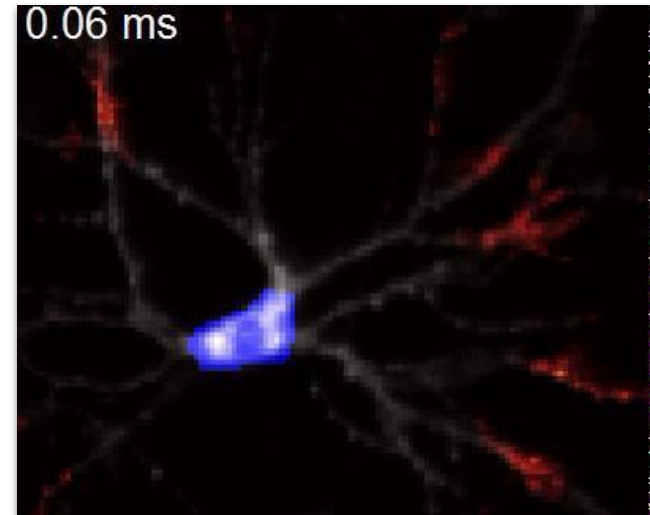
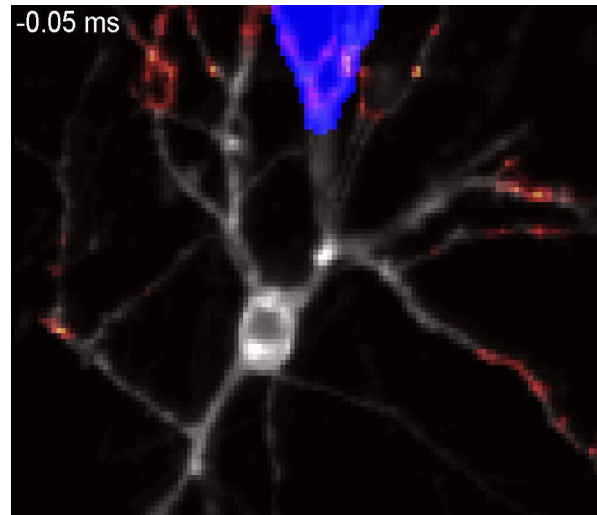
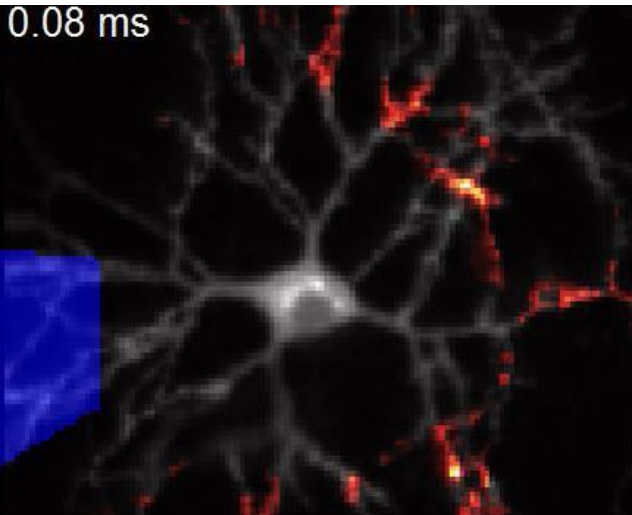


# Scalable Bayesian inference of dendritic