

15th AIMS Conference, Athens

Concentration results for time-dependent SPDEs with Gaussian and fractional noise

Nils Berglund

Institut Denis Poisson, University of Orléans, France



8 July 2026

Collab: Rita Nader (Rennes), Barbara Gentz (Bielefeld) and Alexandra Blessing (Konstanz)



Project
PERISTOCH



View from Lycabettus Hill

Warm-up: Slowly time-dependent 1D SDE

$$dx_t = f(\varepsilon t, x_t) dt + \sigma dW_t$$

- On slow time scale $\varepsilon t \rightarrow t$:

$$dx_t = \frac{1}{\varepsilon} f(t, x_t) dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_t$$

- Stable equilibrium branch $x^*(t)$: $f(t, x^*(t)) = 0$, $\partial_x f(t, x^*(t)) < 0$
 $\sigma = 0 \Rightarrow \exists$ slow solution $\bar{x}(t) = x^*(t) + \mathcal{O}(\varepsilon)$ [Tihonov, Fenichel]
- $\sigma > 0$: write $x_t = \bar{x}(t) + \xi_t$ and Taylor-expand:

$$d\xi_t = \frac{1}{\varepsilon} \left[\bar{a}(t)\xi_t + \underbrace{b(t, \xi_t)}_{=\mathcal{O}(\xi_t^2)} \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_t$$

where $\bar{a}(t) = \partial_x f(t, \bar{x}(t)) = \partial_x f(t, x^*(t)) + \mathcal{O}(\varepsilon) < 0$

- Variations of constants, if $\xi_0 = 0$ and $\bar{\alpha}(t, s) = \int_s^t \bar{a}(u) du$:

$$\xi_t = \underbrace{\frac{\sigma}{\sqrt{\varepsilon}} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} dW_s}_{\xi_t^0: \text{sol of linearised system}} + \underbrace{\frac{1}{\varepsilon} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} b(s, \xi_s) ds}_{\text{treat as a perturbation}}$$

Warm-up: Slowly time-dependent 1D SDE

$$dx_t = f(\varepsilon t, x_t) dt + \sigma dW_t$$

- ▷ On slow time scale $\varepsilon t \rightarrow t$:

$$dx_t = \frac{1}{\varepsilon} f(t, x_t) dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_t$$

- ▷ Stable equilibrium branch $x^*(t)$: $f(t, x^*(t)) = 0$, $\partial_x f(t, x^*(t)) < 0$
 $\sigma = 0 \Rightarrow \exists$ slow solution $\bar{x}(t) = x^*(t) + \mathcal{O}(\varepsilon)$ [Tihonov, Fenichel]
- ▷ $\sigma > 0$: write $x_t = \bar{x}(t) + \xi_t$ and Taylor-expand:

$$d\xi_t = \frac{1}{\varepsilon} \left[\bar{a}(t)\xi_t + \underbrace{b(t, \xi_t)}_{=\mathcal{O}(\xi_t^2)} \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_t$$

where $\bar{a}(t) = \partial_x f(t, \bar{x}(t)) = \partial_x f(t, x^*(t)) + \mathcal{O}(\varepsilon) < 0$

- ▷ Variations of constants, if $\xi_0 = 0$ and $\bar{\alpha}(t, s) = \int_s^t \bar{a}(u) du$:

$$\xi_t = \underbrace{\frac{\sigma}{\sqrt{\varepsilon}} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} dW_s}_{\xi_t^0: \text{sol of linearised system}} + \underbrace{\frac{1}{\varepsilon} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} b(s, \xi_s) ds}_{\text{treat as a perturbation}}$$

Warm-up: Slowly time-dependent 1D SDE

$$dx_t = f(\varepsilon t, x_t) dt + \sigma dW_t$$

- ▷ On slow time scale $\varepsilon t \rightarrow t$:

$$dx_t = \frac{1}{\varepsilon} f(t, x_t) dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_t$$

- ▷ Stable equilibrium branch $x^*(t)$: $f(t, x^*(t)) = 0$, $\partial_x f(t, x^*(t)) < 0$
 $\sigma = 0 \Rightarrow \exists$ slow solution $\bar{x}(t) = x^*(t) + \mathcal{O}(\varepsilon)$ [Tihonov, Fenichel]
- ▷ $\sigma > 0$: write $x_t = \bar{x}(t) + \xi_t$ and Taylor-expand:

$$d\xi_t = \frac{1}{\varepsilon} \left[\bar{a}(t)\xi_t + \underbrace{b(t, \xi_t)}_{=\mathcal{O}(\xi_t^2)} \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_t$$

where $\bar{a}(t) = \partial_x f(t, \bar{x}(t)) = \partial_x f(t, x^*(t)) + \mathcal{O}(\varepsilon) < 0$

- ▷ Variations of constants, if $\xi_0 = 0$ and $\bar{\alpha}(t, s) = \int_s^t \bar{a}(u) du$:

$$\xi_t = \underbrace{\frac{\sigma}{\sqrt{\varepsilon}} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} dW_s}_{\xi_t^0: \text{sol of linearised system}} + \underbrace{\frac{1}{\varepsilon} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} b(s, \xi_s) ds}_{\text{treat as a perturbation}}$$

Warm-up: Slowly time-dependent 1D SDE

$$dx_t = f(\varepsilon t, x_t) dt + \sigma dW_t$$

- ▷ On slow time scale $\varepsilon t \rightarrow t$:

$$dx_t = \frac{1}{\varepsilon} f(t, x_t) dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_t$$

