#### In Silico Model of Neuron-Glia Interaction

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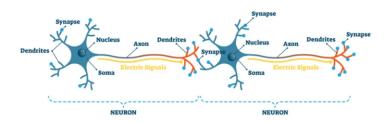
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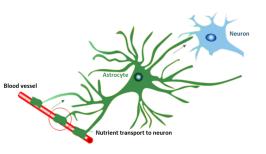
Università degli Studi di Pisa Dipartimento di Scienze Matematiche Fisiche e Naturali Corso di Laurea Magistrale in Fisica



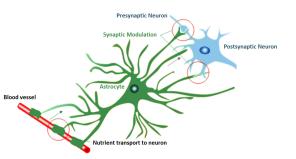
- Central Nervous Systems (CNS) elaborates the external stimuli from the environment to regulate and control biological processes
- Neurons are the specialized cells that accomplish these tasks through the electrical properties of the membrane
- The abrupt and rapid increase of membrane potential (spike) is transmitted from one neuron to other via Synapses



- The glial cells are the main cell type in the CNS. These non excitable cells appear to fall into three different types: oligodendrocytes, microglia and astrocytes
- Astrocytes perform number of biological processes such as neurons-vascular interface



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- Astrocytes perform number of biological processes such as neurons-vascular interface



Astrocytes have a prominent role in the signalling transmission at microscopic (synaptic) scale

- Tripartite Synapse: the astrocyte is the third active element in synaptic communication between the pre- and postsynaptic neurons
- The efforts in Computation Neuroscience lead to built mathematical models to describe the interplay between astrocytes and neurons
- At mesoscopic (network) level of description, our knowledge of the effect involved neuron-glia interaction is not complete:

#### Current works

- Ullah (2009): homeostatic regulation
- Savtchenko (2014): neglect dynamical feedback between neuronal and astrocytic activity
- 3 Stimberg et al. (2017): Neuron-Glia network partially explored

#### Topic

- **Tripartite Synapse**: the astrocyte is the third active element in synaptic communication between the pre- and postsynaptic neurons
- The efforts in Computation Neuroscience lead to built mathematical models to describe the interplay between astrocytes and neurons
- At mesoscopic (network) level of description, our knowledge of the effect involved neuron-glia interaction is not complete:

#### Purpose of present thesis

The present thesis aims to fill the gap between microscopic and mesoscopic descriptions of the interplay between neurons and astrocytes

#### Outline

- Modelling Approach
- Results 1: Tripartite Synapse

3 Results 2: Neuron-Glia Network

4 Conclusion

#### Overview

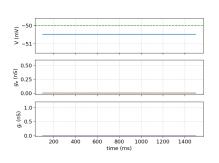
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### Neuron - Conductance base Integrate and Fire

 Neural response to spikes arrival is described by subthreshold dynamics (Integrate) and spike generation mechanism (Fire)

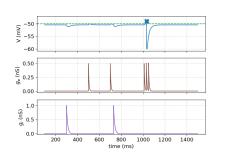


subthreshold dynamics:

$$C_m rac{dV}{dt} = g_I(E_I - V) + \ g_e(E_e - V) + g_i(E_i - V) + I_{ext}(t) \ rac{dg_e}{dt} = -rac{g_e}{ au_e} \ rac{dg_i}{dt} = -rac{g_i}{ au_i}$$

# Neuron - Conductance base Integrate and Fire

 Neural response to spikes arrival is described by subthreshold dynamics (Integrate) and spike generation mechanism (Fire)



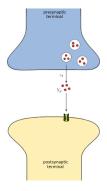
subthreshold dynamics:

$$C_m rac{dV}{dt} = g_I(E_I - V) + g_e(E_e - V) + g_i(E_i - V) + I_{ext}(t)$$
 $rac{dg_e}{dt} = -rac{g_e}{ au_e} \quad ext{spike arrival:} \quad g_e o g_e + w_e$ 
 $rac{dg_i}{dt} = -rac{g_i}{ au_i} \quad ext{spike arrival:} \quad g_i o g_i + w_i$ 

