# Introduction of Gamma Oscillation by Synaptic Inhibition through Wang-Buzsaki Model (1996)

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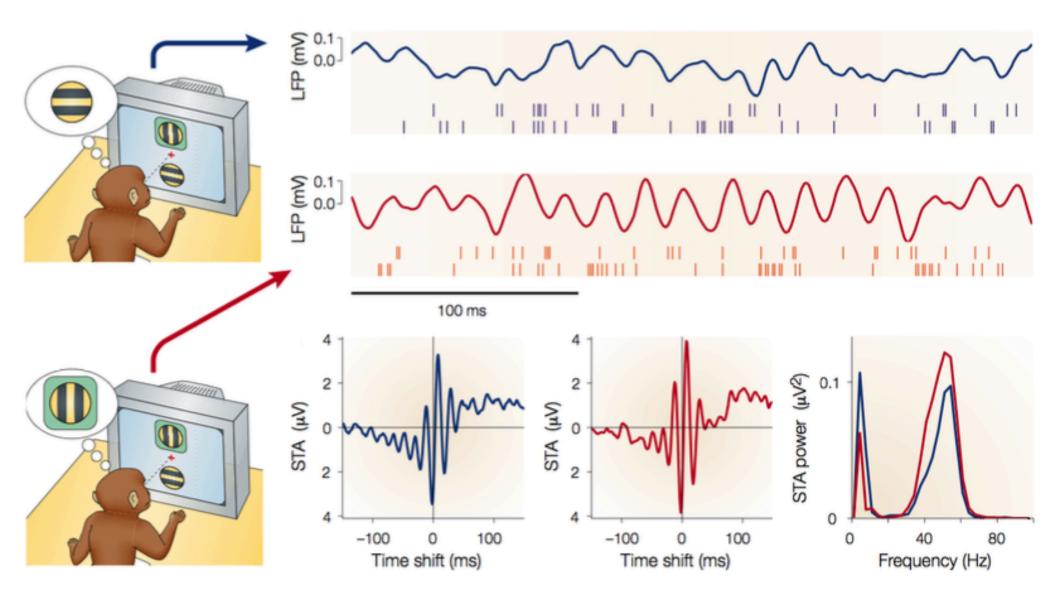
Special Topic: Modeling and Simulation in Science, Engineering, and Economics

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#### Gamma Oscillation

- Fast neuronal oscillations (30-60 Hz) observed in spiking activity of neurons of the neocortex depend on stimulus.
- Under certain behavioral condition, the synchronous rhythm is <u>confined</u> to selective neural populations rather than spatially uniform across the cortex.



Selective attention induces changes in synchrony in the visual cortex [Fries et al. (2001), Wang (2010)]
Red (blue): attention directed inside (outside) the RF

LFP displays enhanced gamma oscillations for preferred spatial location:

- Synchronization in time with the population rhythm
- Changes of firing rates of sensory neurons

# Synchronization of Neural Oscillators

- Brain rhythms as an emerging phenomenon in a network of oscillatory neurons.
- Whereas Gamma oscillations can be synchronized over a range of spatial scales, rhythmogenesis arises locally from microcircuit mechanisms [Fisahn et al. 1998, Atallah and Scanziani 2009].

• Recurrent excitation (E-E); Mutual inhibition (I-I); Feedback Inhibition (E-I-E).

Synchronizing coupled neurons only when synaptic excitation is very fast

WB Model (Type I single neuron and  $GABA_A$  receptor for synaptic coupling)

HH-type & Modified Integrate-and-Fire Model with nonlinear sodium and potassium conductances

# Model Neuron & Synapse

$$c_{\rm m} \frac{dV}{dt} = -I_{\rm Na} - I_{\rm K} - I_{\rm L} - I_{\rm syn} + I_{\rm app}$$

- Spike-generating  $Na^+$  and  $K^+$  voltage-dependent ion currents + transient sodium current.
  - Shift of (in)activation curves and impact  $m_{\infty}$  &  $h_{\infty}$  and  $n_{\infty}$  comparing to Hodgkin–Huxley.
- Gating variables are described in first-order kinetics (e.g.  $\frac{dh}{dt} = \phi \left( \alpha_{\rm h} (1-h) \beta_{\rm h} h \right)$ ).
- 1. Action potential is followed by a brief afterhyperpolarization
  - Relatively small maximal conductance  $g_{\rm K}$  and fast gating process of  $I_{\rm K}$ .
- 2. Interneurons have ability to fire repetitive spikes at high frequencies
  - Fast kinetics of  $I_{\rm Na}$  inactivation &  $I_{\rm K}$  activation and relatively high threshold of  $I_{\rm K}$ .
  - Sensitive to input heterogeneities at smaller  $I_{\rm app}$  values.

