

Reduction of neurons to phases

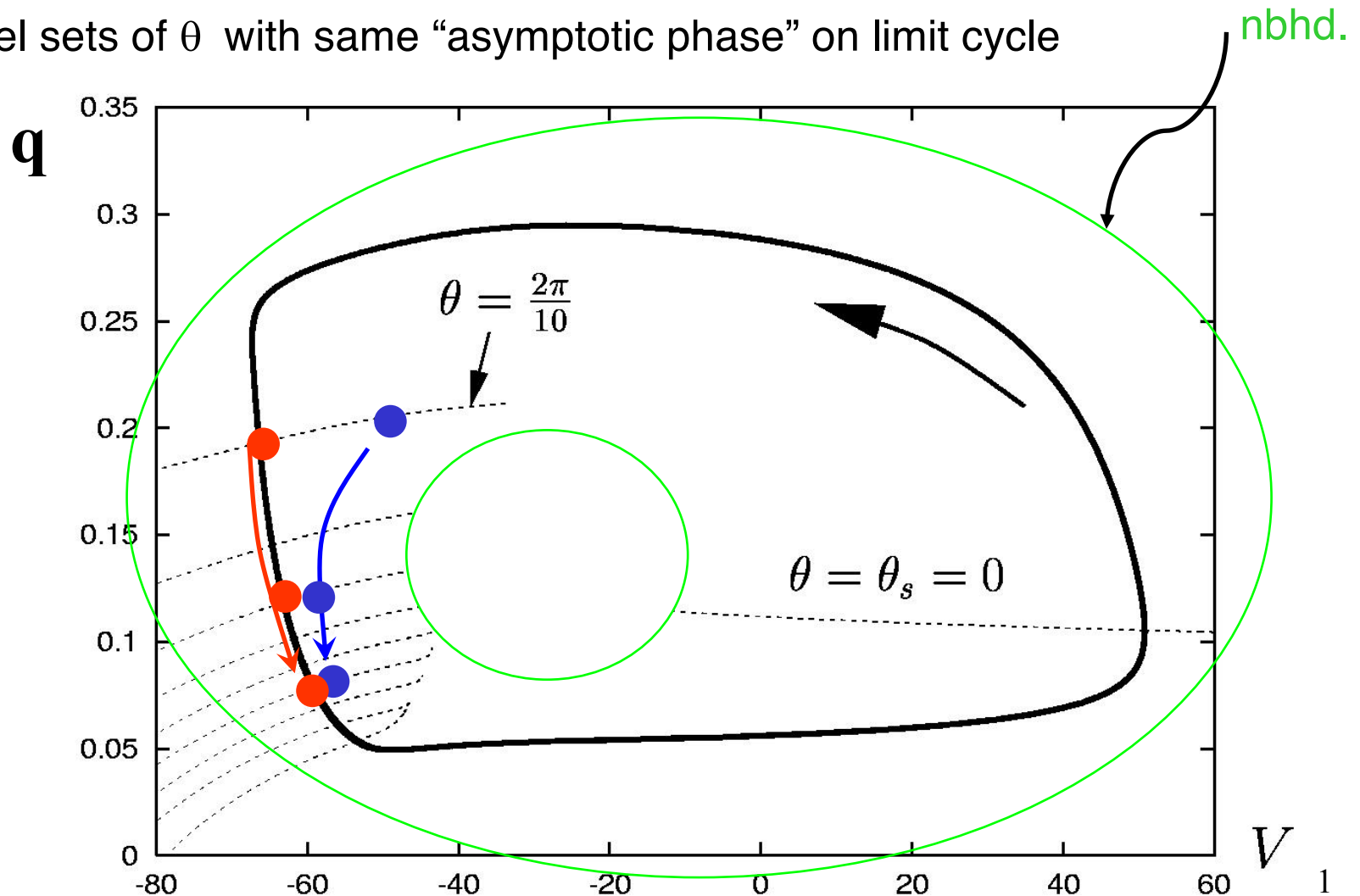
[Coddington and Levinson, 1955, Winfree, 1974, Guckenheimer, 1985]

In nbhd. of limit cycle, define variable $\theta(x)$ such that:

$$\frac{d\theta}{dt} = \omega \text{ along trajectories, where } \omega = \frac{2\pi}{T}$$

Strategy: start on limit cycle itself, where say $V(\theta) = V(\omega t)$.