- spike generation:
  - 1)  $V(t) = V_{thr}$
  - 2) a spike is generate (point-wise event)
  - 3)  $V(t) \rightarrow V_{res}$

# Synapse - Short-Term Plasticity (STP)

- Synapses release a cert amount of neurotransmitters that induce an increase/decrease in post synaptic conductance
- STP: synaptic release  $r_S$  depends on spike timing



x<sub>S</sub>: fraction of available resourcesu<sub>S</sub>: fraction of resources effectively

$$\frac{du_S}{dt} = \Omega_f u_S$$

$$\frac{du_S}{dt} = \Omega_d (1 - x_S)$$

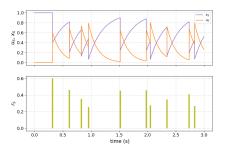
$$\frac{dx_S}{dt} = \Omega_d (1 - x_S)$$
spike arrival:
$$u_S \rightarrow u_S + u_0 (1 - u_S)$$

$$r_S = x_S u_S$$

$$x_S \rightarrow x_S - r_S$$

# Synapse - Short-Term Plasticity (STP)

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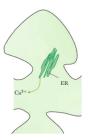
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$$r_S = x_S u_S$$
 
$$r_S \rightarrow x_S - r_S$$
 spike arrival:  

$$u_S \rightarrow u_S + u_0(1 - u_S)$$
 
$$r_S = x_S v_S - r_S$$

•  $r_S$  modulate the conductance's updating:  $g_X \rightarrow g_X + r_S w_X$  with x = e, i

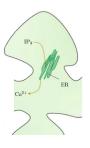
Tsodyks M.V., Markram H. Proc Natl Acad Sci (1997)

### Astrocyte - G-Chl Model



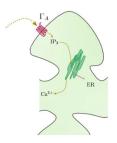
 Endoplasmatic Reticulum (ER) regulates the intracellular Ca<sup>2+</sup> level

### Astrocyte - G-Chl Model



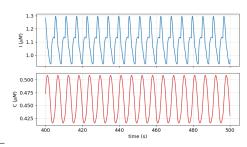
- Endoplasmatic Reticulum (ER) regulates the intracellular Ca<sup>2+</sup> level
- Inositol 1,4,5-trisphosphate (IP<sub>3</sub>) regulates the ER calcium-release

# Astrocyte - G-Chl Model



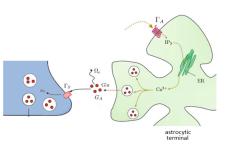
$$\begin{split} \frac{d\Gamma_A}{dt} &= \text{activ.} - \text{inactiv.} \\ \frac{dI}{dt} &= J_{in}(\Gamma_A, C, I) - J_{out}(C, I) \\ \frac{dC}{dt} &= J_{in}(C, I) - J_{out}(C) \end{split}$$

- Endoplasmatic Reticulum (ER) regulates the intracellular Ca<sup>2+</sup> level
- Inositol 1,4,5-trisphosphate (IP<sub>3</sub>) regulates the ER calcium-release
- Membrane receptors  $(\Gamma_A)$  regulates the intracellular IP<sub>3</sub> level



De Pittà et al. Computational Glioscience (2019)

### Gliomodulation of Synaptic Release



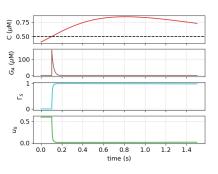
- The astrocyte releases the gliotransmitters
- The gliotrasmitters actives the presynaptic membrane receptors
- The activation/inactivation of membrane receptor modulate u<sub>0</sub>

