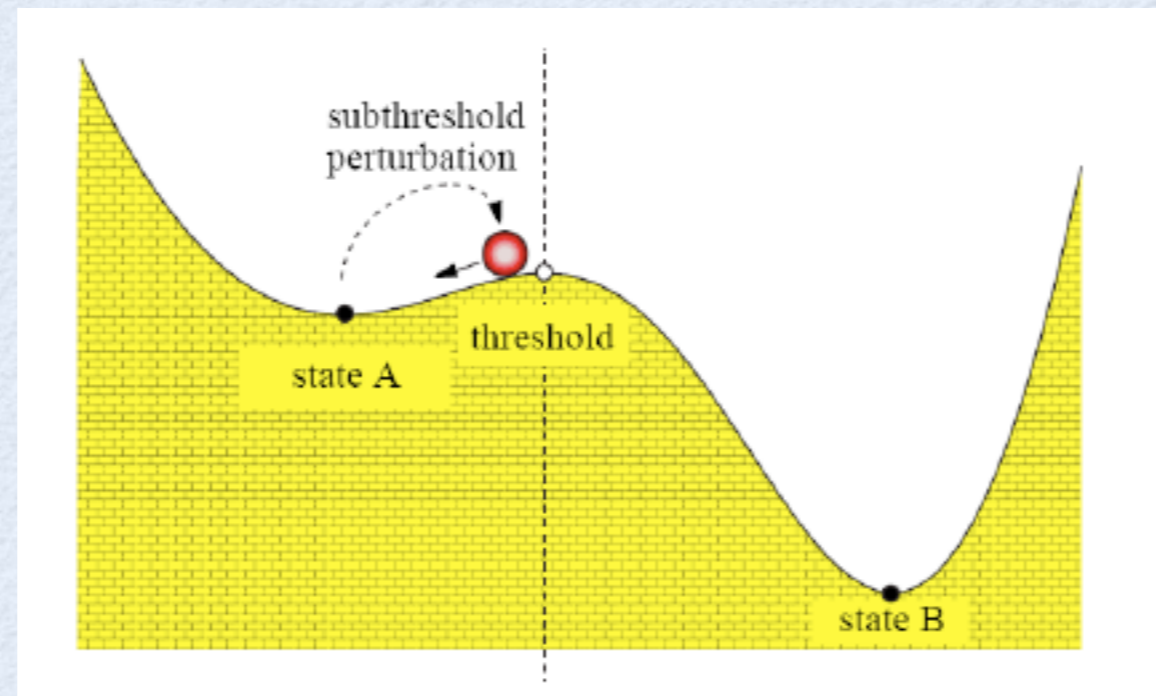
One-dimensional neural model

- Threshold and action potential - mechanistic (energy) interpretation



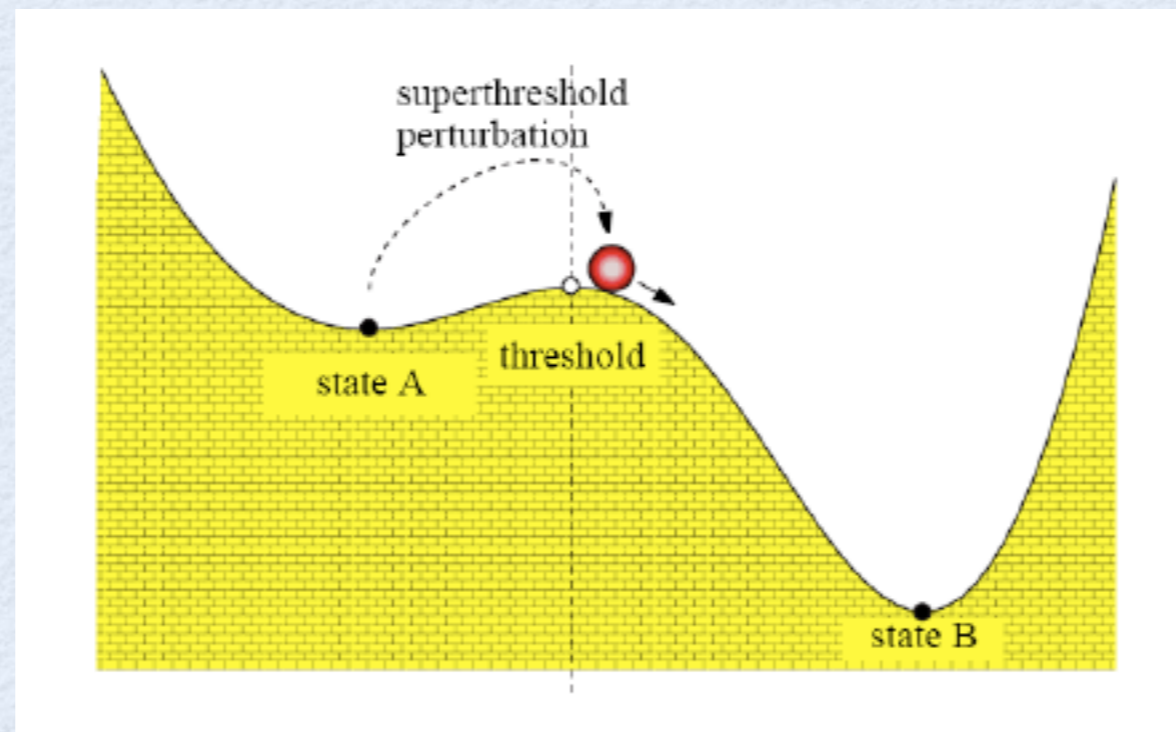
One-dimensional neural model

- Threshold and action potential - mechanistic (energy) interpretation



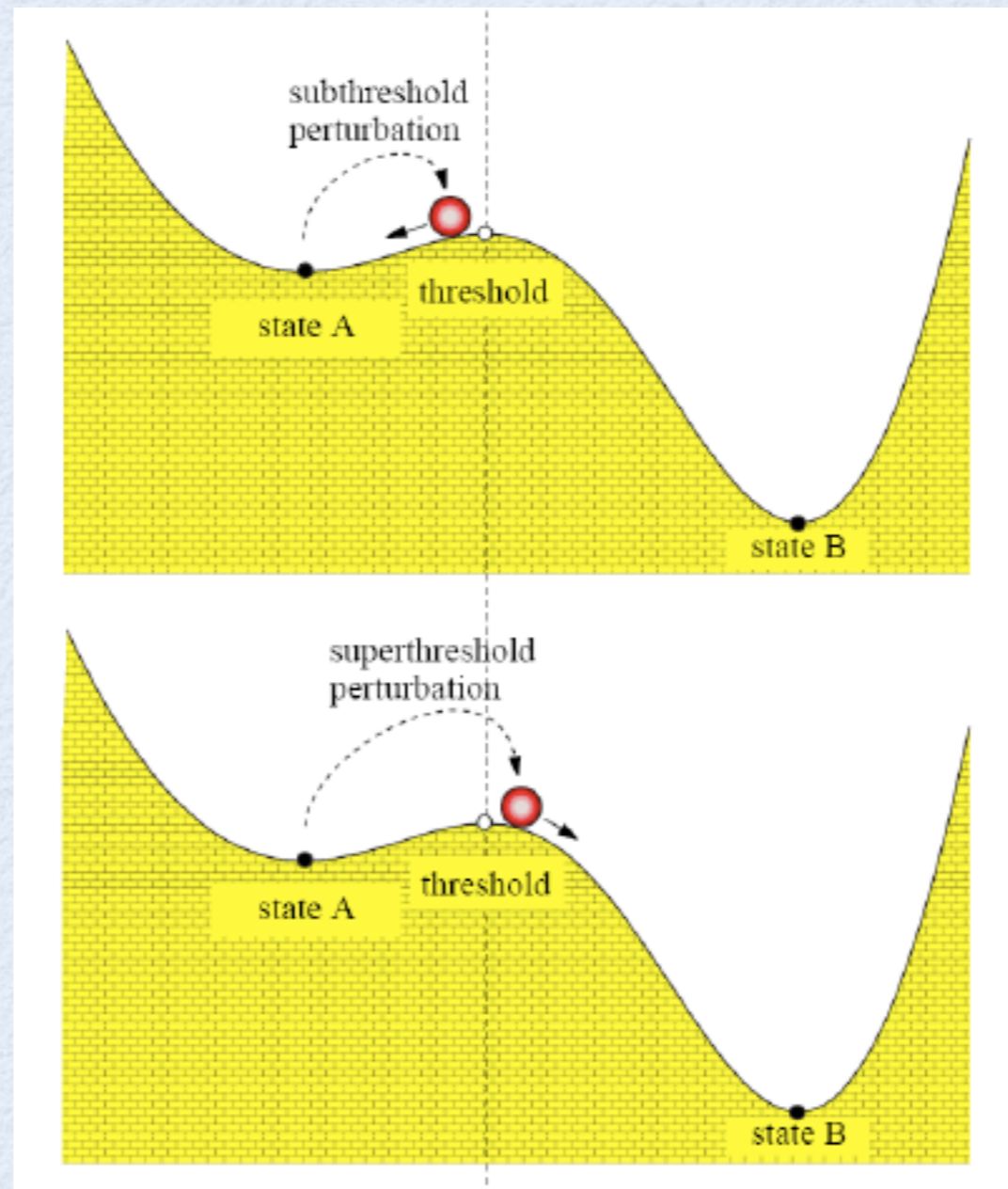
One-dimensional neural model

- Threshold and action potential - mechanistic (energy) interpretation



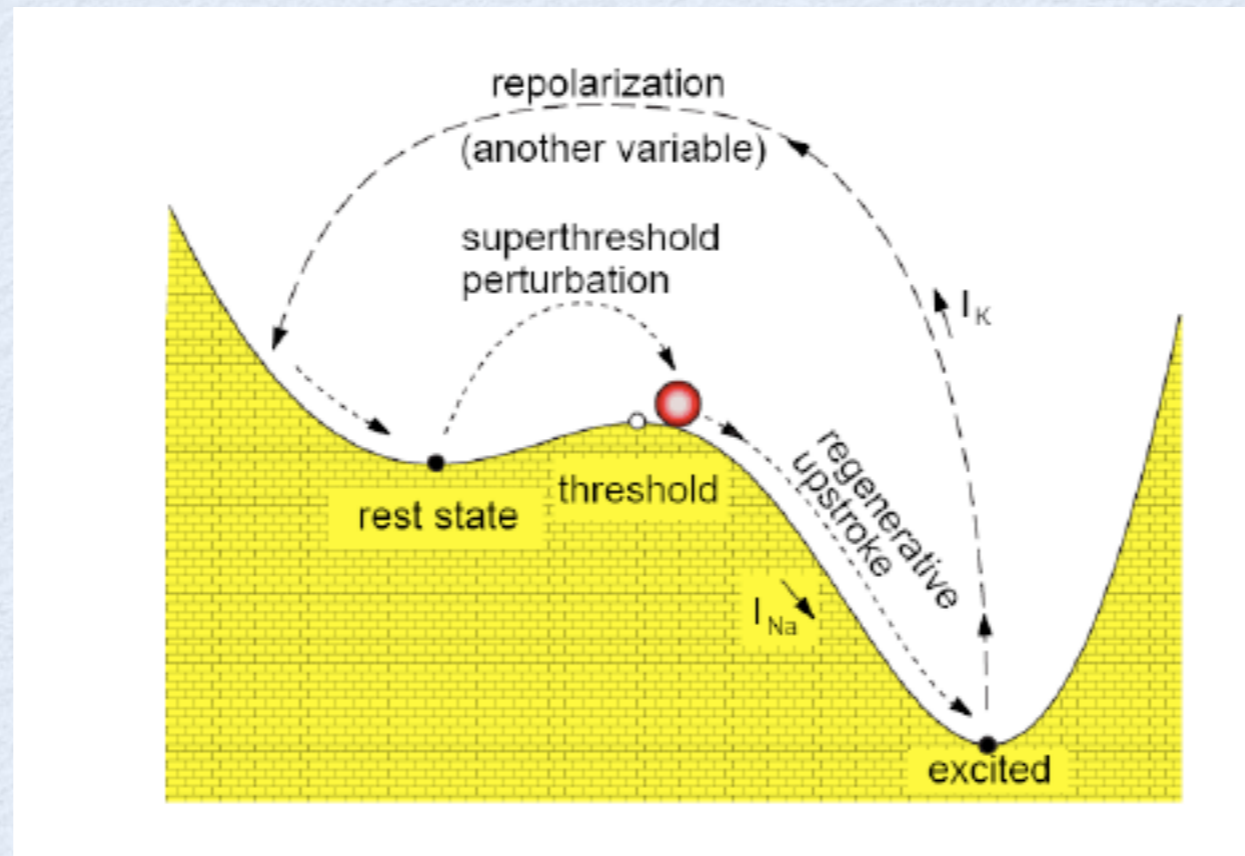
One-dimensional neural model

