

## TD6 – Rate Models

### Practical Information



#### TD Assistant

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#### TD Material

[https://github.com/esther-poniatowski/2223\\_UlmM2\\_ThNeuro](https://github.com/esther-poniatowski/2223_UlmM2_ThNeuro)

### Goals of the TD

This TD aims to study the *Ring Model*, originally introduced for orientation coding in the visual cortex. More generally, this model could apply for any network encoding a circular one-dimensional variable.

### Ring Model

The model consists of a network of neurons responsive to a one-dimensional stimulus  $\theta$  which spans the range  $[0, 2\pi]$ . Each neuron is characterized by a preferred stimulus value. Therefore, neurons can be conceptually aligned along a ring, their position being assimilated to their preferred stimulus value.

In the limit of a large number of neurons and with a uniform distribution of preferred stimuli in the range  $[0, 2\pi]$ , the network is assumed to be continuous. Specifically, for each value of  $\theta$ , there exists one neuron preferring this stimulus value. In this view, the neural activity can be written as a continuous function  $m(\theta, t)$ .

Each neuron receives two types of inputs :

- External inputs which depend on the neuron's position :  $h(\theta)$ .
- Recurrent inputs which depend on the activity of the whole network. The connection strength between two neurons is set by the distance between their preferred stimuli  $\theta_1$  and  $\theta_2$ . It is given by :

$$J_0 + J_1 \cos(\theta_1 - \theta_2)$$

The activity of the network evolves according to the following dynamics :

$$\frac{dm(\theta, t)}{dt} = -m(\theta, t) + f[I(\theta, t)] \quad (1)$$

$$f(x > 0) = x \quad (2)$$

$$f(x < 0) = 0 \quad (3)$$

## 1 Input current & Uniform state

- ① In the general case, express the total input current received by a neuron preferring a stimulus value  $\theta$ .

↙ The network is submitted to a constant, uniform and positive external current  $h(\theta) = h_0$ , which is sufficient for inducing a positive, uniform network activity  $m(\theta, t) = m_0$ .

- ② In this particular case, express the current received by each neuron. Deduce the network activity  $m_0$ . How does it depend on the connectivity parameters ?

## 2 Description through order parameters

To determine whether this uniform state is stable, the network's activity is assumed to be perturbed around its equilibrium :

$$m(\theta, t) = m_0 + \delta m(\theta, t)$$

The network's evolution in this context can be studied through two order parameters :

$$M(t) = \frac{1}{2\pi} \int_0^{2\pi} \delta m(\theta', t) d\theta' \quad (4)$$

$$C(t) = \frac{1}{2\pi} \int_0^{2\pi} \delta m(\theta', t) e^{i\theta'} d\theta' \quad (5)$$

The goal will be to obtain a description of the dynamics in terms of the evolution of these two order parameters.

- ③ Interpret the order parameters  $M$  and  $C$ .

The perturbation is assumed to be uniform :

$$\delta m(\theta, t) = \epsilon$$

- ④ Compute the values of  $M$  and  $C$ .

## 3 Bumpy perturbation

Now, the perturbation is assumed to be a small bump centered around the angle  $\phi$  :

$$\delta m(\theta, t) = \epsilon \cos(\theta - \phi) = \epsilon \frac{e^{i(\theta - \phi)} + e^{-i(\theta - \phi)}}{2}$$

- ⑤ Compute the values of  $M$  and  $C$ .
- ⑥ Linearize the dynamics of the activity around  $m_0$  and express it as a function of  $M(t)$  and  $C(t)$ .
- ⑦ Derive the differential equations governing the evolution of the order parameters.
- ⑧ Determine the conditions under which the uniform activity stable. What happens when either of these conditions is not met ?

Consider  $J_0 < 1$ ,  $J_1 < 2$ .

The network is submitted to an external input with weak modulation :

$$h(\theta) = h_0 + \epsilon \cos(\theta) \quad \epsilon \ll 1$$

- ⑨ Determine is the evolution of the profile of activity of the network.
- ⑩ Based on the order parameters, justify that the activity profile at equilibrium can be written under the form :

$$m(\theta, t) = m_0 + m_1 \cos(\theta)$$

- ⑪ Determine conditions on the connectivity parameters so that the network amplifies the input, i.e.

$$\frac{m_1}{m_0} > \frac{\epsilon}{h_0}$$