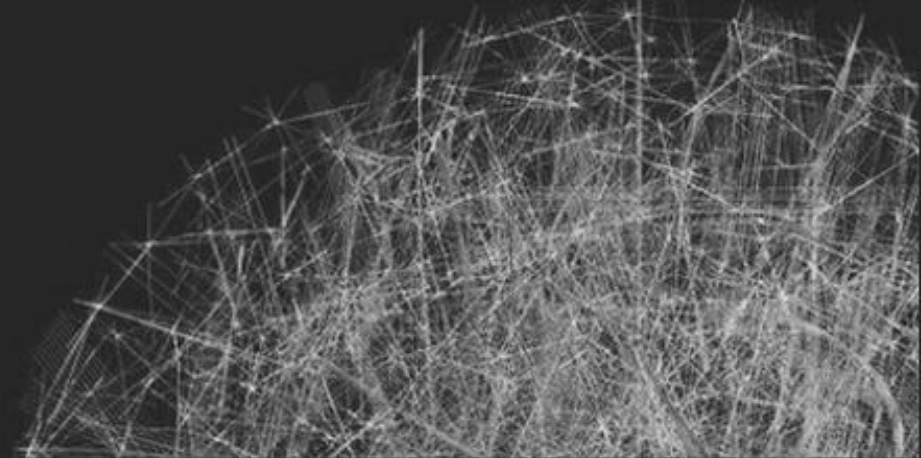


# NeuroMat



## Effect of synaptic plasticity on functional connectivity and global activity of a cortical network model



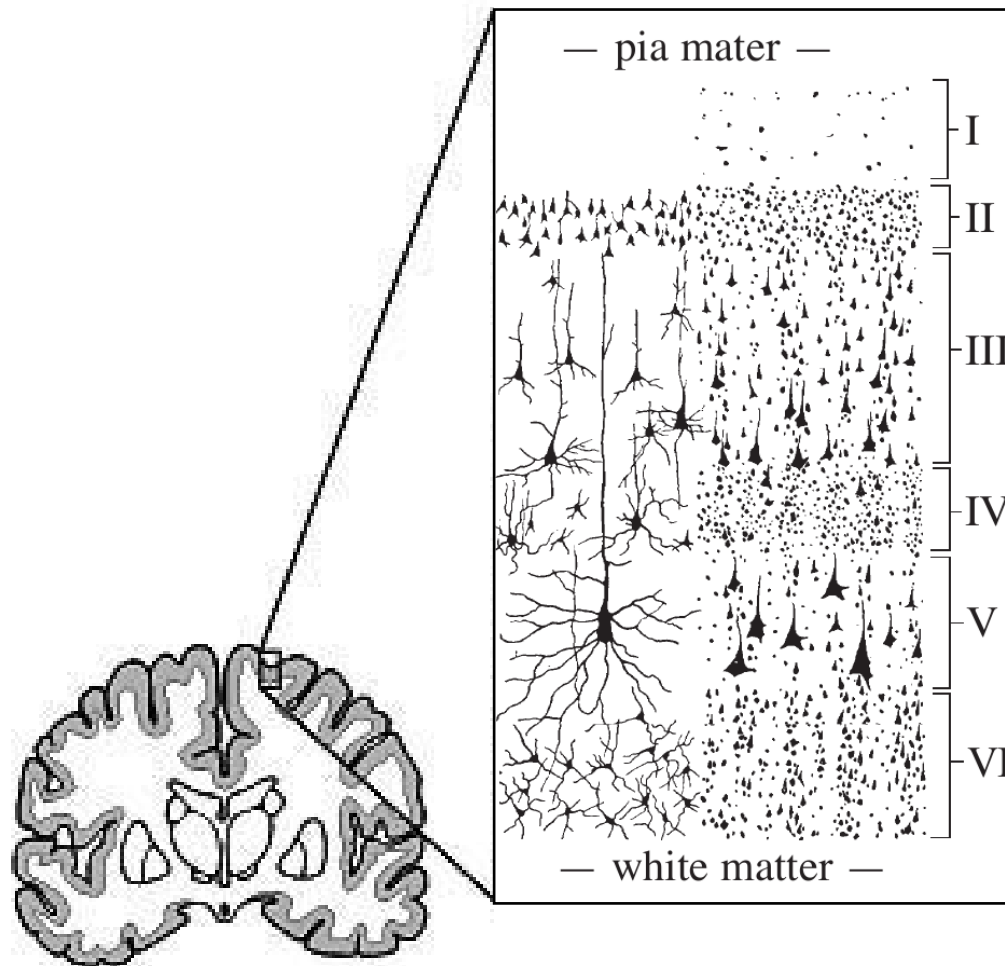
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# Introduction

# Cerebral neocortex – Structural organization



# General background

- Synaptic plasticity is believed to underlie learning and information storage in the brain, as well as neurorecovery after stroke and other brain damage or disease, which are among the main research foci of NeuroMat.
- Long-term plasticity can persist from a scale of seconds to hours or more.