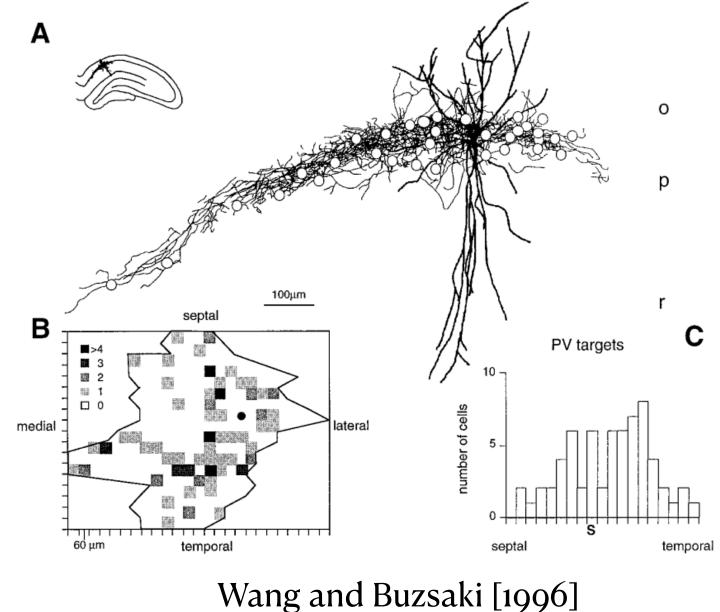
Fraction of open synaptic ion channels.

$$I_{\text{syn}} = g_{\text{syn}} s(V - E_{\text{syn}}), \frac{ds}{dt} = \alpha F(V_{\text{pre}})(1 - s) - \beta s.$$

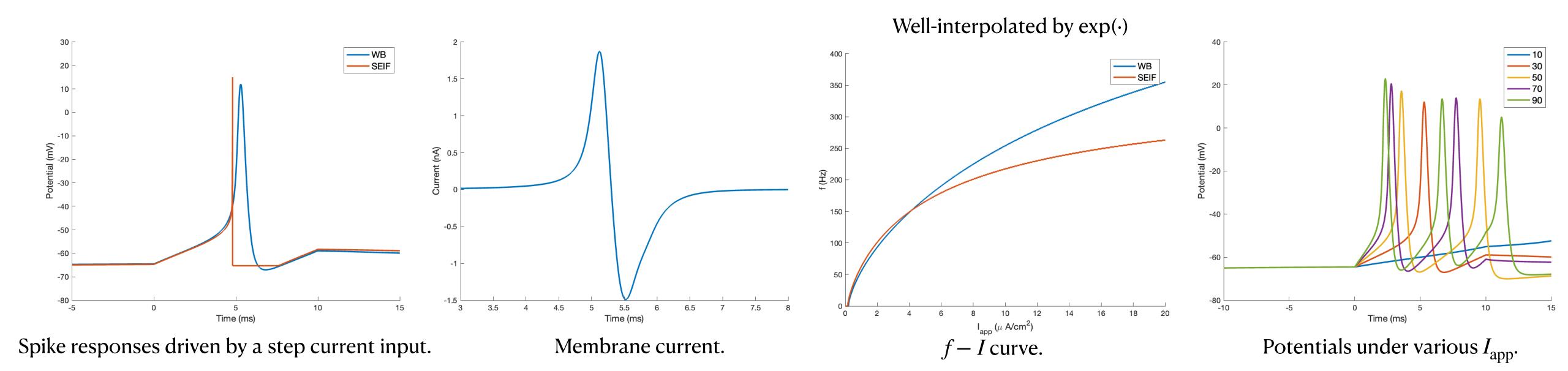
# Modeling Network

- Random network connectivity  $M_{\mathrm{syn}}$  & Fully coupled (all-to-all) connectivity
  - Estimate from CA1 double staining of PV interneuron in hippocampus [Sik et al., 1995].
  - Axonal arborization of an intracellularly labeled basket cell was largely confined in the striatum pyramidale.
- Heterogeneous inputs (Gaussian distribution).
- Network coherence & Population coherence
  - Cross-correlation of spike trains at zero time lag within a time bin of  $\Delta t = \tau$ .