Then define level sets of θ with same “asymptotic phase” on limit cycle

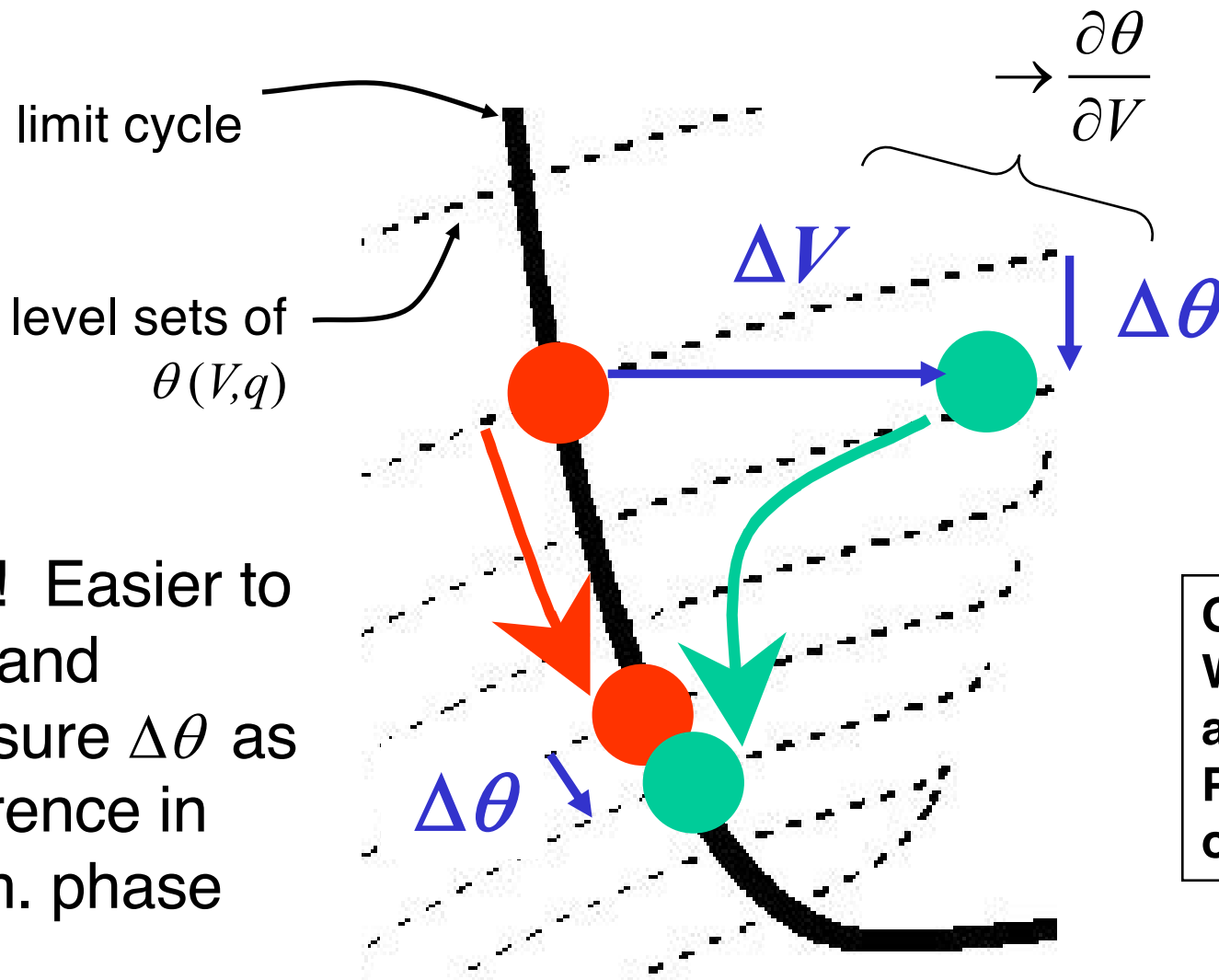


Following
Brown, Moehlis,
Holmes
Neural Comp 04

Finding $z(\theta)$.

Asymptotic phase property of field $\theta(x)$ gives nice way to calculate

$$z(\theta) = \lim_{\Delta V \rightarrow 0} \frac{\Delta \theta}{\Delta V}$$



standard way to calculate partial deriv. -- must know $\theta(x)$ in nbhd. of lim cycle.

BUT! Easier to wait and measure $\Delta \theta$ as difference in asym. phase

Glass and Mackey, Winfree, Ermentrout and Kopell, Izhikevich, Park and Kim, and others

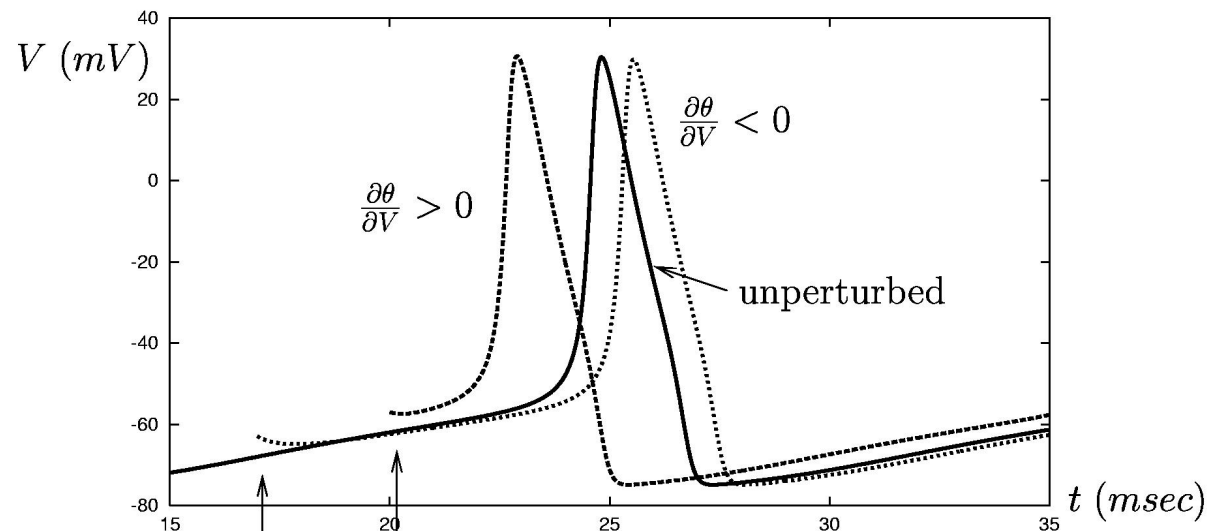
Finding $z(\theta)$.