# Goals

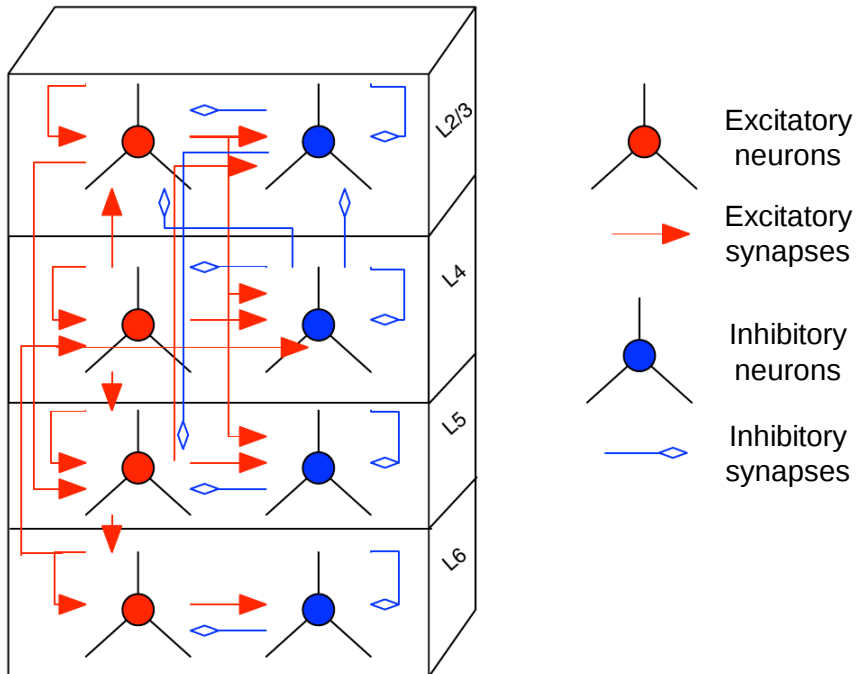
# Goals

- To study the effect of synaptic plasticity rules on the behavior of neural spiking activity patterns in a cortical network model.
- To study the changes in the functional connectivity of the network as disclosed by graph-theoretic measures.



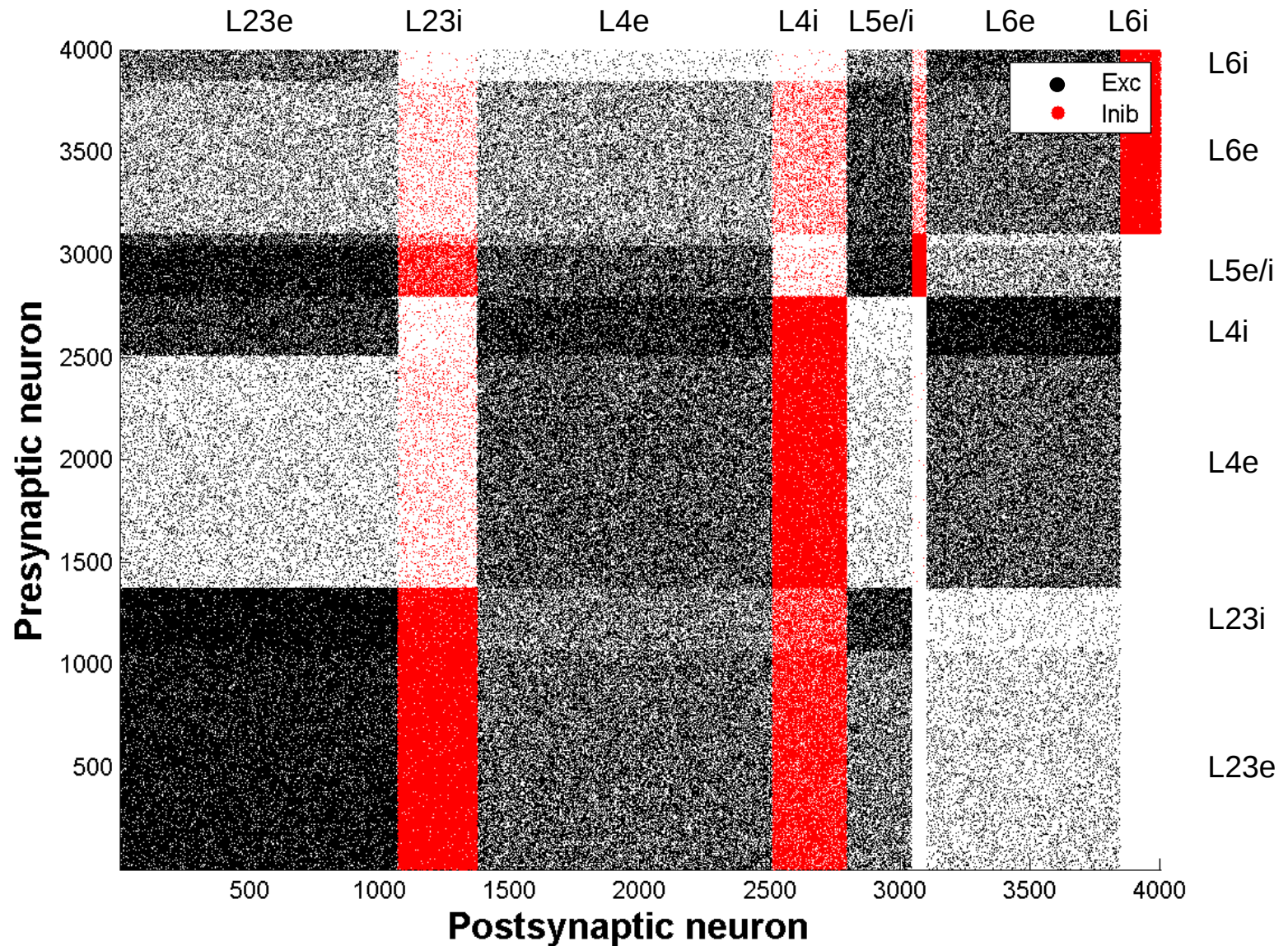
# Methods

# The network



- 4000 neurons;
- Excitatory/inhibitory rate = 4:1;
- ~780.000 synapses.