$$\kappa_{ij}(\tau) = \frac{\sum_{l=1}^{K} X(l)Y(l)}{\sqrt{\sum_{l=1}^{K} X(l) \sum_{l=1}^{K} Y(l)}}$$



#### Response Properties of Single-Compartment Model



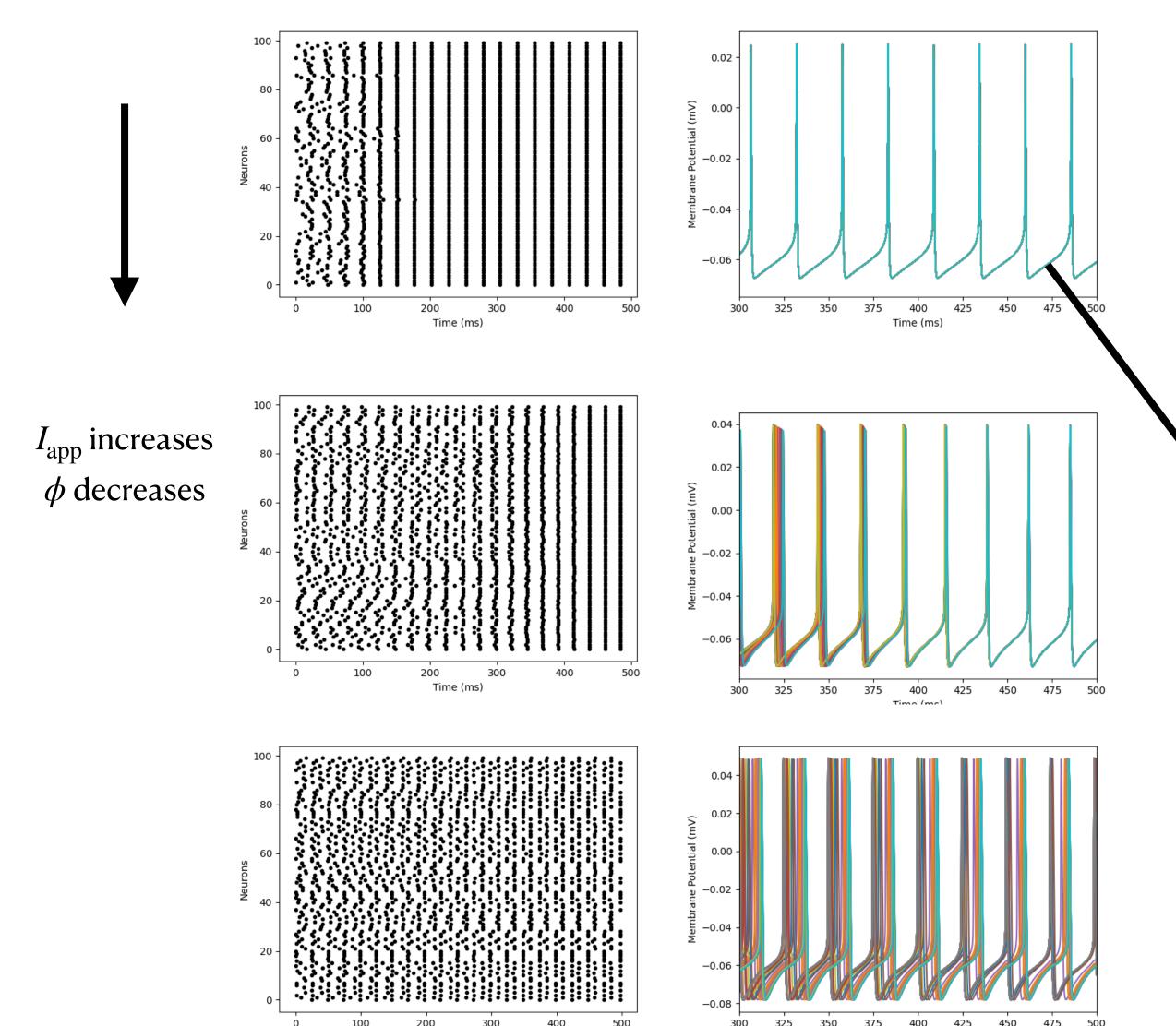
Comparison with (Standard)Exponential Integrate-and-Fire Model [Fourcaud-Trocme et al. (2003)]