- ▷ Stable equilibrium branch $x^*(t)$: $f(t, x^*(t)) = 0$, $\partial_x f(t, x^*(t)) < 0$
 $\sigma = 0 \Rightarrow \exists$ slow solution $\bar{x}(t) = x^*(t) + \mathcal{O}(\varepsilon)$ [Tihonov, Fenichel]
- ▷ $\sigma > 0$: write $x_t = \bar{x}(t) + \xi_t$ and Taylor-expand:

$$d\xi_t = \frac{1}{\varepsilon} \left[\bar{a}(t)\xi_t + \underbrace{b(t, \xi_t)}_{=\mathcal{O}(\xi_t^2)} \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_t$$

where $\bar{a}(t) = \partial_x f(t, \bar{x}(t)) = \partial_x f(t, x^*(t)) + \mathcal{O}(\varepsilon) < 0$

- ▷ Variations of constants, if $\xi_0 = 0$ and $\bar{\alpha}(t, s) = \int_s^t \bar{a}(u) du$:

$$\xi_t = \underbrace{\frac{\sigma}{\sqrt{\varepsilon}} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} dW_s}_{\xi_t^0: \text{sol of linearised system}} + \underbrace{\frac{1}{\varepsilon} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} b(s, \xi_s) ds}_{\text{treat as a perturbation}}$$

Warm-up: Slowly time-dependent 1D SDE

$$dx_t = f(\varepsilon t, x_t) dt + \sigma dW_t$$

- On slow time scale $\varepsilon t \rightarrow t$:

$$dx_t = \frac{1}{\varepsilon} f(t, x_t) dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_t$$

- Stable equilibrium branch $x^*(t)$: $f(t, x^*(t)) = 0$, $\partial_x f(t, x^*(t)) < 0$
 $\sigma = 0 \Rightarrow \exists$ slow solution $\bar{x}(t) = x^*(t) + \mathcal{O}(\varepsilon)$ [Tihonov, Fenichel]
- $\sigma > 0$: write $x_t = \bar{x}(t) + \xi_t$ and Taylor-expand:

$$d\xi_t = \frac{1}{\varepsilon} \left[\bar{a}(t)\xi_t + \underbrace{b(t, \xi_t)}_{=\mathcal{O}(\xi_t^2)} \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_t$$

where $\bar{a}(t) = \partial_x f(t, \bar{x}(t)) = \partial_x f(t, x^*(t)) + \mathcal{O}(\varepsilon) < 0$

- Variations of constants, if $\xi_0 = 0$ and $\bar{\alpha}(t, s) = \int_s^t \bar{a}(u) du$:

$$\xi_t = \underbrace{\frac{\sigma}{\sqrt{\varepsilon}} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} dW_s}_{\xi_t^0: \text{sol of linearised system}} + \underbrace{\frac{1}{\varepsilon} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} b(s, \xi_s) ds}_{\text{treat as a perturbation}}$$

Concentration estimate for linearised process

Properties of OU-like process $\xi_t^0 = \frac{\sigma}{\sqrt{\varepsilon}} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} dW_s$:

- ▷ Gaussian process, $\mathbb{E}[\xi_t^0] = 0$, $\text{Var}(\xi_t^0) = \frac{\sigma^2}{\varepsilon} \int_0^t e^{2\bar{\alpha}(t,s)/\varepsilon} ds$
- ▷ Confidence interval: $\mathbb{P}\{|\xi_t^0| > \frac{h}{\sigma} \sqrt{\text{Var}(\xi_t^0)}\} = \mathcal{O}(e^{-h^2/2\sigma^2})$
- ▷ $\sigma^{-2} \text{Var}(\xi_t^0)$ satisfies ODE $\varepsilon \dot{v} = 2\bar{a}(t)v + 1$

Lemma [B & Gentz, PTRF 2002]

$\bar{v}(t)$ solution of ODE bounded away from 0: $\bar{v}(t) = \frac{1}{-2\bar{a}(t)} + \mathcal{O}(\varepsilon)$

$$\mathbb{P}\left\{\sup_{0 \leq s \leq t} \frac{|\xi_s^0|}{\sqrt{\bar{v}(s)}} > h\right\} = C_0(t, \varepsilon) e^{-h^2/2\sigma^2}$$

where $C_0(t, \varepsilon) = \sqrt{\frac{2}{\pi} \frac{1}{\varepsilon} \left| \int_0^t \bar{a}(s) ds \right|} \frac{h}{\sigma} [1 + \mathcal{O}(\varepsilon + \frac{t}{\varepsilon} e^{-h^2/\sigma^2})]$

Proof based on Doob's submartingale inequality and partition of $[0, t]$

Concentration estimate for linearised process

Properties of OU-like process $\xi_t^0 = \frac{\sigma}{\sqrt{\varepsilon}} \int_0^t e^{\bar{\alpha}(t,s)/\varepsilon} dW_s$:

- ▷ Gaussian process, $\mathbb{E}[\xi_t^0] = 0$, $\text{Var}(\xi_t^0) = \frac{\sigma^2}{\varepsilon} \int_0^t e^{2\bar{\alpha}(t,s)/\varepsilon} ds$
- ▷ Confidence interval: $\mathbb{P}\{|\xi_t^0| > \frac{h}{\sigma} \sqrt{\text{Var}(\xi_t^0)}\} = \mathcal{O}(e^{-h^2/2\sigma^2})$
- ▷ $\sigma^{-2} \text{Var}(\xi_t^0)$ satisfies ODE $\varepsilon \dot{v} = 2\bar{a}(t)v + 1$

Lemma [B & Gentz, PTRF 2002]

$\bar{v}(t)$ solution of ODE bounded away from 0: $\bar{v}(t) = \frac{1}{-2\bar{a}(t)} + \mathcal{O}(\varepsilon)$

$$\mathbb{P}\left\{\sup_{0 \leq s \leq t} \frac{|\xi_s^0|}{\sqrt{\bar{v}(s)}} > h\right\} = C_0(t, \varepsilon) e^{-h^2/2\sigma^2}$$