# Adjacency matrix of topological connections



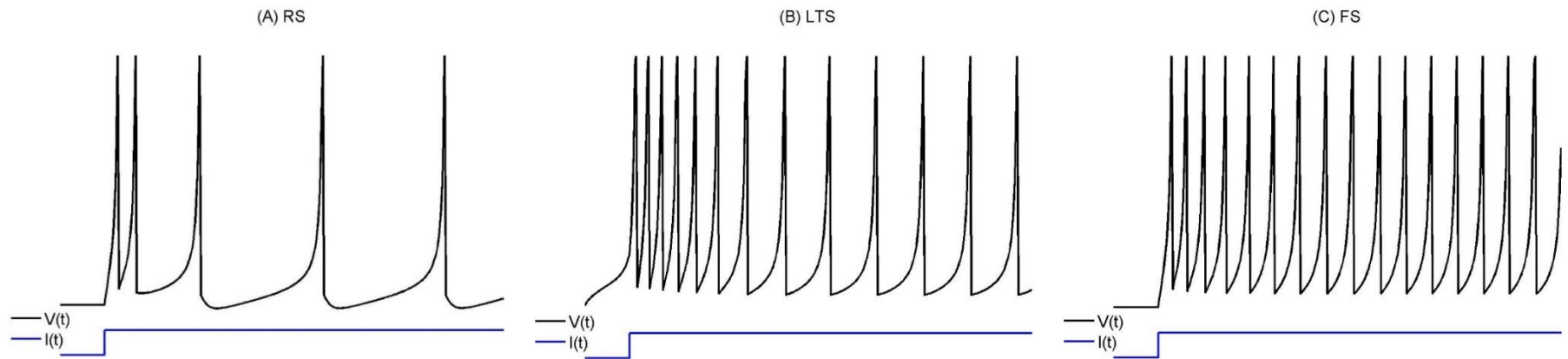
# Neurons: Izhikevich model

$$C \frac{dV}{dt} = k(V - V_{rest})(V - V_{threshold}) - u + I$$

$$\frac{du}{dt} = a\{b(V - V_{rest}) - u\}$$

If  $V \geq 30\text{mV}$ , then:

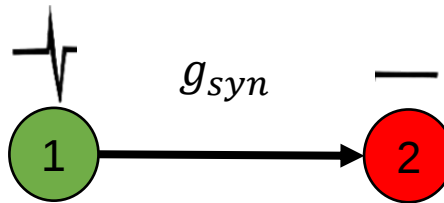
$$\begin{cases} V \leftarrow c \\ u \leftarrow u + d \end{cases}$$



A) regular spiking (RS); B) low threshold spiking (LTS); C) fast spiking (FS).

# Synapse model

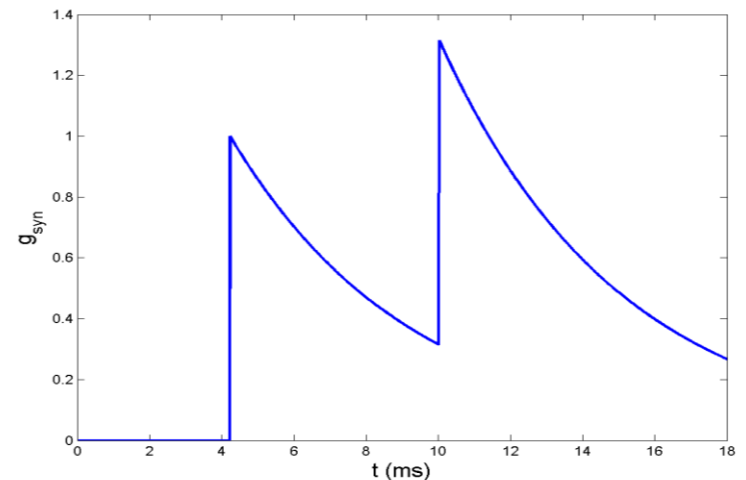
- Event-based model:



If pre- fires a spike,  $g_{syn} \rightarrow g_{syn} + g_{max}$

$$\frac{dg_{syn}}{dt} = -\frac{g_{syn}}{\tau}$$

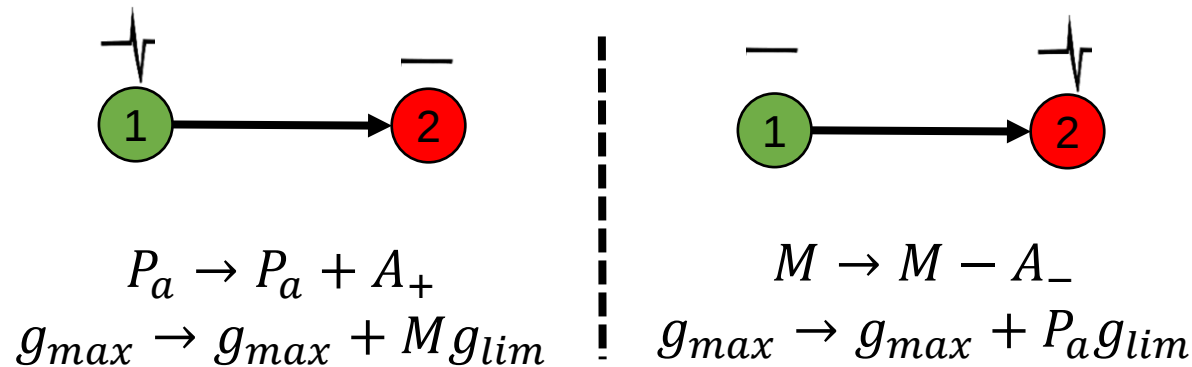
$$I_{syn} = \sum g_{syn}(t)(V(t) - E_{syn});$$