voltage via spatiotemporal recurrent state space models

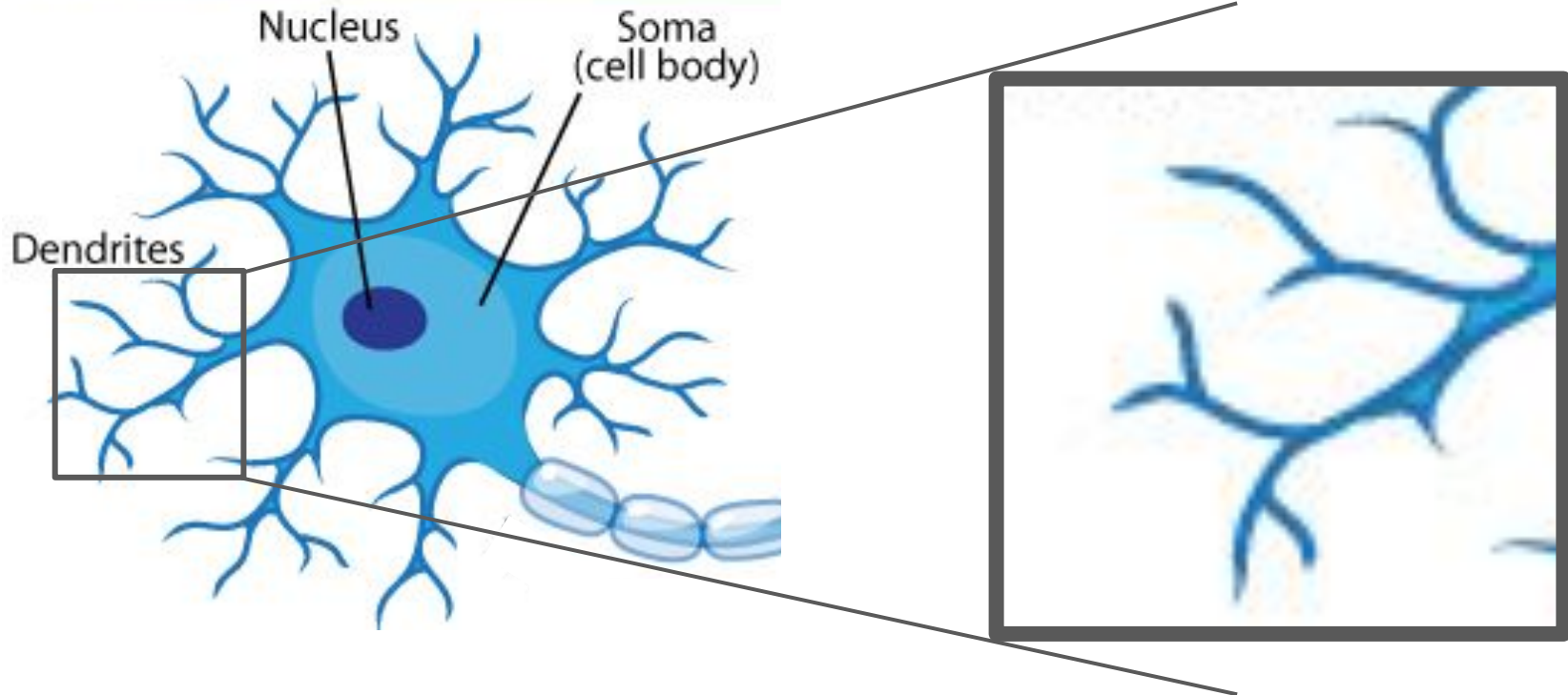
Ruoxi Sun\*, Scott Linderman\*, Ian Kinsella, Liam Paninski

