

FitzHugh-Nagumo-ing the Hodgkin-Huxley model with RNNs

Reproducing with a lower dimensional model

CAMP 2022

Team (still deciding)

The Hodgkin-Huxley model, and friends

$$I = C_m \frac{dV_m}{dt} + \bar{g}_K n^4 (V_m - V_K) + \bar{g}_{Na} m^3 h (V_m - V_{Na}) + \bar{g}_l (V_m - V_l),$$

$$\frac{dn}{dt} = \alpha_n(V_m)(1 - n) - \beta_n(V_m)n$$

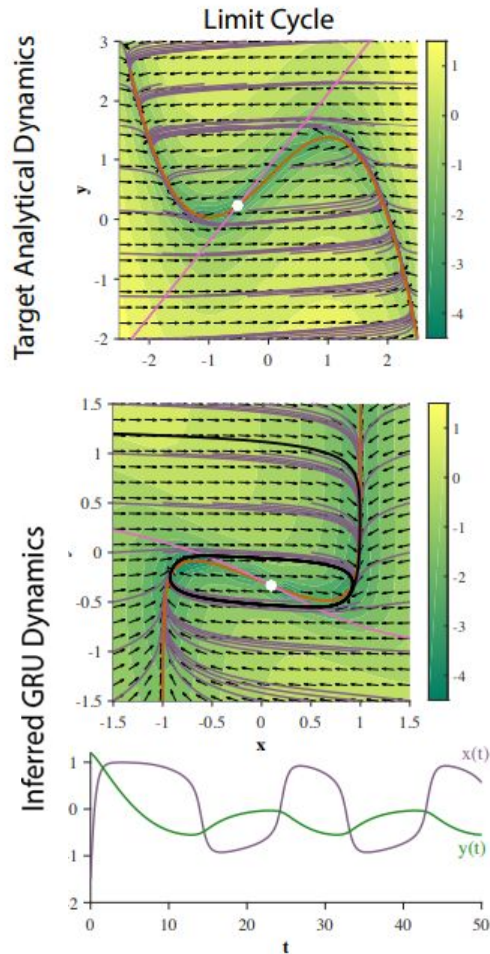
$$\frac{dm}{dt} = \alpha_m(V_m)(1 - m) - \beta_m(V_m)m$$

$$\frac{dh}{dt} = \alpha_h(V_m)(1 - h) - \beta_h(V_m)h$$

Motivation

- Simplification of HH model from 4D
- Exploring the power of RNNs to analyse dynamical systems
- This article:

Jordan et al., **Gated recurrent units viewed through the lens of continuous time dynamical systems.**
Front. Comput. Neurosci, 2021.



Bonhoeffer-van der Pol (BVP) model

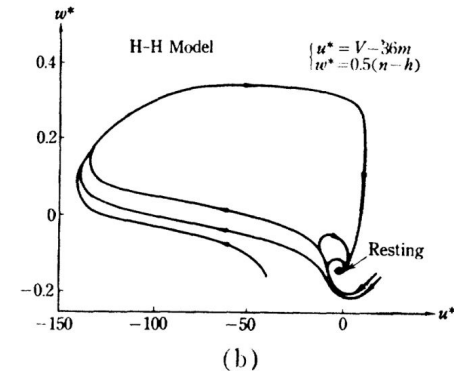
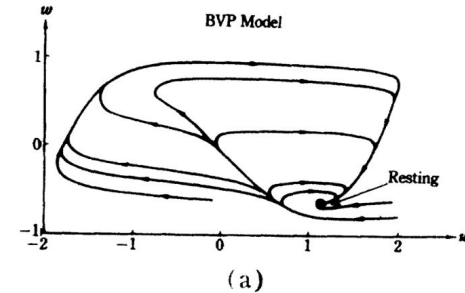
FitzHugh, J. Gen. Physiol, 1960; Thresholds and Plateaus in the Hodgkin-Huxley Nerve Equations.

$$\begin{cases} J = \frac{1}{c} \frac{du}{dt} - w - \left(u - \frac{u^3}{3} \right), \\ c \frac{dw}{dt} + bw = a - u, \end{cases}$$

where $1 > b > 0$, $c^2 > b$, $1 > a > 1 - \frac{2}{3} * b$

Correspondence of BVP and HH
model parameters:

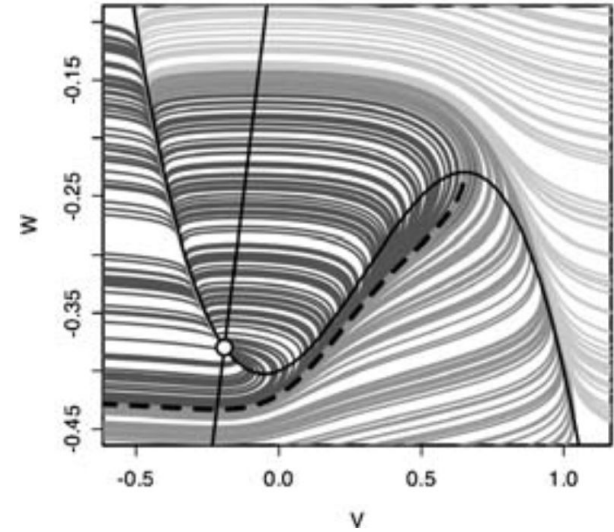
$$u \rightarrow V, m; w \rightarrow h, n; J \rightarrow I$$



J. Nagumo, S. Arimoto, S. Yoshizawa (1962)

FitzHugh-Nagumo model

$$\begin{aligned}\psi \, dV/dt &= V(1 - V)(a + V) - W + c + I, \\ dW/dt &= V - bW,\end{aligned}$$



$$a = 0.1, b = 0.5, c = -0.4, \psi = 0.015, I = 0$$

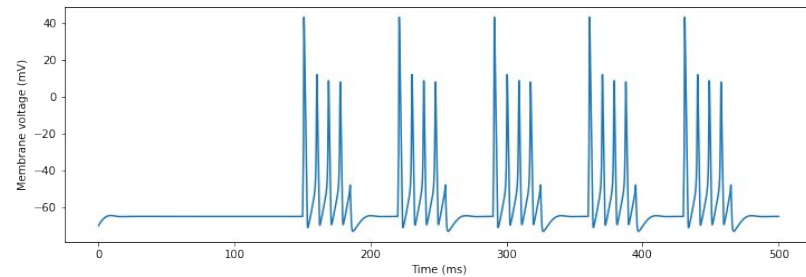
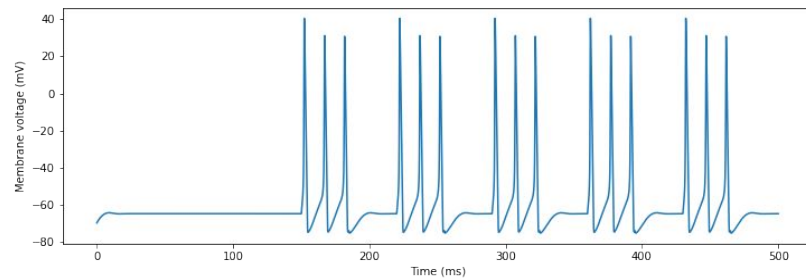
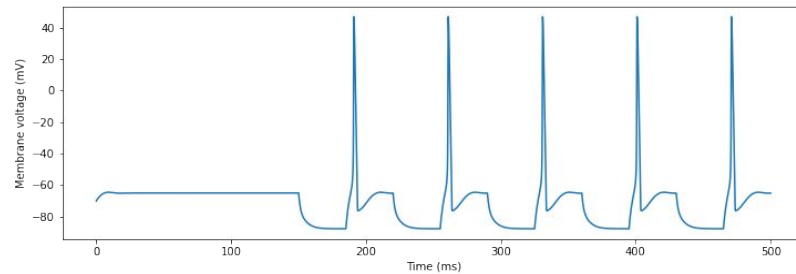
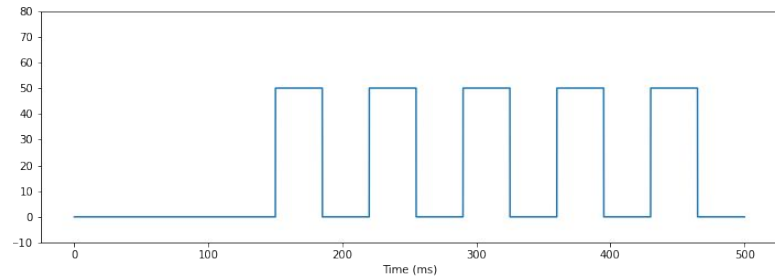
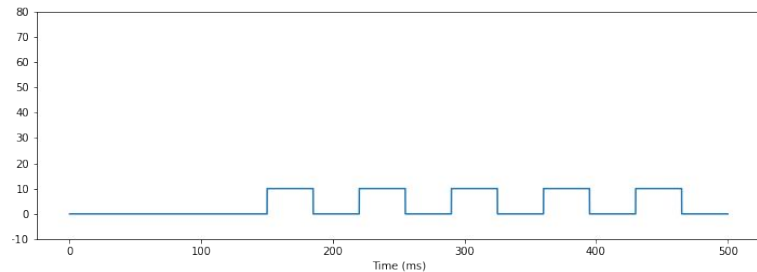
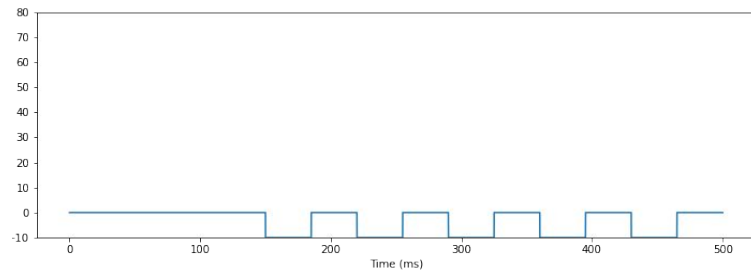
Figure from S. Hong, B. Agüera y Arcas, and A. Fairhall (2007)