# Asymmetric spike-time-dependent plasticity (STDP) rule

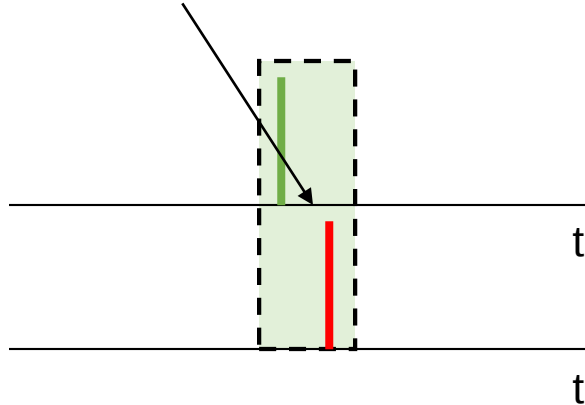
$$\tau_- \frac{dM}{dt} = -M \quad e \quad \tau_+ \frac{dP_a}{dt} = -P_a$$

If:

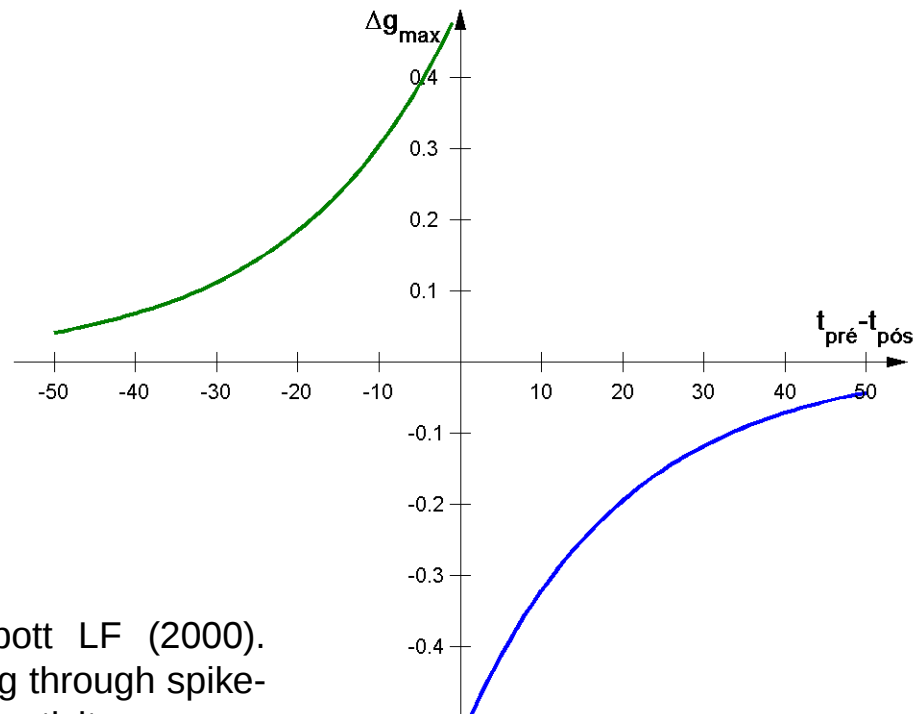
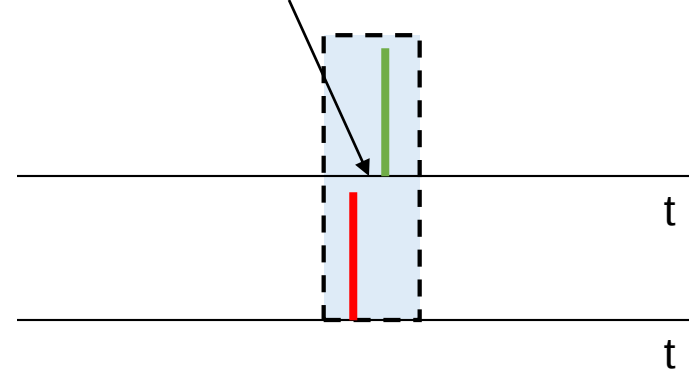


\* Song S, Miller KD, Abbott LF (2000).  
Competitive Hebbian learning through spike-  
timing-dependent synaptic plasticity.

Strengthening



Weakening



\* Song S, Miller KD, Abbott LF (2000).  
Competitive Hebbian learning through spike-  
timing-dependent synaptic plasticity.

# Simulations

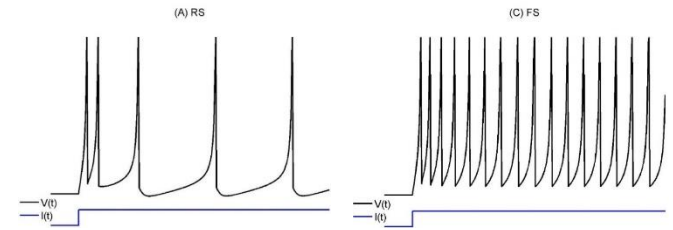
- Duration of simulation: 10000 ms;
- Input: poissonian spike train generated and connected to neurons of layers 4 (L4) and 6 (L6), which is the main input layers of the cortex;
- Electrophysiological classes used:
  - Excitatory: regular spiking (RS);
  - Inhibitory: low-threshold spiking (LTS) or fast spiking (FS).