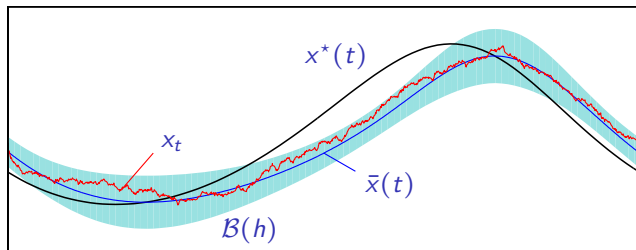
where $C_0(t, \varepsilon) = \sqrt{\frac{2}{\pi} \frac{1}{\varepsilon}} \left| \int_0^t \bar{a}(s) ds \right| \frac{h}{\sigma} [1 + \mathcal{O}(\varepsilon + \frac{t}{\varepsilon} e^{-h^2/\sigma^2})]$

Proof based on Doob's submartingale inequality and partition of $[0, t]$

Concentration estimate for nonlinear process

Nonlinear equation: $d\xi_t = \frac{1}{\varepsilon} [\bar{a}(t)\xi_t + b(t, \xi_t)] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_t$

Confidence strip: $\mathcal{B}(h) = \{|\xi| \leq h\sqrt{\bar{v}(t)} \ \forall t\} = \{|x - \bar{x}(t)| \leq h\sqrt{\bar{v}(t)} \ \forall t\}$



Theorem [B & Gentz, PTRF 2002]

$$C(t, \varepsilon) e^{-\kappa_- h^2 / 2\sigma^2} \leq \mathbb{P}\{\text{leaving } \mathcal{B}(h) \text{ before time } t\} \leq C(t, \varepsilon) e^{-\kappa_+ h^2 / 2\sigma^2}$$

where $\kappa_{\pm} = 1 \mp \mathcal{O}(h)$ and $C(t, \varepsilon) = C_0(t, \varepsilon)[1 + \mathcal{O}(h)]$ (requires $h \leq h_0$)

SPDE on $\mathbb{T}^1$: stable case

$$d\phi(t, x) = \frac{1}{\varepsilon} [\Delta\phi(t, x) + f(t, \phi(t, x))] dt + \frac{\sigma}{\sqrt{\varepsilon}} \underbrace{dW(t, x)}_{\text{cylindrical Wiener process}}$$

- ▷ $f(t, \phi^*(t)) = 0$ for all $t \in I = [0, T]$
- ▷ $a(t) = \partial_\phi f(t, \phi^*(t)) \leq -a_- < 0$ for all $t \in I$

In deterministic case $\sigma = 0$: $\exists$ particular solution $\bar{\phi}(t, x)$ such that

$$\|\bar{\phi}(t, \cdot) - \phi^*(t)e_0\|_{H^1} \leq C\varepsilon \quad \forall t \in I$$

Theorem [B & Nader, SPDE 2023]

Fix $s < \frac{1}{2}$, and let $\mathcal{B}(h) = \{(t, \phi) : t \in I, \|\phi - \bar{\phi}(t, \cdot)\|_{H^s} < h\}$

For any $\nu > 0$

$$\mathbb{P}\{\text{leaving } \mathcal{B}(h) \text{ before time } t\} \leq C(t, \varepsilon, s) \exp\left\{-\kappa \frac{h^2}{\sigma^2} \left[1 - \mathcal{O}\left(\frac{h}{\varepsilon^\nu}\right)\right]\right\}$$

holds for some $\kappa > 0$, $h = \mathcal{O}(\varepsilon^\nu)$ and $C(t, \varepsilon, s) = \mathcal{O}(t/\varepsilon)$.

SPDE on $\mathbb{T}^1$: stable case

$$d\phi(t, x) = \frac{1}{\varepsilon} [\Delta\phi(t, x) + f(t, \phi(t, x))] dt + \frac{\sigma}{\sqrt{\varepsilon}} \underbrace{dW(t, x)}_{\text{cylindrical Wiener process}}$$

- ▷ $f(t, \phi^*(t)) = 0$ for all $t \in I = [0, T]$
- ▷ $a(t) = \partial_\phi f(t, \phi^*(t)) \leq -a_- < 0$ for all $t \in I$

In deterministic case $\sigma = 0$: $\exists$ particular solution $\bar{\phi}(t, x)$ such that

$$\|\bar{\phi}(t, \cdot) - \phi^*(t)e_0\|_{H^1} \leq C\varepsilon \quad \forall t \in I$$

Theorem [B & Nader, SPDE 2023]

Fix $s < \frac{1}{2}$, and let $\mathcal{B}(h) = \{(t, \phi) : t \in I, \|\phi - \bar{\phi}(t, \cdot)\|_{H^s} < h\}$

For any $\nu > 0$

$$\mathbb{P}\{\text{leaving } \mathcal{B}(h) \text{ before time } t\} \leq C(t, \varepsilon, s) \exp\left\{-\kappa \frac{h^2}{\sigma^2} \left[1 - \mathcal{O}\left(\frac{h}{\varepsilon^\nu}\right)\right]\right\}$$

holds for some $\kappa > 0$, $h = \mathcal{O}(\varepsilon^\nu)$ and $C(t, \varepsilon, s) = \mathcal{O}(t/\varepsilon)$.