Columbia University  
NeurIPS 2019

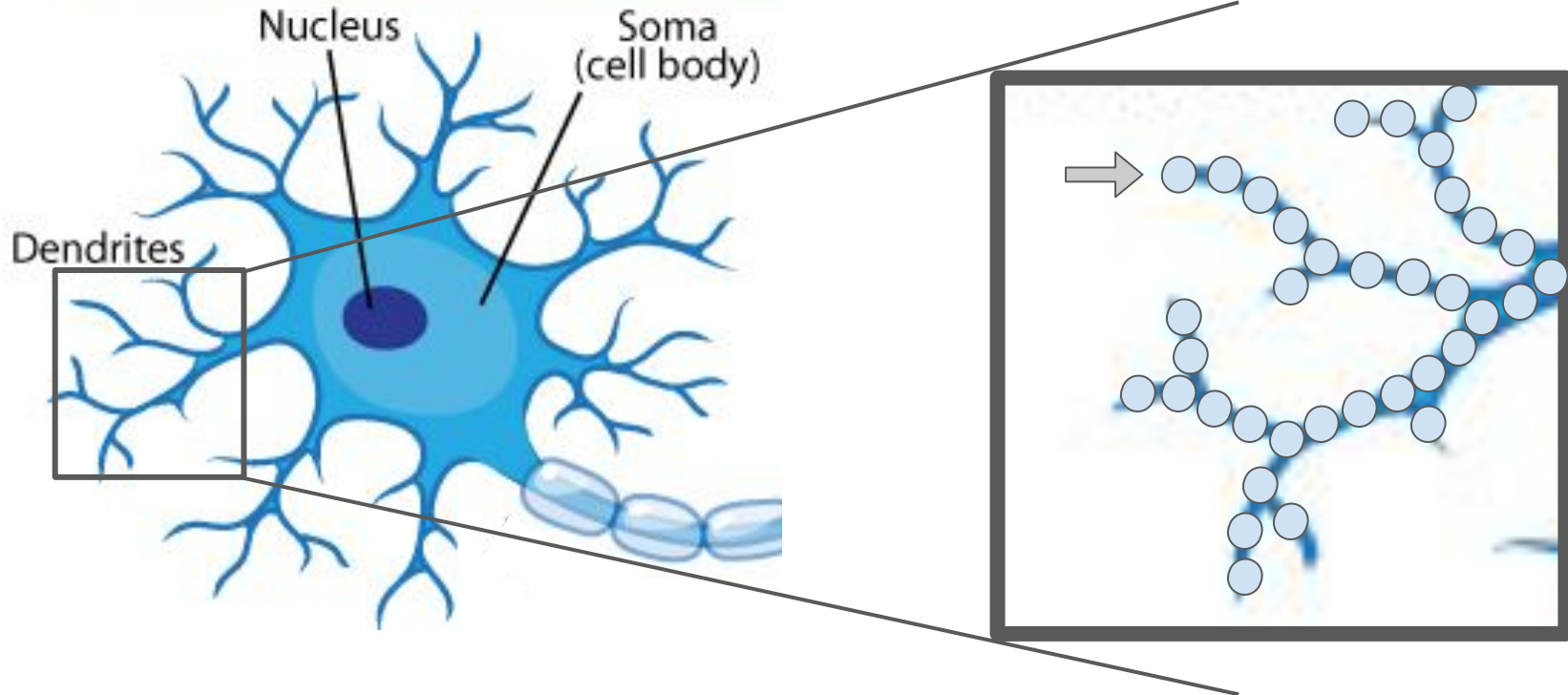
# Dendritic voltage imaging



# Multiple Compartment models

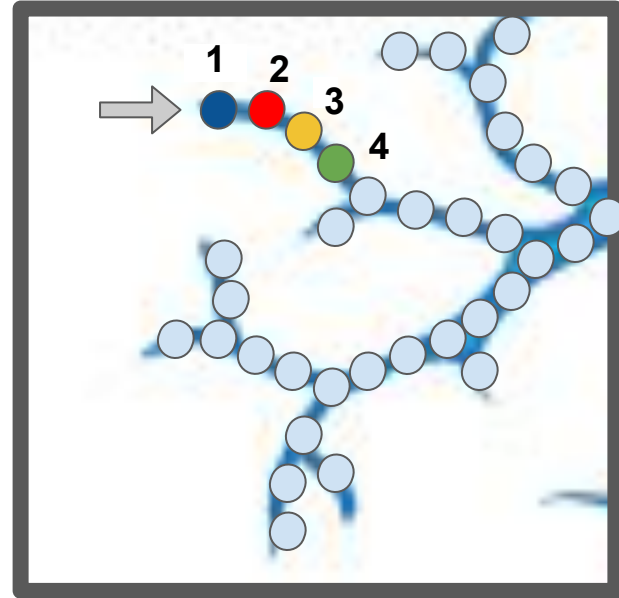
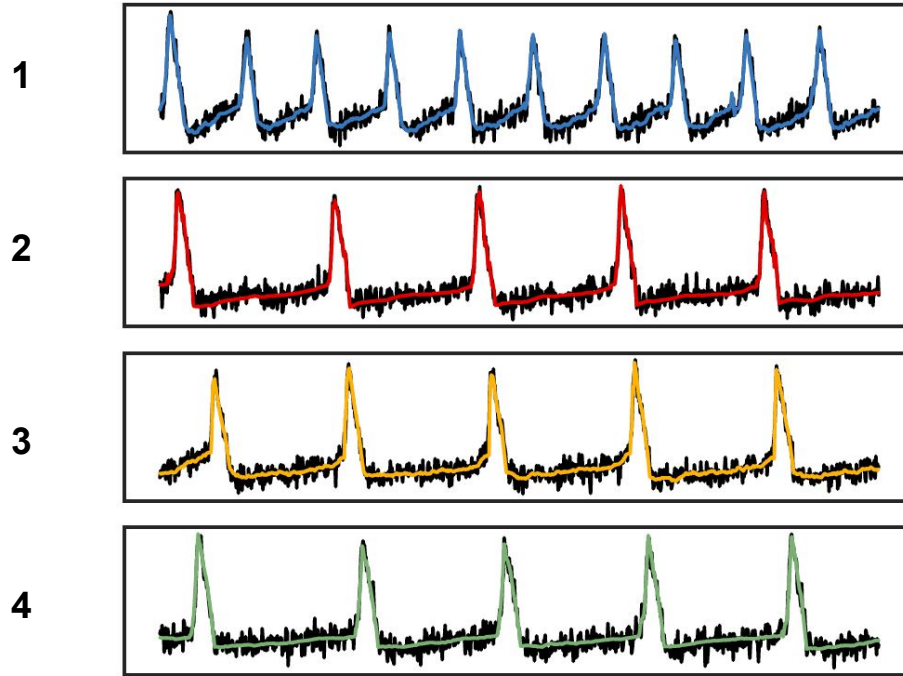