# Simulations

- Two versions of the model were constructed:

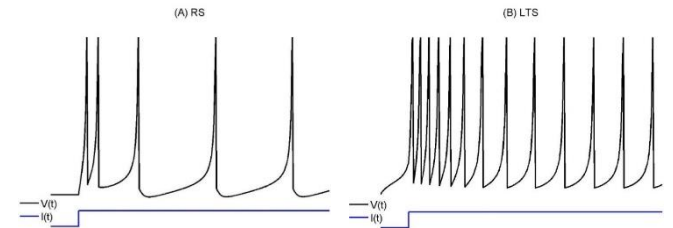
- RS\_FS

- Without synaptic plasticity,
- STDP in excitatory synapses.



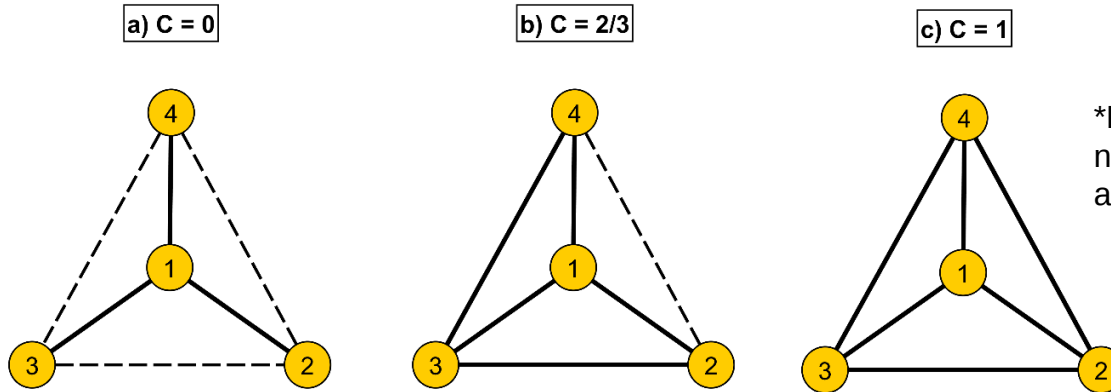
- RS\_LTS

- Without synaptic plasticity,
- STDP in excitatory synapses.



# Measures

## Clustering Coefficient (C)



\*Rubinov M, Sporns O (2010). Complex network measures of brain connectivity: Uses and interpretations.

## Synchrony index (C)

$$X^2(N) = \frac{\sigma_S^2}{\frac{1}{N} \sum_{i=1}^N \sigma_{S_i}^2};$$

$$0 \leq X \leq 1$$

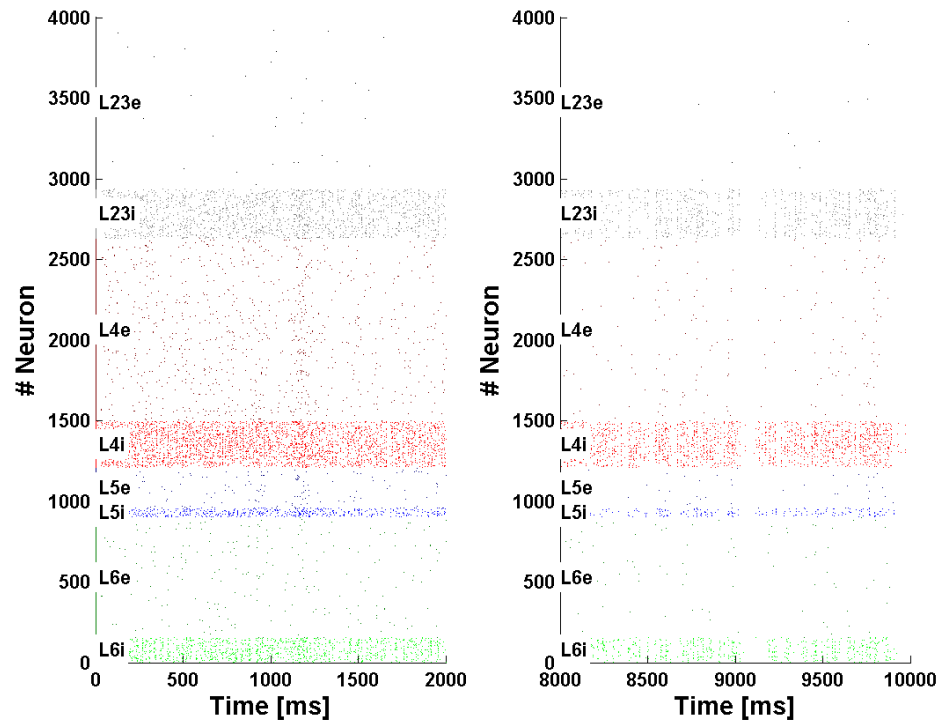
\*Golomb D (2007). Neuronal synchrony measures.