#### Gliomodulation

The astrocytic activity induced a new degrees of freedom to basal release probability

$$u_0 \equiv u_0(G_A) = u_0(\Gamma_S(G_A))$$

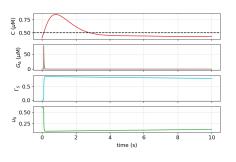
### Release-decreasing effect



- The astrocyte releases the gliotransmitters
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- ullet The activation/inactivation of membrane receptor modulate  $u_0$

$$u_0(t) = (1 - \Gamma_S(t))U_0^*$$

### Release-decreasing effect



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- The activation/inactivation of membrane receptor modulate  $u_0$  $u_0(t) = (1 - \Gamma_S(t))U_0^*$

#### Gliomodulation

The sequence of activation and inactivection of presynaptic receptors  $\Gamma_S$  drive the modulation of basal release probability

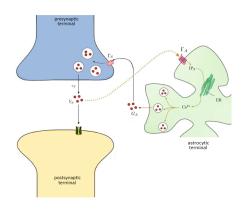
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- Modelling Approach
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3 Results 2: Neuron-Glia Network

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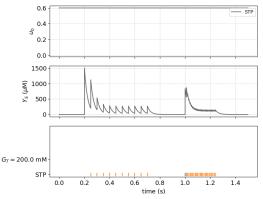
### Tripartite Synapse



- Bidirectional coupling between synapse and astrocyte
- Presynaptic firing rate  $\nu_S$  is the control parameter
- Modulation of basal releae probability:

$$u_0(t) = (1 - \Gamma_S(t))U_0^*$$

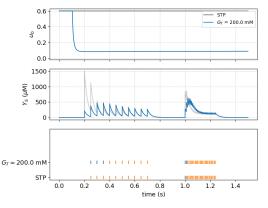
From dynamical perspective, we want to point out how the presynaptic firing rate  $\nu_S$  drives the dynamical behaviour



cine (3)

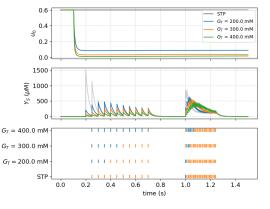
 STP induces depression transmission

Pair Pulse Ratio (PPR) =  $\frac{r_{S_{i+1}}}{r_{S_i}}$ 



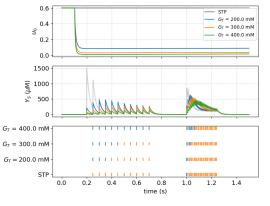
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- Gliomodulation induces facilitation transmission



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- STP induces depression transmission
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- Facilitation depends on coupling strenght  $(G_T)$

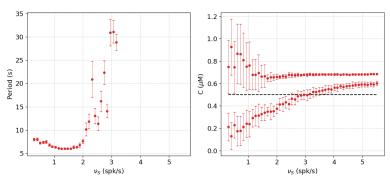


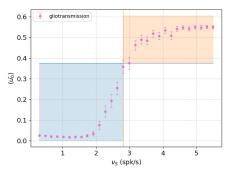
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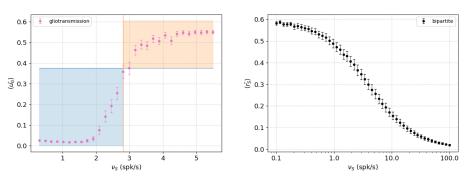
#### Glio-induced facilitation transmission

Astrocytic modulation leads to both depression and facilitation synaptic transmission.

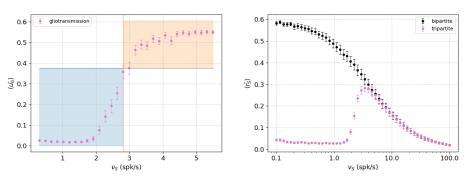
- ullet Analysis of tripartite synapses concerning the control parameter  $u_{\mathcal{S}}$
- ullet Presynaptic firing rate  $u_S$  leads to astrocytic calcium oscillations across the threshold







 The filtering characteristic curve of simple bipartire synapses shows monotonically decrease behaviour (low-pass filter)



- The filtering characteristic curve of simple bipartire synapses shows monotonically decrease behaviour (low-pass filter)
- The tripartite one turns into a bell-shape curve (band-pass filter)