• Positive feedback underlying action potential.

$$c_{\rm m} \frac{dV}{dt} = -g_{\rm L} \left( V - E_{\rm L} \right) + \Delta \exp \left( \left( V - V_{\rm th} \right) / \Delta \right) + I_{\rm app}$$

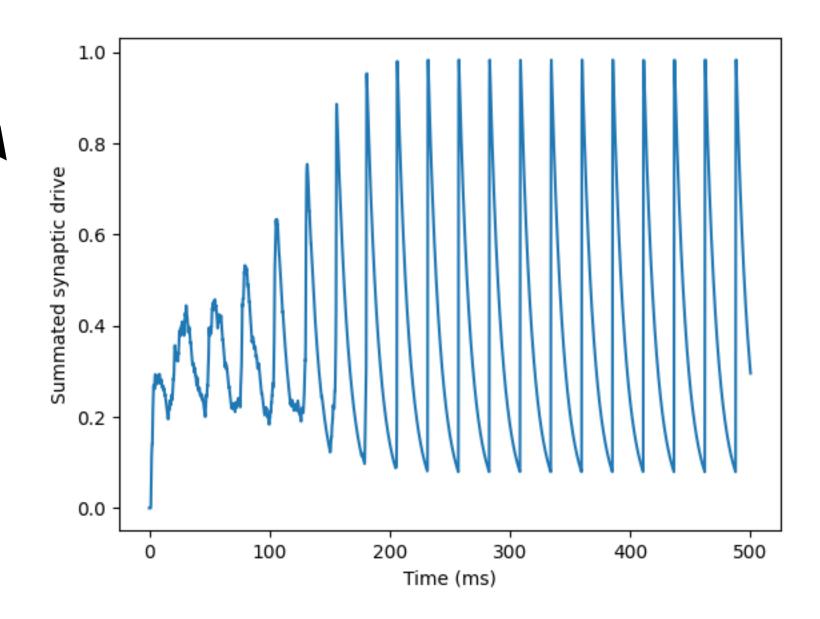
$$r = \beta \sqrt{I_{\rm app} - I_{\rm c}}$$

