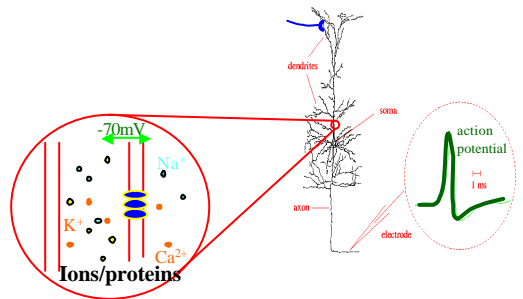


Lecture 3: Two-dimensional neuron models

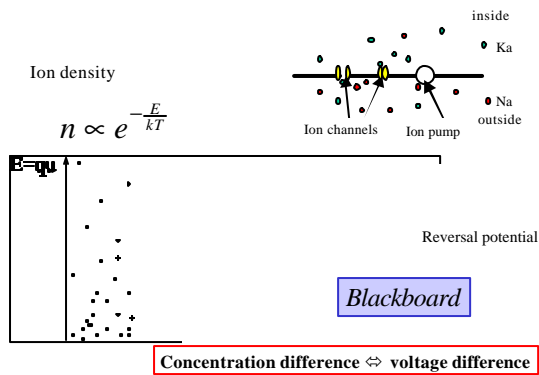
BOOK: Spiking Neuron Models,
W. Gerstner and W. Kistler
Cambridge University Press, 2002

Chapter 3

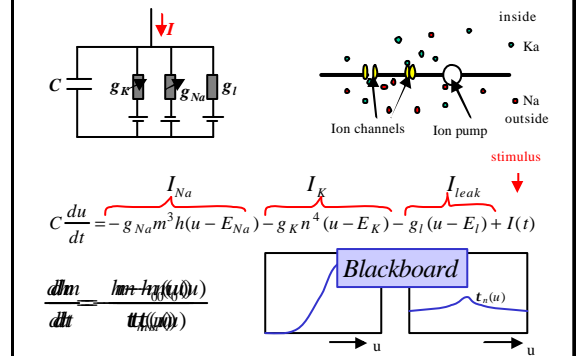
Biophysics of neurons



Hodgkin-Huxley Model



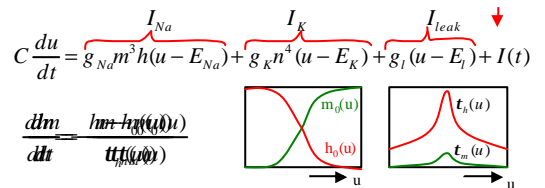
Hodgkin-Huxley Model



Two-dimensional neuron models

-Reduction of Hodgkin -Huxley
-step 1: separation of time scales

Reduction of Hodgkin -Huxley Model



1) dynamics of m is fast $\rightarrow m(t) = m_0(u(t))$

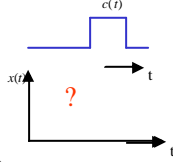
Exercise: separation of time scales

stimulus

$$C \frac{du}{dt} = \overbrace{g_{Na} m^3 h (u - E_{Na})}^{I_{Na}} + \overbrace{g_K n^4 (u - E_K)}^{I_K} + \overbrace{g_l (u - E_l)}^{I_{leak}} + I(t)$$

$$\frac{dm}{dt} = -\frac{m - m_0(u)}{\tau_m(u)}$$

$$\frac{dx}{dt} = -\frac{x - c(t)}{\tau}$$



$$\frac{dm}{dt} = -\frac{m - m_0(u)}{\tau_m}$$

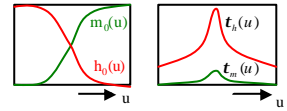
$$\frac{du}{dt} = f(u) - m$$

Reduction of Hodgkin-Huxley Model

stimulus

$$C \frac{du}{dt} = \overbrace{g_{Na} m^3 h (u - E_{Na})}^{I_{Na}} + \overbrace{g_K n^4 (u - E_K)}^{I_K} + \overbrace{g_l (u - E_l)}^{I_{leak}} + I(t)$$

$$\frac{dm}{dt} = \frac{hm - h_{\infty}(u)}{\tau_m(u)}$$



- 1) dynamics of m is fast $\longrightarrow m(t) = m_0(u(t))$
- 2) dynamics of h and n is similar $\longrightarrow 1 - h(t) = a n(t)$