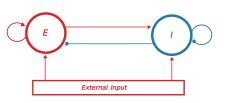
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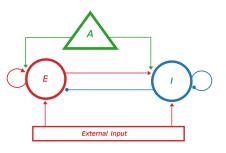
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#### Network Scheme



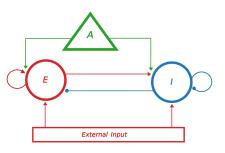
- Baseline: excitatory/inhibitory neural network
- External input: Poisson spike train with  $\nu_{ext}(t)$

#### **Network Scheme**



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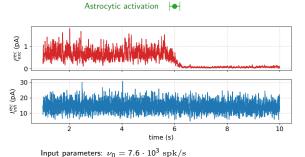
#### Gliomodulation at mesoscopic level of description

Compare the Neural and Neuron-Glia network dynamics through mesoscopic quantity:

Local Field Potential (LFP) 
$$=\sum_{l=1}^{N_e}|I_{exc_l}|+|I_{inh_l}|$$

### Effect of single Gliorelease event

• Constant external input  $(\nu_{ext}(t) = \nu_0)$  leads to regulate recurrent excitatory current  $I_{ext}^{rec}$ 

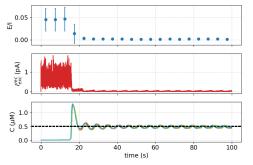


	LFP (a.u.)
baseline	$11.4 \pm 0.3$
gliomodulation	$11.3 \pm 0.3$

imput parameters.  $\nu_0 = 7.0^\circ$  10 SpR/S

#### Long-term effect Gliotransmission

- The regulation  $I_{\text{exc}}^{\text{rec}}$  suggests a modulation of E/I balace
- Long time simulations are mandatory to observe depressing and facilitating gliomodulation effect



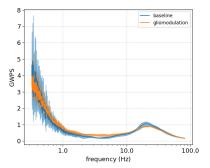
#### Input parameters: $\nu_0 = 7.6 \cdot 10^3 \text{ spk/s}$

#### Modulation of E/I balance

The sequence of depressing and facilitation mechanism induced by astrocytes leads to a persistent modulation of recurrent balance

#### Long-term effect Gliotransmission

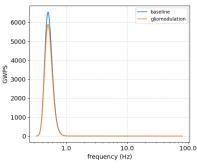
- Network oscillation are investigate through spectrum analysis of LFP
- External input:  $\nu_{ext}(t) = \nu_0$



Input parameters:  $\nu_0 = 7.6 \cdot 10^3 \text{ spk/s}$ 

### Long-term effect Gliotransmission

- Network oscillation are investigate through spectrum analysis of LFP
- External input:  $\nu_{\mathsf{ext}}(t) = \nu_0 + A\sin(2\pi\omega t)$
- Modulation of frequency s:  $M(s) = \frac{GWPS(s)_{\text{gliomodulation}} GWPS(s)_{\text{baseline}}}{GWPS(s)_{\text{baseline}}}$



#### Gliomodulation on NG-network

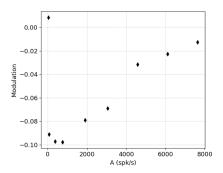
Gliotransmission modulates the periodic external input:

$$M(0.5) = 9.6 \pm 0.6 \%$$

Input parameters:  $\nu_0 = 7.6 \cdot 10^3 \ \mathrm{spk/s}; \ \omega = 0.5 \ \mathrm{Hz}; \ A = 0.1 \ \nu_0$ 

### Long-term effect Gliotransmission

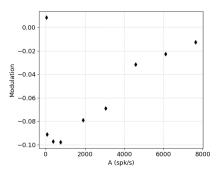
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#### Amplitude Gliomodulation

Amplitude modulation induced by astrocytic activity show "reverse-bell" shape behaviour

Input parameters:  $\nu_0 = 7.6 \cdot 10^3 \; \mathrm{spk/s}; \; \omega = 0.5 \; \mathrm{Hz}$ 