What solutions did we expect?

- 2-neuron RNN which can accurately predict HH model from current values
- Try to replicate spikes, rebound spikes, f-I curves resembling the HH model
- Phase portrait depicting dynamics of the HH system
- Comparison of observed spikes using latency, spike width, f-I curve

Protocol

Injected current



Implementation and Architecture

Hidden state dynamics of the RNN:

$$\mathbf{dh}/dt = f(\mathbf{h}(t), \mathbf{x}(t))$$

where, \mathbf{h} = hidden state

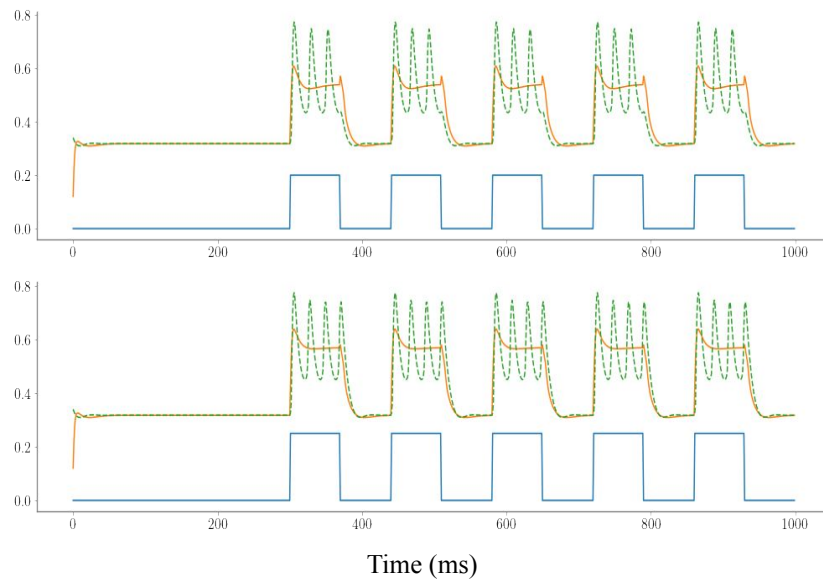
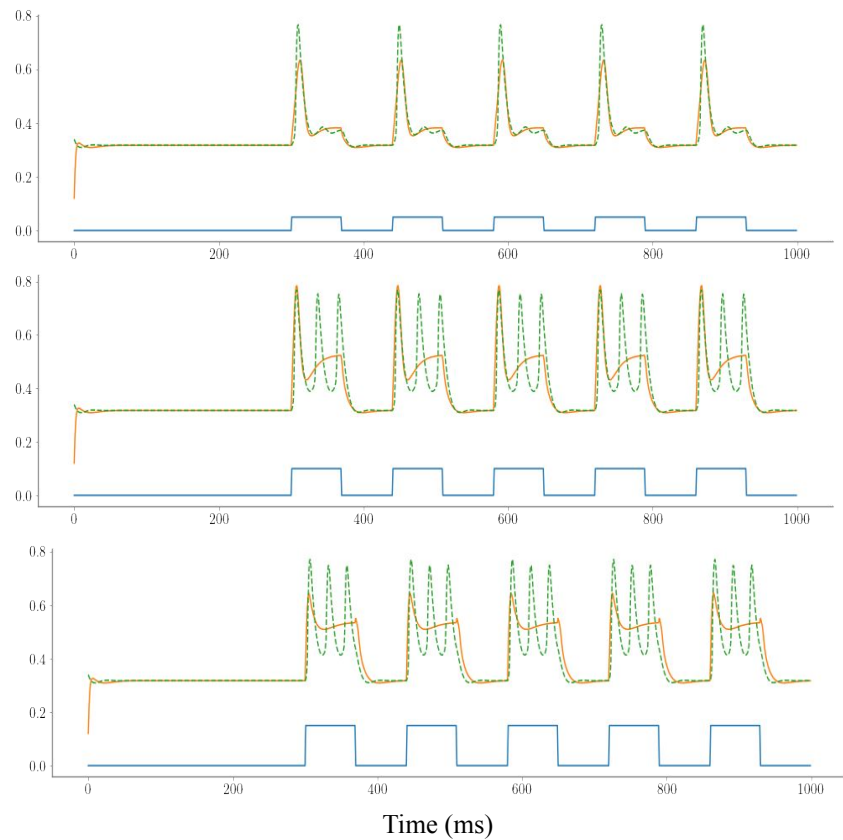
$\mathbf{x}(t)$ = time dependent input