Calculating the phase response curve:

- Perturb an uncoupled neuron with a brief voltage stimulus ΔV at different times in its cycle, parameterized by θ
- Measure the resulting phase-shift $\Delta\theta$ with respect to the unperturbed system

For Hodgkin-Huxley neurons with $I = 10\mu A/cm^2$:

$$z(\theta) = \lim_{\Delta V \rightarrow 0} \frac{\Delta\theta}{\Delta V}$$



perturb with 5 mV stim.

Jeff Moehlis

Have phase dynamics ... that you could directly derive from the laboratory !

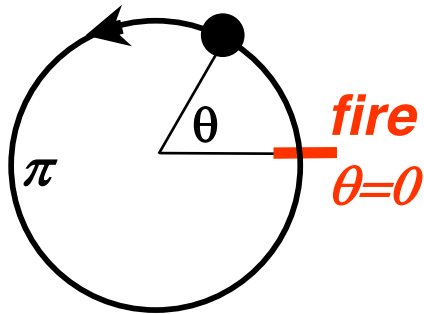
Glass and Mackey, Winfree

$$\frac{d\theta}{dt} = \omega + z(\theta) \times \overline{I_{syn}(t)}$$

natural frequency

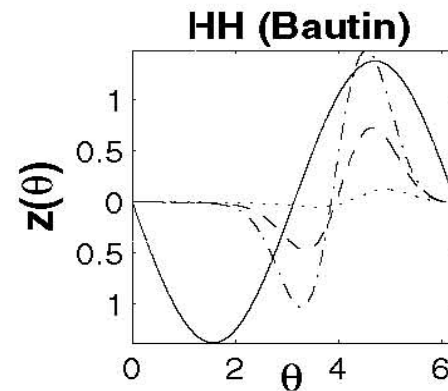
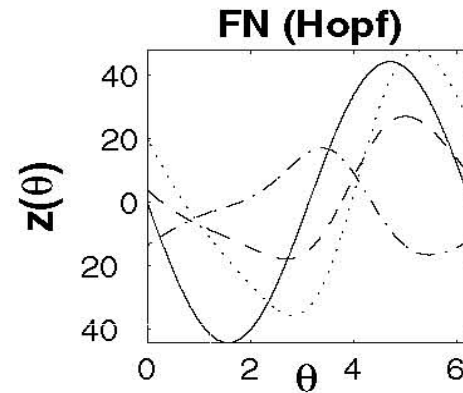
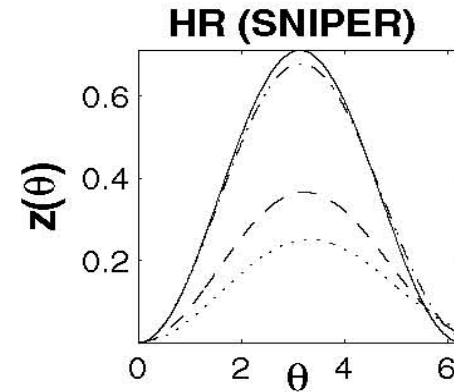
phase response curve
(phase sensitivity curve)

$$z(\theta) = \lim_{\Delta V \rightarrow 0} \frac{\Delta \theta}{\Delta V}$$



Phase response curves for different neurons look very different!

[Ermentrout and
Kopell,
Van-Vreeswick,
Bressloff, Izhikevich,
Moehlis, Holmes, S-B]



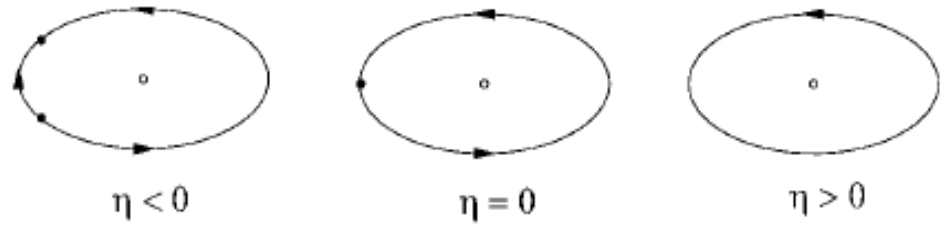
Hodgkin-Huxley

Following
Brown, Moehlis,
Holmes
Neural Comp 04

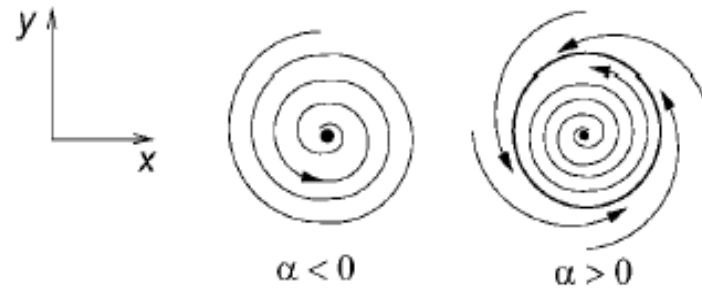
These differences follow from the underlying normal forms

SNIPER (Ermentrout, 1996)

Type I

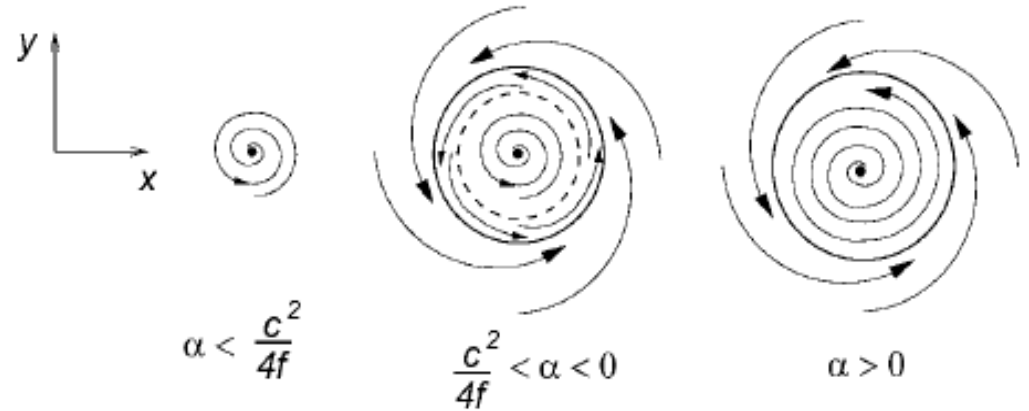


Hopf (Erm. + Kopell, 1984)