# Preliminary results

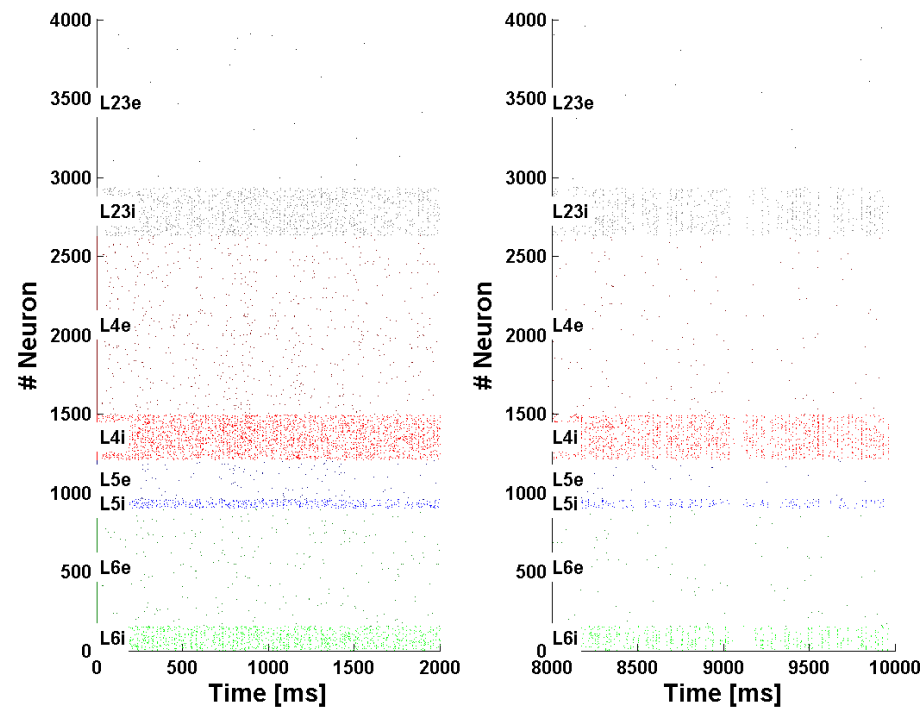
# RS – FS:

No synaptic plasticity



$$f_{\text{mean}} = 0,636 \text{ Hz}; X = 0,026$$

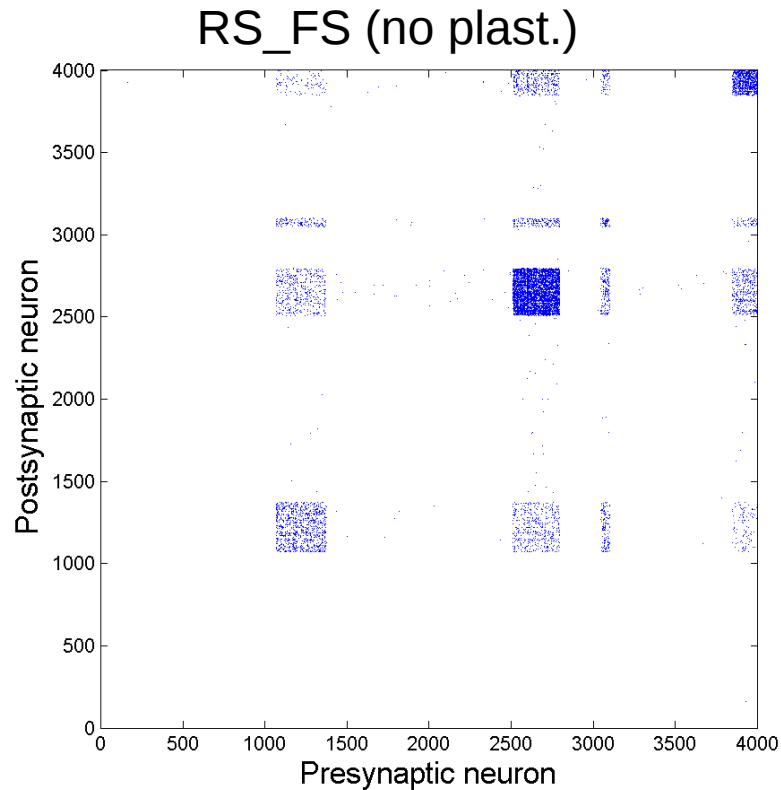
STDP – Excitatory synapses



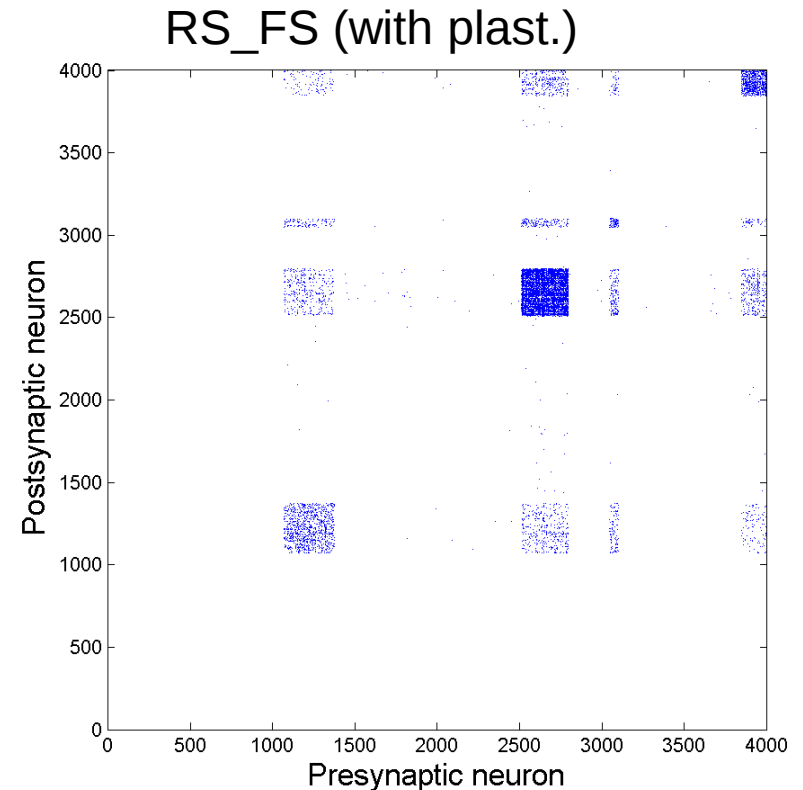
$$f_{\text{mean}} = 0,608 \text{ Hz}; X = 0,027$$