- Threshold and action potential - mechanistic (energy) interpretation



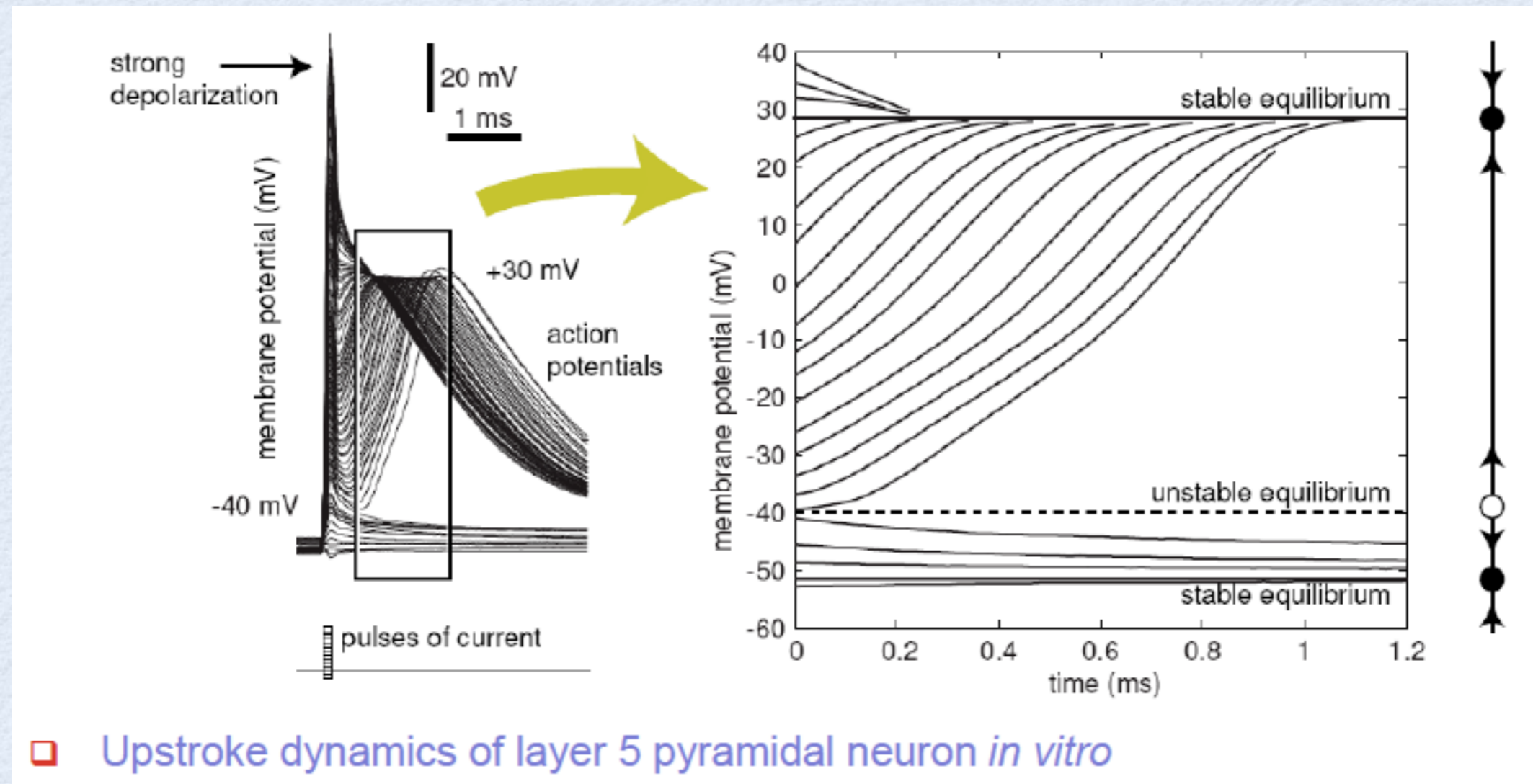
One-dimensional neural model

- Threshold and action potential - mechanistic (energy) interpretation



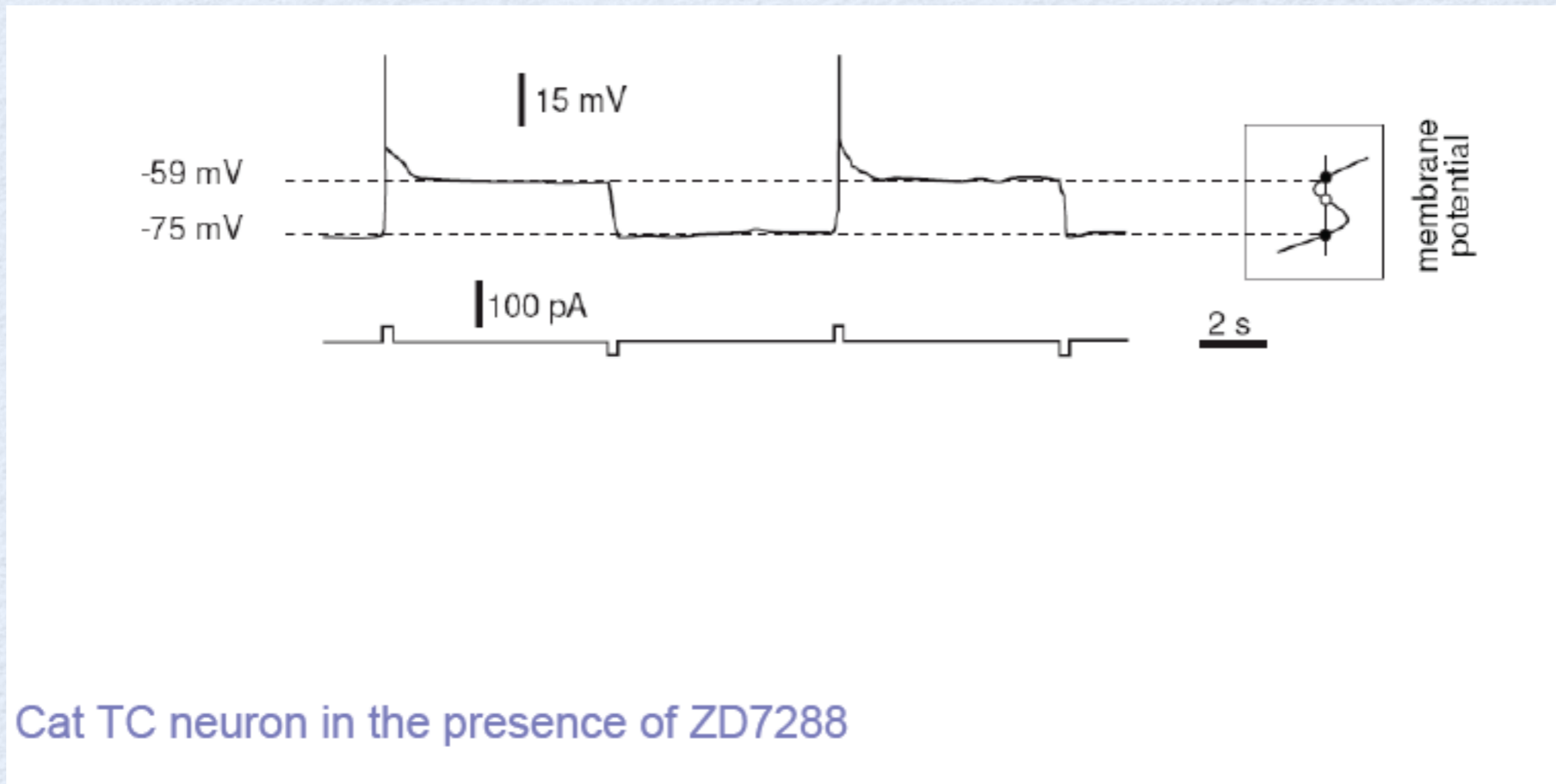
One-dimensional neural model

- Threshold and action potential



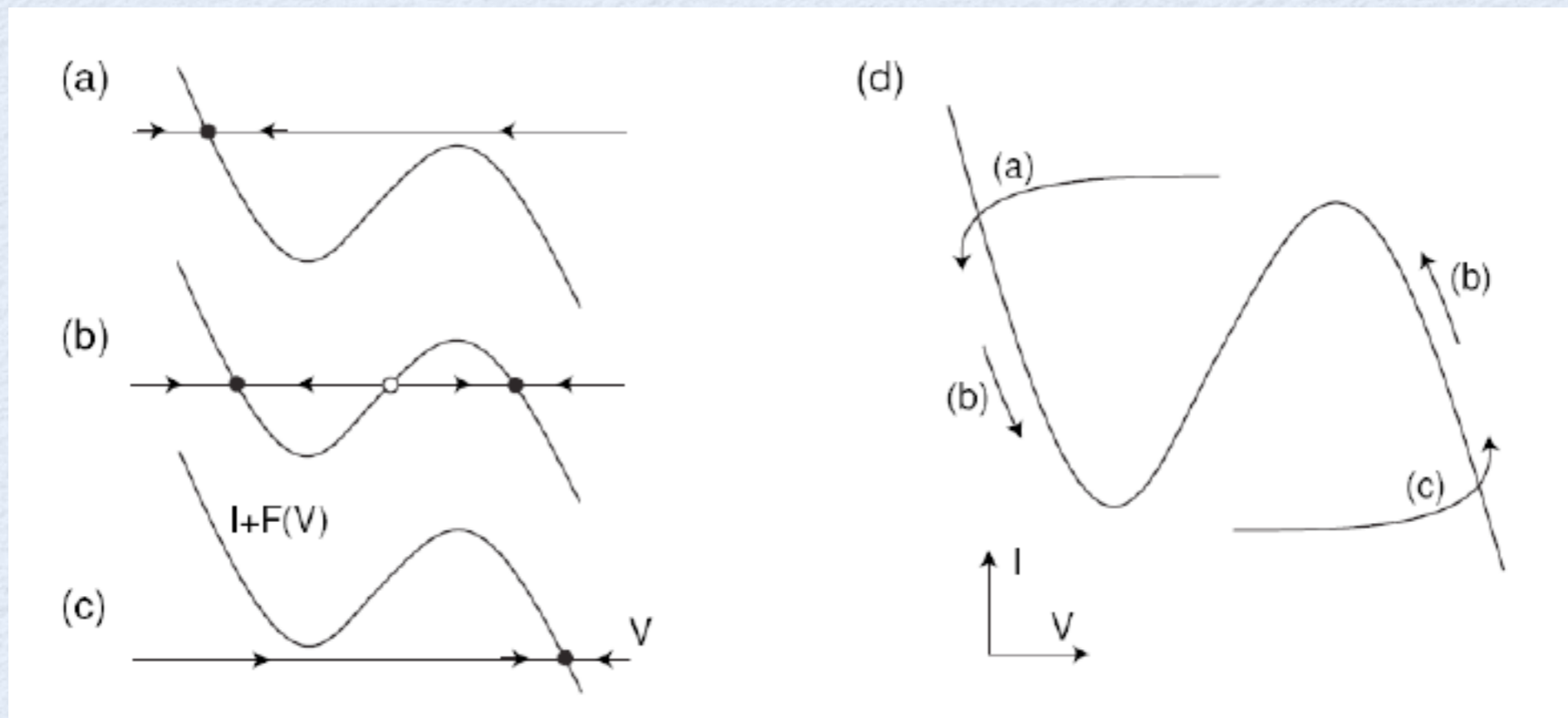
One-dimensional neural model

- Bistability and hysteresis



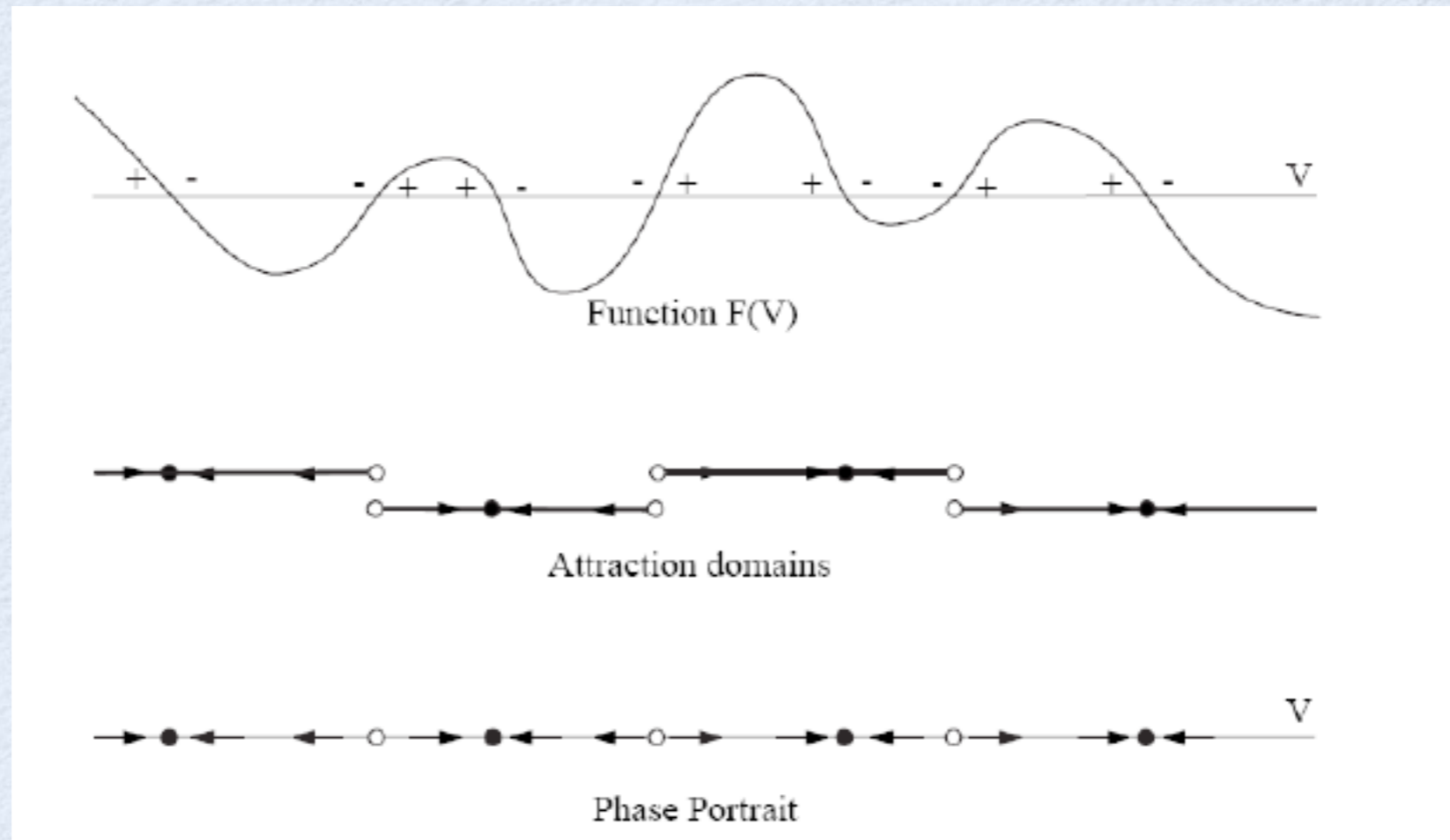
