Nonlinear Dynamics of Neural Firing

BENG/BGGN 260 Neurodynamics

University of California, San Diego

Week 3

- E.M. Izhikevich, *Dynamical Systems in Neuroscience*, MIT Press, 2007, Ch. 1-2, pp. 53-121.
- C. Koch, *Biophysics of Computation*, Oxford Univ. Press, 1999, Ch. 7, pp. 172-192.

One Dimensional Systems

$$C_{m} \frac{dV_{m}}{dt} = I_{ext} - \underline{\overline{g}}_{Na} m_{\infty} (V_{m}) (V_{m} - E_{Na}) - g_{L} (V_{m} - E_{L})$$

$$I_{Na,p} \text{ persistent, fast } Na^{+} (Izh. p.55)$$

$$m_{\infty} = \frac{1}{1 + e^{-(V_{m} - V_{1/2})/k}}$$

$$\Rightarrow \frac{dV}{dt} = F(V)$$

$$F(V)$$

$$I_{m} = \frac{1}{1 + e^{-(V_{m} - V_{1/2})/k}}$$

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Phase Portrait, Equilibria



Bistability and Hysteresis



Figure 3.16: Membrane potential bistability in a cat TC neuron in the presence of ZD7288 (pharmacological blocker of I_b). (Modified from Fig. 6B of Hughes et. al. 1999).



Figure 3.17: Bistability and hysteresis loop as I changes.

Izhikevich, pg. 66

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Saddle-Node Bifurcation



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Two-Dimensional Systems

Morris-Lecar

$$C_{m}\frac{dV_{m}}{dt} = I_{ext} - \bar{g}_{K} \underset{i}{w} (V - E_{K}) - \bar{g}_{Ca} \underset{i}{m} \underset{Na}{m} (V) (V - E_{Ca}) - g_{L} (V - E_{L}) \underset{i}{\overset{ML}{\downarrow}} \underset{Na}{\downarrow} \underset{Na}{\overset{Na}{\downarrow}} \xrightarrow{V} I_{Na,p} + I_{K} \underset{i}{\overset{Na}{\downarrow}} \underset{Na}{\overset{ML}{\downarrow}} \underset{ML}{\overset{ML}{\downarrow}} \underset{Na}{\overset{ML}{\downarrow}} \underset{ML}{\overset{ML}{\downarrow}} \underset{Na}{\overset{ML}{\downarrow}} \underset{ML}{\overset{ML}{\downarrow}} \underset{Na}{\overset{ML}{\downarrow}} \underset{ML}{\overset{ML}{\downarrow}} \underset{Na}{\overset{ML}{\downarrow}} \underset{ML}{\overset{ML}{\downarrow}} \underset{ML}{\overset{ML}{\iota}} \underset{ML}{\overset{ML}{\iota} \underset{ML}{\overset{ML}{\iota}} \underset{$$

$$\frac{dw}{dt}=\frac{w_{\infty}\left(V_{m}\right)-w}{\tau_{w}\left(V_{m}\right)}$$

Dynamic repertoire:

- Stationary Points
- Limit Cycles
- Stable Attractors

No other limits in 2-D (since V and W are bounded)

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$$\Rightarrow \begin{cases} \frac{dV}{dt} &= F(V, W) \\ \frac{dW}{dt} &= G(V, W) \end{cases}$$

Nullclines

$$\begin{cases} \frac{dV}{dt} &= F(V, W) \\ \frac{dW}{dt} &= G(V, W) \end{cases}$$



• Unstable?