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#### Conclusion

### Original Results

#### Microscopic scale:

- Derivation of Mean Field description of Tripartite Synapse
- ② Gliomodulation allow the synapses to transmit the input in depression and facilitation modality
- Characteristics curve points out the filtering of low presynaptic firing rate

#### Mesoscopic scale:

- Gliomodulation filters the network oscillations but does not elicit new ones
- Astrocytes regulate the excitatory/inhibitory recurrent balance

#### Conclusion

### Future Perspective

- The collaboration of Scuola Superiore Sant'Anna, the Roma Sapienza University and the Camerino University aims to investigate the glioblastoma multiform (GBM)
- GBM is an aggressive brain tumour that alters the balance of inhibitory and excitatory neurotransmission and disrupts neural circuits
- Cancer cells release a huge non-physiological glutamate concentration (excitotoxicity)
- Exploiting NG-network can improve our knowledge of underpinning mechanisms of excitotoxicity and the relative alteration in neural network dynamics.

# Thank you for your attention!

### Neural-Glia Network

#### Current Model

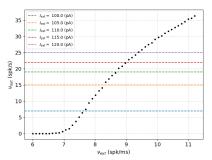
- Homogeneous and constant external input: I<sub>ext</sub>
- ② Astrocytic coupling:  $N_a = N_e$ . The first astrocyte is coupled with all inhibitory neurons.
- Fixed network structure and external connections
- $oldsymbol{0}$  Time simulation  $10~\mathrm{s}$  without persistent astrocytic activity

### Original Model

- **1** Poissonian external input  $\nu_{ext}(t)$ :  $I_{ext} = \tilde{w}_e (E_e V)$
- ② Astrocytic coupling:  $N_a = N_e + N_i$ . Homogeneous neurons-astrocytes coupling
- Network degrees of balance g:  $g = \frac{p_e}{p_i} = \frac{w_i N_i}{w_e N_e}$ Strength of external connection s:  $\tilde{w}_e = s w_e$  on inhibitory neurons
- Long-time simulation (100 s) with analysis of persistent activity.

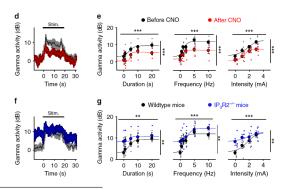
# Neural-Glia Network - Poisson Input

- **0** evaluate the neural activity  $\nu_{out}$  due to constant external current  $I_{ext}$ ;
- **3** plot the input/output curve for different values of  $\nu_{\text{ext}}$ ;
- **4** set  $\nu_{\mathsf{ext}}$  such that the activity in 1. is equal to activity in 2.



### Experimental evidence

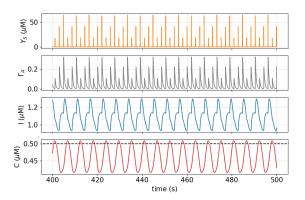
- Astrocytic population in Somatosensory Cortex (S1) mice regulate the stimulus-envoked  $\gamma$ -oscillation in neural network
- The astrocytic activity decrease the intensity of network oscillation in LFP recording



Lines, J., Martin, E.D., Kofuji, P. et al. Nat Commun (2020)

# G-ChI model with homogeneous external input

- Noiseless firing rate  $u_{s} = 0.5 \; \mathrm{spk/s}$  dirve the dynamical behaviour
- The sequence of neurotransmitter release affect  $\Gamma_S$ , I and C
- The oscillation across threshold modulate the release of Y<sub>S</sub>, onset of glio-induced facilitation transmission



