

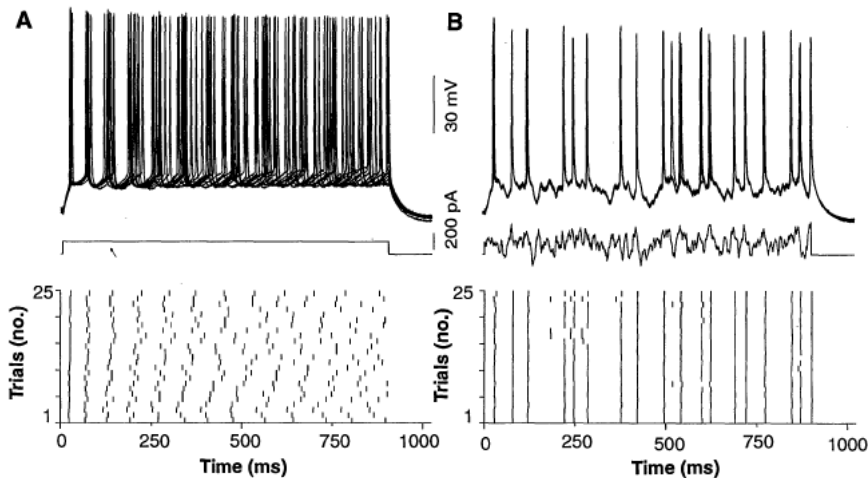
# Robustness of noise-induced synchronization

Ngày 16 tháng 4 năm 2008

- 1 Problem statement
- 2 Modelling “noise”: Stochastic Differential Equations
- 3 The proof
- 4 Limitations of the analysis

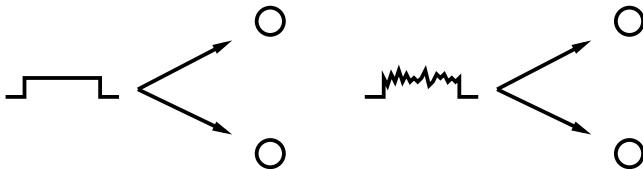
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# Mainen & Sejnowski experiment



[Mainen & Sejnowski, 1995]

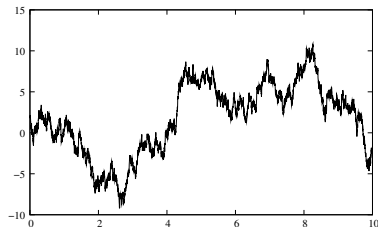
# Synchronization interpretation



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# Random walk and Wiener process

- Random walk (**discrete-time**):  $\mathbf{x}_{t+\Delta t} = \mathbf{x}_t + \xi_t \Delta t$   
where  $(\xi_t)_{t \in \mathbb{N}}$  are Gaussian and mutually independent
- If one is interested in very rapidly varying perturbations,  $\Delta t$  has to be very small
- Wiener process (or Brownian motion) (**continuous-time**): limit of the random walk when  $\Delta t \rightarrow 0$



# Wiener process and “white noise”

- Problem: a Wiener process is **not** differentiable (why?), thus it is not the solution of any ordinary differential equation
- Define formally  $\xi_t$  (“white noise”) = “derivative” of the Wiener process
- Formally:  $W(t) - W(0) = \int_0^t \xi_t dt$  or  $dW/dt = \xi_t$  or  **$dW = \xi_t dt$**



# Stochastic Differential Equations

- We consider processes driven by “white noise”
- We would like to write (but it's not correct, because  $\xi$  **does not** exist)

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}) + \mathbf{g}(\mathbf{x})\xi$$

- In integral form, it may be more correct

$$\mathbf{x}(t) - \mathbf{x}(0) = \int_0^t \mathbf{f}(\mathbf{x})dt + \int_0^t \mathbf{g}(\mathbf{x})dW$$

where the last term is a **Stieltjes** integral against  $W$  (which **does** exist)

- The integral form can also be written in differential form

$$d\mathbf{x} = \mathbf{f}(\mathbf{x})dt + \mathbf{g}(\mathbf{x})dW$$

# Definition of Itô and Stratonovich integrals

- For deterministic function  $\alpha$ , the Stieltjes integral (which generalizes Riemann integrals) against  $\alpha$  is defined as

$$\int_0^T \beta(t) d\alpha = \lim_{N \rightarrow \infty} \sum_1^{N-1} \beta(t_i) [\alpha(t_{i+1}) - \alpha(t_i)]$$