One-dimensional neural model

- Bistability and hysteresis as I changes



One-dimensional neural model

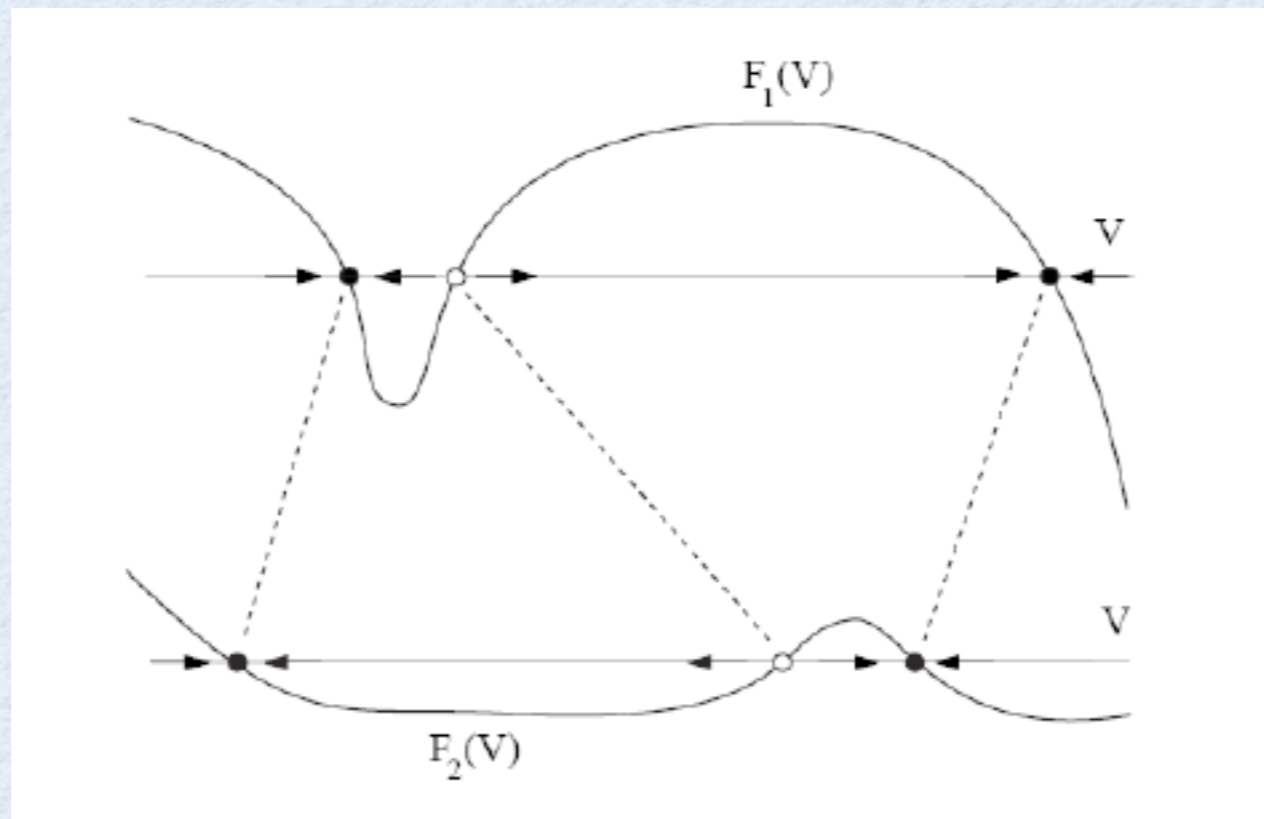
- Multistability



Phase portrait of a one-dimensional system $dV/dt = F(V)$

One-dimensional neural model

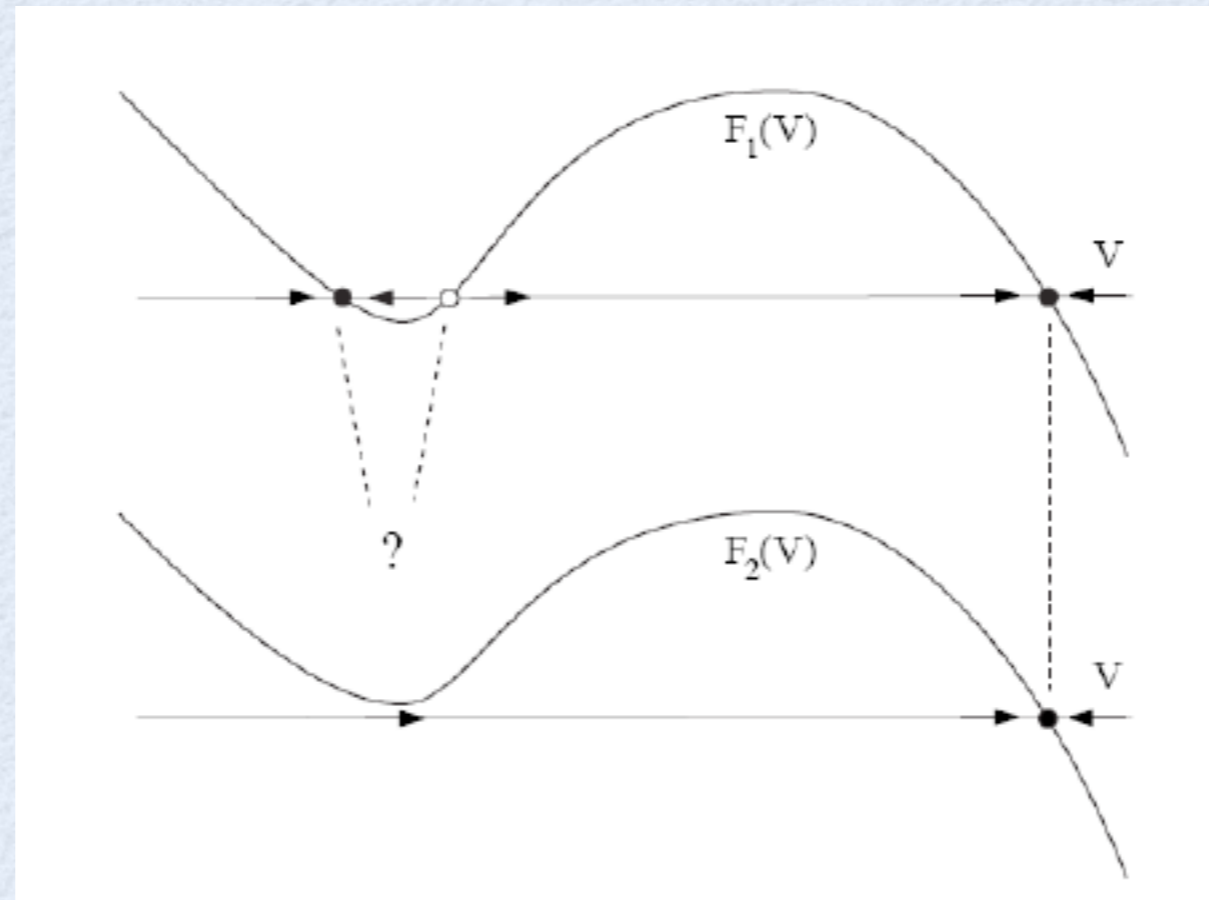
- Topological equivalence



Phase portrait of two seemingly different one-dimensional systems $dV_1 / dt = F(V_1)$ and $dV_2 / dt = F(V_2)$

One-dimensional neural model

- Topological equivalence