Ideas of proof

- ▷ $\phi(x) = \sum_{k \in \mathbb{Z}} \phi_k e_k(x) \Rightarrow \|\phi\|_{H^s}^2 = \sum_{k \in \mathbb{Z}} \langle k \rangle^{2s} \phi_k^2, \quad \langle k \rangle = \sqrt{1 + k^2}$
- ▷ Deterministic case: $\psi = \phi - \phi^* e_0$, $\|\psi\|_{H^1}^2$ is a **Lyapunov** function

Ideas of proof

▷ $\phi(x) = \sum_{k \in \mathbb{Z}} \phi_k e_k(x) \Rightarrow \|\phi\|_{H^s}^2 = \sum_{k \in \mathbb{Z}} \langle k \rangle^{2s} \phi_k^2, \quad \langle k \rangle = \sqrt{1 + k^2}$

▷ Deterministic case: $\psi = \phi - \phi^* e_0$, $\|\psi\|_{H^1}^2$ is a **Lyapunov** function

▷ Linear stoch case:

$$d\psi_k = \frac{1}{\varepsilon} a_k(t) \psi_k dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_k(t), \quad a_k(t) = \bar{a}(t) - \frac{k^2 \pi^2}{L^2} < 0$$

For any decomposition $h = \sum_k h_k$, τ first-exit time from $\mathcal{B}(h)$,

$$\mathbb{P}\{\tau < T\} \leq \sum_k \mathbb{P}\left\{\sup_t \psi_k(t)^2 \geq h_k^2 \langle k \rangle^{-2s}\right\} \leq \sum_k C_k(T, \varepsilon) e^{-\kappa h_k^2 \langle k \rangle^{2-2s} / \sigma^2}$$

Choose $h_k^2 \sim h^2 \langle k \rangle^{-2+2s+\eta}$, $\eta > 0$

Ideas of proof

$$\triangleright \phi(x) = \sum_{k \in \mathbb{Z}} \phi_k e_k(x) \Rightarrow \|\phi\|_{H^s}^2 = \sum_{k \in \mathbb{Z}} \langle k \rangle^{2s} \phi_k^2, \quad \langle k \rangle = \sqrt{1 + k^2}$$

$\triangleright$ Deterministic case: $\psi = \phi - \phi^* e_0$, $\|\psi\|_{H^1}^2$ is a **Lyapunov** function

$\triangleright$ Linear stoch case:

$$d\psi_k = \frac{1}{\varepsilon} a_k(t) \psi_k dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_k(t), \quad a_k(t) = \bar{a}(t) - \frac{k^2 \pi^2}{L^2} < 0$$

For any decomposition $h = \sum_k h_k$, τ first-exit time from $\mathcal{B}(h)$,

$$\mathbb{P}\{\tau < T\} \leq \sum_k \mathbb{P}\left\{\sup_t \psi_k(t)^2 \geq h_k^2 \langle k \rangle^{-2s}\right\} \leq \sum_k C_k(T, \varepsilon) e^{-\kappa h_k^2 \langle k \rangle^{2-2s} / \sigma^2}$$

Choose $h_k^2 \sim h^2 \langle k \rangle^{-2+2s+\eta}$, $\eta > 0$

$\triangleright$ **Schauder** estimate: $\beta \in H^r$, $0 < r < \frac{1}{2} \Rightarrow$

$$\|e^{t\Delta} \beta\|_{H^q} \leq M(q, r) t^{-(q-r)/2} \|\beta\|_{H^r} \quad \forall q < r + 2$$

Consequence: $\psi = \psi^0 + \psi^1$ where nonlinear term satisfies

$$\|\psi^1\|_{H^q} \leq M' \varepsilon^{(q-r)/2-1} \sup_t \|b(t, \psi(y, \cdot))\|_{H^r}$$

SPDE on the 2d torus

$$d\phi(t, x) = \frac{1}{\varepsilon} \left[\Delta \phi(t, x) + \sum_{j=1}^n A_j(t) \phi(t, x)^j \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW(t, x) \quad x \in \mathbb{T}^2$$

SPDE on the 2d torus

$$d\phi(t, x) = \frac{1}{\varepsilon} \left[\Delta \phi(t, x) + \sum_{j=1}^n A_j(t) \phi(t, x)^j \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW(t, x) \quad x \in \mathbb{T}^2$$

▷ SPDE is **not well-posed**, needs to be renormalised

For $N \in \mathbb{N}$, project on $\text{span}\{e_k\}_{|k| < N}$:

$$d\phi(t, x) = \frac{1}{\varepsilon} \left[\Delta \phi(t, x) + \sum_{j=1}^n A_j(t) P_N : \phi(t, x)^j : \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_N(t, x)$$

where $: \phi^n :$ = $H_n(\phi; C_N)$ Wick power, $C_N \sim \log N$ variance of GFF

SPDE on the 2d torus

$$d\phi(t, x) = \frac{1}{\varepsilon} \left[\Delta \phi(t, x) + \sum_{j=1}^n A_j(t) \phi(t, x)^j \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW(t, x) \quad x \in \mathbb{T}^2$$

- ▷ SPDE is **not well-posed**, needs to be renormalised

For $N \in \mathbb{N}$, project on $\text{span}\{e_k\}_{|k| < N}$:

$$d\phi(t, x) = \frac{1}{\varepsilon} \left[\Delta \phi(t, x) + \sum_{j=1}^n A_j(t) P_N : \phi(t, x)^j : \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_N(t, x)$$

where $: \phi^n :$ = $H_n(\phi; C_N)$ Wick power, $C_N \sim \log N$ variance of GFF

- ▷ Let ψ be stochastic convolution:

$$d\psi(t, x) = \frac{1}{\varepsilon} \Delta \psi(t, x) dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_N(t, x)$$

[Da Prato & Debussche '03]: $\phi - \psi$ cv to well-defined function

SPDE on the 2d torus

$$d\phi(t, x) = \frac{1}{\varepsilon} \left[\Delta \phi(t, x) + \sum_{j=1}^n A_j(t) \phi(t, x)^j \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW(t, x) \quad x \in \mathbb{T}^2$$

