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# Inference of Global States Stability in Cortical Networks

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#### The Default Global State of the Cerebral Cortex [1]

#### Phenomenology

▶ Slow Oscillations (≤1Hz)
 ▷ at the neuron level (membrane potential and firing rate)
 ▷ at the network level



#### **Cortical Slices Recordings**





(extracellular electrical activity: LFP and MUA)

Observed in:

slow wave sleep
deep anaesthesia
deafferentation
cortical slices

#### **Key Features**

Bistability

**Motivation** 

- existence of two attractors:UP state and DOWN state
- Intrinsic fluctuations between these attractors
- Regularity of the UP/DOWN states alternance
- $\rightarrow$  behaves as a relaxation oscillator



Figure 1: Slow oscillations recorded from the frontal cortex of an anaesthetized mouse [1]

## Advantages

- Resilience to perturbances:
   the relaxation-oscillator regime acts as an equilibrium of the network.
- Facilitation of the transition towards more connected, awake-like states.

Figure 2: Cortical slice recording setup: (A) cortex, (B) white matter, (C) infragranular layers' electrodes and (D) supragranular layers' electrodes

### **Experimental Conditions** [2]

- ▶ Pharmacological Modulations
   ▷ adding Carbachol (0.5 µM) + Norepinephrine (50 µM)
  - reducing extracellular Calcium (to 0.8-0.9 mM)
  - ▷ Reducing temperature (to 31-32 °C)
- Electrical Stimuli
  - $\triangleright$  150  $\mu \rm A$  pulses every 10 s at layer 5.
  - $\rightarrow$  Experimental model to explore the transitions from a state of slow oscillations towards a higher complexity state (awake-like asynchronous state).



Figure 3: Schematic of the 16-channel SU-8-based flexible microprobe used for the recordings

Can we detect and characterize other global network states apart from the SO regime?

#### **Cortical Networks Parameters: Literature Review**

In fact, the dynamics of the network states are not fully understood [3]. The stability of such states seems to be strongly influenced by:

- the input stimulus [4]
- the connectivity properties of the network
  - either unshaped or structured in clusters [5]
- the excitatory-inhibitory balance
  [6]
- the network architecture
  - either predominantly feedforward or recurrent [7]
- the kind of noise
  - ▷ intrinsic or extrinsic [8]

**Problem:** Although multiple mechanistic hypotheses have been proposed in models, the current analysis tools do not enable us to discern empirically the dynamics of the network states.

Figure 4: Example of extracellular activity (LFP and logMUA) issued from recordings at two different layers under two different experimental conditions

#### **Our Approach**

#### Aims

To identify the stability of the global network states in isolated cortical networks under experimental manipulations that alter key network parameters (excitability, input, connectivity, etc.).

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- To develop a novel theoretical tool which empirically captures the metastable regions of the network, i.e. transient states that temporarily behave as attractors.

#### Methodology

- With the aid of kernel mean embedding techniques for clustering [9], we will detect the convergence regions of the system.
- By studying how the phase portrait of the system evolves when the slow-oscillation regime is perturbed, we will map the bifurcations or transient states with the network parameters.

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