Stability

Equilibria:

$$\begin{cases} F(V_0, W_0) &= 0 \\ G(V_0, W_0) &= 0 \end{cases}$$

intersection of nullclines

Linear Analysis:

$$\begin{cases} V = V_0 + \tilde{v} \\ W = W_0 + \tilde{w} \end{cases}$$

$$\Rightarrow \begin{cases} \frac{d\tilde{v}}{dt} = a\tilde{v} + b\tilde{w} \\ \frac{d\tilde{w}}{dt} = c\tilde{v} + d\tilde{w} \end{cases} \text{ with } \begin{vmatrix} a = \frac{\partial F}{\partial V} \\ c = \frac{\partial G}{\partial V} \end{vmatrix} \begin{vmatrix} v_0, W_0 \\ v_0, W_0 \end{vmatrix} = \begin{vmatrix} \frac{\partial G}{\partial W} \\ v_0, W_0 \end{vmatrix}$$
Eigenanalysis:
$$\mathbf{U} = \begin{pmatrix} \tilde{v} \\ \tilde{w} \end{pmatrix} \quad \mathbf{A} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \qquad \text{eigenvectors eigenvalues} \\ \frac{d\mathbf{U}}{dt} = \mathbf{A} \mathbf{U}, \text{ or } \mathbf{U} = c_1 \mathbf{U}_1 e^{\lambda_1 t} + c_2 \mathbf{U}_2 e^{\lambda_2 t} \quad \text{with } \begin{vmatrix} \mathbf{A} \mathbf{U}_1 \\ \mathbf{A} \mathbf{U}_2 \end{vmatrix} = \lambda_1 \mathbf{U}_1 \\ \mathbf{A} \mathbf{U}_2 = \lambda_2 \mathbf{U}_2 \end{cases}$$

$$\begin{cases} Re \lambda_1 < 0 \text{ and } Re \lambda_2 < 0 \Rightarrow \text{ stable fixed point} \\ Re \lambda_1 > 0 \text{ or } Re \lambda_2 > 0 \Rightarrow \text{ unstable}, \end{cases}$$

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Stability (Continued)

Eigenanalysis:

$$\mathbf{A} \mathbf{U} = \lambda \mathbf{U}$$
$$det (\mathbf{A} - \lambda \mathbf{I}) = 0$$
$$\begin{vmatrix} \mathbf{a} - \lambda & \mathbf{b} \\ \mathbf{c} & d - \lambda \end{vmatrix} = 0$$
$$\lambda^2 - \tau \lambda + \Delta = 0 \quad \text{with} \quad \begin{cases} \tau &= \mathbf{a} + d &: \text{ Trace } (\mathbf{A}) \\ \Delta &= \mathbf{a} d - b c &: \text{ det } (\mathbf{A}) \end{cases}$$
$$\lambda_{1,2} = \frac{\tau \pm \sqrt{\tau^2 - 4\Delta}}{2}$$

 $\Rightarrow \begin{cases} \tau^2 &> 4\Delta: \text{ real, distinct eigenvalues} \\ \tau^2 &= 4\Delta: \text{ real, repeated eigenvalues} \\ \tau^2 &< 4\Delta: \text{ complex conjugate eigenvalues} \end{cases}$

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Stability of 2-D Equilibria



Figure 4.15: Classification of equilibria according to the trace (τ) and the determinant (Δ) of the Jacobian matrix **A**. The shaded region corresponds to stable equilibria.

Izhikevich, pg. 104

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Morris-Lecar and $I_{Na,p} + I_K$

Persistent sodium and potassium $(I_{Na,p} + I_K)$ is equivalent to Morris-Lecar (n = w).



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Figure 4.4: Nullclines of the $I_{Na,p} + I_{K}$ -model (4.1, 4.2) with low-threshold K^+ current in Fig. 4.1b. (The vector field is slightly distorted for the sake of clarity of illustration).

Izhikevich, pg. 93

Graded Action Potentials



Figure 4.7: Failure to generate all-or-none action potentials in the $I_{Na,p} + I_K$ -model.

Izhikevich, pg. 95

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Variable Threshold



Figure 4.8: Failure to have a fixed value of threshold voltage in the $I_{Na,p} + I_K$ -model.

Izhikevich, pg. 96

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Stable Limit Cycle



Figure 4.10: Stable limit cycle in the $I_{Na,p} + I_K$ -model (4.1, 4.2) with low-threshold K^+ current and I = 40.

Izhikevich, pg. 97 Week 3 16 / 16

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