# Synchronization by GABA<sub>A</sub> synapses

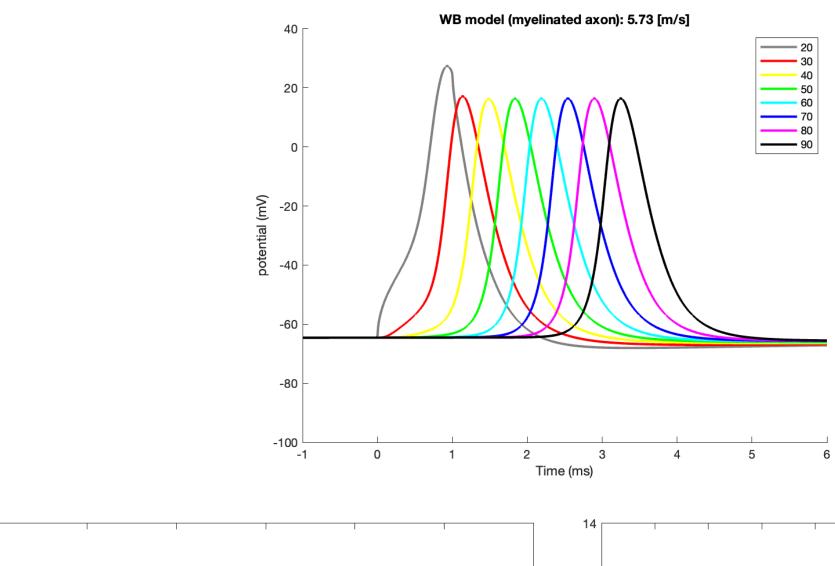


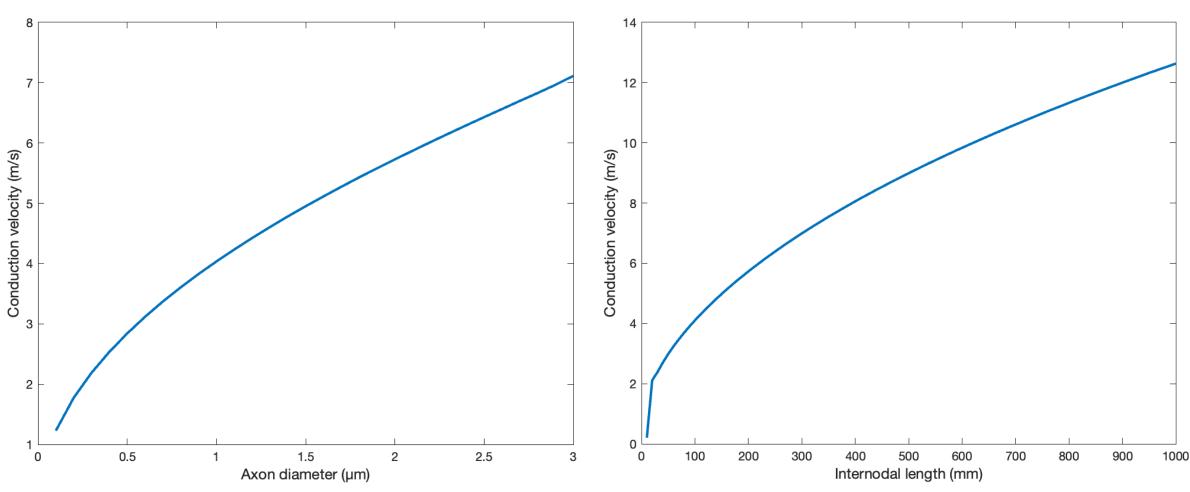
Time (ms)

- Rastergrams & Membrane potentials of cells (N=100) under various  $I_{\rm app}$  and  $\phi$  to preserve a similar oscillation frequency.
- With smaller  $\phi$  values,  $I_{\rm K}$  activation &  $I_{\rm Na}$  inactivation is slower and the AHP amplitude is more negative.
- When  $V_{\rm AHP} < E_{\rm syn}$ , the full synchrony is lost and switches into 2 clusters alternating in time.



# Application of WB Model: Myelinated Axons





- Series of excitable units interconnected with an axial resistance (with negligible capacitance and transmembrane conductance) [Moore et al., 1978].
- Each nodal compartment is described in WB model.
- 1D non-branching axon and stimulus current is injected intracellularly.

Axonal current: 
$$I_{axon^{j}} = g_{axon}^{j-1}(V_{j-1} - V_{j}) + g_{axon}^{j}(V_{j+1} - V_{j})$$

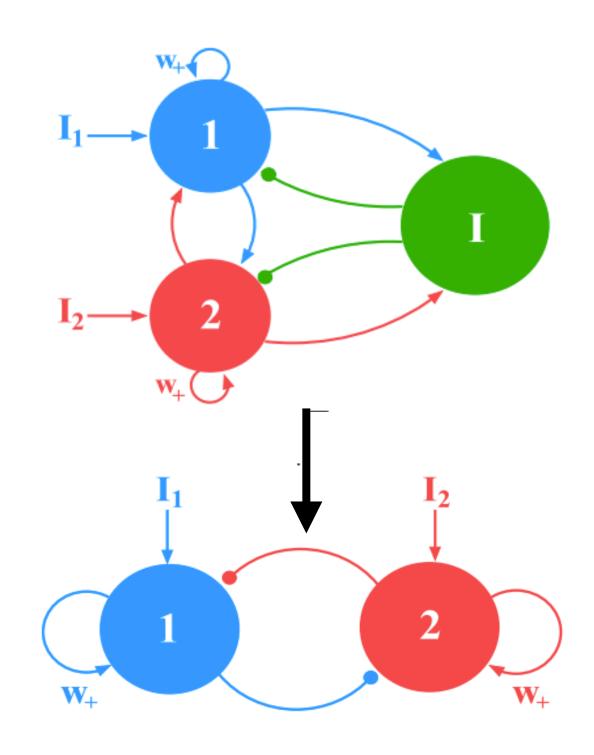
# Preliminary Approach Population Rate Model of Recurrent Neural Circuit