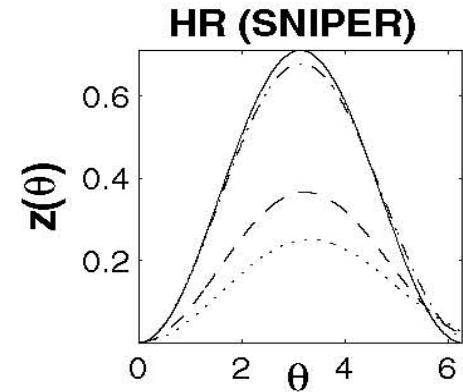


Type II

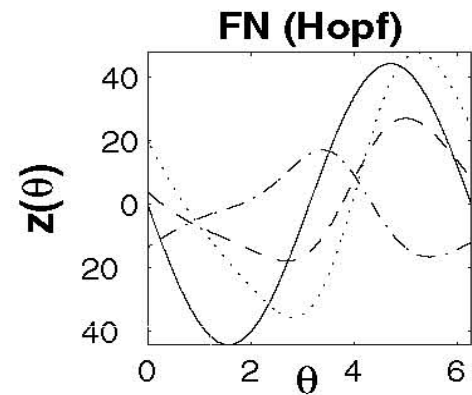
Degenerate Hopf / Bautin



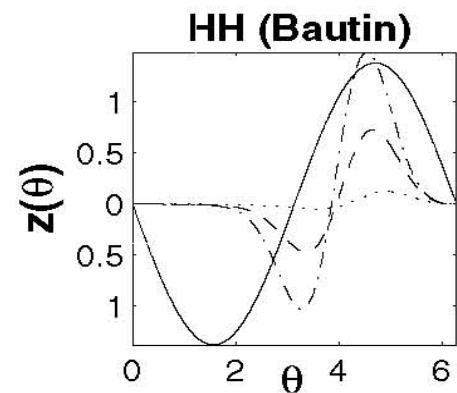
Phase response curves for normal form neuron models