# Multiple Compartment models



# Multiple Compartment models

Compartment

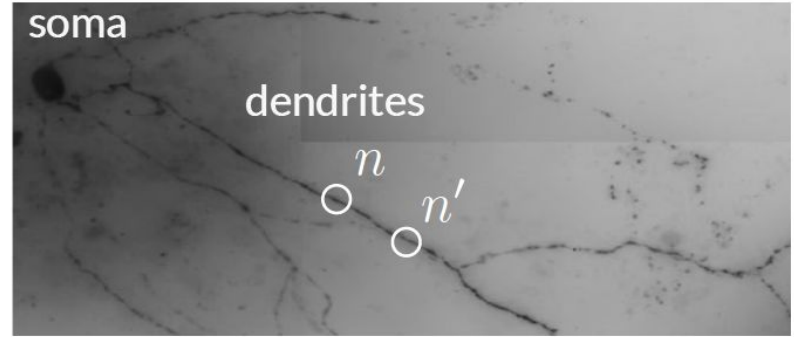


# Biophysics

## Cable equation theory

$$I = \frac{\Delta V}{R} = g\Delta V$$

**g**: conductance; **I**: current; **R**: resistance; **V**: voltage; **C**: capacitance

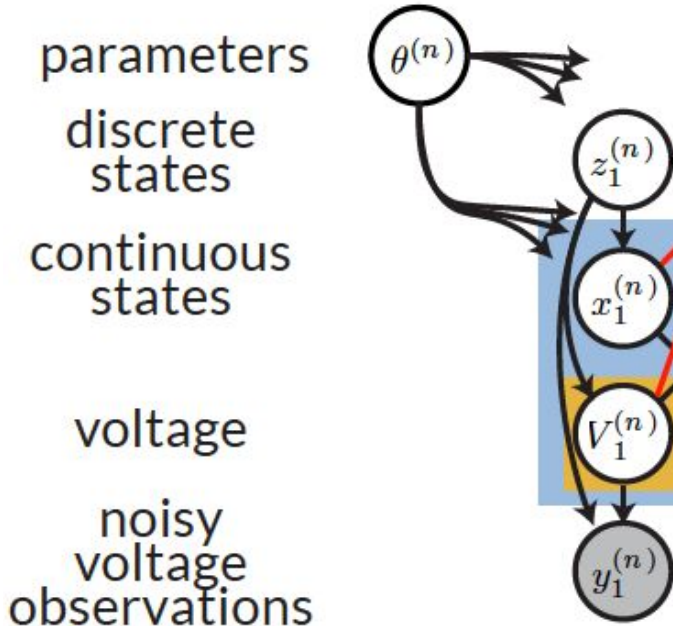


## Compartment n:

$$V_{t+\Delta t}^{(n)} \approx V_t^{(n)} + \frac{\Delta t}{C_n} \left[ \sum_j I_t^{(n,j)} + \sum_{n'=1}^N g_{nn'} \cdot (V_t^{(n')} - V_t^{(n)}) \right]$$

# **Biophysics to Statistics Model**

# Model Single Compartment Dynamics one time step



**theta:** parameters

**Z:** discrete latent variable

**X:** continuous latent variable  
(cycle parameters)

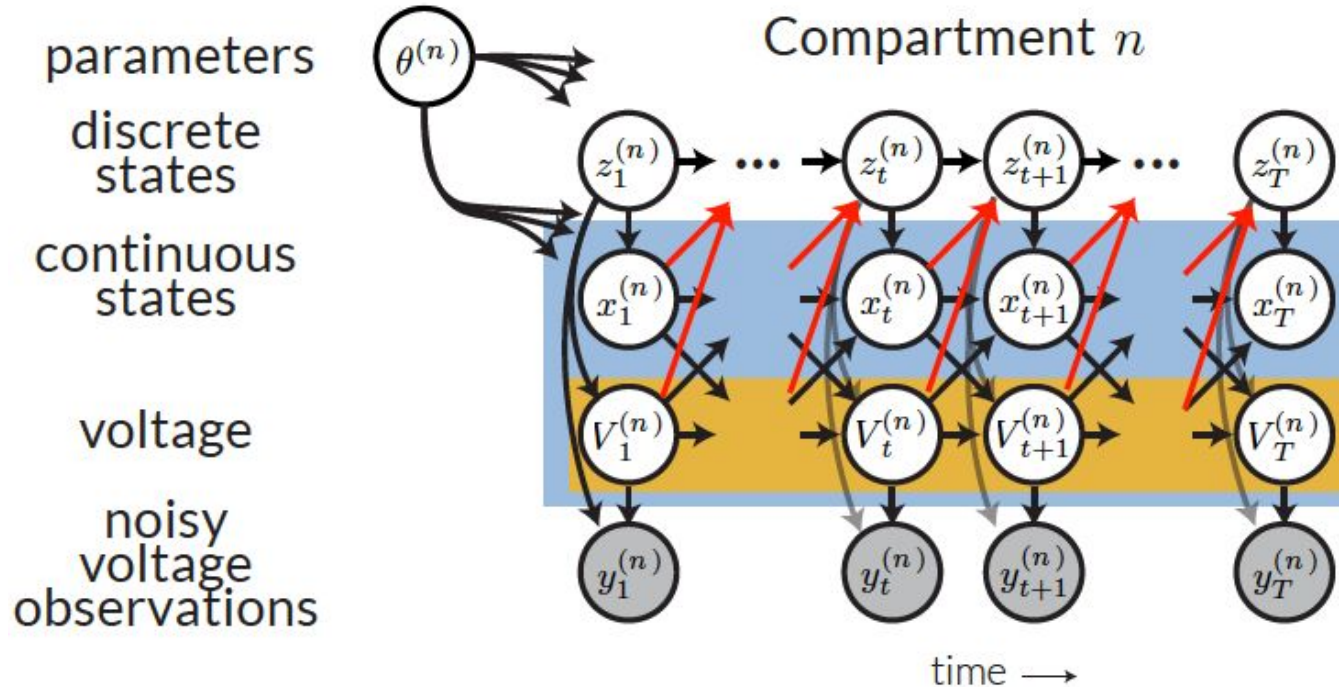
**V:** continuous latent variable  
(denoised voltage)

**Y:** observed variables



# Model Single Compartment Dynamics

- Recurrent **Switching** Linear Dynamical System (rSLDs)



# Statistical Model

- Recurrent **Switching** Linear Dynamical System (rSLDs)