$$\mathbf{z}(t) = \sigma(\mathbf{U}_z \mathbf{h}(t) + \mathbf{b}_z)$$

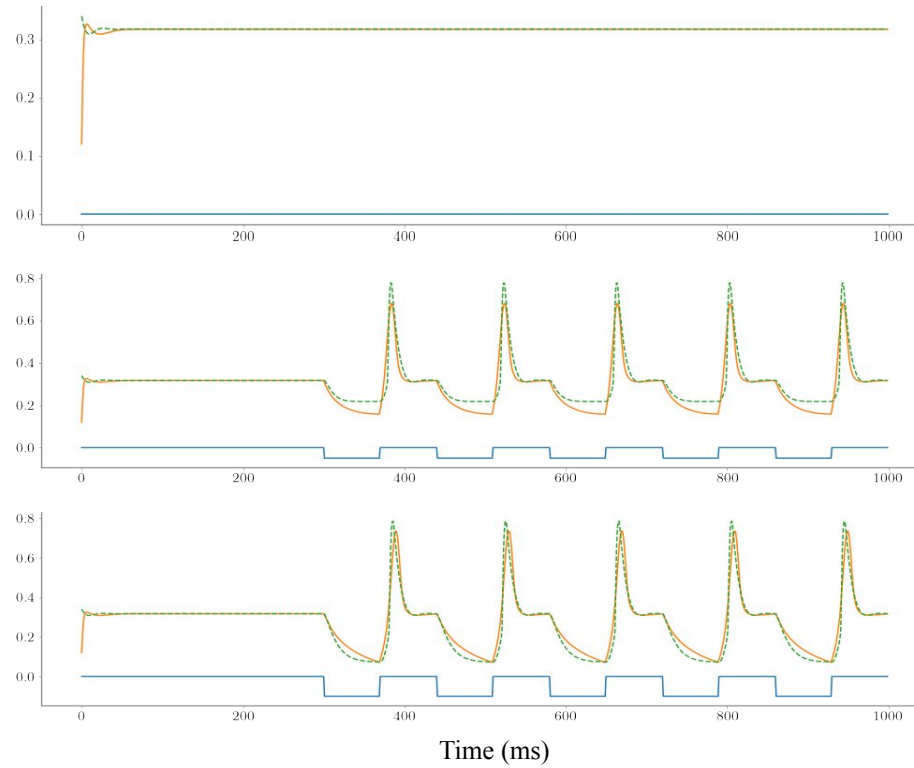
$$\mathbf{r}(t) = \sigma(\mathbf{U}_r \mathbf{h}(t) + \mathbf{b}_r)$$

$$\dot{\mathbf{h}} = (1 - \mathbf{z}(t)) \odot (\tanh(\mathbf{U}_h(\mathbf{r}(t) \odot \mathbf{h}(t)) + \mathbf{b}_h) - \mathbf{h}(t))$$

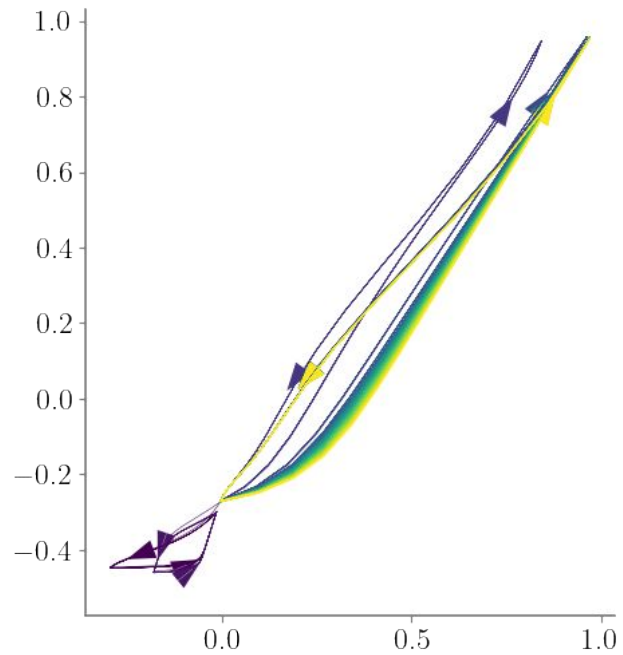
Results



Results



Phase Portrait



Thank you!