- ▷ SPDE is **not well-posed**, needs to be renormalised

For $N \in \mathbb{N}$, project on $\text{span}\{e_k\}_{|k| < N}$:

$$d\phi(t, x) = \frac{1}{\varepsilon} \left[\Delta \phi(t, x) + \sum_{j=1}^n A_j(t) P_N \phi(t, x)^j \right] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_N(t, x)$$

where $:\phi^n: = H_n(\phi; C_N)$ Wick power, $C_N \sim \log N$ variance of GFF

- ▷ Let ψ be stochastic convolution:

$$d\psi(t, x) = \frac{1}{\varepsilon} \Delta \psi(t, x) dt + \frac{\sigma}{\sqrt{\varepsilon}} dW_N(t, x)$$

[Da Prato & Debussche '03]: $\phi - \psi$ cv to well-defined function

- ▷ Use Besov–Hölder spaces $\mathcal{B}_{2,\infty}^\alpha$, $\alpha < 0$, instead of Sobolev spaces H^s :

$$\|\phi\|_{\mathcal{B}_{2,\infty}^\alpha} = \sup_{q \geq 0} 2^{q\alpha} \|\delta_q \phi\|_{L^2} \quad \delta_q \phi = \sum_{2^{q-1} \leq |k| < 2^q} \phi_k e_k$$

SPDE on the 2d torus

Theorem [B & Nader, EJP 2025]

For $\alpha < 0$, $m \in \mathbb{N}$,

$$\mathbb{P}\left\{\sup_{0 \leq t \leq T} \|\psi(t, \cdot)^m\|_{\mathcal{B}_{2,\infty}^\alpha} > h^m\right\} \leq C_m(T, \varepsilon, \alpha) e^{-\kappa_m(\alpha) h^2 / \sigma^2}$$

where

$$\kappa_m(\alpha) \geq c_0 \frac{\alpha^2}{m^7} \quad C_m(T, \varepsilon, \alpha) \leq c_1 \frac{T}{\varepsilon} \frac{m^{3/2} e^m m^m}{|\alpha|}$$

▷ Binomial formula

$$:\psi^m: = H_m(\psi; C_N) = \sum_{|n|=m} \frac{m!}{n!} \prod_{q \geq 0} H_{n_q}(\delta_q \psi; c_q) \quad c_q = \mathcal{O}(1)$$

▷ Doob submartingale inequality for $\sup_{t \in I_\ell} \|\delta_{q_0}(\prod_{q \geq 0} H_{n_q}(\delta_q \hat{\psi}; c_q))\|_{L^2}^2$

where $\hat{\psi}$ martingale approximating ψ on intervals I_ℓ depending on q_0

▷ Upgrade to bound for $\sup_{t \in I_\ell} \|\delta_{q_0}(\prod_{q \geq 0} H_{n_q}(\delta_q \psi; c_q))\|_{L^2}^2$

▷ Bound probability by summing over ℓ , q_0 and n

SPDE on the 2d torus

Theorem [B & Nader, EJP 2025]

For $\alpha < 0$, $m \in \mathbb{N}$,

$$\mathbb{P}\left\{\sup_{0 \leq t \leq T} \|\psi(t, \cdot)^m\|_{\mathcal{B}_{2,\infty}^\alpha} > h^m\right\} \leq C_m(T, \varepsilon, \alpha) e^{-\kappa_m(\alpha) h^2 / \sigma^2}$$

where

$$\kappa_m(\alpha) \geq c_0 \frac{\alpha^2}{m^7} \quad C_m(T, \varepsilon, \alpha) \leq c_1 \frac{T}{\varepsilon} \frac{m^{3/2} e^m m^m}{|\alpha|}$$

▷ Binomial formula

$$:\psi^m: = H_m(\psi; C_N) = \sum_{|n|=m} \frac{m!}{n!} \prod_{q \geq 0} H_{n_q}(\delta_q \psi; c_q) \quad c_q = \mathcal{O}(1)$$

▷ Doob submartingale inequality for $\sup_{t \in I_\ell} \|\delta_{q_0}(\prod_{q \geq 0} H_{n_q}(\delta_q \hat{\psi}; c_q))\|_{L^2}^2$

where $\hat{\psi}$ martingale approximating ψ on intervals I_ℓ depending on q_0

▷ Upgrade to bound for $\sup_{t \in I_\ell} \|\delta_{q_0}(\prod_{q \geq 0} H_{n_q}(\delta_q \psi; c_q))\|_{L^2}^2$

▷ Bound probability by summing over ℓ , q_0 and n

Concentration estimates

Theorem [B & Nader, EJP 2025]

Let $\phi_1 = \phi - \phi^* - \psi$. Then $\forall \gamma < 2, \forall \nu < 1 - \frac{\gamma}{2}, \forall h < h_0 \varepsilon^\nu$

$$\mathbb{P}\left\{\sup_{t \in [0, T]} \|\phi_1(t)\|_{C^{\gamma-1}} > M \varepsilon^{-\nu} h(h + \varepsilon)\right\} \leq C(T, \varepsilon) e^{-\kappa h^2 / \sigma^2}$$

- ▷ Use $\|\phi^\ell : \psi^m : \|_{\mathcal{B}_{2,\infty}^{(2\ell+1)\alpha}} \leq \|\phi\|_{\mathcal{B}_{2,\infty}^\ell}^\ell \|\psi^m\|_{\mathcal{B}_{2,\infty}^\alpha}^\alpha$ to bnd nonlin term in $d\phi_1$
- ▷ Use Schauder estimate and $\mathcal{B}_{2,\infty}^\gamma \hookrightarrow \mathcal{B}_{\infty,\infty}^{\gamma-1} = C^{\gamma-1}$