Physical model

$$V_{t+\Delta t}^{(n)} \approx V_t^{(n)} + \frac{\Delta t}{C_n} \left[ \sum_j I_t^{(n,j)} + \sum_{n'=1}^N g_{nn'} \cdot (V_t^{(n')} - V_t^{(n)}) \right]$$

$$\mathbb{E} \left[ \begin{pmatrix} V_{t+\Delta t}^{(n)} \\ x_{t+\Delta t}^{(n)} \end{pmatrix} \middle| z_t^{(n)} = k, \left\{ V_t^{(n')} \right\}_{n' \neq n} \right]$$

$$= \begin{pmatrix} V_t^{(n)} \\ x_t^{(n)} \end{pmatrix} + \frac{\Delta t}{C_n} \left[ \underbrace{A_k^{(n)} \begin{pmatrix} V_t^{(n)} \\ x_t^{(n)} \end{pmatrix} + b_k^{(n)}}_{\approx \sum_j I_t^{(n,j)}} + \begin{pmatrix} \sum_{n'=1}^N g_{nn'} \cdot (V_t^{(n')} - V_t^{(n)}) \\ 0 \end{pmatrix} \right]$$

**theta**: parameters; **Z**: discrete latent variable; **X**: continuous latent variable (cycle parameters);  
**V**: continuous latent variable (denoised voltage); **Y**: observed variables

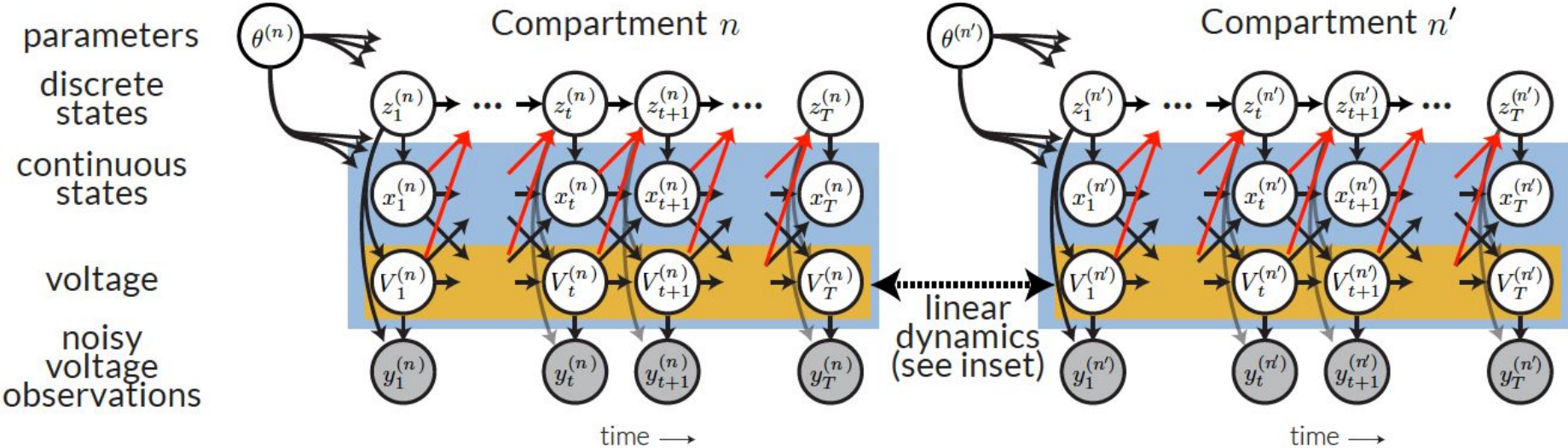
# Statistical Model

- **Recurrent** Switching Linear Dynamical System (rSLDs)

$$p(z_{t+\Delta t}^{(n)} = k \mid V_t^{(n)}, x_t^{(n)}) \propto \exp \left\{ w_k^{(n)\top} \begin{pmatrix} V_t^{(n)} \\ x_t^{(n)} \end{pmatrix} + d_k^{(n)} \right\}$$

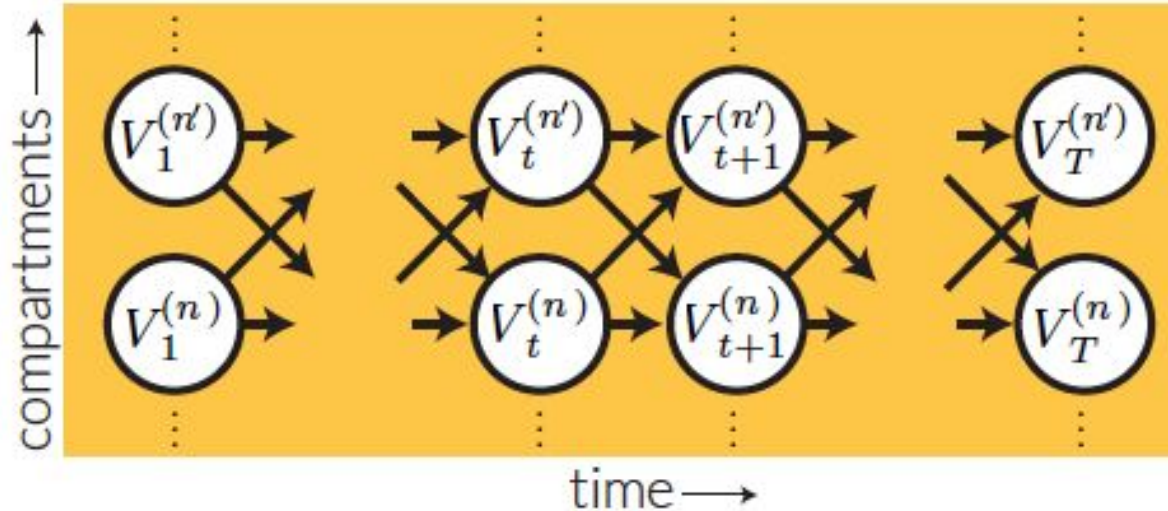
**theta**: parameters; **Z**: discrete latent variable; **X**: continuous latent variable (cycle parameters);  
**V**: continuous latent variable (denoised voltage); **Y**: observed variables