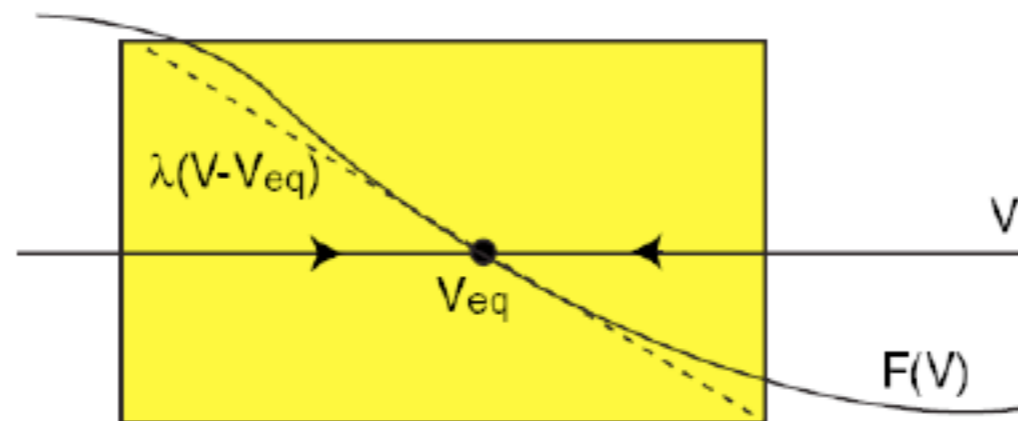


Phase portrait of two seemingly different a one-dimensional systems $dV_1 / dt = F(V_1)$ and $dV_2 / dt = F(V_2)$

One-dimensional neural model

- Local equivalence

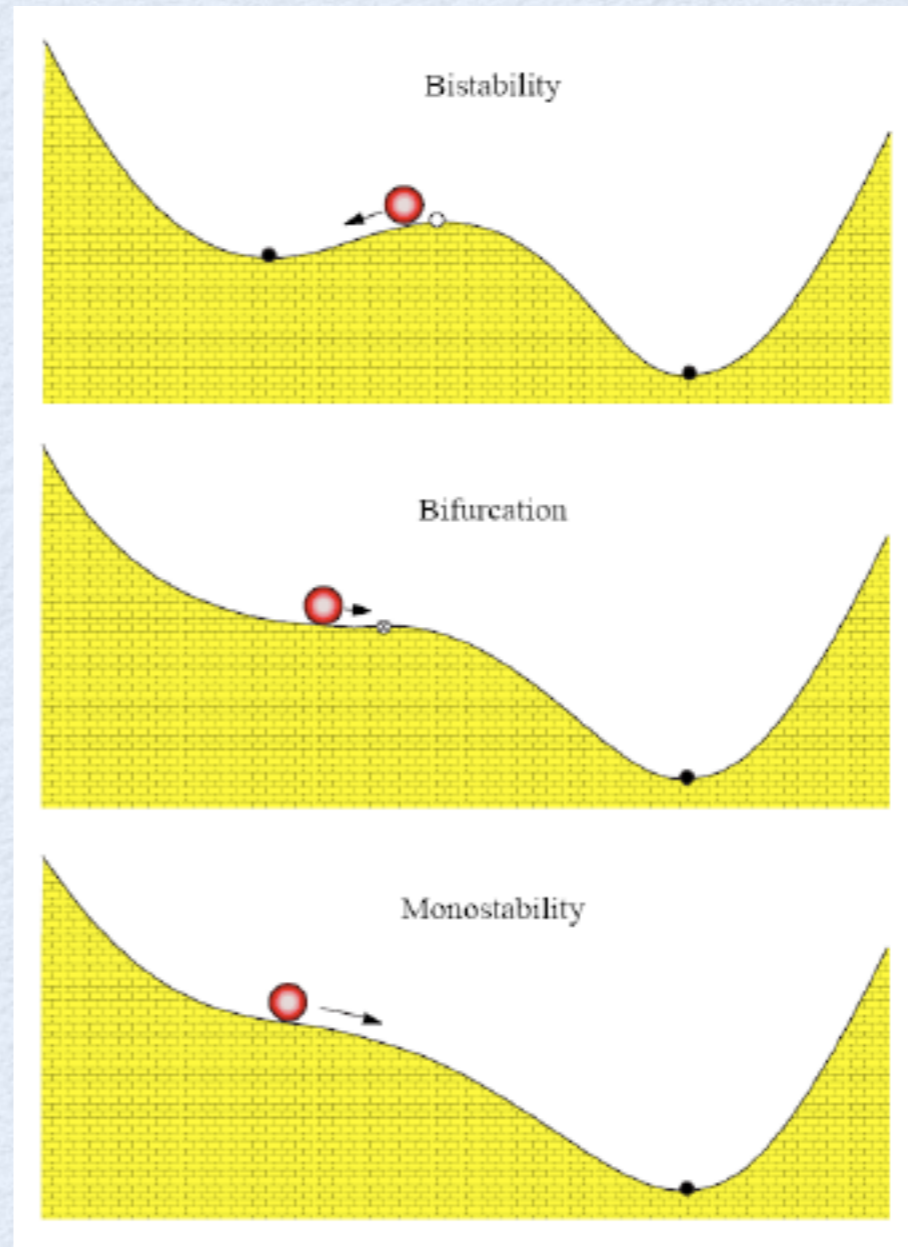
$$\frac{dV}{dt} = F(V) \quad \lambda = F'(V_{eq}) \neq 0 \quad || \quad \frac{dV}{dt} = \lambda (V - V_{eq})$$



Hartman-Grobman theorem: The two systems are topologically equivalent in the local (shaded) neighborhood of the hyperbolic equilibrium