Blackboard

Reduction of Hodgkin-Huxley Model

stimulus

$$C \frac{du}{dt} = \overbrace{g_{Na} m_0(u)^3 (1-w) (u - E_{Na})}^{I_{Na}} + \overbrace{g_K \left(\frac{w}{a}\right)^4 (u - E_K)}^{I_K} + \overbrace{g_l (u - E_l)}^{I_{leak}} + I(t)$$

$$\frac{dw}{dt} = -\frac{w - w_0(u)}{\tau_w(u)}$$

- 1) dynamics of m is fast $\longrightarrow m(t) = m_0(u(t))$
- 2) dynamics of h and n is similar $\longrightarrow 1 - h(t) = a n(t)$

Blackboard

Two-dimensional neuron models

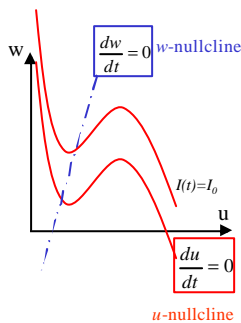
- Reduction of Hodgkin-Huxley
- step 1: separation of time scales
- step 2: effective variable w
- Phase space analysis

Reduction of Hodgkin-Huxley Model: 2 dimensional Neuron Models

stimulus

$$\tau \frac{du}{dt} = F(u, w) + I(t)$$

$$\tau_w \frac{dw}{dt} = G(u, w)$$

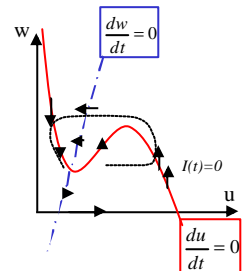


FitzHugh-Nagumo Model 2-dimensional Neuron Models

stimulus

$$\tau \frac{du}{dt} = F(u, w) + I(t)$$

$$\tau_w \frac{dw}{dt} = G(u, w)$$

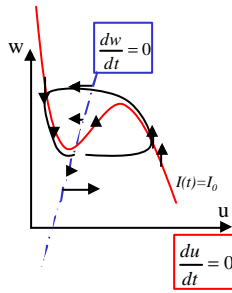


FitzHugh Nagumo Model – limit cycle

$$\begin{aligned} \tau \frac{du}{dt} &= F(u, w) + I(t) \\ \tau_w \frac{dw}{dt} &= G(u, w) \end{aligned}$$

- unstable fixed point
- closed boundary with arrows pointing inside

→ limit cycle



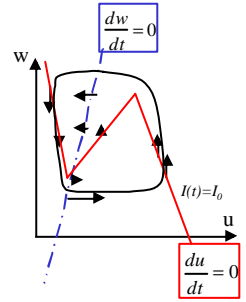
limit cycle

Stability of fixed points -Exercise

$$\begin{aligned} \tau \frac{du}{dt} &= au + bw + I_0 \\ \tau_w \frac{dw}{dt} &= c u + d w \end{aligned}$$

Exercise:

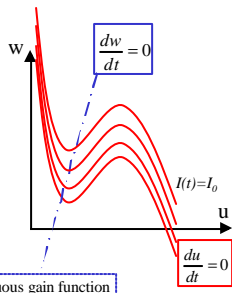
Calculate stability for parameters a, b, c, d



type II Model – constant input

$$\begin{aligned} \tau \frac{du}{dt} &= F(u, w) + I(t) \\ \tau_w \frac{dw}{dt} &= G(u, w) \end{aligned}$$

Stability lost → oscillation with finite frequency



Discontinuous gain function

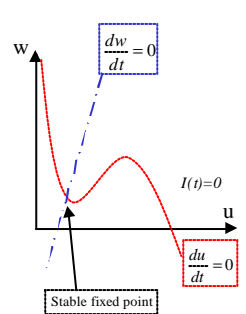
type II Model - pulse input

$$\begin{aligned} \tau \frac{du}{dt} &= F(u, w) + I(t) \\ \tau_w \frac{dw}{dt} &= G(u, w) \end{aligned}$$

pulse input



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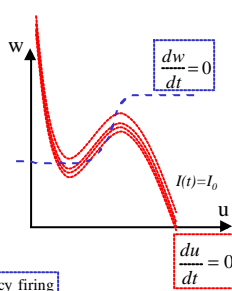


Stable fixed point

type I Model – constant input

$$\begin{aligned} \tau \frac{du}{dt} &= F(u, w) + I(t) \\ \tau_w \frac{dw}{dt} &= G(u, w) \end{aligned}$$

Low-frequency firing



type I Model-pulse input

$$\begin{aligned} \tau \frac{du}{dt} &= F(u, w) + I(t) \\ \tau_w \frac{dw}{dt} &= G(u, w) \end{aligned}$$

pulse input



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