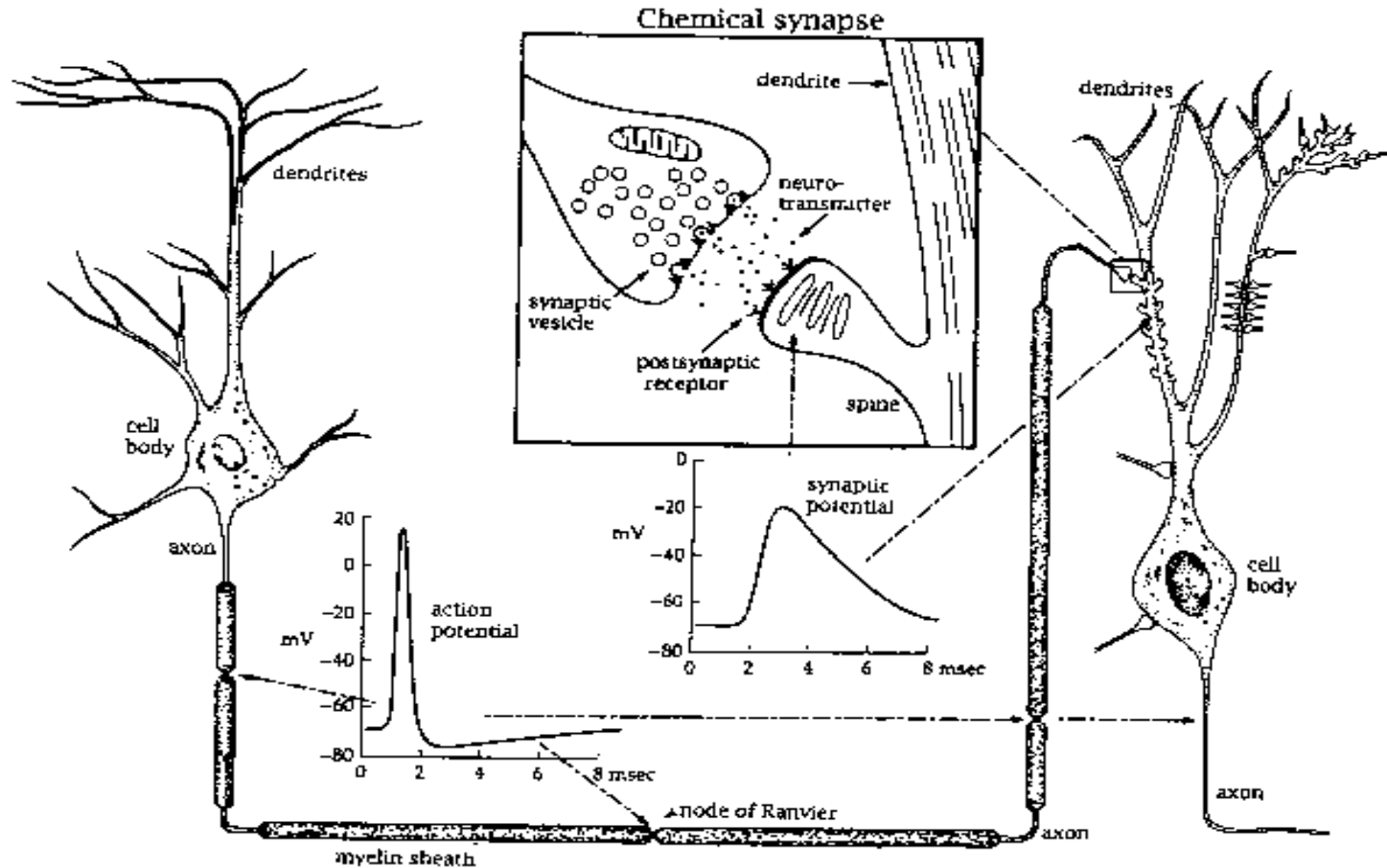
$$k(1 - \cos(\theta))$$



$$k_1 \sin(\theta - k_2)$$



Use phase response curves to study synchrony in “network” of two coupled neurons

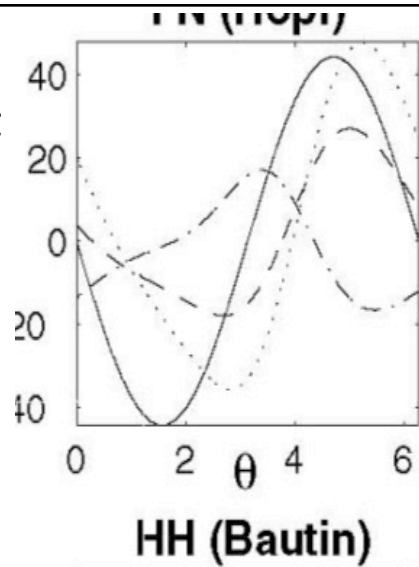


$$\begin{aligned}
 \frac{d\theta_1}{dt} &= \omega + z(\theta_1) * h\delta(t - t_2^j) \\
 \frac{d\theta_2}{dt} &= \omega + z(\theta_2) * h\delta(t - t_1^j)
 \end{aligned}$$

$I_{\text{syn}}(t)$

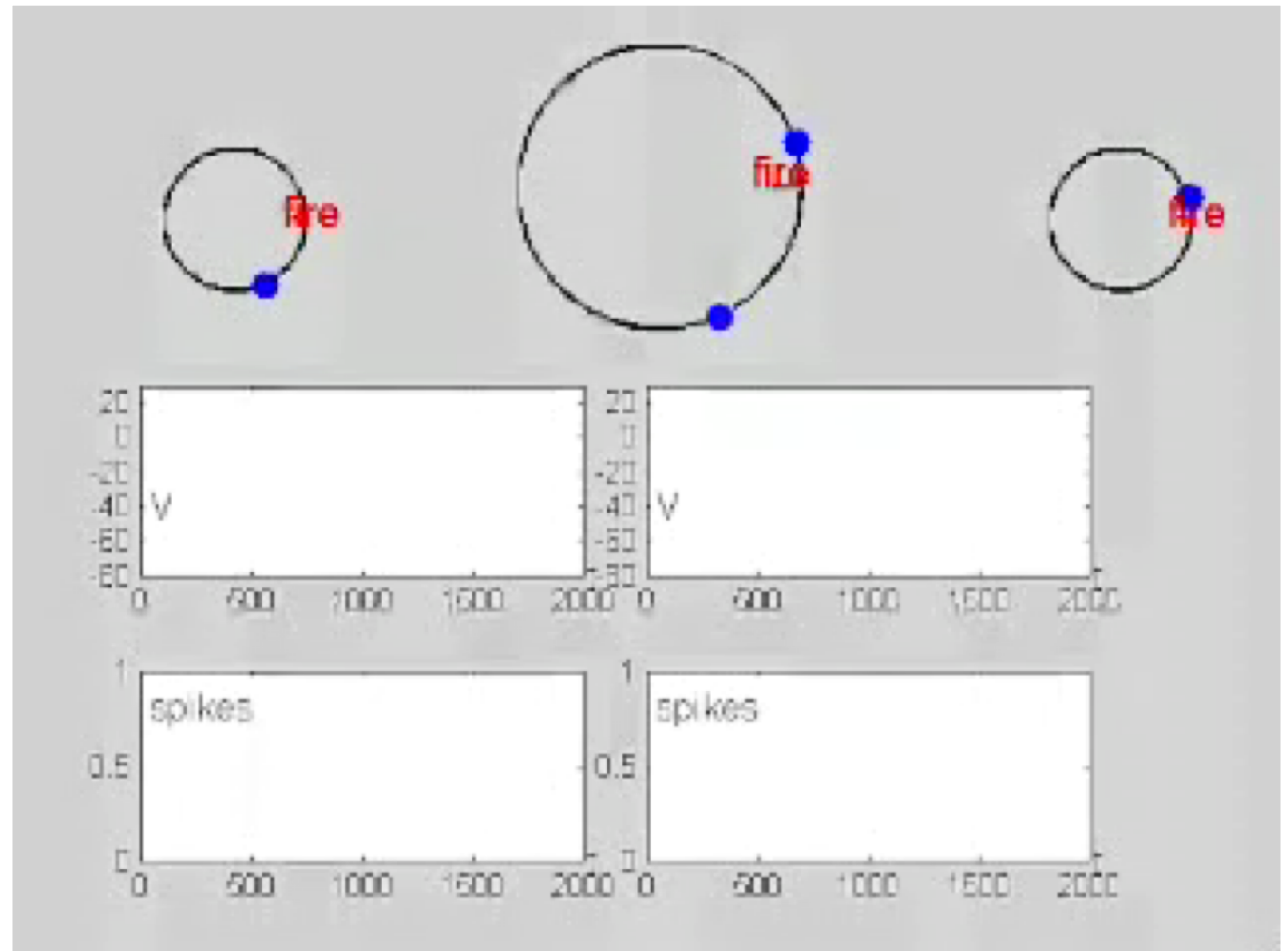
The Hodgkin-Huxley model

PRC
z(θ)