# Model Inter-Compartment Dynamics



# Linear Dependency between Adjacent Compartments

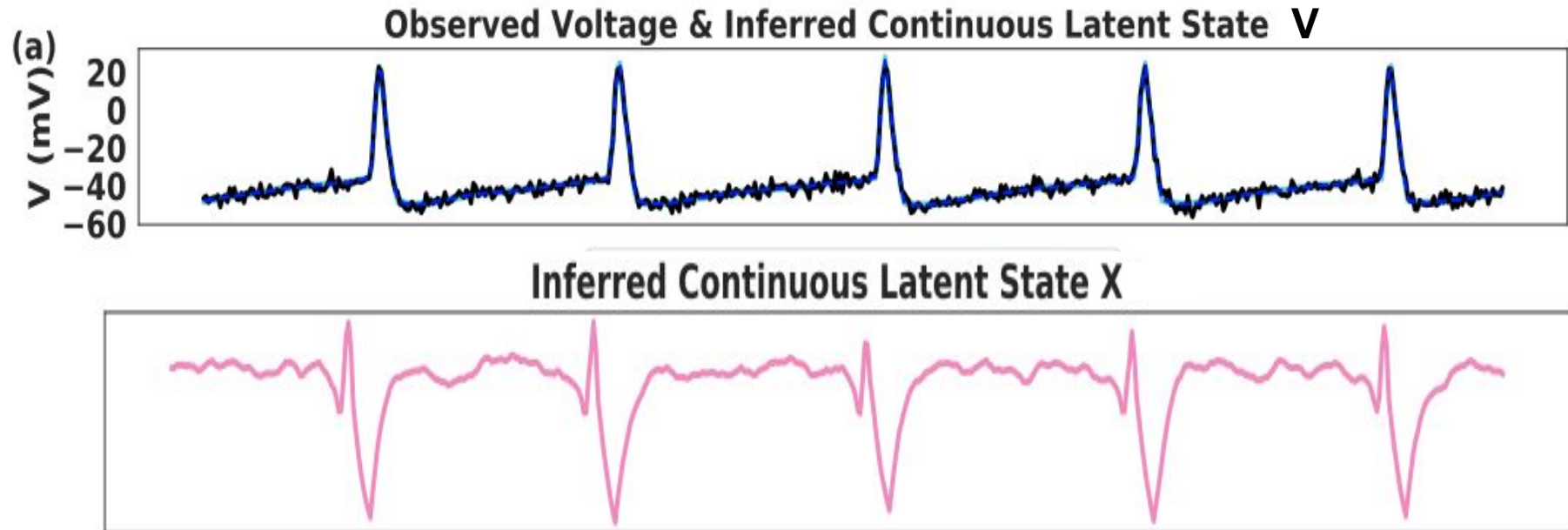
Inset: linear cable eqn. dynamics



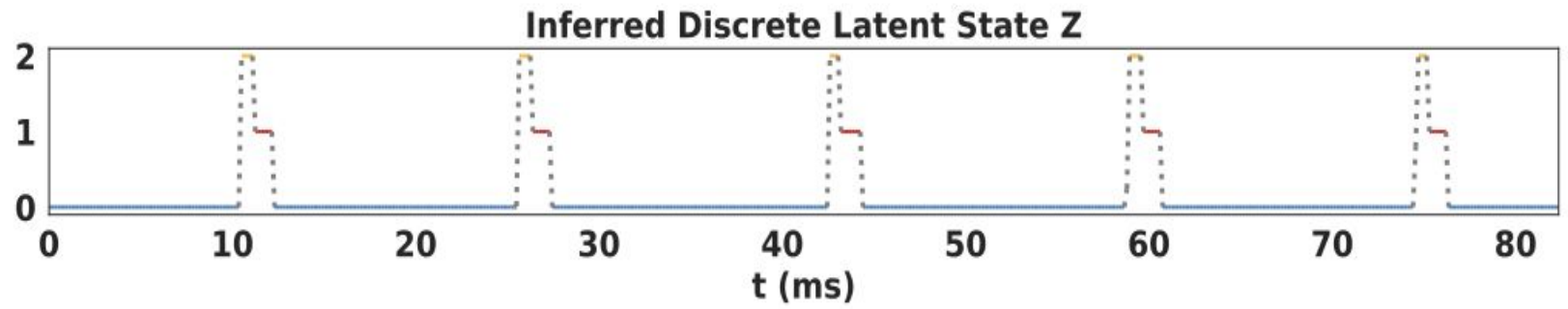
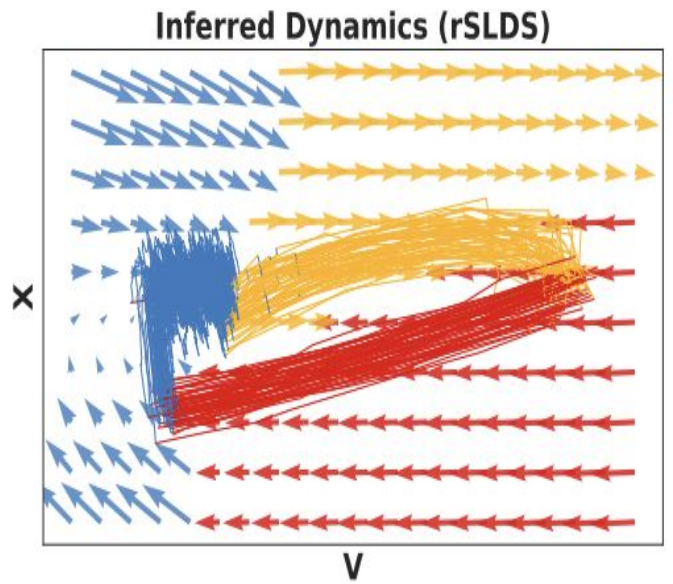
# Results: Single Compartment