- Two excitatory neural assemblies
  - Strong recurrent excitation between neurons with similar stimulus selectivity.
  - Effective mutual inhibition competition.

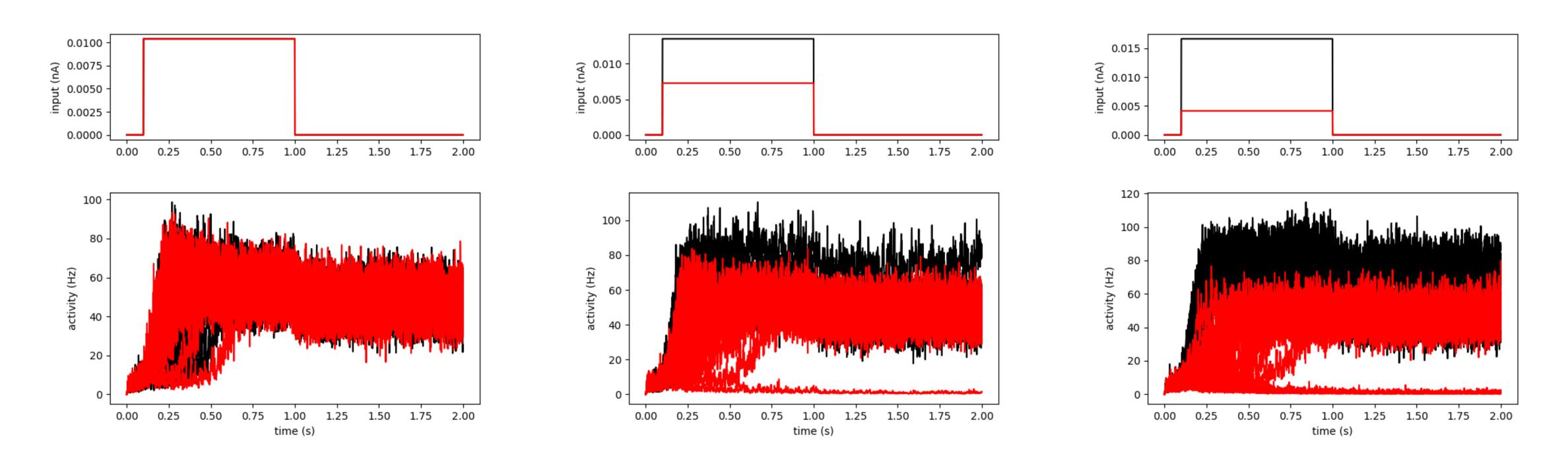
$$\frac{ds_1}{dt} = F(I_1) \gamma (1 - s_1) - s_1/\tau_s, r_i = F(I_i)$$

$$I_1 = g_E s_1 - g_I s_2 + I_{b1} + g_{ext} \mu_1$$

$$\tau_0 \frac{dI_{b1}}{dt} = -(I_{b1} - I_0) + \eta_1(t) \sqrt{\tau_0 \sigma^2} \text{ (Ornstein-Uhlenbeck)}$$



Wong and Wang (2006)



Different formulations of background inputs; n = 20.

### Summary

- Reproduce the behaviors and dynamics of neurons and synapses described in Wang-Buzsaki 1996 model.
- Regenerate the synchronization by  $GABA_A$  synapses in the neuronal network.
- Use WB model to simulate the myelinated axons and characterize the action potential propagation.
- Adapt Wong-Wang 2006 model to describe two excitatory WB neural assemblies and their mutual inhibitory interactions.