One-dimensional neural model

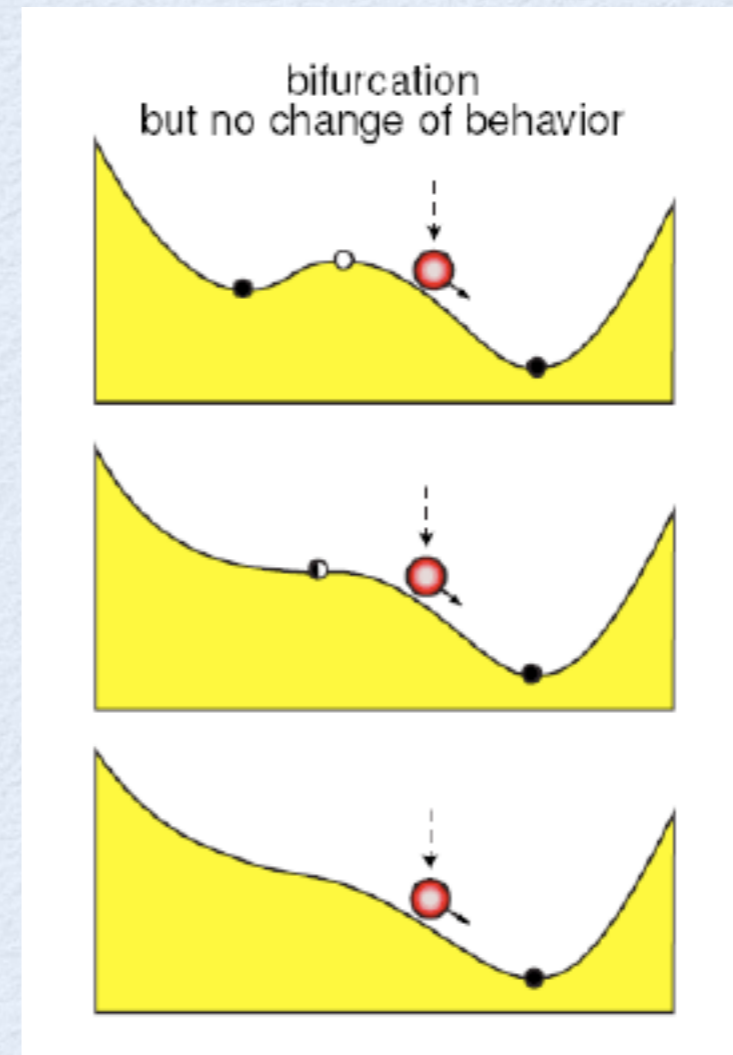
- Bifurcations



Mechanistic illustration of a bifurcation as a change of the landscape

One-dimensional neural model

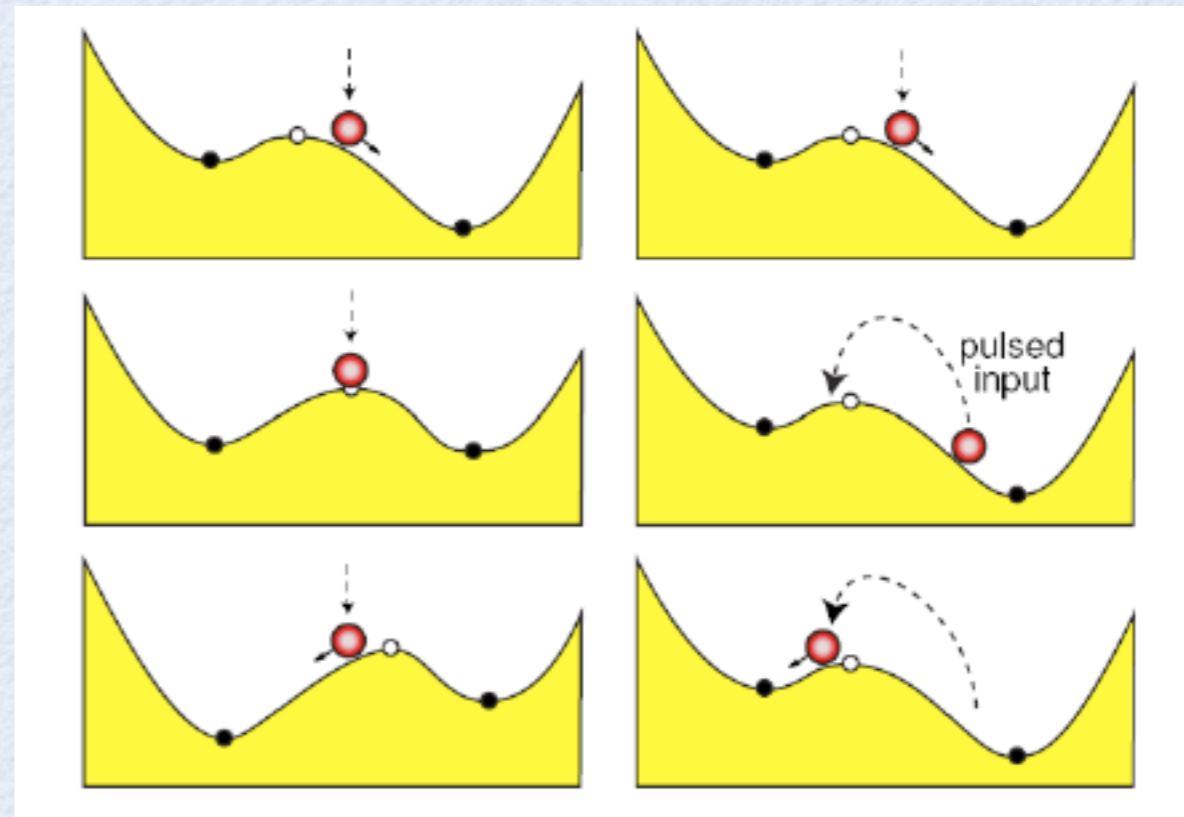
- Bifurcations



No change of behavior

One-dimensional neural model

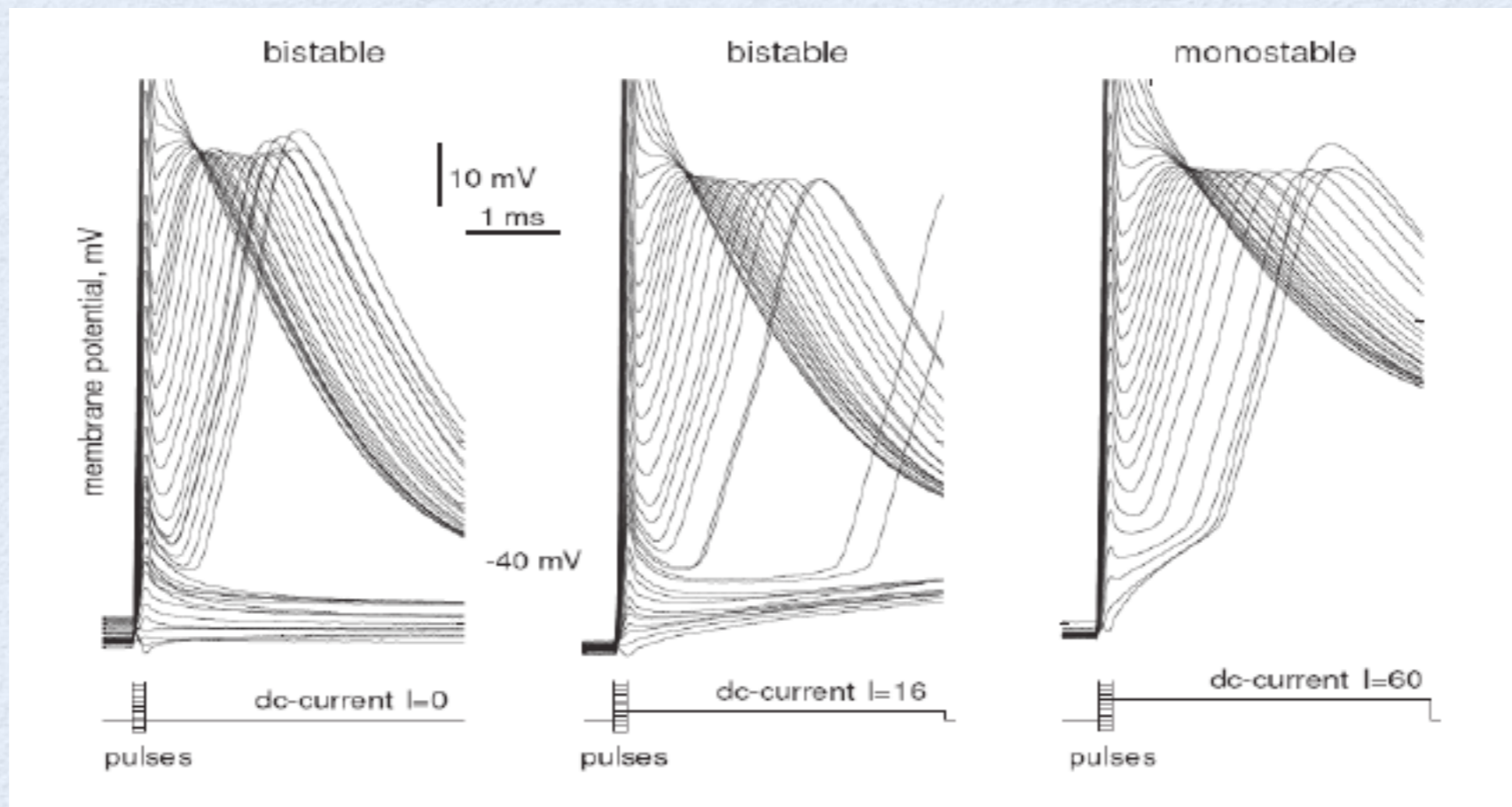
- Bifurcations



Change of behavior

One-dimensional neural model

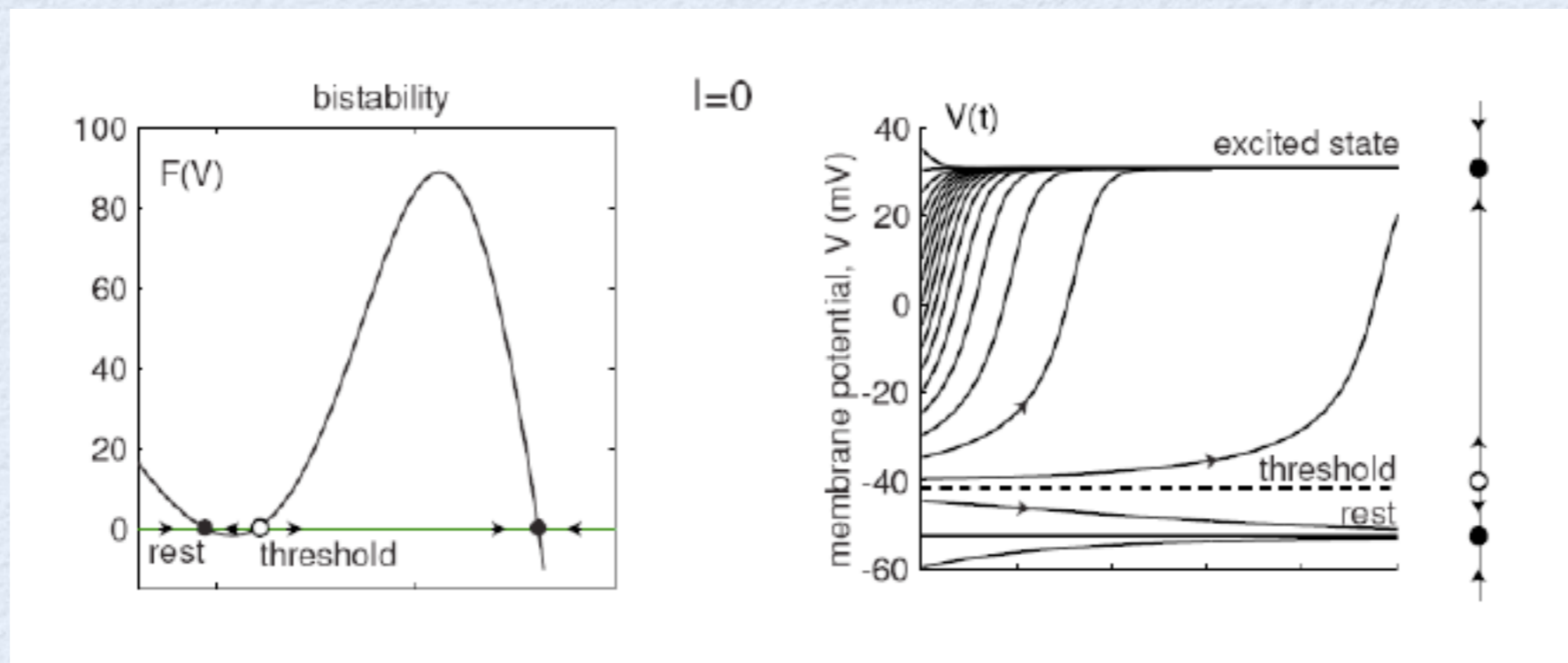
- Bifurcations



Qualitative change of the upstroke dynamics of layer 5 pyramidal neuron from rat visual cortex

