Keio University





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# Discovery of a structural model for neuronal activation

Hideyasu SHIMADZU Ritei SHIBATA Toshinobu SHIMOI Kotaro OKA (Keio University, JAPAN)

### Introduction

- Joint work with neuroscientists in Keio University.
- To construct suitable data driven models of neuronal activation which
- incorporate knowledge in neuroscience;
- develop the joint research further.



# Outline

- Introduction
- Data
- Biological backgrounds
- Model
- Results
- Conclusion



#### Data

Earthworm's membrane potential is measured

- by intra-cellar recording;
- for 40 sec with 0.05 msec time resolution (879000 observations).



Build a model for a cluster.

#### Cluster defined:

- from 500 msec ahead the first spike;
- to 500 msec behind the last spike.



Potential adjusted as both ends take zero.

Comments:

Modelling challenges include:

• the sudden stop of spikes;



#### There have been many attempts that model:

#### Membrane potential

- Conductance based models: Hodgkin & Huxley (1952), Rose & Hindmarsh (1989), Wilson (1999) etc.
- Integrated fire models: Izhikevich (2003, 2004) etc.

#### Spike occurrence time

- Point process models:
  - Cox & Isham (1980), Kass & Ventura (2001), Ventura et al. (2002), Kass et al. (2005) etc.

#### Comments:

No bridges between these two approaches.



# Biological backgrounds (1)



Hodgkin-Huxley model (Hodgkin & Huxley, 1952) assumes that

- the membrane behaves like electric circuits;
- channels switch depending on membrane potential.

$$C \frac{dV(t)}{dt} = I(t) - \sum_{i} g_i(t, V(t)) \left(V(t) - V_i\right)$$

V(t): Membrane potential [V];

I(t): Synaptic current [A];

C: Capacitance [F];

 $g_i(t, V(t))$ : Conductance [S];

 $V_i$ : Battery [V].

# Biological backgrounds (2)

- H-H model shows **two phase** in a spike:
  - I: Firing phase;
  - II: Refractory phase.

#### Comments:

Two phases are **NOT** enough for the description.



# Biological backgrounds (3)



Encyclopedia of Life Sciences Published by John Wiley & Sons, Ltd

#### Two types of inputs:

- Electrical synapse (No delay);
- Chemical synapse (delay, degitise).

#### Membrane potential accumulated as $V(t) \sim L(t) + S(t)$

### A simple system



Input signal: L(t)

Cumulative potential changes caused through chemical synapse is given by  $S(t) = \sum_{i=1}^{N} s(t - T_j)$  Spike occurrence times

$$\{T_j; j=1,2,\ldots,N\}$$

Intensity function is given by

$$\lambda\left(t\right) = \begin{cases} \kappa_1 \left(\frac{dL\left(t\right)}{dt}\right)_+ & t \le \tau_1 \\ \\ \kappa_2 \left(\frac{dL\left(t\right)}{dt}\right)_+ & \tau_1 < t \end{cases}$$

$$\begin{array}{c} \begin{array}{c} \begin{array}{c} C \frac{dV\left(t\right)}{dt} = I\left(t\right) - g\left(t,V\right)\left(V\left(t\right) - E\right) \\ \hline \end{array} \\ \left\{ \begin{array}{c} \left\{g\left(t,V\right) = \text{constant} \\ I\left(t\right) = 0 \end{array}\right\} \\ V\left(t\right) = \alpha e^{\beta t} + \gamma, \\ \left(\alpha = V\left(0\right) - \gamma, \beta = -g/C, \gamma = E\right) \end{array} \\ \left(\alpha = V\left(0\right) - \gamma, \beta = -g/C, \gamma = E\right) \end{array} \\ \hline \end{array} \\ \begin{array}{c} \begin{array}{c} \left\{ \begin{array}{c} \alpha_{0}e^{\beta_{0}t} + \gamma_{0}, & t < \tau_{1} \\ \alpha_{1}e^{\beta_{1}\left(t-\tau_{1}\right)} + \gamma_{1}, & \tau_{1} \leq t < \tau_{2} \end{array}\right\} \\ \left(\alpha_{1}e^{\beta_{1}\left(t-\tau_{1}\right)} + \gamma_{1}, & \tau_{1} \leq t < \tau_{2} \end{array}\right) \\ \hline \end{array} \\ \left( \begin{array}{c} \left\{ \begin{array}{c} \alpha_{0}e^{\beta_{0}t} + \gamma_{0}, & t < \tau_{1} \\ \alpha_{1}e^{\beta_{1}\left(t-\tau_{1}\right)} + \gamma_{1}, & \tau_{1} \leq t < \tau_{2} \end{array}\right\} \\ \left( \begin{array}{c} \left( 0 \right) \\ \end{array} \\ \left( \begin{array}{c} \mathbf{Pre-firing phase} \\ \alpha_{2}e^{\beta_{2}\left(t-\tau_{2}\right)} + \gamma_{2}, & \tau_{2} \leq t \end{array}\right) \\ \hline \end{array} \\ \left( \begin{array}{c} \left( \mathbf{II} \right) \\ \end{array} \\ \left( \begin{array}{c} \mathbf{Refractory phase} \\ \mathbf{Refractory phase} \end{array}\right) \\ \end{array} \\ \end{array} \\ \end{array} \\ \end{array} \\ \end{array}$$



# Model of the input V(t) = L(t) + S(t) + U(t)

Assume the three phase model for the input as

$$L(t) = \begin{cases} a_0 e^{b_0 t} + w_0, & -\infty < t < t^* \\ a_1 e^{b_1 (t - t^*)} + w_1, & t^* \le t < t^{**} \\ a_2 e^{b_2 (t - t^{**})} + w_2, & t^{**} \le t < \infty \end{cases}$$

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# Model of a spike V(t) = L(t) + S(t) + U(t)

$$S(t) = \sum_{j=1}^{N} s(t - T_j)$$

$$s(t) = \begin{cases} \alpha_1 e^{\beta_1 t} + \gamma_1, & T_j < t < \tau_1 \\\\ \alpha_2 e^{\beta_2 (t - \tau_1)} + \gamma_2, & \tau_1 \le t < \tau_2 \\\\ \alpha_3 e^{\beta_3 (t - \tau_2)} + \gamma_3, & \tau_2 \le t < \infty \end{cases}$$













Model checking (1)12



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#### Model checking (2)



$$\lambda(t) = \begin{cases} \kappa_1 \left(\frac{dL(t)}{dt}\right)_+ & t \le \tau_1 \\ \\ \kappa_2 \left(\frac{dL(t)}{dt}\right)_+ & \tau_1 < t \end{cases}$$



$$\Lambda_{j} = \int_{0}^{T_{j}} \hat{\lambda}\left(u\right) du$$

$$Z_j = \Lambda_{j+1} - \Lambda_j$$







# Conclusion

- Contributions from
- Neuroscience
  - H-H model
  - Two types of inputs

#### Data Science

- Pre-firing phase
- Relation between the input and intensity function



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Thank you for kind attention. Comments and suggestions welcomed!

Hideyasu SHIMADZU shimadzu@stat.math.keio.ac.jp http://www.stat.math.keio.ac.jp/~shimadzu/