- Thus one can define, by analogy

$$\int_0^T \beta(t) dW = \lim_{N \rightarrow \infty} \sum_1^{N-1} \beta(\mathbf{t}_i) [W(t_{i+1}) - W(t_i)]$$

which is the **Itô integral**

- But one can also define

$$\int_0^T \beta(t) dW = \lim_{N \rightarrow \infty} \sum_1^{N-1} \beta\left(\frac{\mathbf{t}_i + \mathbf{t}_{i+1}}{2}\right) [W(t_{i+1}) - W(t_i)]$$

which is the **Stratonovich integral**

# Independence properties

- The two above definitions lead to the same result in the **deterministic** case (probably,  $C^1$  is required)
- But there are differences in the **stochastic** case:
- Since  $\beta(t_i)$  (present) is **independent** of  $W(t_{i+1}) - W(t_i)$  (future), one has, for Itô integrals

$$\mathbb{E}(\beta(t_i)[W(t_{i+1}) - W(t_i)]) = \mathbb{E}(\beta(t_i))\mathbb{E}(W(t_{i+1}) - W(t_i)) = 0$$

leading to

$$\mathbb{E}\left(\int_0^T \beta(t) dW\right) = 0$$

This explains Teramae claim “In the Itô formulation, [...], the correlation between  $\phi$  and  $\xi$  vanishes”

# Variable transformation

- The variable transformation (“changement de variable” in French) formula is also different for Itô and Stratonovich integrals
- Consider the function  $y(x)$ . In the deterministic case, one has, for instance

$$\frac{dy}{dt} = \frac{\partial y}{\partial x} \cdot \frac{dx}{dt} \quad \text{or} \quad dy = \frac{\partial y}{\partial x} dx$$

- The same rule is valid for Stratonovich integrals (Teramae's “conventional variable transformation”): if

$$dx = f(x)dt + g(x)dW$$

then

$$dy = \frac{\partial y}{\partial x} (f(x)dt + g(x)dW)$$

# Itô's formula for variable transformation

- Consider the Itô SDE

$$dx = f(x)dt + g(x)dW$$

- Then for a function  $y(x)$ , one has (Itô's formula)

$$dy = \left( \frac{\partial y}{\partial x} f(x) + \frac{1}{2} \frac{\partial^2 y}{\partial x^2} g(x)^2 \right) dt + \frac{\partial y}{\partial x} g(x) dW$$

- This will explain Teramae's "the disappeared correlation is compensated by the new extra drift term  $Z'DZ$ "

# Outline

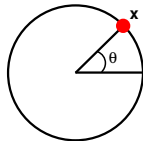
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# Phase reduction

- Consider the system  $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x})$  which has a **limit-cycle**.
- We would like to find a **phase** variable  $\phi(\mathbf{x})$  such that:

$$\frac{d\phi}{dt} = \omega \quad \omega = \text{constant}$$

Example: a mobile travelling on a circle with constant velocity



- General case (using the chain rule):

$$\frac{d\phi}{dt} = \frac{\partial \phi}{\partial \mathbf{x}}(\mathbf{x}) \cdot \frac{d\mathbf{x}}{dt} = \frac{\partial \phi}{\partial \mathbf{x}}(\mathbf{x}) \cdot \mathbf{f}(\mathbf{x}) = \omega$$

- One then has to solve the above **PDE** to find  $\phi$

# Phase reduction(continued)

- Consider now a **small** perturbation  $\xi$

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}) + \xi$$

- Then the equation on the phase becomes

$$\frac{d\phi}{dt} = \frac{\partial\phi}{\partial\mathbf{x}}(\mathbf{x}) \cdot \mathbf{f}(\mathbf{x}) + \frac{\partial\phi}{\partial\mathbf{x}}(\mathbf{x}) \cdot \xi = \omega + \frac{\partial\phi}{\partial\mathbf{x}}(\mathbf{x}) \cdot \xi$$

- This can be converted into a  $\phi$ -only equation using some approximations

$$\frac{d\phi}{dt} = \omega + Z(\phi)\xi$$

- This was equation (2) in [Teramae & Tanaka, 2004]



# Stratonovich to Itô switch

- Actually, the authors could have done everything in Itô!
- Let us compute the phase equation obtained previously but using now Itô's formula (with  $D = \frac{1}{2}g^2$ )

$$d\phi = \left( \frac{\partial \phi}{\partial \mathbf{x}} \mathbf{f}(\mathbf{x}) + D \frac{\partial^2 \phi}{\partial \mathbf{x}^2} \right) dt + \frac{\partial \phi}{\partial \mathbf{x}} dW$$