One-dimensional neural model

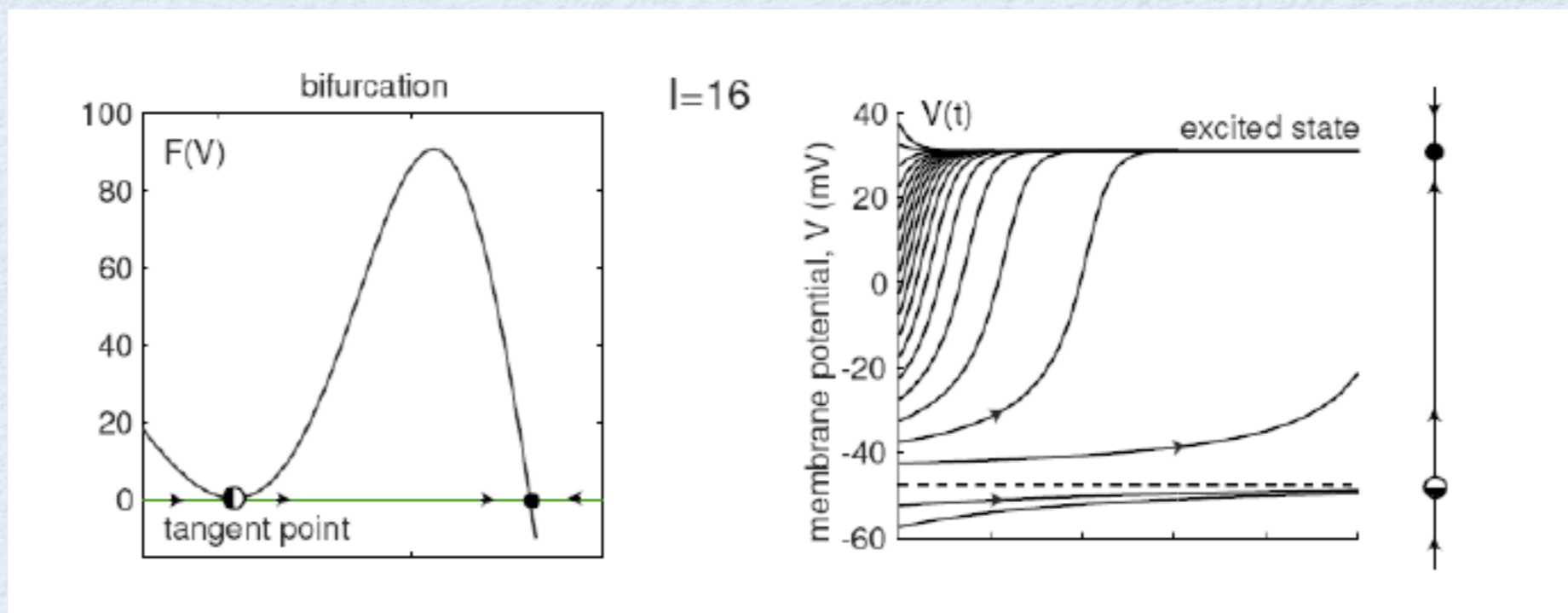
- Bifurcations in the $I_{Na,p}$ model



The resting state and the threshold state coalesce and disappear when the parameter I increases

One-dimensional neural model

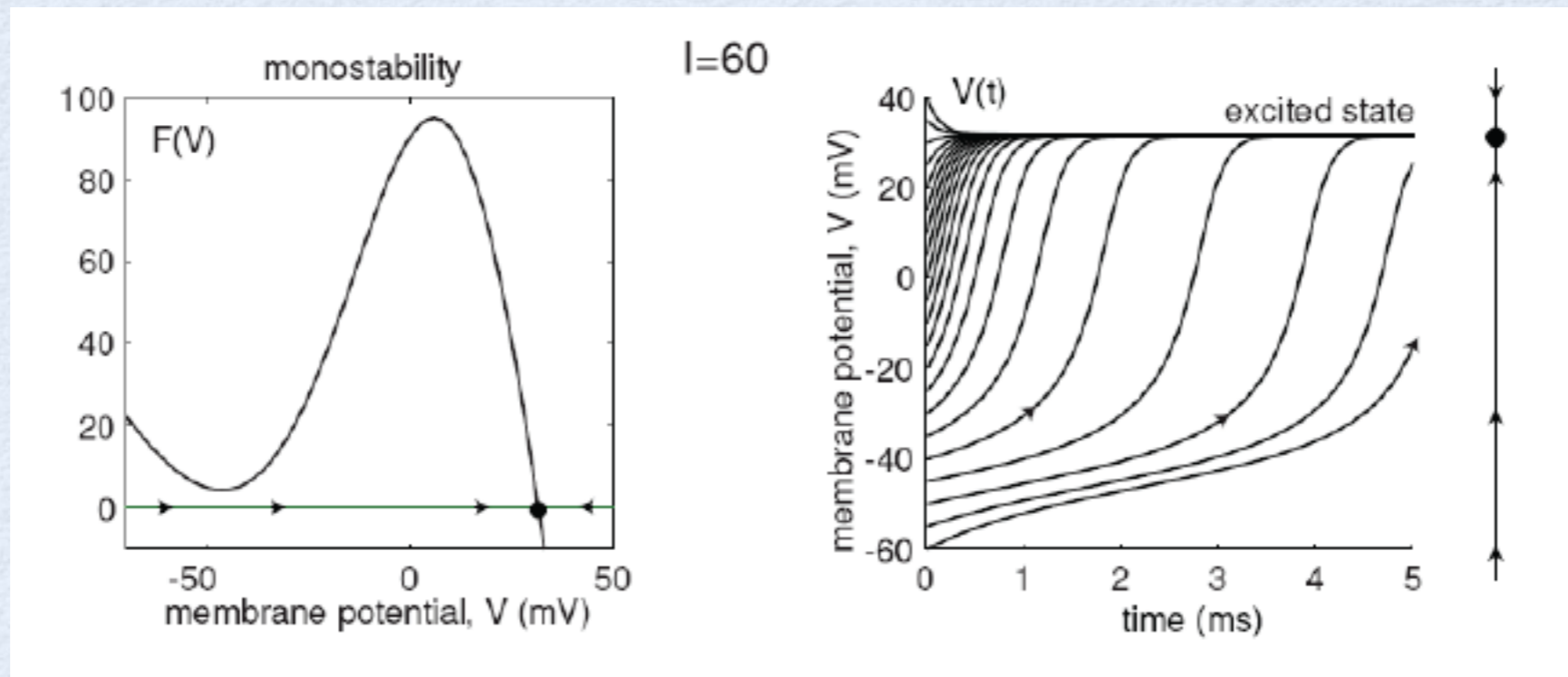
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The resting state and the threshold state coalesce and disappear when the parameter I increases

One-dimensional neural model

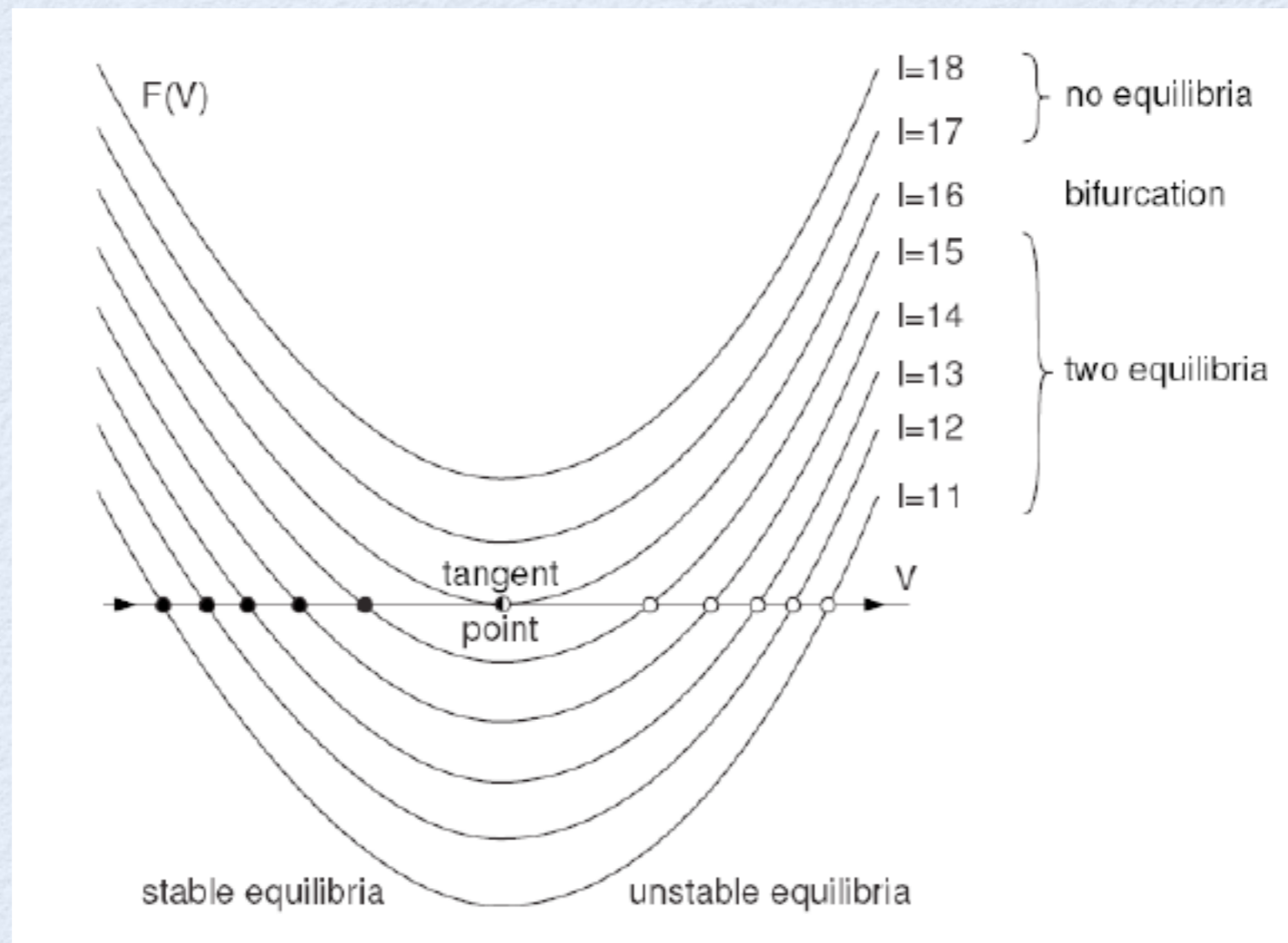
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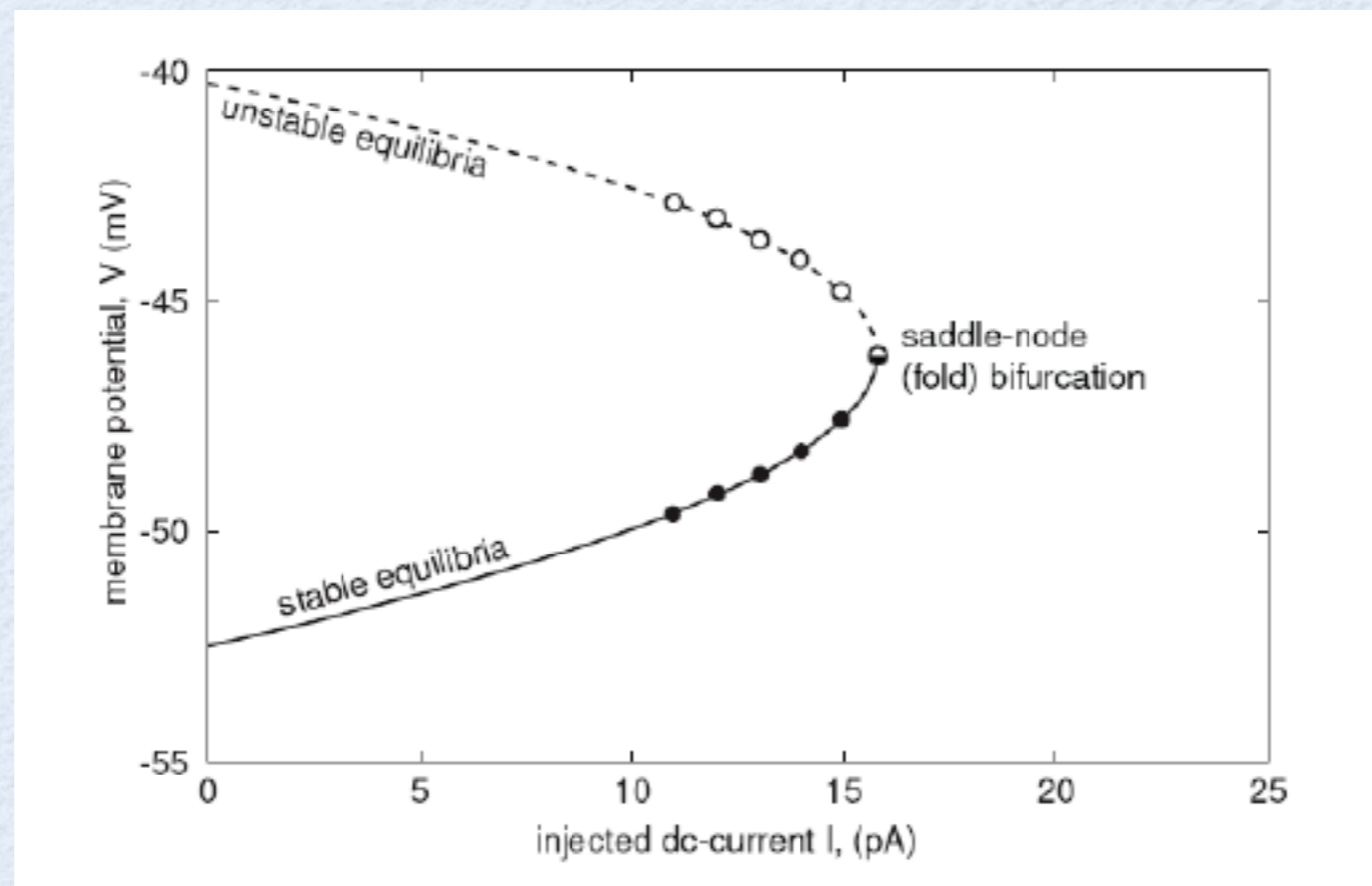
One-dimensional neural model