Example: Dynamic pitchfork bifurcation

$$d\phi(t, x) = \frac{1}{\varepsilon} [\Delta \phi(t, x) + a(t)\phi(t, x) - :\phi(t, x)^3:] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW(t, x)$$

- ▷ Decompose $\phi = \psi_\perp + \phi_1$, where $\psi_\perp$ stoch conv driven by $W_\perp$
- ▷ Decompose $\phi_1 = \phi_1^0 e_0 + \phi_1^\perp$
- ▷ Concentration estimate for $\|\phi_1^\perp\|_{C^{\gamma-1}}$ as long as $a(t) < (2\pi)^2$
- ▷ For ϕ_1^0 , similar results as in SDE case (bif delay of order $\sqrt{\varepsilon \log(\sigma^{-1})}$)

Concentration estimates

Theorem [B & Nader, EJP 2025]

Let $\phi_1 = \phi - \phi^* - \psi$. Then $\forall \gamma < 2, \forall \nu < 1 - \frac{\gamma}{2}, \forall h < h_0 \varepsilon^\nu$

$$\mathbb{P}\left\{\sup_{t \in [0, T]} \|\phi_1(t)\|_{C^{\gamma-1}} > M \varepsilon^{-\nu} h(h + \varepsilon)\right\} \leq C(T, \varepsilon) e^{-\kappa h^2 / \sigma^2}$$

- ▷ Use $\|\phi^\ell : \psi^m : \|_{\mathcal{B}_{2,\infty}^{(2\ell+1)\alpha}} \leq \|\phi\|_{\mathcal{B}_{2,\infty}^\ell}^\ell \|\psi^m\|_{\mathcal{B}_{2,\infty}^\alpha}$ to bnd nonlin term in $d\phi_1$
- ▷ Use Schauder estimate and $\mathcal{B}_{2,\infty}^\gamma \hookrightarrow \mathcal{B}_{\infty,\infty}^{\gamma-1} = C^{\gamma-1}$

Example: Dynamic pitchfork bifurcation

$$d\phi(t, x) = \frac{1}{\varepsilon} [\Delta \phi(t, x) + a(t)\phi(t, x) - :\phi(t, x)^3:] dt + \frac{\sigma}{\sqrt{\varepsilon}} dW(t, x)$$

- ▷ Decompose $\phi = \psi_\perp + \phi_1$, where $\psi_\perp$ stoch conv driven by $W_\perp$
- ▷ Decompose $\phi_1 = \phi_1^0 \mathbf{e}_0 + \phi_1^\perp$
- ▷ Concentration estimate for $\|\phi_1^\perp\|_{C^{\gamma-1}}$ as long as $a(t) < (2\pi)^2$
- ▷ For ϕ_1^0 , similar results as in SDE case (bif delay of order $\sqrt{\varepsilon \log(\sigma^{-1})}$)

The case of fractional noise

- ▷ Fractional Brownian motion (fBm) of Hurst parameter $H \in (0, 1)$:
Centred Gaussian process with covariance

$$\mathbb{E}[W_t^H W_s^H] = \frac{1}{2}(t^{2H} + s^{2H} - |t - s|^{2H})$$

Theorem [B & Blessing, ECP 2025]

Let $dx_t = \frac{1}{\varepsilon} a(t) x_t dt + \frac{\sigma}{\varepsilon^H} dW_t^H$, with $a(t) \leq a_0 < 0$ and $H \in (0, 1)$. Then

$$\mathbb{P}\left\{\sup_{t \in [0, T]} |x_t| > h\right\} \leq C\left(T, \frac{h}{\sigma}, a_0\right) e^{-\kappa h^2 / (2\sigma^2)} \quad \kappa = \frac{1 - \mathcal{O}(\varepsilon)}{2H\Gamma(2H)}$$

Theorem [B & Blessing, ECP 2025]

Let $d\phi(t, x) = \frac{1}{\varepsilon} [\Delta\phi(t, x) + f(t, \phi(t, x))] dt + \frac{\sigma}{\varepsilon^H} dW^H(t, x)$, $H \in (\frac{1}{4}, 1)$
 $\phi^*(t)$ stable eq. branch of f , $\mathcal{B}(h)$ nbh in H^s , $s \in (0, 2H - \frac{1}{2})$, $h = \mathcal{O}(\varepsilon^\nu)$

$$\mathbb{P}\left\{\tau_{\mathcal{B}(h)} < T\right\} \leq C\left(T, \frac{h}{\sigma}, a_0\right) \exp\left\{-(2H - \frac{1}{2} - s) c_0 \kappa \frac{h^2 [1 - \mathcal{O}(h\varepsilon^{-\nu})]}{2\sigma^2}\right\}$$

The case of fractional noise

- ▷ Fractional Brownian motion (fBm) of Hurst parameter $H \in (0, 1)$:
Centred Gaussian process with covariance

$$\mathbb{E}[W_t^H W_s^H] = \frac{1}{2}(t^{2H} + s^{2H} - |t - s|^{2H})$$

Theorem [B & Blessing, ECP 2025]

Let $dx_t = \frac{1}{\varepsilon} a(t) x_t dt + \frac{\sigma}{\varepsilon^H} dW_t^H$, with $a(t) \leq a_0 < 0$ and $H \in (0, 1)$. Then

$$\mathbb{P}\left\{\sup_{t \in [0, T]} |x_t| > h\right\} \leq C\left(T, \frac{h}{\sigma}, a_0\right) e^{-\kappa h^2 / (2\sigma^2)} \quad \kappa = \frac{1 - \mathcal{O}(\varepsilon)}{2H\Gamma(2H)}$$

Theorem [B & Blessing, ECP 2025]