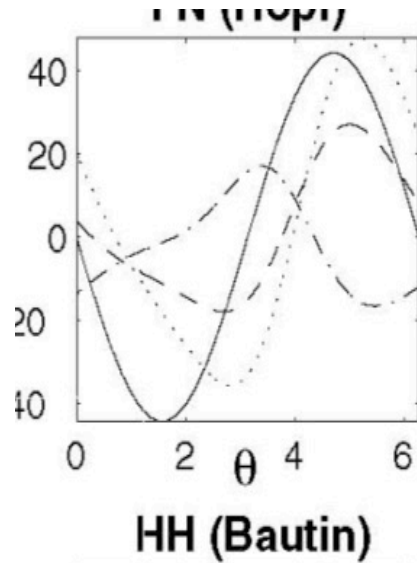
$$\frac{d\theta_1}{dt} = \omega + h * z(\theta_1)\delta(t - t_2^j)$$

$$\frac{d\theta_2}{dt} = \omega + h * z(\theta_2)\delta(t - t_1^j)$$



The Hodgkin-Huxley model

PRC
 $z(\theta)$



$$\frac{d\theta_1}{dt} = \omega + h * z(\theta_1)\delta(t - t_2^j)$$
$$\frac{d\theta_2}{dt} = \omega + h * z(\theta_2)\delta(t - t_1^j)$$

Moral: “Fast” excitatory coupling can synchronize HH neurons ...

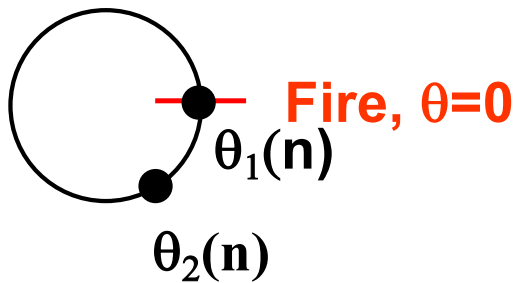
Analyze via Poincare map between firing times of θ_1

$$\frac{d\theta_1}{dt} = \omega + h * z(\theta_1)\delta(t - t_2^j)$$

$$\frac{d\theta_2}{dt} = \omega + h * z(\theta_2)\delta(t - t_1^j)$$

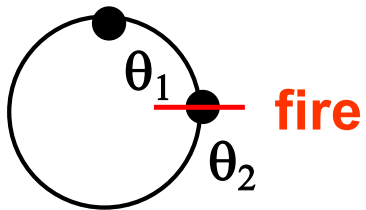
Nancy Kopell, Bard Ermentrout,
-- “weak coupling theory”

Let $\theta_{12} = \theta_1 - \theta_2$

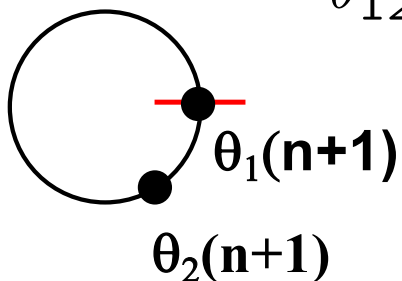


$$\theta_2 \mapsto \theta_2 + h z(-\theta_{12}(n)), \text{ or:}$$

$$\theta_{12}(n) \mapsto \underbrace{\theta_{12}(n) - h z(-\theta_{12}(n))}$$



$$\theta_1 \mapsto \theta_1 + h z[\underbrace{\theta_{12}(n) - h z(-\theta_{12}(n))}]$$



$$\theta_{12}(n+1) = \theta_{12}(n)$$

$$- h z(-\theta_{12}(n))$$

$$+ h z[\theta_{12}(n) - h z(-\theta_{12}(n))]$$

$$= \underline{\theta_{12}(n) + h \{z(\theta_{12}(n)) - z(-\theta_{12}(n))\}} + \mathcal{O}(h^2)$$