- Bifurcations



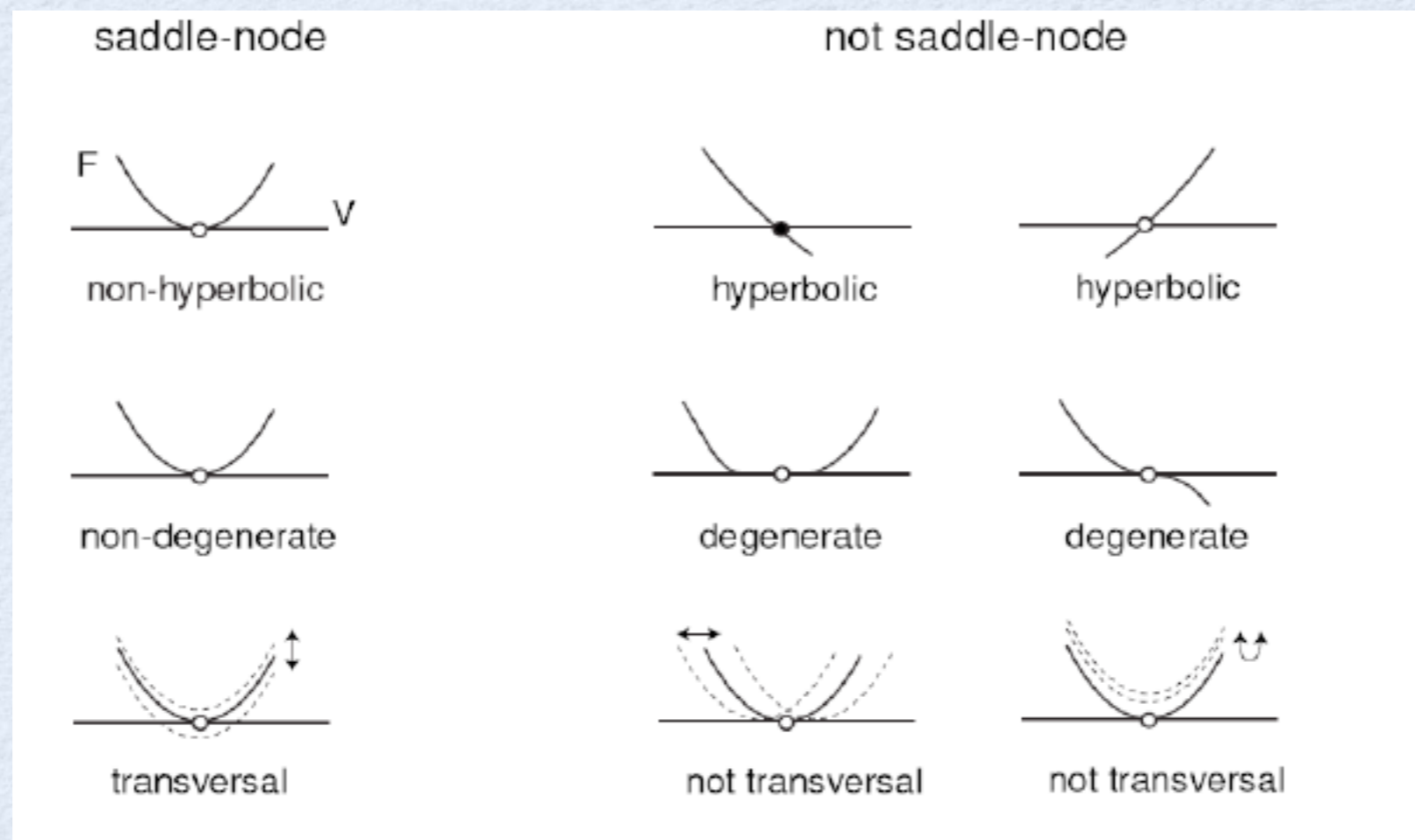
One-dimensional neural model

- Bifurcation diagram



One-dimensional neural model

- Bifurcations: conditions defining saddle-node bifurcations



Arrows denote the direction of displacement of the function $F(V, I)$ as the bifurcation parameter I changes

One-dimensional neural model

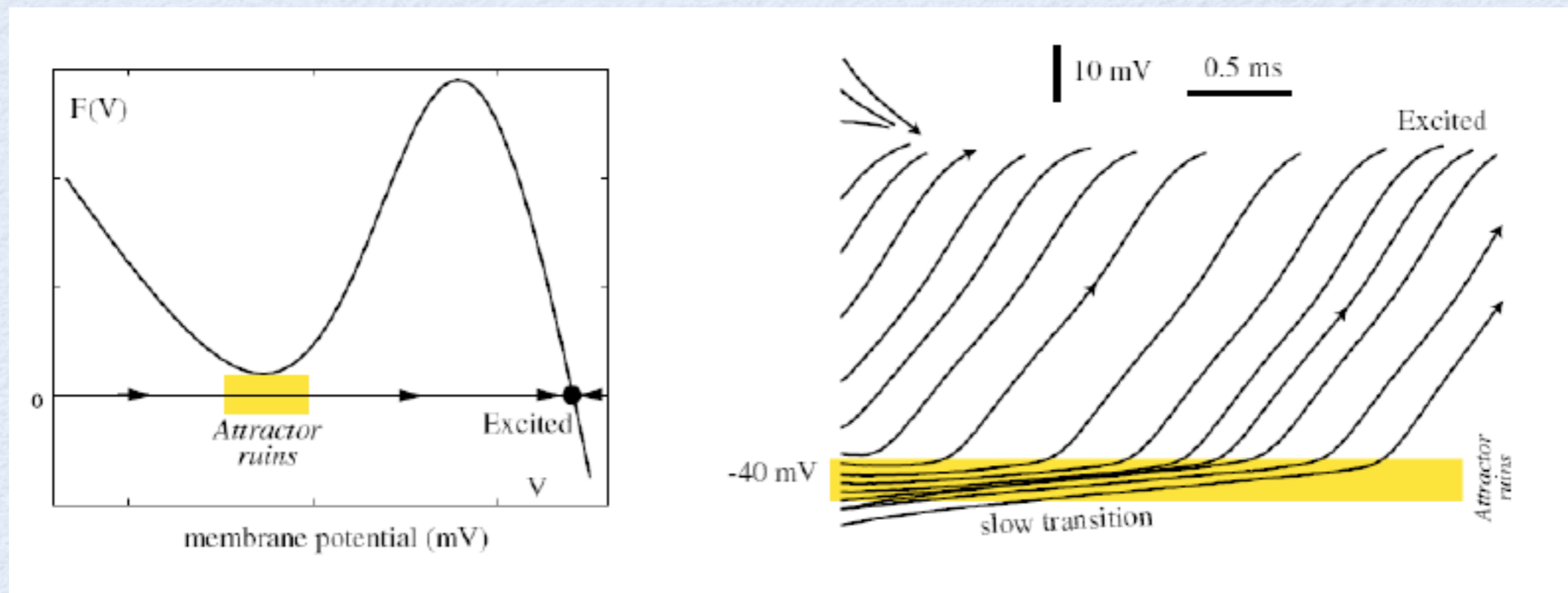
- Topological normal form for the saddle-node bifurcation