Let $d\phi(t, x) = \frac{1}{\varepsilon} [\Delta\phi(t, x) + f(t, \phi(t, x))] dt + \frac{\sigma}{\varepsilon^H} dW^H(t, x)$, $H \in (\frac{1}{4}, 1)$
 $\phi^*(t)$ stable eq. branch of f , $\mathcal{B}(h)$ nbh in H^s , $s \in (0, 2H - \frac{1}{2})$, $h = \mathcal{O}(\varepsilon^\nu)$

$$\mathbb{P}\left\{\tau_{\mathcal{B}(h)} < T\right\} \leq C\left(T, \frac{h}{\sigma}, a_0\right) \exp\left\{-(2H - \frac{1}{2} - s) c_0 \kappa \frac{h^2 [1 - \mathcal{O}(h\varepsilon^{-\nu})]}{2\sigma^2}\right\}$$

The case of fractional noise

- ▷ Fractional Brownian motion (fBm) of Hurst parameter $H \in (0, 1)$:
Centred Gaussian process with covariance

$$\mathbb{E}[W_t^H W_s^H] = \frac{1}{2}(t^{2H} + s^{2H} - |t - s|^{2H})$$

Theorem [B & Blessing, ECP 2025]

Let $dx_t = \frac{1}{\varepsilon} a(t) x_t dt + \frac{\sigma}{\varepsilon^H} dW_t^H$, with $a(t) \leq a_0 < 0$ and $H \in (0, 1)$. Then

$$\mathbb{P}\left\{\sup_{t \in [0, T]} |x_t| > h\right\} \leq C\left(T, \frac{h}{\sigma}, a_0\right) e^{-\kappa h^2/(2\sigma^2)} \quad \kappa = \frac{1 - \mathcal{O}(\varepsilon)}{2H\Gamma(2H)}$$

Theorem [B & Blessing, ECP 2025]

Let $d\phi(t, x) = \frac{1}{\varepsilon} [\Delta\phi(t, x) + f(t, \phi(t, x))] dt + \frac{\sigma}{\varepsilon^H} dW^H(t, x)$, $H \in (\frac{1}{4}, 1)$
 $\phi^*(t)$ stable eq. branch of f , $\mathcal{B}(h)$ nbh in H^s , $s \in (0, 2H - \frac{1}{2})$, $h = \mathcal{O}(\varepsilon^\nu)$

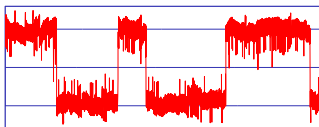
$$\mathbb{P}\left\{\tau_{\mathcal{B}(h)} < T\right\} \leq C\left(T, \frac{h}{\sigma}, a_0\right) \exp\left\{-(2H - \frac{1}{2} - s)c_0\kappa \frac{h^2[1 - \mathcal{O}(h\varepsilon^{-\nu})]}{2\sigma^2}\right\}$$

Stochastic resonance in an SDE

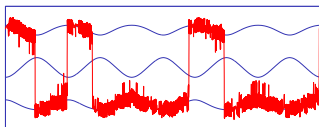
$$\begin{aligned} dx_t &= \underbrace{\left[-x_t^3 + x_t + A \cos(\varepsilon t) \right]}_{= -\frac{\partial}{\partial x} \left[\frac{1}{4} x^4 - \frac{1}{2} x^2 - A x \cos(\varepsilon t) \right] \Big|_{x_t}} dt + \sigma dW_t \end{aligned}$$

- ▷ Ice Ages: deterministically bistable climate [Croll, Milankovitch]
- ▷ random perturbations due to weather [Benzi-Sutera-Vulpiani, Nicolis-Nicolis]

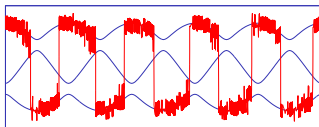
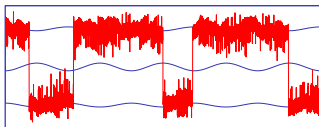
Sample paths $\{x_t\}_t$ for $\varepsilon = 0.001$:



$A = 0, \sigma = 0.3$



$A = 0.24, \sigma = 0.2$

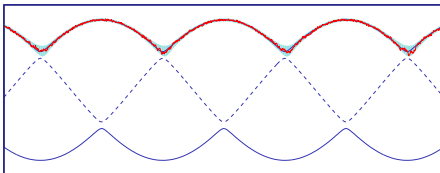


The synchronisation regime

$A_c = \frac{2}{3\sqrt{3}}$, $A = A_c - \delta$, $0 < \delta \ll 1$. Critical noise intensity: $\sigma_c = \max\{\delta, \varepsilon\}^{3/4}$

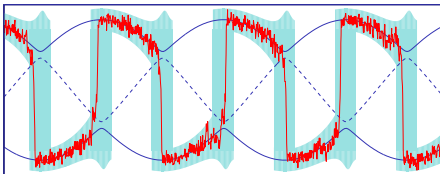
$\sigma \ll \sigma_c$:

transitions unlikely



$\sigma \gg \sigma_c$:

synchronisation



Theorem [B & Gentz, Annals App. Proba 2002]

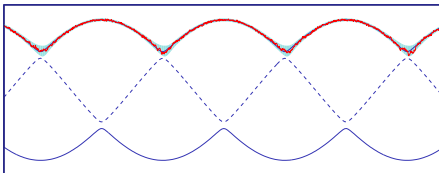
- ▶ Away from (avoided) bifurcations, sample paths concentrated in σ -neighbourhood of deterministic stable periodic solutions
- ▶ $\sigma \ll \sigma_c$: transition probability per period $\leq e^{-\sigma_c^2/\sigma^2}$
- ▶ $\sigma \gg \sigma_c$: transition probability per period $\geq 1 - e^{-c\sigma^{4/3}/(\varepsilon|\log \sigma|)}$