- As above, let  $Z(\phi) = \frac{\partial \phi}{\partial \mathbf{x}}$ . Then

$$\frac{\partial^2 \phi}{\partial \mathbf{x}^2} = \frac{\partial}{\partial \mathbf{x}} Z(\phi) = \frac{\partial Z}{\partial \phi} \frac{\partial \phi}{\partial \mathbf{x}} = Z'(\phi) Z(\phi)$$

- Thus

$$d\phi = (\omega + Z'(\phi) D Z(\phi)) dt + Z(\phi) dW$$

which is equation (3) (after formal division by  $dt$ )

- Consider  $\dot{\phi}_1 = f(\phi_1)$  and  $\dot{\phi}_2 = f(\phi_2)$
- Then (using the Taylor expansion assuming  $\phi_1 - \phi_2$  very small)

$$\begin{aligned}\dot{\phi}_1 - \dot{\phi}_2 &= f(\phi_1) - f(\phi_2) \\ &= f(\phi_1) - (f(\phi_1) + (\phi_2 - \phi_1)f'(\phi_1)) \\ &= f'(\phi_1)(\phi_1 - \phi_2)\end{aligned}$$

- This explains equation (4) if we set  $\psi = \phi_1 - \phi_2$

- Consider two infinitesimally close trajectories. The Lyapunov exponent  $\lambda$  verifies (intuitively)

$$\|\delta\phi(t)\| \simeq e^{\lambda t} \|\delta\phi_0\|$$

- If  $\lambda > 0$ , then nearby trajectories **diverge** = instability
  - If  $\lambda < 0$ , then nearby the trajectories **converge** = stability
- (Remark: if a system is **contracting**, then  $\lambda < 0$ )

# Lyapunov exponent (continued)

- Let us manipulate the above expression:

$$e^{\lambda t} \simeq \frac{\|\delta\phi(t)\|}{\|\delta\phi_0\|} \quad \lambda t = \ln \left( \frac{\|\delta\phi(t)\|}{\|\delta\phi_0\|} \right) \quad \lambda = \frac{1}{t} \ln \left( \frac{\|\delta\phi(t)\|}{\|\delta\phi_0\|} \right)$$

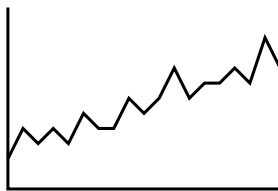
- Actually, the Lyapunov exponent is defined as (because we are interested in long-time behaviour)

$$\lambda = \lim_{t \rightarrow \infty} \frac{1}{t} \ln \left( \frac{\|\delta\phi(t)\|}{\|\delta\phi_0\|} \right)$$

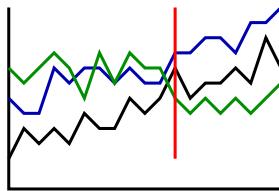
# Ergodic hypothesis

- Consider a stochastic process  $\mathbf{x}(\omega, t)$
- Any physicist knows that (**ergodic hypothesis**):  $\forall \omega, t$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T \mathbf{x}(\omega, t') dt' = \int_{\Omega} \mathbf{x}(\omega', t) d\omega' = \mathbb{E}(\mathbf{x}(t))$$



Time average



Ensemble average

# Ergodic hypothesis (continued)

- Remark now that  $y$  in equation (5) is defined as

$$y = \ln(\psi) = \ln(\phi_1 - \phi_2) = \ln(\delta\phi)$$

- Remark that

$$\int_0^T \dot{y} = y(T) - y(0) = \ln(\delta\phi(T)) - \ln(\delta\phi(0)) = \ln \frac{\delta\phi(T)}{\delta\phi(0)}$$

thus

$$\frac{1}{T} \lim_{T \rightarrow \infty} \int_0^T \dot{y} = \frac{1}{T} \lim_{T \rightarrow \infty} \ln \frac{\delta\phi(T)}{\delta\phi(0)} = \lambda$$

By the ergodic hypothesis, we then have

$$\lambda = \mathbb{E}(\dot{y})$$

which explains the first line in equation (6).

# Probability density

- Let  $P(\phi, t)$  denotes the time-dependent probability density of the random variable  $\phi \in [0, 2\pi]$
- $P$  is constant intuitively means that  $\phi$  has equal probability of being anywhere in  $[0, 2\pi]$

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# Some limitations

- A lot of unproven statements (ergodicity, uniform distribution of  $\phi$  in steady state,...). Perhaps those statements are evident for physicists!
- There is a mistake in the computation of the phase equation, as pointed out by [Yoshimura & Arai, 2007] (Thank you, **Francis!**). However, this mistake does not alter the result.
- The analysis is only valid when  $Z$  is continuously differentiable up to the second-order, which is not verified for e.g. resetting neuron models (Integrate and Fire, Izhikevich,...)