$$\frac{dV}{dt} = I + V^2$$

The system has a saddle-node bifurcation when $V=0$ and $I=0$

One-dimensional neural model

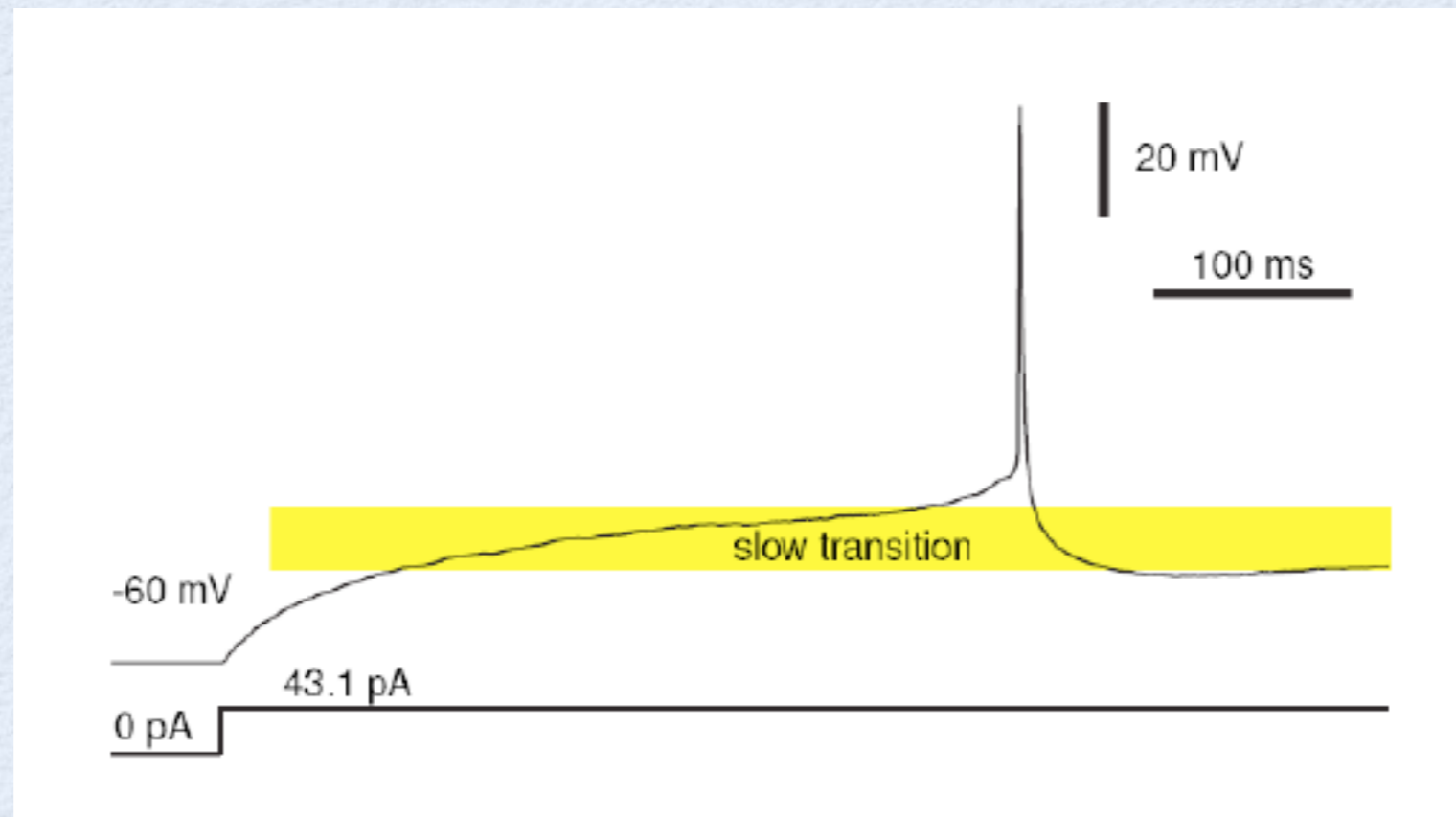
- Slow transitions



Slow transition through the ghost of the resting state attractor in a cortical pyramidal neuron with $I=30$ pA

One-dimensional neural model

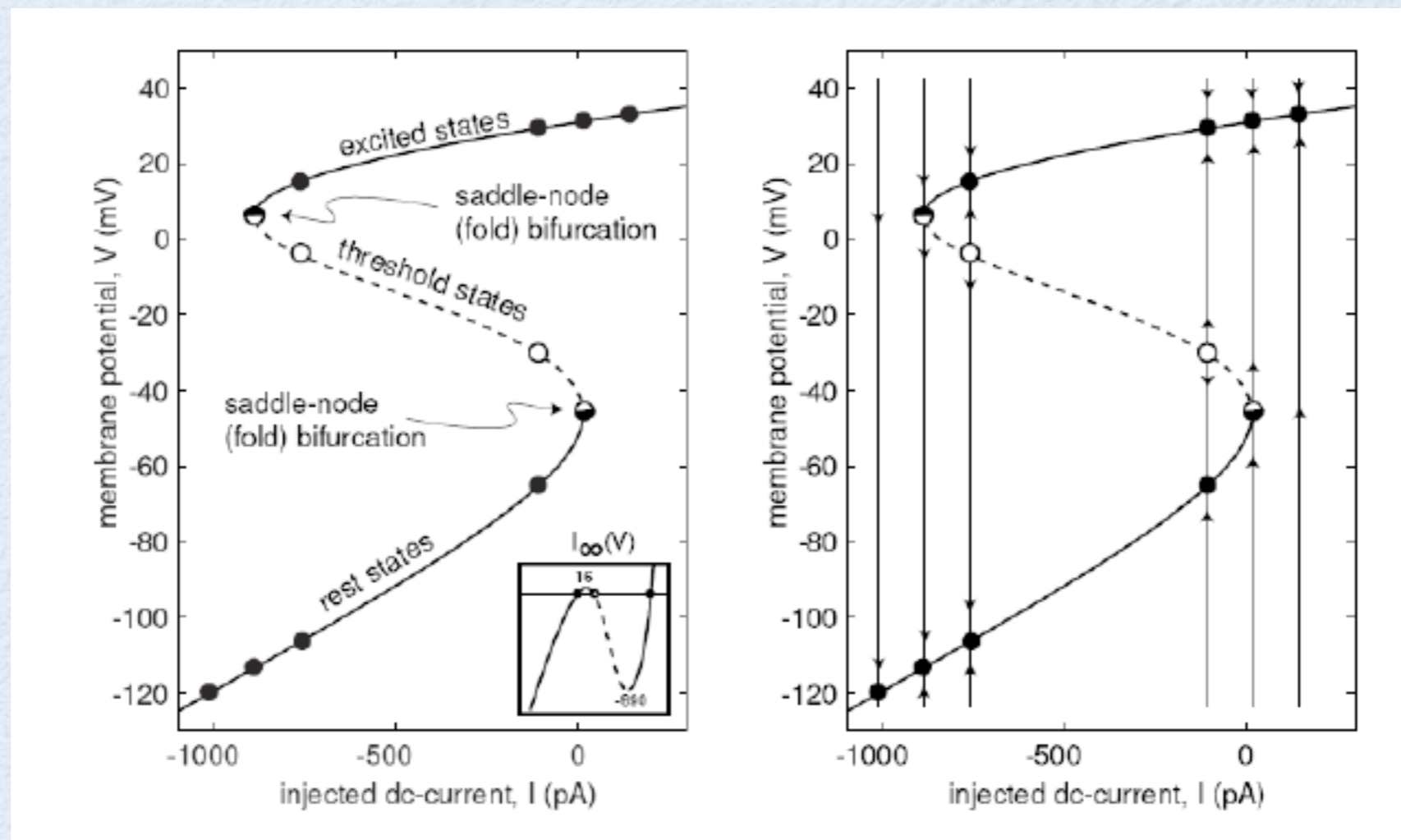
- Slow transitions



A 400 ms latency in a layer 5 pyramidal neuron of rat visual cortex

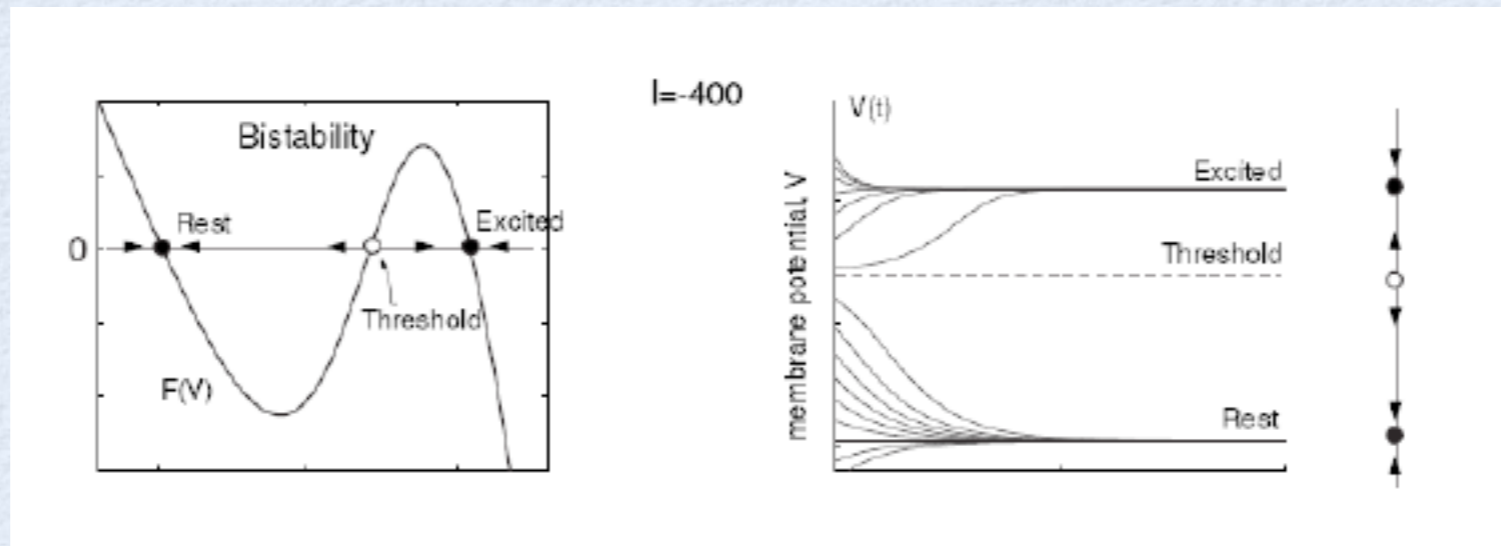
One-dimensional neural model

- Bifurcation diagram of the $I_{Na,p}$ model



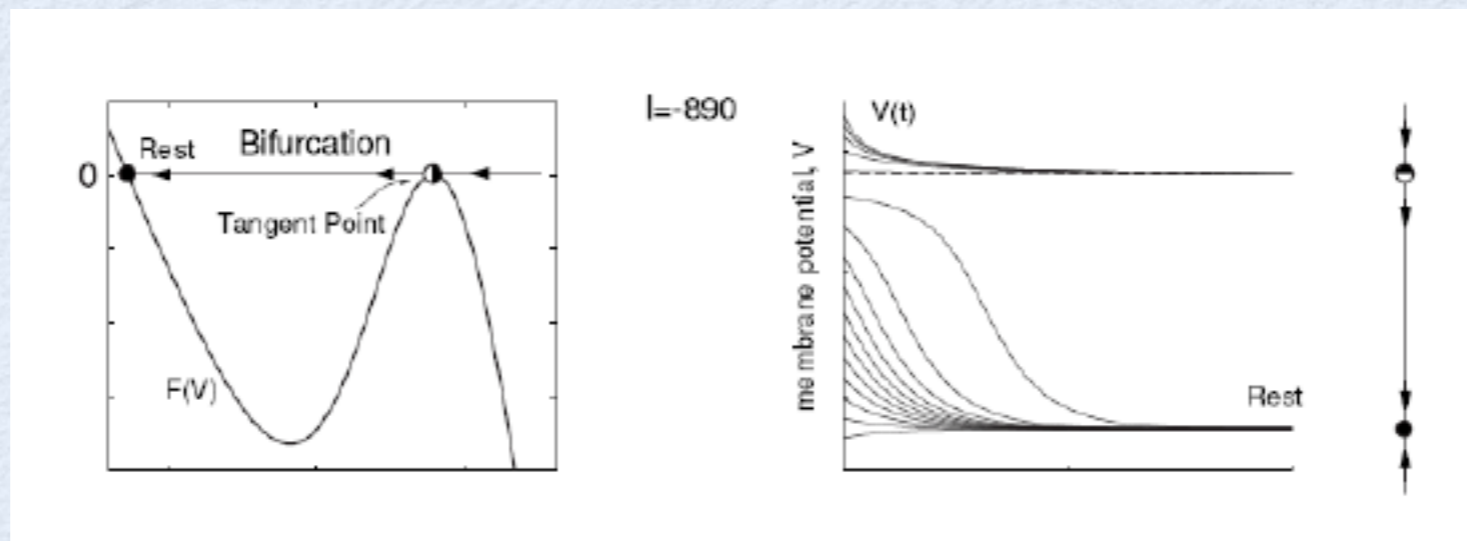
One-dimensional neural model

- Bifurcation diagram of the $I_{Na,p}$ model



One-dimensional neural model

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One-dimensional neural model

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