The synchronisation regime

$A_c = \frac{2}{3\sqrt{3}}$, $A = A_c - \delta$, $0 < \delta \ll 1$. Critical noise intensity: $\sigma_c = \max\{\delta, \varepsilon\}^{3/4}$

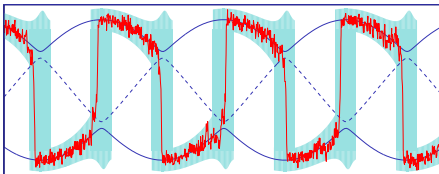
$\sigma \ll \sigma_c$:

transitions unlikely



$\sigma \gg \sigma_c$:

synchronisation



Theorem [B & Gentz, Annals App. Proba 2002]

- ▷ Away from (avoided) bifurcations, sample paths concentrated in σ -neighbourhood of deterministic stable periodic solutions
- ▷ $\sigma \ll \sigma_c$: transition probability per period $\leq e^{-\sigma_c^2/\sigma^2}$
- ▷ $\sigma \gg \sigma_c$: transition probability per period $\geq 1 - e^{-c\sigma^{4/3}/(\varepsilon|\log \sigma|)}$

Stochastic resonance in stochastic PDEs

$$d\phi(t, x) = \left[\Delta\phi(t, x) + \phi(t, x) - \phi(t, x)^3 + \underbrace{A \cos(\varepsilon t)}_{h(\varepsilon t)} \right] dt + \sigma dW(t, x)$$

Simulation available at youtu.be/eN3NWiEjBK8

Stochastic resonance in bistable SPDEs on $\mathbb{T}^1$

$$d\phi(t, x) = [\Delta\phi(t, x) + f(\varepsilon t, \phi(t, x))] dt + \sigma dW(t, x)$$

- ▷ $\phi = \phi(t, x) \in \mathbb{R}$, $\varepsilon t \in [0, T]$ or f is T -periodic, $x \in \mathbb{T} = \mathbb{R}/L\mathbb{Z}$, $L > 0$
- ▷ $\phi \mapsto f(s, \phi)$ bistable, $\mathcal{C}^2$, confining, e.g. $f(s, \phi) = \phi - \phi^3 + A \cos(s)$
- ▷ $dW(t, x)$ space-time white noise on $\mathbb{R}_+ \times \mathbb{T}$
- ▷ $0 < \varepsilon, \sigma \ll 1$
- ▷ δ measures closeness to bifurcation (e.g. $A_c - A$)

Theorem [B & Nader, SPDE 2025]

- ▷ Away from bifurcations, solutions are concentrated around deterministic solutions in Sobolev H^s -norm for any $s < \frac{1}{2}$
- ▷ $\sigma \ll \sigma_c = \max\{\delta, \varepsilon\}^{3/4}$: transition probability per period $\leq e^{-\sigma_c^2/\sigma^2}$
- ▷ $\sigma \gg \sigma_c$: transition probability per period $\geq 1 - e^{-c\sigma^{4/3}/(\varepsilon|\log \sigma|)}$

Stochastic resonance in bistable SPDEs on $\mathbb{T}^1$

$$d\phi(t, x) = [\Delta\phi(t, x) + f(\varepsilon t, \phi(t, x))] dt + \sigma dW(t, x)$$

- ▷ $\phi = \phi(t, x) \in \mathbb{R}$, $\varepsilon t \in [0, T]$ or f is T -periodic, $x \in \mathbb{T} = \mathbb{R}/L\mathbb{Z}$, $L > 0$
- ▷ $\phi \mapsto f(s, \phi)$ bistable, $\mathcal{C}^2$, confining, e.g. $f(s, \phi) = \phi - \phi^3 + A \cos(s)$
- ▷ $dW(t, x)$ space-time white noise on $\mathbb{R}_+ \times \mathbb{T}$
- ▷ $0 < \varepsilon, \sigma \ll 1$
- ▷ δ measures closeness to bifurcation (e.g. $A_c - A$)

Theorem [B & Nader, SPDE 2025]

- ▷ Away from bifurcations, solutions are concentrated around deterministic solutions in Sobolev H^s -norm for any $s < \frac{1}{2}$
- ▷ $\sigma \ll \sigma_c = \max\{\delta, \varepsilon\}^{3/4}$: transition probability per period $\leq e^{-\sigma_c^2/\sigma^2}$
- ▷ $\sigma \gg \sigma_c$: transition probability per period $\geq 1 - e^{-c\sigma^{4/3}/(\varepsilon|\log \sigma|)}$

Related result

- ▷ Case $x \in \mathbb{T}^3$: Dimitri Faure, *Concentration around a stable equilibrium for the non-autonomous Φ_3^4 model*, Stoch. & PDEs: Anal. & Comp. Online first (2026)

References

- ▷ N. B. & Barbara Gentz, *A sample-paths approach to noise-induced synchronization: Stochastic resonance in a double-well potential*, Ann. Appl. Probab., **12**(1):1419–1470, 2002
- ▷ ———, *Noise-Induced Phenomena in Slow-Fast Dynamical Systems. A Sample-Paths Approach*, Springer, Probability and its Applications (2005)
- ▷ N. B. & Rita Nader, *Stochastic resonance in stochastic PDEs*, Stoch. & PDEs: Anal. & Comp. **11**:348–387 (2023)
- ▷ ———, *Concentration estimates for slowly time-dependent singular SPDEs on the two-dimensional torus*, Electronic J. Probability **29**:1–35 (2024)
- ▷ N. B. & Alexandra Blessing, *Concentration estimates for SPDEs driven by fractional Brownian motion*, Electronic Comm. Probability **30**:1-13 (2025)
- ▷ N. B., *An Introduction to Singular Stochastic PDEs*, EMS Press (2022)