# Output of the model for Single Compartment model

- Observed Voltage ( $y$ )
- Inferred Continuous Latent State:  $V$  (voltage) and  $X$  (cycle)

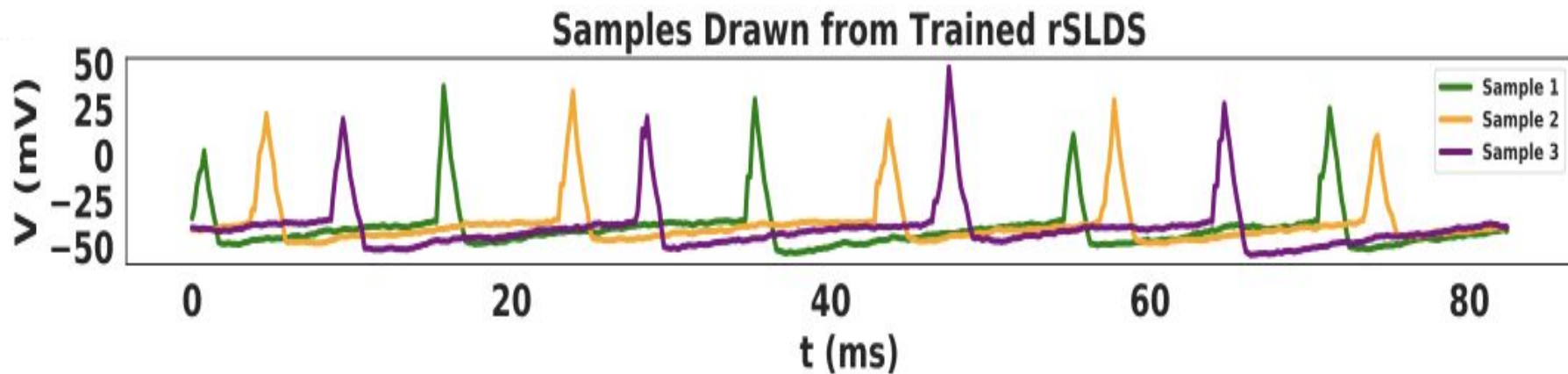
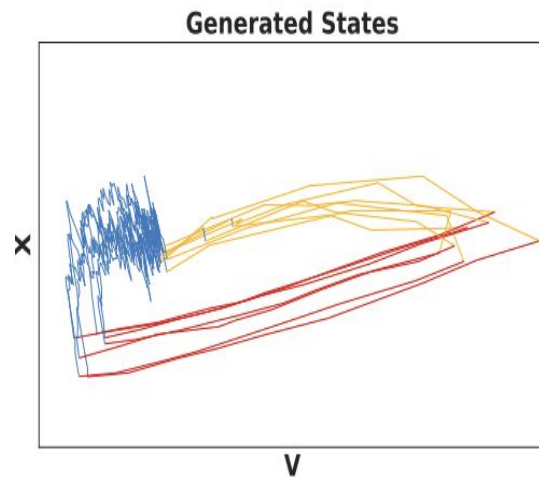


- Inferred Discrete Latent State (Z)





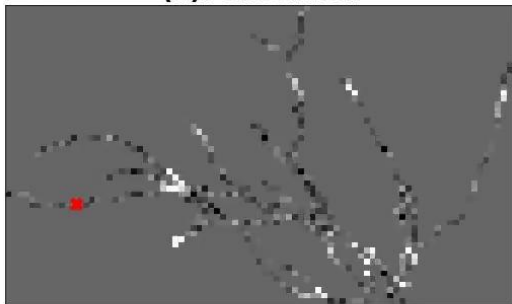
- Generated new spike (voltage)