# Mean Field description of Bipartite Synpases

Dynamics of mean value of variables over trial :

$$\begin{aligned} \frac{d\bar{u}_{S}}{dt} &= \Omega_{f}(u_{0} - \bar{u}_{S}) + \frac{u_{0}}{n_{S}} \sum_{l=1}^{n_{S}} \sum_{k} (1 - u_{S_{l}}) \, \delta(t - t_{l_{k}}) \\ \frac{d\bar{x}_{S}}{dt} &= \Omega_{d}(1 - \bar{x}_{S}) - \frac{1}{n_{S}} \sum_{l=1}^{n_{S}} \sum_{k} u_{S_{l}} x_{S_{l}} \, \delta(t - t_{l_{k}}) \end{aligned}$$

• Dynamics of mean value of variables over time:

$$egin{split} rac{d\langlear{u}_S
angle}{dt} &= \Omega_f(u_0 - \langlear{u}_S
angle) + u_0(1 - \langlear{u}_S
angle)\,
u_S(t) \ rac{d\langlear{x}_S
angle}{dt} &= \Omega_d(1 - \langlear{x}_S
angle) - \langlear{u}_S
angle\langlear{x}_S
angle\,
u_S(t) \end{split}$$

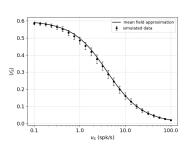
# Mean Field description of Bipartite Synapses

• Steady-state with constant firing rate  $\nu_S(t) = \nu_S$ :

$$\begin{split} \langle \bar{u}_S \rangle &= \frac{u_0(\Omega_f + \nu_S)}{\Omega_f + \nu_S \, u_0} \\ \langle \bar{x}_S \rangle &= \frac{\Omega_d}{\Omega_d + \langle \bar{u}_S \rangle \, \nu_S} \\ \langle \bar{r}_S \rangle &= \frac{u_0 \Omega_d(\Omega_f + \nu_S)}{(\Omega_f + \nu_S \, u_0)(\Omega_d + \langle \bar{u}_S \rangle \, \nu_S)} \end{split}$$

• The slope of  $\langle \bar{r}_S \rangle$  determine the threshold value for depression an facilitation transmission:

$$u_{\theta} = \frac{\Omega_d}{\Omega_d + \Omega_f}$$



# Mean Filed description of Tripartite Synapse

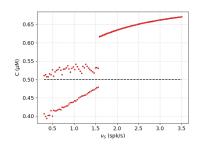
- Mean field description starts from the average over time ( $\langle \cdot \rangle$ ) and trials  $(\bar{\cdot})$  of synaptic variables
- Assuming the statistical independence between  $u_0$  and  $u_S$  we obtain:

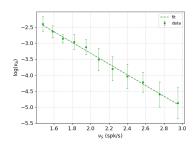
$$\begin{split} &\frac{d\langle \bar{u}_S\rangle}{dt} = \Omega_f\big(\langle \bar{u}_0\rangle + \langle \bar{u}_S\rangle\big) + \langle \bar{u}_0\rangle\big(1 - \langle \bar{u}_S\rangle\big)\,\nu_S \\ &\frac{d\langle \bar{x}_S\rangle}{dt} = \Omega_d\big(1 - \langle \bar{x}_S\rangle\big) - \langle \bar{u}_S\rangle\langle \bar{x}_S\rangle\,\nu_S \\ &\frac{d\langle \bar{x}_A\rangle}{dt} = \Omega_A\big(1 - \langle \bar{x}_A\rangle\big) - U_A\langle \bar{x}_A\rangle\nu_A(\nu_S) \\ &\frac{d\langle \bar{\Gamma}_S\rangle}{dt} = J_SU_A\langle \bar{x}_A\rangle\big(1 - \langle \bar{\Gamma}_S\rangle\big)\nu_A(\nu_S) - \Omega_G\langle \bar{\Gamma}_S\rangle \\ &\frac{d\langle \bar{\Gamma}_S\rangle}{dt} = U_0^* + (\alpha - U_0^*)\langle \bar{\Gamma}_S\rangle \end{split}$$