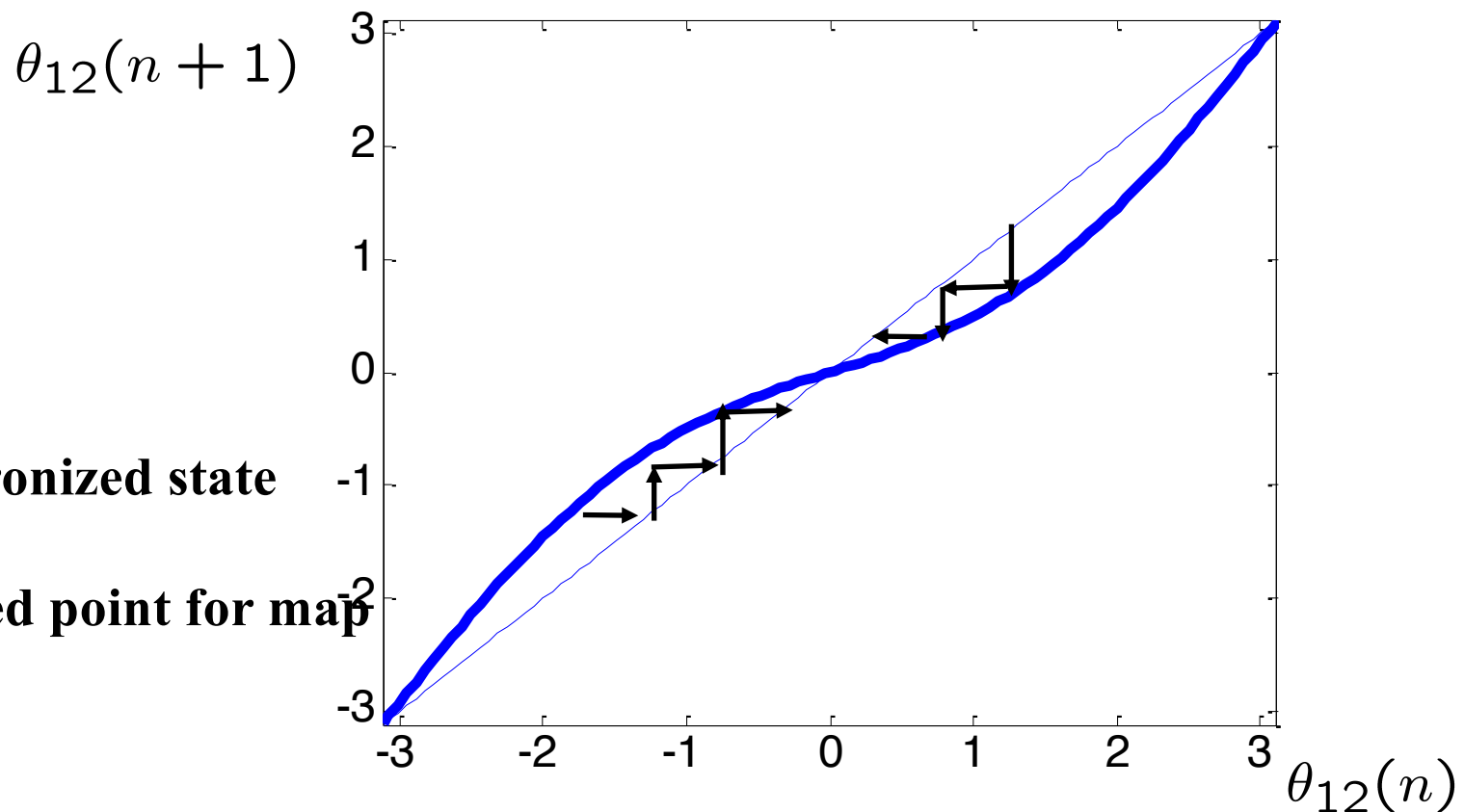
Phase-difference map

$$\theta_{12} = \theta_1 - \theta_2$$

$$\theta_{12}(n+1) = \theta_{12}(n) + h \{z(\theta_{12}(n)) - z(-\theta_{12}(n))\} + \mathcal{O}(h^2)$$

E.g., for HH neuron, $z(\theta) \sim -\sin(\theta)$, so

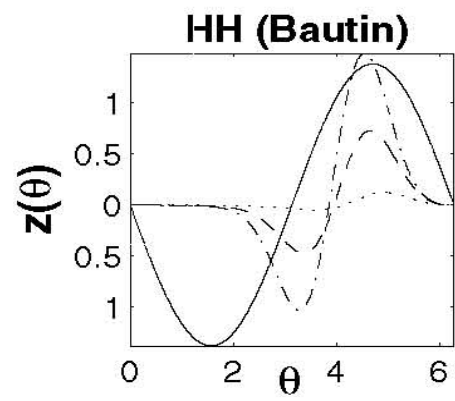
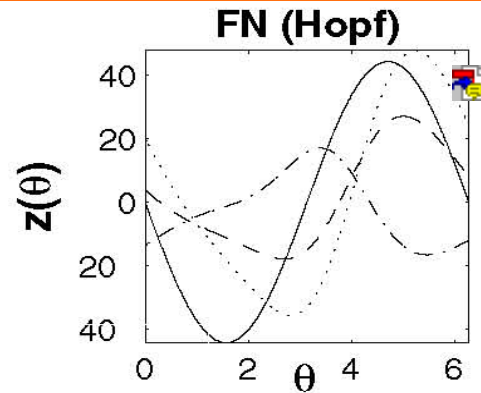
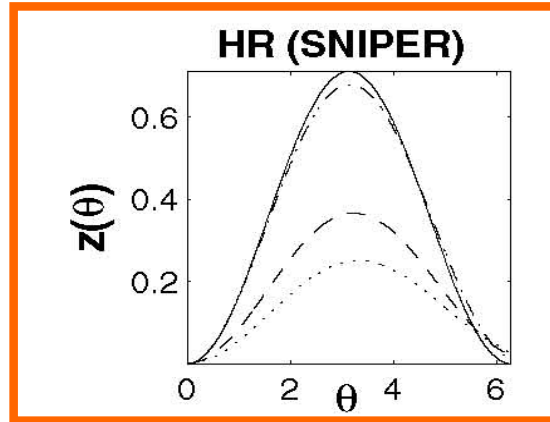
$$\theta_{12}(n+1) \approx \theta_{12}(n) - 2h \sin(\theta_{12})$$