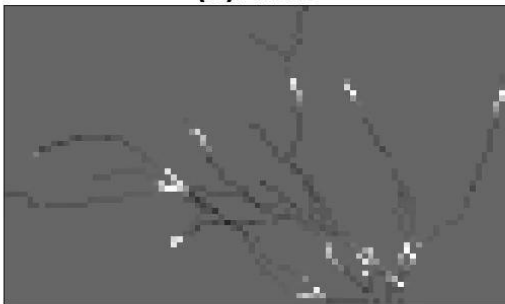
# **Results: Multiple Compartments**

# Multiple Compartment denoising

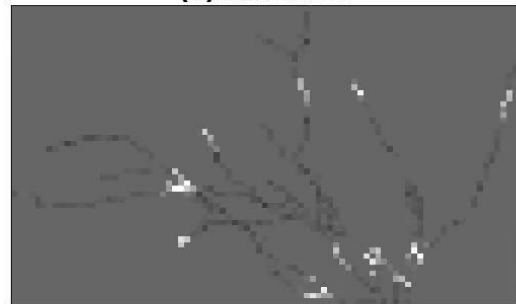
(a). Observed



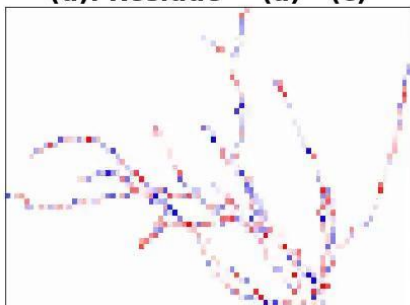
(b). True



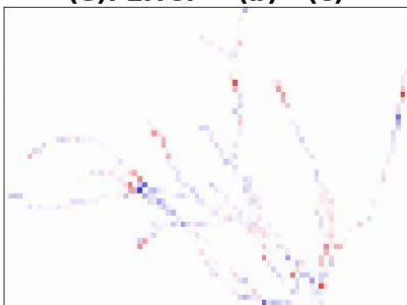
(c). Inferred



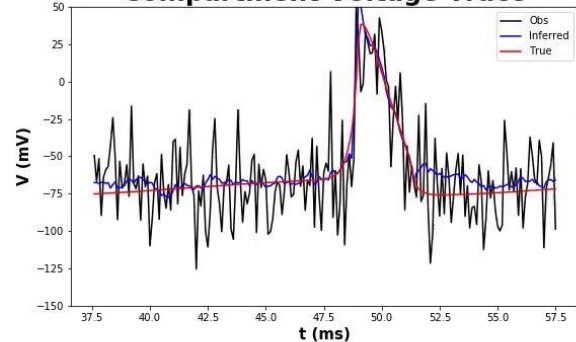
(d). Residue = (a) - (c)



(e). Error = (b) - (c)

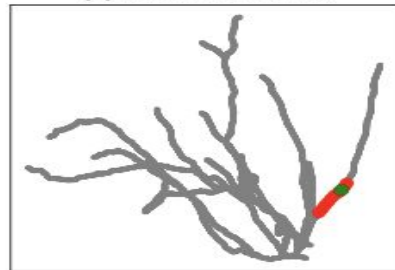


Compartment Voltage Trace

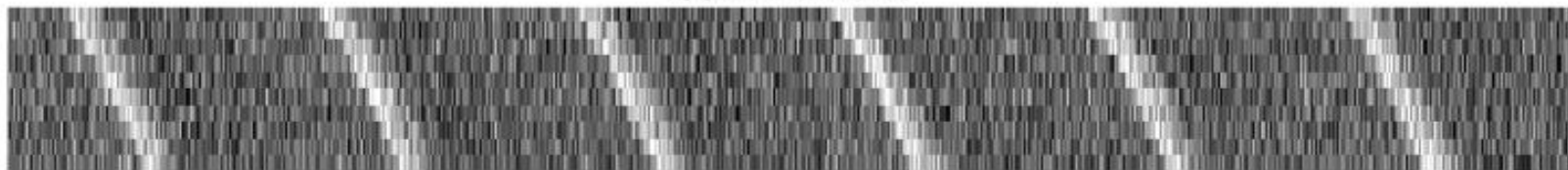


# Inferred Voltage

(f) Dendritic tree



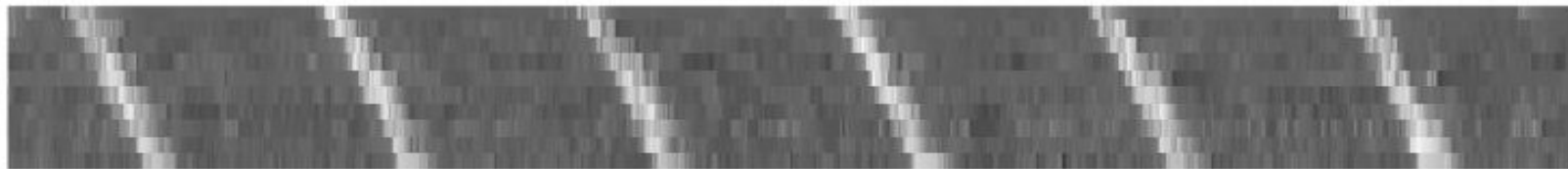
(i) Observed



(j) True



(k) Inferred



Compartment ID

Thank you!

Poster: #147

Code: [https://github.com/SunRuoxi/Voltage\\_Smoothing\\_with\\_rSLDS](https://github.com/SunRuoxi/Voltage_Smoothing_with_rSLDS)

# Previous Biophysical work

- Hodgkin Huxley
- Fitzhugh-Nagumo