# Mean Field desciprtion of Tripartite Synapse

- The dynamical behaviour of G-ChI model with noiseless input give insights regarding biological function  $\nu_A(\nu_S)$
- ullet Qualitative dynamical change for the bifurcation value  $u_{\mathsf{S}}^{\mathit{bif}}$
- According we propose the guess fuction:

$$\nu_{A} = \left\{ \begin{array}{ll} \nu_{A} = \nu_{A_0} & \text{with } \nu_{S} \leq \nu_{S}^{bif} \\ \nu_{A} = \nu_{A_0} e^{-\tau_{A}(\nu_{S} - \nu_{S}^{bif})} & \text{with } \nu_{S} > \nu_{S}^{bif} \end{array} \right.$$

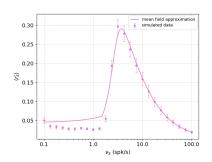




# Mean Field desciprtion of Tripartite Synapse

 The mean field approximation well describe the shape of characteristc curve

$$\begin{split} \langle \bar{u}_S \rangle &= \frac{\langle \bar{u}_0 \rangle (\Omega_f + \nu_S)}{\Omega_f + \nu_S \langle \bar{u}_0 \rangle} \\ \langle \bar{x}_S \rangle &= \frac{\Omega_d}{\Omega_d + \langle \bar{u}_S \rangle \nu_S} \\ \langle \bar{x}_A \rangle &= \frac{\Omega_A}{\Omega_A + U_A \nu_A (\nu_S)} \\ \langle \bar{\Gamma}_S \rangle &= \frac{J_S \Omega_A U_A \nu_A (\nu_S)}{\Omega_A \Omega_G + (J_S \Omega_A + \Omega_G) U_A \nu_A (\nu_S)} \\ \langle \bar{u}_0 \rangle &= U_0^* + (\alpha - U_0^*) \langle \bar{\Gamma}_S \rangle \end{split}$$



# NG-network activity with external periodic signal

• Statistical independence between  $u_0$  and  $u_S$  has been estimated by the Cauchy-Swarz:

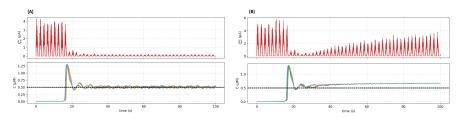
$$\frac{|\langle ab \rangle - \langle a \rangle \langle b \rangle|}{\langle a \rangle \langle b \rangle|} \leq \mathsf{CV}_a \, \mathsf{CV}_b$$

$$\mathsf{CV}_{u_S}^2 = \frac{\Omega_f (1 - \langle u_0 \rangle)^2 \nu_S}{(\Omega_f + \nu_s)(2\Omega_f + \langle u_0 \rangle(2 - \langle u_0 \rangle) \nu_S}$$

$$\mathsf{CV}_{u_0}^2 = \frac{\langle \Gamma_S \rangle^2}{(1 - \langle \Gamma_S \rangle)^2} \frac{\Omega_G^2}{(g + (1 - \beta)\nu_A)(\Omega_G + (1 + \beta)\nu_A)}$$

# NG-network activity with external periodic signal

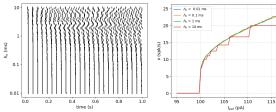
- The persistent astrocytic activity depends on Amplitude of external periodic input
- Intense amplitude does not further sustain gliotransmission. The amplitude gliomodulation tends to 0.



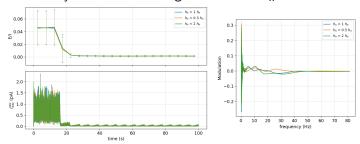
External Input:  $\nu_{\rm ext}(t) = \nu_0 + A \sin(2\pi\omega t)$ . Parameters:  $\omega = 0.5~{\rm Hz}$ ;  $\nu_0 = 7.6 \cdot 10^3~{\rm spk/s}$ ; A)  $A = 0.4 \, \nu_0$ ; b)  $A = 0.6 \, \nu_0$ 

# Numerical Integration

ullet IF generates different firing rate concerning the integration step  $h_n$ 



• NG-network dynamics in the neighborhood of  $h_n = 0.05 \text{ ms}$ 



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