The version with synaptic plasticity has a little higher frequency and synchrony, but not considerable.

# Adjacency matrices of functional connections



$C = 0,021$

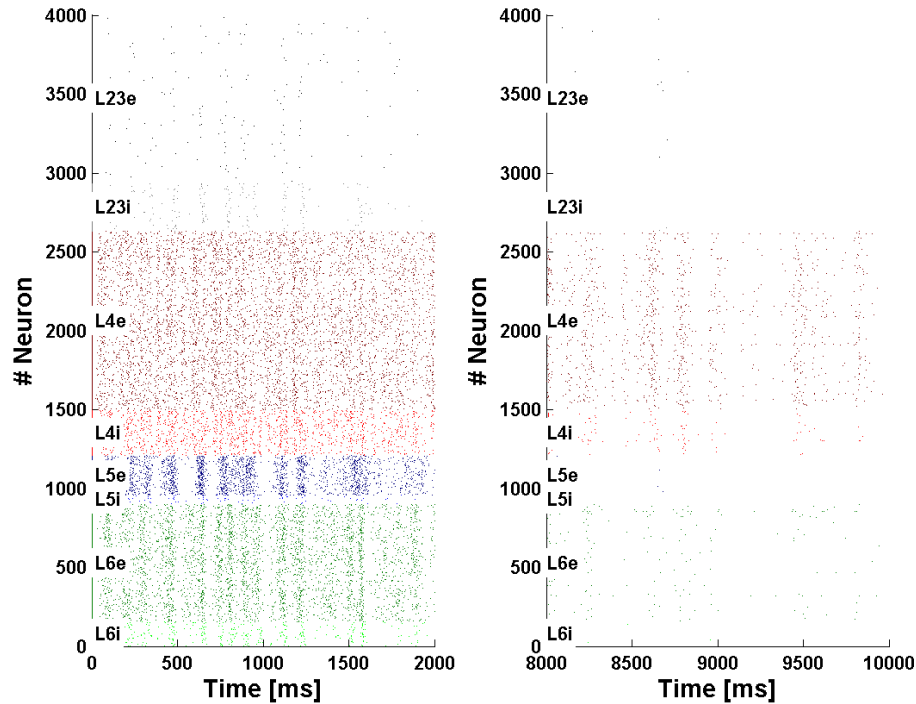


$C = 0,024$

The functional matrices didn't show considerable changes.

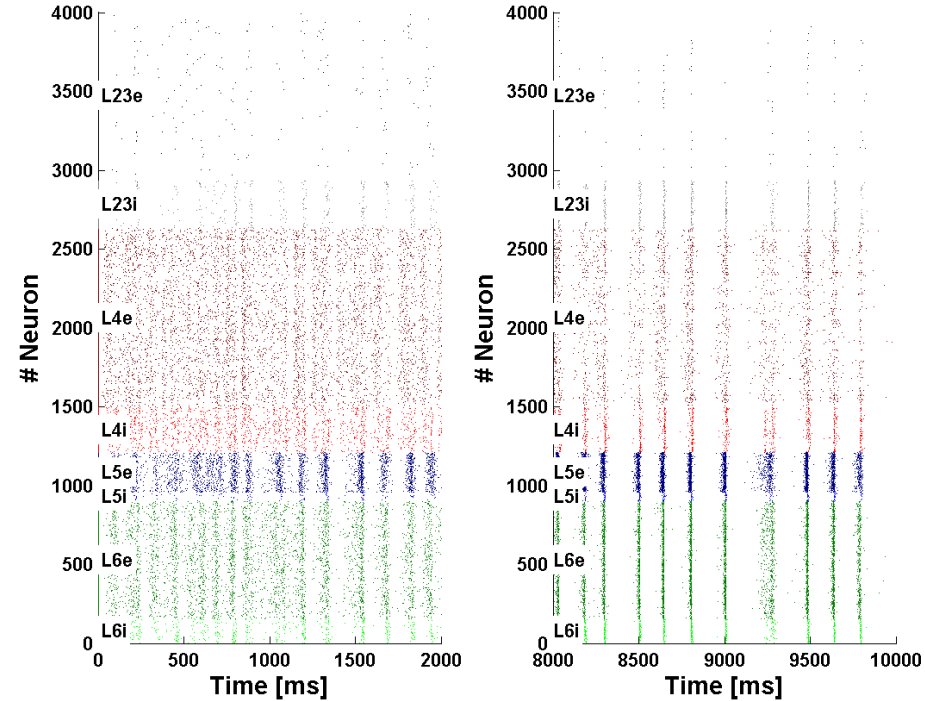
# RS – LTS:

No synaptic plasticity



$$f_{\text{mean}} = 0,744 \text{ Hz}; X = 0,032$$

STDP – Excitatory synapses

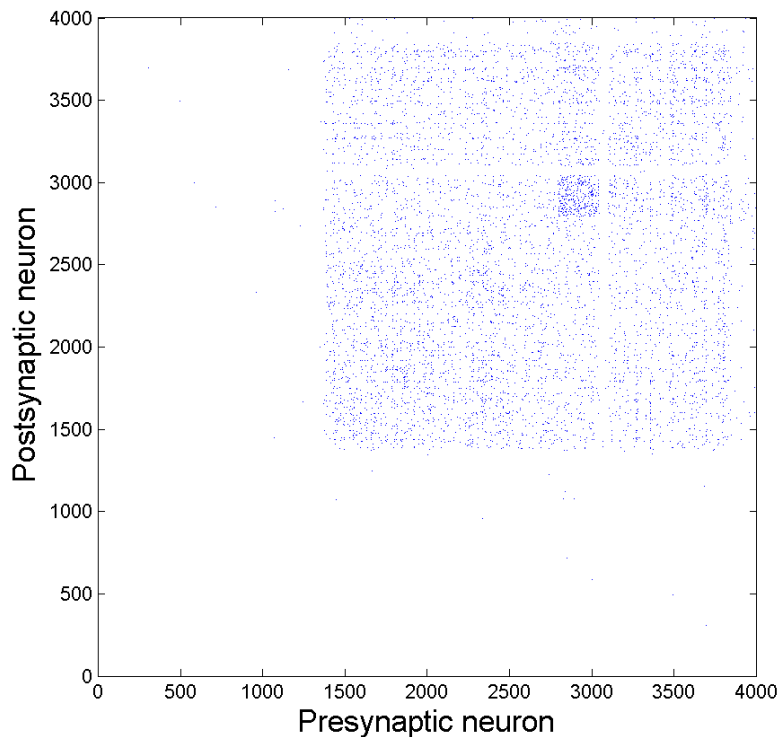


$$f_{\text{mean}} = 1,668 \text{ Hz}; X = 0,067$$

The version with synaptic plasticity has a higher mean frequency. And is possible to see a more synchronous activity of the network.

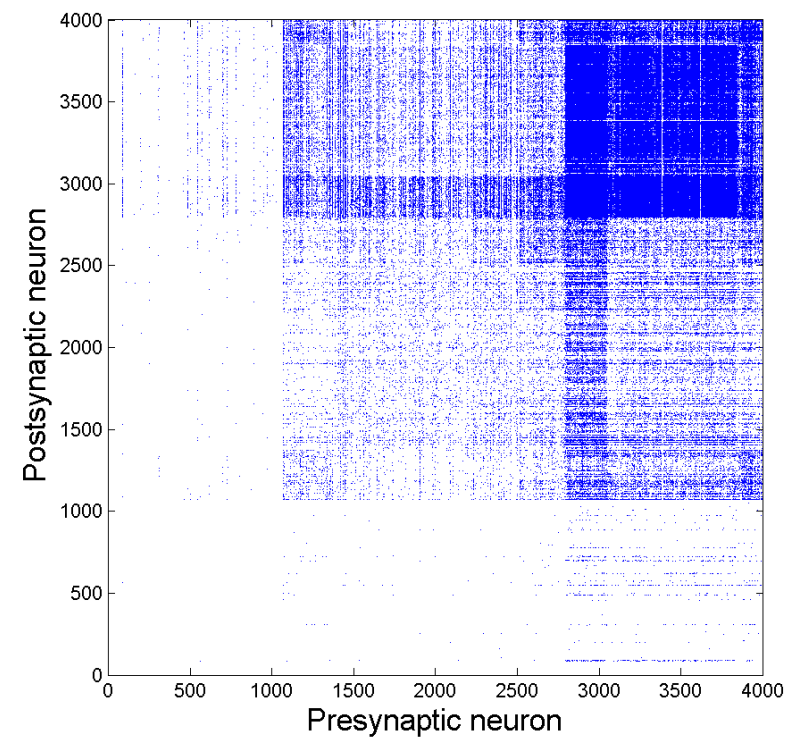
# Adjacency matrices of functional connections

RS\_LTS (no plast.)



$C = 0,014$

RS\_LTS (with plast.)



$C = 0,246$

The formation of clusters of synchronous neural activity was facilitated for the case with synaptic plasticity.

# Partial conclusions

- Even without specific input patterns, STDP may induce changes in the functional connectivity of the cortical network with impact on its global activity;
- The network composition in terms of electrophysiological classes of the neurons has influence on the global activity;



Next steps

# Next steps and future studies

- To characterize the topological and functional matrices of the network with graph-theoretic measures;
- To add STDP on inhibitory synapses and study the effect in this neocortical architecture;
- To induce lesions in the network and study the effects of synaptic plasticity in the reorganization of activity.



# References

# References

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- Izhikevich EM (2007). Dynamical Systems in Neuroscience: The Geometry of Excitability and Bursting. *MIT Press, Cambridge, MA*.
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- Rubinov M, Sporns O (2010). Complex network measures of brain connectivity: Uses and interpretations. *NeuroImage*, 52:1059-1069.