See: synchronized state
 $\theta_{12}=0$
is stable fixed point for map

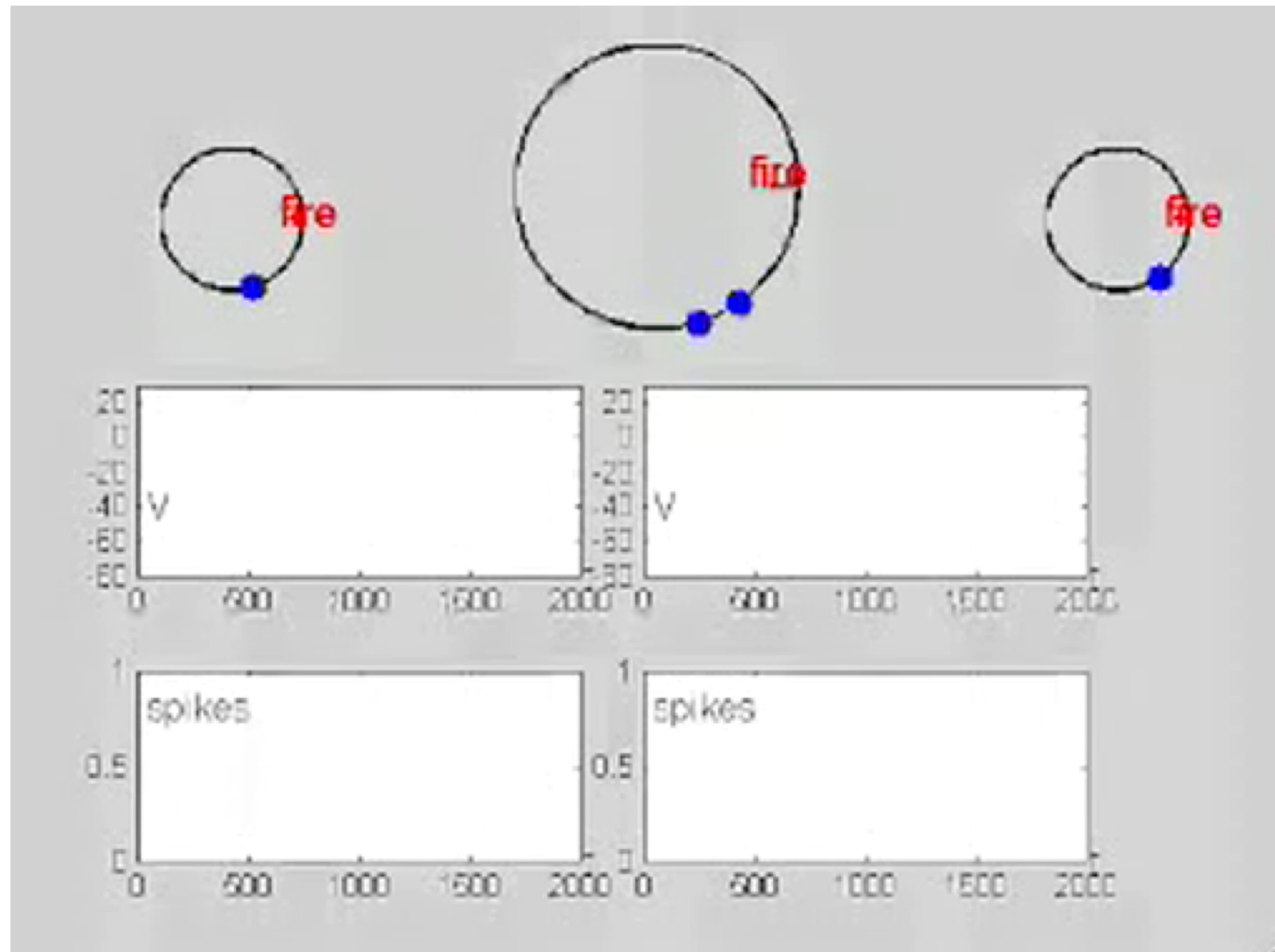
Let's try (as our last example)

Very common
in neural
models ...

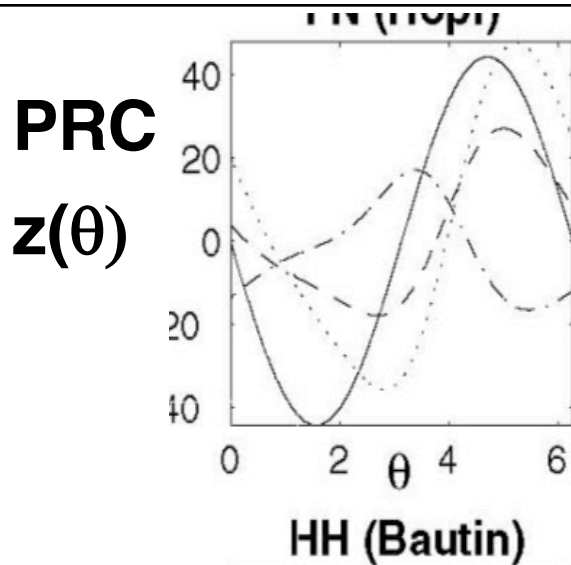


The “Hodgkin Huxley plus A current” model

$$\frac{d\theta_1}{dt} = \omega + h * z(\theta_1)\delta(t - t_2^j)$$
$$\frac{d\theta_2}{dt} = \omega + h * z(\theta_2)\delta(t - t_1^j)$$



The “Hodgkin Huxley plus A current” model



$$\frac{d\theta_1}{dt} = \omega + h * z(\theta_1)\delta(t - t_2^j)$$

$$\frac{d\theta_2}{dt} = \omega + h * z(\theta_2)\delta(t - t_1^j)$$

**Moral: Excitatory coupling actually
DEsynchronizes HH neurons with A currents**

Stable “anti-synchronized” state

However, inhibition *does* synchronize ...

**“When inhibition, not excitation,
synchronizes...” Van Vreeswijk et al 1995**